The effect of hiring top workers on productivity: What is the role of absorptive capacity?

Magnus Lodefalk and Aili Tang
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The Effect of Hiring Top Workers on Productivity: What is the Role of Absorptive Capacity?∗

Magnus Lodefalk†
Aili Tang‡

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Abstract

We examine heterogeneous productivity effects of hiring top workers on small and medium-sized enterprises, using longitudinal employer-employee data. We find the productivity effect to be stronger for firms with higher absorptive capacity in terms of having a well-educated workforce, being in a knowledge-intensive industry or performing R&D. Technological laggards within an industry benefit more strongly from hiring top workers if their workforce is more well-educated.

JEL Classification: D22, D24, D83, J24, J62

Keywords: recruitment, knowledge spillover, firm growth, productivity, SME, absorptive capacity

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†Corresponding author: Lodefalk: Örebro University, SE-70182, Örebro, and Ratio Institute, Box-3203, SE-10364, Stockholm, Sweden (email: magnus.lodefalk@oru.se)

‡Tang: Örebro University (email: aili.tang@oru.se)
1 Introduction

Firms may acquire knowledge through recruitment that enables them to enter into higher productivity trajectories (e.g., Almeida and Kogut 1999; Gidehag and Lodefalk 2016; Parrotta and Pozzoli 2012). Recruitment and its usefulness for the hiring firm is likely to vary across workers but also across donating and receiving firms. As for the recruiting firm, it arguably needs a considerable level of organisational capacities to be able to capitalise on knowledge acquisition from external sources. The seminal contributions by Cohen and Levinthal (1989, 1990) emphasised the issues of capabilities and knowledge gaps of firms in effectively managing inwards technology transfer.\(^1\)

For small and medium-sized enterprises (SMEs), who dominate employment, hiring leading personnel can be a way to tap into the much larger pool of knowledge of ‘better’ firms.\(^2\) Yet such recruitment is understandably difficult, not only in terms of attracting such workers but to absorb their accumulated experience.\(^3\) SMEs are known to generally be less productive, more prone to exit and to have difficulties to hire qualified workers (Eurostat 2011; OECD 1998).

We study the role of the absorptive capacity for SMEs’ productivity effect from hiring managers and professionals – top workers likely to be key for profit-maximising firms’ growth (e.g., Glass and Saggi 2002; Dahl and Klepper 2015).\(^4\) Our results rest on employing state-of-the-art estimation of total factor productivity and within-firm and quasi-experimental techniques to a comprehensive longitudinal employer-employee dataset. The dimensions of the dataset enables us to pay attention to the characteristics of workers and donor firms and, importantly, the absorptive capacity of receiving firms.

We find the hiring effect on productivity to be stronger for firms with higher absorptive

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1Firms’ absorptive capacity, is a concept that has later subsequently been broadened to include a firm’s overall capacity for acquisition, assimilation, transformation and exploitation (Zahra and George 2002).

2Bloom and van Reenen (2010) substantiate the importance of management practices for firm performance and, i.a., find a strong correlation between human capital and good practices.

3There is plenty of evidence that firms’ own research and development is not only important in itself for productivity but for being able to absorb the insights from others’ R&D (e.g., Escribano et al 2009 and Griffith et al (2003))

4Mion and Opromolla (2014) note the unfortunate lack of research on the nexus between manager mobility and firm productivity. Neither Gidehag and Lodefalk (2016), who pursue the issue, nor the related study by Parrotta and Pozzoli (2012) on recruitment of technicians and post-secondary educated workers consider the role of the absorptive capacity of firms.
capacity in terms of having an well-educated workforce, being in an knowledge-intensive industry or performing R&D. Technological laggard firms within an industry benefit more strongly than other firms from hiring top workers if they have a more well-educated workforce.

