EMPLOYER SURVEY

Snapshot—2016

Highlights from Six Low- and Middle-Income Countries
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World Bank regional teams were led by Meskerem Mulatu (Armenia, Azerbaijan and Georgia), Johannes Koettl and Indhira Santos (Macedonia), Halil Dundar (Sri Lanka), and Xiaoyan Liang (Yunnan Province, China).

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Measuring Skills for Employment and Productivity

How does skills mismatch affect employers in the modern economy?

This snapshot features a descriptive analysis of the data collected during the STEP Employer Survey. The issues highlighted here can be further investigated by researchers and policy-makers interested in using survey tools to discuss skills availability and needs in the labor and education policy contexts. Patterns are provided from the six countries where the survey was deployed—namely, Armenia, Azerbaijan, Georgia, Sri Lanka, Vietnam and China (Yunnan Province only). Detail on each individual country is available in reports featured online. The following sections will discuss how employers view jobs mismatch, identify skills employers’ value in their workers, provide insights into innovative employers and, finally, illustrate training provision by employers.

Economic growth depends on skills being put to productive use. In recent years, research on labor outcomes and education has converged on the unanimous conclusion that there is a substantial mismatch between the supply of and demand for skills around the world (Cappelli, 2014; McIntosh and Vignoles, 2001). This mismatch affects more than just wages or individual job satisfaction. Skill mismatches have an impact on productivity and growth at both the firm level and the macro-economic levels (Quintini, 2014). Reports show that firms around the world consider skill challenges to be an impediment to the operation and development of their business. A cause for further concern is that this constraint seems to be disproportionately affecting the more dynamic and innovative employers, signifying a potentially negative impact on job creation and technological progress (World Bank, 2012).

It is essential to provide employers with a voice in the skills mismatch debate. The debate over the “skills gap” (a deficiency in basic skills inherent in the flow of workers entering the workforce), the “skills shortage” (a lack of technical skills in the stock of workers available for specific occupations) and the “skills mismatch” (a case of under-supply or over-supply of qualified workers) has thus far been primarily focused on improving individual well-being (Cappelli, 2014). However, in order to make the most of investing in the skill set of a workforce, there is a need to understand how these skills will be put to use in the current labor market (Quintini, 2014). The potential contribution of skills to improving productivity and employment outcomes needs to be fully exploited, and this depends largely on understanding the ‘demand’ side of skills. Most attempts to measure this disparity have been focused on measuring the level of ‘education’ of workers in various occupations (Quintini, 2014) or comparing earnings across these occupations (Potter, 2002). However, as the focus shifts to the skills-shortage, there is a growing need to separately identify the role of ‘skills’ in this mismatch and, especially, to collect evidence on the labor market value of any particular skills-profile (McIntosh and Vignoles, 2001).

Identifying the most relevant competencies and employable skills will help inform labor force and education policy. Current data measuring skill demands are largely posited on the ‘stated preferences’ of the employers surveyed. However, this approach does not provide enough evidence to show what skill set is actually valued or utilized by the workers in a role (Cunningham, Villaseñor, and World Bank, 2014). This exposes the need to capture data that could suggest the ‘revealed preference’ employers have for a certain skill, for example, when deciding to retain an employee beyond the probation period, and also exposes the need to identify the skills that existing employees use most regularly. This information could be used to guide policies targeted at skill development for the current workforce, re-employment strategies, preparation of workers entering the job market, and so on (Peterson et al., 2001).
The Skills Towards Employment and Productivity (STEP) Skills Measurement Program aims to contribute to the knowledge base and inform policies targeted at improving the alignment between skill supply and demand. Tools developed for the STEP Program comprise an Employer Survey (designed to identify skill demand and constraints using data from employers) and a Household Survey (developed to measure the generic skills of the working-age population).
The STEP Employer Survey (STEP ES) is designed to enable us to better understand how employers view the skills and characteristics of their workers and the challenges employers face in recruiting workers with the desired skills. The survey collects information on the skills that employers look for when they recruit staff, the skills that staff most often apply, the skills-related constraints that employers face, and the way employers attempt to mitigate skill constraints, such as through training and exchanges with education and training providers. The survey also captures firm characteristics, the environment in which employers operate, and the structure of their workforce.

