HUMAN CAPITAL AND INNOVATION IN DEVELOPING COUNTRIES: A FIRM LEVEL STUDY

Annelies van Uden
a.vanuden@fm.ru.nl

Joris Knoben
j.knoben@fm.ru.nl

Patrick Vermeulen
p.vermeulen@fm.ru.nl

Radboud University Nijmegen
Institute for Management Research
P.O. Box 9108
6500 HK Nijmegen
The Netherlands

---WORKING PAPER JUNE 2014---

This is a working paper from the Co-ordinated Country Case Studies: Innovation and Growth, Raising Productivity in Developing Countries research programme, funded by the UK’s Department for International Development (DFID).

\(^1\) Please send all comments to the first author.
HUMAN CAPITAL AND INNOVATION IN DEVELOPING COUNTRIES: A FIRM LEVEL STUDY

ABSTRACT

In this paper we test whether human capital endowments of firms and additional practices of firms, such as formal training and employee slack time, have a positive relation with the innovative output of firms. This paper contributes to the literature about factors that influence innovation at the firm level and the literature about human capital and innovation at the national level. Furthermore, we study this relation in developing countries, while most studies about innovation have been done in developed countries. We test this relation in Kenya, Tanzania and Uganda with data stemming from the Enterprise Surveys of the World Bank. Our results show that there exists a positive relation between human capital and innovation. In particular, the role of practices of firms such as offering formal training and employee slack time are conducive for innovative output.
INTRODUCTION

Innovation is widely believed to be a key factor for economic growth (Schumpeter, 1934; Solow, 1956), especially in developing countries (Crespi & Zuniga, 2011; Lee & Kang, 2007; Robson, et al., 2009). It is therefore crucially important to understand the determinants of innovation at the country level, but also at the level at which innovations are developed, namely the firm. Studies at the national level highlight, among others, human capital as a driving force for innovation (e.g. Dakhli & De Clercq, 2004). At the firm level, however, this determinant of innovation has received scarce attention (Schneider et al., 2010). Instead, most firm level studies focused on the role of R&D activities, technology acquisition, firm size and age as determinants of innovation (Hirsch-Kreinsen et al., 2005; Shefer & Frenkel, 2005). The few studies that have considered the role of human capital as a key factor fostering innovation at the firm level have all taken place in developed countries (e.g. Grimpe & Sofka, 2009; Liu & Buck, 2007). Yet, there is a striking dearth of such studies in developing countries. This is surprising because stimulating education levels and building human capital is the cornerstone of many development initiatives and policies in developing countries (UNCTAD, 2014). As such, firm level studies in developing countries regarding the relation between human capital and innovation are much needed.

First, a firm level approach is needed because it is not necessarily possible to generalize macro level evidence to the micro level, due to the risk of ecological fallacies (Beugelsdijk, 2007; Robinson, 1950). Studies on human capital in developing countries often focus on the impact on economic growth, the growth of total factor productivity (Benhabib & Spiegel, 1994; Goedhuys, Janz, & Mohnen, 2008) or foreign direct investment (Noorbakhsh et al., 2001). However, these studies are mostly at the country level and shed little light on the relationship between human capital and innovation at the firm level. Firm level studies that have looked at this relationship almost exclusively focus on firms in developed economies. For instance, Leiponen (2005) and Vinding (2006) examine the impact of education of employees on innovation and find a positive relation (in Finland and Denmark respectively). Schneider et al., (2010) consider the relation between innovation and the educational level of employees and training provided by firms in Germany and found a positive relation as well.
Second, studies in developing countries are needed, because an increasing body of literature provides evidence that innovation and productivity are key factors to grow out of poverty (Crespi & Zuniga, 2011; Hegde & Shapira, 2007; Lee & Kang, 2007; Robson et al., 2009). However, firms in developing countries operate “substantially below the technology frontier, with lower levels of human capital” (Goedhuys et al., 2008), which means that those countries can catch up by technology acquisition and imitation (Bell & Pavitt, 1993; Katz, 1986). Therefore, a focus solely on R&D investments is unlikely to tell the whole story in developing economies (Crespi & Zuniga, 2011). The lack of formal R&D expenditures and technology investments in most developing countries can partly be explained by the poor supply of human capital (Chaminade & Vang, 2006). Yet, the role of human capital in innovation has received limited attention in developing countries. Some studies take the human capital endowments of the firm into account, but their measures of human capital are often restricted to formal schooling (see for instance Robson et al., 2009). We argue that not only the human capital endowments of a firm (such as the level of education of employees) play a role, but that a firm can also invest in human capital by offering formal training or by providing slack time to employees. These firm-level practices could, next to the level of schooling, increase the level of human capital within the firm and influence the innovative output of that firm. Hence, this study extends the above-mentioned studies by taking into account other factors that could improve human capital within the firm and by applying these arguments at the firm level in developing countries.

