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EVALUATION FFF – IMPACT ANALYSIS

BACKGROUND REPORT 3.2

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Evaluation FFF

Impact Analysis

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Content

LIST OF TABLES	5
LIST OF FIGURES.....	6
1 INTRODUCTION	7
1.1. Background.....	7
1.1.1 Database	7
1.1.2 Structure of the Report	8
1.2. Relevance of the FFF in the context of the Austrian funding system.....	9
2 FFF-FUNDED PROJECTS - PATTERNS OF PARTICIPATION	10
2.1. The data	10
2.2. Applications and approvals from 1995 to 2003	11
2.3. Rejections and Approvals: a project perspective.....	13
2.4. Rejections and Approvals: the firm perspective.....	16
2.5. Rejections, Approvals, and funding: a multivariate approach	21
2.5.1 A binary participation model.....	23
2.5.2 A linear funding model.....	26
3 INPUT ADDITIONALITY	28
3.1. What is the problem?	28
3.2. Theoretical Considerations	28
3.2.1 Typical results: a quick literature survey	30
3.3. The data base	31
3.4. The model.....	33
3.4.1 Standard errors revisited: a bootstrap approach.....	36
3.4.2 Additionality – a function of firm size?	36
3.4.3 Additionality in firms with sporadic R&D activities.....	38
3.4.4 Concluding remarks.....	39
4 OUTPUT ADDITIONALITY	41
4.1. Factors explaining the Level of R&D Subsidies	41
4.1.1 Hypotheses about the agency's allocation rule.....	41
4.1.2 Data and descriptive results.....	43
4.1.3 Empirical results	45
4.1.4 Concluding remarks.....	47
4.2. Productivity Effects of R&D subsidies.....	47
4.2.1 Introduction.....	47
4.2.2 Empirical model and hypothesis	48
4.2.3 Data and descriptive results.....	49
4.2.4 Results of the productivity effects of the amount of R&D subsidies.....	50
4.2.5 Concluding remarks.....	51
5 BEHAVIOURAL ADDITIONALITY	52

5.1.1	Project additionality	53
5.1.2	Behavioural changes in case of rejection	55
5.1.3	Collaboration and networking	56
5.1.4	Preceding and subsequent projects	57
5.1.5	Econometric evidence from the FFF-panel data set	58
5.1.6	Concluding remarks	62
6	IMPLICATIONS OF THE FFF-SUPPORT: RESULTS OF A SURVEY OF FFF CUSTOMERS	63
6.1.	The survey and sampling	63
6.1.1	Some basic features of the sample firms	64
6.1.2	Objectives and problems with R&D activities of FFF customers	68
6.2.	Characterisation of FFF-funded Projects	69
6.2.1	Main motivation for submission: high costs of R&D	69
6.2.2	FFF-funded projects lead to R&D extension	71
6.2.3	FFF-funded projects are technologically more difficult, more expensive and of longer duration	72
6.2.4	Majority of Firms: FFF funded projects have no impact on other R&D projects	73
6.2.5	FFF-Funding Results: New or improved products/services are most frequent	74
6.2.6	To open up a market: primary goal for commercialisation	75
6.2.7	Firms' motivations and project advancement: differences between successful and unsuccessful FFF submissions	76
6.3.	Licenses, Products and Processes	78
6.3.1	Time to market	78
6.3.2	Licenses	79
6.3.3	Products and Processes	80
7	MAIN FINDINGS	83
8	REFERENCES	86
	APPENDIX I: DEFINITION OF NACE-CODES	89
	APPENDIX II: SAMPLE DISTRIBUTION BY SECTORAL AFFILIATION AND SUB-BRANCH	90

List of Tables

Table 1: Applications and approvals by type of applicant, S 1995-2003 (present).....	10
Table 2: Descriptive statistics of model variables, 1998-2003 (present).....	22
Table 3: Estimation results of the probit model of participation.....	23
Table 4: simulated percentage difference in approval rates vis a vis the “benchmark project”.....	25
Table 5: Estimation results of the OLS model of the funding share (present value of funding as % of total project costs).....	27
Table 6: Econometric results, geographically differentiated.....	30
Table 7: Results of the fixed-effects panel regression.....	35
Table 8: Results of the bootstrapping: descriptive statistics and Kernel density approximation.....	36
Table 9: Model results disaggregated by firm size.....	37
Table 10: Additionality of firms with intermittent R&D performance.....	39
Table 11: R&D intensity and R&D subsidy ratio (supported firms), 1995-2002.....	43
Table 12: R&D intensity and subsidies ratio by firm size (supported firms).....	44
Table 13: Summary statistics (supported firms).....	44
Table 14: Summary statistics (dummy variables, percentage share of total).....	45
Table 15: Determinants of the ratio of R&D subsidies to total R&D: Panel estimates.....	46
Table 16: Change in labour productivity and R&D subsidy ratio: Panel estimates.....	51
Table 17: Implementation/non-implementation if application was/is rejected: analysis by sector-affiliation.....	53
Table 18: Implementation/non-implementation if application was/is rejected: analysis by firm-size.....	54
Table 19: Behavioural Additionality: adaptations if application was/is rejected.....	55
Table 20: Behavioural Additionality: collaboration and networking (%).....	57
Table 21: Behavioural Additionality: chain effects of public funding (in %).....	58
Table 22: Panel estimates of the determinants of log(scientific R&D-personnel).....	59
Table 23: Panel estimates of the determinants of the logarithm of scientific R&D personnel by size.....	60
Table 24: Impact of subsidies on scientific R&D personnel: Dynamic panel estimates ^{a)}	61
Table 25: Increase in FFF-subsidies necessary to employ one additional R&D-worker.....	62
Table 26: Sample for the survey.....	63
Table 27: Response rate.....	64
Table 28: Joint distribution of sector-affiliation and firm-size.....	65
Table 29: Year of foundation (percentile-distribution) (N=1104).....	65
Table 30: Aggregate sales (in Mio. €).....	66
Table 31: Aggregate export performance (levels in Mio. €) ^{a)}	66
Table 32: R&D-intensity of sample firms.....	67
Table 33: Experience with FFF.....	68
Table 34: Success categories: absolute frequencies and shares.....	68
Table 35: R&D-activities are aiming at.....	68
Table 36: Problems related to companies' R&D-engagement (in %, N=1232).....	69
Table 37: Important / not-important reasons to submit a project to FFF (in %, multiple responses).....	70
Table 38: Characteristics of successfully submitted projects (in %, multiple responses).....	71
Table 39: Comparison of FFF-funded projects and not externally funded projects (in %).....	73
Table 40: FFF funded projects: impact on other R&D projects (in %, multiple responses).....	74
Table 41: Results of the FFF funded projects (in %, multiple responses).....	75
Table 42: Commercialisation of the results of R&D: the four most important goals (in %).....	76

List of Figures

Figure 1: Applications and Costs of FFF-funded Projects, 1995-2003 (present).....	11
Figure 2: Approval and funding rates, 1995-2003 (present)	12
Figure 3: Funding mix, 1995-2003 (present).....	13
Figure 4: Approval and funding rates by size of project and type of fund, 1995-2003 (present).....	13
Figure 5: Applications, costs and funding rates by economic activity, 1998-2003 (present)	14
Figure 6: Applications and approvals by economic activity, 1998-2000 vs. 2001-2003 (present)	16
Figure 7: Repeated Applications, 1995-2003 (present).....	17
Figure 8: Applications and approvals by firm size and age, 1995-2003 (present).....	18
Figure 9: Applications and approvals by firm location 1995-2003 (present).....	19
Figure 10: Applications and approvals by firms' export share and R&D share, 1995-2003 (present)	20
Figure 11: Effects of R&D Subsidies on Total R&D Expenditures	29
Figure 12: Constructing time series by the „Relay method“	32
Figure 13: Number of firms by number of applications to the FFF, 1995-2003	33
Figure 14: Implementation/non-implementation if application was rejected: analysis by firm-size.....	37
Figure 15: PV rates by size class.....	38
Figure 16: Correlations between the growth rate of labour productivity in the following two years and the initial R&D subsidy-sales ratio.....	49
Figure 17: Correlation between the initial R&D subsidy ratio in t-2 and the growth rate of R&D personnel in the following years ^{a)}	58
Figure 18: Motivations - Differences between successful and unsuccessful submissions (in %)	77
Figure 19: Project advancement – Differences between successful and unsuccessful submissions (in %)	77
Figure 20: Time to market from the end of project to market entry (N=855) (1995-2003).....	78
Figure 21: Licenses by size and sector in % (N=862) (1995-2003)	79
Figure 22: License revenues by size and sector for the period 1995-2003(N=118).....	79
Figure 23: Product/Process development by size and sector (1995-2003).....	80
Figure 24: Impact of FFF funding on product development (y-axis = cumulative percentage of answering firms)	81
Figure 25: Share of new products out of FFF funding on total turnover (last business year).....	81
Figure 26: Impact of FFF funding on product development (y-axis = cumulative percentage of answering firms)	82

1 Introduction

1.1. BACKGROUND

The Austrian Industrial Research Promotion Fund (FFF) occupies a central role in the Austrian Innovation System. One of its main activities is to provide public funding to industrial R&D projects. One of the principal tasks of the FFF is thus to support promising innovative projects at the limit of a firm's capabilities in order to encourage continuing R&D activities.

More research and development to generate new products, processes and services are among the most important factors of business success. We therefore pursue forcefully our legally prescribed objective of encouraging firms to undertake on a lasting basis more research and more ambitious projects and to support them in doing so. Our support is mainly financial, backed by other measures (e.g. networking, partner search). (Mission of the FFF).

This study is part of a larger evaluation of the FFF set by the Ministry for Transport, Innovation and Technology (BMVIT). The study is aimed at giving a thorough description of the patterns of R&D funding by the FFF, identifying parameters which influence the provision of funds and presenting the direct, indirect and broader effects of FFF funding. It was thus one of the main interests to use the concept of additionality to analyze different (additionality) aspects of firms that have received funding from the FFF.

The question of additionality has been hotly discussed in recent years resulting out of the growing interest in R&D subsidies and the measurement of their impact. 'What difference does it make?' – the rationale of additionality still is the main touchstone of design and outcome of public policy. It is the most obvious guideline for government R&D policy and is reflected in most of the mission statements of R&D funding organizations: profit-maximizing firms underinvest in R&D.

The concept of additionality has a wide set of intentions which we try to consider within the present study. It focuses on (i) the extent to which public support of private R&D lead to an increase in overall research expenditures by the funded firms. Input additionality thus investigates whether publicly funded R&D is complementary and thus 'additional' to privately funded R&D spending. Another intention (ii) is output additionality, which focuses on the questions whether publicly funded R&D had an impact on both research output and overall productivity growth. Behavioural additionality (iii) broadens the traditional additionality concept by looking at permanent changes in the conduct of a company, possibly mirrored in a more formal institutionalization of innovation and R&D-activities. It is obvious that these elements are more challenging since they are more intangible, hard to measure and based on more subjective assessments.

1.1.1 Database

We used mainly two databases for our analysis:

- Project- and firm-level data, which were provided by the FFF and comprise the period from 1995 to November 2003.
- Results of a questionnaire survey conducted among funded firms as well as firms, which submitted a research proposal to the FFF but were rejected.

A detailed description of both databases can be found at the beginning of the relevant chapters.

We are well aware of the fact that some of the issues mentioned above are quite difficult to mention without a valid ‘control group’. Hence, the fundamental question ‘can the attained advance in R&D be credited to public intervention, or would it have been taken place anyway?’ is all but trivial and leads to measurement problems when using the additionality concept. This is because there are great difficulties in estimating the returns to R&D and the nature of the problem as a counterfactual. The latter leads to the problem of finding a valid control group when seeking to estimate impacts based on panel data. However, the reasons for not using a control group in our study are the following:

- As the present study is part of a larger evaluation of the FFF an additional survey of a relevant control group would have allocated too much resources on this issue and ignored other relevant issues of the evaluation.
- Austria is a small country and the FFF is the major funding agency. As we covered in our survey all firms, which submitted a proposal at the FFF in the period 1995–2003 we simply presumed that it would be hard to find firms, which are innovative (or undertake R&D) and had never submitted a proposal at the FFF. We guessed that most of the firms, which do R&D are aware of the FFF for supporting business R&D. We therefore concluded that creating a sample of firms, as a control group for FFF funded firms would be enormous costly if not impossible.

1.1.2 Structure of the Report

The study is divided into different sections led by different hypotheses and based on different databases.

The **first section** concerns the analysis of who are the participants of FFF funding. We thereby look at the application numbers and rejection rates from two perspectives: the project perspective includes the solicited amount, the types of funds, the sectoral activity of the project, and whether it is a follow-up to a previous project or a completely new application; the firm perspective deals with the firms’ geographical location, firm type, size and age, and previous research activities. To complement these ‘one-dimensional’ approaches, a multi-variate model of binary choice will be estimated in the second part. It aims at identifying parameters, which influence the Fund’s decision on whether to accept or reject a certain proposal.

The following chapters deal with the additionality of public R&D Funding. The **second section** is on input additionality and addresses the question: do public contributions to private research boost total R&D expenditures – and if so, do they boost them by an amount which is larger than the amount of public money which was used in this way?

The **third section** deals with output additionality and consists of two parts: the first part presents results on factors explaining the level of R&D subsidies. It looks at the evolution of the R&D subsidy ratio as well as the R&D intensity among FFF supported firms and quantifies econometrically the main factors behind the amount of R&D subsidies. The second part investigates the relationship between privately and publicly funded R&D on labor productivity growth.

The **fourth section** is on behavioral additionality. This section uses data from the survey as well as information of the FFF project and firm database. It explores various dimensions of behavioural additionality resulting from FFF-subsidies and investigates whether participation in FFF funded projects influences the R&D-related behaviour of the firms in a significant manner. The focus is on the building of innovative capabilities and competence building in general and on the companies’ ability to make use of new technologies and R&D procedures elsewhere. In the case of such effects this may strengthen the firm’s ability to absorb new knowledge, i.e. its absorptive capacity.

The **fifth section** is based on the results of the survey and presents the relevant results of the questionnaire. The first part deals with the implementation of the survey and describes our approach and sampling. It further summarizes the basic features of the sample firms. The following parts deal with the following questions:

- Are there firm (or categories)-specific motivations to submit a research proposal to the FFF?
- What are the characteristics of FFF-funded projects, what are the aims and what are the distinct differences between FFF-funded projects and projects funded by the firm solely?
- What are the technical results of FFF-funded projects?
- What are the distinct differences between FFF-funded projects and not FFF-funded projects as regards reasons of submission and project advancement?

The final part is related to the FFF funding in total for the period 1995-2003 and aims at assessing the direct economic effects in terms of licenses, new developed products and services and new process innovations. Although quite difficult to assess the respondents tried to calculate the proportion of the FFF in relation to all new developed products, services and process for the period 1995-03.

The **sixth section** summarizes the main findings.

1.2. RELEVANCE OF THE FFF IN THE CONTEXT OF THE AUSTRIAN FUNDING SYSTEM

According to the OECD-MSTI database, Business Expenditure on R&D (BERD) amounted to € 2,107 Million in 1998, the most recent year for which official BERD data are available. Of this sum, 5.5 % or about € 116 Million were financed by government (64.4 % were financed by industry, 30.1 % were financed by abroad. 0.1 % were financed by other national sources.

In 1998, the FFF supported business R&D with subsidies and loans worth a total of € 187 Million, equivalent to a present value of € 76.7 Million. Thus, the FFF accounted for about 66 % of that part of Austrian BERD which was financed by government. Adding those funds which are provided by other sources but which are managed by the FFF, viz. € 8.5 Million from the OeNB, the Austrian Central Bank, and € 8.4 Million by the ITF, the Ministry of Innovation's *Innovation and Technology Fund*, raises the FFF's share in government-financed BERD to more than 80 %.

Deducing from this impressive number, it is probably safe to assume that the FFF can count the overwhelming majority of innovating firms in Austria as their clients. Of innovating firms, which have applied for public R&D subsidies, probably all have approached the FFF at least once during the last decade.

2 FFF-funded Projects - Patterns of Participation

2.1. THE DATA

The project-level data were provided by FFF; they comprise the period from 1995 to November of 2003. The data base consists of two parts, one containing information on the project itself, the other information on the applying firms. Not all projects, however, can be unambiguously linked to this data base of applying firms: 260 projects were submitted by research co-operations of more than one firm (*Arbeitsgemeinschaften*, or *ARGEs*). A further 938 applications (or almost 10 % of the total) were filed by applicants which were not contained in the firm data base. According to the FFF, these applicants are individuals, University institutes or “very old” firms. The following Table 1 provides an overview.

Table 1: Applications and approvals by type of applicant, S 1995-2003 (present)

type of applicant	# applicants	# applications	# applications / applicant	costs / project [k€]	approval rate [%]	costs / approved project [k€]	funding rate [%]	PV rate [%]
Single Firms	3 138	8 769	2.8	510	75	551	45	22
ARGE	191	260	1.4	458	76	513	42	24
Firm data n.a.	711	938	1.3	404	60	440	49	26
total	4 028	9 967	2.5	498	74	542	45	22

Source: FFF; own calculations

In total, the data base lists 4028 applicants submitting a total of 9967 proposals, 7384 or 74 % of which eventually received funding. The overwhelming majority of projects (88 %) is submitted by single firms; less than 3 % are filed by ARGEs. The rest is filed by applicants which are not contained in the data base.

The costs per submitted project amount to about 500 000 €. Interestingly, at 460 000 €, the typical project submitted by ARGEs is somewhat below that average.

Approval rates are very similar for single firms and ARGEs; only projects for “other applicants”, at 60 %, face markedly below-average approval rates. For all types of applicants, the average costs of approved projects are about 10 % higher than the average costs of proposed projects: obviously, more expensive projects face somewhat higher chances of approval than cheaper ones.

The average funding rate (i.e., the share of project costs eligible for subsidies) is 45 % of the costs of approved projects. It is quite homogenous for all applicants, as is the Present Value rate (PV) (the share of the subsidies’ present (cash) value, which on average amounts to 22 % of total project costs). The PV rate is lower than the funding rate, because only part of the subsidies is non-repayable (the rest consists of loans by the FFF or FFF-subsidised loans by private banks; for more on the “funding mix”, cf. below).

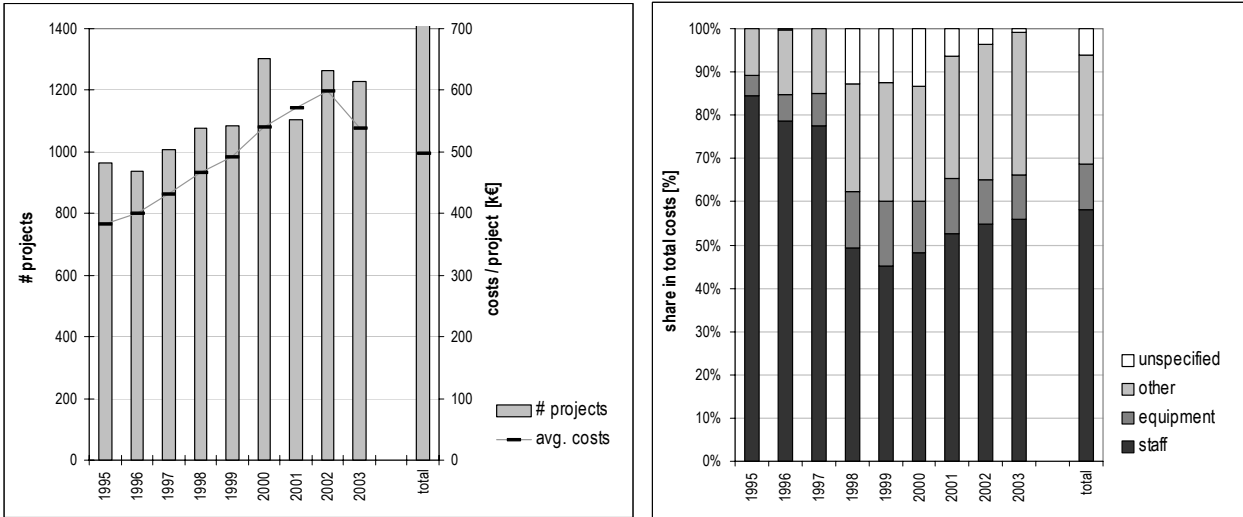
Not all subsidies granted by are paid for out of FFF funds proper: additionally to these funds, the FFF manages funds provided by the Austrian Central Bank (OeNB) and by the Ministry for Transport, Innovation, and Technology (BMVIT; the so-called *Innovations- und Technologiefonds*, ITF).

88 % of approved projects were financed out of FFF funds, 6 % received funding from OeNB sources and a further 6 % were funded under the ITF program. All three types of funds, however, are managed by the FFF on identical grounds. Therefore, in the following analysis, they are lumped together as “FFF-funded projects”.

Besides the 9967 applications which were thus managed, a further 472 applications were submitted for dedicated programs initiated and financed by BMVIT, 212 or 45 % of which successfully. Though managed by the FFF, these programs follow their own set of rules; accordingly, they will largely be omitted from the following analyses.

2.2. APPLICATIONS AND APPROVALS FROM 1995 TO 2003

Figure 1: Applications and Costs of FFF-funded Projects, 1995-2003 (present)

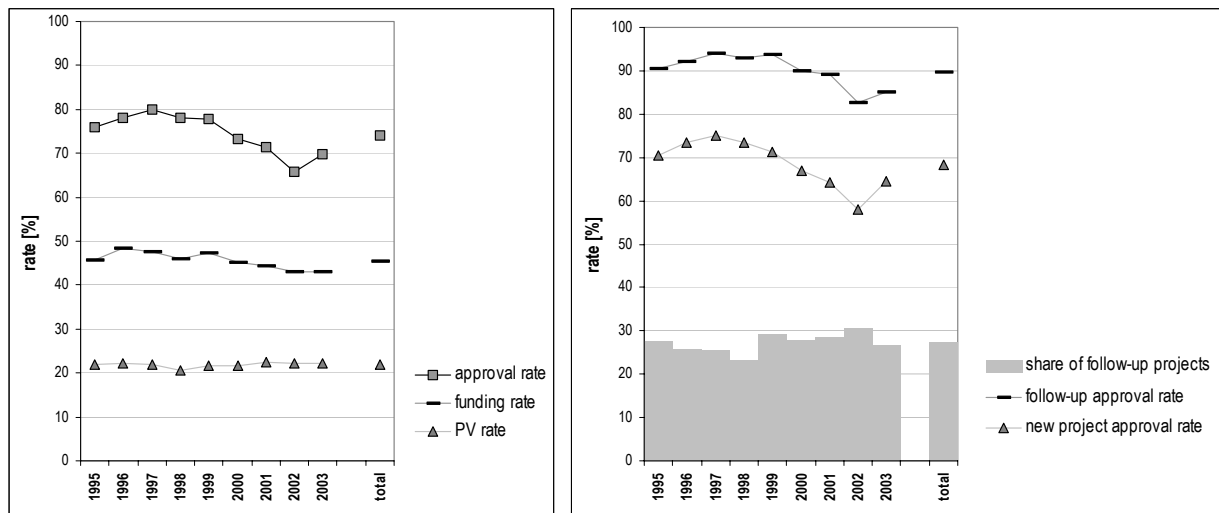


Source: FFF; own calculations

The number of annual applications has risen over the past 9 years, from under 1000 in 1995/96 to more than 1200 in 2002 and 2003 (rem.: for 2003, submissions are included until November only). A rising trend can also be seen in the average solicited amount, from under 400 000 € per project in 1995 to almost 600 000 € in 2002, an annual increase of 6.6 %. In 2003, however, average project size dropped appreciably, to a level last seen in the year 2000.

Project applications show a remarkable change in their cost structure: in the first three years, staff costs accounted for some 80 % of total costs, with equipment and “other costs” amounting to 6 and 14 % respectively. Starting in 1998, staff costs dropped sharply, to about 50 %; equipment and other costs roughly doubled their share. Additionally, some 8 % of the solicited amount remained “unspecified”.

Figure 2: Approval and funding rates, 1995-2003 (present)



Source: FFF; own calculations

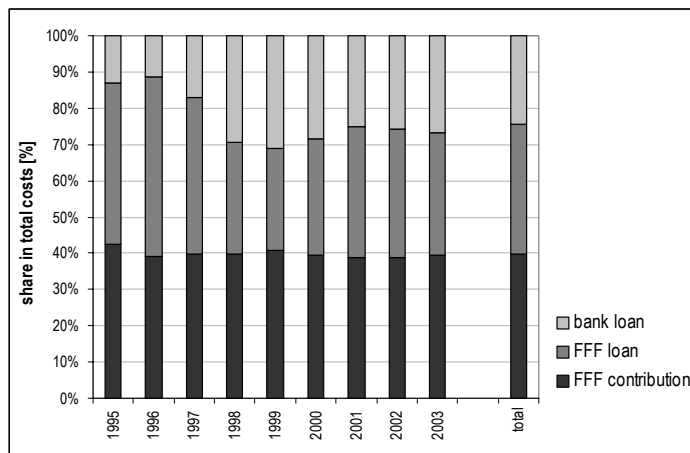
Whereas application numbers and solicited amount both went up, approval rates markedly decreased, from just under 80 % in 1995-97 to about 70 % in 2001-03¹. Approval rates are, however, quite different for new and for follow-up projects: not surprisingly, follow-ups experience a drastically higher rate of approval, beating new projects by about 20 percentage points. Both types, however, exhibit the same pattern over time.

The share of follow-up projects is quite stable, if somewhat higher in 1999-2003 than in 1995-1998. At 27 % on average, it is also quite high. This can be explained by the FFF's policy of favouring shorter projects, with a duration of 6-18 months. Longer projects are then financed via follow-up proposals.

Also shown in Figure 2 are two rates pertaining to the funding of approved projects. The first, the *funding rate*, gives the share of the amount accepted for funding. Funding, however, consists of a mix of 3 instruments: direct (non-repayable) subsidies, (refundable) loans by the FFF, and bank loans, which are guaranteed (and subsidised) by the FFF. As most projects are financed by a mix of non-refundable contributions (from the FFF) and refundable loans (either from the FFF or a private bank; cf. Figure 3 below), the *Present Value* of the approved subsidies is smaller than their nominal amount. Consequently, although the funding rate hovers around 45 %, the PV of the funding mix is much lower, at a stable 20-23 % of the approved project costs.

¹ In fact, the possible funding decisions are not just "approved" and "rejected"; two further types of decision exist: "rejected for lack of funds" and "retracted". Both outcomes, however, are of only marginal importance: 1.3 % of projects are rejected by the FFF citing funding problems. Even fewer projects are retracted by the applicant: of almost 10 000 applications, only 4 suffered this fate. Therefore, all three types of negative decision are lumped together.

Figure 3: Funding mix, 1995-2003 (present)



Source: FFF; own calculations

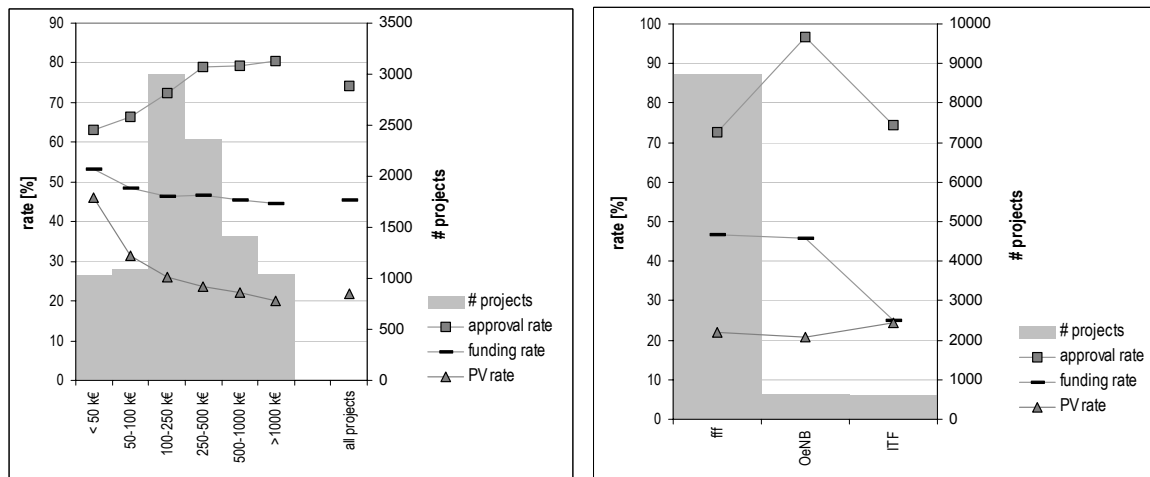
Figure 3 shows the funding mix of the average project: of all the costs approved for funding, about 40 % are disbursed as non-refundable subsidies. The rest is divided between loans by either the FFF itself or by private banks (the FFF's contribution to private bank loans consists of guarantees and allowances towards credit costs). Whereas the share of FFF contributions remained stable, bank loans gained in importance (from 15 % in 1995-97 to more than a quarter afterwards).

The following two chapters will examine rejections and approvals from the perspective of the application and of the applicant.

2.3. REJECTIONS AND APPROVALS: A PROJECT PERSPECTIVE

Both approval and funding rates, however, are not homogeneous: rather, they correlate with project size and also the type of fund, as the following Figure 4 shows.

Figure 4: Approval and funding rates by size of project and type of fund, 1995-2003 (present)



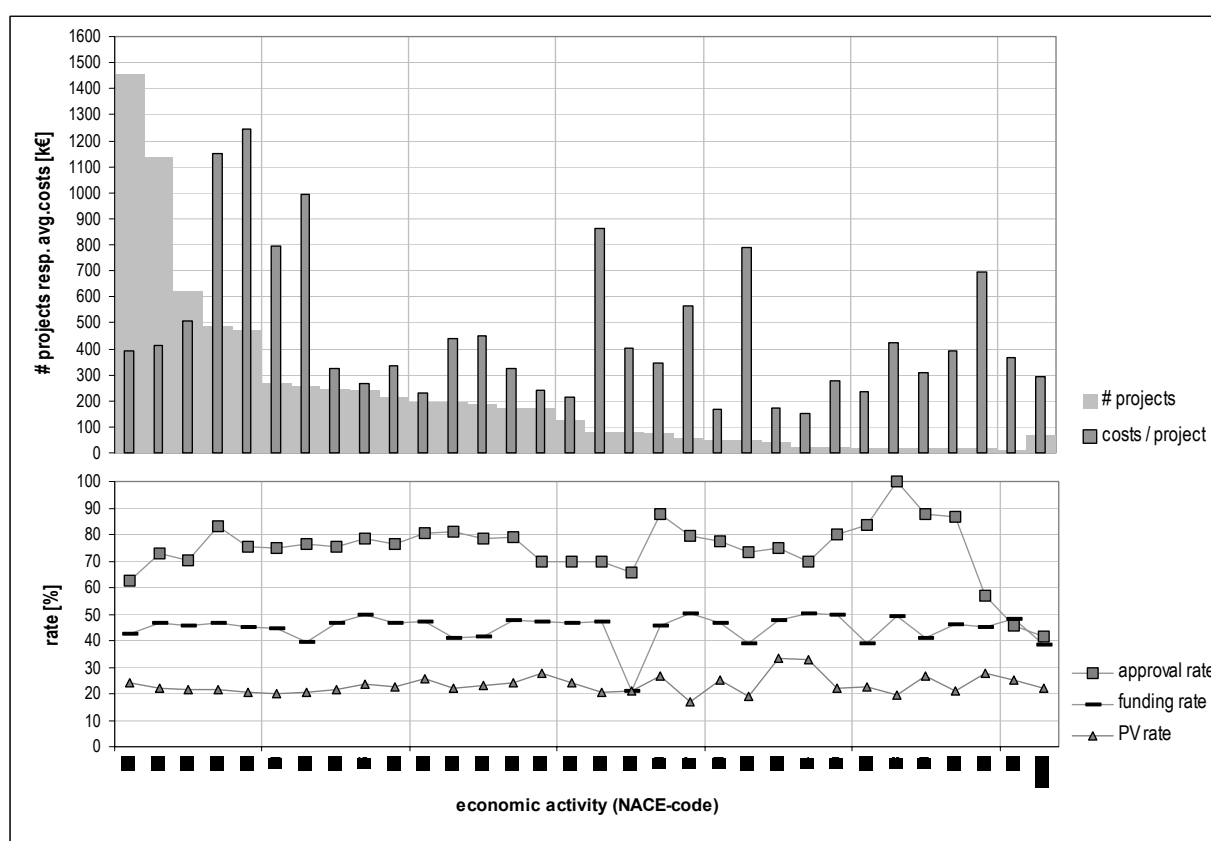
Source: FFF; own calculations

The larger the project, the higher the rate of approval, but the lower the funding rate and – even more so – the PV rate. The funding mix of smaller projects is much more biased towards non-repayable contributions. As a consequence, the PV of small projects amounts to 47 % of fundable costs, vs. an average of only 22 % for all projects. Also, more of their solicited allowance is recognised as fundable in the first

place (at 54 %, their funding rate is higher than the average 46 %). The most expensive applications get rejected only in about 20 % of cases; on the other hand, their subsidies' PV is less than 20 % of approved costs.