2 Empirical approach

We set out to estimate the effect of knowledge inflow – through the hiring of new managers and professionals – on total factor productivity (TFP). To carefully address the endogeneity and simultaneity issues in estimating TFP – our response variable – we employ a slightly extended version of the semi-parametric estimator of Ackerberg el al. (2015) that pays additional attention to unobserved firm heterogeneity, drawing on Vandenberghe (2013). Basically, we first predict value added after estimating a production function – with physical capital and skilled and unskilled labour – plus a second-order polynomial of inputs as well as a proxy variable (material), which is assumed to costlessly adjust to short-term productivity shocks. Using the predicted value added, we then employ non-linear optimisation to estimate the output elasticity parameters. Finally, we estimate TFP, which then will be our response variable.\(^5\)

Our log-linear econometric model is specified as follows:

\[
E[a_{it-0}|MaP_{it-1}, O_{it-1}, Z_{it-2}, \upsilon_i] = \zeta_{MaP} MaP_{it-1} + \zeta_O O_{it-1} + Z_{it-2} \zeta_s + I_{it} \zeta_I + \upsilon_i \tag{1}
\]

where the expected conditional TFP \(a_{it-0}\) of firm \(i\) at time \(t-0\) is considered to be a function of the employment of managers and professionals \(MaP_{it-1}\) as well as other workers \(O_{it-1}\) at time \(t-1\); time-variant pre-hiring firm characteristics of row vector \(Z_{it-2}\) at time \(t-2\), and heterogeneity at the industry, year, municipality and firm levels of row vector \(I_{it}\) and the scalar \(\upsilon_i\).\(^7\) The key parameter of interest is the semi-elasticity \(\zeta_{MaP}\) that captures within-firm TFP-effects of hiring in the previous year, conditioned on

\(^5\)Gidehag and Lodefalk (2016) fully account for the method.

\(^6\)Log of firm size and age; multinational affiliation; and legal form.

\(^7\)The model is lagged partly since knowledge transfer through new recruits and its TFP pay-off is likely to hinge on the intensity and length in interaction between the recruit and the hiring firm, the pay-off, and partly to attenuate endogeneity concerns (e.g., Granovetter 1973; Lodefalk 2016).
covariates established two years prior.⁸

To analyse differential TFP-effects according to the absorptive capacity of the hiring firm, we classify firms according to the educational level of their workforce (1, 0 for > 50% having ≥ 3 years of post-secondary education), their knowledge-intensity (1, 0 for being in a knowledge-intensive service industry, if a services firm)⁹ and their research and development (R&D) status (1, 0 for spending on R&D in the pre-hiring years 2001-2002).

Additionally, we consider how the technological distance of a firm to the frontier firm of its 3-digit industry affects the TFP-effect of hiring top workers, and if higher absorptive capacity moderates the impact of such a distance on the hiring effect. Andrews et al (2015) find substantial and widening productivity gaps of firms within industries across time. We would expect that hiring top workers—especially hiring from frontier firms—could potentially help laggarding SMEs to catch-up. For our analysis, we define a firm’s distance $Tech - gap_{ijt}$ to the technological frontier $\text{max}TFP_{jt}$ as an index:¹⁰

$$Tech - gap_{ijt} = \text{max}TFP_{jt} - RTFP_{ijt}$$  (2)

where $RTFP_{ijt} = TFP_{ijt} - \overline{TFP}_{ijt}$ and $\overline{TFP}_{ijt} = \exp\left\{\frac{1}{n} \sum_{n=ijt} \ln(TFP_{ijt})\right\}$.¹¹

On the one hand, being a laggard firm may arguably increase the marginal benefit from acquiring knowledge, on the other hand it implies less absorptive capacity. Possibly, among laggards, those with higher absorptive capacity are more able to assimilate and exploit newly acquired knowledge. To examine this conjecture, we interact the technology gap and the educational level of the firm.

Finally, we test the robustness of our estimation results to remaining endogeneity concerns by employing a quasi-experimental technique. We employ a nearest neighbour propensity score matching and difference-in-difference (DiD) estimator.¹² We then compare the growth ($t - 1$ versus $t + 1$ and $t + 2$) of firms that hire managers and professionals with

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⁸The recruitment variables are not in logs because most SMEs do not recruit a particular year.
⁹We follow Eurostat’s classification at the 3-digit industry level.
¹⁰The index draws on the approach of Griffith et al (2005), which is to attenuate endogeneity concerns.
¹¹That is, first, the firm’s TFP is evaluated relative to a common reference point (the geometric mean TFP of the industry), and, second, compared with the maximum TFP of the industry.
¹²For details, see the online appendix.
the growth of matched–‘twin’–firms that were as likely to hire leading personnel but instead hired other workers.

