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**When subsidized R&D-firms fail,
do they still stimulate growth?
Tracing knowledge by following
employees across firms**

Abstract:

Public R&D subsidies aim to target particularly risky R&D and R&D with large externalities. One would expect many such projects to fail from a commercial point of view, but they may still produce knowledge with social value. Such knowledge is likely to be embodied in workers or teams of workers. I utilize a large matched employer-employee data set and test for knowledge diffusion from subsidized technology firms transmitted through the labor market. The specific case analysed is a series of Norwegian IT-programs so far considered unsuccessful, but which have been linked to the rise of a new generation of successful IT-firms. It has been argued that know-how and networks built up in leading companies during the programs still 'fertilize' the IT-industry even though many of the companies have exited. I find limited support for this claim. On the positive side, the market value of work experience from subsidized firms does not seem to have been reduced by the fact that the firms did not succeed commercially, but workers from subsidized firms have not outperformed similar workers without this experience, either. Furthermore, firms that are spin-offs from formerly subsidized firms seem to perform below, rather than above average.

Keywords: R&D-subsidies, Knowledge spillovers, Human capital, Labor mobility,

Displaced workers, Spin-off firms, IT-industry, Program evaluation, Matched employer-employee data

JEL classification: J24, J31, J62, O32

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1 Introduction

Public R&D subsidies aim to target particularly risky R&D and R&D with large externalities. One would expect many such projects and firms to fail from a commercial point of view, but they may still produce knowledge with social value. Such knowledge is likely to be embodied in workers or teams of workers. In order to assess it's value, one need to trace workers as they move across firms and industries seeking to maximize the returns to their human capital. Consider the early days of the semiconductor industry as an example of the potential importance of this approach. If evaluating the social returns to R&D contracts awarded pioneering firms such as Sprague Electric, Shockley or Fairchild based on the performance of these firms alone, it seems clear from historical accounts that the return would appear modest. Yet, it is well documented in case studies that key technologies later utilized in the semiconductor industry by tremendously successful companies like Intel, was developed in these early entrants and transferred by employees to new firms better suited to exploit the technologies commercially, see e.g. Holbrook et al. (2000), Jackson (1997) or Saxenian (1994).

A possible 'scrap value' associated with unsuccessful R&D projects and firms may significantly influence the social returns to R&D and reduce the overall risk associated with technology programs. This issue has so far not been investigated in the technology program evaluation literature, nor has there been much empirical analysis of labor market knowledge flows or spin-off firms in general. The recent availability of large matched employer-employee data sets, however, makes it possible to analyze statistically the importance of human capital and employee mobility.

This paper 're-evaluates' a series of Norwegian technology programs in the 1980s that subsidized IT manufacturing firms. A previous evaluation by Klette and Møen (1999) concluded that "the IT-programs were largely unsuccessful". Later, however, claims have been made that the growth of the Norwegian IT-industry in the late 1990s was stimulated by knowledge built up in formerly subsidized firms. In particular, employees of the fallen industry leader, Norsk Data, have been pointed to as key contributors in a new generation of successful firms. Norsk Data was a 'national champion' and a leading minicomputer company. It was considered the third most profitable computer company in the world in the mid 1980s, but had considerable difficulties in adapting to the technology shift in the late 1980s represented by the introduction of PCs and open standards. In 1989 mass layoffs were unavoidable, in 1991 it closed down its manufacturing plants and in 1993 what little was left of the company went bankrupt.

One expression of the idea that Norsk Data had a lasting impact on the industry,

can be found in a publication from the Research Council of Norway (2000) presenting IT (ICT) firms and technologies that have benefitted from R&D subsidies. In the introduction the Council states that¹

“[t]he bankruptcy in Norsk Data received much attention, and left the impression that the Norwegian ICT industry was severely injured. This was not the case. Know-how was embedded in the employees, and these employees were rather quickly absorbed by other Norwegian ICT-firms.”

It may not be very surprising that the Research Council in this way tries to improve upon the public impression of Norsk Data, given that the firm had received massive subsidies². A similar, but even stronger statement, however, was made by Norway’s leading engineering magazine, *Teknisk Ukeblad*, one year earlier. In the fall of 1999, this bulletin of the Norwegian Engineering Association wanted to elect the ‘engineering achievement of the century’³. Second of ten nominees was Norsk Data. The magazine argued that this ‘industrial adventure ... left behind a thousand professionals whose knowledge still fertilize Norwegian information technology’⁴.

It seems that the statements quoted above are based on knowledge about a handful of cases. Both the Research Council and *Teknisk Ukeblad* mention e.g. Dolphin Interconnect Solutions, a company that came out of the R&D department in Norsk Data when it closed down. In 2000 a part of Dolphin was sold to Sun Microsystems and in the business press, the price was pictured as sensational. Such ‘spin-off returns’ from previous investments cannot be captured by ordinary microeconomic program evaluation methodologies which focus on the performance of the subsidized firms. In

¹In my translation.

²Norsk Data was the largest recipient among firms subsidized by the National Program for Information Technology lasting from 1987 to 1990, and received more than 12 percent of the budget allocated to commercial R&D under the program. Given the size of the company, this does not necessarily imply that the subsidies were large relative to Norsk Data’s private R&D investments, but money from the National Program for Information Technology came on top of subsidies from preceding programs and substantial public procurements which were used actively to help the company develop new technology throughout its history. Cf. Harlem et al. (1990) and Bjerkan and Nergård (1990).

³Cf. Valmot (1999). A list of all nominees is given in the same journal (*Teknisk Ukeblad*), August 12th 1999, pp 10-11.

⁴My translation. Spelled out in more detail: “All over Norway we see spin-off effects from the Norsk Data era; thousands of people that worked in or with Norsk Data built up know-how whose existence it is hard to imagine without this company. Many of these people started new firms together with old colleagues or business contacts, others have contributed with their experience in other sectors of the economy.” The article was titled “The lighthouse of the Norwegian IT-industry”.

order to evaluate whether Dolphin and similar cases are representative, a quantitative framework utilizing matched employer-employee data is called for.

Using such data, I find that scientists and engineers with experience from subsidized IT-firms to a much larger extent than other scientists and engineers in high-tech industries migrated to the rapidly growing IT service industry. They have not performed bad, but there is no evidence indicating that these scientists and engineers played a particularly prominent role in the growth process, either. Nor do spin-off firms from the subsidized firms perform particularly well. In fact, they seem to have performed below, rather than above, average. One possible explanation for these discouraging results is that the technology shift in the late 1980s rendered much of the intellectual human capital built up under the programs obsolete.

The rest of this paper is organized as follows: The next section discuss the data, the empirical approach and the definition of key variables. Section three describes the flow of scientists and engineers out of subsidized and non-subsidized firms. Section four analyze the value of experience from subsidized IT-firms using wage regressions on a sample of scientists and engineers with experience from high-tech and IT-industries. Section five analyze the performance of spin-off firms, while section six concludes.

2 Data and empirical approach

2.1 Data

The data used in this study come from four main sources: Governmental administrative records prepared by Statistics Norway, the biennial R&D survey conducted by the Royal Norwegian Council for Scientific and Industrial Research and Statistics Norway, the manufacturing statistics of Statistics Norway, and the statistics of accounts for non-financial joint-stock companies prepared by Creditinform and Statistics Norway. The Norwegian data are extraordinary in the sense that the entire working population can be traced across employers over more than a decade, and in the sense that extremely rich information is available both about the workers and about their employers. Cf. the data appendix in the working paper version, Møen (2002), for further details and descriptive statistics.

2.2 Hypothesis and approach

Figure 1 compares employment growth in subsidized firms with employment growth in other categories of high-tech firms. There is a strong decline in employment in sub-

sidized firms⁵. Given this picture, the dismal conclusion of Klette and Møen (1999), evaluating the technology programs based on firm level data, are not surprising. However, as suggested in the quotes in the previous section, this interpretation may be misleading. A more positive way to read Figure 1 is to stress that workers were leaving the subsidized firms on a large scale, and that they may have contributed to growth elsewhere. Figure 2 pictures the growth in the Norwegian IT industry, as defined by OECD, from 1995 to 1999. In these years the IT service industry grew considerably faster than the rest of the private sector. The hypothesis under consideration in this paper, is that the boom in R&D subsidies and R&D investments in the Norwegian IT manufacturing industry in the mid and late 1980s, later caused growth in this or other sectors of the economy through transfer of knowledge embodied in people. Establishing a possible causal link of this type is demanding and involves constructing a counterfactual situation for the firms and workers involved.