Disaggregation by type of fund shows that projects financed by the OeNB fund are rejected only very rarely. This, however, might be a “statistical artefact”: all projects are put through the same approval process; it is only afterwards that projects are assigned to a specific fund (according to the content of an application). Therefore, the assignment of rejected proposals to the different types of funds might conceivably be somewhat arbitrary. ITF funds, however, do exhibit a significant difference: their funding rate is equal to their PV rate. The reason for this is that projects assigned to the ITF fund receive only non-refundable contributions; loans, either from the FFF or from private banks, are not part of their funding mix.

Figure 5: Applications, costs and funding rates by economic activity², 1998³-2003 (present)



Source: FFF; own calculations

According to content, projects are assigned codes for economic activity⁴. Most projects (20 %) are *computer-related* projects (NACE 72), followed by NACE 29 (*machinery and equipment*; 16 %). The top 4 activities (which, besides the ones already mentioned, comprise *medical, precision, and optical instruments* and *chemicals and chemical products*, NACE 33 and 24) account for 52 % of all submitted projects, a share which rises to 67 % for the top 7 and to 75 % for the top 10 activities (in total, 47 different activities were mentioned).

² For a complete list of 2-digit NACE codes, see Annex A.

³ In 1995, the classification system was changed from BS68 to NACE. However, BS68 codes were obviously still used for project classification in 1995 and, though less so, in 1996 and 1997. It was only in 1998 that the old BS68 codes seem to more or less vanish from the FFF data base. Consequently, for the part of the analysis dealing with economic activity, we restricted the time frame to the years 1998-2003.

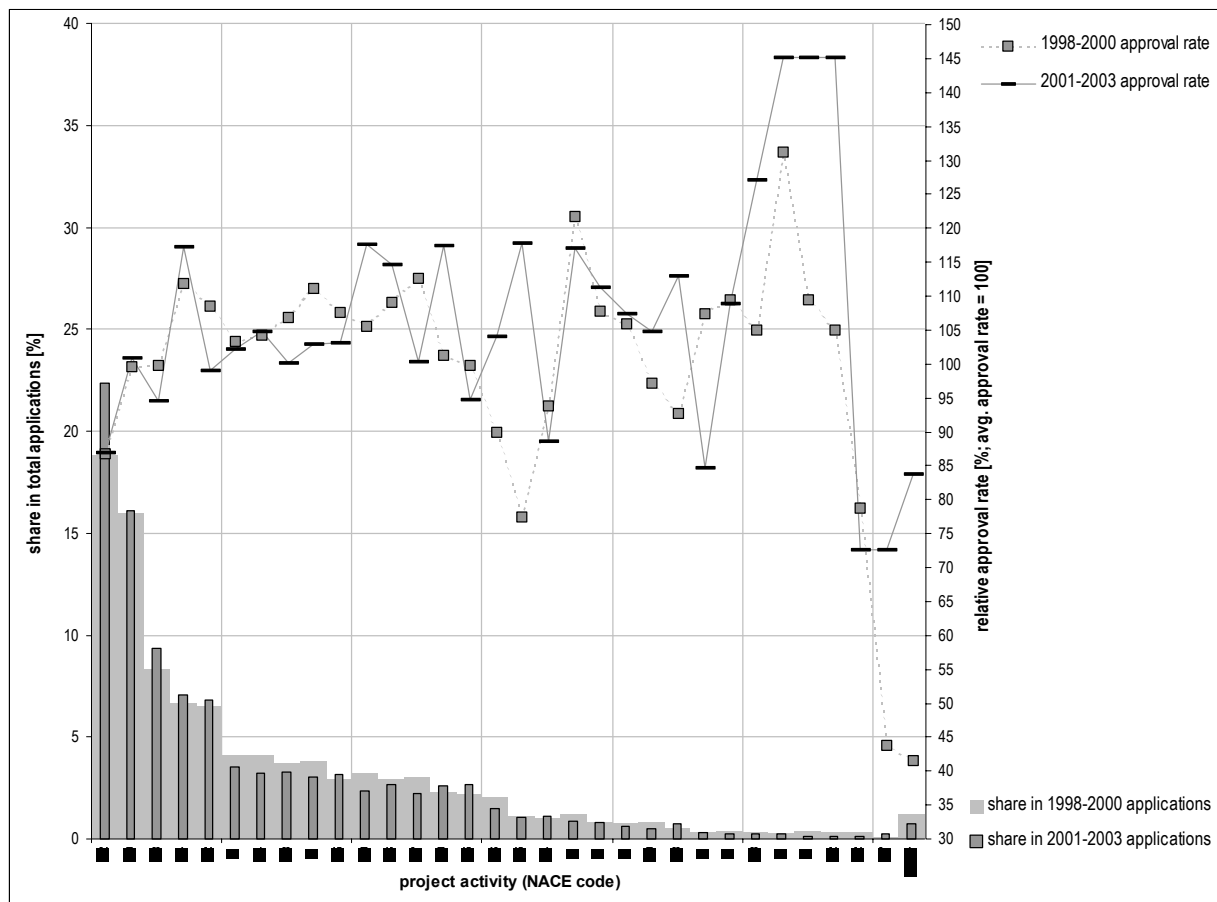
⁴ Unfortunately, the FFF assigns codes for economic activity only to the project, not to the applicant.

Projects with different activities show a huge variation in size: typical proposals to do with *chemicals and chemical products* and the *manufacture of radio, television and communication equipment and apparatus* (NACE 24 and 32) ask for more than 1.1 million €, more than double the overall average of 540 000 €. In contrast, projects in the top 3 activities, at under 450 k€, are comparatively low cost projects.

As for approval rates, these show a substantial variation as well. For the less important sectors of activity, this might well be due to the relatively small number of applications (the least numerous activities which are presented separately represent only 10 applications. In this case, a single rejected project reduces the approval rate by 10 percentage points). More interesting are the relatively low rates of approval for the top 3 activities, and especially for *computer-related* activities (NACE 72): a reason for the low approval rate for this activity might be that in this field, the boundary between innovations (especially process innovations) and investments is conceivably rather blurred – with the latter not included in the FFF’s task list (probably, unclear distinction between innovation and investment is also behind the below-average approval rates of activities #2 and 3, applications for *machinery and equipment* (NACE 29) and *medical, precision, and optical instruments* (NACE 33)).

Another reason, of course, might be that the high number of applications in the top-ranking activities allows (and, in way, even forces) the FFF to pick and choose. Figure 6 below examines changes in the structure of applications over time.

Figure 6: Applications and approvals by economic activity⁵, 1998⁶-2000 vs. 2001-2003 (present)



Source: FFF; own calculations

Although between 1998/2000 to 2001/2003, the ranking of activities remained more or less constant, the distribution of applications' activities became more unequal: the top 5 activities could all expand their share (in 2001/2003, together they accounted for 62 % of all applications, up from 56 % in 1998/2000). Relative approval rates⁷, however, remained roughly unchanged (it is only for less important activities which exhibit higher variation in their approval rates, which, however, might conceivably be attributable to their lower absolute number). This implies that changes in the activity mix of approved projects are predominantly the consequence of a shift in the activity mix of submitted projects, and not of a “conscious” decision on part of the FFF.

2.4. REJECTIONS AND APPROVALS: THE FIRM PERSPECTIVE

The last chapter looked at approval rates from a project's point of view. In this chapter, the firm perspective will be taken: what distinguishes successful from unsuccessful applicants?

As already mentioned, the data base contains 9967 applications. These were submitted by a total of 4028 different applicants. Only 8769 proposals, however, can be matched to firms which are included in the FFF's firm data base. 260 are submitted by co-operations of two or more firms, or *Arbeitsgemeinschaften*, *ARGEs*; the rest is filed by individuals, University institutes, or “very old firms”. Whatever the

⁵ For a complete list of 2-digit NACE codes, see Annex A.

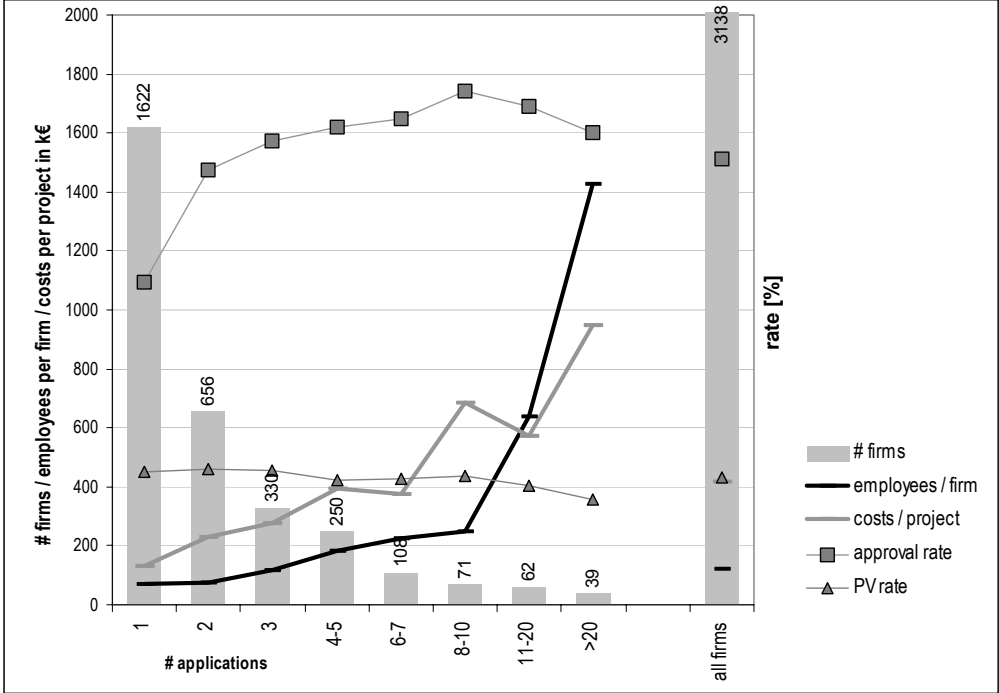
⁶ In 1995, the classification system was changed from BS68 to NACE. However, BS68 codes were obviously still used for project classification in 1995 and, though less so, in 1996 and 1997. It was only in 1998 that the old BS68 codes seem to more or less vanish from the FFF data base. Consequently, for the part of the analysis dealing with economic activity, we restricted the time frame to the years 1998-2003.

⁷ relative approval rates had to be used as the overall approval rate decreased, from 76 to 69 % of all applications.

reason, these projects cannot be analysed from the “firm perspective”. Consequently, they will be omitted from the following analysis. These 8769 proposals were submitted by 3138 different firms. Thus, the typical firm applies repeatedly for funding: on average, in 1995 to 2003, each firm applied 2.8 times.

Beneath this average, a huge variability is hidden: 1622 firms were one-time-only applicants. On the other hand, 39 firms submitted more than 20 applications (one firm even boasts 110 proposals). The following Figure 7 shows the statistics.

Figure 7: Repeated Applications, 1995-2003 (present)



Source: FFF; own calculations

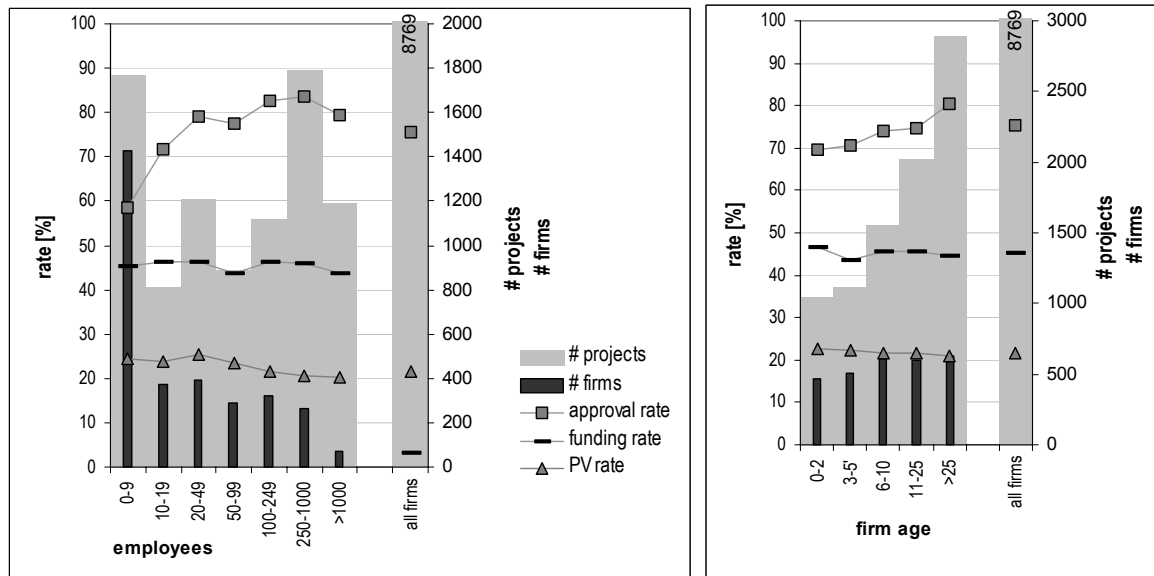
As noted above, more than half of all FFF customer firms have applied only once; including firms which applied twice, this share rises to almost 75 %. Not surprisingly, the number of applications correlates closely with firm size: one-time applicants employ on average 70 people; this number increases to more than 1400 employees in firms with more than 20 applications. Probably more surprisingly, costs per project also correlate quite closely with application numbers: larger firms not only apply more often, they also ask for more money per project. Single applications, on average, amount to 130 000 €; at firms with more than 20 submissions, the average project size is almost 950 000 € (the average price of all projects is 420 000 €). Additionally, the largest firms face above-average approval rates of 80 % vs. 75 % for all projects (although firms with 4 to 20 applications experience even higher rates).

The worst approval rate, at 55 %, is faced by firms with single applications. This, however, seems logical: such firms are smaller, probably younger, and conceivably less R&D intensive. As a consequence, they lack experience in dealing with the FFF (not to mention the possibility that their projects are of somewhat inferior quality).

The only index which unambiguously favours firms with fewer applications is the PV rate: at 23 %, this is above the average of 21.5 %. For firms with more than 20 applications, in contrast, the present value of FFF funding amounts to only 18 % of project costs.

In the light of the last paragraphs, the following Figure 8 offers only little surprising information:

Figure 8: Applications and approvals by firm size and age, 1995-2003 (present)



Source: FFF; own calculations

The left diagram shows applications and approvals for different firm size. As the number of applications per firm correlates closely with firm size, approval and funding rates show a pattern which is quite similar to Figure 7: the larger the firm, the higher the approval rate, and the lower the PV rate (though not consistently so). Most projects are submitted by small (<9 employees) and medium-large firms (250-1000 employees), each with a share of 18 %. The other size classes provide between 8 and 12 % of all applications.

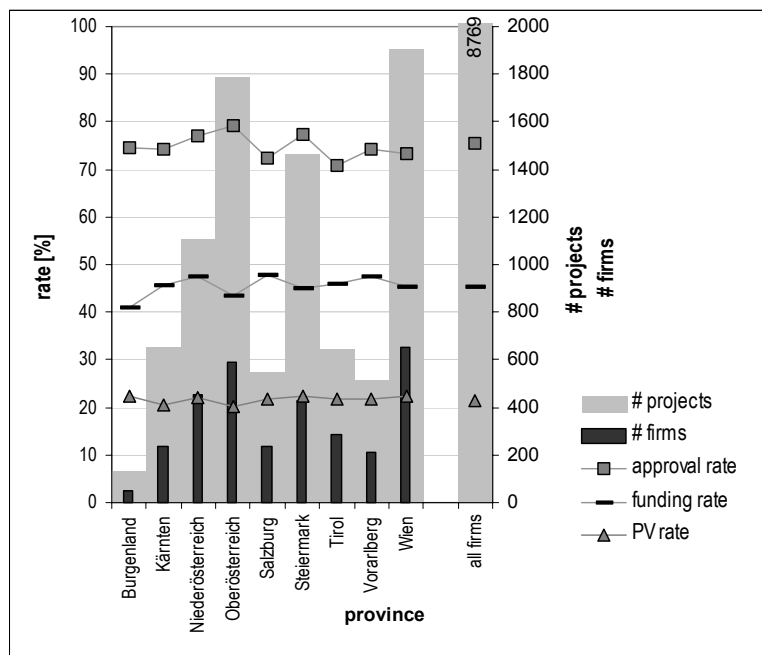
Most of FFF's customers (45 %) are small firms with an average of 1.2 projects each. The largest firms, with almost 17 projects per firm, represent only 2.2 % of all applicants (but submit 14 % of all applications).

The diagram to the right, applications and approvals by firm age, shows a similar picture: the older a firm, the higher the approval rate, but the lower the PV rate. Most projects (30 % of the total) are submitted by firms which are older than 25 years; a share which rises to more than half for firms older than 10 years.

The age structure of FFF customers is quite homogeneous⁸: each age group contains between 15 and 23 % of all applicants. Nevertheless, the average number of projects per firm correlates closely with firm age: start-ups (< 2 yrs old) submitted 1.6 projects each, whereas the oldest firms managed almost 4 applications. Again, this is straightforward: older firms have had more time and/or more experience in submitting proposals; additionally, they tend to be larger than new ones.

⁸ As firm age is defined as a firm's age at the time of an application, in this diagram, the total number of firms is larger than in the other diagrams. The reason for this is double counting: a firm which repeatedly applies for FFF funding can show up in different age groups.

Figure 9: Applications and approvals by firm location 1995-2003 (present)

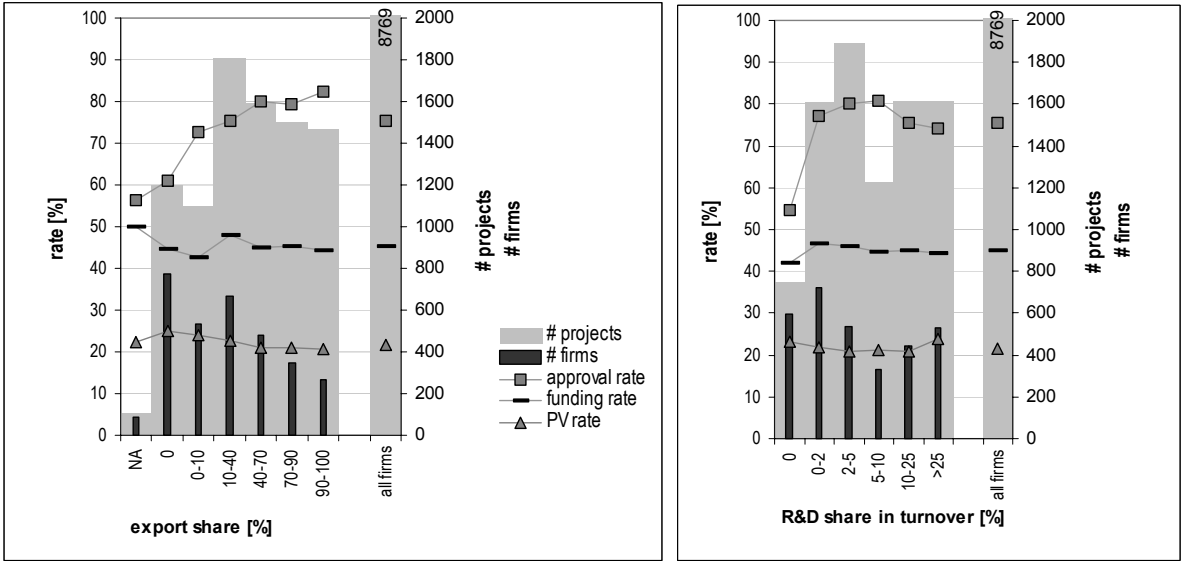


source: FFF; own calculations

Figure 9 shows the project distribution by geographical location of the applying firm. The provinces with most proposals are Wien and Oberösterreich with 20 and 18 % of all applications, respectively. The fewest proposals, less than 2 %, originate with firms from Burgenland. The variation in the number of projects per firm is not very pronounced and ranges from 2.4 in Niederösterreich to 3.4 in Steiermark. The same holds true for the provinces' approval and funding rates: both stay reasonably close to their respective Austrian values.

Projects by type of firm show a huge domination of *GmbHs* (limited companies), which account for 96 % of all proposals. They submit an average of 2.7 projects per firm. *AGs* (incorporated companies), though comparatively few in number (only 18 FFF customers belong to this type), submitted 228 projects, or almost 13 projects per firm. The other types are of marginal importance: together, they account for less than 1 % of all applications.

Figure 10: Applications and approvals by firms' export share and R&D share, 1995-2003 (present)



Source: FFF; own calculations

More export-oriented firms are definitely better FFF customers: the higher a firm's export share, the more project it submits, and the higher the approval rate it faces. The PV rate, however, is inversely related with export share.

The diagram to the right, however, shows a more egalitarian picture. Both approval and PV rates are much more homogeneous. With the exception of previously non-R&D performing firms (R&D share = 0), which get approval for only half of their projects, the approval rate is in the range of 74-81 %. The R&D newcomers, however, face above-average PV rates, second only to firms with R&D rates in excess of 25 %. Most applications per firm can be found in the 2-25 % classes. The reasons seem clear: firms with less than 2 % are certainly not in a position to accommodate high numbers of FFF funded projects, whereas highly research intensive firms, as a tendency, are younger.

2.5. REJECTIONS, APPROVALS, AND FUNDING: A MULTIVARIATE APPROACH

In the last chapter, applications and approvals were looked at from a variety of perspectives: size, content, and type of the projects; size, location, age, and type of applying firms. All perspectives, however, were one-dimensional: in what way are applications and approvals different with respect to only a single discriminatory variable?

This is necessarily incomplete: as we have seen, older firms face above-average approval rates. But is this due to their being more mature firms, or is it due to the fact that such firms tend to be bigger – because, as we also have seen, bigger firms also experience higher-than-average rates of approval. Of course, the “true” reason might also be a combination of these two factors.

To disentangle the effects of the various characteristics, pertaining both to a project itself, but also to its submitting firm, a multi-variate approach is called for. This is exactly what will be attempted with the following analysis.

All variables are defined as class variables. For some of the variables, this is straightforward: location and type of firm, for example, cannot be defined otherwise; the same is true for project characteristics such as the type of fund or whether it is a follow-up project or a completely new one. The reason for the classification of continuous variables, such as project size, company turnover, export share, etc. is that their inclusion as class variables allows for a highly “non-linear” response of the dependent variable, i.e. the decision of the FFF whether to approve of a project or not. For example, it might be that small companies face higher approval rates than medium companies, but lower rates than large companies. To allow for this response in a continuous setting, the size variable would have to be included in linear and quadratic, possibly even in cubic terms; and even then, the implicit “smooth” response of the decision with respect to the size variable might be a serious mis-specification. Including size as a class variable (e.g., small, medium-small, medium-large, large), though no cure-all, allows for a more flexible response.

The following Table 2 presents the definitions and descriptive statistics of the class variables used in the multivariate model. The dependent variable in this model is binary: it is 0 for rejected, 1 for approved projects. Accordingly, the model was estimated as a probit model (which among other desirable features restricts simulated values of the dependent variable to fall within the [0,1]-range).

In addition to the binary participation model, a linear model for the PV rate (the ratio of the funding’s present value to total project costs) will be presented.

Table 2: Descriptive statistics of model variables, 1998-2003 (present)

Variable	Description	Sum	Mean	Std. Dev.	Variable	Description	Sum	Mean	Std. Dev.
Granted	1 if project approved, 0 if rejected	4591	0.741	0.438	DAGE1	1 if firm age < 2 yrs	765	0.123	0.329
FORTS	1 if follow-up project, 0 if new proje	1801	0.291	0.454	DAGE2	1 if firm age 2-5 yrs	766	0.124	0.329
DY1998	1 if submitted in 1998	950	0.153	0.360	DAGE3	1 if firm age 5-10 yrs	1057	0.171	0.376
DY1999	1 if submitted in 1999	962	0.155	0.362	DAGE4	1 if firm age 10-25 yrs	1557	0.251	0.434
DY2000	1 if submitted in 2000	1149	0.185	0.389	DAGE5	1 if firm age > 25 yrs	1992	0.321	0.467
DY2001	1 if submitted in 2001	973	0.157	0.364	DBL1	1 if firm located in Burgenland	100	0.016	0.126
DY2002	1 if submitted in 2002	1118	0.180	0.385	DBL2	1 if firm located in Kärnten	469	0.076	0.264
DY2003	1 if submitted in 2003	1047	0.169	0.375	DBL3	1 if firm located in Niederösterreich	810	0.131	0.337
DNA01	1 if project activity is NACE01	39	0.006	0.079	DBL4	1 if firm located in Oberösterreich	1237	0.200	0.400
DNA02	1 if project activity is NACE02	12	0.002	0.044	DBL5	1 if firm located in Salzburg	381	0.061	0.240
DNA14	1 if project activity is NACE14	22	0.004	0.059	DBL6	1 if firm located in Steiermark	1023	0.165	0.371
DNA15	1 if project activity is NACE15	220	0.035	0.185	DBL7	1 if firm located in Tirol	442	0.071	0.257
DNA17	1 if project activity is NACE17	55	0.009	0.094	DBL8	1 if firm located in Vorarlberg	391	0.063	0.243
DNA18	1 if project activity is NACE18	16	0.003	0.051	DBL9	1 if firm located in Wien	1344	0.217	0.412
DNA19	1 if project activity is NACE19	18	0.003	0.054	DEXP1	1 if firm's export share = 0 %	843	0.136	0.343
DNA20	1 if project activity is NACE20	163	0.026	0.160	DEXP2	1 if firm's export share > 0, < 10 %	806	0.130	0.336
DNA21	1 if project activity is NACE21	38	0.006	0.078	DEXP3	1 if firm's export share 10-40 %	1272	0.205	0.404
DNA22	1 if project activity is NACE22	14	0.002	0.047	DEXP4	1 if firm's export share 40-70 %	1100	0.177	0.382
DNA23	1 if project activity is NACE23	15	0.002	0.049	DEXP5	1 if firm's export share 70-90 %	1017	0.164	0.370
DNA24	1 if project activity is NACE24	430	0.069	0.254	DEXP6	1 if firm's export share > 90 %	1053	0.170	0.376
DNA25	1 if project activity is NACE25	197	0.032	0.175	DFUE1	1 if firm's R&D share = 0	471	0.076	0.265
DNA26	1 if project activity is NACE26	186	0.030	0.171	DFUE2	1 if firm's R&D share > 0, < 2 %	1273	0.205	0.404
DNA27	1 if project activity is NACE27	172	0.028	0.164	DFUE3	1 if firm's R&D share 2-5 %	1403	0.226	0.418
DNA28	1 if project activity is NACE28	224	0.036	0.187	DFUE4	1 if firm's R&D share 5-10 %	822	0.133	0.339
DNA29	1 if project activity is NACE29	1030	0.166	0.372	DFUE5	1 if firm's R&D share 10-25 %	1084	0.175	0.380
DNA30	1 if project activity is NACE30	40	0.006	0.080	DFUE6	1 if firm's R&D share > 25 %	1146	0.185	0.388
DNA31	1 if project activity is NACE31	255	0.041	0.199	DGF1	1 if firm type is AG	157	0.025	0.157
DNA32	1 if project activity is NACE32	431	0.070	0.254	DGF2	1 if firm type is Einzelfirma	5	0.001	0.028
DNA33	1 if project activity is NACE33	542	0.087	0.282	DGF4	1 if firm type is GmbH	6012	0.970	0.171
DNA34	1 if project activity is NACE34	227	0.037	0.188	DGF5	1 if firm type is KG	19	0.003	0.055
DNA35	1 if project activity is NACE35	67	0.011	0.103	DGF6	1 if firm type is OHG	6	0.001	0.031
DNA36	1 if project activity is NACE36	107	0.017	0.130	DMA1	1 if firm size is < 10 employees	1277	0.206	0.404
DNA37	1 if project activity is NACE37	6	0.001	0.031	DMA2	1 if firm size is 10-19 employees	565	0.091	0.288
DNA45	1 if project activity is NACE45	136	0.022	0.146	DMA3	1 if firm size is 20-49 employees	912	0.147	0.354
DNA51	1 if project activity is NACE51	16	0.003	0.051	DMA4	1 if firm size is 50-99 employees	673	0.109	0.311
DNA72	1 if project activity is NACE72	1217	0.196	0.397	DMA5	1 if firm size is 100-249 employees	793	0.128	0.334
DNA73	1 if project activity is NACE73	36	0.006	0.076	DMA6	1 if firm size is 250-999 employees	1234	0.199	0.399
DNA74	1 if project activity is NACE74	64	0.010	0.101	DMA7	1 if firm size is > 1000 employees	745	0.120	0.325
DNA90	1 if project activity is NACE90	147	0.024	0.152					
DNAOTHER	1 if project activity is 'other'	57	0.009	0.095					
DK1	1 if project costs < 50 k€	625	0.101	0.301					
DK2	1 if project costs 50-100 k€	755	0.122	0.327					
DK3	1 if project costs 100-250 k€	1956	0.316	0.465					
DK4	1 if project costs 500-1000 k€	1461	0.236	0.424					
DK5	1 if project costs > 1000 k€	816	0.132	0.338					
DK6	1 if project costs 50-100 k€	586	0.095	0.293					
DF1	1 if application for FFF funds	5396	0.870	0.336					
DF2	1 if application for OeNB funds	480	0.077	0.267					
DF10	1 if application for ITF funds	323	0.052	0.222					

Source: FFF; own calculations

2.5.1 A binary participation model

The model was estimated for applications submitted from 1998 to 2003. The first three years, 1995-1997, had to be discarded due to problems with the correct classification of projects' economic activity (cf. above). Results of the estimation process are presented in Table 3.