We investigate the relationship between human capital and innovation in three developing countries in the East African region: Kenya, Tanzania and Uganda. We use the Enterprise Surveys of the World Bank, which are harmonized questionnaires conducted in the manufacturing and services sectors in several developing countries. The latest version (from 2013) also consists of data about the innovative output of firms and human capital. The results indicate that human capital is an important factor for innovation. Firm-specific practices in particular have a positive effect on the innovative output of firms in developing countries.

Our study has two significant contributions. First, we contribute to the literature about human capital and innovation by an in-depth investigation of the different aspects of human capital within a firm. We show how these distinct aspects of human capital are conducive for innovation. We do not
only examine the human capital endowments, but also the practices of firms to improve the level of human capital and the interaction between certain aspects. Second, we contribute to the literature by analyzing the role of human capital in firms in developing countries.

THEORY AND HYPOTHESES

Human capital and innovation

Human capital refers to the skills, abilities and knowledge of individuals (Becker, 1964). In general, it has been argued that human capital is an important source of innovative activities that may result in a competitive advantages for firms and nations (Coleman, 1988; Gimeno et al., 1997).

In this study, we specifically focus on the role of human capital for innovation at the firm level, because it is conducive for the development of new knowledge (Smith et al., 2005) and it supports the ability of firms to absorb knowledge (Cohen & Levinthal, 1990). This knowledge can be accumulated by R&D conducted within the firm or can derive from the skills and abilities of employees (Zahra & George, 2002), which we refer to as human capital. Notwithstanding the role of R&D, we specifically consider the role of human capital for innovation. Previous studies already showed the positive effect of R&D on innovation in developed economies (e.g. Amara et al., 2008; Raymond & St-Pierre, 2010) and in developing countries (e.g. Goedhuys, 2007; Shefer & Frenkel, 2005), but the role of human capital in innovation is studied less explicitly and few empirical studies focus on the relation between human capital and innovative output (Schneider et al., 2010). The relation between human capital and innovation has mainly been studied at the national level (e.g. Dakhli & De Clercq, 2004) or, it is used as a control variable in studies at the firm level (e.g. Goedhuys & Veugelers, 2012).

Yet, Schneider et al. (2010) argue that a firm level approach is needed to better understand the relation between human capital and innovation. Their empirical results indeed show that better trained employees will provide firms with more innovative output. Similarly, in the context of developing countries, Mahemba & Bruijn (2003) and Robson et al. (2009), showed the importance of training for the innovative performance of firms. However, most studies using human capital focus predominantly on formal education or years of schooling as a measure of human capital (Marvel & Lumpkin, 2007).
We aim for a more sophisticated measure of human capital and argue that human capital within a firm does not only consists of the level of education or schooling, but also consists of firm level practices that are geared towards the development of human capital within the firm. We consider the role of practices like formal training and providing employees with slack time as ways to improve the level of human capital within the firm. As such, these practices positively affect the innovative performance of these firms.

**Employee schooling**

Employee schooling refers to the level of schooling that employees of the firm possess. The level of schooling of employees within a firm may benefit the firm, because education enhances the ability to understand, create and process information quicker compared to individuals without education (Nelson & Phelps, 1966). Furthermore, a workforce that has a certain level of schooling is better able to absorb knowledge and exploit opportunities compared to a workforce without any schooling (cf. Cohen & Levinthal, 1990). Previous research in the agricultural sector in developing countries shows a positive relation between education and innovation (see for instance (Knight et al., 2003). Liu & Buck (2007) included the level of schooling to explain innovative output in China and found a significant effect as well. Hence, the level of schooling within a firm could be conducive in absorbing knowledge and transforming this knowledge into innovation. Therefore, we propose the following hypothesis:

H1: The higher the share of employees with at least secondary schooling within a firm, the higher its probability to produce innovative output.

**Formal training**

We consider formal training as another crucial element of human capital within a firm. Formal training is the extra training that employees receive from the firm in order to instruct employees in specific processes and learn them specific skills and abilities. This increases the human capital of the firm by updating knowledge and capabilities of employees (Cohen & Levinthal, 1990). Furthermore, if a firm would like to introduce a new product or service, the employees should learn new skills in order
to produce these new products and services. Therefore, training of employees enhances the success of innovation (Freel, 2005). Especially in low- and medium-technology industries training seems to be crucial for innovation (Santamaría et al., 2009). Yet, empirical evidence is not conclusive about the role of formal training. For instance, Caloghirou et al., (2004) do not find a significant effect of formal training in developed countries, while Santamaría et al. (2009) find a significant effect of training on innovation in Spain. Goedhuys (2007) does not find a significant relation between training and product innovation in Tanzania. However, we expect that the relation will be positive and significant in developing countries, because formal training of employees can compensate for the lower degree of education of employees (Goedhuys & Srholec, 2010). Hence, we propose the following hypothesis.