Compared to the standard program evaluation literature, cf. e.g. Heckman, Lalonde and Smith (1999), several complications are present. First, the ‘treatment’ is not dichotomous. R&D investments have both an intensity dimension and a time dimension. Moreover, there is no clear-cut start of the program as various technology programs have replaced each other for several decades prior to the period that can be observed⁶. Also, the selection problem, fundamental to all program evaluation where participation is not randomized, has a peculiar twist. There is a ‘double selection’ process where firms are selected into programs, and workers self-select into firms. Deciding on a relevant and valid comparison group under these circumstances is difficult.

My responses to the problems listed above will be as follows: First, with respect to the intensity and time dimension of treatment, I will use a regression framework so that continuous variables can be utilized in addition to a dichotomous classification, based on cut-off values. Next, with respect to missing data for previous programs, little can be done. I will, however, argue below that this is not a severe obstacle. Finally, my response to the potential selection problem will be to allow for individual fixed effects. A more explicit approach to the selection problem is difficult due to lack

⁵Employment in non-subsidized R&D firms and other R&D firms appears to fluctuate more than the other two categories simply because there are fewer workers behind these graphs. The strong decline in employment for non-subsidized IT R&D firms from 1992 to 1993 is driven by one single firm that ran into trouble. Much of the subsequent growth is due to the same firm recovering. The negative employment growth in subsidized IT R&D firms is not driven by Norsk Data alone. Leaving out this company does not alter the picture significantly. Furthermore, looking at sales growth gives a very similar picture, but then one cannot keep track of plants which change industry classification from manufacturing to services.

⁶Cf. Klette and Møen (1999) for details.

of good instruments for subsidy awards and career choice. However, positive selection creates a bias against my conclusion that the programs were not successful and negative selection is not particularly relevant since R&D programs are meant to stimulate high quality research⁷.

2.3 Defining ‘treatment’

Defining IT R&D-firms, other R&D-firms and high-tech Treatment, in the context of this paper, is having work experience from a subsidized R&D firm in the IT manufacturing industry. In principle, therefore, we would like to compare similar workers with experience from IT-firms with and without subsidies. However, it is difficult to define an IT-industry since information technology does not constitute a separate class in standard industrial classification schemes⁸. Too narrow a set of classes will leave out a lot of true IT-firms, whereas a broader set will include a lot of non-IT firms. I get around this problem by utilizing a unique variable in the R&D surveys which identify the IT-content in each firm’s R&D investments. Using this variable in combination with R&D man-years, I define IT R&D-firms in the manufacturing sector as firms with an intensity of IT-related R&D above 10 percent⁹. This definition is designed to exclude a large number of firms that perform small IT projects without having information technology as their main focus or being technologically advanced. Similarly, I define R&D-firms in general as firms with an intensity of total R&D (IT and non-IT) above 10 percent¹⁰.

Almost without exception, units classified as IT R&D-firms according to the above definition belong to ISIC 382-385 (NACE 29-35), i.e. the machinery, electronics, transportation equipment and technical instruments industries. I will hereafter refer to these industries together as ‘high-tech’.

Defining subsidized firms Since subsidies are awarded unevenly among recipients, there is also a problem of how to define a subsidized IT R&D-firm (hereafter referred

⁷This is not to say that negative selection could not exist. Various political economy processes may lead the subsidies to troubled firms, cf. Klette and Møen (1999) for a discussion. Then, however, the programs would not look successful, nor be successful.

⁸Cf. e.g. OECD (2000).

⁹The R&D-variables are not available, nor as relevant, for the IT service sector. This sector will be defined using the OECD definition based on industrial classification codes.

¹⁰R&D intensity is measured as R&D man-years per employee (per year) at the three-digit line of business level within firms. Cf. the data appendix in Møen (2002) for more information. In the text, I will not distinguish between firms and lines of business within firms.

to as a subsidized firm). For a subsidy to have an effect on a firm's research activities, it must be of some significance. Hence, any subsidy should not qualify, and I define the treatment group as IT R&D-firms with an intensity of subsidized IT-related R&D above 0.5 percent. For a treatment firm with an intensity of IT-related R&D at the lower limit, i.e. 10 percent, this implies that at least 5 percent of the firm's IT-R&D must be subsidized¹¹. The criteria is designed so that all large subsidy recipients known from other sources, that can be identified in the data, are included.

Defining the treatment period Data on individual workers start in 1986, and the era of large R&D subsidies ended in 1990. Hence, I will consider the years 1986 to 1990 to be the 'treatment period'. As mentioned, there were targeted IT-programs prior to 1986, but I do not believe the lack of data from these early years is a severe restriction. The largest R&D subsidy program were in effect from 1987 to 1990, and the largest IT R&D contracts were awarded in the years 1985 to 1987. Furthermore, with some stability in employment relationships, a certain persistence in program participation, and both a lag and some persistence in the effect of subsidies, there will be a positive correlation between the unobserved and the observed treatment.

Categorizing workers I want to assess the value of the core technological know-how built up in the subsidized firms. This know-how is likely to be possessed by scientists and engineers, and my analysis will therefore focus on this group. With the treatment period lasting from 1986 to 1990, many scientists and engineers will have had several employers, and firms may also have changed subsidy status within this time interval. I categorize scientists and engineers as having 'experience from subsidized firms' if they are attached to a subsidized firm in at least one year. Similarly scientists and engineers are categorized as having 'experience from IT R&D-firms' and 'experience from R&D-firms' if they have at least one year experience from such firms in 1986 to 1990.

Sample size The employee sample consists of male scientists and engineers born after 1935 and employed full time in a high-tech firm at least one of the years 1986 to 1990¹². Altogether there are 3784 scientists and engineers in the sample. 3419 of these

¹¹I know for each firm the share of R&D that is classified as IT, but not the share of subsidies used in IT-projects. However, since the government had IT high on its agenda, I assume that R&D-subsidies awarded to firms that report to do IT R&D is related to their IT-projects. If subsidies exceed a firm's IT R&D-investments, the excess subsidies are excluded.

¹²I have excluded women because they are known to have different career patterns and preferences than men, and do not constitute a large share of the labor stock in these industries.

are in firms with known R&D-investments. There are 1755 scientists and engineers with experience from R&D-firms. Out of these 1290 have experience from IT-R&D firms. In this group 1095 have experience from subsidized firms. About a quarter of the workers in subsidized firms were employed by the industry leader, Norsk Data.

There are altogether 1173 plants (constituting 957 firms) with known R&D in the high-tech industries in the period 1986 to 1990. 197 plants belong to ‘R&D firms’, i.e. having an intensity of total R&D above 10 percent. Out of these, 108 belong to ‘IT R&D-firms’ and 79 belong to subsidized firms.

There are on average 4.0 observations of each plant in the years 1986 to 1990. Note that firms, and thereby plants, can change category between years. When giving the number of plants in different categories above, plants are counted as belonging to an R&D firm or IT R&D firm if it has this status in at least one of the years 1986-1990.

Continuous treatment variables The firm categories defined above are based on cut-off values for R&D intensities that are somewhat arbitrary, and that conceal a significant amount of variation in research and ‘program’ exposure. The intensity of R&D and subsidies varied between firms within each category, and within firms over time. Furthermore, workers may have stayed with several employers during the program years. In many of the analyses that follow it is possible to use such continuous variation in treatment, and therefore I construct a stock measure of experience in addition to the dummies. This is done by attaching to each worker information about his employers R&D investments, and adding up intensities in R&D, IT R&D and subsidized IT R&D over the years 1986 to 1991. I use these sums as measures of the human capital accumulated¹³.

2.4 A description of workers and firms by treatment category¹⁴

IT R&D firms are concentrated in the following industries: Computer and office machinery, Other machinery, Radio, TV and communication equipment, Insulated cables

¹³Since the intensities are measured in man-years per employee per year, the unit of the ‘experience stocks’ are years. This should not be interpreted literally, however. It will only be a precise measure of individual R&D experience if all workers participate equally in the firms’ R&D projects. This is obviously not the case, and one should rather think of R&D intensities as proxies for how much there is to learn in a firm at a given time. Summing the intensities over the time dimension then gives a measure of on-the-job learning.