Table 3: Estimation results of the probit model of participation

Dependent Variable: Decision01
Method: ML - Binary Probit (Quadratic hill climbing)
Included observations: 6199 after adjusting endpoints

Variable	Coeff	S.E.	z-stat	prob	
constant	-0.366	0.299	-1.225	0.22	
follow-up?	0.629	0.047	13.399	0.00	
decision year	1999	-0.088	0.070	-1.252	0.21
	2000	-0.200	0.066	-3.030	0.00
	2001	-0.280	0.069	-4.070	0.00
	2002	-0.520	0.066	-7.912	0.00
	2003	-0.346	0.068	-5.088	0.00
	activity of proposed project (2-digit NACE-code)	1	0.573	0.296	1.936
2		0.369	0.455	0.812	0.42
14		0.408	0.351	1.163	0.24
15		0.789	0.203	3.890	0.00
17		0.690	0.269	2.570	0.01
18		8.015	#####	0.000	1.00
19		0.695	0.414	1.677	0.09
20		0.781	0.210	3.720	0.00
21		1.008	0.327	3.078	0.00
22		0.796	0.474	1.677	0.09
23		1.033	0.458	2.255	0.02
24		0.649	0.191	3.403	0.00
25		0.599	0.204	2.929	0.00
26		0.777	0.208	3.730	0.00
27		0.463	0.211	2.195	0.03
28		0.534	0.200	2.676	0.01
29		0.478	0.180	2.650	0.01
30		0.434	0.279	1.552	0.12
31		0.512	0.197	2.603	0.01
32		0.359	0.190	1.894	0.06
33		0.453	0.185	2.448	0.01
34		0.422	0.204	2.070	0.04
35		0.648	0.264	2.450	0.01
36		0.516	0.221	2.335	0.02
37		0.553	0.563	0.981	0.33
45		0.534	0.211	2.533	0.01
51		1.123	0.491	2.285	0.02
72		0.361	0.179	2.015	0.04
73		0.792	0.297	2.667	0.01
74		0.338	0.255	1.325	0.19
90	0.830	0.213	3.902	0.00	
project costs	50-100 k€	-0.064	0.078	-0.827	0.41
	100-250 k€	0.009	0.066	0.137	0.89
	250-500 k€	0.025	0.069	0.356	0.72
	500-1000 k€	0.008	0.077	0.107	0.91
	>1000 k€	0.030	0.084	0.362	0.72
type of fund	OeNB	0.910	0.112	8.117	0.00
	ITF	0.340	0.091	3.752	0.00

Variable	Coeff	S.E.	z-stat	prob	
firm age	2-5 years	-0.117	0.072	-1.625	0.10
	5-10 years	-0.160	0.070	-2.289	0.02
	10-25 years	-0.193	0.067	-2.873	0.00
	> 25 years	-0.149	0.073	-2.044	0.04
firm based in ...	Kärnten	0.046	0.160	0.286	0.77
	Niederösterreich	0.151	0.154	0.982	0.33
	Oberösterreich	0.139	0.152	0.915	0.36
	Salzburg	0.029	0.162	0.181	0.86
	Steiermark	0.131	0.153	0.857	0.39
	Tirol	0.000	0.161	0.002	1.00
Vorarlberg		-0.042	0.162	-0.262	0.79
	Wien	0.022	0.151	0.148	0.88
export share	0-10 %	0.175	0.070	2.511	0.01
	10-40 %+	0.230	0.064	3.578	0.00
	40-70 %	0.274	0.071	3.828	0.00
	70-90 %	0.223	0.076	2.912	0.00
	> 90 %	0.297	0.079	3.778	0.00
R&D share in turnover	0-2 %	0.338	0.080	4.209	0.00
	2-5 %	0.372	0.081	4.584	0.00
	5-10 %	0.490	0.087	5.617	0.00
	10-25 %	0.436	0.082	5.339	0.00
	> 25 %	0.536	0.081	6.647	0.00
firm type	Einzelfirma	-8.334	#####	0.000	1.00
	GmbH	-0.296	0.153	-1.934	0.05
	KG	-0.391	0.341	-1.146	0.25
	OHG	0.069	0.640	0.108	0.91
firm size (# employees)	10-19	0.299	0.071	4.237	0.00
	20-49	0.524	0.067	7.809	0.00
	50-99	0.456	0.077	5.964	0.00
	100-249	0.496	0.079	6.260	0.00
	250-999	0.434	0.077	5.636	0.00
	> 1000	0.231	0.087	2.666	0.01

McFadden R-squared	0.128	
Obs with Dep=0	1608	Total obs
Obs with Dep=1	4591	6199

coefficient significant at 10% level
coefficient value significant at 5% level

Source: FFF; own calculations

The fit of the model, at a (McFadden-)R² of 0.13, is not overwhelming, although a majority of coefficients is significantly different from zero at least at the 10 %-level^{9,10}.

Coefficients have to be interpreted as the effects of a project's structure from the (virtual) "standard project" implied by the model specification: this standard project has all characteristics which are NOT explicitly included in the model. Therefore, the standard project is a new project, submitted in 1998, pursues some "other activity", and asks for less than 50 000 € from FFF funds. The submitting firm is a start-up (less than 2 years old), is located in Burgenland, and has zero export share. Previous R&D rates are zero, and is an AG with less than 10 employees.

Some of the coefficients give rise to a revision of the results from one-dimensional analysis: approval rates, for example, seem not to depend on the costs of the submitted project, despite earlier result indicating such a dependency (cf. Figure 4). One of the reasons that small projects are more likely to be rejected is that only 8 % of them are follow-up projects, whereas for other projects, this share is 29 % (Recall that follow-up projects face significantly lower rejection rates than new projects, cf. Figure 2). Also, the share of applications from computer-related activities (NACE 72), at 25 %, is higher than the respective overall share of 19 %. This attribute also lowers the chances of approval.

As it is, the parameter values – aside from their sign – are hard to interpret: a value of 0.63 for the follow up-dummy does not imply that such a project faces chances which are 63 %-points better than the chances of a new project. Such "percentages" have to be calculated by simulations. The following Table 4 gives the results of just such an exercise.

The decision year clearly makes a difference: except for 1999, all other year dummies are significantly negative, implying rising rejection rates (in 2002 and 2003, approval rates were 17 resp. 12 %-points less than in the benchmark year of 1998). As for the activity dummies, all of them exhibit positive signs, most of them significantly so. This means that the benchmark activity ("other activity") faces the least chances of approval. Differences are between 13 and 40 % (as a tendency, at the upper end of this range activities with fewer applications can be found). As already mentioned, computer-related activities (NACE 72, with a share of 19 % of applications the single most important activity), at only 14 %, are at the lower end of this range.

Contrary to the one-dimensional results, project costs exert only small and insignificant influence on the decision. Both other types of fund, however, OeNB and ITF, offer significantly higher approval rates than the standard FFF fund (although caution should be taken in this interpretation: all three types of funds are subject to the same procedure. Projects do not even apply for a specific fund; rather, the type of fund is assigned after the decision process. In this light, the estimated differences between funds are probably spurious).

As for the applicant firm, being a start-up (less than 2 years old) is an advantage: applications from other firms are rejected more often (their approval rates are 4-7 % lower). The geographical location, however, does not make a significant difference (all coefficients are statistically undistinguishable from zero). Export orientation and (previous) R&D activities are both rewarded by higher approval rates: non-R&D performing firms are distinctly more often rejected than other firms, as are, though to a smaller extent,

⁹ A cautionary remark: it would be wrong to interpret the coefficients causally. For example, it would not be correct to say that as the dummy AGE30m has negative sign, the fact that a coordinator is younger than 30 years causes her project to face worse chances than if the coordinator were in her forties, say. The correct interpretation would be simply that young coordinators typically submit proposals which get rejected more often. This CAN be brought about by the coordinator's youth, of course, but also by some other fact: maybe young coordinators are not experienced enough to formulate winning projects. The model is not in a position to distinguish between these two (and possibly other) reasons for rejection.

In this sense, the estimated differences are EX-POST rather than EX-ANTE.

non-exporting firms. The type of firm is not a significant influence; most firms, however, are GmbHs: they account for some 97 % of all applications (AGs provide a further 2 %, the insignificant rest is submitted by firms of some other type).

The influence of firm size is more ambiguous: although all other firms face higher approval rates than the benchmark (0-9 employees), both small (10-19) and very large firms (> 1000 employees) are rejected somewhat more often than medium small to medium large firms.

Table 4: simulated percentage difference in approval rates vis a vis the “benchmark project”¹¹

Variable		% difference in approval rates	Variable		% difference in approval rates
	follow-up?	24.7			
decision year	1999	-3.2	firm age	2-5 years	-4.3
	2000	-7.1		5-10 years	-5.8
	2001	-9.8		10-25 years	-6.9
	2002	-16.9		> 25 years	-5.4
	2003	-11.9			
		1	22.5	firm based in ...	Kärnten
	2	14.4	Niederösterreich		5.8
	14	15.9	Oberösterreich		5.3
	15	30.7	Salzburg		1.1
	17	27.0	Steiermark		5.0
	18	64.3	Tirol		0.0
	19	27.2	Vorarlberg		-1.6
activity of proposed project (2-digit NACE-code)	20	30.4	Wien	0.8	
	21	38.2	export share	0-10 %	6.7
	22	30.9		10-40 %+	8.9
	23	39.1		40-70 %	10.6
	24	25.4		70-90 %	8.6
	25	23.5		> 90 %	11.5
	26	30.2	R&D share in turnover	0-2 %	13.2
	27	18.1		2-5 %	14.5
	28	21.0		5-10 %	19.2
	29	18.7		10-25 %	17.1
	30	17.0		> 25 %	21.0
	31	20.1	firm type	Einzelfirma	-35.7
	32	14.0		GmbH	-10.3
	33	17.7		KG	-13.3
34	16.5		OHG	2.6	
35	25.4	firm size (# employees)	10-19	11.6	
36	20.3		20-49	20.6	
37	21.7		50-99	17.9	
45	21.0		100-249	19.4	
51	41.8		250-999	17.0	
72	14.1		> 1000	8.9	
73	30.8				
74	13.2				
90	32.1				
project costs	50-100 k€	-2.4			
	100-250 k€	0.3			
	250-500 k€	0.9			
	500-1000 k€	0.3			
	>1000 k€	1.1			
type of fund	OeNB	35.0			
	ITF	13.2			

coefficient significant at 10% level
coefficient value significant at 5% level

Source: FFF; own calculations

¹⁰ A re-estimation of the model with only significant variables included yielded only marginally different coefficients.

¹¹ All percentage values must be interpreted “ceteris paribus”: a simple addition of a certain combination of variables is not permissible, because the probit model is NOT linear in its parameters! The effect of a certain combination would again have to be simulated.

2.5.2 A linear funding model

The last chapter aimed at disentangling the influences which firm and project characteristics exert on an application's probability of acceptance, the *approval rate*. In this chapter, a closer look will be taken on a specific attribute of accepted projects, the *PV rate*: the ratio of the present value of FFF's contributions to the total costs of an approved project. The PV rate is substantially lower than the funding rate (the ratio of FFF funding to total costs, typically 50 or 60 %, depending on the type of firm): as most projects are financed by a mix of non-refundable contributions (from the FFF) and refundable loans (either from the FFF or a private bank; cf. Figure 3 above), the Present Value of the approved subsidies is smaller than their nominal amount. The share of the non-refundable part depends positively on the FFF's assessment of a project's riskiness and technological "new-ness" and negatively on economic potential.

The structure of this model is similar to the participation model: it uses the same design matrix on which to regress the dependent variable. However, in this case the independent variable is no longer a binary 0/1-variable (approval/rejection) but rather a continuous variable, the PV rate (defined as $100 \cdot PV / \text{total costs}$). Accordingly, the model can be estimated using standard linear OLS. Here, interpretation is straightforward: the coefficients can be interpreted as the difference (in percentage points) in PV rate vis a vis the "benchmark project", which is the same as above: it is a new project, submitted in 1998, pursues some "other activity", and asks for less than 50 000 € from FFF funds. The submitting firm is a start-up (less than 2 years old), is located in Burgenland, and has zero export share. Previous R&D rates are zero, and is an AG with less than 10 employees.

The results of the estimation are presented in Table 5 below. With an R^2 of 0.25, the estimation shows satisfactory fit. Most of the coefficients are significant at least at the 10 % level.

Interestingly, although the participation model shows follow-up project to exhibit appreciably larger approval rates, follow-ups seem to face significantly lower PV rates, of about -2.5 %. Again contrary to the tendencies in approval rates, the PV rate seems to steadily rise over time: in 2003, it was a good 3 % higher than in the benchmark year of 1997.

The highest PV rates seem to be granted for projects in NACE activities 37, 14, 23, and 73. But as these activities are rather "rare" (each of these 4 activities represents less than 0.6 % of all projects), statistically this result might be less than solid.

Larger projects face uniformly higher PV rates, by a margin of some 1 to 2 %, than small projects, whereas firm age does not seem to make any difference (contrary to the one-dimensional analysis, where the PV rate seems to be inversely correlated with firm size, cf. Figure 8).

A clear pattern can be seen with respect to firm size: the larger a firm, the lower its project's PV rates (firms with more than 1000 employees are confronted with an almost 3 % lower PV rate than the smallest firms). A similar though less regular pattern applies to export and R&D share: the higher a firm's export share (or its R&D share), the lower the PV rate of its projects (for both characteristics, this tendency roughly mirrors approval rates: whereas approval rates tend to rise with export and R&D shares, PV rates fall).

Although a firm's location does not seem to bear significantly on its approval rates (see above), it seems to influence PV rates: all other provinces exhibit higher PV rates than the benchmark Burgenland, three of them significantly so (projects by Styrian firms by more than 3 %).

Table 5: Estimation results of the OLS model of the funding share (present value of funding as % of total project costs)

Dependent Variable: PV funding/Total Project Costs*100

Method: Least Squares

Included observations: 5164 after adjusting endpoints

	Variable	Coeff	S.E.	z-stat	prob		Variable	Coeff	S.E.	z-stat	prob
	constant	41.67	2.49	16.72	0.000						
	follow-up?	-2.38	0.32	-7.54	0.000						
decision year	1999	0.88	0.51	1.74	0.082	firm age	2-5 years	0.87	0.58	1.50	0.135
	2000	0.98	0.49	2.00	0.046		5-10 years	0.65	0.56	1.16	0.245
	2001	1.90	0.52	3.69	0.000		10-25 years	0.38	0.53	0.71	0.475
	2002	2.81	0.50	5.61	0.000		> 25 years	0.65	0.56	1.16	0.247
	2003	3.61	0.52	6.98	0.000		Kärnten	0.77	1.24	0.62	0.536
activity of proposed project (2-digit NACE-code)	1	7.56	2.40	3.15	0.002	firm based in ...	Niederösterreich	2.75	1.19	2.31	0.021
	2	11.82	4.18	2.83	0.005		Oberösterreich	1.88	1.18	1.60	0.110
	14	13.86	2.95	4.69	0.000		Salzburg	0.95	1.27	0.75	0.454
	15	7.33	1.82	4.02	0.000		Steiermark	3.42	1.18	2.89	0.004
	17	4.14	2.19	1.89	0.059		Tirol	2.22	1.26	1.76	0.079
	18	2.16	3.12	0.69	0.488	Vorarlberg	0.57	1.27	0.45	0.651	
	19	3.23	2.98	1.08	0.278	Wien	2.67	1.17	2.27	0.023	
	20	9.26	1.87	4.94	0.000	export share	0-10 %	-1.77	0.59	-2.98	0.003
	21	3.52	2.37	1.48	0.138		10-40 %+	-1.64	0.54	-3.01	0.003
	22	6.87	3.27	2.10	0.036		40-70 %	-2.33	0.58	-4.00	0.000
	23	11.94	3.19	3.74	0.000		70-90 %	-2.06	0.62	-3.31	0.001
	24	3.67	1.74	2.11	0.035	> 90 %	-2.11	0.62	-3.38	0.001	
	25	4.77	1.84	2.60	0.009	R&D share in turnover	0-2 %	-0.47	0.73	-0.65	0.515
	26	4.16	1.84	2.27	0.024		2-5 %	-2.18	0.72	-3.00	0.003
	27	5.65	1.88	3.01	0.003		5-10 %	-1.81	0.76	-2.39	0.017
	28	3.45	1.82	1.90	0.058		10-25 %	-1.96	0.74	-2.65	0.008
	29	3.44	1.69	2.03	0.043		> 25 %	0.32	0.73	0.45	0.655
	30	2.30	2.45	0.94	0.349	firm type	Einzelfirma	-28.49	10.22	-2.79	0.005
	31	2.86	1.80	1.59	0.113		GmbH	-21.09	1.06	-19.97	0.000
32	3.70	1.75	2.11	0.035	KG		-11.30	3.03	-3.73	0.000	
33	3.43	1.73	1.98	0.047	OHG		-14.67	4.34	-3.38	0.001	
34	3.41	1.84	1.86	0.063	firm size (# employees)	10-19	-0.50	0.61	-0.81	0.416	
35	4.04	2.16	1.87	0.062		20-49	-0.57	0.54	-1.05	0.294	
36	5.18	1.97	2.63	0.009		50-99	-1.62	0.62	-2.62	0.009	
37	17.46	6.08	2.87	0.004		100-249	-1.42	0.62	-2.28	0.022	
45	8.51	1.90	4.48	0.000		250-999	-2.84	0.61	-4.61	0.000	
51	8.64	3.19	2.71	0.007		> 1000	-2.79	0.68	-4.10	0.000	
72	4.12	1.70	2.43	0.015		R-squared		0.247			
73	10.26	2.43	4.21	0.000		Total Observations		5164			
74	6.84	2.25	3.04	0.002							
90	9.38	1.89	4.95	0.000							
type of fund	50-100 k€	0.95	0.61	1.56	0.118						
	100-250 k€	1.76	0.52	3.40	0.001						
	250-500 k€	1.71	0.54	3.19	0.001						
	500-1000 k€	1.90	0.60	3.20	0.001						
	>1000 k€	1.55	0.64	2.42	0.016						
	OeNB	-0.88	0.53	-1.67	0.095						
	ITF	0.67	0.73	0.92	0.360						

coefficient significant at 10% level
coefficient value significant at 5% level

Source: FFF; own calculations

3 Input Additionality

3.1. WHAT IS THE PROBLEM?

The main problem which shall be addressed in this section is the additionality (or, more precisely, the *input additionality*) of R&D subsidies: do public contributions to private research boost total private R&D expenditures - and if so, do they boost them by an amount which is larger than the amount of taxpayers' money which was used in this way?

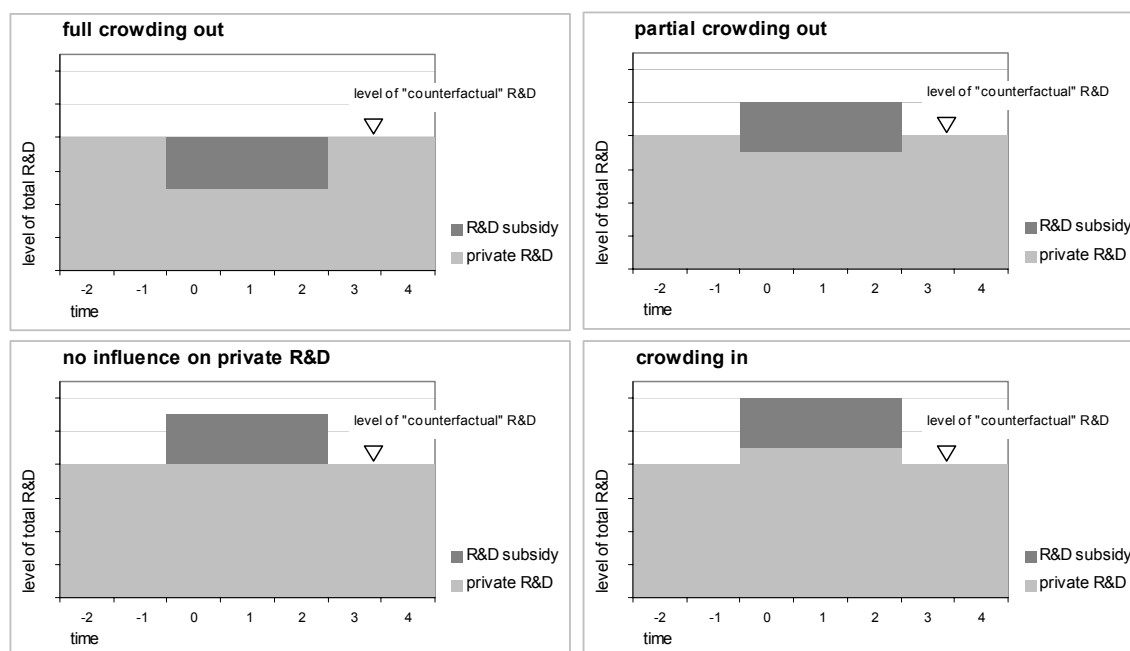
Besides input additionality, there are other concepts of "additionality" as well, notably *output* additionality (what is the effect of the subsidies research on a firm's turnover, profit, etc.) and *behavioural* additionality (in how far does the existence and availability of public subsidies alter firms' research decisions). In the first part of this section, however, only the question of input additionality will be addressed. Moreover, input additionality will be defined primarily in a *contemporaneous* way: what is the immediate effect of a subsidy on R&D expenditures? For reasons to do with data availability, the longer term (how total R&D expenditures are influenced by subsidies in the long run) will not be dealt with.

The chapter starts out with some brief (and rather non-technical) theoretical considerations about input additionality, followed by a chapter on the data used for its estimation. The main chapter presents the model and estimation results.

3.2. THEORETICAL CONSIDERATIONS

It is important to bear in mind that the level of R&D expenditures is the result of an internal decision process within in the firm; so are the reactions to R&D subsidies. Therefore, subsidies do not (or only partially) influence R&D directly, but rather indirectly: for the firm as a whole, the subsidy implies an outward shift of the budget constraint. The allocation of the additional funds within the firm, then, is subject to considerations involving "marginal benefit". Therefore, the effect of the subsidy on own R&D expenditures depends on many (internal and external) circumstances.

The following Figure 11 presents possible reactions of own R&D expenditures to a subsidy.

Figure 11: Effects of R&D Subsidies on Total R&D Expenditures¹²

Full crowding out occurs when firms perceive the subsidy as “windfall gains”: in the face of a subsidy, firms do not change their R&D plans, but rather use the subsidy to reduce their own spending¹³.

Partial crowding out occurs if firms raise their total R&D expenditures, but by less than the amount of the subsidy. This is probably the likeliest effect for firms which are not “liquidity constrained”, meaning that their R&D plans are not kept down by (external) budget constraints (e.g., the inability to get bank credit). In the presence of liquidity constraints, a possible reaction to a subsidy might be an *unchanged level of own R&D expenditures*: the firm would like to do more R&D than it is able to afford because of banks’ unwillingness to finance it. In this case, the firm would use the subsidy to extend total research by the full amount of the subsidy. If, additionally, the fact that the firm managed to secure a subsidy somehow results in a loosening of the liquidity constraint (if, say, banks perceive the grant as a positive signal, a “seal of quality”, which leads to an extension of the credit line), a result might be a *crowding in*.

Reasons for *crowding in* might also be found in the internal decision process. When a firm allocates its total budget to its different departments (marketing, production, research,...), the shares each department is awarded is the result of an internal “struggle” between departments. If, again, the R&D grant acts as a stamp of approval, this might improve the research department’s bargaining power, resulting in a larger budget share than would otherwise have been attainable.

¹² For simplicity, the level of the counterfactual R&D expenditures (i.e., those expenditures which would have been observed in the absence of any subsidy) is held constant over time

¹³ In the context of the present analysis, “more than full” crowding out can be ruled out: it would imply that firms reduce their own R&D expenditures by more than the amount of the subsidy; total R&D spending (own expenditures + subsidy) would fall. This has been demonstrated in only a few very special cases, notably the SEMATECH program, which was set up in the 1980s to co-ordinate the research efforts of US-american semiconductor firms in order to counter the “Japanese menace”. By reducing duplicate research, this programme seems to have had a (significant) negative influence on total R&D expenditures on the part of participating firms (see Irwin and Klenow, 1995).

3.2.1 Typical results: a quick literature survey

The econometric evidence of the substitutability or complementarity effects of public R&D funding is very inconclusive (following David et al. (2000), “substitutability” is taken to imply (even partial) crowding out; “complementarity” implies crowding in).

The empirical evidence on the effects of public subsidy is rather limited consisting of various ‘additionality studies’ with different methodological approaches (cf. David et al., 2000). However, to be able to provide a common background only the firm level studies are mentioned. One can think of, among others, Czarnitzki and Fier (2001), Meeusen and Janssens (2001), Lach (2000), and Irwin and Klenow (1996), which extend the important work of David et al. (2000).

The comparison of the company-level studies indicates the difficulties of measuring leverage effects: roughly half of the studies indicate complementarity and substitution respectively. An interesting difference, though, can be observed between European and US-american studies.

Table 6 shows the results of the 18 econometric studies split into European and US studies. The difference is highly visible. The total of studies with substitution effects is 7 whereof 6 are studies analyzing US data and only one is a European study. The contrary is the case with complementarity of public R&D funding, where 5 out of 7 studies comprise data from European countries. Referring to David et al. (2000) this could be partly due to the fact that US studies very often measure the impact of government *contract* R&D on private R&D spending, whereas in Europe firms get government grants and loans instead of direct R&D contracts.

Table 6: Econometric results, geographically differentiated

study results	substitutability	complementarity	mixed results
USA	6	2	3
Europe	1	5	1
Total	7	7	4

It has to be noted, though, that the firm-level studies employ different methods and look at different sets of data at different periods of time, thus are not strictly comparable.

As to the size of the additionality effect¹⁴, the studies in the survey exhibited a wide range of estimated values: this ranges from -6.5 (implying that an additional monetary unit of subsidies leads to a reduction of own R&D expenditures to the tune of 6.5 monetary units) to +8. Both extreme values look implausible: indeed, from the theoretical exposition above, a range of -1 (full crowding out) to, maybe, +2 or +3 seems more appropriate. Indeed, the -6.5 are the results of a study by Toivanen and Niinen (1998), in which they estimated additionality to be between -6.5 and +4.0, depending on firm type and specification¹⁵. The only other study to find more than full crowding out, at -2, is the SEMATECH-study by Irwin and Klenow (1996). In this case, the large negative effect seems more plausible¹⁶, as SEMATECH in essence constituted an R&D consortium: member firms pooled part of their R&D efforts. As this construction allowed for more efficient R&D in

¹⁴ Numerical values for the additionality effects, in the sense that “1 unit of subsidies leads to x units of additional own R&D expenditures”, could not be provided for all papers; David et al. (2000) do list additionality effects for most of the papers in their survey, but they included this effect as an *elasticity*, which is not very informative (the net effect is hard to estimate if the result is that “an additional 1 % of subsidies results in an additional 0.07 % of own R&D expenditures”)

¹⁵ Summing up, they conclude that “there is additionality for at least some firms”

¹⁶ Although their study drew some heavy criticism for comparing large firms within the SEMATECH consortium with small firms outside the consortium, thereby implying problems with selection bias.

the sense of a prevention of some duplicate R&D, the “fuller than full” crowding out could be the result of this increased efficiency.

Closest to full crowding out, at a reduction in own R&D expenditures of 82 cents for every dollar of R&D subsidies, came the study of the Small Business Innovation Research program (SBIR) by Wallsten (2000). His conclusion was that SBIR subsidies mainly financed R&D projects which would anyway have been undertaken by the funded firms, because the funded projects were highly successful in commercial terms.

Most other studies in the survey exhibited modest-to-fair amounts of crowding in, of between +0.1 to some +2.5 of additional own R&D expenditures for every unit of subsidies. The extreme value of +7 was estimated in a study of 86 Italian firms by Antonelli (1989).

3.3. THE DATA BASE

The data base was provided by the FFF¹⁷. It comprises two parts: the first part contains information on the 3138 firms which applied (whether successfully or not) for FFF-funding during 1995 and September of 2003. The second part consists of information on the 8769 projects for which applications were filed during the aforementioned period.

The firm level data contain information which has to be provided when submitting an application. This includes general firm characteristics:

Turnover, cashflow, exports, number of employees, year of foundation, legal form, and location

Besides, R&D specific variables are collected:

R&D expenditures and R&D personnel

This information has to be provided for the three years prior to the application of a project. After the submission of the project, no further data are collected on the firm level.

On the project level, the data include:

classification of the project according to the NACE-definition of economic activity, planned duration of the project, planned project costs (disaggregated into personnel, equipment, other), and, if appropriate, a reference to the original project (for applications requesting continued funding for longer projects).

For successful applications, additional data are included:

time period for which funding is granted (for longer projects, funding is typically not granted for the whole period. After the approved funding period, an application for continued funding has to be submitted), the total amount of funding (nominal and present value), and the “funding mix”.

The last point necessitates some explanation: typically, funding is granted to the tune of 50 % of a project’s costs¹⁸ (60 % in some cases). So, the *nominal* amount of funding is 50 (or 60) %. Most projects, however, are financed by a mix of non-refundable contributions (from the FFF) and refundable loans (either a subsidised loan from the FFF or a business loan from a private bank, in

¹⁷ The authors would like to thank Mag. Klaus Schnitzer and DI.Mag. Reinhard Zeilinger from the FFF for their co-operation.

¹⁸ These are “reviewed” costs: it is not necessarily the amount which the applicant asked for in his proposal. Rather, it is the costs which are “negotiated” between the applicant and the FFF.

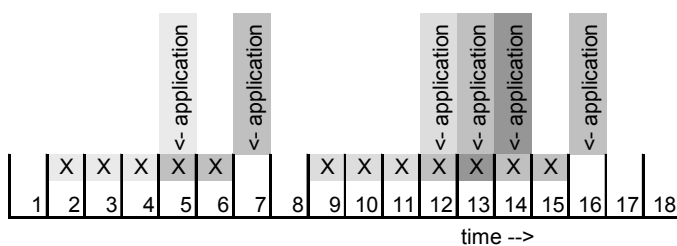
which case the FFF's contribution consists in a debt guarantee or in allowances towards the loan's annuities, or both); together, these finance instruments amount to the aforementioned 50 % of project costs. Therefore, the *present value* (PV) of the approved subsidies is smaller than their nominal amount. The share of the non-refundable part depends positively on the FFF's assessment of a project's riskiness and technological "new-ness" and negatively on economic potential. On average, the PV of funding represents 22 % of total project costs (or about 47 % of nominal subsidies). In all of the analyses, it is the reaction of R&D expenditures to this PV which will be of interest, not the reaction to the nominal amount.

From this description, two problems associated with this data base should be obvious. The first one has to do with the different periodicity of firm level and project level data: whereas the former contains (discrete) annual data, the latter is based on "continuous time": a project can start and end at any day (or, rather, month) of the year. To solve this discrepancy, the subsidies' PV is proportionally distributed over the approved funding period: for example, if the funding period starts in November of 1997 and ends in June of 1999, thus spanning 20 months, 10 % (i.e., 2/20) of total PV are counted as "funding in 1997", 60 % (=12/20) are assigned to 1998, leaving 30 % (=6/20) for 1999. This assumption of a linear deduction is certainly not "realistic" in the sense that firms use up their research funds in this linear fashion. However, given our ignorance about the "true" course of each project, this seemed to be the best solution (and it is certainly more realistic than simply allotting the whole amount to, e.g., the first project year).

The second problem is harder to solve: it has to do with the fact that from the way the firm level data are collected, firm level data and project level data cover completely separate periods: the firm level data span the three years *prior* to the project, leaving the period when the firm actually receives funding completely uncovered - not a very promising situation to start from when trying to estimate the effect of funding on the firms' total R&D expenditures.

To solve this paradox, we have to rely on firms which have repeatedly applied for funding. For such firms, overlapping time series of both R&D and funding data might be constructed in the following way: say, a firm had applied for funding in 1997. This would imply that this firm had to report company statistics for the years 1994-1996. If this were the last application this particular firm had made, it would be the end of the story. If, on the other hand, this firm again approached the FFF in, say, the year 2000, the company statistics for the years 1997-1999, which the firm would have to report for the new application, could be used to obtain the information necessary for the evaluation of the project applied for in 1996; in an athletic analogy, this might be termed "relay method".

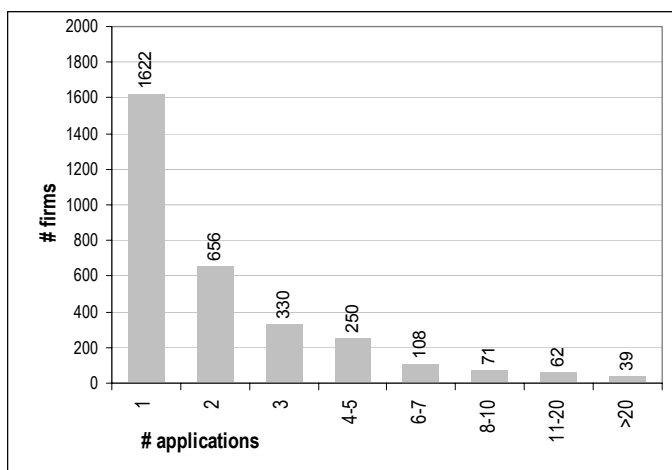
Figure 12: Constructing time series by the „Relay method“



X ... data available

Firms with repeated applications to the FFF are quite numerous: on average, each firm submitted almost 3 projects. The following Figure 13 shows a histogram of such "repeat offenders"

Figure 13: Number of firms by number of applications to the FFF, 1995-2003



Source: FFF; own calculations

Clearly, all 1600 firms (representing one half of all firms) which only applied once are “lost” for this analysis¹⁹. Unfortunately, firms with repeated applications also do not automatically qualify for inclusion in this analysis: their applications have to be “close enough” to provide for the required overlap of firm- and funding data (as an example: suppose in 1996 a firm submitted and started a project which lasted for one year. If this firm then came back with another application in 2002, this would be too late: firm level data would be available for 1993-1995 and then again from 1999-2001, whereas funding data would cover the years 1996 and 1997).

Applied to the FFF data base, this method produces a total of some 1100 firms, for which contemporaneous firm- and project data are available for at least one year (for different reasons, the sample of firms actually included in the analyses will be smaller still).