Several innovations have been applied to the survey design as compared to traditional enterprise surveys. Firstly, the information about skill use, skill demand, training, and remuneration is gathered concerning two randomly selected types of workers from among the types of workers each firm employs. The STEP ES identifies the two workers types, as follows:

1. **Type A or white-collar workers**: Workers whose job profiles which generally require higher skill levels, including managers, professionals, and technicians.

2. **Type B or blue-collar workers**: Workers whose job profiles which tend to require medium or low skill levels; this type includes clerical support workers, service workers, sales workers, skilled agricultural workers, craft and related trade workers, plant and machine operators, and elementary occupations.

Secondly, the survey identifies and assesses three main skill domains:

1. Cognitive skills,
2. Job-related technical skills (both of which are categorized under job-related skills), and
3. Socio-emotional skills (categorized under personality traits).

In addition to this, personal characteristics of hired employees are also assessed.

Thirdly, in addition to generating data on the hiring process, the survey collects information on skills used by experienced workers by asking questions on the day-to-day activities of randomly selected occupations. This allows for an identification of the actual importance of particular skill types across different occupations. Finally, the survey questionnaire has been designed with straight-forward questions, applicable to employers of all sizes.

In order to determine whether an employer can be categorized as innovative, questions related to research and development (R&D) have been added to the survey’s background characteristics module. Innovative employers are defined as those who have, in the past three years, introduced any new or significantly improved products, services, manufacturing or logistical processes, management practices, or other similar innovations and have spent on R&D.

The findings included in this snapshot are based on the first two waves of the STEP Employer Surveys carried out in six countries: Armenia, Azerbaijan, China (Yunnan Province only), Georgia, Sri Lanka, and Vietnam. The information presented is

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not meant to be exhaustive or definitive, but rather serves to illustrate the scope and depth of policy-relevant information that may be derived from the STEP ES. While differences in the applied sampling methodology across surveys imply that findings are not fully comparable across countries, the reporting of results for the different surveys nonetheless allows an identification of broad patterns that occur across countries.³

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2. Sri Lanka excludes real estate. Azerbaijan excludes land, water and air transportation; wood processing, pulp and paper; tobacco manufacturing; and recycling. Georgia only includes tourism; construction; IT; and telecommunication. Armenia only includes food processing; fishing; mining; manufacturing; electricity, gas and waterworks; construction; wholesale; retail trade; repair of motor vehicles, motorcycles and household goods; hotels and restaurants; transportation; financial services; and real estate.

3. See Annex 2—Technical Supplement and Tables for further information on sampling design and results.
Key Findings:

- Substantial shares of employers face difficulties when trying to recruit new staff due to a “lack of required skills in candidates.” Interestingly, employers tend to be somewhat more positive about the qualifications of white collar workers rather than blue collar workers. Other constraints include wage expectations, low applicant numbers and dissatisfaction with working conditions.

- Large shares of employers find that the general education and TVET system does not provide individuals with the required ‘practical’ skills and consider the general education and, especially, the TVET of their workers as a constraint on the operation and growth of their business.

- Employers in all countries tend to attach the most value to job-specific technical skills of their workers.

- White collar workers apply more skills in their day-to-day activities than blue collar workers. The incidence and level of complexity of computer use is also comparatively higher for white collar workers.

- Significantly, innovative firms tend to experience problems in recruiting staff much more than traditional firms and are more critical of the current skill-levels of workers. Innovative employers also report a higher provision of training and skill intensity for both worker types across all countries.

- In most countries, relatively few firms have regular interactions with education and training institutions. These interactions are primarily focused on serving employers’ short-term needs, e.g., recruiting.

- Employers rely to a large extent on in-firm training rather than external training to upgrade the skills of their current employees.
Learning How Employers View Job Mismatch in the Economy

Visualizing the Gaps in Skill Demand and Supply

The debate over the “skills gap” (a deficiency in basic skills inherent in the flow of workers entering the workforce), “skills shortage” (a lack of technical skills in the stock of workers available for specific occupations) and “skills mismatch” (a case of under-supply or over-supply of qualified workers) has become an essential component of labor market discussions (Cappelli 2014). The STEP Employer Surveys are designed to provide employers with a voice in this debate, specifically in assessing the level of skills prevalent among job applicants and typical workers at these firms.