¹⁴Detailed tables with descriptive statistics on workers and firms are given in the data appendix in Møen (2002).

and wires, Professional and scientific instruments, and Photographic and optical goods. Except for computers, non of these industries are dominated by IT R&D firms, however. Subsidized and non-subsidized IT R&D-firms coexist in most industries mentioned. Other R&D firms and non-R&D firms are represented in a wider set of subindustries than the IT R&D firms. These industries comprise the production of various types of machinery, electrical equipment and transport equipment¹⁵.

An important thing to notice is that the larger part of the IT-industry received subsidies. There are 1095 scientists and engineers with at least one year of experience from subsidized IT-firms and 195 that only have IT experience from non-subsidized firms. Given that the authorities were determined to stimulate the IT-industry, this is perhaps not surprising, but it leaves a relatively small, and possibly non-random, control group. That being said, however, there are very few observable differences between workers in subsidized and non-subsidized IT R&D-firms. Furthermore, my analysis is not dependent on this dichotomous classification, as I also utilize continuous experience variables as explained above¹⁶.

Subsidized firms are somewhat larger, more unionized and more likely to have a rural location than non-subsidized firms. They are also more often foreign owned and younger. The most interesting difference, however, is that subsidized firms had significantly higher growth rates in the years preceding the awarded subsidies. Presumably, recent success must have been an important criteria when subsidies were awarded. With respect to intensity in R&D and IT-R&D the two groups of firms are close to identical. ‘Other R&D firms’ are somewhat less R&D intensive than IT R&D-firms and have a slightly lower educational level, but they are more capital intensive. Non-R&D-firms have an even lower educational level than R&D-firms and are more unionized and less often foreign owned. Non-R&D firms are clearly the oldest group of firms.

With respect to educational composition, subsidized firms are slightly more diversified with respect to the human capital they possess than non-subsidized firms. All R&D-firms, however, even non-IT firms, are highly intensive in electrotechnical engineering skills. Non-R&D firms also employ many workers of this type, but mechanical engineers is the most dominant skill group in these firms.

Summing up the differences between subsidized and non-subsidized IT R&D-firms, the main impression left by the descriptive statistics is that workers in subsidized and non-subsidized firms are quite similar, although there are some differences between the two types of firms. In particular, the technology programs seem to have favored firms

¹⁵About 82 percent of the worker-year observations are from firms with R&D information available.

¹⁶This creates substantial variation, as subsidies were very unevenly distributed across firms. This was part of a long tradition where ‘national champions’ were considered important catalysts for growth.

with rapid growth.

3 Tracing workers out of the subsidized firms

A natural first step when analyzing potential growth effects brought about by labor mobility out of subsidized firms, is to see where the technical expertise became employed later on. The results of such an analysis are presented in Table 1. The first column shows the industry of occupation in 1997 for scientists and engineers with experience from subsidized firms. The main comparison group is scientists and engineers with experience from IT R&D-firms that were not subsidized. These are tabulated in column 2. Columns 3 and 4 give mobility patterns for scientists and engineers with experience from other R&D-firms in the high-tech industries, i.e. firms whose research activities were not strongly IT-related, and scientists and engineers without experience from R&D-intensive firms.

The main difference between subsidized and non-subsidized IT R&D-firms is that a much higher share of scientists and engineers from the subsidized firms has moved to IT-service industries. 30 percent of scientists and engineers from subsidized IT-firms became employed in the IT-service industry¹⁷ versus 14 percent of scientists and engineers with experience from non-subsidized IT-firms. The other columns show that the less IT and R&D intensive the firms, the less likely are the scientists and engineers to move to the IT service sector. The table suggests that the subsidized IT-activities were service related, or at least that the IT-service industry offered the best opportunities for scientists and engineers from subsidized firms when these firms closed down.

3.1 A brief summary of some ‘non-wage’ labor market outcomes

The main message to take away from Table 1, is that *the possibility* of a link between R&D subsidies awarded in the 1980s and growth in the IT-service sector in the 1990s, is present in the data. Next, I investigate how workers from the subsidized firms actually performed in the labor market. Were e.g. workers from the subsidized firms “rather quickly absorbed” in the labor market, as claimed by the Research Council? Some indicators that can throw light on this issue are reported in Table 2. Row 1 reports the share of displaced workers that had to move to a different municipality to find a new job. Row 2 reports the share of workers who participated in active labor market

¹⁷Looking separately at workers from Norsk Data, the share is as high as 46 percent.

programs. Row 3 reports the average employment rate following the program, row 4 reports the share of workers who took further education and finally row 5 reports the share of workers that became self-employed¹⁸. Taken together, the results do not suggest that workers from subsidized firms had any particular difficulties in finding new jobs. Having established this, I will move on to analyze earnings.

4 Wage regression analyses

If know-how built up in the subsidized firms was not firm-specific and provided a basis for growth in other firms later on, we would expect experience from subsidized firms to have higher value in the labor market than experience from other firms. This assertion can be tested using extended Mincer (1974) wage regressions. Lacking a ‘pre treatment’ period, I start out exploring the effect of R&D and R&D subsidies on scientists and engineers’ wages during the program. Next, I investigate wage growth following the program and check the results obtained from these two analyses against the wage level after the program. Given that know-how built up in the industry leader Norsk Data has been considered particularly valuable, and that about one quarter of all scientists and engineers with experience from subsidized firms have worked for this company, I investigate the robustness of all results with respect to leaving out these workers.

4.1 The effect of R&D and subsidies on wages during the program

Several mechanisms related to R&D, IT and subsidies may possibly have affected wages during the program period. First and foremost, if scientists and engineers expected to accumulate more general knowledge in subsidized firms (or in IT firms in general) than in other firms, they should be willing to pay for this through lower wages¹⁹. To the extent that subsidized firms promoted more advanced technologies, and technologies considered to have a large future potential, such investments in general human capital are conceivable, although risk aversion and liquidity constraints on the worker side may reduce the effect. Another mechanism, possibly affecting the wages, is that subsidized firms may have employed scientists and engineers of better (unobserved) quality. High-

¹⁸These numbers may be artificially low. Presumably, they do not include workers who are employed in joint-stock companies that they own themselves. Self-employed are included in the wage analyses presented in the next section.

¹⁹This follows from classical human capital theory, cf. Becker (1962, 1964) and the discussion in Møen (2005).

ability workers are necessary to develop frontier technologies, but high-ability workers may also have a preference for working in a technologically advanced environment²⁰. The net effect of this on wages is not obvious. On one hand, high-ability workers have better outside options, but workers with a preference for technologically advanced firms may, on the other hand, accept wages below their outside option²¹. A final possible mechanism is unions. The wage level in subsidized firms would be affected if the workers were able to negotiate higher wages and thereby extract some of the subsidies as rents.

Table 3, column 1, 2, 5 and 6 explores the wage level for prime aged male scientists and engineers in high-tech industries in the program years by including measures of R&D, IT R&D and subsidized IT R&D in a standard wage regressions. Both a dummy variable approach (column 1 and 5) and a specification with continuous variables (column 2 and 6) are reported. The dummy approach utilizes the dummies for R&D firm, IT R&D firm and subsidized IT R&D firm described in section 2. Note that these dummies are nested in the sense that a subsidized firm is also an IT R&D firm which is also an R&D firm. In specifications with continuous variables, I use intensities measured as the share of the work force doing R&D, IT R&D and subsidized IT R&D. These variables are also nested, so that in order to find the total effect of a marginal increase in IT R&D due to a subsidized project, all three of the reported coefficients should be added.

In all regressions, workers in non-R&D firms is the baseline comparison group. Non-reported control variables are listed in the subtext to the table. Among these variables are 15 dummies for different academic degrees, hence, scientists and engineers are compared within detailed educational groups.