3.4. THE MODEL

Given the type of data as described in the previous section (time series data on quite a large number of individual firms), a logical framework for the estimation of the effect of FFF subsidies on firms’ R&D expenditures is given by panel regressions. Under the assumption that (known and unknown) characteristics influence firms’ R&D behaviour in a firm-specific but time-invariant way, incorporating firm fixed effects (i.e., a different constant for every cross-section unit) allow for the implicit modelling of these characteristics. This is quite convenient: although the data base contains information on some firm characteristics (turnover, export share, employees), most variables which might exert some influence are missing (most notably, firms’ sector of activity). In the fixed-effects framework, such unobserved but time-invariant variables should be captured by the inclusion of firm-specific fixed effects.

Additionally, this model allows for every firm to act, in a way, as its own “control firm”, in effect providing information on the firm’s behaviour vis a vis different levels of support. This allows to overcome a major problem of the data base, the almost complete absence of firms which have some R&D activities but which did not get any subsidy.

¹⁹ This is a pity because such “one-timers” conceivably represent a type of firm - those who only occasionally perform R&D - whose reaction to R&D subsidies is especially interesting. To facilitate future analyses, the FFF might contemplate to collect firm level data not only before the start of a project, but at the end as well.

Given the enormous range of firms in the database (from “owner-only” firms to companies with a couple of thousand employees), adequate correction for any potential (non-time invariant) size-effect must certainly accounted for. It was found that a polynomial in annual turnover (averaged over two years) seems to provide this correction. Using the number of employees instead of annual turnover yielded roughly the same results; however, as turnover is the one variable which is available for every firm (data on employees were missing in about 5 % of firm-years), turnover was used in the final specification²⁰. Once size was “sufficiently” provided for, the inclusion of additional variables (employees, export share) seemed not to make much difference to the estimation results.

To allow for the disregard for the calendar year of the typical R&D project, lagged R&D expenditures are included. Lastly, year dummies were included to account for the panel’s “unbalancedness”: as data are not available for all firms and all years, each year’s data comprise a slightly different sample of firms. As a tendency, the larger the firm, the more data-years are available.

The model, then, can be written as

$$R \& D_{i,t} = \lambda R \& D_{i,t-1} + \alpha \text{subsidies}_{i,t} + \sum_{k=1}^4 \beta_k ((\text{turnover}_{i,t} + \text{turnover}_{i,t-1})/2)^k + \\ + \sum_{t=1998}^{2002} D_t + \gamma_i + \varepsilon_{i,t}$$

The model was estimated for the years 1997-2002. Although project data were available since 1995, the years 1995 and 1996 were not used in the estimation process. The reason for this is the fact that the typical period for which FFF funding is provided is about 18 months. Therefore it cannot be ruled out (in fact, it is more than likely) that pre-1995 funding persists in the following years. To prevent this unknown source of funding from “contaminating” the estimates, the first two years were dropped.

The final sample comprised 495 firms. These were selected according to the following criteria:

- a minimum of 4 observations in 1997-2002, to preserve the “time-series” flavour of the panel regression
- no “problematic” values of their R&D expenditures, defined as an amount of R&D expenditure which is less than the contemporaneous amount of (approved) project costs as recorded in the data base.
- no “problematic” values of annual turnover. A few firms reported sales which amount to more than a million euro per employee. Although such values are not strictly impossible, they were interpreted as indicators of possibly erroneous data (the cut-off was actually set at 500 000 euro/employee).
- included were only firms which consistently reported positive R&D expenditures. The reason behind this restriction is the idea that habitual R&D performers react differently to R&D subsidies than intermittent performers: as an extreme case, suppose a firm had performed only a single R&D project which was supported by the FFF. This firm, then,

²⁰ Cashflow, which also would have been available, was disregarded. The reason for this is that quite a few definitions of cashflow are in use; it was not clear whether all firms used the same one. Additionally, as cashflow is an accounting concept similar to profits, R&D

should exhibit R&D expenditures which are about twice the nominal amount of the granted sum (typically, 50 % of project costs are covered by FFF subsidies) and about 4-5 times the amount of the subsidy's present value (as the typical funding mix consists of grants and loans, the present value, at about 22 % on average, is less than the nominal amount). For this reason, the effect of FFF funding on non-habitual R&D performers is suspected to be larger than for firms which perform R&D on a more regular basis²¹.

Altogether, 495 firms fulfilled the complete set of criteria, 35 of which did not receive any FFF funding during the observation period (despite their being regular R&D performers).

Using this set of 495 firms, the model was estimated using GLS with cross-section weighting. The results are presented in Table 7 below.

Table 7: Results of the fixed-effects panel regression

<i>dependent variable:</i>	R&D expenditure		
<i>estimation period:</i>	1997-2002		
<i>estimation method:</i>	GLS (cross-section weights)		
	<i>coefficient</i>	<i>s.e.</i>	<i>prob-value</i>
R&D expenditure(-1)	0.701	0.02	0.00
FFF funding (present value)	1.400	0.07	0.00
(turnover+turnover(-1))	0.008	0.00	0.00
(turnover+turnover(-1)) ²	-2.80E-08	7.40E-09	0.00
(turnover+turnover(-1)) ³	5.40E-14	1.28E-14	0.00
(turnover+turnover(-1)) ⁴	-1.40E-20	3.98E-21	0.00
Dummy 1998	-18.10	3.53	0.00
Dummy 1999	-13.34	3.83	0.00
Dummy 2000	-27.91	4.19	0.00
Dummy 2001	-85.36	8.38	0.00
Dummy 2002	89.92	15.04	0.00
# cross-section units	495		
# observations	2194		

Source: FFF data base; own calculations

According to the estimation results, one additional Euro of funding (or, rather, of its present value) leads to an increase in (total) R&D expenditures of 1.40 Euros – or, put differently, an additional 40 cents of private R&D expenditures for each Euro of funding. FFF funding and private R&D, therefore, seem to be complementary, with a “leverage effect” of about 40 %. The complementarity can only be established for the *present value* of FFF funding: for the *nominal* amount of FFF subsidies, a substitution effect has to be admitted (the present value being about half of the nominal subsidy, the coefficient of the nominal funding would be calculated at about 0.7; a re-estimation of the model using nominal funding instead of the present value confirms this value).

All coefficients are highly significant. With an estimated standard error of 0.07, the 95 %-range of the funding coefficient is about 1.26-1.54, comfortably above a value of 1.0 which would constitute the boundary between “substitutability” and “complementarity”: if the coefficient were less than 1.0, it would have to be concluded that firms substitute R&D subsidies for own expenditures (at least partially).

expenditures, being a cost component, enter the calculation of cashflow, thus potentially introducing problems with “simultaneity”.

²¹ Estimation results seem to vindicate this assumption: for the 33 firms which were identified as “intermittent R&D performers” (defined as firms which report at least 2 years of zero R&D), the level of additionality is indeed estimated to be considerably higher; cf. below.

3.4.1 Standard errors revisited: a bootstrap approach

This happy result, that the funding is complementary “almost certainly”, warrants some cautionary remarks: it is estimated under the usual assumption that the residuals are normally distributed. Conceivably, this might be implausible: the estimation is performed on the *level* of R&D expenditures as well as turnover; even if the size effect is properly taken care of, the residuals certainly remain affected by the level of a firm’s typical R&D expenditures. Additionally, they are probably distributed as log-normal rather than normal (if they are normally distributed at all). To assess the extent to which the standard errors’ estimates might be biased, a bootstrap exercise was performed. In this, the original sample of 495 firms was resampled 1000 times. For each bootstrapped sample, the model was re-estimated. The distribution of the 1000 coefficients of FFF funding were then statistically analysed.

The following Table 8 presents the summary statistics of this bootstrap.

Table 8: Results of the bootstrapping: descriptive statistics and Kernel density approximation

	bootstrapped funding coefficient (n=1000)
Mean	1.497
Median	1.422
Std. Dev.	0.594
Maximum	4.151
Minimum	-0.122
2.5% limit	0.534
97.5% limit	2.771
Jarque-Bera	95.153

The results are clear: the average of the bootstrapped coefficients, at about 1.50, is somewhat higher than the point estimate for the whole sample of 495 firms. The real difference, however, can be found for the estimated range of the funding coefficient: whereas our original results indicated a narrow range of [1.26,1.54] for the 95 % interval, the bootstrap yields, also at the 95 % level, a range of [0.53, 2.77] – in other words, more than 10 times the range of the original estimate. Accordingly, the previous conclusion that complementarity of FFF funding is “almost certain” (provided that our model is valid, of course), has to be downgraded to “most probably” (about 80 % of the bootstrapped coefficients are larger than 1).

In the next two chapters, a closer look will be taken at two specific questions. The one is if additionality is a matter of firm size; the other, whether firms which do not perform R&D on a continuous basis exhibit different reactions to FFF funding.

3.4.2 Additionality – a function of firm size?

For the whole sample of 495 firms, a leverage of about 40 % was estimated. In this chapter, the model will be re-estimated for samples of firms of different size. Firms were assigned to 4 different categories according to the average reported number of employees: less than 10, 10-50, 50-250, more than 250. The following Table 9 gives the results from the 4 panel regressions.

Table 9: Model results disaggregated by firm size

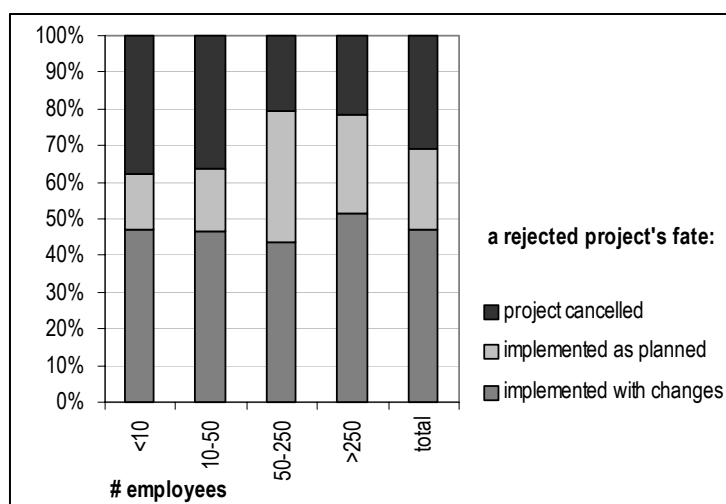
	< 10 employees		10-50 employees		50-250 employees		>250 employees	
	coefficient	s.e.	coefficient	s.e.	coefficient	s.e.	coefficient	s.e.
R&D expenditure(-1)	0.088	0.04	0.413	0.03	0.458	0.04	0.771	0.03
FFF funding (present value)	1.621	0.15	1.293	0.09	1.048	0.17	1.955	0.26
(turnover+turnover(-1))	0.153	0.06	0.046	0.02	0.022	0.02	0.003	0.00
(turnover+turnover(-1)) ²	-1.9E-05	1.35E-04	5.9E-06	5.91E-06	2.3E-06	7.77E-07	-1.6E-08	1.11E-08
(turnover+turnover(-1)) ³	-5.4E-08	9.72E-08	-9.9E-10	6.62E-10	-5.5E-11	1.26E-11	3.8E-14	1.59E-14
(turnover+turnover(-1)) ⁴	1.4E-11	1.78E-11	3.4E-14	2.38E-14	2.6E-16	6.17E-17	-9.9E-21	4.73E-21
Dummy 1998	-1.454	4.42	8.818	4.80	7.137	12.85	-42.569	5.94
Dummy 1999	3.924	4.89	3.029	5.12	-18.311	21.66	-42.933	22.80
Dummy 2000	10.019	5.02	9.139	5.33	-55.753	24.44	-5.919	26.30
Dummy 2001	15.679	5.41	17.954	6.66	-62.808	29.33	-202.459	54.47
Dummy 2002	11.650	5.90	22.043	9.20	-98.867	27.53	120.214	85.19
# cross-section units	66		146		143		136	
# observations	259		607		651		746	

bold numbers indicate significance at the 10 % level

All firm sizes exhibit complementarity, though to a differing degree. Interestingly, it is the smallest and the largest firms which exhibit the highest leverage, medium-sized firms show only small additionality. At first sight, this is puzzling: one would probably suspect a homogeneously falling reaction of additionality with respect to firm size.

Candidates for the solutions to this puzzle can come from various corners. The first – and easiest – is certainly that this result is something of a statistical artefact, probably due to an inadequate model²². Of course, this cannot be ruled out. On the other hand, this U-shaped function reflects results from a survey which was conducted specifically for the purpose of evaluating the “behavioural effects” of FFF funding. In this survey, firms were asked – among other things - whether rejected projects were conducted despite this negative decision, if they were conducted in a modified way, or not at all. Interestingly, the answers, presented in Figure 14, show a similar pattern when disaggregated into size classes.

Figure 14: Implementation/non-implementation if application was rejected: analysis by firm-size



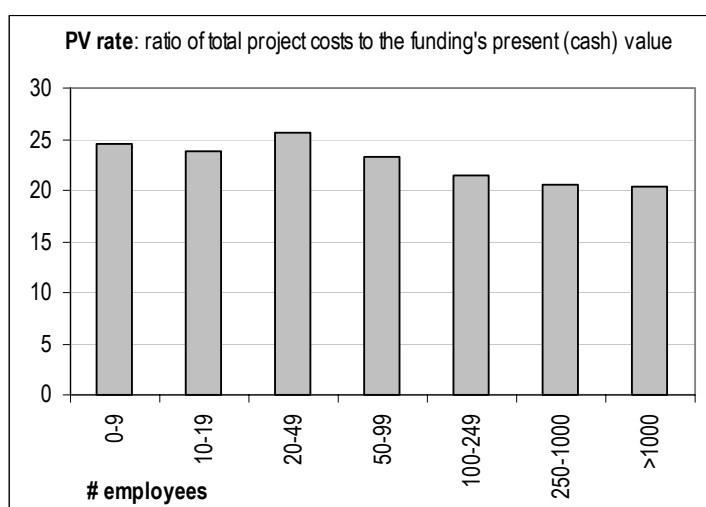
Source: Survey

²² Although different model specifications exhibited similar patterns of additionality with respect to firm size.

According to the survey, medium-sized firms, more often than large and much more often than smaller firms, report that they implemented a project “without changes” even when confronted with a rejection. They also cancelled projects less often (although only slightly less often than large firms)²³. Although this result does not “explain away” the whole difference in additionality between firms of different sizes, it seems to hint at similar tendencies.

A third explanation, then, is certainly the fact that large firms face a lower PV rate than smaller firms: the ratio of total project costs to the funding’s present value is some 15 % lower than that of medium-sized firms (which enjoy the highest PV rates). In fact, the PV rate, disaggregated by firm size resembles an inverted U-shape (as such, it is the mirror-image of the additionality by size class):

Figure 15: PV rates by size class



Source: FFF; own calculation

A lower PV rate, then, implies higher additionality for each project: as the FFF’s share in total project costs becomes smaller, the “funding leverage” becomes higher: an PV rate of 25 % implies a leverage of 400 %; an PV rate of 20 %, one of 500 %. Together with the different reaction to a rejection, this might provide a partial solution to the “leverage puzzle”.

3.4.3 Additionality in firms with sporadic R&D activities

In the estimation of the model as presented above, only such firms were included which performed R&D on a regular basis. A firm was categorized as “regularly R&D performing”, if in every year for which data on turnover were provided, a positive amount for R&D expenditures was reported. This restriction was justified on the ground that firms which regularly perform R&D might conceivably exhibit a weaker reaction to funding than firms which perform R&D only intermittently. In this chapter, this hypothesis shall now be dealt with.

For this, firms with intermittent R&D were identified as firms which reported zero R&D for at least one year for which data on turnover is available (and which fulfilled the other requirements

²³ From the cancellation rate as reported in the survey and a couple of bold assumptions, a (very) rough value for funding additionality can be derived: in total, about 30 % of rejected projects were aborted. Assume that 100 projects were submitted. If all had been rejected, only 70 of them would have been conducted (either unchanged or modified), 30 would have been dropped altogether. Suppose further, each project would carry costs of 100 euros. Then, with the 30 dropped projects, 3000 euros of R&D expenditures would not have been spent. The funding of all 100 projects, at an average NPV rate of 22 %, would have cost the FFF 2200 euros. This translates into an “R&D leverage” of $3000/2200 = 1.36$, or about 36 cent of additional private R&D for every euro of FFF subsidies. But, of course, this is only a naive assessment.

for “sensible data” as stated at the beginning of chapter 3.4). Additionally, firms with at least 2 and 3 years of zero R&D were identified. The number of such firms is small: only 61, 33, and 18 could be identified, respectively.

Table 10: Additionality of firms with intermittent R&D performance

<i>dependent variable:</i>		R&D expenditure					
<i>estimation period:</i>		1997-2002					
<i>estimation method:</i>		GLS (cross-section weights)					
	at least 1 year of 0 R&D		at least 2 years of 0 R&D		at least 3 years of 0 R&D		
	<i>coefficient</i>	<i>s.e.</i>	<i>coefficient</i>	<i>s.e.</i>	<i>coefficient</i>	<i>s.e.</i>	
R&D expenditure(-1)	0.236	0.08	-0.232	0.13	0.110	0.14	
FFF funding (present value)	1.560	0.29	6.603	1.06	3.571	0.67	
(turnover+turnover(-1))	0.039	0.03	0.509	0.10	0.161	0.07	
(turnover+turnover(-1)) ²	-2.6E-07	2.14E-06	-2.3E-05	5.00E-06	-8.4E-06	5.57E-06	
(turnover+turnover(-1)) ³	-1.6E-11	5.72E-11	4.3E-10	9.99E-11	2.7E-10	2.10E-10	
(turnover+turnover(-1)) ⁴	2.0E-16	4.60E-16	-2.8E-15	6.94E-16	-3.0E-15	2.49E-15	
Dummy 1998	28.976	24.09	8.785	19.37	-16.723	63.72	
Dummy 1999	83.721	22.90	47.019	29.78	-31.301	63.40	
Dummy 2000	89.017	22.95	30.593	33.83	30.103	64.64	
Dummy 2001	115.574	24.54	-28.556	48.75	-34.116	67.04	
Dummy 2002	181.217	42.48	51.380	87.71	49.112	99.49	
# cross-section units	61		33		18		
# observations	226		119		65		

From the results it could be inferred that firms which do not perform R&D on a regular basis indeed exhibit a more pronounced response to R&D subsidies: although firms with at least 1 R&D-free year show only slightly higher leverage than regular R&D performers (1.56 vs. 1.40), firms with 2 or 3 years of zero R&D are credited with markedly higher estimates. Nevertheless, it has to be borne in mind that the panels on which these results were obtained are quite small; the estimated standard errors of the funding coefficient are quite large, implying a wide confidence interval (which furthermore, as argued in chapter 3.4.1, is likely to be estimated as much too narrow).

3.4.4 Concluding remarks

What, now, can be said in answering the question whether FFF funding acts as a complement or a substitute to privately financed R&D? The evidence can be interpreted as leaning towards complementarity: the results of the panel regression model certainly point in this direction, even if a bootstrap exercise adds some qualifications to this result. As for the leverage of FFF funding’s numerical value, the analysis seems to place it at about 40 %; 1 additional euro of funding (or, to be more precise, of its present value) induces firms to contribute an additional 40 cents of their own money. Both very small and large firms seem to exhibit higher leverage, small and medium-sized firms smaller leverage. Additionally, the leverage estimates for firms which perform R&D only occasionally are higher than for regular R&D performers.

In the whole analysis, there are some sources of shakiness: first of all, funding from sources other than the FFF are unknown. Although the FFF is by far the most important source of public subsidies to private R&D, it is certainly not the only one. The direction of the bias thus introduced is not completely clear: if funding by the FFF and funding by other sources are positively contemporaneously correlated, the analysis is likely to overstate the complementary effect; conversely, a

negative correlation would dampen the estimated effect²⁴. Whatever the direction, this unknown influence is unlikely to completely alter the results of the analysis (after all, the FFF accounts for about 80 % of all public subsidies to private R&D). Also, the (somewhat arbitrary) linear distribution of the subsidies over the respective funding periods certainly introduces some (unavoidable) “fuzziness”. Longer time series would be of special importance to alleviate this problem by allowing some “averaging out” of mis-distribution introduced via the linear method.

The next is the choice of firms to enter the analysis. As it is, the standards for data quality are set rather high (at least 4 years of “sensible” data, etc.). As the bootstrapping exercise showed, even a varying set of firms which fulfill the same standard of data quality leads to a rather wide range of estimates. It shall not be concealed that the choice of functional form is of crucial importance as well; the results as presented in this paper are based on only one of quite a few specifications which were tested in the course of this work (albeit the one which was deemed to be the “best specification”, of course).

Using the lagged endogenous variable, furthermore, might introduce Nickell-Bias, resulting in additionality estimates which are probably too conservative. Also, the dynamic formulation of the model conceivably renders the results as the “lower limit” of additionality.

But whatever the “right” specification actually might turn out to be, any econometric analysis is bound to be confronted with a vast amount of “noise” hidden in this data base (which is not to say that the data are “inaccurate”; rather, too much of what actually gets on inside any firm and which influences the amount of realised R&D remains necessarily unknown).

²⁴ If funding by other sources coincides with FFF funding, the subsidies from the other sources would be part of total R&D expenditures, thus raising the estimated effect of FFF funding on total R&D expenditures. Conversely, if funding by the FFF and other sources took place in different years, the other sources would raise total R&D spending in years without FFF funding, thereby dampening the estimated “jump” in R&D expenditures resulting from FFF funding.

4 Output Additionality

The following chapter consists of two sub-sections. The first aims at analysing the factors explaining the intensity of R&D subsidies and the second investigates the productivity effects of both privately and publicly funded R&D.

4.1. FACTORS EXPLAINING THE LEVEL OF R&D SUBSIDIES

The decision on the level of R&D subsidies will be affected by both the funding agency's objectives and firm characteristics. Thus, the amount of R&D subsidies is not exogenous but endogenously determined by the funding agency's selection rule. According to technology guidelines and budget constraints, a funding agency decides which projects to subsidize, as well as what amount and what kind of subsidy the projects will receive. Based on certain criteria, the agency ranks the applications and funds the best project. According to the fund's strategy, the size, age, industry affiliation and performance of the firm as well as the firms' R&D intensity may all influence the level of R&D subsidies. Due to data limitations, we will investigate determinants of the level of R&D subsidies of supported firms rather than the factors influencing the probability to receive subsidies.²⁵

4.1.1 Hypotheses about the agency's allocation rule

The funding agency's strategy depends on a number of factors. In principle, the FFF's strategy is to open all industrial R&D projects. The FFF decides whether to give subsidies to a R&D project. If the decision is positive, the amount of the subsidies will be determined. The decision whether or not to subsidize an R&D project as well as the decision on the level of subsidies may depend on specific selection criteria. According to the FFF's guidelines, the fund encourages R&D in small firms and in start-up firms. Furthermore, a special grant program called "start-up" is established, favoring technology-oriented companies established during the past three years and consisting of less than 50 employees. The support of small firms can be justified by the presence of capital market imperfections and by the fact that R&D may involve fixed set-up costs, part of which may be sunk costs. Another grant program, "R&D dynamics", is designed to provide financing to firms with a lower-than-average R&D intensity but with high R&D growth in the past.

Another program initiative is the micro technology program. Therefore, we would expect industry variation as well. In particular because the potential for profitable R&D projects in micro technology is higher, we expect a higher R&D subsidy ratio in R&D-intensive industries using micro technologies such as the electronic industry and instruments, opposed to industries not using micro technologies. Other selection criteria taken into consideration include measures of firm performance such as sales growth rates and cash flow in the past. In order to test the impact of past firm performance, we include the cash flow to sales ratio (lagged two years) and the annual growth rate of total sales in the last two years as additional explanatory variables.

In the following, five hypotheses concerning the determinants of the R&D subsidy ratio are formulated that will be evaluated in the following empirical work:

²⁵ For an analysis of the probability of receiving subsidies for an R&D project based on the FFF project database, see section 2.2.

- H1: We expect that both small firms and young firms receive higher R&D subsidies to total R&D.
- H2: The R&D subsidy ratio decreases with increasing firm size.
- H3: The R&D subsidy ratio may be higher in the software (Nace 72) and the electronic industry (NACE 30-33) than in other industries.
- H4: Fast-growing firms as well as firms with a high cash flow sales ratio in the past are expected to receive higher amounts of R&D subsidies.
- H5: The R&D subsidy ratio is significantly negatively related to the current R&D intensity.

The general specification used in the following empirical implementation relates the log ratio of R&D subsidies to total R&D expenditures to log R&D intensity, average annual growth rate of total sales in the past two years, cash flow ratio in the past two years as well as a set of appropriate control variables. In addition, we control for time and fixed effects:

$$\log(RDSUB_{it} / RD_{it}) = \alpha_1 \log(RD_{it} / Y_{it}) + \alpha_2 ((\log(Y_{it}) - \log(Y_{it-2}))) + \alpha_3 CFR_{it-2} + \sum_{k=1}^K \beta_k X_{kit} + \theta_i + \lambda_t + \varepsilon_{it},$$

where i and t are indexes of firm and year, respectively. θ_i and λ_t denote fixed and time effects. $\log(RDSUB_{it} / RD_{it})$ is the natural logarithm of the ratio of R&D subsidies to total R&D expenditures. CFR_{it-2} denotes the cash flow to sales ratio, $\log(RD_{it} / Y_{it})$ is the natural logarithm of R&D intensity, measured as the ratio of R&D to turnover. $(\log(Y_{it}) - \log(Y_{it-2}))$ is the growth rate of total sales in the last two years and X_{kit} are other explanatory variables that are considered relevant. Appropriate control variables, which may affect the level of R&D subsidies, include firm age, legal status, industry affiliation and size dummies. In particular, firm age represents a good control variable because one of FFF's strategies is to support newly founded firms. Age effects are captured by a dummy variable indicating whether firms were founded less than six years before the survey year. Furthermore, we include a dummy variable indicating whether the firm is a company with limited liability (GmbH). Year dummies control for common shifts over time. In order to account for the firms' industry affiliation (i.e. types of activities), we use eleven industry dummies. Firm size is defined by the number of employees and firms are divided into six size classes: the reference group has less than 10 employees, the three medium-sized classes are defined as 10-24, 25-49, 50-249, while large firms are defined to have 250-499 and more than 500 employees. As normalization we exclude one of the industry dummies and size classes each in the estimation.

To estimate the factors explaining R&D subsidies and their productivity effects, we use the IV techniques introduced by Hausman and Taylor (1981). Recall that the Hausman Taylor strategy is to divide the time-varying variables and time-invariant variables into exogenous (i.e. independent of individual effects) and endogenous, therefore possibly correlated with fixed effects. In our specification of the R&D subsidy equation, the endogenous explanatory variables are the growth rate of total sales lagged two years, the cash flow ratio lagged two years, the current R&D intensity and the duration of the R&D project. Dummy variables for firm size, industry affiliation and year are all assumed to be exogenous.

4.1.2 Data and descriptive results

The data used in this study are based on a unique data set containing all firms involved in R&D applying for R&D subsidies from the FFF. The support of the FFF comprises non-repayable grants, loans and guarantees for bank loans. Loans and guarantees are measured in net present value terms. Note that grants account for 90 percent of total R&D subsidies. Projects are supported with up to 50 % of their total R&D project costs.

The FFF project database allows us to identify exactly whether a R&D project is subsidized. This database also includes total project costs, duration in months, total amount of R&D subsidies, level of grants, loans and liabilities and the NACE code of firm specific activity. The FFF project database has been linked with information on the FFF firm database. Firms applying for an R&D project were requested to give information on sales, number of total employees, total R&D expenditures, R&D personnel, foundation year, cash flow and total exports for the last three years before the application for an R&D project. The sample size is 12,333 observations on 3,585 firms. It can be considered approximately representative of all firms doing R&D in Austria. The database includes all firms with at least one employee. The linked project-firm database is constructed in various steps. Since the duration of R&D projects usually exceeds one year, the subsidies granted have to be distributed equally between the years. Next, we sum up the amount of R&D subsidies by firm and year. Finally, the project database is merged with the firm database leaving us with information on about 1,250 firms with 3,500 observations. Furthermore, we exclude firms with a ratio of R&D subsidies to total R&D expenditures above two. Similarly we exclude firms with R&D to sales ratio above one. This leads to a final sample of 1,125 firms with 3,179 observations.

Table 11 reports the evolution of the R&D subsidy ratio as well as R&D intensity among the supported firms for the period 1995-2002. The aggregate mean R&D subsidy ratio is quite stable about 4 percent²⁶. The aggregate average R&D intensity of the supported firms is 5 percent and appears to be increasing in 2001 and 2002.

Table 11: R&D intensity and R&D subsidy ratio (supported firms), 1995-2002

	R&D/sales			R&D subsidies/total R&D			# of obs.
	mean (aggregate)	mean	median	mean (aggregate)	mean	median	
1995	5.0	12.6	5.6	2.6	13.6	7.4	294
1996	5.0	11.7	5.0	3.0	13.0	8.4	357
1997	4.6	12.6	5.1	3.7	14.1	9.0	407
1998	4.1	11.8	5.0	4.2	15.1	10.5	471
1999	4.7	12.8	5.2	4.1	16.4	10.5	477
2000	4.4	12.4	5.2	4.0	16.2	11.0	486
2001	6.1	14.4	5.8	4.4	14.8	10.5	456
2002	8.5	15.8	6.5	3.4	15.2	10.2	231
total	5.1	12.9	5.3	3.7	14.9	9.7	3,179

Notes: Firms receiving R&D subsidies are included. Number of firms is 1,125.

Source: Linked FFF project-firm database, own calculations.

²⁶ The aggregate mean ratio is calculated as the ratio of the sum of R&D subsidies to the sum of total R&D expenditures.

Table 12: R&D intensity and subsidies ratio by firm size (supported firms)

	R&D/sales			R&D subsidies/total R&D			# of obs.
	mean (aggregate)	mean	median	mean (aggregate)	mean	median	
Firm size:							
0-9	20.9	32.7	27.2	18.0	24.9	18.0	487
10-24	13.6	20.3	13.2	14.9	19.6	14.5	459
25-49	11.2	14.8	8.1	12.8	18.8	14.1	302
50-99	6.1	7.8	4.2	12.1	18.9	13.0	353
100-249	4.8	5.5	3.3	7.5	12.1	8.7	573
250-499	3.9	4.9	2.9	7.6	9.4	6.7	475
>500	5.1	5.5	3.1	2.4	4.8	3.1	530
Total	5.1	12.9	5.3	3.7	14.9	9.7	3,179
Firm age:							
Last 5 years	4.5	23.0	12.3	5.8	19.9	13.8	
Six and more	5.1	9.8	4.7	3.5	13.4	8.7	

Notes: Number of firms: 1,125.

Source: Linked FFF project-firm database, own calculations.

Table 12 presents the breakdown of both R&D intensity and R&D subsidy ratio by firm size and firm age. Small and medium-sized firms possess the expected higher R&D subsidy ratio than large firms. The R&D subsidy ratio ranges between 18 percent (aggregate means) in the smallest size class (0-9 employees) and 2.4 percent in the largest size class. Furthermore, we find that firms that are five years old or less have an R&D subsidy ratio of about 5.8 percent as compared to that of firms that are more than six years old with an R&D subsidy ratio of about 3.5 percent. The variation of the R&D subsidy ratio across firm size and firm age is in line with the fund's objectives.

Table 13: Summary statistics (supported firms)

	Mean	Median	Std. Dev.	Min	Max
estimation sample of the R&D subsidy equation (# of obs.: 2,483, # of firms: 909)					
Ratio of R&D subsidies to total R&D	14.5	9.6	18.5	0.1	198.1
R&D/sales	11.1	5.0	15.6	0.0	99.2
Growth rate of total sales, t, t-2	15.1	8.5	30.2	-71.0	236.7
Cash flow/total sales, t-2	8.0	7.9	13.8	-99.1	94.1
Duration of the project in months	14.4	13.8	3.0	3.0	28.0

Notes: Variables are expressed as a percentage of 100.

Source: Linked FFF project-firm database, own calculations.

Table 13 reports averages and standard deviations of key variables used for the estimation. Table 14 contains information on the firms' industry affiliation, size classes, year dummy variables, a dummy variable indicating whether the firm is less than 6 years old and interaction effects between year dummies and the dummy variable for newly founded firms. About 12.5 percent of the 2,483 observed firms have less than 10 employees. About 15 and 18 percent, respectively, belonged to the large firm size class with 250-499 and more than 500 employees. With respect to the industry affiliation, the sample of firms doing R&D is broken down into eleven sub sectors. The share of firms is the highest in machinery (21 percent), followed by electrical machinery and instruments (19.5 percent), chemicals (12.4 percent) and the software industry (12 percent). Furthermore, 16 percent of the firms are five or less years old.

