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# What firm characteristics determine women's employment in manufacturing? Evidence from Bangladesh

Salma Ahmed<sup>\*</sup>, Simon Feeny<sup>†</sup>, Alberto Posso<sup>‡</sup>

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## Abstract

**Purpose** – This study investigates the principal determinants of women's employment in the manufacturing sector of Bangladesh using a firm-level panel data from the World Bank's 'Enterprise Survey' for the years 2007, 2011 and 2013. The paper sheds light on the demand-side factors, mainly firm-level characteristics, which also influence this decision.

**Design/methodology/approach** - We estimate a fractional logit model to model a dependent variable that is limited by zero from below and one from above.

**Findings** - The results indicate that firm size, whether medium or large, and firms' export-oriented activities, have an important impact on women's employment in the manufacturing sector in Bangladesh. Moreover, we find that women are significantly more likely to work in unskilled-labour intensive industries within the manufacturing sector.

**Research limitations/implications** – The research is limited to Bangladesh; however, much of the evidence presented here has implications that are relevant to policymakers in other developing countries.

**Practical implications** – The study identifies factors that affect female employment, that is, where the main constraints to increase female labour force participation. The study focuses on the demand-side factors, which has been somewhat neglected in recent years. As such, it has practical policy implications.

**Social implications** – Focusing on female employment in Bangladesh also sheds light on the nexus between labour market opportunities and social change within a country that is characterised by extreme patriarchy, which has wide-reaching implications.

**Keywords** Bangladesh, female employment, manufacturing firms

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## 1. Introduction

On April 24, 2013, Rana Plaza, an eight-storey commercial building in Savar, a sub-district of the Greater Dhaka Area, Bangladesh, collapsed. The building, owned by the family of a prominent politician, housed a large number of garment factories that employed approximately 5,000 people, of whom 1,129 died and 2,515 were seriously injured. In the days that followed this, the deadliest garment-factory accident in history, garment workers across the industrial areas of Dhaka, Chittagong and Gazipur rioted (according to a report on the BBC website from May 3, 2013; <http://www.bbc.com/news/world-asia-22394094>). However, the uproar did not end in Bangladesh, with political leaders, NGOs, and religious organisations around the world not only criticising working conditions in the country, but also criticising multinational garment brands such as Benetton, Mango, and Walmart for engaging ‘sweatshops’ to manufacture their clothes. In the immediate aftermath of this tragedy, yet another terrible statistic emerged: more than half of the victims were women, children and many more who were at nursing facilities in the building (Nelson, 2013).

Labour-intensive manufacturing industries with poor working conditions often spring up in developing countries when they embark on export-oriented development strategies. Studies using household-level surveys have shed light on many supply-side reasons—individual, demographic and household-related—why workers, mainly women, opt to work in industries with poor working conditions (Kabeer and Mahmud, 2004; Salway *et al.*, 2003), with the literature in the context of Bangladesh having focused primarily on the garment sector (Amin *et al.*, 1997; Kabeer, 1991; Kabeer and Mahmud, 2004; Kibria, 1995). However, very little is known about the demand-side factors, mainly firm-level characteristics, which also influence employment and participation rates, particularly for women. Fasih and Ghazalian (2015) provide a review of the literature that does exist on this issue. They show that Bratti *et al.* (2005), find importance of demand-side and job-related factors in explaining labour market participation of mothers in Italy and that Buchanan *et al.* (2010) also demonstrate that demand-side factors affect female employment and workforce participation rates. They also document Abe’s (2013) finding that in addition to demand-side factors, supply-side factors determine female labour force participation rates for the case of Japan. Pissarides *et al.* (2005) show that low female employment and workforce participation rates may be able to be explained by employers’ preferences and characteristics, while Lee *et al.* (2008) provide a similar explanation for the low labour force participation rates found among married women in South Korea. Fasih and Ghazalian (2015) themselves demonstrate that demand-related factors, such as private ownership and exporting activities, are important predictors of women’s employment in manufacturing firms located in the Middle East and North Africa (MENA). This paper complements this small existing body of knowledge by providing a firm-level analysis of female employment in the manufacturing sector of Bangladesh.