Employers report a lack of sufficiently skilled job candidates. The lack of sufficiently skilled candidates poses the greatest constraint on recruitment, according to employers. Across most countries surveyed, employers also report that their firms face a greater challenge hiring adequately skilled white-collar workers than blue-collar workers. The exception is Yunnan Province (China) where employers reported a slightly higher challenge recruiting blue-collar workers with required skill sets.

Hiring difficulties in general are reported by substantial shares of employers across all the countries surveyed—with the highest share being 95% of sampled firms in Yunnan Province (China) in regard to both white- and blue-collar workers. A notable exception is Georgia, where fewer firms, sampled from the tourism, construction, IT, and telecommunication industries, found hiring to be a challenge.

Substantial shares of employers across all countries surveyed consider their employees to be “not fully-qualified.” These shares range from one-third to two-thirds of reporting firms. It is interesting to note that, in contrast to the recruiting concerns noted above, when asked about their current workers employers are more critical about the qualifications of their blue collar workers.

Employers have a negative attitude toward the general education and TVET systems, primarily

Figure 1: Applicants Lacking Required Skills as the Greatest Recruitment Challenge, Armenia and Vietnam

Figure 2: Substantial Recruiting Problems Reported Across All Countries Surveyed

4. See Annex 1 Figure 1A, “Perceptions of blue collar worker qualifications across all countries surveyed.”
because they believe these institutions are unable to produce candidates with the requisite ‘practical skills.’ There is also concern regarding a lack of up-to-date knowledge and the general level of skills imparted. It is worth noting, however, that few employers feel that these systems produce graduates lacking ‘good attitudes’; the focus of their criticism is on job-specific technical skills.

Both employers reporting hiring constraints and those more critical of their workers’ qualifications are both more likely than other employers to consider the general education and TVET system as ‘problematic.’ This holds true across all countries surveyed.

A substantial share of employers across countries feel that the general education and the TVET system (in particular) pose a constraint to the operation and growth of their firms. The lowest share of

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5. See Annex 1 Figure 1B, “Share of employers reporting both recruiting constraints and negative perceptions of the general education and TVET, Georgia.”

6. See Annex 1 Figure 1C “Employers considering general education and TVET as a constraint on firm operation and growth across all countries.”
employers holding this view is in Vietnam, but even there as much as 48% and 58% of surveyed employers regard general education and TVET, respectively, as posing constraints. Compared to other labor constraints however, the general education and TVET systems vary in significance.

Finding workers with previous experience is a significant labor constraint on business operation and growth. This difficulty in finding experienced workers features as the top constraint in all countries except Armenia, where payroll taxes are considered a greater deterrent. This may also suggest further dissatisfaction with the level of skills the workforce currently possesses as well as indicate a potential unwillingness to provide employees with training upon hiring.

Among other constraints to doing business, such as tax rates and political uncertainty, the perceived impacts of labor constraints vary substantially among the countries surveyed.
**Identifying Skills Valued by Employers: The Right Skills for the Job**

**Understanding the Demand**

Current data measuring skill demands are posited on employers’ ‘stated preferences,’ yet this is often considered as a shortcoming when attempting to identify the skill-profiles valued by employers (Cunningham, Villaseñor, and World Bank, 2014). To avoid this, the STEP Employer Survey module on new hiring attempts to uncover possible ‘revealed preference’ in skills, by determining what skills employers look for when deciding to retain an employee beyond the probation period. The survey module further reviews those skills that are used most regularly by the existing workforce.

**Job-specific technical skills are prioritized by employers when they make retention decisions for both white- and blue-collar workers.** The secondary criteria for retaining new hires vary among worker types. Across all countries, leadership skills and numeracy skills are ranked as most important to retaining white-collar workers. For blue-collar workers, communication skills and the ability to work independently are most valued.

When asked to compare job-relevant skills against personality traits and personal characteristics, employers place job-relevant skills such as communication and problem-solving on the job relatively higher. This result has been the same across all countries surveyed. An interesting difference is that although preferences follow the same ranking pattern for the two types of workers, personal characteristics play a more pivotal role than personality traits for blue collar workers.8

**White collar workers use a larger number of skills and apply them more regularly in their jobs.** This holds true across almost all countries with the exception of Yunnan Province, where blue collar workers are reported to require the use of an equally high

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7. The personality traits identified in the STEP ES are conscientiousness, agreeableness, emotional stability, extraversion and openness to experience. Personal characteristics included are age, appearance, gender, and family relations/personal ties.