Table 14: Summary statistics (dummy variables, percentage share of total)

Food, beverages, textiles and clothing (15-19)	6.4	year dummy 1996	11.9
Wood, paper, publishing (20-22)	5.3	year dummy 1997	14.1
Chemicals, rubber (23-25)	12.4	year dummy 1998	16.5
Non-metallic mineral products (26)	4.1	year dummy 1999	16.5
Metals, fabricated metal products (27-28)	9.5	year dummy 2000	17.0
Machinery (29)	21.0	year dummy 2001	15.8
Electrical machinery, instruments (30-33)	19.5	year dummy 2002	8.3
Transport equipment (34-35)	6.3	other indicators	
Other manufacturing (36)	1.4	company with limited liability (GmbH)	98.6
Computer services (72)	12.0	Enterprises founded in last five years	16.1
Firm size distribution in terms of employees, L		founded in last five yrs * year dummy 1996	1.9
L <10 employees	12.3	founded in last five yrs * year dummy 1997	2.5
10 ≤ L <25 employees	13.9	founded in last five yrs * year dummy 1998	2.6
25 ≤ L <50 employees	9.6	founded in last five yrs * year dummy 1999	2.6
50 ≤ L <100 employees	11.5	founded in last five yrs * year dummy 2000	2.4
100 ≤ L <250 employees	19.7	founded in last five yrs * year dummy 2001	2.7
250 ≤ L <500 employees	15.1	founded in last five yrs * year dummy 2002	1.5
L ≥ 500 employees	18.0		

Notes: Number of observations is 2,483, number of firms is 909. Variables are expressed as a percentage of 100.
Source: Linked FFF project-firm database, own calculations.

4.1.3 Empirical results

In order to quantify the main factors behind the amount of R&D subsidies, the logarithm of the R&D subsidy ratio is regressed against potential explanatory variables discussed above. As noted earlier, the sample includes only firms receiving R&D subsidies. Table 15 reports the results for the Hausman-Taylor instrumental variable estimator as well as the standard fixed effects model. We also report results of a different specification with interaction terms between year dummies and the indicator on newly founded firms.

The most important factors explaining the logarithm of R&D subsidy ratio are firm size, the log current R&D intensity, firm age and the growth rate of total turnover in the past two years. The ratio of R&D expenditures to total turnover is significantly negative suggesting that firms with a high R&D intensity have a lower R&D subsidy ratio. Note that this is consistent with the fund's objectives. Furthermore, we find that the R&D subsidy ratio continuously decreases with firm size, which probably reflects another one of the public agency's goals. For instance, the R&D subsidy ratio is 37 percent lower for firms with 25-49 employees than for firms with 0-9 or 10-24 employees. The difference between small firms and medium-sized firms with 50-99 employees is 53 percent. The differences for the other size classes (100-249 employees, 250-499 employees and 500 and more employees) are 78, 82, and 91 percent, respectively²⁷. Firms founded in the last five years have a significantly higher R&D subsidy ratio of about 31 percent on average compared to firms that are 6 or more years old²⁸. This is also consistent with the funding agency's strategy. The interaction variables between year and the indicator on firm age show that the subsidy effect of newly founded enterprises is the highest in 2000 and appears to be declining in 2001 and 2002. The significantly negative coefficient on the change in turnover indicates that firms with a higher sales growth rate during the past two years have a lower current R&D subsidy ratio. This means

²⁷ The percentage effect is calculated as $\exp(\text{coefficient})-1$.

that fast-growing firms get lower subsidies than slowly growing firms. Surprisingly, the lagged ratio of cash flow to turnover is not a major determinant of the R&D subsidy ratio. The legal form is not significant either. Firm- specific activities in electrical machinery, instruments and software tend to have a significantly higher R&D subsidy ratio.

Table 15: Determinants of the ratio of R&D subsidies to total R&D: Panel estimates

	Hausman-Taylor IV estimates				Fixed effects estimates			
	(1)		(2)		(3)		(4)	
	coeff.	t-value	coeff.	t-value	coeff.	t-value	coeff.	t-value
Log R&D to sales ratio, %	-0.48**	-10.31	-0.45**	-10.61	-0.49**	-10.70	-0.50**	-10.87
Change in sales, t, t-2, %	-0.37**	-4.11	-0.34**	-4.07	-0.38**	-4.33	-0.40**	-4.47
Cash flow to sales ratio, t-2, %	0.10**	0.51	0.08**	0.47	0.07**	0.37	0.11**	0.58
Founded in the last five years	0.27**	3.56			0.20**	1.89		
Log project duration in months	2.94**	3.04	1.89**	2.11				
Company with limited liability (GMBH)	0.86**	1.68	0.22**	0.49				
Year dummy 1996 (ref. 2002)	-0.07**	-0.79	-0.06**	-0.77	-0.05**	-0.55	-0.07**	-0.81
Year dummy 1997	-0.02**	-0.26	-0.06**	-0.70	0.00**	-0.05	-0.07**	-0.85
Year dummy 1998	-0.02**	-0.21	-0.03**	-0.39	-0.02**	-0.28	-0.07**	-0.84
Year dummy 1999	0.05**	0.60	0.04**	0.52	0.05**	0.59	0.01**	0.15
Year dummy 2000	0.07**	0.98	0.04**	0.59	0.06**	0.75	0.01**	0.09
Year dummy 2001	0.07**	0.97	0.04**	0.55	0.05**	0.62	0.02**	0.20
Founded in the last five yrs * yr 96			0.33**	2.45			0.20**	1.09
Founded in the last five yrs * yr 97			0.26**	1.90			-0.04**	-0.17
Founded in the last five yrs * yr 98			0.17**	0.88			-0.43**	-1.54
Founded in the last five yrs * yr 99			0.12**	0.88			0.10**	0.61
Founded in the last five yrs * yr 00			0.39**	3.07			0.39**	2.60
Founded in the last five yrs * yr 01			0.24**	1.95			0.26**	1.66
Founded in the last five yrs * yr 02			0.23**	1.79			0.14**	0.86
Industry, size dummies								
Food & bev., textiles and clothing	0.09**	0.55	0.04**	0.26				
Wood, paper, publishing	0.13**	0.72	0.14**	0.78				
Chemicals, rubber	-0.25**	-1.71	-0.25**	-1.77				
Non-metallic mineral products	0.27**	1.16	0.11**	0.48				
Metals, fabricated metal products	-0.11**	-0.73	-0.14**	-0.93				
Electrical machinery, instruments	0.33**	2.40	0.22**	1.70				
Transport equipment	0.26**	1.32	0.16**	0.82				
Other manufacturing	0.44**	1.28	0.28**	0.82				
Computer services	0.65**	3.33	0.44**	2.38				
Other industries	-0.16**	-0.64	-0.14**	-0.57				
10-24 employees	-0.12**	-0.80	-0.22**	-1.58				
25-49 employees	-0.46**	-2.71	-0.58**	-3.58				
50-99 employees	-0.75**	-4.55	-0.83**	-5.11				
100-249 employees	-1.52**	-9.84	-1.53**	-10.11				
250-499 employees	-1.74**	-10.31	-1.76**	-10.63				
> 500 employees	-2.43**	-14.68	-2.47**	-15.17				
Constant	-11.63**	-4.27	-8.03**	-3.21	-3.96**	-27.24	-3.94**	-26.92

²⁸ The percentage effect is calculated as $\exp(0.27)-1$.

*Notes: Dependent variable is log R&D subsidies to total firms R&D expenditures. Number of observations is 2,483. The references for industry and size dummies are machinery and 0-9 employees, respectively. * denotes significance at the 10% level; ** denotes significance at the 5% level. Coefficients can be interpreted as elasticities.*

4.1.4 Concluding remarks

This section of the study provides first systematic evidence on the determinants of the amount of R&D subsidies. The empirical evidence comes from panel data provided by the FFF. In particular, we investigated the determinants of R&D subsidies if firms had received subsidies. R&D subsidies are measured as grants plus the net present value of loans. The empirical research indicates that the R&D subsidy ratio is significantly negatively related to both firm size and the R&D intensity, but not to the cash flow ratio in the past. Furthermore, we find that newly founded enterprises as well as firms in electrical machinery and computer services have a significantly higher ratio of R&D subsidies to total R&D. In contrast, fast output growth in the past does lead to a higher R&D subsidy ratio. Overall, the results are consistent with the fund's strategy.

4.2. PRODUCTIVITY EFFECTS OF R&D SUBSIDIES

4.2.1 Introduction

Policymakers increasingly recognize the importance of research and development (R&D) as a driver of productivity growth. Governments use both indirect and direct measures to stimulate technological activity.

In recent years, there has been growing interest in R&D subsidies and the measurement of their impact. Previous studies of the possible impact of publicly funded R&D activities can be divided into two main groups: input additionality analysis (see chapter 3) and output additionality analysis. Output additionality analysis assesses the impact of publicly funded R&D on both research output (i.e. patents) and overall productivity growth. Input additionality analysis investigates whether publicly funded R&D is complementary and thus “additional” to privately funded R&D spending (see Arvanitis 2003; David et al., 2000, for a review). There have been few studies investigating in detail the effects of privately vs. publicly funded R&D on productivity growth based on non-U.S. data. Griliches (1998) summarizes the results of extensive econometric studies of rates of return to privately and publicly funded R&D in the United States. These rates of return range between 18 – 20 percent. The author suggests that there is no differential effect of publicly financed versus private R&D on the levels and rates of growth of total factor productivity at the firm level, although the differences are evident at the industry level. Using U.S. data, Lichtenberg and Siegel (1991) and Nadiri (1993) find that the productivity effects of publicly funded R&D were lower than the estimated effects of privately funded business R&D. Using data for eleven finish manufacturing industries, Niininen (2000) finds that the productivity effect of publicly financed R&D is similar to that of privately financed R&D. Combined, total industrial R&D accounts for nine percent of total factor productivity growth. Using aggregate data for 16 OECD countries, Guellec and van Pottelsberghe (2001) show that increases in both private and public R&D have a positive and significant impact on the change in total factor productivity. Using industry panel data for German manufacturing, Bönte (2003) found positive and significant productivity effects of publicly financed R&D. More recently, both OECD (2001) and Bassanini and Scarpetta (2001) have reported cross-country regressions that suggest a negative return on public sector R&D. Guellec and Van Pottelsberghe (2001) suggest that considerable caution is needed in drawing policy conclu-

sions from empirical analysis at the aggregate level. In this section, we will investigate the impact of public - thus FFF - support for R&D at the firm level.

In this section we investigate the relationship between privately and publicly funded R&D and labor productivity growth. The analysis is based on large sample of Austrian firms doing R&D provided by the FFF for the period 1995-2002.

4.2.2 Empirical model and hypothesis

The empirical model is based on the R&D capital stock model developed on Griliches (1979). Many empirical studies of the productivity effects of R&D relate the change in total factor productivity to the R&D expenditure output ratio because it avoids the problems of measuring the R&D capital stock. The model can be written as:

$$\dot{TFP} = \eta + \rho(RD_{it} / Y_{it})$$

where \dot{TFP} denotes average annual growth of total factor productivity. Since data on investment are not available, we work with labour productivity growth rather than total factor productivity growth. Furthermore, we use the logarithm of the R&D subsidy to sales ratio rather than the level of the R&D subsidy to sales ratio. It is important to keep in mind that the impact of R&D subsidies is lagging and would surface after a certain period *after* the utilization of the R&D subsidies. Following Guellec and van Pottelsberghe (2001), we assume a two-year time lag for the impact of both publicly and privately financed R&D. The resulting production function relates the average annual growth rate of labour productivity to lagged levels of both the adjusted R&D intensity and the R&D subsidy sales ratio:

$$\Delta \log(Y_{it} / L_{it}) = c_1 \log((RD_{it-2} - RDSUB_{it-2}) / Y_{it-2}) + c_2 \log(RDSUB_{it-2} / Y_{it-2}) \\ + c_3 CFR_{it-2} + \sum_{k=1}^K c_k Z_{kit} + \theta_i + \lambda_t + \varepsilon_{it}$$

where i and t are indexes of the firm and the year, respectively. $\Delta \log(Y_{it} / L_{it})$ denotes the average annual growth rate of labour productivity between t and $t-2$. θ_i and λ_t denote fixed and time effects. R&D subsidies are subtracted from total R&D expenditures in order to avoid double counting²⁹. This gives the logarithm of the adjusted R&D intensity lagged two years: $\log((RD_{it-2} - RDSUB_{it-2}) / Y_{it-2})$. $\log(RDSUB_{it-2} / Y_{it-2})$ is the natural logarithm of R&D subsidies as percentage of total turnover lagged two years. CFR denotes the cash flow to sales ratio lagged two years and Z_k are other explanatory variables (i.e. firm age, legal status, industry affiliation and firm size). Since output prices are difficult to get at the two-digit level, one can use the year dummies to proxy the unknown output deflator. The main hypothesis is that the productivity effects of both privately and publicly financed R&D are positive and significant, but we expect that these effects are lower for publicly funded R&D than those estimated for privately financed R&D.

²⁹ The average annual growth rate is calculated as: $\Delta \log(Y_{it} / L_{it}) = \exp[\log(Y_{it} / L_{it}) - \log(Y_{it-2} / L_{it-2})] / 2$

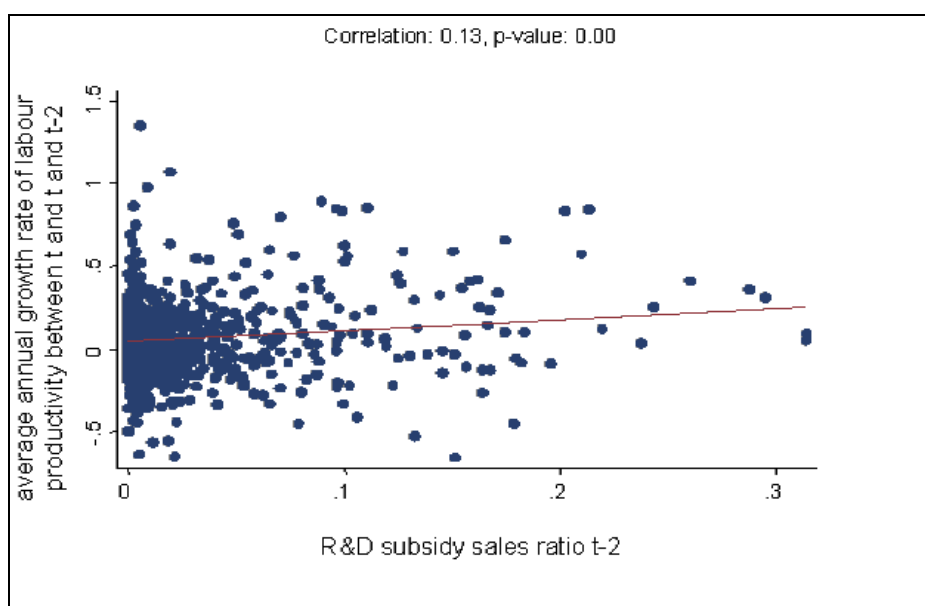
To estimate the factors explaining R&D subsidies and their productivity effects, we use the IV techniques introduced by Hausman and Taylor (1981). Recall that the Hausman Taylor strategy is to divide the time-varying variables and time-invariant variables into exogenous (i.e. independent of individual effects) and endogenous, therefore possibly correlated with fixed effects. In the production function, the endogenous explanatory variables are R&D subsidies to sales ratio, the adjusted R&D intensity, firm age and legal status. Dummy variables for firm size, industry affiliation and year are all assumed to be exogenous.

4.2.3 Data and descriptive results

The data used in this study are based on a unique data set containing all firms involved in R&D applying for R&D subsidies from the FFF. The FFF project database has been linked with information on the FFF firm database. Firms applying for an R&D project were requested to give information on sales, number of total employees, total R&D expenditures, R&D personnel, foundation year, cash flow and total exports for the last three years before the application for an R&D project. The sample size is 12,333 observations on 3,585 firms. It can be considered approximately representative of all firms doing R&D in Austria. The database includes all firms with at least one employee. The linked project-firm database is constructed in various steps. Since the duration of R&D projects usually exceeds one year, the subsidies granted have to be distributed equally between the years. Next, we sum up the amount of R&D subsidies by firm and year. Finally, the project database is merged with the firm database leaving us with information on about 1,250 firms with 3,500 observations. Furthermore, we exclude firms with a ratio of R&D subsidies to total R&D expenditures above two. Similarly we exclude firms with R&D to sales ratio above one. This leads to a final sample of 1,125 firms with 3,179 observations.

In order to get a first insight into the relationship between the change in labour productivity in the following years and the initial R&D subsidy sales ratio we provide a scatterplot with a regression line. We find a significantly positive relationship between the between lagged R&D subsidy ratio the growth rate of labour productivity in the following years (see Figure 16).

Figure 16: Correlations between the growth rate of labour productivity in the following two years and the initial R&D subsidy-sales ratio.



Source: FFF data base; # of obs. 1521. Six firms with R&D subsidy sales ration above 0.4 are excluded

4.2.4 Results of the productivity effects of the amount of R&D subsidies

As stated in the previous section, we investigate the productivity effects of public R&D in a production function framework. The change in labour productivity is a function of the logarithms of the ratio of privately and publicly financed R&D to sales, ten sector dummies, six size dummies, five year dummies, dummy variables for firms founded during the last five years and legal status. Results for Hausman-Taylor IV estimator and the standard fixed effects model are reported in Table 16. Note that both R&D variables are allowed to be correlated with the individual effects. The results indicate that both privately financed R&D and publicly financed R&D make significant contributions to future labour productivity growth as indicated by the elasticities of 0.11 and 0.05. Moreover, we find that newly founded firms have higher labour productivity growth than firms with six or more years.

The magnitude of the productivity effect of R&D subsidies is quite large given the median R&D subsidy sales ratio of 0.4 %. To give an example of the magnitude, we consider a ten percent increase in the R&D subsidy ratio. This would lead to an increase in the growth rate of turnover of about 0.5 percentage points per year given the elasticity of 0.05. Given the elasticities of the production function, one can also calculate how much of the observed change in output per worker can be attributed to the effects of publicly and privately funded R&D. Combined, both funding sources of R&D account for 24 percent of the change in output per worker per year³⁰. However, the contribution of publicly funded R&D is less than 2 percentage points, while the contribution of privately funded R&D is 22 percentage points.

Furthermore, for a number of reasons the large productivity effects of privately funded R&D should be regarded with caution. As noted by Hall and Mairesse (1995), the use of flow rather than stock data tends to overestimate the productivity effects. Furthermore, the relative large productivity effects may partly be due to the omission of capital. Thus, the productivity effects seem to be the upper limit. Furthermore, the sample is reduced heavily due to the inclusion of two-year lags and may be no longer representative. Another concern arises out of the use of gross sales rather than value added as a proxy for output. However, in one of the few studies that used data on both gross sales and value added, Hall and Mairesse (1995) found only small differences in the estimated productivity effects based on value added rather than sales.

³⁰ Elasticities are multiplied by the average sample means and are then divided by the average change in output per worker.

Table 16: Change in labour productivity and R&D subsidy ratio: Panel estimates

	Hausman Taylor IV estimates		Fixed effects estimates	
	coeff.	t-value	coeff.	t-value
Log R&D subsidies to sales ratio, t-2, %	0.05**	6.24	0.05**	6.46
Log adj. R&D to sales ratio, t-2, %	0.11**	7.85	0.11**	8.21
Founded in the last five years	0.13**	3.32	0.11**	2.74
Company with limited liability	1.90**	1.60		
Year dummy 1998 (ref. 1997)	-0.02**	-0.77	-0.01**	-0.63
Year dummy 1999	-0.02**	-1.06	-0.02**	-0.92
Year dummy 2000	0.00**	0.20	0.01**	0.56
Year dummy 2001	-0.02**	-0.85	-0.02**	-1.13
Year dummy 2002	-0.09**	-3.68	-0.10**	-4.06
Food, beverages, textiles and clothing ¹	0.11**	1.64		
Wood, paper, publishing	0.18**	1.65		
Chemicals, rubber	0.04**	0.56		
Non-metallic mineral products	0.17**	1.07		
Metals, fabricated metal products	0.08**	1.39		
Electrical machinery, instruments	-0.10**	-2.18		
Transport equipment	0.00**	0.02		
Other manufacturing	0.10**	0.88		
Computer services	-0.20**	-3.24		
other industries	0.03**	0.31		
10 -24 employees ²	0.04**	0.75		
25-49 employees	0.09**	1.32		
50-99 employees	0.09**	1.35		
100-249 employees	0.03**	0.43		
250-499 employees	0.06**	0.82		
> 500 employees	0.05**	0.69		
Constant	-1.14**	-0.97	0.77**	12.55

Notes: The dependent variable is average annual change in labor productivity between time t and $t-2$. Number of observations is 1,527 and the number of firms is 581. ^{1,2} The references for industry and size dummies are machinery and 0-9 employees, respectively. * denotes significance at the 10% level; ** denotes significance at the 5% level.

4.2.5 Concluding remarks

This section provides first systematic evidence on the productivity effects of the amount of R&D subsidies. The empirical evidence comes from a unique panel data provided by the FFF. In particular, we investigate the impact of R&D subsidies on the growth of output per worker from 1997 to 2002. Using a Cobb-Douglas function to assess the productivity of input factors, we find that the amount of R&D subsidies as well as privately financed R&D expenditures (both expressed as a percentage of sales) have a significant and positive effect on output growth in the following years. Furthermore, the impact of R&D subsidies is relatively large considering the amount of subsidies spent. A ten percent increase in the R&D subsidy sales ratio would lead to a rise in the growth rate of output per worker of about 0.5 percentage points in the next two years. Furthermore, we find that the productivity effects of publicly funded R&D were much lower than those estimated for privately funded R&D.

5 Behavioural Additionality

So far the direct economical effects of FFF-support have been explored. First, the input additionality has been investigated, i.e. the quantification of the extent to which public support stimulates new R&D activity in comparison to the counterfactual of no government funding. Another way of measuring the success of the FFF-scheme is to determine the economic impact of public R&D-support (output additionality), i.e. the focus lies on the gross effects such as increased turnover, enhanced productivity, stronger competitiveness, improved market positions and the like.

Rigby (2003) has proposed to require high output additionality as a necessary first-order condition for the provision of public money, while high input additionality is to be treated as a kind of second-order condition. In view of the scarcity of public funds, he (among others) argues that the second order test is to make sure that publicly funded R&D is not simply substituting for, or actually crowding-out private R&D-investment³¹ and that the latter additionality concept is “a measure of the leverage effect of public money on the private resources of the firm”. Without questioning the ultimate need for efficient use of public resources, there is increasing awareness that R&D-promotion schemes must be judged on the basis of more than immediately materializable returns or directly measurable indicators. Instead, intangible social returns should also be taken into consideration, general competence building and networking included, since this may lead companies to spend more resources on innovation and R&D-projects in the future.³² Accordingly, this chapter addresses long-term behavioural changes emerging from FFF-participation, so-called “behavioural additionality”.

The concept of behavioural additionality as originally introduced by Buisseret et al. (1995) is an attempt to steer the notion of additionality away from a narrow focus on either immediate commercial effects, or various kinds of secondary, nevertheless materializable, effects. Instead, the traditional additionality concept is broadened by investigating whether participation in assisted projects influences the R&D-related *behaviour* of FFF-funded firms in a significant manner. Thus, attention is not so much drawn on the actual economical effects or the triggering of private economic funding of the projects themselves; the decisive point is rather if participation in FFF-funded projects has made the actors become more involved in R&D-activities, if there have been permanent changes in the conduct of a company and particularly on the institutionalization of innovation and R&D-activities (Aslesen et al., 2001). In short, the focus is on the building of innovation capabilities and competence building in general and on the companies’ ability to make use of new technologies and R&D-procedures elsewhere. If such was the case, this may strengthen the company’s ability to absorb new knowledge (their ‘absorptive capacity’). It should be noted that this form of competence building may also benefit other participants in the innovation system, including customers and collaboration partners, thus contributing to a permanent and sustainable increase in the level of Austrian R&D-investment.

This section explores various dimensions of behavioural additionality resulting from FFF-assistance, including project additionality, acceleration additionality, scope and scale additionalities.³³ In this context we present some descriptive evidence from the FFF-survey on, first, the behavioural consequences in case of rejection, second, on collaboration and networking when FFF-assisted projects are implemented, and, third, on the degree to which proposals are based on

³¹ For a survey of the econometric evidence on this issue see David et al. (2000).

³² Georghiou (1997, 2000, 2002), Luukkonen (2000), Papaconstantinou et al. (1997), Sakakibara (1997).

³³ For the motivation of these concepts see Davenport et al. (1998) and Georghiou (2002).

assisted projects are implemented, and, third, on the degree to which proposals are based on preceding projects and/or result in subsequent projects, respectively. Last, data from the FFF's linked firm-project database is exploited to estimate the effect of FFF-subsidies on the stock of scientific R&D-personnel.

5.1.1 Project additionality

Within this evaluation the FFF's customers have been surveyed as regards their appraisal of the working of the FFF. Details on the survey and an overview description of the sample are presented in the next chapter and will not be discussed at this stage. One of the neat features of the survey is that questions in relation to behavioural changes in case of rejection do not take the subjunctive only, but that we can compare the results with answers from those respondents who have actually experienced rejection of their proposals. Ultimately, hypothetical questions are very likely to be answered in a strategic way and the (control) group of actually unsuccessful candidates allows us to uncover such strategic answering behaviour.³⁴

One of the most compelling issues in the context of publicly funded projects is the question of implementation/non-impementation in the (hypothesized) situation of no public assistance. This dimension of additionality is occasionally referred to as "project additionality" (see Davenport et al., 1998). Table 17 below present detailed evidence on this issue derived from the FFF-survey data. Analyses have been undertaken by sector-affiliation, as well as by firm-size and contrast hypothesized outcomes (first column) with actual outcomes (second column).

Table 17: Implementation/non-implementation if application was/is rejected: analysis by sector-affiliation

	Hypothetical Scenario ^{a)}	Actual Consequences ^{b)}
Carry out project without changes (total)	15.13	21.85
Traditional industries	19.77	34.02
R&D-intensive industries	16.32	23.61
Traditional services	16.36	18.75
Knowledgege-intensive service	8.93	12.12
Carry out a revised version of the project (total)	57.16	46.79
Traditional industries	60.47	46.39
R&D-intensive industries	59.64	47.92
Traditional services	49.09	41.67
Knowledgege-intensive service	54.29	47.73
Cancel project (total)	27.72	31.35
Traditional industries	19.77	19.59
R&D-intensive industries	24.04	28.47
Traditional services	34.55	39.58
Knowledgege-intensive service	36.79	40.15
Number of sample firms	985	421

^{a)} Sample comprises companies that never have been rejected by the FFF and answered the relevant question 41;

^{b)} Sample comprises companies that have experienced rejection by the FFF and answered the relevant question 56.

Source: Survey

The first thing to note is that the readiness to carry out a revised version of the rejected proposal is broadly overestimated by around 10 percentage points (total sample). In fact, nearly every third

³⁴ Since respondents have an interest in the continuation of public support, there is an incentive to over-emphasize the effects of public assistance measures. On the other hand one could argue that companies are reluctant to admit their dependency on public funds. In either case there is wide scope for speculation, i.e. we simply cannot tell the direction of distortion (Sakakibara, 1997).

respondent of the set of companies that have experienced rejection at least once reports full additionality, meaning that the project has been cancelled without FFF-support (as opposed to around 28 % in the hypothetical scenario). Likewise 22 % of the respondents say that the project has been implemented anyway, regardless of public assistance, while within the sample of actually supported firms only a fraction of 15 % admit that ultimately no additional R&D-activities are encouraged by means of FFF-sponsoring. Hence, the case of full additionality (“cancel project if FFF-funding is denied”) is systematically underestimated, as is the ultimate willingness or ability to realize the project even if no support was granted. Firms’ wrong assessment with regard to their actual readiness to carry out a revised version of the project is particularly severe in case of the traditional industries and for companies with more than 10 but below 100 employees,³⁵ while the micro-sector firms are generally characterized by more realistic self-assessments.

Public R&D-assistance proves to be most crucial for servicing companies, while – quite by surprise- manufacturing industries are more likely to execute the projects unaltered when FFF-funding is denied, i.e. with the same scale and timetable. Conversely, the share of respondents who claim that the project would have been/has been dropped entirely without FFF-funding is lowest within this subgroup (see Table 17). Presumably these findings are “by construction”. Recall that the questionnaire has only been sent to firms that have ever applied for FFF-support. It is very likely that R&D-intensive firms, or servicing firms, respectively, apply for FFF-support on regular grounds, while traditional manufacturing firms only do so if the project is deemed to be successful and without risk. This would explain our finding that traditional industries are least dependent on FFF-funding. More sophisticated methods are in order to deduce unambiguous evidence on the additionality aspect of public R&D-support.³⁶

Table 18: Implementation/non-implementation if application was/is rejected: analysis by firm-size

	Hypothetical Scenario ^{a)}	Actual Consequences ^{b)}
Carry out project without changes (total)	15.10	21.93
< 10 employees	8.15	14.08
10 – 99 employees	14.07	19.15
100 – 249 employees	21.64	42.31
250 and more employees	25.44	26.97
Carry out a revised version of the project (total)	56.47	46.70
< 10 employees	49.22	47.18
10 - 99 employees	59.55	43.97
100 - 249 employees	56.72	44.23
250 and more employees	62.72	51.69
Cancel project (total)	28.43	31.37
< 10 employees	42.63	38.73
10 - 99 employees	26.38	36.88
100 - 249 employees	21.64	13.46
250 and more employees	11.83	21.35
Number of sample firms	1020	424

^{a)} Sample comprises companies that never have been rejected by the FFF and answered the relevant question 41;

^{b)} Sample comprises companies that have experienced rejection by the FFF and answered the relevant question 56.

Source: Survey

³⁵ Here, discrepancies between hypothesized and actual outcomes range between 14-16%

³⁶ For a survey of the econometric evidence over the past 35 years see David et al. (2000).

A second, likewise unexpected result is that it is *not* the largest firms which prove to be the least reliant on FFF-funding (see Table 18).

It is true that within this size-category the share of respondents who claim that they would carry out the project irrespective of funding opportunities is highest and, conversely, the share of companies reporting that projects would be cancelled in case of FFF-rejection is lowest. However, when we leave the hypothetical scenario and take a look at the actual consequences, we find additionality to be lowest in case of companies with above 100 and below 250 employees. Irrespective of this irregularity, it remains true that micro-sector firms are naturally the most vulnerable if FFF-support is/was withdrawn, or – to put it the other way round - the role of FFF-assistance is quite decisive in encouraging additional R&D-activities within very small firms.

5.1.2 Behavioural changes in case of rejection

Hypothetically asked, 61 % of the set of companies which never failed to qualify for FFF-support state they would seek for alternative public assistance to further promote the rejected project proposals, while not even every third respondent within this subset states the contrary.^{37,38} In this respect differences between various branches turn out statistically insignificant. Repeated analysis by firm-size reveals that the largest size-category is characterized by the highest share of respondents stating that they would indeed send in rejected proposals to other R&D-promotion schemes in case their FFF-application was turned down. But in fact only every fourth company of the total sample actually did seek for alternative support funds, suggesting that dependency on public R&D-assistance is broadly overestimated.³⁹ However, evidence on that issue is not that conclusive since for the latter finding the relevant sample is reduced to only about 300 firms as compared to 1131 firms in case of the hypothetically asked question.⁴⁰

Table 19 below displays conceived, as well as actual consequences, respectively, in case FFF-application turned out unsuccessful. Note that the sample is reduced to the set of respondents who claim that the project would be/has been implemented in a revised form.⁴¹

Table 19: Behavioural Additionality: adaptations if application was/is rejected

At issue		Hypothetical Scenario ^{a)}	Actual Consequences ^{b)}
Starting date of the project	postponed	32.03	43.35
	remains unchanged	61.02	48.77
Duration of the project	Longer	50.68	61.08
	remains unchanged	30.34	23.15
Scale of the project	Smaller	74.07	60.10
	remains unchanged	21.53	32.51
Technical demands	less sophisticated	48.81	39.90
	Remain unchanged	46.78	51.72
Accessibility of project results	Later	62.71	63.55
	at no later point in time	31.02	27.59

^{a)} Refers to Question 42; N = 590; very few firms report an earlier starting date, shorter duration of the project, greater scale of the project or higher technical demand. The share of missing answers ranges between 3%-4%; ^{b)} Refers to Question 57; N=203; very few firms report an earlier starting date, shorter duration of the project, greater scale of the project or higher technical demand. The share of missing answers ranges between 4%-7%. Source: Survey

³⁷ Refers to question 50 of the questionnaire (unreported results). Share of missing answers: 8%,

³⁸ untabled results

³⁹ Refers to question 60 of the questionnaire (unreported results). Share of missing answers below 1%.