**H2:** A firm that provides formal training to its employees has a higher probability to produce innovative output compared to firms that do not provide formal training.

**Employee slack time**

Employee slack time refers to human resources that are not necessary for the daily operations of a firm (Bourgeois, 1981). This employee slack is the time that employees can spend on other explorative activities instead of their daily activities, because it is the excess time that employees have. The effect of slack resources in general on innovation is still a point of discussion (Anderson et al., 2014). Some find a inverted U-shape relation of slack resources on innovation (Herold et al., 2006; Nohria & Gulati, 1996), while others find an negative relation (Latham & Braun, 2009) or do not find a significant effect (Alpkam et al., 2010; Mousa & Chowdhury, 2014).

We focus on employee slack time in particular, because it is an element of building human capital within a firm; it gives employees the opportunity to develop themselves and obtain new skills and knowledge. Positive effects show that slack time gives employees the opportunity to explore new domains and activities, which might result in new ideas about innovative products and services (Nohria & Gulati, 1996). During this time, employees have the opportunity to be creative, which can lead to innovation (Amabile, 1996). Hence, the basic idea of the positive relation is that employee slack time gives employees the opportunity to develop their own ideas during working hours, which
may eventually result in innovative output. Negative effects show that slack time is a waste of resources (Williamson, 1963) and may interrupt certain processes (Mosakowski, 2002), which could result in a negative impact on innovation.

There is some anecdotal evidence about Google, where employees spend 10% of their working time on a self-chosen project, which shows a positive relation between slack time and innovation. Another company that implements slack time for its employees is 3M. Employees have to spend 15% of their time on their own projects or ideas. Garud et al. (2011) identify this 15% rule as one of the key characteristics of the company that fosters innovative output.

Employee slack time enhances the creativity of employees. It has been argued that especially in low- and medium technology industry, creativity is one of the drivers of innovation instead of technological knowledge (Santamaría et al., 2009). Most developing countries have a comparative advantage in low- and medium technology industries (Goedhuys et al., 2014), which indicates that creativity is an even more important factor in those countries. Hence, we expect the following relation:

\[ H_3: \text{A firm that gives slack time to its employees has a higher probability to produce innovative output compared to firms that do not give slack time to their employees.} \]

**Formal training and employee slack**

There is reason to believe that a combination of human capital development practices in firms could influence the result of those practices. In most studies there is an implicit assumption that if factors spur innovation, the higher the level of that factor, the higher the innovative output (Anderson et al., 2014). However, it could be that combinations of factors have less favorable results and do not reinforce each other. Therefore, more research is needed to indicate how different variables interact with each other (Anderson et al., 2014).

We argue that the combination of formal training and employee slack time may have counterproductive effects. Employees that receive formal training are provided with information and guidelines how they have to proceed in their work. We argued that this training helps employees to successful implement innovations. However, if employees receive formal training they will be trained
in terms of procedures and guidelines, which could diminish their creativity. During training sessions, codified knowledge will be transmitted to employees in the formalized structure of the training. These formalized structures can result in more rigidities which can limit creativity (Klijn & Tomic, 2010). The slack time that employees receive to develop themselves and come up with innovative ideas is based on the assumption that some spare time will enhance creative thinking and result in new ideas (Amabile, 1996). Therefore, we expect that the effect of employee slack time on innovation diminishes if the firm provides formal training to its employees as well. Thus, the combination of formal training and employee slack time can result in less favorable results. This results in the following hypothesis:

H4: Employee slack time in combination with formal training within a firm diminishes each other’s effect on the probability to produce innovative output.

Employee schooling and employee slack
The education of employees provides a certain level of basic knowledge for individuals. It has been argued that a certain level of knowledge can contribute to creativity (Amabile, 1996; Ford, 1996). Education gives individuals the opportunity to receive a certain kind of general knowledge, while formal training provides employees with specific guidelines and routines. Hence, we expect that the combination of employee schooling and employee slack time has a different effect than the combination of formal training and employee slack time.

Employee slack time gives employees the opportunity to develop their skills and come up with innovative ideas. It seems therefore reasonable to argue that employees that already have a certain level of knowledge will come up with more creative ideas during their slack time than individuals without this level of knowledge. While the knowledge component and the time available for creativity has been widely recognized theoretically, empirical studies on the interaction between the knowledge base of employees and time related to creativity remains scarce (Anderson et al., 2014). We expect that employee slack time and formal education will reinforce each other’s positive effect on innovation. This results in the following hypothesis:
H5: Employee slack time in combination with the amount of employee schooling within a firm strengthens each other’s effect on the probability to produce innovative output.