8. See Annex 1 Figure 2A, “Ranking job-specific skills, personality traits and personal characteristics for staff retention decisions across all countries surveyed.”
number of skills in their roles. A ranking of skills most used on the job reveals the top skills to be teamwork, numeracy, and, to a lesser extent, reading. This holds true for both types of workers.

The survey goes further to determine if these patterns exist in firms of different sizes. It finds that whereas for white-collar workers skill usage for each identified skill typically increases as firm size increases, for blue-collar workers the patterns of skill intensity and firm sizes vary across countries. For example, in Vietnam the use of mathematics by blue-collar workers is higher in smaller firms (47%) than in larger firms (25%). This pattern varies with each identified skill and by country.

Figures 8a and 8b: Skill Usage by Firm Size for Each Worker Type, Sri Lanka and Vietnam

9. See Annex 1 Figure 2B, “Number of different skills that workers regularly use in Azerbaijan and Yunnan, share of workers.”
Both the incidence and the level of complexity of computer use are higher for white-collar workers across all countries. The complexity of computer use also increases with firm size in all countries with the exception of Armenia and Yunnan Province, where it is relatively higher in medium-sized firms. (Both of these patterns are captured for the share of workers who are required to use a computer as part of their daily role.)
Capturing Requirements for Innovative Firms

Factoring in the Potential

The World Bank’s *World Development Report 2013: Jobs* confirmed that there is indeed a correlation between firms’ engaging in innovative activities and having skills constraints. The STEP Employer survey compares employers’ perception of skill mismatch in innovative firms with that in traditional firms. It further compares the constraints faced by innovative firms as well as the unique skills profile required by innovative employers. The responses are very interesting as there are striking differences in findings from innovative and traditional firms.

Employers in innovative firms are typically less satisfied than traditional employers with the present qualifications of their workers and the effectiveness of general education and TVET systems. This perception is observed across all countries surveyed and is particularly significant when concerning the general education and TVET institutes. In almost all observed countries, more than 50% of innovative employers responded that the general education and TVET systems constrain the operation and growth of their business.\(^{10}\)

When assessing the skills level of current employees, employers in innovative firms in Azerbaijan, Armenia, Georgia and Vietnam were more critical of white-collar workers, whereas innovative employers in Sri Lanka and Yunnan Province were less satisfied with the qualifications of their blue-collar workers.

Innovative firms are more exposed than traditional ones to recruiting constraints, across all countries surveyed, in seeking white-collar workers. This observation also holds for recruiting blue-collar workers in all countries except Azerbaijan, where traditional employers found recruiting for these workers to be problematic for business, Yunnan Province and Sri Lanka.\(^{10}\)

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1. See Annex 1 Figure 3A, “Comparison of innovative and traditional firms reporting general education and TVET systems are problematic for business, Yunnan Province and Sri Lanka.”
Figure 12: Comparison of Innovative and Traditional Firms Reporting Recruitment Problems Across All Countries

Skill intensity is higher for both worker types in innovative firms across all countries surveyed. The share of workers applying skills regularly is reported to be higher in innovative firms. However, a few specific skills are exceptions, such as writing skills positions a greater constraint. In the case of Azerbaijan, this observation is determined for all enterprises except those in the following sectors: land, water and air transportation; wood processing; pulp and paper; tobacco manufacturing; and recycling.
among white-collar workers and team work and use of foreign languages among blue-collar workers, which are reportedly used more often in traditional firms.

The same is the case with the level and complexity of computer use. Workers tend to use computers more regularly in innovative firms and more of the use is typically for specialized and complex tasks, as compared with traditional firms, where usage is reported to be more straightforward or only moderately complex.

‘Critical thinking’ is a unique skill that is valued more highly by innovative firms than by traditional firms for both white- and blue-collar workers. This value is revealed when making retention decisions for new hires. For white-collar workers, innovative firms attach importance to technical skills, team work, and critical thinking. Similarly, for blue-collar workers, the skills these firms generally value more are critical thinking and leadership.
Learning If and How Employers Are Supplementing Their Training Needs

Accounting for All Tools and Resource

Recent literature points to a growing trend of employers spending fewer resources on training either new hires or existing workers and choosing instead to recruit employees already possessing the required skill sets (Cappelli, 2014). The STEP Employer Survey explores the propensity of firms to provide internal and external trainings to their employees in an effort to highlight any attempts being made to meet skill requirements.