⁴⁰ In the actual scenario, companies that dropped their proposal after FFF-rejection are deleted from the relevant sample.

⁴¹ For the hypothetically asked question N=590, while 203 respondents report what actually happened when a revised version of the rejected proposal had been carried out.

First, we observe great unanimity when it comes to the accessibility of project results, i.e. in this respect conceived and actual consequences do not really differ from each other. Two out of three respondents agree that project results could only be exploited at some later date than originally aimed at, supporting the notion that so-called “acceleration additionality”⁴² really matters.⁴³ We find acceleration additionalities to originate as an immediate consequence of postponed starting dates and prolonged implementation phases in case of no public sponsoring. In fact, delays generally turn out much more severe than expected. More than 60% of the unsuccessful candidates admit prolonged project duration.

Most interestingly, it is again the firm-size category of above 100 and below 250 employees which deviates from the norm.⁴⁴ Not even one in three respondents report extended implementation phases and one in five even claims that the project has even been finished earlier than originally aimed at. Furthermore, when asked hypothetically, more than three in four respondents of this size-category claim they would still stick to the original time schedule and only one in five companies conceive that the starting point would have to be postponed. Quite similar results on (hypothesized) starting date and project duration, respectively, are obtained for traditional servicing companies. In summary, there seems to be some evidence that large (but not the largest) firms, as well as traditional servicing firms, do not really *change their behaviour* when experiencing FFF-rejection, but rather make concessions to the time horizon of the project. This notion is supported by results from a previous customer content analysis based on the same survey data. Here, firms of the size-category 100-250 employees were found to invest the least time-input when seeking for FFF-support. They might thus simply be unable to respond in another than a rather pragmatic way.

A final good news from Table 19 is that the least concession are made when it comes to the technical demands. In fact actual consequences turn out less severe than hypothetically conceived and the same is true for the “scale”-issue. Still, 60 % of the rejected firms state to have carried out the project on a smaller scale when FFF-assistance had been denied, hence so-called “scale additionalities” are also prevailing to a considerable degree.

5.1.3 Collaboration and networking

Collaboration and networking must be considered as key aspects of project participation. They involve both, individual and organisational learning, generate network externalities and thereby influence the competences of the actors, as well as their future behaviour (Aslesen et al., 2001, Georghiou, 1997). Furthermore, working closely with R&D-institutions enhances the companies’ absorptive capacity as regards scientific knowledge. At least every other respondent claims that collaboration has taken place in one or the other way (see Table 20). Collaboration may take the form of joint project proposals (true for 52 % of the FFF-funded firms), cooperation with research institutes or with other companies (true for 51 % and 55 %, respectively). It is, however, not that conclusive whether *additional* partnerships have been established, or if existing partnerships have simply been continued. A previous empirical study undertaken in New Zealand, for example, revealed that only around 10 % of the respondents claimed that the government-assisted project resulted in a new partnership (Davenport et al., 1998).

⁴² Acceleration Additionality: when R&D-assistance is speeding up the course of the project (Georghiou, 2002).

⁴³ Within the traditional servicing companies the respective fraction amounts to only 42% when asked hypothetically, however, actual consequences do not differ in a statistically significant way.

⁴⁴ Detailed results are not tabled.

Table 20: Behavioural Additionality: collaboration and networking (%)

	yes	No	Missing
Joint project proposal with customers/suppliers/research partners	52.4	37.7	9.8
Cooperation with research institutes	50.8	38.1	10.9
Cooperation with other companies	55.3	33.6	10.8
Building of research networks	31.3	56.2	12.3

Source: Survey

A detailed analysis verifies that it is especially the more traditional companies, both within the manufacturing sector as well as within the servicing sector, which aim to make good own deficiencies and make use of the particular knowledge of research institutions by respective collaboration. Most surprisingly, however, this hypothesis (i.e. levelling own comparative disadvantages by means of collaboration with the scientific community) does not prove true for the microsector, but rather the contrary is the case. Instead, these obviously highly specialized small firms are particularly prone to the building of own research networks, i.e. collaboration is rather complementary than substitutive in nature. Generally spoken, servicing firms are more involved into research networks and are more responsive to teamwork with other companies. In other words, supporting servicing firms are more likely to generate so-called “scope additionalities”⁴⁵ as the benefits of public R&D-assistance spill over to collaboration partners as well, while the diffusion effects are moderate in case of the manufacturing firms.

5.1.4 Preceding and subsequent projects

If FFF-participation actually enhanced a company’s efforts as regards own R&D-activities, such increased commitment should ultimately result in subsequent projects. In total 485 firms, accounting for 43 % of the relevant sample, report that participation in the FFF-funding scheme has resulted in successive projects which are directly based on preceding, FFF-sponsored research projects (see Table 21).⁴⁶ Likewise, roughly the same fraction states that the FFF-sponsored project just happens to be the later project, i.e. the proposal had been based on some project that had been carried out before. Chain effects (in both directions) are the most likely to occur within R&D-intensive industries, or knowledge-intensive services, respectively, where at least every other firm bases FFF-proposals on former projects, or vice versa. Furthermore, three out of four firms within the R&D-intensive industries, or knowledge-intensive services, respectively, appreciate the value of FFF-support in as far as R&D-activities could be extended to new areas. Such extensions imply changes at the strategic level (i.e. to move into a new area of activity) as well as changes at the level of competences to be acquired in the future. More traditional firms, on the other hand, show a significantly lower readiness to enter unknown research territory (hence the term “traditional”) so that in total “only” about two in three companies extend their research scope. Still, the triggering effect of FFF-participation must be regarded as a definite success.

⁴⁵ Scope additionalities refer to the outcome that “the coverage of an activity is expanded to a wider range of applications or markets than would have been possible without government assistance (including the case of creating a collaboration in place of a single-company effort)” (Georghiou, 2002).

⁴⁶ A fraction of 38% (i.e. 182 firms) has been successful to get public support for the following project as well (no significant difference across sector-affiliation or firm-size). Out of these 79% have again managed to attract FFF-subsidies, 44% have received financial assistance from other Austrian R&D-promotion schemes and 12% have been funded by the European Union. Note that some companies must have been assisted by more than one institution.

Table 21: Behavioural Additionality: chain effects of public funding (in %)

	Yes	No	missing
FFF-project is based on former R&D-projects of our company (Question 38)	44.3	44.4	11.1
FFF-Project has resulted in subsequent projects (Question 47)	42.8	49.9	7.1
Project allowed us to extend R&D-activities to new areas (Question 38)	62.8	26.4	10.7

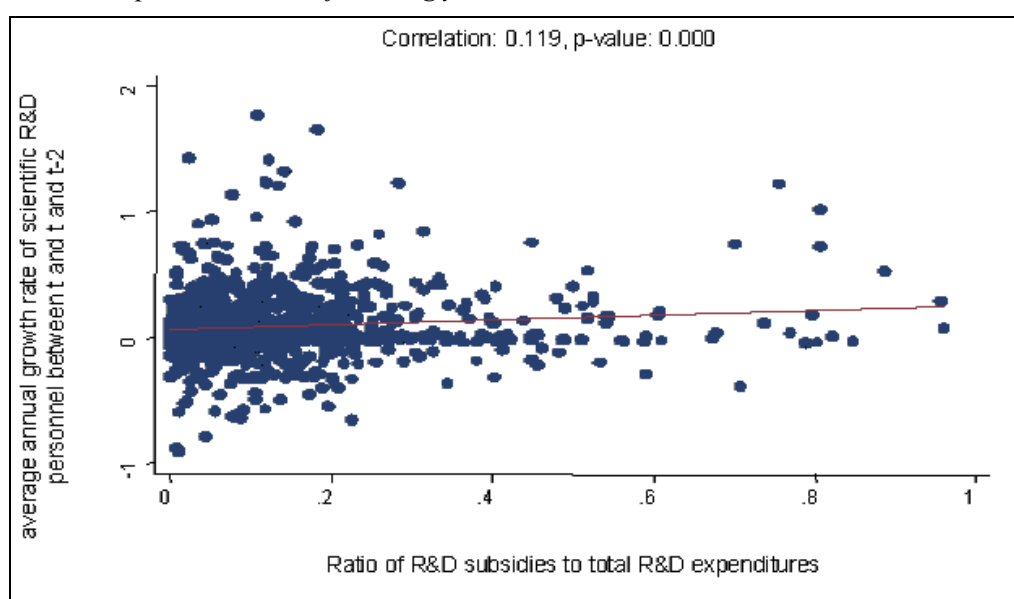
Source: Survey

5.1.5 Econometric evidence from the FFF-panel data set

The standard econometric approach to input additionality is to regress measures of private R&D-resources (expenditure or scientific labour input) on public assistance correcting for selectivity. Behavioural Additionality deals with “the difference in firm behaviour resulting from the intervention” (Georghiou, 2003). As straightforward as appealing this definition may appear, it is not that easy to make operational for econometric purposes. After all, Behavioural Additionality aims to capture *intangible* benefits, such as training of researchers, increased awareness of R&D opportunities in general and establishment of informational networks (Sakakibara, 1997). In search for a suitable left hand-side variable we scanned the linked firm-project FFF-database⁴⁷ and eventually came up with $\log(\text{R\&D-personnel})$. True, this specification once again sheds some more light on the issue of input additionality in an orthodox sense. But additional R&D-personnel will certainly facilitate “increased awareness of R&D opportunities”, the “establishment of informational networks” and the like and will hence improve the firm’s absorptive capacity with respect to new knowledge. Ultimately, the differences between input additionality and behavioural additionality are hard to tell and even harder to quantify.

Appendix IV below presents some preliminary bivariate evidence on the relationship between the initial R&D-subsidy ratio (i.e. FFF-subsidies as a share in total R&D-expenditure) and the average annual growth rate of R&D-personnel. Though the correlation turns out statistically significant and positive, as expected, the coefficient of correlation is very small in magnitude.

Figure 17: Correlation between the initial R&D subsidy ratio in $t-2$ and the growth rate of R&D personnel in the following years^{a)}



Source: FFF database; ^{a)} number of observations (firms) is 1405

⁴⁷ For a description of the dataset, see the preceding chapters on input and output additionality.

For a multivariate analysis we fit a fixed effects model with time-invariant variables as introduced by Hausman and Taylor (1981), results are displayed in Table 22. Coefficients on the year dummies indicate that the number of scientific R&D-staff has increased by some $(\exp(0.28)-1)=32\%$ ⁴⁸ between 1995 (the left-out reference year) and 2002. Results on the other control variables are likewise plausible: companies undertaking R&D in computer-related services, or in high-tech electrical machinery, respectively, employ significantly more R&D-workers as compared to the reference group and apparently the number of scientific R&D-personnel is increasing in firm size (a tautological relationship). With respect to the actual variable of interest, viz. $\log(\text{R\&D-subsidies})$, a highly significant coefficient of only very small magnitude confirms the first impression: a one-percent increase in the amount of R&D-subsidies granted will induce firms to increase its scientific R&D-staff by 0.04 %. To illustrate, suppose the company in concern employs 40 R&D-workers.⁴⁹ One additional worker would increase the stock of R&D-personnel by 2.5%. Hence, in order to achieve this 2.5 % increase – or to hire exactly one more R&D-worker, respectively- the net present value of FFF-subsidies would have to increase by 62.5 %. We conclude that the demand for high-skilled R&D-personnel is only marginally affected by additional FFF-assistance. Instead, it is rather driven by fundamental performance indicators such as total annual sales, or by the NACE-classification of the supported project.

Table 22: Panel estimates of the determinants of $\log(\text{scientific R\&D-personnel})$

	coeff.	t-value	p-value
Year dummy 1996 (ref. 1995)	0.02	0.66	0.51
Year dummy 1997	0.06	2.08	0.04
Year dummy 1998	0.08	2.91	0.00
Year dummy 1999	0.13	4.23	0.00
Year dummy 2000	0.17	5.46	0.00
Year dummy 2001	0.23	7.21	0.00
Year dummy 2002	0.28	7.37	0.00
Founded in the last five years	0.00	-0.07	0.95
\log R&D subsidies (net present value)	0.04	4.31	0.00
\log total sales	0.19	10.89	0.00
Food & bev., textiles and clothing	-0.17	-1.05	0.30
Wood, paper, publishing	-0.60	-3.23	0.00
Chemicals, rubber	0.26	1.85	0.06
Non-metallic mineral products	0.06	0.29	0.77
Metals, fabricated metal products	-0.41	-2.68	0.01
Electrical machinery, instruments	0.83	6.88	0.00
Transport equipment	0.61	3.36	0.00
Other manufacturing	0.08	0.22	0.83
Computer services	1.06	7.26	0.00
Other industries	-0.15	-0.66	0.51
Company with limited liability (GMBH)	0.36	0.90	0.37
10 -24 employees	0.46	3.89	0.00
25-49 employees	0.73	5.19	0.00
50-99 employees	0.67	4.54	0.00
100-249 employees	0.81	5.62	0.00
250-499 employees	1.29	8.02	0.00
> 500 employees	2.38	13.39	0.00

⁴⁸ See Halvorsen et al. (1980) on the correct interpretation of dummy variables in semilogarithmic equations.

⁴⁹ This is the average number of scientific R&D-personnel in the linked firm-project database.

Log project duration in months	1.23	1.53	0.13
Constant	-5.13	-2.44	0.02
# of obs (firms)	3031(1064)		

Notes: Dependent variable is log scientific R&D personnel. Estimates are based on the Hausman-Taylor estimator (Fixed effects with time-invariant variables). The references for industry and size dummies are machinery and 0-9 employees, respectively.

Source: Survey

In qualitative terms, the above findings are confirmed if the same regression is run on various firm-size samples (less than 25 employees, 25-99 employees and 100 and more employees – see Table 23). Comparing the crucial coefficients across size-categories we find that the demand for scientific R&D-labour is the more elastic with respect to FFF-sponsoring the smaller firms. In particular, a statistically insignificant coefficient on log(subsidies) points at the fact that the largest firms hire additional R&D-labour irrespective of additional FFF-funding.

Table 23: Panel estimates of the determinants of the logarithm of scientific R&D personnel by size

	100 and more employees			25-99 employees			Less than 25 employees		
	coeff.	t-value	p-value	coeff.	t-value	p-value	coeff.	t-value	p-value
Year dummy 1996 (ref. 2002)	0.03	0.71	0.48	-0.03	-0.39	0.70	0.05	0.76	0.45
Year dummy 1997	0.06	1.71	0.09	0.04	0.53	0.59	0.08	1.16	0.25
Year dummy 1998	0.08	2.05	0.04	0.09	1.21	0.23	0.07	1.05	0.29
Year dummy 1999	0.10	2.78	0.01	0.05	0.62	0.54	0.21	2.88	0.00
Year dummy 2000	0.16	3.98	0.00	0.07	0.77	0.44	0.22	3.01	0.00
Year dummy 2001	0.23	5.56	0.00	0.14	1.55	0.12	0.27	3.47	0.00
Year dummy 2002	0.26	5.56	0.00	0.16	1.58	0.11	0.34	3.83	0.00
Founded in the last five years	-0.08	-1.45	0.15	0.01	0.09	0.93	0.08	1.45	0.15
log subsidies (net present value)	0.02	1.52	0.13	0.05	2.22	0.03	0.07	3.83	0.00
log total sales	0.24	7.70	0.00	0.26	6.04	0.00	0.14	5.05	0.00
Food & bev., textiles and clothing	-0.51	-1.95	0.05	0.25	0.80	0.43	0.19	0.75	0.45
Wood, paper, publishing	-1.12	-3.80	0.00	-0.10	-0.25	0.81	-0.12	-0.48	0.63
Chemicals, rubber	0.09	0.35	0.73	0.57	1.87	0.06	0.07	0.42	0.68
Non-metallic mineral products	-0.05	-0.13	0.90	-0.14	-0.32	0.75	0.29	0.78	0.43
Metals, fabricated metal products	-0.65	-2.68	0.01	-0.12	-0.40	0.69	-0.28	-1.17	0.24
Electrical machinery, instruments	0.95	3.94	0.00	1.12	4.83	0.00	0.46	3.26	0.00
Transport equipment	0.69	2.44	0.02	0.24	0.60	0.55	0.12	0.44	0.66
Other manufacturing	-0.25	-0.50	0.62	0.38	0.34	0.73	0.42	0.95	0.34
Computer services	0.69	1.55	0.12	1.36	4.94	0.00	0.69	5.27	0.00
Other industries	-0.74	-1.58	0.11	-0.01	-0.02	0.98	0.18	0.74	0.46
limited liability comp. (GMBH)	6.35	3.37	0.00	-0.53	-0.98	0.33	0.00	0.01	1.00
10-24 employees							0.43	4.92	0.00
25-49 employees									
50-99 employees				-0.07	-0.44	0.66			
100-249 employees									
250-499 employees	0.45	2.60	0.01						
> 500 employees	1.46	7.69	0.00						
Log project duration in months	0.79	0.41	0.68	1.08	1.13	0.26	-0.07	-0.12	0.91
constant	-9.32	-1.67	0.09	-4.06	-1.60	0.11	-1.20	-0.83	0.41
# of obs (firms)	1501 (390)			638 (237)			892 (437)		

Notes: Dependent variable is log scientific R&D personnel. Estimates are based on the Hausman-Taylor estimator (Fixed effects with time-invariant variables). The references for industry and size dummies are machinery and 0-9 employees (specification 1), 50-99 (specification 2) and >500 (specification 3), respectively.

Last, some evidence on a dynamic specification resulting from a partial adjustment model is presented in Table 24.⁵⁰ Here, coefficients on year dummies turn out insignificant and the trend effect (persistence) is captured by the lagged endogenous variable instead. We find that contemporaneous, as well as lagged subsidies have a positive short-run effect on the number of scientific R&D-workers of 0.02 %. The long-run effect of 0.06% is calculated as the sum of contemporaneous and (one-period) lagged effects divided by the partial-adjustment coefficient.⁵¹ These effects match the results of the static approach (Table 22) remarkably well and once again support the finding that increased FFF-sponsoring impacts only marginally on the additional demand for high-skilled R&D-labour.

Table 24: Impact of subsidies on scientific R&D personnel: Dynamic panel estimates^{a)}

	(1)			(2)		
	Coeff	t-value	p-value	coeff	t-value	p-value
$\Delta \log$ R&D personnel (t-1)	0.22	1.76	0.08	0.24	1.98	0.05
$\Delta \log$ R&D subsidies (t)	0.02	1.65	0.10	0.02	1.68	0.09
$\Delta \log$ R&D subsidies (t-1)	0.02	2.12	0.04	0.02	2.12	0.03
$\Delta \log$ sales (t)	0.06	2.42	0.02	0.06	2.18	0.03
$\Delta \log$ sales (t-1)	0.05	2.03	0.04			
newly founded	0.00	0.04	0.97	0.01	0.44	0.66
year dummy 1997 (ref 2001)	-0.01	-0.44	0.66	-0.02	-0.64	0.52
year dummy 1998	-0.02	-1.10	0.27	-0.02	-1.16	0.25
year dummy 1999	0.02	0.85	0.40	0.02	0.71	0.48
year dummy 2000	0.02	1.01	0.31	0.02	0.92	0.36
Constant	0.02	1.16	0.25	0.02	1.51	0.13
nobs (firms)	1186 (422)			1186 (422)		
Implied long run subsidy ffect	0.06			0.06		

^{a)} One-step estimates based on heteroscedasticity robust standard errors. Data consist of all firm-year observations with non-missing and non-zero values of subsidies and sales.

Given the overall disappointing and fairly robust results of the econometric exercise, is FFF-funding at least a useful instrument to boost R&D-employment within small firms? The answer is “yes, in relative terms” and clearly “no in absolute terms”. To illustrate, take the extreme case of a one-person firm. One additional R&D-employee translates into a doubling of the existing stock of scientific labour (+100 %). A one-percent increase would induce the labour stock to rise by 0.07 %. To let it rise by 100 % FFF-subsidies would have to grow by $(100/0.07)=1,428.6$ %. Table 25 below illustrates this argument.

How to reconcile the unambiguous evidence of high behavioural additionality from the FFF-survey data with the moderate effects from the econometric approach? The first problem deals with the near impossibility to capture what Papaconstantinou und Polt (1997) call the “soft side of innovation” (networking, learning effects, cooperation, innovative behaviour... etc.). Except for the growth of R&D-personnel there is no such variable in the FFF-project database which could be regarded as a proxy for firms’ improved abilities to absorb new knowledge and accordingly we just estimated what is estimable. A second, arguably more technical problem, is introduced by the fact that the FFF requires companies to reveal certain performance indicators and firm characteristics only when they apply for funds. At this stage information from the three preceding accounting years are mandatory. But ex post figures are only implicitly available, and that only for supported firms repeatedly turning in new research proposals. For our purposes therefore the relevant sample

⁵⁰ This specification has been suggested by Lach (2000).

⁵¹ Partial adjustment coefficient = $(1 - \text{coefficient on the lagged endogenous variable})$.

set is highly selective and biased in favour of more or less continuous R&D-performers. Even if further behavioural changes for these were not subject to the law of diminishing returns, the need for an ever greater R&D-staff certainly is.

Table 25: Increase in FFF-subsidies necessary to employ one additional R&D-worker

	Size-classes	Coefficient on log(subsidies)	Number of scientific R&D-personnel (various scenarios)	Increase in FFF-subsidies necessary to hire one more R&D-worker ^{a)}
	all	0.04	40	62.5
7, columns 8-10	smaller	0.07	1	1428.6
7, columns 8-10	firms	0.07	12	119.0
7, columns 8-10	only	0.07	24	59.5
7, columns 5-7	medium-	0.05	25	80.0
7, columns 5-7	sized	0.05	62	32.3
7, columns 5-7	firms	0.05	99	20.2
7, columns 2-4	larger	0.02	100	50.0
7, columns 2-4	firms	0.02	300	16.7
7, columns 2-4	only	0.02	500	10.0
8, short-run effect	all	0.02	40	125.0
8, long-run effect	all	0.06	40	41.7

^{a)} The following formula applies: $((X+1)*100/X) - 100/c$, where X denotes the existing stock of scientific R&D-personnel and c gives the estimated coefficient on log(subsidies).

5.1.6 Concluding remarks

This section has addressed long-term behavioural changes emerging from FFF-participation, so-called “behavioural additionality”. Descriptive evidence from the survey data revealed that FFF-funding is indeed generating various dimensions of behavioural additionality:

- Around 80-85 % of the sample firms experience some degree of project additionality.
- Acceleration additionalities arise for two in three firms.
- The share of companies appreciating scale additionalities ranges between 60-74 %.
- At least every other firm reports scope additionalities to have arisen from collaboration and a fraction of over 62 % benefits from scope additionalities in as far as new research areas could be entered with the financial help of the FFF-scheme.

Results from some subsequent econometric exercises based on the linked company-project FFF-database turned out not that conclusive, however. In this context the first problem refers to the unavailability of appropriate measures for the mostly intangible merits of behavioural additionality. A second problem is introduced by the general unavailability of ex-post information which makes it hard to systematically evaluate additionality effects of FFF-funding. Conceivably, the greatest effects of FFF-funding on firms’ demand for high-skilled R&D-labour should be observable for firms that do not undertake R&D-activities on regular grounds. Unfortunately, however, it is exactly this type of firm which is hardest to assess and the relevant data set consists of “routine” R&D-performers only. Even if further behavioural changes for these were not subject to the law of diminishing returns, the need for an ever greater R&D-staff certainly is.

The FFF is therefore recommended to condition the provision with public assistance on the obligation to give ex post information.

6 Implications of the FFF-support: results of a survey of FFF customers

6.1. THE SURVEY AND SAMPLING

The main purpose of the survey is to get some idea on how the customers evaluate the working of the FFF. Consequently, we developed a questionnaire for only those companies which have ever sent in a research proposal to apply for R&D-funds.

For this purpose the FFF handed down to the project team a database containing the data set entries of all applicants since 1995. This database held more than 5000 entries. After deleting entries which held incomplete contact information (addresses, contact person) or were referring to joint ventures, 4190 entities remained in the gross sample.

Table 26: Sample for the survey

	N	Percent
gross sample (number of questionnaires dispatched)	4190	100 %
not part of the basic population	48	1 %
contact person retired, dead, has quit, is unknown	44	1 %
firm closed or substantially re-organised	49	1 %
no valid address	397	9 %
revised gross sample	3652	87 %

Source: FFF

During fieldwork the gross sample was further reduced by about 13 % to a revised gross sample of 3652 units. The main reason for this further reduction was that 9 % of the addresses were outdated. A small fraction (in particular university departments) of the original gross sample was actually not part of the basic population.

The questionnaire was a result of an intensive co-operation of the project team, mainly occurring on several joint workshops. The questionnaire was further send to the FFF with their helpful comments included in the final version of the questionnaire. To assure for comprehensibility, a pretest was carried out (see box).

BOX: PRETEST

For pretesting, the project team selected a sample of 50 FFF applicants randomly. The sample included 20 companies which were successful for FFF support with all their project proposals, another 23 companies which had experienced both success and failure while the remaining 7 companies did not receive any FFF support.

End of August, within two days, 44 of the 50 companies were contacted (by phone) and asked to participate in the pretest. It turned out that 14 companies would not be willing or able to complete the draft questionnaire in particular due to a demanding time frame (full response within a week). For another 7 companies a proper respondent was not available because of holidays.

The remaining 19 companies received the draft questionnaire via fax and completed it within one week. Companies were encouraged to comment in particular on wording, availability of data and time required for answering; furthermore, they were asked to earmark any question difficult to answer or to interpret.

The project team contacted the respondents of the draft questionnaire by phone in order to also improve the understanding of how each question is being interpreted. The length of these interviews varied between 10 and about 50 minutes; some of the interviewees had only minor problems while others discovered several critical points.

Most of the interviewees needed between 30 minutes and one hour for completion of the questionnaire. In particular respondents from larger firms saw a need to involve colleagues from at least one other organisational unit. Having received the feedback from the pretest, the project team modified the questionnaire. As a result, 54 questions remained unchanged while modifications became necessary for 7 questions and 4 new questions were added to the final version of the questionnaire.

After completion of the pretest fieldwork started in week 36 2003 as a prenotice letter was sent to the respondents. In the following week the questionnaires were dispatched. The receiver of the questionnaire was the technical project leader or the business administration manager. In week 40 a reminder letter with enclosed replacement questionnaire was posted to the firms, which had not yet responded.

Table 27: Response rate

	N	Percent
revised gross sample	3652	100 %
realised sample (filled out questionnaires received on time)	1298	36 %
latecomers	158	4 %
refusals	425	12 %

Source: FFF; survey

Based on the revised gross sample, a quite satisfying response rate of 36 % could be realized. Another 4 % of the revised gross sample returned completed questionnaires, but unfortunately too late for including them into the analyses. 12 % of the revised gross sample refused to participate in the survey.

6.1.1 Some basic features of the sample firms

As we use for the analysis some categorical variables the following section describes the sample firms according to these categories. It should be proofed if the categories are comprehensible and are useful for further analysis.

Sectoral affiliation

Out of the total sample of 1298 firms we could attribute to 1213 firms, in total accounting for 93.6 % of the sample firms, either one of the following four distinct aggregate branches, viz. traditional industries (23.9 %), R&D-intensive industries (30.3 %), traditional services (11.2 %) and knowledge-intensive services (28.2 %). 72 firms (5.6 %) are attached to the primary sector and 11 companies (0.8%) failed to report their sectoral affiliation (see Table 28).⁵²

Firm size

Second, four size-groups have been introduced, viz. companies with (i) less than 10 employees (synonymously referred to as the micro-sector), (ii) 10 and more, (iii) 100 and more and (iv) 250 and more employees. Table 28 below presents the joint distribution of the sample by sector-affiliation and by firm-size. Every third sample firm belongs to the micro-sector, a fraction of

⁵² For the classification of aggregate sectors, as well as detailed analysis see Appendix II.

broadly 70 % does not engage more than 100 employees and 14 % must be considered as large companies. FFF-application of the latter is subject to particular guidelines.

Table 28: Joint distribution of sector-affiliation and firm-size

Sector-affiliation	firm-size (by number of employees)					Total across size-category
	< 10	10 - 99	100 - 249	250 and more	missing	
Traditional industries	51	110	65	77	7	310
	16.5	35.5	21.0	24.8	2.3	100
	11.4	23.2	43.6	42.1	15.9	23.9
R&D-intensive industries	110	149	55	71	8	393
	28.0	37.9	14.0	18.1	2.0	100
	24.7	31.4	36.9	38.8	18.2	30.3
Traditional services	63	48	10	18	6	145
	43.5	33.1	6.9	12.4	4.1	100
	14.1	10.1	6.7	9.8	13.6	11.2
Knowledge-intensive Services	187	137	12	11	18	365
	51.2	37.5	3.3	3.0	4.9	100
	41.9	28.9	8.1	6.0	40.9	28.2
others/miscellaneous	31	28	6	5	2	72
	43.1	38.9	8.3	6.9	2.8	100
	7.0	5.9	4.0	2.7	4.6	5.6
Missing	4	2	1	1	3	11
	36.4	18.2	9.1	9.1	27.3	100
	0.9	0.4	0.7	0.6	6.8	0.9
Total across branches	446	474	149	183	44	1296
	34.4	36.6	11.5	14.1	3.4	100
	100	100	100	100	100	100

Source: Survey

Firm-age

Though the survey data provides no information on either firm-age, or, equivalently, the year of foundation, this deficiency can be met by merging the survey data with the FFF-database containing basic characteristics of every firm that ever sought for FFF-support. For 1104 sample firms the year of foundation could be reconstructed. The respective percentile-distribution is given in Table 29.

Table 29: Year of foundation (percentile-distribution) (N=1104)

Percentile	1%	5%	10%	25%	50%	75%	90%	95%	99%
Year of foundation	1850	1911	1945	1973	1990	1996	1999	2000	2001

Source: Survey

Ownership & location

The survey contains some valuable information on ownership issues and, in parts, on factory location. Every third sample firm (N=417) is a group company. Roughly two-third of the combine subset (N=268) report the parent company to be located in Austria and, again, two-third of the latter subset (N=182) claim that some of the factories are located outside Austria.

Aggregate sales

Table 30: Aggregate sales (in Mio. €)

Branch	2001	2002	growth rate	Na)
Traditional industries	15783.8	15838.5	0.3	269
R&D-intensive industries	15499.1	16010.7	3.3	343
Traditional services	1619.1	1764.5	9.0	100
Knowledge-intensive services	3424.1	3757.8	9.7	258
others/miscellaneous	1276.3	1298.3	1.7	48
Missing	139.1	148.8	6.9	7
Total	37741.4	38818.5	2.9	1025

Source: Survey ^{a)} Balanced sample: refers to the set of firms reporting data on sales and export rate for both accounting years 2001 and 2002.

Export performance

With respect to companies' export performances, 13 % of the respondents state that sales are exclusively realized within Austria, while about the same share did not comment on export activities. 38 % of the total sample figure that on average at least half of their total sales are earned through exports.

Table 31: Aggregate export performance (levels in Mio. €)^{a)}

Branch	2001	2002	growth-rate	implied export rate	
				2001	2002
Traditional industries	10112.2	10866.0	7.5	64.1	68.6
R&D-intensive intensive industries	11489.3	12199.6	6.2	74.1	76.2
Traditional services	471.2	493.0	4.6	29.1	27.9
Knowledge-intensive services	2820.6	3090.6	9.6	82.4	82.2
others/miscellaneous	577.9	613.5	6.2	45.3	47.3
Missing	90.8	98.8	8.8	65.3	66.4
Total	25562.0	27361.5	7.0	67.7	70.5

Source: Survey ^{a)} Balanced sample: refers to the set of 1025 firms reporting data on sales and export rate for both accounting years 2001 and 2002.