3 Data and empirical results

We use matched employer-employee panel data that encompasses all Swedish firms with \( \geq 1 \) and \(< 250\) employees and all individuals of those firms who are at least 16 years old, with the data covering the 2001–2010 period.\(^{13}\) All datasets are from Statistics Sweden. Our matching is done using unique identifiers of individuals and firms. A snapshot of the dataset in year 2010 contains approximately 167,000 SMEs and 1,362,000 individuals.\(^{14}\) Most SMEs are typical micro-enterprises, without external engagements.\(^{15}\)

In Table 1, we present the results from estimating equation (1). The upper third of the table presents results for firms with higher versus lower expected absorptive capacity in terms of education, the middle section presents results for firms in and outside of knowledge-intensive industries, and the lower third of table displays results for firms with and without R&D expenditure.

On the role of absorptive capacity for the hiring effect, we find that hiring professionals is associated with within-firm TFP growth but mainly in firms with higher absorptive capacity. Comparing firms with a high or low level of education, and comparing knowledge-intensive and other services industry firms suggests that only the more \textit{ex ante} absorptive firms gain in terms of TFP from hiring professionals. The within-firm semi-elasticity is approximately 0.4 and statistically significant at least at the 5%-level, that is, hiring an additional professional is associated with 0.4% higher TFP next year. The pattern of results for firms with and without R&D is similar, although there is a statistically significant result also for the latter category of firms. Second, hiring managers is not linked to subsequent TFP, irrespective of the classification, in line with Gidehag and Lodefalk (2016).

\(^{13}\)Additionally, we require included firms to grow organically in the studied period to avoid the result being driven by factors related to merges and acquisitions; here, we make use of the Firm and Plant Dynamics register of Sweden.

\(^{14}\)The summary statistics, additional results and robustness checks are available in the online appendix. The dataset is similar to the one in Gidehag and Lodefalk (2016).

\(^{15}\)Firms with R&D-expenditures or that are active in knowledge-intensive services industries stand out in terms of hiring and employment of leading personnel.
Table 1: Hiring top workers, absorptive capacity and firm productivity - within-firm estimation results

<table>
<thead>
<tr>
<th>Absorptive capacity</th>
<th>Levels</th>
<th>Managers$_{t-1}$</th>
<th>Professionals$_{t-1}$</th>
<th>Others$_{t-1}$</th>
<th>Obs.</th>
<th>Adj.$R^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Education</td>
<td>High educ.</td>
<td>-0.00636</td>
<td>0.00438***</td>
<td>0.000458</td>
<td>174,136</td>
<td>0.656</td>
</tr>
<tr>
<td></td>
<td>Others</td>
<td>0.00753</td>
<td>0.00142</td>
<td>0.00566***</td>
<td>225,147</td>
<td>0.705</td>
</tr>
<tr>
<td>Knowledge-intensity</td>
<td>Intensive</td>
<td>0.00434</td>
<td>0.00381**</td>
<td>0.00128</td>
<td>134,571</td>
<td>0.662</td>
</tr>
<tr>
<td></td>
<td>Less-intensive</td>
<td>-0.00278</td>
<td>-0.00162</td>
<td>0.000363</td>
<td>150,143</td>
<td>0.695</td>
</tr>
<tr>
<td>R&amp;D</td>
<td>With expenditure</td>
<td>-0.00176</td>
<td>0.00679**</td>
<td>0.00169**</td>
<td>33,156</td>
<td>0.638</td>
</tr>
<tr>
<td></td>
<td>Without expenditure</td>
<td>-0.00218</td>
<td>0.00286*</td>
<td>0.00127**</td>
<td>332,713</td>
<td>0.712</td>
</tr>
</tbody>
</table>

Notes: TFP (log) is regressed on the hiring variables and confounding factors (log of firm size, firm age, multinational affiliation and legal form), while controlling for industry, year and firm-specific heterogeneity. High education (1, 0) corresponds to educational category 4, see Section 2. Robust and firm-clustered standard errors in parentheses. * p<0.1, ** p<0.05, *** p<0.01.
In Table 2, we present within-firm estimation results for a version of equation (1) that is augmented with the measure of the technology gap of the firm of equation (2). First, we find that the augmentation somewhat magnifies the coefficient of hiring professionals, at least compared with the results for more absorptive firms in terms of education and knowledge intensity (Col 1). Second, although quantitatively very minor, the coefficient of the interaction between the hiring of professionals and the technology gap indicates that hiring professionals has a smaller impact on productivity for laggard firms.