DATA AND METHODS

Data

The data that we use to test our theoretical ideas stems from the Enterprise Surveys 2013 conducted in Kenya, Tanzania and Uganda. The Enterprise Surveys have been developed by the World Bank to collect data that can be harmonized among developing countries. Since 2002, the World Bank has conducted interviews with top managers and business owners of 130,000 firms in 135 economies.

Ideally, the population consists of all eligible firms within the country, based on the population that is registered in their statistical offices. Sometimes, other lists have been used to come up with the population. The World Bank uses stratified random sampling as sampling methodology. The strata that have been used are firm size, business sector and geographic region within a country. This resulted in a representative sample for the countries and industries involved. In total 2076 firms have been surveyed in our sample, 713 from Kenya, 723 from Tanzania and 640 located in Uganda.

Dependent Variable

To operationalize a firm’s innovation outcomes, we used self-reported measures of innovativeness that were developed for the Community Innovation Survey (CIS) (Brouwer & Kleinknecht, 1996). Specifically, to measure whether companies are innovative we utilized two sequential questions. First, respondents were asked “Did you introduce new or significantly improved products or services to the market in the last three years?”. A three-year period was chosen to avoid bias resulting from measuring accidental or one-off innovation. Respondents answering in the affirmative to this question were subsequently asked “Was this new or significantly improved product or service also new to your main market?”. Companies answering ‘yes’ on both questions were coded with a ‘1’ all other companies with a ‘0’. This measurement is in line with generally accepted definitions of incremental and radical innovation and prior research has shown that this perception based measure of innovation

2 For more information about the methodology and sampling see www.enterprisesurveys.org.
outcomes is highly reliable and correlates heavily with other (objective) measures of innovation outcomes (Hagedoorn & Cloodt, 2003).

Independent Variables

**Employee Schooling.** The level of education of the employees was measured by asking the respondents “What percentage of your full-time workers has completed their high school?”. The resulting variable ranges between 0 and 100 by design.

**Formal training.** The presence of formal training practices within the company was assessed by asking “In the last fiscal year did your company offer formal training programs to your full-time permanent employees?”. Companies answering ‘yes’ to this question were coded with a ‘1’ all other companies with a ‘0’.

**Employee Slack.** The presence of the practice of giving employees slack time to work on creative new ideas was measured by asking “During the last three years, did your establishment give employees time to work on new ideas?”. Companies answering ‘yes’ to this question were coded with a ‘1’ all other companies with a ‘0’.

Control Variables

**Size.** We control for the size of the company as generally bigger companies have more resources at their disposal can more easily free up personnel and resources for innovative activities (Hansen, 1992). The size of the company was measured by the natural log of the number of full-time permanent employees of the company.

**Age.** We control for the age of the company as it is often argued that older companies are more inert and less flexible and will therefore be less likely to innovate (Hansen, 1992). A company’s age was determined by asking for the year of establishment of the company and subtracting this from the year in which the survey was performed (i.e. 2013).

**Subsidiary.** We also control for whether the company as an independent economic unit or part of a larger organizational entity. We do so by asking the question “Is your establishment part of a
larger firm?”. Companies answering ‘yes’ to this question were coded with a ‘1’ all other companies with a ‘0’.

**R&D.** A firm’s internal capacity to generate and process knowledge is also likely to impact on its innovation outcomes (Cohen & Levinthal, 1989). As such, we included a dummy variable that took the value ‘1’ if the responding firm indicated that in the last three year it had spend any money on formal R&D activities and ‘0’ in all other cases.

**Foreign presence or foreign owned.** We used a question about the percentage of the company that is owned by private foreign individuals, companies or organizations to construct two control variables. First, for any company that answered any value greater than 0% to the above question we coded the control variable ‘foreign presence’ as ‘1’ and ‘0’ otherwise. Second, for any company that answered any value greater than 50% we coded the control variable ‘foreign owned’ as ‘1’ and ‘0’ otherwise. We control for foreign presence and foreign ownership because firms in emerging economies often highly benefit from technological knowledge available from their international headquarter and research labs (Isobe et al., 2000).

**Country and industry dummies.** Finally, we include dummy variables to control for differences between countries (Uganda being the reference category) and between industries (services being the reference category).