R&D-intensity

Table 32 depicts average R&D-intensity of various sub-sectors (panel (i)) and firm-size categories (panel (ii)). R&D-intensity is measured both, in terms of the average firm-wise share of R&D-personnel in total employment, and in terms of the average firm-wise R&D-expenditure share in total sales, respectively. Apparently, both measures generate fairly robust images of the inherent degree of R&D-intensity. Within the traditional industries and services the (absolute) majority of firms are characterized by very little R&D-activities. R&D-intensive industries by the majority fall into the group of medium R&D-performers, while the knowledge-intensive industries are consistently classified as top-performers. Likewise, the analysis by firm-size yields robust orderings concerning the inherent R&D-intensity, but the size-pattern is not that conclusive as compared to the branch-pattern. As a matter of fact, small denominators generate high ratios, so that the mere size-effect dilutes the expressiveness of the measures.

Table 32: R&D-intensity of sample firms

i) by sector-affiliation	a) Average share of R&D-personnel in total employmenta)			
	up to 5%	up to 30%	more than 30%	missing
Traditional industries	57.1	27.7	7.4	7.7
R&D-intensive industries	25.7	46.6	22.9	4.8
Traditional services	43.5	25.5	17.2	13.8
Knowledge-intensive services	11.2	28.2	52.3	8.2
others/misc.	33.3	30.6	19.4	16.7
Missing	27.3	27.3	18.2	27.3
ii) by firm-size				
less than 10	10.3	30.5	51.8	7.4
10 and more	31.9	44.1	19.6	4.4
100 and more	62.4	28.2	6.7	2.7
250 und more	64.5	25.7	6.0	3.8
Missing	2.3	0.0	0.0	97.7
Total	31.6	33.5	26.6	8.3
i) by sector-affiliation	b) Average R&D expenditure share in total sales b)			
	up to 2.5 %	more than 2.5 %	15 % and more	missing
Traditional industries	49.0	24.5	9.4	17.1
R&D-intensive industries	24.2	38.4	21.6	15.8
Traditional services	40.0	25.5	12.4	22.1
Knowledge-intensive services	11.2	22.5	45.8	20.6
others/misc.	33.3	25.0	25.0	16.7
Missing	36.4	18.2	18.2	27.3
ii) by firm-size				
less than 10	15.0	21.8	43.1	20.2
10 and more	29.5	32.3	21.1	17.1
100 and more	47.0	36.9	8.1	8.1
250 and more	51.4	31.7	6.6	10.4
Missing	6.8	6.8	6.8	79.6
Total	28.9	28.2	24.6	18.3

Source: Survey^{a)} average firm-wise number of R&D-employees in 2001 and 2002 (Question 9) measured as a share of average firm-wise total employment in 2001 and 2002 (Question 6); ^{b)} average firm-wise R&D-expenditure in 2001 and 2002 (Question 10) measured as a share of average firm-wise total sales in 2001 and 2002 (Question 4).

FFF participation

In this context an additional important categorical variable is experience with the FFF, i.e. companies are grouped according to their former experiences with the FFF-support scheme. 1070 or 83 % of the sample firms have been provided with FFF-subsidies in the preceding 8-9 years, while proposals have been (repeatedly) turned down for 206 (16 %) of the survey participants. Since the ratio of FFF-treated to untreated firms amounts to ca. 5:1, the characteristics of FFF-subsidized will – by construction – not significantly deviate from the characteristics of the survey participants.

58 % report that they have been provided with FFF-funds since 1995 and that none of their proposed projects has ever been turned down (“highly successful”, see Table 34). A very small group of 154 firms in total (12 %) is characterized by solely negative feedback as none of their applications has been accepted since 1995. Finally, nearly 30 % of the sample firms have experienced both, acceptances, as well as rejections, respectively (“mixed feedback”).

Table 33: Experience with FFF

		Application has been rejected at least once?			
		Yes	No	Missing	total
Qualified for FFF-support at least once?	Yes	320	717	33	1070
	No	135	55	16	206
	Missing	3	6	11	20
	Total	458	778	60	1296

Source: Survey

Table 34: Success categories: absolute frequencies and shares

FFF-feedback	Abs. Frequency	Percent
highly successful	756	58.33
never successful	154	11.88
mixed	375	28.94
missing	11	0.85
Total	1296	100

Source: Survey

In spite of increasing rejection rates in the last two years, the chances to be provided with FFF-support are still fairly good. While in 2000 three out of four submitted proposals have been accepted, the average rejection rate in 2001 and 2002 amounts to 33 %. In our sample 1070 companies report to have successfully applied for FFF-assistance at least once, and 458 admit to have been rejected at least once. These figures translate into shares of 83 % and 35 %, respectively. Though the rejection rate of the surveyed firms approximately coincides with the average rejection rate in 2001 and 2002, the sample is most probably biased in favour of the successful candidates. After all, the rejection rate as calculated from the survey data is *accumulative* in nature.

6.1.2 Objectives and problems with R&D activities of FFF customers

70 % of the sample firms target at a greater variety of products with their R&D activity in general. From Table 35 the general notion arises that traditional and larger companies are able to list more particular goals which are to be pursued through engagement in R&D. Arguably, R&D is an integral part of usual business activities within the R&D-intensive industries and knowledge-intensive services. But traditional manufacturing firms, as well as traditional servicing companies, seem to take the challenge of R&D to upgrade their production processes or to introduce process-innovations, respectively.

Table 35: R&D-activities are aiming at ...

	Replacement of prod.	Greater variety of prod.	Quality impr. of prod.	Improvements of processes	Process innovation	other
Size						
0-9	18.6	61.9	45.3	18.4	34.5	7.4
10-99	31.9	73.6	53.0	31.2	43.7	4.9
100-249	42.3	77.9	65.1	43.6	48.3	4.7
250+	44.8	78.7	65.6	55.2	55.2	5.5
Sector						
Traditional industry	33.2	74.2	53.9	47.7	56.1	2.9
R&D intensive industry	41.0	76.8	59.0	31.8	38.2	4.6
Traditional services	22.8	66.2	55.2	33.8	42.1	55.5
Knowledge intensive services	21.4	61.4	46.8	18.1	37.3	9.6
Total	29.9	70.1	53.1	31.3	42.7	5.9

^{a)} Refers to question 13 of the questionnaire. Percentage figures give the share within subgroups agreeing to selected “aims of R&D”.

Companies undertaking R&D apparently struggle with a great deal of problems. Only 29 firms failed to comment on the respective question and merely 35 respondents explicitly state to not face one or the other difficulty in the context of R&D. Below some selected problems are ranked in descending order of importance (see Table 36). The overall impression emerging is that the most frequently mentioned problems basically boil down to the number-one obstacle, viz. high cost of R&D-projects. A perceived “high commercial risk”, “low financial capacity”, or a “too long time to market” can be considered as dimensions of the cost problem. As expected, the most crucial obstacle to R&D-activities of micro-sector firms is given by their limited financial resources. The technological risk in turn is of little importance for micro-sector firms but it is more pronounced within the manufacturing sector.

Table 36: Problems related to companies' R&D-engagement (in %, N=1232)

high cost	63.0
high commercial risk	49.8
time to market is too long	36.4
financial resources are limited	35.6
high technological risk	34.2
little protection against risk of imitation	27.9
amortization of R&D-expenditures takes too long	24.8
no sufficient number of R&D-workers	14.0
little opportunities to cooperate with public R&D-institutions	10.1
little opportunities to cooperate with private companies undertaking R&D	8.7
internal problems to organise R&D	7.6
other miscellaneous problems	3.1

Source: Survey

6.2. CHARACTERISATION OF FFF-FUNDED PROJECTS

The following chapters aim at characterising FFF-funded projects in several respects based on a descriptive analysis. Firstly, we question whether there are firm (grouped to categories)-specific motivations to submit a project. Secondly, we describe FFF-funded projects in relation to projects funded by the firm solely. Thirdly, we look at the technical results of FFF-funded projects. Fourthly, we point at the differences between FFF-funded and not FFF-funded projects as to reasons for submission and project advancement.

One should bear in mind that the empirical analysis of the following chapters are related to one project - the last finalized project funded by the FFF (Question 33 of the questionnaire). In the case of not yet finalized projects we asked the respondent to think in their answers of the still ongoing project. This is for reasons of comparison as we wanted the respondents not to think only of the most successful project which would have lead to a bias towards overestimation of success.

6.2.1 Main motivation for submission: high costs of R&D

High costs of R&D is the most important reason to submit a project for FFF funding; this is valid for all firm categories (see Table 37). To reduce the R&D risk of technical feasibility and FFF's reputation as a quick funding possibility are two further important reasons why firms apply for FFF support. Furthermore the survey showed that the percentage of firms frequently submitting

R&D projects for FFF funding is small (usual way of R&D funding).⁵³ Also possible spillovers to competitors are a rather unimportant reason for asking support in order to compensate external effects.

Table 37: Important / not-important reasons to submit a project to FFF (in %, multiple responses)

Categories	Important reasons			Not important reasons	
	R&D costs	Technical risks	Quick aid	Usual way of R&D funding	Spillovers
Size					
0-9	91.6	61.3	63.9	9.4	18.3
10-99	88.1	68.7	57.9	6.3	17.9
100-249	74.8	66.1	69.2	6.7	15.7
250+	75.6	71.6	70.7	9.2	14.9
Sector					
Traditional industry	85.1	72.0	62.3	7.4	20.0
R&D intensive industry	82.7	68.5	69.6	7.2	17.2
Traditional services	86.7	62.4	63.1	6.3	22.0
Knowledge intensive services	85.6	60.5	57.5	9.7	12.5
Export (in % sales)					
Null	88.7	62.0	60.6	5.9	17.7
1-49	86.8	64.9	59.3	6.6	19.2
50+	81.3	68.7	67.2	8.8	15.8
Competition					
Intense	85.6	69.8	62.1	7.6	20.8
Not-intense	84.1	61.4	62.5	8.2	12.8
R&D (in % sales)					
0-2.5	81.0	71.0	60.2	7.8	20.4
2.6-14	83.5	70.2	65.4	5.7	15.4
15+	89.2	58.9	61.6	8.8	13.6
Firm-structure					
Domestic multinational	75.3	64.0	71.8	6.5	15.7
Domestic concern	80.0	65.1	64.1	14.3	23.8
Foreign multinational	74.5	70.5	61.2	5.8	10.7
Domestic individual firm	89.6	66.2	61.8	7.7	17.6
Total	85.1	66.2	63.5	7.8	17.1

Source: FFF-survey 2003, basis: FFF supported firms. Domestic multinational (head office in Austria with branch(es) abroad), domestic concern (head office in Austria with national branch(es) only), foreign multinational (head office abroad with branch(es) in Austria), domestic individual firm (individual firm in Austria).

Firms of different size show the greatest differences with respect to the various reasons for a finally successful submission. Small firms evaluate high costs of R&D as more important than large firms. This is quite different to the other important motives. Technical R&D risks and quick aid motivate larger firms (100+) stronger than smaller ones. Using FFF support as a usual way of R&D funding is found to be less frequent in middle sized firms than in large firms (250+) and small firms (0-9). Technical spillovers to competitors play a greater role for smaller firms than for larger ones.

The relative importance of different motivations is not related to sectors. For all but one sector (R&D intensive industry) reduction of R&D costs is more important than reduction of technical risks and access to quick aid. R&D costs are relatively unimportant, technical risks and quick aid relatively important for firms with high export rates (50+). In contrast, R&D costs are relatively more important for firms without exports. Fierce competition leads to a higher assessment of the motive of R&D costs, technical risks and – as expected – to a greater importance of spillovers to

⁵³ There is no automatism both on the firms side (to submit) and on the FFF side (to fund).

competitors compared with firms in a less competitive environment. Firms with low R&D intensity (0-2.5) see technical risks and spillovers as relatively more important reasons for a submission than firms with greater R&D intensity. For latter R&D costs are relatively more important. Firm structure leads to greater differences in the evaluation of R&D costs, quick aid, routine and spillovers. While individual firms report R&D costs as the relatively most important reason for a submission, Austrian firms with international subsidiaries are relatively stronger motivated by quick aid. Austrian concerns (with subsidiaries only in Austria) submit more regularly projects for FFF funding than other types of firms. And they also see FFF-funding as a compensation for spillovers.

6.2.2 FFF-funded projects lead to R&D extension

FFF funded R&D projects primarily help firms to extent their R&D to new fields (71 %), 62 % of firms stated that the FFF funded project was conducted in cooperation with other firm(s) (see Table 38). And 58 % reported that the idea for the funded project was developed together with customers or suppliers or research partners. Which firm segments differ from this general picture?

FFF funded projects of larger firms are carried out more often in cooperation with research facilities than in cooperation with other firm(s). The opposite tendency is observed for smaller firms. This indicates that projects of larger firms are mainly cooperations between partners of different knowledge levels (basic vs. applied), while projects from smaller firms focus on cooperations on a similar knowledge level. No size related-differences are detected with respect to the sources of ideas for R&D projects (“in-house” or in cooperation with other actors).

Table 38: Characteristics of successfully submitted projects (in %, multiple responses)

Categories	Idea (cooperation)	Idea (firm)	Lead to R&D extension	Cooperation (research facilities)	Cooperation (firms)
Size					
0-9	53.9	53.6	70.5	52.6	66.5
10-99	58.6	51.6	68.8	52.4	62.4
100-249	57.9	51.6	69.4	61.1	54.6
250+	64.0	46.2	73.9	71.1	62.1
Sector					
Traditional industry	56.8	52.3	67.6	64.3	56.1
R&D intensive industry	58.0	52.7	71.4	52.9	61.8
Traditional services	55.3	55.8	58.8	60.6	71.7
Knowledge intensive services	60.3	47.6	76.6	55.4	64.7
R&D (in % sales)					
0-2.5	59.9	47.3	60.8	59.4	60.9
2.6-14	55.1	55.0	72.8	55.0	58.4
15+	61.4	46.0	77.0	54.7	64.3
Total	58.1	50.9	70.5	57.2	62.2

Source: FFF-survey 2003, basis: FFF supported firms

Traditional industries and traditional services are in some ways different from other sectors. Funded projects from the traditional industry primarily aim (like in other sectors) at R&D extension, but in contrast to other sectors they cooperate quite often with research facilities. Traditional services show a very different picture compared to the general observation (see above). Their FFF funded projects are primarily conducted in cooperation with other firms. The next important characteristic is cooperation with research facilities and “R&D extension” can be found only at the third place. In addition the innovative idea is more often developed within the firm than in cooperation with other actors. So one might conclude that the FFF funds more conservative organised projects focused on well known research areas in the traditional service sector as in other sectors.

Quite interesting is the great difference between “idea (cooperation)” (60 %) and “idea (firm)” (48 %) and the importance of R&D extension (77 %) in the knowledge intensive service sector. In combination with the high “ranking” for R&D extension this might reflect the open attitude and the relatively good development of this sector in the last years. The R&D intensity of firms do not influence the ranking of the mentioned characteristics. One can say that here are no exemptions from the rule (except one⁵⁴). The results for firms with the greatest R&D intensity (15+) confirm the general picture as described above: 77 % of the FFF funded projects lead to R&D extension, 64 % are carried out in cooperation with other firms and the idea for these projects were born in 61 % of the cases in cooperation with other actors (customer, suppliers or research partners).

6.2.3 FFF-funded projects are technologically more difficult, more expensive and of longer duration

Based on the answers from firms with both FFF-funded projects and projects funded by the firm solely, it can be stated that FFF funded projects are *in general* more expensive, the technological problems are more difficult and takes longer to carry out the project (see Table 39). In addition the majority of firms report that there are no differences at to the type of research (applied or basic) and the easiness of commercialisation. Basically this holds true independent of how we classify the firms: according to size, sector, export intensity, competition, R&D intensity or firm structure. Nevertheless there are some exemptions: the majority (51 %) of very small firms (0-9) detects no differences in project costs between the two types of projects. Also funded projects of the largest firms (250+) are more oriented towards basic research than in other size classes. This supports the result that successfully submitted projects from large firms are very often carried out in cooperation with research facilities (see Table 38).

Grouped to industrial sectors, firms of the traditional service sector and the knowledge intensive service sector do not see differences in project costs between FFF aided and not externally supported projects (projects funded by the firm solely). Also more firms in the traditional service sector report that the two types of projects do not show a different duration. While firms in the industry sectors reflect the above mentioned “rule”, firms in the service sectors show some different patterns which may be due to different product characteristics and/or research processes.

⁵⁴ 60.9% of firms with a low R&D intensity (0-2.5) stated that FFF funded projects are conducted in cooperation with firms, 60.8% stated that projects led to R&D extension. The “rule” is the other way round: “lead to R&D extension” should be higher ranked than “cooperation (firms)”. The figures are rather similar so that one can hardly speak of a real exemption.

Table 39: Comparison of FFF-funded projects and not externally funded projects (in %)

Categories	Project-costs			Technology			Research			Project duration			Commercialisation		
	High	Nor	Low	Diff.	Nor	Easy	Basic	Nor	Appl.	Long	Nor	Short	Diff.	Nor	Easy
<i>Size</i>															
0-9	44.6	51.4	4.1	51.1	45.3	3.6	25.7	59.2	15.1	47.5	43.0	9.5	31.1	61.3	7.7
10-99	53.8	40.9	5.4	63.8	34.6	1.6	30.6	54.5	15.0	57.6	36.6	5.7	35.4	58.0	6.7
100-249	54.1	41.8	4.1	63.4	34.2	2.4	37.5	53.3	9.2	48.3	47.5	4.2	40.5	52.1	7.4
250+	50.0	46.8	3.2	71.0	28.4	0.7	40.4	45.7	13.9	63.0	34.4	2.6	33.1	63.0	3.9
<i>Sector</i>															
Traditional ind.	54.1	41.4	4.5	68.9	30.2	0.9	34.4	51.6	14.0	60.7	35.7	3.7	40.0	54.0	6.1
R&D intensive ind.	54.6	40.9	4.5	64.1	33.5	2.4	33.0	52.6	14.4	57.4	38.1	4.5	34.5	60.7	4.8
Traditional serv.	46.7	50.7	2.7	52.0	42.7	5.3	25.3	58.7	16.0	41.3	49.3	9.3	25.0	63.2	11.8
Knowl. intensive serv.	42.3	52.4	5.3	53.9	43.8	2.4	33.0	53.9	13.1	49.8	40.1	10.1	32.7	59.6	7.7
<i>Competition</i>															
Intense	46.9	47.6	5.5	62.7	35.6	1.7	31.9	55.4	12.7	54.8	39.5	5.8	38.3	55.0	6.7
Not-intense	55.3	42.0	2.7	59.9	37.1	3.0	31.5	53.0	15.5	53.6	38.8	7.6	30.2	63.1	6.7
<i>R&D (in % sales)</i>															
0-2.5	49.2	47.6	3.3	60.7	37.7	1.6	31.4	53.3	15.3	53.3	43.4	3.3	36.3	58.0	5.7
2.6-14	57.0	37.6	5.4	68.8	29.0	2.2	31.9	56.7	11.5	59.3	34.8	5.9	36.2	57.6	6.3
15+	43.5	53.4	3.1	56.5	41.4	2.1	35.6	50.3	14.1	49.7	40.3	10.0	31.6	58.6	9.8
Total	50.6	45.0	4.4	61.5	36.3	2.2	32.2	53.7	14.1	54.3	39.7	6.0	34.4	59.0	6.6

Source: FFF-survey 2003, basis: firms with FFF-aided projects and with projects funded solely by the firm

The competitive environment and the R&D intensity of a firm do not cause a substantial deviation from the general pattern as well. Differences are only detected in relation to the project costs. The majority of firms exposed to fierce competition and also the majority of firms with greater R&D intensity (+15) see - in deviation from the “rule” - no differences in project costs between FFF-funded and solely internally funded projects.

6.2.4 Majority of Firms: FFF funded projects have no impact on other R&D projects

In most of the cases FFF funded projects have no impact on other R&D projects within the same firm. They are carried out as planned (58 %). This gives an impression that firms plan their research activity independent of FFF support and/or FFF funded projects are not of immediate importance. Latter is supported by the fact that only 7 % of the firms stated that they postpone other projects in order to start with the FFF funded project. Only in very small firms (0-9) and firms with an export intensity between 1 and 50 % FFF funded projects have clearly an above-average importance (see Table 40). Furthermore it has to be considered that 21 % of the firms have solely FFF funded projects. For these firms the just mentioned argument is not valid. Especially smaller (34 %) and individual firms (28 %) and firms in the traditional service sector (38 %) have only FFF funded projects.

In 25 % of the cases FFF funding leads also to an extension of existing projects or/and new R&D projects. This means also that in most of the cases FFF funding contributes to existing firm-specific “research-paths” and to a lesser extent to broaden the research activities of a firm. This somehow contradicts to the overall research aim of firms to extent their R&D fields (see Table 38). It seems that FFF funding is important and intensifies R&D but it is not enough financial support to start “radically” new projects. This average assessment applies to smaller individual firms with rather low R&D intensity and null or few export activities. In contrast, FFF funding has a much greater impact on other R&D projects in larger firms (39 %), in R&D intensive Industries

(31 %), in export intensive firms (30 %) and in R&D intensive firms (29 %). Also individual firms notice (to above 30 %) that FFF funding has a broad impact on their R&D activities.

Table 40: FFF funded projects: impact on other R&D projects (in %, multiple responses)

Categories	Postpone other projects	Add on and/or start new projects	No impact on other projects	No other projects
Size				
0-9	9.0	20.8	43.1	34.2
10-99	7.1	22.2	59.9	21.7
100-249	3.7	26.1	69.4	11.2
250+	4.2	38.7	72.0	4.2
Sector				
Traditional industry	5.1	21.4	66.5	17.5
R&D intensive industry	8.7	31.0	59.4	15.5
Traditional services	5.6	21.5	43.0	38.3
Knowledge intensive services	6.9	22.7	55.2	24.2
Export (in % sales)				
Null	3.5	19.8	45.7	36.2
1-49	10.3	22.0	54.8	22.7
50+	5.0	29.6	67.6	12.5
Firm-structure				
Domestic multinational	3.5	32.2	72.5	7.0
Domestic concern	4.2	33.8	62.0	16.9
Foreign multinational	4.4	30.7	72.8	6.1
Domestic individual firm	8.2	21.5	51.3	27.8
Total	6.7	25.0	58.0	21.2

Source: FFF-survey 2003, basis: firms with FFF-aided projects. Domestic multinational (head office in Austria with branch(es) abroad), domestic concern (head office in Austria with national branch(es) only), foreign multinational (head office abroad with branch(es) in Austria), domestic individual firm (individual firm in Austria).

6.2.5 FFF-Funding Results: New or improved products/services are most frequent

63 % of all firms stated that they have achieved their technical goals in an FFF funded project totally, 35 % partly and 2 % have admitted a failure. That is a quite impressive performance for R&D projects. But what are the main achievements in greater detail?

More frequently FFF support leads to new or improved products or services (54 %), especially in middle-sized and large firms. Firms in the knowledge-intensive service sector (63 %) reached this goal more often than firms in other sectors. This is quite in accordance with the dynamic development of this sector in the last few years, which was heavily based on service/product innovations. Also exporting firms – most of the time exposed to fierce competition – as well as R&D intensive firms show an above-average performance in product/service innovations (see Table 41).

FFF funding resulted in a prototype in 37 % of the cases. Mainly middle-sized firms (39 %) and firms in the R&D-intensive industries (40 %) achieved this result. Serial prototypes are more often reached in the R&D intensive industry (36 %) than in any other sector. Also export intensive firms clearly show a better performance in serial prototypes than less export intensive firms or firms that do not export at all. This result may be due to some sectoral/branch effects⁵⁵ which can not be detected in this descriptive analysis.

⁵⁵ It might be necessary in some branches to build a serial prototype and in some other branches not. In case that serial prototypes are more important in branches with high export rates than in others, this would deter the descriptive results and make an interpretation difficult.

To apply for a patent does not seem to be of high priority for FFF funded projects - on average 28 %. Medium-sized firms (29 %), the traditional service sector (33 %) and export intensive firms (32%) are above the average. Although internationally oriented firms tend to apply for patents more often than firms mainly active on national markets, firm structure may also play an important role in this respect. Multinational firms more often apply for patents than other firms do. Thus firms' patent policy is probably not related to FFF funding. It is more a question of organisational routines of how to protect R&D results.

Table 41: Results of the FFF funded projects (in %, multiple responses)

Categories	Feasibility study	Patent	Prototype	Serial prototype	New/improved product/service	Process innovation
Size						
0-9	50.0	27.3	27.3	31.8	45.5	13.6
10-99	25.2	23.6	34.7	26.4	51.6	18.2
100-249	24.2	28.7	38.5	28.9	56.5	28.7
250+	26.3	23.3	36.8	28.6	51.9	33.1
Sector						
Traditional industry	23.3	32.9	36.1	33.7	48.2	45.8
R&D intensive industry	27.0	32.9	40.4	36.2	53.9	21.6
Traditional services	31.5	33.3	37.0	23.2	48.2	29.6
Knowledge intensive services	26.5	15.4	34.9	19.1	62.5	15.8
Export (in % sales)						
Null	38.6	21.1	36.0	19.3	51.8	16.7
1-49	26.7	24.8	36.8	29.9	58.7	29.1
50+	22.8	32.3	37.8	30.9	54.2	29.0
R&D (in % sales)						
0-2.5	28.9	26.1	30.4	27.9	52.5	36.1
2.6-14	24.4	29.5	40.6	32.1	58.1	27.0
15+	26.1	24.1	36.0	26.9	59.3	20.6
Total	26.3	27.5	36.7	28.6	54.4	27.0

Source: FFF-survey 2003, basis: FFF supported firms

27 % of the firms stated that their last FFF funded project brought out in a new or improved process. Process innovation point at the efficiency aspect of a firm, which lays in the core of larger firms. So it is not surprising that larger firms, firms in the traditional industry sector and export intensive firms reported this more often than others. Feasibility studies are a rather seldom result of FFF funded projects (26 %). Very small firms (0-9) and firms in the traditional service sector reached this R&D result most often. It seems that they are the result of smaller R&D projects or a first step on the way to a more comprehensive R&D project.

6.2.6 To open up a market: primary goal for commercialisation

Firms commercialise FFF funded projects primarily to open up new markets (67 %). This is a "logical" continuation of the FFF-funding. Successfully submitted R&D projects extent the R&D fields of the firm in 71 % of the cases (see Table 38). So it is obvious that firms' R&D goals and FFF funding practice are coherent. This is valid for all kind of firms especially for firms with high R&D intensity (15+) (see Table 42). The second important goal for commercialisation is to increase/maintain market shares abroad (51 %). R&D seems to be the key for maintaining/augmenting activities abroad for many Austrian firms. More often than other large firms and firms in the industry sector (R&D intensive and traditional) as well as export intensive firms and firms with international branches commercialise FFF-funded project in order to strength their

international presence. This is of primary importance for the largest firms (250+) and Austrian firms with international branches.

Larger firms in a small, open economy like Austria need to be well positioned on international markets. So it does not wonder that the third important goal is again related with markets abroad. Firms commercialise the results of R&D in order to maintain/increase their export share. Again this is more important for larger firms, firms in the R&D intensive and traditional industry, export intensive firms and internationally fragmented Austrian firms and Austrian firms with national branches only. Latter don not have branches abroad but nevertheless export markets are important.

Table 42: Commercialisation of the results of R&D: the four most important goals (in %)

Categories	Open up new Markets		Maintain/increase market share abroad		Maintain/increase export share		Maintain/increase national market share	
	Very imp.	Not imp.	Very imp.	Not imp.	Very imp.	Not imp.	Very imp.	Not imp.
Size								
0-9	69.8	2.0	45.2	13.1	44.1	15.6	47.1	8.0
10-99	68.3	1.1	49.2	4.8	49.3	8.0	43.3	7.8
100-249	63.5	1.6	54.8	8.1	50.0	7.3	37.1	7.3
250+	62.3	2.0	64.9	4.6	53.9	7.7	39.2	7.2
Sector								
Traditional industry	60.2	2.5	55.7	5.2	50.9	7.3	41.5	9.2
R&D intensive industry	68.5	0.6	64.0	2.6	59.3	3.5	40.3	9.6
Traditional service	71.0	3.0	42.9	13.3	48.0	12.2	58.0	6.0
Knowledge intensive service	68.6	1.6	35.9	13.3	36.3	17.1	40.2	6.4
Export (in % sales)								
Null	65.1	1.9	23.7	24.7	22.7	29.9	47.0	7.0
1-49	66.7	0.8	48.9	5.6	50.3	8.4	53.1	3.6
50+	67.6	2.5	64.2	2.3	55.6	4.2	32.9	10.5
Firm-structure								
Domestic multinational	64.2	1.9	65.0	4.4	56.9	5.6	39.2	8.2
Domestic concern	73.0	3.2	63.9	6.6	62.9	9.7	43.6	4.8
Foreign multinational	69.8	1.9	56.6	2.8	44.9	9.4	31.1	12.3
Domestic individual firm	66.3	1.3	45.4	9.3	45.3	11.3	46.0	7.1
Total	66.8	1.7	51.3	7.8	48.3	10.2	43.1	7.8

Source: FFF-survey 2003, basis: FFF supported firms which plan to commercialise the results of R&D. The firms answered based on a five point likert-scala. The table shows the two points: very important and not important. Domestic multinational (head office in Austria with branch(es) abroad), domestic concern (head office in Austria with national branch(es) only), foreign multinational (head office abroad with branch(es) in Austria), domestic individual firm (individual firm in Austria).

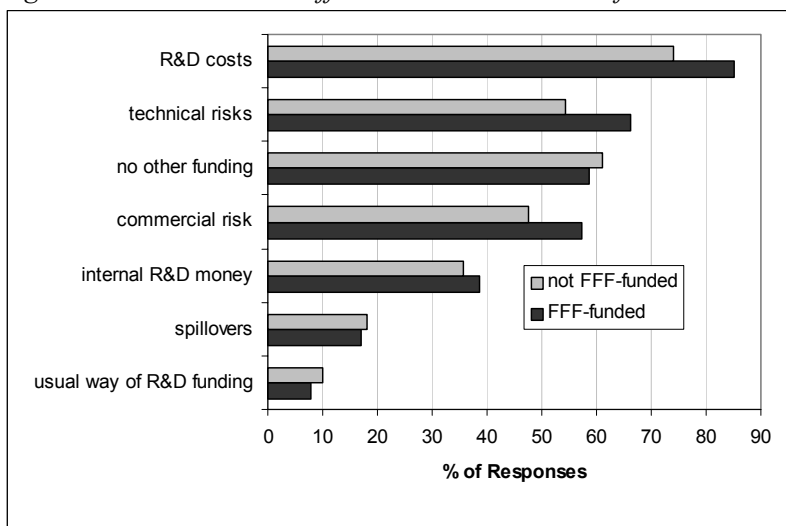
Very small firms (0-9) commercialise the results of R&D very often in order to maintain or increase their national market share. This is for them of secondary importance right after “open up new markets”. Firms in the traditional service sector and exporting firms (1-50) also assess this goal to be of above-average importance. Furthermore it is interesting that the commercialisation goals don not vary much with the R&D intensity of firms. This means that these goals are set rather independent of the size or relative importance of the R&D department. Also it is worth mentioning that “to increase productivity” is only a goal of some importance.

6.2.7 Firms’ motivations and project advancement: differences between successful and unsuccessful FFF submissions

There are no significant differences in motivations between firms which successfully submit R&D projects for FFF-funding and those which unsuccessfully do so (see Figure 18). In both cases high R&D costs are the main motivation for submission. The motives “decrease of the risk of commer-

cialisation” or “easy access to R&D money” or “compensation of R&D spillovers to competitors” are ranked at fourth, fifth and sixth position, respectively. Only the motivation to cover “technical risk” and “no other funding” possibility, change the position in dependence of successfully or unsuccessfully submitted projects; FFF-funded projects are more often motivated by “technical risk” reduction than not FFF-funded projects. The contrary is valid for “no other funding”.