Table 2: Hiring top workers, technology gap, absorptive capacity and firm productivity - within-firm estimation results

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Managers&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>-0.00185</td>
<td>-0.000324</td>
</tr>
<tr>
<td></td>
<td>(0.00352)</td>
<td>(0.00361)</td>
</tr>
<tr>
<td>Professionals&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>0.00595***</td>
<td>0.00567***</td>
</tr>
<tr>
<td></td>
<td>(0.00131)</td>
<td>(0.00139)</td>
</tr>
<tr>
<td>Others&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>0.00224***</td>
<td>0.00222***</td>
</tr>
<tr>
<td></td>
<td>(0.000402)</td>
<td>(0.000456)</td>
</tr>
<tr>
<td>Share Post-Sec&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>0.00112***</td>
<td>0.00119</td>
</tr>
<tr>
<td></td>
<td>(0.000754)</td>
<td>(0.000875)</td>
</tr>
<tr>
<td>Technology gap&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>1.05e-07***</td>
<td>1.01e-07***</td>
</tr>
<tr>
<td></td>
<td>(1.85e-08)</td>
<td>(2.16e-08)</td>
</tr>
<tr>
<td>(Share Post-Sec * Tech-gap)&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>4.18e-09*</td>
<td>4.68-09*</td>
</tr>
<tr>
<td></td>
<td>(1.93e-09)</td>
<td>(1.96e-09)</td>
</tr>
<tr>
<td>(Managers * Tech-gap)&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>1.94e-08</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(2.75e-08)</td>
<td></td>
</tr>
<tr>
<td>(Professionals * Tech-gap)&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>-2.09e-08*</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(1.15e-08)</td>
<td></td>
</tr>
<tr>
<td>(Managers * Share Post-Sec * Tech-gap)&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>2.13e-10</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>(1.96e-09)</td>
</tr>
<tr>
<td>(Professionals * Share Post-Sec * Tech-gap)&lt;sub&gt;t-1&lt;/sub&gt;</td>
<td>1.75e-10*</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>(7.37e-11)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th></th>
<th>Obs.</th>
<th>Adj. $R^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>326,417</td>
<td>0.686</td>
</tr>
</tbody>
</table>

**Notes:** TFP (log) is regressed on the hiring variables, previously included confounding factors (log of firm size, firm age, multinational affiliation and legal form) and educational shares, tech-gap and their interactions, while controlling for industry, year and firm-specific heterogeneity. Robust and firm-clustered standard errors in parentheses. * p<0.1, ** p<0.05, *** p<0.01.

Third, we have also included the absorptive capacity of the firms, in terms of educational level, and its interaction with the technology gap (Col 2). The resulting coefficient for the interaction term that concerns professionals is positive and statistically significant at the 10%-level, suggesting the increasing importance of absorptive capacity for more absorptive firms.

<sup>16</sup>Measured as an indicator (1, 0) for > 50% of the workforce having ≥ 3 years of post-secondary education.
laggard firms’ productivity, but the magnitude of the coefficient is not economically substantial. Fourth, we add a triple interaction to test our conjecture that a higher absorptive capacity enhances the benefit for laggards of hiring professionals. Although economically small, the sign and statistical significance of the triple interaction would seem to suggest to confirm our conjecture. Taking the additional variables and interactions into account, we find the average marginal effect (at means of firm size, age, multinational affiliation and legal form) equivalent to an additional newly hired professional being linked to a 0.07% higher productivity. 17

As an extension, we have analysed the role of the firms that ‘donate’ leading personnel to SMEs. We conjecture that recruits from ‘better’ firms would contribute more strongly to the hiring firm’s productivity than recruits from elsewhere, 18 and that the contribution should be even stronger for firms with higher absorptive capacity. Overall, our within-firm estimation results would seem to confirm the two conjectures. 19 Professionals recruited from ‘better’ firms–enterprise groups, multinational enterprises and foreign-trading firms–are more strongly associated with the subsequent productivity of the hiring SME than are recruits from elsewhere. 20 Typically, the linkage is statistically significant only for hiring SMEs with higher absorptive capacity.