White-collar workers receive more external training and blue-collar workers receive more internal training, across all countries surveyed. External training refers to any training organized by the firm outside the workplace including but not limited to vocational training by public or private technical institutes, via equipment suppliers, or NGOs and international organizations. Internal training refers to training received on the premises of the workplace, such as on-the-job training or training by managers, other employees, dedicated firm trainers or even external consultants. Firms are found to rely more on internal than external training to upgrade the skills of either type of worker. Larger firms provide more training opportunities to their workers than smaller firms do.

Innovative firms provide significantly more internal and external training to their employees than traditional firms. This difference between innovative and traditional firms is most pronounced concerning external training opportunities. An interesting exception is Yunnan Province, where internal training for both types of workers is approximately equally common in traditional and innovative firms.

Across most countries surveyed, less than 25% of employers have regular interactions with education and training providers. The only exception is Yunnan Province, where close to 60% of employers have linkages with such institutes. Effective interactions with education and training providers is considered essential for improving the alignment of the
Figure 16: Incidence of Internal and External Training in Innovative and Traditional Firms, Azerbaijan and Yunnan Province

Figure 17: Share of Employers Regularly Interacting with General Education and TVET Institutes, across All Countries Surveyed

Figure 18: Share of Innovative and Traditional Employers Regularly Contacting General Education and TVET Institutes, Georgia and Azerbaijan
demand and supply of skills. However, in most countries the share of employers engaging with these institutes ranges from 4% to 22% and the engagement is most often related to white-collar workers.

Compared to traditional employers, innovative employers have more regular interactions with training institutes. In Georgia, innovative employers are four times as likely as traditional ones to have this contact.

These interactions are primarily focused on serving employers' short-term needs, such as recruiting, work placements, and employee training. Far fewer firms engage in curriculum development or student testing, that is, in efforts to structurally improve the quality and relevance of skill supply.
References


Resources

The STEP Skills Measurement Data
The STEP Skills Measurement datasets are available in the World Bank’s Microdata Catalog (link provided below), including relevant background and implementation documents for the surveys conducted in each country. The Microdata Catalog is regularly updated as new findings and materials related to STEP are made available.

Country Reports
In addition to the STEP Methodology Note (Pierre et al., 2014) and the STEP Snapshot 2014 the following country reports have also been published and are available online:


Links
Annex 1—Additional Figures

Annex Figure 1A: Perceptions of Blue-Collar Worker qualifications across All Countries Surveyed

Annex Figure 1B: Share of Employers Reporting Recruiting Constraints and Also Having Negative Perceptions of the General Education and TVET Systems, Georgia
Annex Figure 1C: Employers Considering General Education and TVET as a Constraint on Firm Operation and Growth across All Countries

Annex Figure 2A: Ranking of Job-Specific Skills, Personality Traits and Personal Characteristics for Staff Retention Decisions across All Countries Surveyed
Annex Figure 2B: Comparison of the Number of Different Skills White- and Blue-Collar Workers Regularly Use in Azerbaijan and Yunnan, Share of Workers

Annex Figure 3A: Comparison of Innovative and Traditional Firms Reporting General Education and TVET Systems Are Problematic for Business, Yunnan Province and Sri Lanka
Annex 2—Technical Supplement and Tables

Overview of the Questionnaire:

45–60 MINUTES

The STEP Employer Survey—Modules

The Employer Survey is designed to better understand the demand for skills emanating from the private sector. It is divided into the following five modules.

Information and Workforce Characteristics (Module 1)

This module helps interviewers establish whether or not they are talking to a person knowledgeable about the recruiting practices and skills development processes in a firm. It seeks information on the firm being surveyed in order to determine its type and the economic sector in which it operates. The module then goes into details about the workforce of the firm, in particular requesting details on the share of females, the share of foreign workers, and changes in the workforce.

Skills Used by the Current Workforce (Module 2)

This module starts with the random selection of two occupations, which the questionnaire will then focus on. The respondent is asked to select one occupation from Type A (managers, professionals, and technicians), and another from Type B (clerical support workers, service workers, sales workers, skilled agricultural, craft and related trades workers, plant and machine operators, and elementary occupations) through a random sticker method. The module goes on to ask a series of questions about specific skills that workers in these two types of occupations are currently using.