**Analyses**

Our dependent variable has a discrete distribution. We therefore employed logistic regression analysis to estimate the effects of our independent variables on the likelihood of a firm being innovative. The basic form of a logistic regression equation is represented in equation 1. To make this function estimable it is transformed into equation 2.

\[
Y = \frac{e^{b_0 + b_1 X_1 + b_n X_n + \varepsilon}}{1 + e^{b_0 + b_1 X_1 + b_n X_n + \varepsilon}}
\]  

(1)

\[
log \left[ \frac{Y}{1 - Y} \right] = b_0 + b_1 X_1 + b_n X_n + \varepsilon
\]

(2)
As is evident from equation 1 and 2, logistic models are highly non-linear. Therefore, formal hypothesis tests using logistic regression models have to take into account that the strength and direction of effects depend on the values of all other variables in the model. We follow recommendations by (Hoetker, 2007) and estimated average marginal effects across all observed values for the other variables in the model. This approach improves on the common practice of setting all other variables at their mean. The latter can be problematic because the mean ignores the actual dispersion of values. In addition, in the case of categorical variables, the mean tends to be a value for which the variable is not defined.

We used improvement of overall model fit to identify appropriate models for hypothesis tests based on log-likelihood ratio tests (Long & Freese, 2006). For our formal hypothesis tests, we report conditional effect specific relevant values of the independent variables (Bowen & Wiersema, 2004; Long & Freese, 2006). In addition, we provide graphs that show their effects across the full observed range of variable values.

RESULTS

Table 1 reports pooled descriptive statistics and correlations. The descriptive statistics indicate that 41% of the firms in our sample report to be innovative. Only 23% of the firms performed any formal R&D in the last three years. However, a surprisingly large share of 44% of the firms formally offers their employees time to work on new and creative ideas. As such, it seems like a large part of the R&D is done informally. Finally 34% of the companies offered formal training to their personnel and for the average firm about 53% of the personnel at least holds a high school degree.

Table 2 reports the results of the binary logistic regressions performed to test our hypotheses. Models 1 through 3 are hierarchical logistic regression. Model 1 is the baseline model including only the control variables. Model 2 adds to direct effects of the three independent variables to the model, whereas model three adds to interaction effects between employee slack and the two other independent
variables. Models 4-5 and 6-7 are identical to model 2 and 3 except that models 4 and 5 include only manufacturing firms, whereas models 6 and 7 include only service firms.

Insert Table 2 here

With regard to the control variables most results are as expected. Firms size has a positive effect on the likelihood of being innovative, whereas firm age has a negative effect. R&D has the expected positive effect, but the size of the effect is surprisingly small. Marginal effect analyses reveal that the difference in the likelihood of being innovative between firms that do perform R&D and firms that do not is only 7.2%. Compared to the effect sizes of some of the human capital variables we will discuss later, this effect is modest indeed. This further underlines the notion that formal R&D is relatively unimportant as a driver of innovation in developing countries.

The main effects of our independent variables are highly similar across all models. With regard to the interaction effects, comparing model 3 to 2 reveals that the full model (model 3) has a superior model fit. However, comparing models 4-5 and 6-7 reveals that this superior model fit is completely driven by a better model fit for the manufacturing firms. For the service firms, the interaction effects are insignificant. As such, we will be interpreting the interaction effects separately for both industries.

Employee schooling has a marginally significant effect on a firm’s likelihood of being innovative. Moreover, the effect is very small. Marginal effects analyses reveal that a one standard deviation increase in employee schooling increases the likelihood of being innovative by about 1%-point. So even though we find some statistical support for our hypothesis 1 we conclude that employee schooling is a relatively unimportant determinant of firm innovation.

We find very strong support for our hypothesis that formal training is of influence on a firm’s innovativeness (hypothesis 2). The size of the effect of this variable is depicted in Figure 1. Having a formal training program makes the likelihood of a firm being innovative increase from 23% to 47%, which is more than a doubling.
The same conclusion holds for firms that offer their employees slack time to work on new and creative ideas. Figure reveals that the size of this effect is even more profound than that of formal training. Offering employees slack time results in an increase in the likelihood of being innovative from 23% to 54%. These findings offer strong support for hypothesis 3.

In hypothesis 4 we predicted that offering both formal training and employee slack would be counterproductive as formal training might reduce individual creativity which employee slack relies on. We find some support for this hypothesis but only in the manufacturing industry. The effect size analysis reported in Figure 1 clearly reveals the differences between the manufacturing and the service industries in this regard. Individually, the effects of formal training and employee slack are bigger in the manufacturing industry than in the service industry. The effect of using both formal training and employee slack, however, is bigger in the service industry. It is important to note, however, that even in the manufacturing industry firms offering both formal training and employee slack are better off than firms offering only one of the two. So even though the two diminish each other’s effect in the manufacturing industry it is not the case that there is a formal trade-off between the two practices.

In hypothesis 5 we predicted that having more educated employees and offering employee slack would be reinforce each other’s positive effects. We find evidence of the opposite, but only in the manufacturing industry. Effect size analyses reported in Figure 2 reveal that for firms that offer employee slack the effect of employee schooling actually turns negative. This is an intriguing finding to which we will get back in detail in the discussion section. However, it is important to note that, for any value of employee schooling, offering employee slack will increase a firm’s likelihood of being innovative (i.e. the black line is always above the grey line in Figure 2). However, for firms that already offer employee slack, a strategy of hiring more educated employees might have negative consequences for the innovativeness.