Table 3, column 1 and 2, does not distinguish between subsidized and non-subsidized IT R&D, and from Part A of the table, using the full sample, we see that the wage level in IT R&D firms is significantly below the wage level in other R&D firms. The average discount is between 2 and 4 percent. Non-IT R&D, however, does not seem to affect wages. When distinguishing between subsidized and non-subsidized IT R&D, a puzzling pattern appears. The dummy approach suggests that the lower wage level is

²⁰The work of Almeida and Kogut (1999), Stern (1999) and others suggests that scientists and technical personnel have preferences regarding the technological environment they work in.

²¹Rosen (1986) provides a review of the theory of compensated differentials (equalizing differences). Stern (1999) shows that this mechanism has relevance for scientists in the private sector. This is, in the setting of my paper, supported by Steine (1992) who states that the company policy of Norsk Data was to pay the same as similar firms, or somewhat less. He adds, “[i]t was attractive to work in Norsk Data, so why be a wage leader?” (p. 50, my translation).

associated with work in subsidized firms while the specification with continuous variables suggests that the lower wage level is associated with work in non-subsidized firms. A clue as to how these conflicting results can be reconciled can be found in Part B of the table where workers from Norsk Data are excluded. Column 1 and 2, suggest that the observed lower wage level in IT R&D firms is driven mainly by workers in Norsk Data. If Norsk Data received enough IT subsidies per worker to be classified as a subsidized firm, but had, relative to other firms, far higher total investments in IT R&D per worker, this may explain the observed coefficients in Part A, column 5 and 6. This is not inconceivable. When sources like Bjerkan and Nergård (1990) describe Norsk Data as a thoroughly subsidized company, they are not so concerned with direct R&D subsidies as with preferential public procurement, and Norsk Data is in this respect a special case²². The company is also special in a different respect relevant for my analysis. The company was famous for rewarding their employees with shares, something that received much attention in the business press. The discount that the employees received when buying shares was counted as taxable labor income and is therefore included in my wage measure²³, but the stock market price of the shares increased so rapidly and for so many consecutive years, that the employees were likely to value the opportunity to buy shares in the company highly and trade this off against ordinary wage compensation. Hence, some (but probably not all, cf. footnote 21) of the apparent discount associated with Norsk Data may be an artifact of the company's unusual compensation scheme and not a true compensating differential²⁴.

Looking at Table 3B, column 5 and 6, we see that even when workers from Norsk Data are excluded, there is still a wage discount associated with workers in subsidized firms. Both the dummy specification and the intensity specification suggest that the discount is slightly less than 2 percent compared to non-subsidized IT R&D firms, although only the intensity specification produces a significant coefficient²⁵. Above I have mentioned several mechanisms that may be behind this. In order to distinguish between some of these possible mechanisms, the analysis is extended by interacting R&D variables with experience, thereby examining wage profiles rather than average

²²Cf. footnote 2.

²³Cf. Steine (1992, p. 54-55).

²⁴As far as I know, this wage policy was unique for Norsk Data at the time, as were their consistently rising stock price. Stock options were not much used in the sample years. Due to a very unfavourable tax treatment between 1991 and 1999, it was not much used in later years, either. For these reasons, labor earnings is likely to be a fairly accurate measure of monetary compensation in other companies than Norsk Data.

²⁵For the intensity specification, the discount is derived by multiplying the coefficient -0.488 with 0.036, the employers' average intensity in subsidized IT R&D, cf. table A2 in Møen (2002).

wage levels. If the wage discount in subsidized firms is due to workers investing in general human capital, one would expect it to be associated with young workers taking a wage cut when entering the firms and then experiencing stronger wage growth as their expectations about the value of on-the-job training become fulfilled²⁶.

Table 3, column 3, 4, 7 and 8 gives the results of including R&D, IT R&D and subsidized IT R&D, interacted with workers' experience. In column 3 and 4, we see that scientists and engineers have a steeper wage profile in IT R&D firms than in other firms. Consistent with the idea that IT is a general technology, cf. e.g. Bresnahan and Trajtenberg (1995), these firms appear to offer lower wages early in the career in exchange for higher wage growth thereafter. The beginning wages in IT R&D-firms are about 10 percent lower than in other R&D firms, and the annual wage growth is about 0.5 percent higher²⁷. Interestingly, there are no significant differences between R&D firms that don't specialize in IT and non-R&D firms.

Moving on to column 7 and 8, distinguishing between subsidized and non-subsidized IT R&D firms, one finds that the wage profile in subsidized firms is less steep than the wage profile in non-subsidized firms. Hence, there is nothing in the data suggesting that investments in general human capital were particularly large for workers in subsidized firms.

4.2 The effect of experience from subsidized firms on wages later in the career

A key issue in this study is how experience from subsidized firms affected wages later in the career. Table 4 contains the results of an analysis of ten year wage growth from 1986 and 1987 to 1996 and 1997. The advantage of looking at wage growth is that potential differences in ability between workers are accounted for, and looking at the full ten year interval takes one from one boom in the economy to the next. This is

²⁶Workers may also pay for learning through lower wages later in their career, but that will be difficult to separate from the wage premia they receive on their previous human capital investments, cf. Møen (2005). From a theoretical point of view, their willingness to invest in human capital should fall gradually towards retirement.

²⁷The dummy and the intensity specification give very similar results. Taking into account the special wage policy of Norsk Data discussed above, and looking instead at part B, it may seem as if 10 percent is rather large. If the correct wage discount for entering workers is between 6 and 7 percent, and the wage growth is between 0.4 and 0.5 percent, as suggested in Part B, this imply a pay-back period of about 15 years. IT R&D-intensity is a noisy variable, and as a proxy for human capital, it probably becomes increasingly noisy the further into the career a worker has reached. This implies that measurement errors will bias the coefficient on the interaction term towards zero.

desirable, since it may be difficult to capture the full program effect before the labor market has adjusted to the many mass layoffs caused by the recession.

The sample consists of full time working male scientists and engineers, having at least one year full time experience from high-tech or IT industries, including services, in 1986-1997. Workers who only have experience from IT services are included for two reasons. First, it has some interest to compare workers entering the expanding IT service industries with background from manufacturing high-tech industries to workers who have acquired most of their work experience within the IT-service industries²⁸. Second, these workers help identify the many control variables in the wage regression, such as experience and dummies for industries, altogether 72 coefficients²⁹. Given the relatively small number of workers with experience from non-subsidized IT firms, it is important to identify common coefficients as precisely as possible³⁰.

At first sight, the results in Table 4A, column 1 and 2, seem to imply that workers in IT R&D firms have had significantly higher wage growth than other workers. Looking, however, at column 3 and 4, and Part B, we see a pattern very similar to the one found in Table 3 and discussed in detail above. This suggests that the significant growth results are driven by a possible mismeasurement of compensation for workers in Norsk Data in the beginning of the period. When excluding these workers, there is only a small and non-significant wage growth effect left, i.e. workers with experience from IT R&D-firms have a slightly higher wage growth than workers with experience from other firms, and workers with experience from subsidized IT R&D-firms have a slightly higher wage growth than workers with experience from non-subsidized IT R&D-firms, without any of these differences being significantly different from zero.

Table 5 reports the results of an analysis of the effect of experience from R&D, IT R&D and subsidized IT R&D-firms in the program years on the wage level in the years 1996 and 1997. Consistent with Tables 3 and 4, the results show that there

²⁸As it turns out, there does not seem to be any important differences between these groups, and I have not tabulated separate coefficients for workers that only have experience from IT service industries. On average, these workers seem to receive slightly lower wages than workers with experience from high-tech manufacturing.

²⁹The industry dummies do not follow a particular NACE or ISIC level. Within high-tech and IT-industries I use a detailed categorization, usually at the five digit level. In less advanced sectors, with fewer observations in the sample, the dummies are usually at the two or three digit level. Cf. the subtext to Table 4 for a full list of control variables and other details regarding the regression.

³⁰The assumption that there is a common experience profile, common industry effects and so on, is of course not obvious, but it seems to be a reasonable approximation. Furthermore, my conclusions are robust to reducing the sample size by excluding workers without experience from firms that have invested in IT R&D.

are no significant differences related to the various types of experience. In particular, workers with experience from subsidized firms, started out with a small but significant (using the intensity specification) average wage discount, and had slightly higher, but not significantly higher, wage growth, and they have ended up with a slightly lower, although not significantly lower, wage level as reported in Table 5³¹. Changing the specification in Table 5 by including firm specific fixed effects, and thereby asking whether workers with experience from subsidized firms have ended up in the best paid positions within their firms, give very similar results to the specification without firm specific fixed effects and is not reported.