Information on New Hires (Module 3)

To understand skills needs at the firm, this module focuses on new hires in the two selected occupation types. It asks respondents to rank skills in order of
importance when they are deciding to keep a worker beyond the probation period. Skills are grouped under three categories: (i) personal characteristics (age, appearance, gender, networks); (ii) cognitive and job-relevant skills (reading, writing, mathematics, English, foreign language, technical skills, communication skills, leadership skills, team work skills, creative and critical thinking, problem solving, ability to work independently, time management); and (iii) personality traits (Conscientiousness, Emotional stability, Agreeableness, Extraversion, Openness to experience). The module also asks for the education level of the most recent hire.

In addition to obtaining information on the skills that are valued by the firm, the module seeks to obtain information on the firm’s recruitment strategies and any difficulties it might be experiencing when hiring. Subsequent questions collect more information on how the firm recruits: the places from which it recruits, the time and number of offers it takes to fill a vacancy, whether salary is negotiable at entry, and the remuneration offered to the most recent hire.

**Training and Compensation (Module 4)**

This module collects information on training provided by the firm, both on the premises and outside the workplace. It asks for the share of workers who benefited from such training programs, the number of days such training involved, and the costs and type of training provided.

The module also contains questions to assess the overall quality of the formal education system (technical and vocational as well as general). It asks a series of questions aimed at determining whether the firm feels that the education system provides adequate skills and whether it has direct contacts with the education system and why.

**Firm Background (Module 5)**

Given the length of the survey and the usual difficulties in collecting financial information from firms, the fifth and last module features only two simplified questions with respect to the firm’s accounts asking about the current financial performance of the company and prospects in the coming three years. Furthermore, the module includes detailed questions about the various ways in which labor market constraints could impair business, as well as questions that situate labor market constraints with respect to the other investment-climate constraints.

**Methodology and Technical Standards**

**Sampling Strategy and Weighting Procedure**

The units of observation were workplaces, and the target populations were low- and medium-income countries. The sectors covered varied by country depending on the core-sector composition of the respective national economy (see Annex Table A1, below). Geographic coverage primarily included the respective capitals and other urban areas and, for some countries, a wider range of provinces, sometimes including rural areas.

The sampling frames in the surveys were derived from national firm registries. This had the advantage of being cost and time-effective compared with other methods. However, since firm registries are not always updated, a substantial share of registered firms did not exist by the time of the survey, leading to resampling and potentially biased results in some countries. Nevertheless, the survey teams collected the intended number of observations, and sample sizes ranged from 300 to 640 units.

The stratification was undertaken using firm size and geographical strata (for most but not all surveys). As explicit strata the sampling distinguished among large, medium and small workplaces and firms within the capital and other urban areas. Some countries also included provinces as geographical strata. Moreover, the samples were further stratified by sectors and geographic subareas to ensure a higher level of representativeness. In most countries, firms with fewer than five workers were excluded.

The data was weighted using post-stratification weights. The population weights were derived from the firm registries and adjusted to account for nonresponse. All surveys, except the one in Sri Lanka, were weighted. The latter was not weighted because it simultaneously used two different firm registries that were not similar in their structure. The weights for data from Yunnan Province appear to be unadjusted for nonresponse.
Annex Table A1:

<table>
<thead>
<tr>
<th>Target Population</th>
<th>Georgia</th>
<th>Azerbaijan</th>
<th>Armenia</th>
<th>Sri Lanka</th>
<th>Vietnam</th>
<th>Yunnan</th>
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<tbody>
<tr>
<td>Non-government workplace</td>
<td>Non-agriculture private workplaces</td>
<td>Non-government workplaces</td>
<td>Workplaces</td>
<td>Workplaces</td>
<td>Non-state owned enterprises</td>
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**Sampling Frame**
- List of businesses from the State Department of Statistics including approx. 3735 firms
- List from Azeri Statistical Office which was updated by WB team in 2013 including approx. 3365 firms
- List of registered businesses with the Social Security State Agency from 2012 including approx. 9100 firms
- Department of Census and Statistics (DCS) and Associations list for tourism sector including approx. 46524 firms
- GSO's dataset of the enterprise census 2009 including approx. 29914 firms
- List of basic units of Kunming in 2010 including 32245 firms