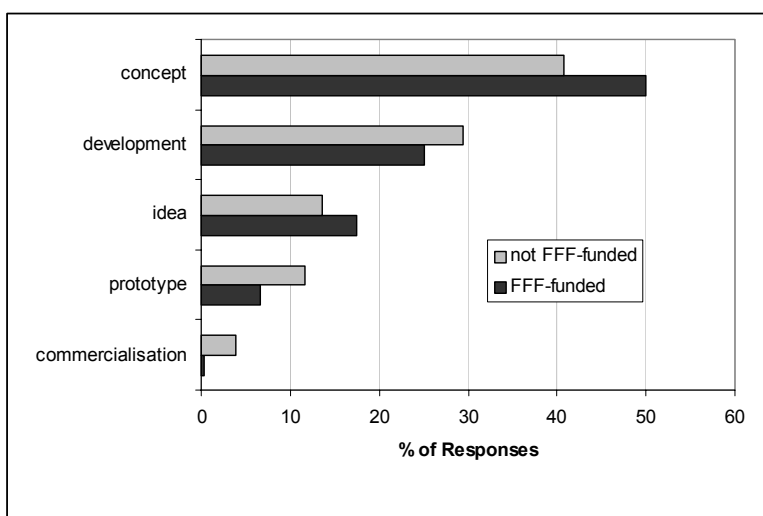
Figure 18: Motivations - Differences between successful and unsuccessful submissions (in %)



Source: FFF-survey 2003, basis: all firms

The more advanced a R&D project is the more information (technical, market possibilities etc.) is available and the lower is the risk of a failure. In Figure 19 different stages of a R&D process are mapped. An idea is followed by a concept, which has to be developed before a prototype can be built. Finally commercialisation shows whether the new/modified product/service or process is a market-success or a market-failure. Based on this serial concept of an R&D process it can be seen that R&D projects are funded by the FFF on every stage of development.

Figure 19: Project advancement – Differences between successful and unsuccessful submissions (in %)



Source: FFF-survey 2003, basis: all firms

R&D projects submitted at the idea-stage received more often a FFF-funding than not. Also at the following concept-stage a positive decision was more likely than a negative one (see Figure 19). This picture changes with the development-stage, the prototype-stage and the commercialisation-

stage. On these levels the FFF issued more often a negative funding decision than a positive one. The differences are rather great with respect to the prototype- and commercialisation-stage. That means that the differences increase the more advanced the R&D project is. Ceteris paribus R&D projects are more likely to be funded than the less advanced ones. Having in mind that an R&D project becomes less risky the closer it comes to commercialisation, one can conclude that the FFF funds more risky projects more often than less risky ones, which should be part of the FFF-mission.

6.3. LICENSES, PRODUCTS AND PROCESSES

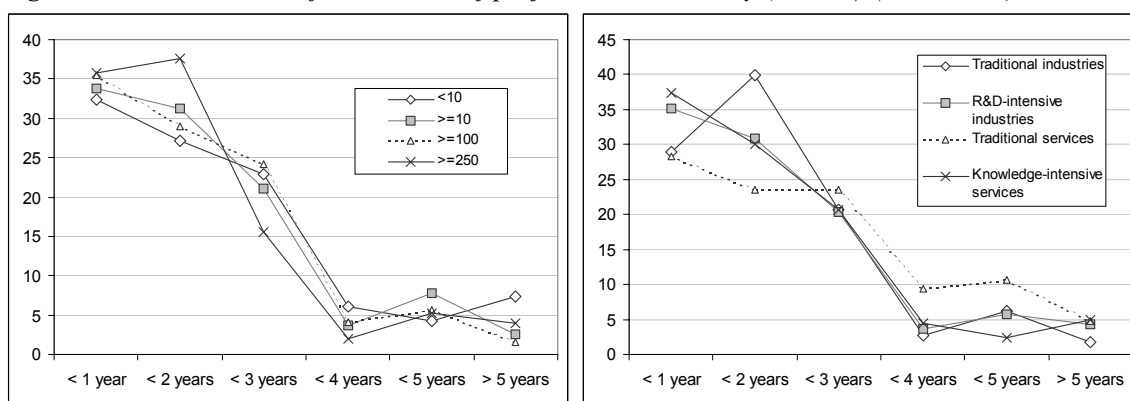
The main aim of the FFF funding for industrial R&D projects is to ‘... strengthen the competitiveness of business in Austria and thus help to secure employment, production and wealth in Austria.’ (FFF-mission). Research and development is among the most important factors of business success. It rises directly or indirectly the technological innovation ability of the industry in order to generate new products, processes and services.

The following section deals with the results of the FFF-support and is based on several questions of the survey related to all funded and already finalized projects for the period 1995 – 2003. We thus asked the respondent to think in his or her answers of all of the funded and already finalized projects and to give in the case of more than one project an assessment of the average.

6.3.1 Time to market

The time to market for the project results was an issue explored in the survey. Based on all funded and already finalized projects the respondents were asked to give the length of time when the products were expected to enter the market and/or new processes were introduced and economically utilized. The majority (65 %) of the project results were expected to enter the market within two years after project end. Within this group almost half reported the entry to the market during the first year after the end of the project. The average time length between the end of the project and market entry was 2,2 years (Median=2 years).

Figure 20: Time to market from the end of project to market entry (N=855) (1995-2003)



Source: survey

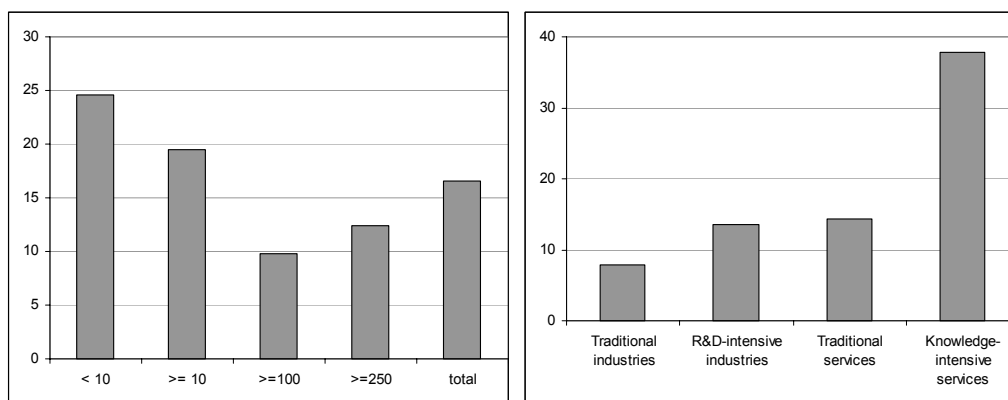
In general, there also does not seem to be major differences in the time to market by firm size. At least 85 % of all projects in all firm size categories had launched the product to market within three years after project end. However, whereas 59 % of the micro firms reported the market entry within two years this ratio increases to 72 % within the group of the big firms. Based on the an-

swers given to this question it can be concluded that big firms are more efficient in commercializing the research results and are faster in terms of time to market.

6.3.2 Licenses

Out of the projects funded and finalized between 1995 and 2003 we were interested in those firms which achieved revenues from licenses. We started this block by asking the firm whether they received license revenues (yes or no). About 18 % of the respondents indicated this question with yes. The analysis by firm size however shows a significant difference between the classes: while 25 % of the micro firms achieved revenues this ratio decrease to about 11 % within the bigger firms.

Figure 21: Licenses by size and sector in % (N=862) (1995-2003)

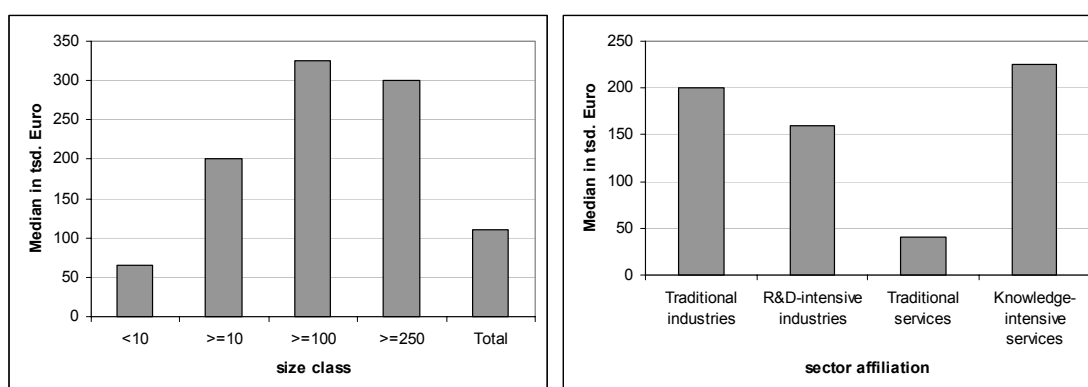


Source: survey

An even higher difference shows the comparison between the sectors of activity: nearly 38 % of the firms in the knowledge-intensive services reported license revenues as a project outcome. Thus it confirms the assumption that revenues from licenses are concentrated on the knowledge-intensive service sector (software, data-processing etc.).

The respondents were further asked to give an estimate of the total license revenues out of the FFF projects for the whole period (1995 – 2003). Although it is difficult to answer 118 firms filled in and gave a rough estimate of the revenues: due to a high standard deviation we focused on the median value. The median value of revenues out of licenses for the whole period was 110 tsd. Euro. The following Figure 22 shows the allocation of revenues by firm size and sector. Although the knowledge-intensive sector exhibit the highest share of firms reporting license revenues the volume of the revenues shows a focus on bigger firms in the R&D-intensive sector.

Figure 22: License revenues by size and sector for the period 1995-2003(N=118)



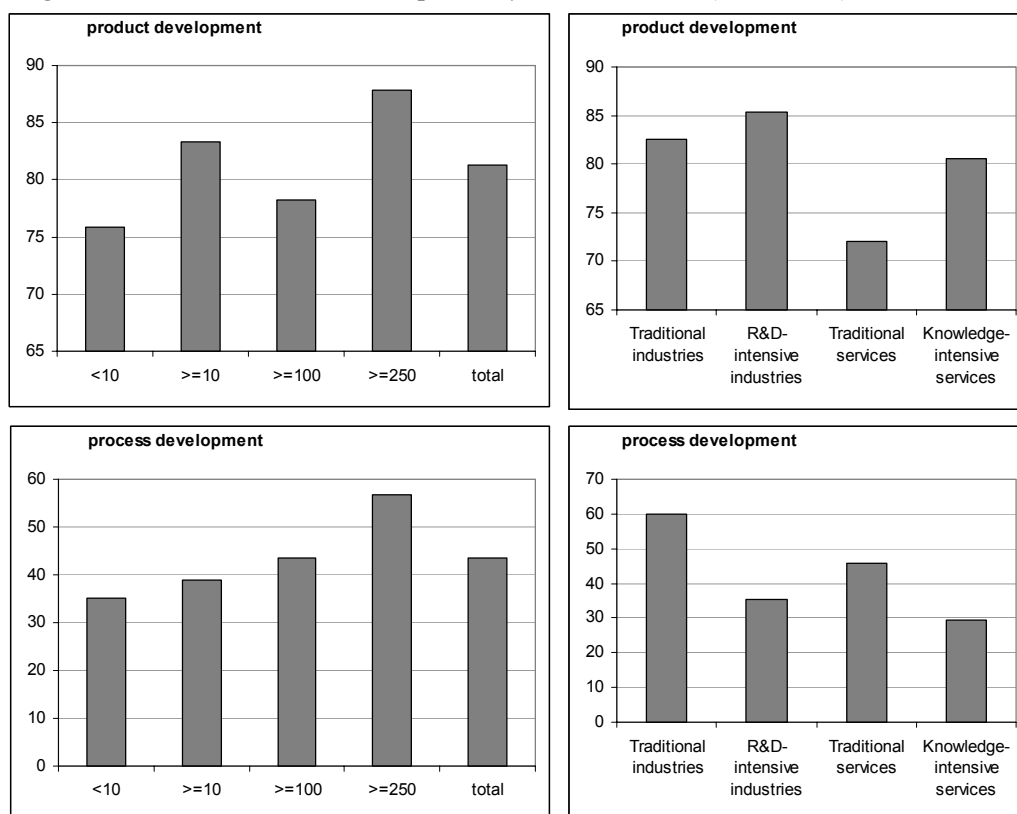
Source: survey

6.3.3 Products and Processes

The general orientation of the projects was assessed by asking the respondents whether a marketable product, service or process was developed as a project outcome during the period 1995 – 2003 (yes or no). In general, the overwhelming part of the funded projects was market-oriented: 81 % of the respondents answered that a marketable product/service was developed out of the results of a FFF-funded project. 42 % of the respondents said that new processes were developed.

Product development does not seem to depend very much on firm size, although bigger firms indicated a 10-percentage points higher rate of product developers than within the group of smaller firms. The development of new processes tends to be more oriented towards larger firms. Among R&D-intensive industries, the likelihood of having a marketable product as the result of a FFF-project was 85 % whereas for process development the sector of traditional industries exhibits the highest probability (Figure 23).

Figure 23: Product/Process development by size and sector (1995-2003)



Source: survey

We further asked the respondents for an estimation of the following issues (see Figure 24).

- how many new marketable products were developed out of FFF-projects since 1995;
- the share of marketable products out of FFF-funded projects on all new products since 1995;
- finally we asked the respondents for an estimation of the share of the marketable products out of FFF-funding (1995-2003) on the total turnover of the last business year.

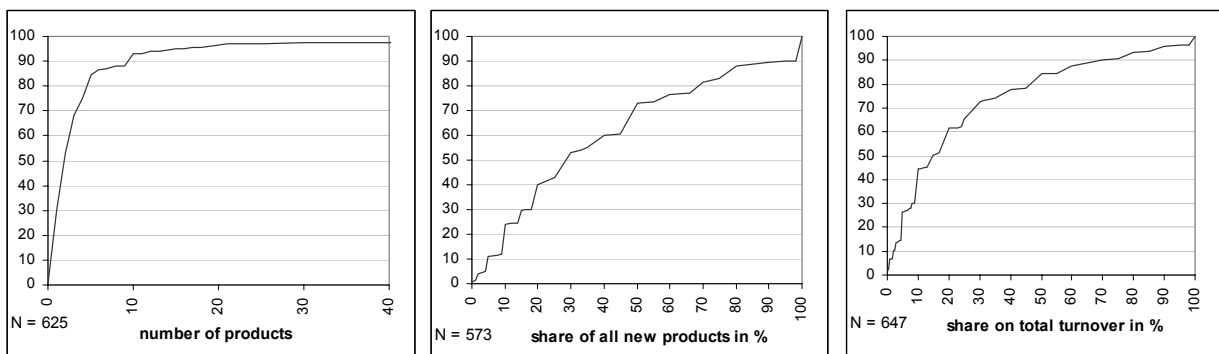
The response rate on these questions was astonishing high. Concerning the first issue marketable products were rather concentrated on a few products: for 30 % of the respondents one marketable

product resulted out of the FFF-funding for the whole period 1995 - 2003. 90 % answered that not more than 9 products emerged out of the funding. Only 10 % of the answering firms (N = 625) reported that more than 10 products were developed out of FFF-projects – with one firm, which answered that 300 products were developed (!).

The share of new and marketable products out of FFF funding on all new developed products is quite high for some firms. For 25 % of the respondents the share lies between 0 and 10 % of all new products. Two thirds of the answering firms see the ratio up to 50 %. But for another 20 % of the respondents the share on new products is between 70 and 100 % of all new products or services.

Correspondingly the share of FFF induced products on the total turnover of the last business year reaches for some firms a ratio of 100 %. However, for 25 % of the respondents the share on total turnover lies between 0 and 5 %. Up to 50 % of the responding firms assess the share between 0 and 15 % and for 20 % of the firms the share lies between 50 and 100 % of the total turnover.

Figure 24: Impact of FFF funding on product development (y-axis = cumulative percentage of answering firms)

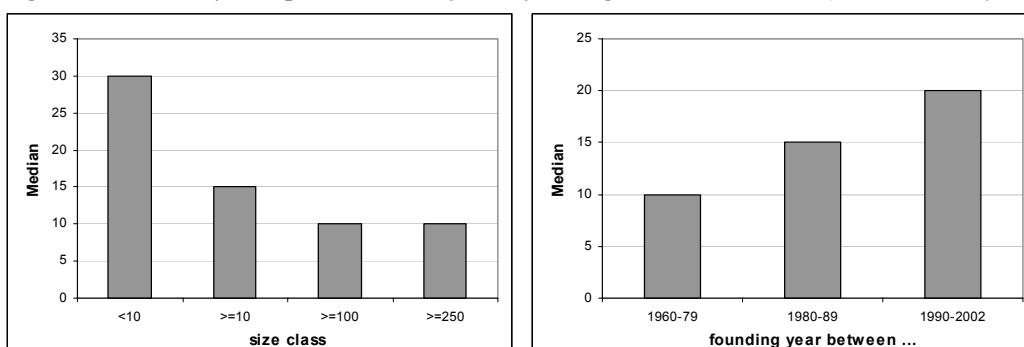


Source: survey

Now the question emerges of what are the characteristics of the firms, which attribute a high share of new developed products and/or a high share of turnover to formerly FFF funded projects?

We focused on the average share of new products out of FFF funding on total turnover of the last business year. We analyzed the sample according to two categories: by size classes and by founding year. The results are presented in Figure 25 and show that for young firms as well as small firms the share of new products out of FFF projects is higher compared with the other categories. The Median within the group of micro-firms is 30 – meaning that 50 % of the answering micro firms indicate higher and 50 % indicate lower shares. The same approach can be applied to firms according to their founding year. Younger firms exhibit a higher share than older firms.

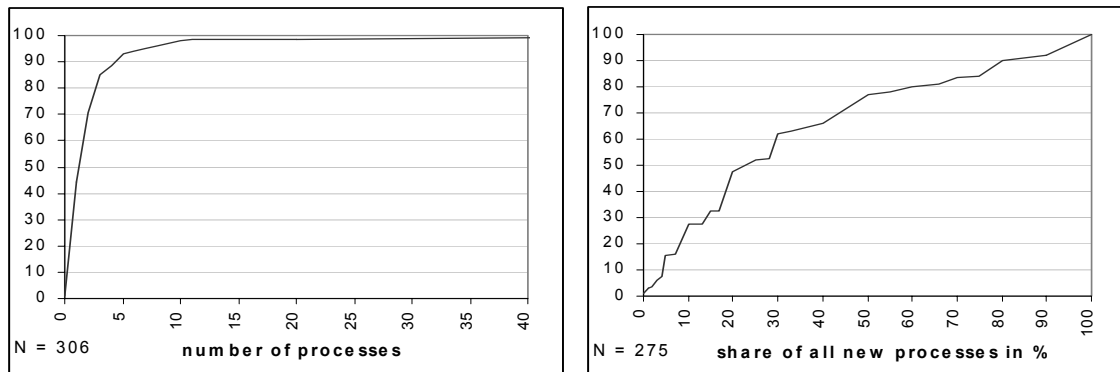
Figure 25: Share of new products out of FFF funding on total turnover (last business year)



Source: survey

The same kind of analysis was used for calculating the attribution of the FFF for the period 1995-2003 in relation to process developments. About 90 % of the respondents assessed that not more than 5 process innovations came out of the FF funded projects. Concerning the share on all process innovations half of the responding firms said that the FFF contributed to about 20 % of all their process innovations in the period 1995-2003.

Figure 26: Impact of FFF funding on product development (y-axis = cumulative percentage of answering firms)



Source: survey

7 Main findings

As to the effect of FFF subsidies on firms' own R&D expenditures, the *input additionality*, the evidence of our exercise can be interpreted as leaning towards complementarity: the results of the econometric analysis certainly point in this direction. Although any econometric analysis is bound to be confronted with a vast amount of 'noise' hidden in the data base and all other effects and sources of R&D funding are not completely covered in the data base it is unlikely that this would completely alter the results – after all, the FFF accounts for about 80 % of all public subsidies to private R&D.

As for the leverage of FFF funding's numerical value, the analysis seems to place it at about 40 %; 1 additional euro of funding (or, to be more precise, of its present value) induces firms to contribute an additional 40 cents of their own money. Both very small and large firms seem to exhibit higher leverage, small and medium-sized firms smaller leverage. Additionally, the leverage estimates for firms which perform R&D only occasionally are higher than for regular R&D performers.

The aggregate mean R&D subsidy ratio, i.e. the ratio of the sum of R&D subsidies to the sum of total R&D expenditures, is about 4 % on average and quite stable over time. Furthermore, the R&D subsidy ratio ranges between 18 % in the smallest firm size class (0-9 employees) and 2.4 % in the largest size class (500 or more employees). Firms that are five years old or less have an R&D subsidy ratio of about 5.8 % as compared to that of firms that are more six or more years old with an R&D subsidy ratio of about 3.5 %.

The results further show that the ratio of R&D subsidies to total R&D expenditures is significantly negatively related to both firm size and the current R&D to total turnover ratio (i.e. R&D intensity). This suggests that firms with a high R&D intensity have a lower R&D subsidy ratio. Furthermore, the R&D subsidy ratio continuously decreases with firm size, i.e. the bigger the firms the smaller the R&D subsidy ratio. Firms founded in the last five years have a significantly higher R&D subsidy ratio of about 31 % on average compared to firms that are 6 or more years old. In contrast, fast output growth in the past is associated with a lower R&D subsidy ratio. This indicates that the fund gives higher subsidies to firms with a low output growth in the past.

Concerning the productivity effects, or *output additionality*, we find that the amount of FFF subsidies as well as privately financed R&D expenditures (both expressed as a percentage of sales) have a significant and positive effect on the labour productivity growth rate in the following years. In particular, we find that the productivity effects of publicly funded R&D were much lower than those estimated for privately funded R&D as indicated by the elasticities. However, the magnitude of the productivity effect of R&D subsidies is quite large given the median R&D subsidy sales ratio of 0.4 %. Given the elasticities, we also calculate how much of the observed change in output per worker can be attributed to the effects of publicly and privately funded R&D. Combined, both funding sources of R&D account for 24 % of the change in output per worker per year. However, the contribution of publicly funded R&D is less than 2 %-points, while the contribution of privately funded R&D is 22 %-points.

FFF funding has effects on *behavioural additionality*, i.e. long-term behavioural changes emerge from FFF-participation. Descriptive evidence from the survey data revealed that FFF-funding is generating various dimensions of behavioural additionality:

- Around 80-85 % of the sample firms experience some degree of project additionality, i.e. in the case of rejection the project would have been cancelled or carried out in a revised version.
- Acceleration additionalities arise for two in three firms, i.e. FFF-funding is speeding up the course of the project. To put it the other way round, two out of three respondents agree that in the case of rejection the project results could only be exploited at some later date than originally aimed at.
- The share of companies appreciating scale additionalities ranges between 60-74 %. Thus rejected firms state to have carried out the project on a smaller scale when FFF-support had been denied.
- At least every other firm reports scope additionalities to have arisen from collaboration and a fraction of over 62 % benefits from scope additionalities in as far as new research areas could be entered with the financial help of the FFF-scheme.

Results from some subsequent econometric exercises based on the linked company-project FFF-database turned out not that conclusive, however. It turns out that the demand for high-skilled R&D personnel is only marginally affected by additional FFF-assistance. Instead, it is rather driven by fundamental performance indicators such as total annual sales. Especially larger firms hire additional R&D-labour irrespective of additional FFF-funding.

Some remarks on the FFF data base

The analyses of input and output additionality, as well as parts of the work on behavioural additionality, made extensive use of the FFF's data base of firm-level variables, i.e. turnover, cash-flow, export share, number of employees, number of research staff. When applying for FFF funding, firms have to provide these data for the three years prior to the application. By this design, firm-level and project-level data refer to completely separate periods. To be able to estimate the effect of FFF funding on firm characteristics (R&D expenditures in the case of input additionality, growth in turnover for output additionality), the analyses had to rely on firms which repeatedly approached the FFF, thus allowing for the construction of overlapping firm level-project level time series. By this approach, firms which ask for funding only rarely are in essence lost to this sort of analysis. Unfortunately, this introduces significant bias, as it is predominantly smaller and/or newer firms which are more likely to be "irregular" FFF customers.

To close this gap, the collection of firm-level data at later stages in the funding process should be introduced; the "final report" which has to be submitted at the end of each project would lend itself as a logical opportunity for collecting the "missing years".

To assess the (commercial) results of funded projects, moreover, the FFF might contemplate to add an additional layer to its report system. As a survey of FFF customers showed, a majority of projects reaches marketability some 2-3 years after the completion of the research phase. Thus, a short, standardized questionnaire (probably modelled on an abridged version of the survey used in this evaluation) could be sent to the performers of FFF-funded projects, asking about commercial results. The standard time frame for this follow-up survey should be about 4 years after the final report. For very long-term (and costly) projects, a revisit after maybe a further 2 to 6 years could provide valuable information on the long horizon.

This questionnaire, of course, could also be used to construct even longer time series of firm-level data. For all firm-level data, at whatever stage they are collected, a higher degree of standardisation would be welcome: for example, it is not clear which definition of cash-flow should

be used. Similarly, when asking about the number of research staff and research expenditures, firms should be provided with a short explanation of who and what should be included in these numbers – and also whether research staff is supposed to be a simple head count or rather measured in full-time equivalents. Additionally, the FFF should require in the application form the NACE code of economic activity of the firm (in addition to the assignment of codes for the economic activity of the project).

Concerning the implications of FFF-funding from the clients' point of view, as reported in a *survey* of 1,300 firms, the following findings can be summarized:

- To cover high costs of R&D, to reduce the risk of technical development and the FFF's reputation for providing quick aid, are the main motivations for firms submitting projects to the FFF. Furthermore the empirical results showed that the FFF fulfils these expectations: the FFF primarily funded technologically more advanced (difficult) projects, which were more expensive and of a longer duration.
- The majority of firms reported that FFF funded projects have no impact on other R&D projects within the firm. On average only $\frac{1}{4}$ of the firms said that FFF funded projects led to an extension of existing projects or were the reason to start a new one. This implies that FFF funding is important and intensifies R&D activities of the funded firms. But it also shows that FFF funding has no direct impact on other R&D projects in $\frac{3}{4}$ of the funded firms. So the FFF support helps to pursue the already existing "research path" within the firm but has not enough financial power to encourage firms to start "radically" new projects or to push a lasting technological change within the economy.
- 63 % of all firms stated that they have achieved their technical goals in an FFF funded project totally, 35 % partly and 2 % have admitted a failure.
- Commercialisation: In more than half of the cases FFF support leads to a new or improved product or service. The commercialisation of the results of FFF funded projects enables the firm in around $\frac{2}{3}$ of the cases to open up new markets.
- R&D projects are funded by the FFF at every stage of development. R&D projects are more likely to be funded than less advanced one. Having in mind that an R&D project becomes less risky the closer it comes to commercialisation, one can conclude that the FFF funds more risky projects more often than less risky ones.

An average assessment of FFF-support based on all funded and already finalised projects for the period 1995-03 shows the following:

- The average time length between end of project and the commercialisation of the product or service was about 2 years whereby bigger firms are more time-efficient in commercialising the research results, i.e. they are fast in terms of time to market than smaller firms.
- About 18 % of the clients reported licence revenues out of the FFF project. The median value of revenues was 110 tsd. € for the total sample.
- Four out five respondents reported a market-oriented project outcome, i.e. a marketable products / services or processes.
- Related to the share of new products out of FFF funding on all new developed products and on the total turnover the FFF turns out to be important for smaller as well as younger firms.

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Appendix I: Definition of NACE-codes

NACE	Definition
1	Agriculture, hunting and related service activities
2	Forestry, logging and related service activities
5	Fishing, operation of fish hatcheries and fish farms; service activities incidental to fishing
10	Mining of coal and lignite; extraction of peat
11	Extraction of crude petroleum and natural gas; service activities incidental to oil and gas extraction excluding surveying
12	Mining of uranium and thorium ores
13	Mining of metal ores
14	Other mining and quarrying
15	Food products and beverages
16	Manufacture of tobacco products
17	Manufacture of textiles
18	Manufacture of wearing apparel; dressing and dyeing of fur
19	Tanning and dressing of leather; manufacture of luggage, handbags, saddlery, harness and footwear
20	Manufacture of wood and of products of wood and cork, except furniture; Manufacture of articles of straw and plaiting materials
21	Manufacture of pulp, paper and paper products
22	Publishing, printing and reproduction of recorded media
23	Manufacture of coke, refined petroleum products and nuclear fuel
24	Manufacture of chemical and chemical products
25	Manufacture of rubber and plastic products
26	Manufacture of other non-metallic mineral products
27	Manufacture of basic metals
28	Manufacture of fabricated metal products, except machinery and equipment
29	Manufacture of machinery and equipment n. e. c.
30	Manufacture of office machinery and computers
31	Manufacture of electrical machinery and apparatus n. e. c.
32	Manufacture of radio, television and communication equipment and apparatus
33	Manufacture of medical, precision and optical instruments, watches and clocks
34	Manufacture of motor vehicles, trailers and semi-trailers
35	Manufacture of other transport equipment
36	Manufacture of furniture; manufacturing n. e. c.
37	Recycling
40	Electricity, gas, steam and hot water supply
41	Collection, purification and distribution of water
45	Construction
50	Sale, maintenance and repair of motor vehicles and motorcycles; retail sale of automotive fuel
51	Wholesale trade and commission trade, except of motor vehicles and motorcycles
52	Retail trade, except of motor vehicles and motor cycles; repair of personal and household goods
55	Hotels and Restaurants
60	Land transport; transport via pipelines
61	Water transport
62	Air transport
63	Supporting and auxiliary transport activities; activities of travel agencies
64	Post and telecommunications
65	Financial intermediation, except insurance and pension funding
66	Insurance and pension funding, except compulsory social security
67	Activities auxiliary to financial intermediation
70	Real estate activities
71	Renting of machinery and equipment without operator and of personal and household goods
72	Computer and related activities
73	Research and development
74	Other business activities
75	Public administration and defence; compulsory social security
80	Education
85	Health and social work
90	Sewage and refuse disposal, sanitation and similar activities
91	Activities of membership organizations n. e. c.
92	Recreational, cultural and sporting activities
93	Other service activities
95	Private households with employed persons
99	Extra-territorial organizations and bodies

Appendix II: Sample distribution by sectoral affiliation and sub-branch

Sectoral affiliation and subbranch	N	% within branch	% within grand total
Traditional industries	310	100.0	23.9
Manufacture of food, beverages & tobacco	42	13.6	3.2
Manufacture of textile & leather	16	5.2	1.2
Manufacture of wood & wood products	35	11.3	2.7
Manufacture of paper & paper products	15	4.8	1.2
Printing, publishing & allied industries	8	2.6	0.6
Manufacture of rubber & Plastic	36	11.6	2.8
Manufacture of non-metallic mineral products	22	7.1	1.7
Basic metal & alloys industries	50	16.1	3.9
Manufacture of metal products and parts	68	21.9	5.2
Manufacture of furniture, jewelry & musical instruments	18	5.8	1.4
R&D-intensive industries	393	100.0	30.3
Manufacture of basic chemicals and chemical products	57	14.5	4.4
Manufacture of machinery & equipment	151	38.4	11.7
Man. of office computing & accounting machinery and parts	9	2.3	0.7
Man. of apparatus for generation and transmission of electricity	24	6.1	1.9
Manufacture of apparatus for radio broadcasting, TV transmission & communication engineering	16	4.1	1.2
Manufacture of medical, surgical, and scientific and measuring equipment	100	25.5	7.7
Manufacture of transport equipment & parts	36	9.2	2.8
Traditional services	145	100.0	11.2
Recycling, power- and water-supply	23	15.9	1.8
Building trade and civil engineering	68	46.9	5.2
Wholesale trade, retail trade, trade & repair of motor vehicles	23	15.9	1.8
Tourism & hotel business	3	2.1	0.2
Transport and traffic	10	6.9	0.8
Sewage and rubbish disposal and other disposal	18	12.4	1.4
Knowledge-intensive services	365	100.0	28.2
News transmission, broadcasting, TV	6	1.6	0.5
Credit & insurance agencies and allied services	1	0.3	0.1
Software, data-processing and database	160	43.8	12.3
Research & development	106	29.0	8.2
Enterprise-related services	63	17.3	4.9
Teaching, instruction & education	10	2.7	0.8
Health, veterinary medicine and social services	10	2.7	0.8
Cultural industries, sports and entertainment	9	2.5	0.7
Other branches/miscellaneous	72	100.0	5.6
Agriculture, hunting, forestry and fishery	13	18.1	1.0
Extraction of mineral oil, natural gas and allied services	4	5.6	0.3
Mining & quarrying	11	15.3	0.8
Other services	44	61.1	3.4
Missing industry affiliation	11	100.0	0.8
Grand total	1296		100.0

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