Before concluding the wage analysis, one should reflect on how the results in Table 4 and 5 relates to Table 3, column 3, 4, 7 and 8, which indicated that workers in IT R&D-firms, whether subsidized or not, accepted a wage discount at the start of their career and experienced higher wage growth later on. If the estimated wage growth associated with a career in IT R&D firms had continued after the program period, it obviously should have caused a significant positive coefficient on experience from IT R&D firms both in Table 4 and 5³². When there is no such positive effect, it implies that these workers did not receive the return they expected. One possible interpretation is that their expectations did not come through because of the technology shifts in the IT-industry in the late 1980s.

Is the glass half empty or half full? Tables 3 through 5, can be summarized in one sentence: *Scientists and engineers with experience from subsidized IT R&D-firms performed about as good, or rather as bad, as workers from non-subsidized firms.* Workers in all IT R&D firms seem to have ‘co-financed’ their employers’ R&D investments by accepting wages below their alternative wage, presumably believing that work experience from these firms would provide general human capital. The expected wage

³¹If including the years 1994 and 1995 in addition to 1996 and 1997, the coefficient on experience from subsidized firms in column 4 becomes marginally significant. With respect to workers with experience from Norsk Data, a detailed investigation of Table 5, contrasting Part A with Part B in light of the previous discussion of subsidies and IT R&D investments in this company, suggests that these workers have wages below the average for other workers with experience from subsidized firms. If running a similar regression for skilled workers with secondary technical education, however, I find a significant positive wage premium for workers with experience from Norsk Data. This may suggest that scientists and engineers accumulate more firm specific human capital, and is more exposed to technological risk than workers with secondary technical education.

³²In Table 5 this is so because the average worker with experience from IT R&D-firms, even if continuing to invest in on-the-job training by staying in such a firm, should have caught up with and passed workers without such experience by 1996/97.

growth, however, did not materialize after the program period, and they received no monetary payback from their investment. With respect to workers in subsidized firms, they do not seem to have gained anything from participating in the subsidized projects. Consequently, *my analysis does not support the idea that the IT R&D programs created significant benefits for workers with experience from subsidized firms*. On the positive side, however, workers in subsidized firms did not perform particularly bad, either, even though many of them became displaced in the late 1980s as shown in Figure 1³³. Hence, the fact that many of the subsidized firms did not succeed commercially does not seem to have reduced the market value of work experience from these firms. This could be interpreted as a positive program effect, but we would expect the effect to be stronger if these workers were truly ‘fertilizing’ the IT-industry with their human capital.

5 The performance of spin-off firms

A complementary approach to looking at the performance of individual workers, is to focus on the performance of spin-off firms defined by groups of workers that have stayed together. When several workers from the same firm continue to work together, it is reasonable to assume that they are exploiting know-how built up in their previous work environment, and that there are positive complementarities between them that make them stay together. It is also possible that firm profits is a better performance measure than wages, particularly if the spin-off firms to some extent are worker-owned. Low tax rate on capital income relative to labor income may induce employee-owners to substitute wages for return on stocks³⁴, and employee-owners may also sacrifice wages in order to finance firm growth³⁵.

³³Note that I control for displacement in the wage regressions in Table 4 and 5, but the variable is not significantly different from zero. Distinguishing, however, between workers with experience from subsidized firms who have stayed with the same firm, and separators, I find a modest negative effect for separators (not reported). In the stock specification this negative effect is significant.

³⁴Note, however, that the Norwegian tax system have detailed rules in order to avoid this type of tax evasion.

³⁵One may also think that employee stock options plans would reduce the relevance of taxable labor income as an earnings measure, and show up in firm profits. This kind of options, however, has been very unusual in Norway due to an unfavorable tax regime, cf. footnote 24.

5.1 Sample and definition of spin-offs

Table 6 present the results of my analysis of spin-off firms. Roughly speaking, i.e. leaving out some of the finer details to be laid out below, I define *a spin-off firm as a firm that was not originally subsidized, but where at least 25 percent of the employees have experience from a firm that was subsidized.*

The sample period is 1994-1997, i.e. the years when the IT industry recovered according to Figure 2. The sample consists of all non-financial joint-stock companies with more than one employee and at least one scientist or engineer, in industries with at least one ‘program firm’, a firm that to a large extent draw on human capital with experience from subsidized IT R&D firms. Formally, I define *program firms* as firms that have, at some point, had at least a 25 percent share of employees with experience from subsidized firms, and at least one scientist or engineer with experience from a subsidized firm. A definition of this type will necessarily be a bit arbitrary, but the idea is to identify firms that draw significantly on knowledge that was built up under the program.

The definition of program firms does not distinguish between continuing subsidized firms that has retained experienced workers, and new firms, spin-offs, employing workers with experience from subsidized firms. This is because I want to start out by tracing all firms drawing on ‘program know-how’. Utilizing information about plants, however, I can identify those of the program firms that represent a continuation of originally subsidized firms³⁶. I label these ‘continuing or reorganized subsidized firms’. This group of firms is defined as program firms that contain one or more plant that in 1986-1990 belonged to a subsidized firm. Program firms that do not fall into this category are defined as spin-off firms. According to the above definitions, there are altogether 109 program firms in the sample, 76 of these are spin-off firms and 33 are continuing or reorganized subsidized firms.

5.2 Results

Program firms are somewhat larger, more capital intensive, more R&D intensive, and more intensive in use of scientists and engineers, than non-program firms. They are also

³⁶Firm identification numbers represent legal units, and will change if one firm or plant is bought by another firm, etc. Plant identification numbers, on the other hand, will change only if the production is physically moved or substantially altered with respect to industry classification. Not all registers that are matched to produce my data set, however, use the same plant and firm identification number system. For this reason the match between plants and firms, and the tracking of units over time, is slightly imperfect.

somewhat younger and less often in a rural location. Spin-off firms are significantly younger and smaller than continuing or reorganized subsidized firms, as one would expect. Spin-offs are also less R&D-intensive, but more human capital intensive. This reflects that a larger fraction of the spin-off firms belong to service industries. 37 percent of the spin-off firms can be identified as spin-offs from Norsk Data.

The first performance measure I consider is sales growth. The results are reported in Table 6, panel A. Program firms perform slightly better than other firms along this dimension, but the difference is not significant. Moving on to profitability, Table 6, panel B-D presents return on sales, return on assets and return on equity, respectively. It shows that program firms are significantly *less* profitable than other firms. On average they have 1.2 percent lower return on sales, 3.2 percent lower return on assets and 15.5 percent lower return on equity.

Looking separately at spin-offs and continuing or reorganized subsidized firms, we see that the significant negative results are exclusively associated with the spin-off firms. It is difficult to explain these coefficients, but one possibility is that spin-off firms mostly consist of troubled remnants of previously subsidized units, and that they are kept running because their core know-how has low alternative value³⁷. Analyzing wages in spin-off firms (not reported), I find some support for this hypothesis. Scientists and engineers with experience from subsidized firms that work in spin-off firms, have a small wage discount. Workers with experience from subsidized firms that work in continuing or reorganized subsidized firms, on the other hand, have a significant wage premium. This may suggest that the most valuable know-how built up under the program is to be found in the surviving plants and not in the spin-off firms. In any case, *my analysis does not give support to the idea that important returns from the IT-program ended up outside the originally subsidized firms through labour mobility.*

5.3 Robustness

In all the firm performance analyses presented above, I have controlled for firm age, firm size, intensity in use of scientists and engineers, current R&D-investments, business cycle effects, and industry differences. The main results are robust to leaving out these control variables, but without controls, also continuing or reorganized subsidized firms have a profitability below average.