Finally, we have tested the robustness of our results to various sources of biases and inconsistencies. Paying attention to any remaining endogeneity of the hiring variables, we employ the matching and DiD estimator–comparing the productivity growth of like firms, subsequent to hiring leading personnel. The results in Table 3 lead us to confirm a positive and statistically significant productivity impact of hiring leading personnel for firms with higher absorptive capacity. Next, we consider whether the productivity impact is from firing rather than hiring leading personnel. However, controlling for fires of leading personnel in \( t - 1 \) corroborates our main findings, why we conclude that the findings are from an inflow of leading personnel. 21 Lastly, we test the sensitivity of our findings to confounders related to the municipality of the firm and time-variant industry shocks, to non-linearity in the hiring-productivity link, as well as to a dynamic specification, only

17 A result statistically significant at the 10%-level.
18 As in Gidehag and Lodefalk (2016).
19 Results are available in Tables B5–B7 in the online appendix.
20 However, using the R&D-status measure of absorptive capacity, which is based only on those firms that existed in the pre-hiring years of 2001-2002, this difference is not upheld.
21 Table A2 in the online appendix.
to find the results to be qualitatively robust.\textsuperscript{22}

Table 3: DiD matching estimator results (2001-2010)

<table>
<thead>
<tr>
<th>Absorptive capacity</th>
<th>Managers and professionals</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$t + 1$</td>
</tr>
<tr>
<td>High educ.</td>
<td>0.184*</td>
</tr>
<tr>
<td></td>
<td>(0.0752)</td>
</tr>
<tr>
<td>Knowledge-intensity</td>
<td>0.199*</td>
</tr>
<tr>
<td></td>
<td>(0.0862)</td>
</tr>
<tr>
<td>With R&amp;D expenditure</td>
<td>0.164*</td>
</tr>
<tr>
<td></td>
<td>(0.0764)</td>
</tr>
</tbody>
</table>

Notes: Presented are the average treatment effects on the treated from DiD matching estimations, using nearest neighbour matching with the common support assumption. \textit{Dependent Variable}: Total factor productivity (log). At education level (4), the number of treated firms (hiring managers and professionals) is 89, and the number of control firms (hiring other workers) is 66. For knowledge-intensive firms, the number of treated firms is 86, and the number of control firms is 58. For firms with R&D expenditure, the number of treated firms is 66, and the number of control firms is 48. Robust and firm-clustered standard errors in parentheses. Significance levels: * p<0.1, ** p<0.05, *** p<0.01.

4 Concluding remarks

Most firms are small, and small firms are less productive than larger firms. Furthermore, there are indications of smaller firms lagging behind more established firms within their industries, and that laggards may have increasing difficulties to learn from frontier firms (Andrews \textit{et al} 2015). Therefore, it is important to identify drivers for the productivity of SMEs, such as the infusion of external knowledge through the hiring of leading personnel. We exploit very detailed total population employer-employee data for Sweden in the 2001-2010 period to analyse the productivity impact of hiring managers and professionals and, in specific, the role of a firm’s absorptive capacity for such an impact. Employing within-firm as well as matching and difference-in-difference estimators as well as various measures of absorptive capacity, we consistently find that firms with higher absorptive capacity experience a higher productivity impact from hiring professionals than do firms with lower capacity. Moreover, technological laggard firms within an industry benefit more strongly from hiring professionals if they have a relatively well-educated workforce. Our findings may contribute to the explanations behind the increasing productivity gap

\textsuperscript{22}Tables A3–A5 in the online appendix. Additionally, correlation analysis in Table A1 suggests that multicollinearity between the key recruitment variables and confounding factors does not seem to be an issue driving the estimated productivity impact of hiring.
between laggard and frontier firms, with the diffusion of external knowledge mostly benefitting the ‘best’ firms within an industry.

REFERENCES