**Geographic Coverage**
- Capital (Tbilisi) and other urban regions (Abkhazia and South Ossetia excluded)
- Capital (Baku) and other urban regions
- Capital (Yerevan) and other urban
- All regions
- Hanoi, Hochiminh and 9 provinces
- Urban area of Kunming (including 4 administrative districts: Wuhua, Panlong, Guandu and Xishan)

**Size of Workplaces**
- Large: >100 workers, Medium: 20–100 workers, Small: < 20 workers
- Large: >100 workers, Medium: 20–99 workers, Small: < 10 workers
- Large: >51 workers, Medium: 11–50 workers, Small: < 10 workers
- Large: >51 workers, Medium: 11–50 workers, Small: 5–20 workers
- Large: >100/50 workers, Medium: 11–50/100 workers, Small: < 10 workers

**Sample Sizes**
- 354
- 316
- 384
- 640
- 330
- 300

**Sampling Details**
- Initial sample of 400 firms followed by 3 replacement rounds out of reserve sample of another 430 firms. 45% of attempted firms completed the interview.
- Response rate based on activated eligible firms is about 20.3%.
- Initial sample of 402 firms and reserve sample of another 1200 firms, 19.7% of attempted firms completed the interview. Response rate based on activated eligible firms is 44%
- Initial sample of 400 firms and 2 reserve samples of 400 and 304 firms respectively 36% of attempted firms completed the interview. Response rate based on activated eligible firms is 44%
- The target sample was 600 firms, resampling took place 3 times. The size of the reserve sample is unclear, ranging from approx. 1100-1400 firms. Sectors came from different source frames and were recorded separately or not at all leading to some inconsistency.
- Initial sample of 330 firms and the two reserve samples of together 130 firms, 56% of attempted firms completed the interview. Response rate based on activated eligible firms is 64%
- Initial target sample was 300 and a reserve sample of 68 firms, attempted. 91% of attempted firms completed the interview. Response rate 91%

**Sector Details**
- Excludes: Land, water and air transportation, Wood processing, Pulp and Paper, Tobacco Manufacturing & Recycling
- Only includes: Food processing, Fishing, Mining, Manufacturing, Electricity gas and waterworks, Constructions, Wholesale, Retail trade, Repair of motorcycles, motor vehicles and household goods, Hotels and restaurants, Transportation, Financial services & Real estate.
- Excludes: Real estate

**Notes**
- Sampling weight follows stratification and is adjusted for non response
- Sampling weight follows stratification and is adjusted for non response
- Sampling weight follows stratification and is adjusted for non response
- No (basic) weights calculated. The sampling frame provided by the DCS was outdated (from 2002) and contained only few cases for the tourism sector thus a different frame was used. It was not possible to sample using a similar stratification for all sectors, also the sizes of the target population by stratum were not consistent.
- Sampling weight follows stratification and is adjusted for non response.
- The strata is not entirely clear as documentation is vague. There is no information about the number of firms that were not eligible
- Weight is not adjusted for non-response, its construction follows the sampling sector.
Some country-specific differences did evolve in the implementation of the surveys. To the extent that these are documented, the sampling specifics and differences are displayed in the following two tables: Annex Table A1 focuses on the sampling design that was used in each country and Annex Table A2 explains the sampling results and particularities of each country.

More detailed sampling results, and response rates in particular, are presented in Annex Table A2, which displays the relation between the target samples, attempted interviews, and the actual completed interviews. Response rates of the survey were moderate. For most countries, there is also information on the shares of firms that did not complete an interview (and why this was the case). The most common reasons for not completing an interview were refusal, the firms’ being closed or not existing, false contact details, and firms’ being ineligible (e.g., they had too few employees to fit the profiling or their actual business was registered under a different sector of operation).

In Azerbaijan, for example, the response rate was particularly low. In addition to the initial sample of 402 workplaces a reserve sample of 1,200 firms had to be activated. In fact, 47% of the 1,602 sampled firms could not be found or contacted, and another 30% refused the interview altogether, leaving the sampling with a response rate of 20.3% based on activated eligible firms. This excludes firms that were closed, that were oil-companies, or that were state-owned or not eligible (i.e., 47 firms).

### Calculation of Selected Variables

**Innovation**

Innovation is measured by the extent to which firms have adopted innovative measures in the past three years. In the question, the firm can name up to four given measures of innovation (new technologies, new processes, new products, new services) The analysis uses a binary variable. It defines firms that have adopted two or more measures as innovative, and all others as traditional firms.