DISCUSSION

In general, our results support the relation between human capital and innovation. The findings indicate that all direct effects of our dependent variables have a positive relation with innovation. However, our study also revealed a few surprising results. For instance, the effect of the level of
education on innovation is less strong compared to the effects of formal training and employee slack
time on innovation. This implies that firm-specific practices are very conducive for innovation in
developing countries. Furthermore, the interaction between secondary schooling and employee slack
time has a different effect than we anticipated. We briefly review our main findings and provide
suggestions for further research.

The positive relation between secondary education and innovation is in line with Knight et
al. ’s (2009) also introduced higher levels of education, which showed an even more significant effect
on innovation. Our results indicate that schooling has a positive relation with innovation in developing
countries, but that firm-specific practices (providing employee slack time and formal training) have a
more profound effect. Although we did not have data about higher levels of education, we would
expect the effect between schooling and innovation to be stronger if higher levels of education are
introduced. This is a limitation of our study and could be interesting for future research.

Our results show a strong positive effect of formal training on innovation. Previous studies
were not conclusive about the effect of formal training on innovation. Free (2005) and Santamaría, et
al., (2009), for instance, found a positive effect, while Caloghirou et al., (2004) did not find an effect
at all. Our study adds to the limited amount of studies about the positive role of training for innovation
(Santamaría et al., 2009). One explanation could be that, especially in developing countries, formal
training supports innovation and compensates for lower levels of education of employees. This is in
line with previous studies (e.g. Goedhuys and Srholec, 2010) that argued that formal training is a
supplement to the lower degree of education in developing countries.

Our results show an even stronger effect of employee slack time on innovation (compared to
formal training and schooling), which implies that employee slack time is pivotal for innovative
output, also in developing countries. The suggestion that slack time results in opportunities to explore
new domains and activities which can result in new ideas about innovation (Nohria & Gulati, 1996)
and that, theoretically, time supports creativity thinking (Shalley & Gilson, 2004) has been considered
crucial for innovation (Amabile, 1996). However, previous studies that empirically examined the role
of slack resources for innovation in developed countries showed contradicting results (Anderson et al.,
2014). This study shows that employee slack time has a positive relation with innovation in developing countries (Greve, 2003).

Summarizing, the direct effect of all variables indicate a positive relation between human capital and innovation. The limited effect of educational level compared to the effect of formal training and employee slack could indicate that these firm-specific practices are complements to the low human capital endowments of the employees of the firm.

We also hypothesized two interaction effects. We found that the combination of formal training and employee slack diminish each other’s effect in the manufacturing industry, which could mean that the creativity related to employee slack time is diminished by formal training. The combination of these two practices has then less favorable results, because formal training may enforce employees with certain guidelines that are not conducive for creativity. Yet, in both manufacturing and services, providing both practices to employees increases the likelihood of innovation. Until now, there was a lack of studies examining the interaction effect between these variables (Anderson et al., 2014). Our results give some insight in the interaction between formal training and employee slack time provided by firms.

The combination of employee slack time and secondary education gives overall a higher innovative output. However, the higher the percentage of employees with secondary schooling, the lower the effect of employee slack time on innovation, which contradicts our expectation that secondary schooling and employee slack time would reinforce each other. This could imply that employees that did not attend secondary education have other characteristics that are even more beneficial in combination with employee slack time. We argued that education results in a basic level of knowledge that is conducive for creativity. However, another way in which employees could acquire knowledge is by job experience (Tierney & Farmer, 2002). It could be that employees without secondary education have spend more time working and therefore have more job experience gaining more knowledge related to the job. Being creative is also related to expertise (Amabile, 1988), which could indicate that employees that did not attend secondary school have more job experience and therefore more expertise within a certain domain. This enhances their creativity and therefore the effect of slack time could be higher for employees without secondary schooling. Further research can
investigate this line of reasoning if data is available about the expertise of employees within the firm compared to the level of secondary education within the firm. This result reveals that the combination of employee slack time and formal training diminish each other’s effect in the manufacturing industry, which adds to the literature about the interaction between multiple factors that are conducive in innovation (Anderson et al., 2014).

The foregoing discussion shows that human capital is of significant importance in developing countries. In particular the role firms play in improving the level of human capital within the firm by specific practices such as formal training and providing employee slack time. This implies that especially in developing countries, firms can enhance their innovative output by giving employees the ability to develop themselves by formal training or employee slack time.