Complementing the extant literature with insights into the firm-level determinants of female employment rates in Bangladesh is important for three reasons. Firstly, as was suggested above, Bangladesh's manufacturing employment, catering mainly to women, has had an impressive rate of growth in recent years. Manufacturing employment grew by 9% per annum over the period 1995–2009, with female employment in manufacturing exhibiting a growth rate of 6% per annum over the same period (BBS, 1996, 2011). Secondly, in spite of these impressive growth rates, scholars have focused only on the determinants of women's employment within the garment sector (Amin *et al.*, 1997; Kabeer, 1991; Kabeer and Mahmud, 2004; Kibria, 1995). This is an important omission, because, although economic liberalisation has been associated with a significant expansion of women's paid employment in this sector, there has also been considerable growth in a number of other manufacturing industries, including food, chemicals and pharmaceuticals. Thirdly, focusing on female employment in Bangladesh also sheds light on the nexus between labour market opportunities and social change within a country that is characterised by extreme patriarchy, which has wide-reaching implications. The analysis in this paper sheds light on these issues using data from the World Bank's 'Enterprise Survey' for Bangladesh, which is a firm-level panel survey that has recently been made available publically for the years 2007, 2011 and 2013.

We analyse the demand-side determinants of female employment in the manufacturing sector using the fractional logit model, which is estimated using the Quasi-Maximum Likelihood Estimator (QMLE). The paper builds on the work of Fasih and Ghazalian (2015), who analyse the demand-side determinants of female employment in the MENA manufacturing sector. However, we depart from the study by Fasih and Ghazalian in two significant ways. First, because we have panel data (more than one observation for each manufacturing firm), we control for firm fixed effects when estimating QMLE. This also allows us to control for the various unobservable and time-invariant features of the firm that tend to be correlated (positively or negatively) with female employment (e.g., Wagner, 2003). Second, we also estimate the fractional logit model using the generalised estimating equation (GEE) to account for heteroskedasticity and serial correlation in the standard errors within the panel dataset (Cui, 2007; Papke and Wooldridge, 1996, 2008).<sup>1</sup> Indeed, while our empirical results are robust, they show some notable differences between the GEE and QMLE results.

The remainder of the paper is organised as follows. Section 2 provides a brief overview of Bangladesh's social and economic context. Section 3 describes the relevant data and outlines firm-level characteristics. Section 4 discusses the estimation strategy. Section 5 presents our main findings, and Section 6 concludes the paper.

## **2. The context of Bangladesh**

Bangladesh is part of a region that practices extreme patriarchy. The societies in this part of Asia tend to be characterised by the practice of female seclusion, patri-lineal principles of descent and inheritance, patrilocal principles of marriage, and strict patriarchal authority structures within the family. Restrictions on women's mobility in the public domain mean that they work either as unpaid family labour or in forms of paid work that can be carried out within the home. The invisibility of such work has meant that the female labour participation rates in these regions have tended to be extremely low. For example, official labour statistics show that women's share of the total employment in 1995 was five million (14%), increasing to 16.2 million (30%) by 2009 (BBS, 1996, 2011).

However, like any other form of social relations, patriarchal relationships can be modified, intensified or transformed over time. While progress on many fronts has been slow, others have seen remarkable achievements. For instance, the gender disparities in gross enrolment ratios have been eliminated at the primary school level and reduced at the secondary level. Bangladesh has also pioneered microcredit programs which lend to millions of women from poor and landless households on the basis of group-based collateral. These programs have expanded women's opportunities for self-employment in rural areas. However, there are still social barriers to women's participation in paid work outside the home, and returns to women's labour in these off-farm activities continue to be low (Ahmed and Maitra, 2010, 2015; Ahmed and McGillivray, 2015; Asadullah, 2006; Hossain and Tisdell, 2005; Kabeer, 2001; Kapsos, 2008; Rahman and Khandker, 1994).

As a result, many women migrate to urban areas in search of work, either with their husbands if they are married, or on their own if they have been widowed, divorced or abandoned. Community-based constraints are usually less severe in urban areas, resulting in higher rates of female participation in paid work. These trends are present in the formal manufacturing sector, particularly in the garment sector and the export processing zones of Bangladesh. Of course, these employment opportunities were due to a greater export-orientation, which generated considerable employment opportunities, particularly for women. The share of female employees in total manufacturing employment in 1995 was 1.3 million (24%), increasing to two million (28%) by 2009 (BBS, 1996, 2011). This not only represents a change in human resources allocation and economic productivity at the population level, but also has implications for individual and household well-being. However, women employed in these industries experience poor working conditions because they hold low-skilled jobs across different manufacturing industries and earn less than men for similar work, in spite of the anti-discrimination laws enacted in 1972 (Majumder and Zohir 1993, 1994).