**Annex Table A2:**

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<th>Sampling Profiles</th>
<th>Georgia</th>
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</tr>
</thead>
<tbody>
<tr>
<td>Targeted Interviews</td>
<td>400</td>
<td>400</td>
<td>402</td>
<td>594</td>
<td>600</td>
<td>300</td>
</tr>
<tr>
<td>Completed Interviews</td>
<td>351</td>
<td>384</td>
<td>316</td>
<td>330</td>
<td>624</td>
<td>300</td>
</tr>
<tr>
<td>In % of Target</td>
<td>88%</td>
<td>96%</td>
<td>79%</td>
<td>56%</td>
<td>104%</td>
<td>100%</td>
</tr>
<tr>
<td>In % of Attempted</td>
<td>45%</td>
<td>36%</td>
<td>20%</td>
<td>56%</td>
<td>n.a</td>
<td>91%</td>
</tr>
<tr>
<td>Attempted Firms</td>
<td>783</td>
<td>1064</td>
<td>1602</td>
<td>594</td>
<td>n.a</td>
<td>330</td>
</tr>
<tr>
<td>Closed or Not Existing</td>
<td>14%</td>
<td>10%</td>
<td>1%</td>
<td>Incl. in not eligible</td>
<td>n.a.</td>
<td>n.a.</td>
</tr>
<tr>
<td>Companies Not Found/ Contacted</td>
<td>10%</td>
<td>27%</td>
<td>47%</td>
<td>Incl. in refused</td>
<td>n.a.</td>
<td>n.a.</td>
</tr>
<tr>
<td>Not Eligible</td>
<td>2%</td>
<td>8%</td>
<td>2%</td>
<td>13%</td>
<td>n.a</td>
<td>n.a.</td>
</tr>
<tr>
<td>Refused (all cases)</td>
<td>29%</td>
<td>19%</td>
<td>30%</td>
<td>31%</td>
<td>n.a</td>
<td>9%</td>
</tr>
</tbody>
</table>
Firm Size

The variable *firm size*, used for disaggregation, is defined as follows:

‘Firms with 10 or less workers are small; firms with 11 to 100 workers are medium; and firms with more than 100 workers are defined as large.’

It should be noted that this variable is based on the questionnaire’s responses and differs from the firm-size stratum.

Indexes on Importance of Skills

The analysis of the skills that employers look for when retaining their workers is based on indexes. They are calculated as follows:

1. The question asks the employers to rank a defined set of skills as “most important,” “second most important,” and so on; for example, for job-related skills this goes on until “fifth most important.” At least one or more skills from the list are not selected by employers.

2. For each skill, the weighted percentage shares of how important it was rated to be is listed.

3. Each share is then “weighted” as follows: 0 for (all) skills not mentioned; the highest number for the “most important” category (5 in the case of most important job-related skill); the second highest number for the “second most important” category (4 in the example); and so on.

4. The index results from the sum of the weighted shares for each skill.

Example:

<table>
<thead>
<tr>
<th>Literacy</th>
<th>Share</th>
<th>Index-Weight</th>
<th>Index Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Most Important</td>
<td>24%</td>
<td>5</td>
<td>1.2</td>
</tr>
<tr>
<td>2nd Most Important</td>
<td>6%</td>
<td>4</td>
<td>0.24</td>
</tr>
<tr>
<td>3rd Most Important</td>
<td>4%</td>
<td>3</td>
<td>0.12</td>
</tr>
<tr>
<td>4th Most Important</td>
<td>5%</td>
<td>2</td>
<td>0.1</td>
</tr>
<tr>
<td>5th Most Important</td>
<td>3%</td>
<td>1</td>
<td>0.03</td>
</tr>
<tr>
<td>Not Mentioned</td>
<td>59%</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Total</td>
<td>100%</td>
<td>SUM</td>
<td>1.69%</td>
</tr>
</tbody>
</table>

The variable for how important or problematic the various labor constraints are to business operations and growth is calculated similarly.

Regular Use of Skills

The variable of the number of regularly used skills is computed as follows:

1. The employer is asked if the job of the respective Type A and B worker regularly involves a set of seven skills.

2. Each firm is assigned a value from 0 to 7 depending on the number of skills that the workers regularly use, which results in two variables; that is, one variable for Type A workers and one for Type B workers.