Policy Implications

Our results reveal that firm level practices, such as formal training and employee slack time, have a more profound impact on innovation than traditional factors like schooling and R&D. This indicates that policymakers, who would like to promote innovation at the firm level, should stimulate investments in formal training and employee slack time. Policymakers could introduce tax advantages or subsidies that are favorable for firms that introduce these practices. For example, in Kenya there are policies that encourage in-house R&D activities of firms (Technopolis Group, 2014). Our study points out that policies directed towards encouraging formal training or slack time may be more beneficial for innovative output than policies focusing on R&D expenditures.

The results of our study have managerial implications as well. Managers, who wish to improve innovative output, seem to be better off when introducing formal training or employee slack time, than by increasing R&D expenditures. For future studies, it could be an interesting approach to investigate the causality between human capital and innovation in developing countries. Due to data limitations it was not possible in our study to conduct analyses using panel data. Such an analysis would give an indication about the causality between human capital and innovation.
REFERENCES


### Table 1
Descriptive Statistics and Bivariate Correlations

<table>
<thead>
<tr>
<th></th>
<th>Mean</th>
<th>St. Dev.</th>
<th>Min</th>
<th>Max</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>9</th>
<th>10</th>
<th>11</th>
<th>12</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Innovation</td>
<td>0.41</td>
<td>0.49</td>
<td>0</td>
<td>1</td>
<td>-</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2</td>
<td>Size (ln)</td>
<td>2.58</td>
<td>1.41</td>
<td>0</td>
<td>8.61</td>
<td>0.15</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>3</td>
<td>Age (ln)</td>
<td>2.79</td>
<td>1.33</td>
<td>0</td>
<td>7.61</td>
<td>-0.02</td>
<td>0.08</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>4</td>
<td>Subsidiary</td>
<td>0.14</td>
<td>0.34</td>
<td>0</td>
<td>1</td>
<td>0.03</td>
<td>0.15</td>
<td>0.08</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>5</td>
<td>Foreign Presence</td>
<td>0.10</td>
<td>0.30</td>
<td>0</td>
<td>1</td>
<td>0.05</td>
<td>0.20</td>
<td>0.05</td>
<td>0.14</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>6</td>
<td>Foreign Owned</td>
<td>0.06</td>
<td>0.24</td>
<td>0</td>
<td>1</td>
<td>0.07</td>
<td>0.15</td>
<td>0.01</td>
<td>0.12</td>
<td>0.78</td>
<td>-</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>7</td>
<td>R&amp;D</td>
<td>0.23</td>
<td>0.42</td>
<td>0</td>
<td>1</td>
<td>0.23</td>
<td>0.25</td>
<td>0.08</td>
<td>0.16</td>
<td>0.14</td>
<td>0.10</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>8</td>
<td>Tanzania</td>
<td>0.35</td>
<td>0.48</td>
<td>0</td>
<td>1</td>
<td>-0.10</td>
<td>-0.21</td>
<td>0.01</td>
<td>-0.11</td>
<td>-0.12</td>
<td>-0.12</td>
<td>-0.20</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>9</td>
<td>Kenya</td>
<td>0.34</td>
<td>0.47</td>
<td>0</td>
<td>1</td>
<td>0.01</td>
<td>0.27</td>
<td>0.06</td>
<td>0.15</td>
<td>0.03</td>
<td>-0.01</td>
<td>0.13</td>
<td>-0.53</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>10</td>
<td>Manufacturing</td>
<td>0.50</td>
<td>0.50</td>
<td>0</td>
<td>1</td>
<td>0.04</td>
<td>0.19</td>
<td>0.10</td>
<td>-0.03</td>
<td>0.08</td>
<td>0.05</td>
<td>0.07</td>
<td>0.01</td>
<td>0.02</td>
<td></td>
<td>-</td>
</tr>
<tr>
<td>11</td>
<td>Employee Schooling</td>
<td>53.83</td>
<td>39.12</td>
<td>0</td>
<td>100</td>
<td>0.08</td>
<td>0.28</td>
<td>-0.01</td>
<td>0.05</td>
<td>0.04</td>
<td>0.02</td>
<td>0.15</td>
<td>-0.30</td>
<td>0.42</td>
<td>-0.06</td>
<td>-</td>
</tr>
<tr>
<td>12</td>
<td>Formal Training</td>
<td>0.34</td>
<td>0.47</td>
<td>0</td>
<td>1</td>
<td>0.30</td>
<td>0.24</td>
<td>0.03</td>
<td>0.13</td>
<td>0.10</td>
<td>0.05</td>
<td>0.29</td>
<td>-0.08</td>
<td>0.14</td>
<td>0.02</td>
<td>0.15</td>
</tr>
<tr>
<td>13</td>
<td>Employee Slack</td>
<td>0.44</td>
<td>0.50</td>
<td>0</td>
<td>1</td>
<td>0.37</td>
<td>0.21</td>
<td>0.04</td>
<td>0.07</td>
<td>0.05</td>
<td>0.05</td>
<td>0.39</td>
<td>-0.11</td>
<td>0.12</td>
<td>0.10</td>
<td>0.13</td>
</tr>
</tbody>
</table>