### 3. Data and variables

The data used in this study were obtained from the World Bank's Enterprise Surveys. The surveys provide the most comprehensive firm-level panel data in emerging markets and developing countries, and include firm-level characteristics, gendered employment, annual sales, workforce composition, infrastructure, innovation and technology, business–government relationships, and performance measures. In Bangladesh, the first wave of the survey was carried out in 2007, while the second and third waves were conducted in 2011 and 2013, respectively.<sup>2</sup> The survey respondents were the business owners and managers of 120 manufacturing firms that were interviewed in all three rounds, resulting in 360 observations. Of these firms, 117 are located primarily in the two main cities of Bangladesh: Dhaka and Chittagong. The manufacturing subsectors that are covered by the data set include food, textiles and garments, leather, chemicals and pharmaceuticals, electrical, and other manufacturing. Data were pooled for the three years 2007, 2011 and 2013. After dropping observations with missing values for the dependant variables and other covariates, we end up with 303 observations. The estimating sample is an unbalanced panel, with an average of 2.9 observations per firm.<sup>3</sup>

The dependent variable, female employment in manufacturing, is defined in three different ways: (1) the fraction of females among all full-time permanent workers; (2) the fraction of females among all full-time production workers; and (3) the fraction of females among all full-time non-production workers.<sup>4</sup> Our measures are consistent with those of Fasih and Ghazalian (2015), who analyse female employment in the MENA manufacturing sector.

There are a number of firm-related factors that can determine female employment rates. Following Lee *et al.* (2008), the firm size is included using two dummy variables (medium and large firms), with the reference category being 'small firms'. Larger firms (i.e., with at least 100 employees) tend to face greater regulatory scrutiny, and therefore might employ a greater proportion of women in order to comply with gender-equity regulations. At the same time, as noted by Fasih and Ghazalian (2015), some studies have suggested that larger firms have more unpleasant working environments because of specialised divisions of labour and an impersonal working environment which could reduce the female supply of labour for such firms (Masters, 1969; Schmidt and Zimmermann, 1991). Thus, the direction in which firm size affects female employment is theoretically ambiguous. The firm's profit ratio (defined as total profit/total sales), is also included, to control for firm performance. It is expected that firm performance should increase the demand for labour, though it is not clear how it will contribute to female employment. Following Fasih and Ghazalian (2015), firm age is also included in the model.

Other firm-level characteristics that may affect female employment in Bangladesh include whether the owner of the firm is female,<sup>5</sup> the number of years of experience of the manager, whether the firm has a website, and whether it exports its output (Fakih and Ghazalian, 2015). It is natural to expect that firms with a female presence in ownership may hire more women than men. In addition, a greater presence of women in ownership in a country may reflect better opportunities for women, and hence, a greater gender parity in law. The use of a website serves as a proxy for the role of computers in firm operations (computer usage for business purposes is not measured in the data directly), as Bresnahan *et al.* (2002) find that the use of technologies such as computers is complementary to workplace reorganisation in measuring productivity and the demand for skilled labour. Hence, IT use may have negative implications for women's employment than for men's employment. At the same time, there are a few studies that suggest that technological change might have a positive effect on female labour force participation by facilitating communications and the flow of information between firms and the labour market (e.g., Olivetti, 2006).

Exporting firms tend to hire greater proportions of female workers than non-exporting firms. One reason for this could be that, in many developing countries, exports are still labour intensive (relative to non-exports) and less skill-intensive. This is favourable for female employment, as, on average, women are less educated and skilled, and are more likely to work for lower wages (Barro and Lee, 2013). Firm ownership, and specifically whether the firm is owned by the government or foreign owned, is also included. In this case, the reference group is 'private domestic ownership'. It is generally hypothesised that government-owned firms tend to employ more female workers than other firms, since they offer jobs that are considered to be 'family friendly' (Hewlett and Rashid, 2010). Foreign-owned firms also usually employ more female workers than private domestic firms, in order to take advantage from women's under-utilised skills (Hewlett and Rashid, 2010). Finally, the firm's composition of production workers is also included. Following Fakih and Ghazalian (2015), this variable is constructed in two ways: the fraction of skilled production workers among total production workers and the fraction of non-production workers relative to total employment. Importantly, the fraction of 'skilled' production workers must be treated solely as a subjective indicator of skill-intensity, as it is based on the question 'At the end of the fiscal year, how many permanent, full-time individuals working in this establishment were skilled or unskilled?'. Therefore, the skill-intensity value obtained from this variable may not necessarily reflect the definition of skill-intensive industries, such as high-end chemical manufacturing, generally used by economists (see Ariff and Hill, 2011). Furthermore, 'production workers' generally refers to workers who are engaged in fabrication, assembly and related activities, and excludes supervisors, administration staff and sales staff. That is, production workers can be either skilled or unskilled, within a more general definition. Nevertheless, the manager's perception of the firm's skill-intensity may be a relevant demand-side factor that could influence the decision to hire more or fewer women.