Since the exact definition of program and spin-off firms is based on a somewhat arbitrary cutoff value for the share of employees that has experience from firms that

³⁷E.g. sales or service departments, or production teams, that either are reclassified with respect to industry or move to a new location, and try to continue on their own.

received subsidies, it is particularly important to test the robustness of the results with respect to these definitions. I have tried both a more inclusive definition, looking at firms with a 10 percent share of employees with experience from subsidized firms, and a more exclusive definition looking at firms with a 50 percent share of employees with experience from subsidized firms. In both cases, the main results in Tables 6 hold true. Defining spin-offs based on the share of engineers with experience from subsidized firms, rather than the share of employees with experience from subsidized firms reduces the significance of the negative coefficients. Finally, I have looked specifically at spin-offs from Norsk Data. If anything, these firms have a weaker performance than other spin-off firms. With respect to a possible time trend in performance, cf. the strong industry growth present in Figure 2, I find that the profitability of the spin-off firms is falling over time.

Given that the returns to innovation is known to have a very skewed distribution, one may also question whether the regression analyses reported above correctly represent *aggregate* profits for the different categories of firms. A few large and profitable spin-off firms could possibly more than outweigh the low profits in the many small firms dominating the sample. One simple way to explore this issue is to pool all spin-off firms, all continuing or reorganized subsidized firms, and all non-subsidized and non-spin-off firms, in order to compute the joint performance of the various groups. The result of this exercise is graphed in Figure 3. When assessing the joint performance this way, spin-off firms as a group have a higher return on sales than non-spin-off firms, but they perform worse with respect to sales growth, return on assets and return on equity.

A final question one may ask with respect to robustness, is whether the results are specifically related to the subsidized IT R&D firms, or whether any spin-off from firms that invested in IT R&D in the late 1980s have performed similarly bad. I have looked at this question by defining spin-offs from all R&D firms and all IT R&D firms in the same manner as I have defined spin-offs from subsidized IT R&D firms. This analysis (not reported) show that the negative results are most strongly associated with spin-offs from subsidized firms. There are, however, very few spin-offs from non-subsidized IT R&D firms in the sample. In a related analysis (also not reported) I have regressed firm profitability on a continuous measure of different types of R&D experience among the firms' scientists and engineers. In this analysis, R&D-, IT R&D- and subsidized IT R&D experience is measured in the same way as in the wage regressions presented in Tables 4 and 5. The results do not confirm the negative effect of subsidies found in the spin-off analysis, but nor do firms whose scientists and engineers have particularly much experience from subsidized firms perform significantly better.

5.4 Remarks on profitability as performance measure

An objection to the spin-off analysis might be that current sales and profitability are not relevant performance measures in the IT industry, and that the spin-off firms may become successes in the long run. Admittedly, numerous companies in the “New Economy” were unprofitable, and still highly valued in the stock market due to large investments in intangible capital. These arguments are not entirely convincing, however, as the extraordinary high stock market value of such firms did not last. Also, private owners buying a company where previous owners have lost their money, may make the company look successful and produce positive profits, without there being a positive *social* return to the historical R&D investments that produced the technology. Comparing total investments to expected future profits is difficult and requires case studies.

A particularly interesting case in the Norwegian IT-industry is Dolphin Interconnect Solutions. This company has been considered the most successful spin-off from Norsk Data, cf. section 2, but did not make positive profit in any of the sample years. The founding engineers started to develop the ‘Dolphin SCI technology’ in 1988 while still working for Norsk Data, and 1999 was the first year in history that the company generated positive profits³⁸. Rough calculations suggest that total investments in Dolphin amounts to about NOK 500 million³⁹. In 2000 a major part of Dolphin was sold to Sun Microsystems and the price, NOK 171 million, was considered very favorable. Per employee, the price was NOK 8 million, something which is more than 10 times the cost of an engineering man-year. However, if the ‘major’ part of the company sold to Sun represents more than one third of the total value of the company, the rate of return to Dolphin as an investments project has been negative. A market based evaluation, therefore, is not likely to make Dolphin come out as a large success.

6 Conclusion

This paper illustrates how matched employer-employee data can be used to assess whether human capital built up in subsidized firms is general, and whether subsidies to firms that fail stimulate growth elsewhere through labor mobility. The case con-

³⁸Cf. <http://www.dolphinics.com>.

³⁹This number is calculated on the basis of articles written about Dolphin in the major newspapers Aftenposten, Dagens Næringsliv and Bergens Tidene in the years 1991-2001. The number is adjusted for inflation. Using an additional 7 percent discount factor, the total investment amounts to NOK 800 mill. About 20 percent of the investments seems to have been financed by public subsidies.

sidered is a series of Norwegian IT-programs from the mid and late 1980s. I find no evidence suggesting that experience from subsidized firms has been rewarded with a wage premium. Scientists and engineers with experience from subsidized firms receive on average the same wage as otherwise similar workers without such experience. This suggests that the human capital built up in subsidized firms was general enough that the commercial failure of these firms did not hurt the average workers' career in the long run. It does not, however, support the idea that the human capital built up in the subsidized firms was particularly valuable and have constituted an important stimulus to growth. Analyzing the performance of spin-off firms reinforces this negative conclusion. Spin-offs from subsidized firms are less profitable than other firms, suggesting that the identified spin-offs to a large extent consist of troubled remnants of previously subsidized units. My analysis, therefore, does not give support to the idea that important returns from the IT-programs ended up outside the originally subsidized firms through labor mobility.

One question my analysis cannot answer, is how the subsidies affected the overall supply of IT-engineers in Norway. The IT-programs may have stimulated talented workers to specialize in IT rather than in some other field, and this may have pushed down wages for IT-personnel in the years after the program. Indirectly, this may have aided later growth in the industry, even though this growth did not draw on technologies developed under the program. A more complete analysis of the programs would have to assess whether the experience of the workers attracted to the IT industry by the programs, would have been better or worse in some other industry, and whether firm entry and growth in the IT industry in the 1990s was sensitive to the observed industry wage level. Answering such counterfactual questions requires out of sample analyses that is beyond the scope of the present study.⁴⁰ The main goal of the programs, however, targeting general technologies with a large potential for growth, did not succeed.

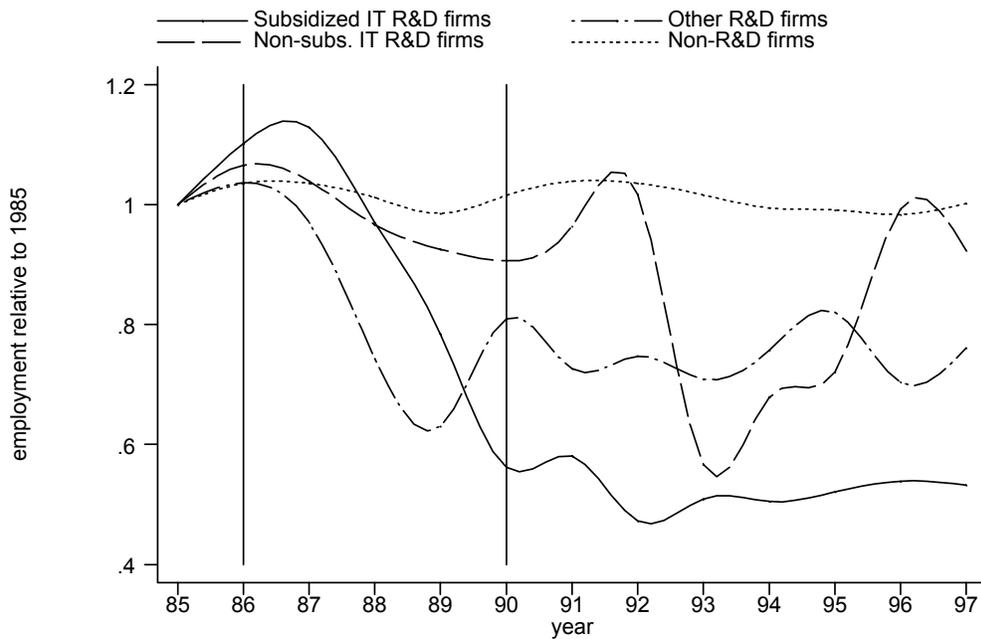
⁴⁰A complete analysis should also assess other spillover channels than labour mobility, although this is the mechanism emphasized by proponents of the program.