Bivariate correlations
## Table 2
Logistic Regression Results

<table>
<thead>
<tr>
<th>Variables</th>
<th>All firms</th>
<th>Model 2</th>
<th>Model 3</th>
<th>Model 4</th>
<th>Model 5</th>
<th>Model 6</th>
<th>Model 7</th>
</tr>
</thead>
<tbody>
<tr>
<td>Control variables</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Size (ln)</td>
<td>0.211***</td>
<td>(0.036)</td>
<td>0.083*</td>
<td>0.081*</td>
<td>0.121*</td>
<td>0.126*</td>
<td>0.047</td>
</tr>
<tr>
<td>Age (ln)</td>
<td>-0.053</td>
<td>(0.036)</td>
<td>-0.078*</td>
<td>-0.077*</td>
<td>-0.001</td>
<td>-0.009</td>
<td>-0.121</td>
</tr>
<tr>
<td>Subsidiary</td>
<td>0.055</td>
<td>(0.136)</td>
<td>-0.149</td>
<td>-0.145</td>
<td>0.000</td>
<td>-0.003</td>
<td>-0.263</td>
</tr>
<tr>
<td>Foreign Presence</td>
<td>-0.197</td>
<td>(0.245)</td>
<td>-0.340</td>
<td>-0.343</td>
<td>-0.261</td>
<td>-0.332</td>
<td>-0.566</td>
</tr>
<tr>
<td>Foreign Owned</td>
<td>0.434</td>
<td>(0.301)</td>
<td>0.554†</td>
<td>0.551†</td>
<td>0.420</td>
<td>0.452</td>
<td>0.817</td>
</tr>
<tr>
<td>R&amp;D</td>
<td>0.336***</td>
<td>(0.129)</td>
<td>0.335***</td>
<td>0.356***</td>
<td>0.256</td>
<td>0.289†</td>
<td>0.441*</td>
</tr>
<tr>
<td>Tanzania</td>
<td>-0.465**</td>
<td>(0.112)</td>
<td>-0.442***</td>
<td>-0.446***</td>
<td>-0.509***</td>
<td>-0.511***</td>
<td>-0.387*</td>
</tr>
<tr>
<td>Kenya</td>
<td>-0.363***</td>
<td>(0.111)</td>
<td>-0.576***</td>
<td>-0.582***</td>
<td>-0.780***</td>
<td>-0.792***</td>
<td>-0.424*</td>
</tr>
<tr>
<td>Manufacturing</td>
<td>0.096</td>
<td>(0.093)</td>
<td>0.046</td>
<td>0.043</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Independent variables</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Employee Schooling (H1)</td>
<td>0.001</td>
<td>(0.001)</td>
<td>0.004*</td>
<td>0.001</td>
<td>0.005*</td>
<td>0.001</td>
<td>0.003</td>
</tr>
<tr>
<td>Formal Training (H2)</td>
<td>0.817***</td>
<td>(0.111)</td>
<td>0.878***</td>
<td>0.767***</td>
<td>1.099***</td>
<td>0.845***</td>
<td>0.652***</td>
</tr>
<tr>
<td>Employee Slack (H3)</td>
<td>1.203***</td>
<td>(0.108)</td>
<td>1.582***</td>
<td>1.203***</td>
<td>1.797***</td>
<td>1.215***</td>
<td>1.377***</td>
</tr>
<tr>
<td>Formal Training * Employee Slack (H4)</td>
<td>-0.101</td>
<td>(0.220)</td>
<td>-0.542†</td>
<td>(0.306)</td>
<td>0.356</td>
<td>0.324</td>
<td></td>
</tr>
<tr>
<td>Employee Schooling * Employee Slack (H5)</td>
<td>-0.006**</td>
<td>(0.002)</td>
<td>-0.008**</td>
<td>(0.003)</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

| N                        | 2076      | 2076    | 2076    | 1042    | 1042    | 1034    | 1034    |
| Log-likelihood           | -1372.31  | -1210.05| -1206.89| -614.45 | -610.23 | -592.68 | -591.38 |
| $\chi^2$/Log-likelihood | -         | 324.52***| 6.30*   | -       | 8.44**  | -       | 2.60    |

a: Robust standard errors in parentheses

† p < .10
* p < .05
** p < .01
*** p < .001
Figure 2
Schooling, Employee Slack and Innovation for Manufacturing Firms

Pr(Innovation = 1) vs % of employees with secondary schooling

- Predicted effect without Employee Slack
- Predicted effect with Employee Slack
- Lower bound 95% CI
- Upper bound 95% CI