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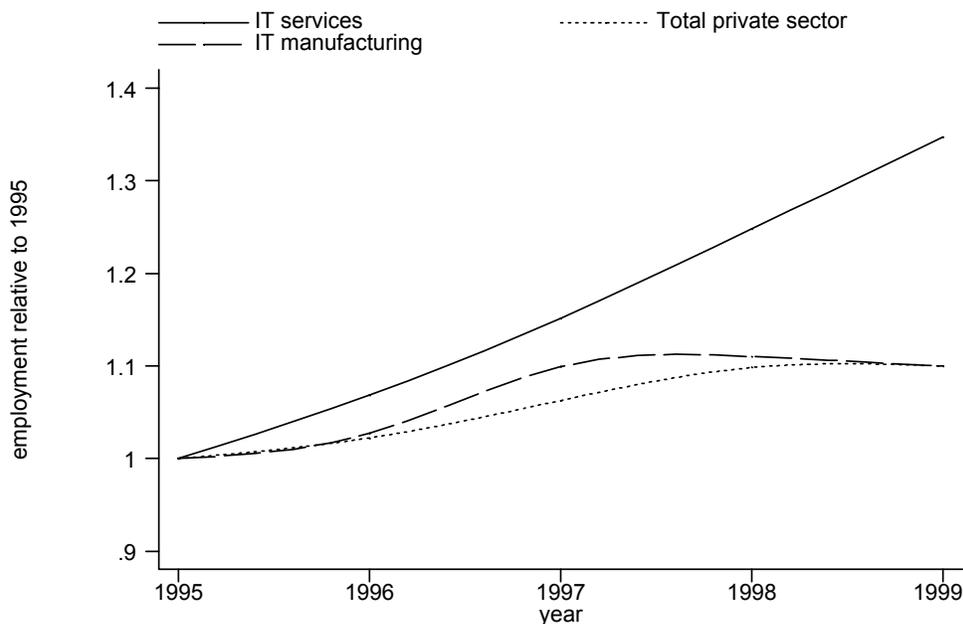
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Figure 1: Employment growth 1985-1997 in subsidized IT R&D-firms vs. other categories of firms in the high-tech industry



In 1985 there were about 11 100 workers in subsidized IT R&D firms, 1 800 workers in non-subsidized IT R&D firms, 5 800 workers in other R&D firms and 58 600 workers in non-R&D firms. Firms with unknown R&D-intensity are excluded. Firms that change industry classification are kept in the sample. Annual data points are connected using a cubic spline.

Figure 2: Employment growth in IT vs. all private industries in 1995-1999



Source: Statistics Denmark (2000) updated with numbers from Statistics Norway (www.ssb.no). Annual data points are connected using a cubic spline.

Figure 3: Joint growth and profitability of spin-off firms vs. non-spin-off firms in 1994 to 1997

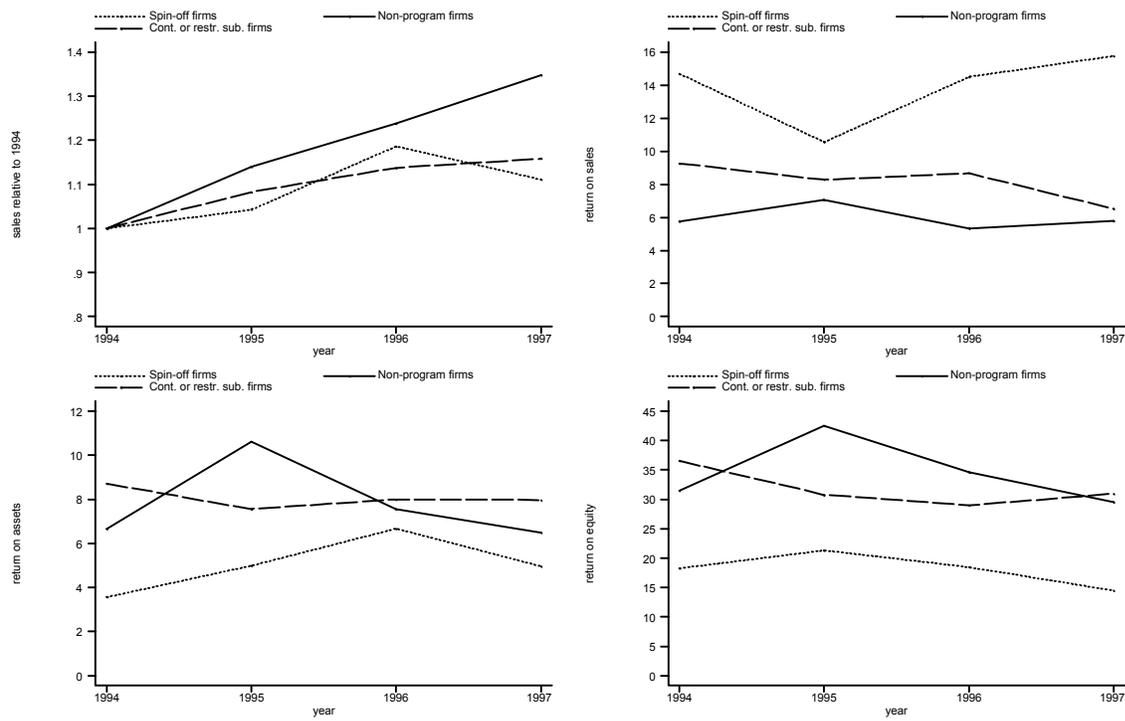


Table 1: Industry of occupation in 1997 for scientists and engineers with experience from high-tech industries in 1986-1990

	Workers from subsidized IT R&D firms	Workers from non-subsidized IT R&D firms	Workers from other R&D firms	Workers from non-R&D firms
High-tech manufacturing industries	40%	53%	49%	44%
Other manufacturing industries	2%	4%	3%	7%
IT services industries	30%	14%	10%	6%
Other services industries	12%	14%	15%	23%
Public sector	5%	5%	6%	4%
Other industries or unknown	2%	1%	8%	8%
Not in the sample	9%	9%	9%	7%
Number of scientists and engineers	1095	195	465	1664

Workers are classified in the leftmost column applicable. Workers who are not observed in 1997 are classified according to their industry of occupation in 1996, if possible. Otherwise they are classified as not in the sample. Workers that only have experience from firms with unknown R&D-intensity are excluded.

Table 2: Non-wage labour market outcomes for scientists and engineers with experience from high-tech industries in 1986-1990

	Workers from subsidized IT R&D firms	Workers from non-subsidized IT R&D firms	Workers from other R&D firms	Workers from non-R&D firms
Average employment rate 1988-1997 [‡]	88%	87%	89%	89%
Participated in active labour market programs 1988-1997	13%	11%	11%	11%
Displaced workers that were re-employed in a different municipality ^{**}	11%	11%	14%	14%
Re-educated or further educated by 1997	2.1%	2.1%	1.7%	1.7%
Self-employed in at least one year after 1990	.01%	.01%	.02%	.01%
Number of scientists and engineers	1095	195	465	1664

Workers are classified in the leftmost column applicable. Workers that only have experience from firms with unknown R&D-intensity are excluded.

[‡] Those not employed include everyone who is not employed and not under education, regardless of whether they are registered as unemployed or not. Part time workers are counted as part time unemployed.

^{**} A displaced worker is defined as a worker with at least two year tenure who left a plant that downsized at least 25 percent in that year or over that year and next year.

Table 3: The effect of R&D, IT and IT-subsidies on the *wage level and wage profile* for scientists and engineers in high-tech industries in 1986-1990

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
<i>A: All observations</i>	Dummy	Intensity	Dummy	Intensity	Dummy	Intensity	Dummy	Intensity
R&D	.007 (.009)	.048 (.043)	.015 (.016)	.081 (.074)	.005 (.009)	.025 (.045)	.013 (.015)	.016 (.075)
R&D * experience			-.001 (.001)	-.004 (.005)			-.001 (.001)	.0001 (.005)
IT R&D	-.043*** (.008)	-.245** (.061)	-.109*** (.015)	-.600*** (.104)	.010 (.015)	-.270*** (.065)	-.088*** (.023)	-.696*** (.108)
IT R&D * experience			.006*** (.001)	.031*** (.008)			.007*** (.002)	.037*** (.009)
Subsidized IT R&D					-.040*** (.014)	.229* (.121)	-.027 (.022)	.837*** (.205)
Subsidized IT R&D * exp.							-.001 (.002)	-.051*** (.017)
R-squared	.50	.50	.51	.51	.51	.51	.51	.51
Number of observations	11 386	11 386	11 386	11 386	11 386	11 386	11 386	11 386
<i>B: Without workers with experience from Norsk Data</i>								
R&D	-.008 (.009)	.022 (.044)	-.004 (.015)	.057 (.075)	-.009 (.009)	.051 (.046)	-.005 (.015)	.049 (.075)
R&D * experience			-.0004 (.001)	-.004 (.005)			-.0004 (.001)	-.001 (.005)
IT R&D	-.015* (.008)	-.108* (.063)	-.064*** (.015)	-.426*** (.109)	.0002 (.015)	-.013 (.067)	-.070*** (.023)	-.426*** (.113)
IT R&D * experience			.004*** (.001)	.028*** (.008)			.006*** (.002)	.035*** (.009)
Subsidized IT R&D					-.019 (.015)	-.488*** (.125)	.009 (.022)	.138 (.211)
Subsidized IT R&D * exp.							-.003 (.002)	-.050*** (.017)
R-squared	.50	.50	.51	.51	.50	.50	.51	.51
Number of observations	10 513	10 513	10 513	10 513	10 513	10 513	10 513	10 513

The dependent variable is ln (real annual earnings). Control variables included in the regression, but not reported are a quartic in experience, a quadratic in tenure, a dummy for job relationships whose starting date is censored at April 30th 1978 together with its interactions with the two tenure variables, dummies for 15 different academic degrees, a quadratic in plant number of employees, dummies for 3 different regions, year dummies, year dummies interacted with experience, 6 industry dummies, 3 dummies denoting whether the R&D, IT or subsidy variable is missing and these dummies interacted with experience. The coefficients are estimated using ordinary least squares. Huber-White robust standard errors allowing for clustering of errors by individuals are in parentheses.

- *** Significant at the 1% level
- ** Significant at the 5% level
- * Significant at the 10% level

Table 4: The effect of R&D, IT and IT-subsidies on *wage growth 1986-1997* for scientists and engineers in high-tech and IT industries

	(1) Dummy	(2) Stock	(3) Dummy	(4) Stock
<i>A: All observations</i>				
R&D-experience	-.017 (.018)	-.005 (.031)	-.018 (.018)	.003 (.030)
IT R&D-experience	.042** (.018)	.069* (.038)	.003 (.028)	.093** (.042)
Subsidized IT R&D experience			.047* (.025)	-.155* (.072)
R-squared	.23	.23	.23	.23
Number of observations	7 130	7 130	7 130	7 130
<i>B: Without workers with experience from Norsk Data</i>				
R&D-experience	-.010 (.018)	.015 (.031)	-.011 (.018)	.010 (.030)
IT R&D-experience	.016 (.018)	-.003 (.039)	.011 (.028)	-.008 (.034)
Subsidized IT R&D experience			.007 (.025)	.045 (.077)
R-squared	.23	.23	.23	.23
Number of observations	6 762	6 762	6 762	6 762

The dependent variable is the first difference of ln (real annual earnings) between year t and year $t-10$ in the period 1986 to 1997. Control variables included in the regression, but not reported are a quartic in experience, a quadratic in tenure, a dummy for job relationships whose starting date is censored at April 30th 1978 together with its interactions with the two tenure variables, year dummies and dummies for 15 different academic degrees, a dummy for having experience from IT service, but not from high-tech manufacturing in 1986-1990, a dummy for not having experience from high-tech manufacturing, nor from IT service in 1986-1990, a dummy for being displaced in one of the years 1986 to 1993, 28 dummies for industry of occupation at time t , 28 dummies for industry of occupation at time $t-10$, two dummies denoting whether R&D or IT R&D is missing for those with experience from manufacturing firms and a similar dummy for subsidized IT R&D in column 4. The coefficients are estimated using ordinary least squares. Huber-White robust standard errors allowing for clustering of errors by individuals are in parentheses.

- *** Significant at the 1% level
- ** Significant at the 5% level
- * Significant at the 10% level

Table 5: The effect of R&D, IT and IT-subsidies in 1986-1990 on the *wage level* for scientists and engineers in 1996 and 1997 in high-tech and IT industries

	(1) Dummy	(2) Stock	(3) Dummy	(4) Stock
<i>A: All observations</i>				
R&D-experience	.012 (.017)	.035 (.027)	.012 (.017)	.037 (.036)
IT R&D-experience	-.007 (.017)	-.021 (.035)	.004 (.026)	-.014 (.038)
Subsidized IT R&D experience			-.012 (.024)	-.041 (.082)
R-squared	.21	.21	.21	.21
Number of observations	10 109	10 109	10 109	10 109
<i>B: Without workers with experience from Norsk Data</i>				
R&D-experience	.011 (.017)	.031 (.027)	.011 (.017)	.033 (.027)
IT R&D-experience	-.003 (.017)	-.009 (.036)	.005 (.026)	.004 (.043)
Subsidized IT R&D experience			-.009 (.025)	-.059 (.093)
R-squared	.22	.22	.22	.22
Number of observations	9 632	9 632	9 632	9 632

The dependent variable is ln (real annual earnings). Control variables included in the regressions, but not reported are a quartic in experience, a quadratic in tenure, a dummy for job relationships whose starting date is censored at April 30th 1978 together with its interactions with the two tenure variables, year dummies and dummies for 15 different academic degrees, a quadratic in plant number of employees, a dummy for being displaced in one of the years 1986 to 1993, dummies for 3 different regions, a dummy for having experience from IT service, but not from high-tech manufacturing in 1986-1990, a dummy for not having experience from high-tech manufacturing, nor from IT service in 1986-1990, 28 industry dummies, two dummies denoting whether R&D or IT R&D is missing for those with experience from manufacturing firms and a similar dummy for subsidized IT R&D in column 4. The coefficients are estimated using ordinary least squares. Huber-White robust standard errors allowing for clustering of errors by individuals are in parentheses.

- *** Significant at the 1% level
- ** Significant at the 5% level
- * Significant at the 10% level

Table 6: Performance in 1994-1997 in firms that employ knowledge developed in the subsidized IT R&D firms

<i>A: Sales growth</i>	(1)	(2)
Dummy for program firm	.064 (.044)	
Dummy for continuing or reorganized subsidized firm		.075 (.086)
Dummy for spin-off firm		.060 (.045)
R-squared	.93	.93
Number of observations	3 641	3 641
<i>B: Return on sales</i>		
Dummy for program firm	-1.22 (1.00)	
Dummy for continuing or reorganized subsidized firm		1.57 (1.48)
Dummy for spin-off firm		-2.56** (1.24)
R-squared	.08	.08
Number of observations	3 719	3 719
<i>C: Return on assets</i>		
Dummy for program firm	-3.15* (1.67)	
Dummy for continuing or reorganized subsidized firm		1.25 (2.51)
Dummy for spin-off firm		-5.26*** (2.00)
R-squared	.07	.08
Number of observations	3 719	3 719
<i>D: Return on equity</i>		
Dummy for program firm	-15.51** (7.57)	
Dummy for continuing or reorganized subsidized firm		2.57 (11.79)
Dummy for spin-off firm		-24.19*** (8.68)
R-squared	.06	.06
Number of observations	3 719	3 719

In panel A the dependent variable is $\ln(\text{Sales})$, $\ln(\text{Sales}_{t-1})$ is included as a regressor. Control variables included in all regressions, but not reported are a quartic in firm age, a quartic in firm no. of employees, a quartic in the share of employees that are scientists and engineers, a dummy for positive R&D-investments, a dummy for R&D-intensity above 0.05, a dummy for R&D-intensity above 0.2, a dummy for no information about R&D investments, year dummies and 38 NACE industry dummies. Firm age is deliberately censored at 30 and firm no. of employees is censored at 1000. The coefficients are estimated using ordinary least squares. Huber-White robust standard errors allowing for clustering of errors by firms are in parentheses. In panel B-D, the influence of outliers is reduced by replacing values for return on sales, assets and equity below the 5th percentile with the 5th percentile, and values above the 95th percentile with the 95th percentile.

- *** Significant at the 1% level
- ** Significant at the 5% level
- * Significant at the 10% level

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