Following convention, we also include industry and time dummies. Industry dummy variables (with the reference category being textiles/garments) are used to control for unobservable time-invariant differences across industries (such as in the rate of technological progress), while time dummy variables are used to control for economy-wide shocks to the labour demand (such as financial shocks or reforms). We also include a dummy variable for Chittagong city (the commercial capital of Bangladesh), to control for regional differences in women's labour force participation.<sup>6</sup> The reference category is 'Dhaka'. The variations in local labour market conditions play an important role in women's labour supply decisions, by providing different structures and opportunities to work in particular occupations or industries, together with inter-city differences in wage levels.<sup>7</sup>

Table 1 provides descriptive statistics for the variables described above. These statistics indicate that 36% of the females in the sample are full-time permanent workers, and 38% of females are full-time production workers, whereas only 19% of females are employed as full-time non-production workers over the three surveys. In general, manufacturing firms are larger and more likely to be export-oriented (55%). The latter finding is consistent with those of Moghadem (2005) for the MENA region. It is striking that about 96% of the firms are domestically owned, 66% use their own website, and 41% are owned by women. It is not surprising that 63% of the manufacturing firms in our sample belong to the textiles/garments industry. Table 2 presents standard multicollinearity tests to inform our empirical strategy. The Variance Inflation Factor (VIF) scores reported in the table are small, with averages of 1.43–1.53, indicating that multicollinearity is not a statistical issue in our data.<sup>8</sup> A full list of the variables, along with their definitions, can be found in Appendix Table A1.

[Table 1 about here]

[Table 2 about here]

#### **4. Empirical strategy**

In this section, we specify the statistical model that is used to estimate the determinants of female employment in manufacturing. The fractional nature of the dependent variable necessitates the use of the fractional logit model proposed by Papke and Wooldridge (1996).<sup>9</sup> As was discussed above, our approach is closely related to that of Fasih and Ghazalian (2015). However, we extend this previous work in two significant ways. First, exploiting the panel dimension of our data, we estimate our specification with fixed effect model in order to control for the various unobservable and time-invariant features of the firm that tend to be correlated with female employment. This is done as follows:



Let  $R_{i,t} \in [0,1]$  denote a fractional variable of female employment for firm  $i$  in year  $t$ , conditional on  $X_{it}$  and  $\alpha_i$ :

$$E(R_{it} | X_{it}, \alpha_i) = F(X_{it}\beta + \alpha_i), \quad (1)$$

where  $X_{it}$  are variables that vary across firms,  $\alpha_i$  is the fixed effect for firm  $i$ , and  $F(\cdot)$  is a cumulative distribution function (c.d.f.) that is assumed to be a logit c.d.f.<sup>10</sup> We estimate the parameters in Eq. (1) using the Quasi-Maximum Likelihood Estimator (QMLE), where the likelihood for an observation is specified as the Bernoulli likelihood:

$$L_i = [F(X_{it}\beta + \alpha_i)]^{R_{it}} [1 - F(X_{it}\beta + \alpha_i)]^{1-R_{it}}. \quad (2)$$

The QMLE of  $\beta$  and  $\alpha_i$  is consistent as long as the conditional expectation in Eq. (1) is specified correctly even if the Bernoulli specification in Eq. (2) is incorrect. The asymptotic variance-covariance matrix of the QMLE estimates is estimated by maintaining only first-moment assumptions, without any additional second moment assumptions. The exponents  $R_{it}$  and  $1 - R_{it}$  represent the fractions of females and males, respectively, in firm  $i$  at time  $t$ .

Second, we estimate the fractional logit model using the generalised estimating equation (GEE) to allow for correlations between observations, such that valid standard errors of the parameter estimates can be obtained, as was described by Liang and Zeger (1986). For our purposes, we focus on a particular correlation matrix that is well-suited for panel data applications with a few time periods. The GEE literature refers to it as an ‘exchangeable’ correlation pattern, where we act as if standardised errors have constant correlations (e.g., Papke and Wooldridge, 2008).

Thus, we provide estimates from three separate regressions: QMLE, QMLE controlling for firm fixed effects, and GEE. We have conducted Hausman specification tests for the random effect and fixed effect models. For all specifications, random effects models are rejected. The empirical test provides  $\chi^2$  test statistics of 26.09 ( $p = 0.004$ ) for female permanent workers, 41.68 ( $p = 0.001$ ) for female production workers, and 56.59 ( $p = 0.000$ ) for female non-production workers.<sup>11</sup> In all models, the standard errors are adjusted for clustering at the firm-level, but it is quite possible that the observations may be correlated within firms. We check this by running regression models without the cluster option, and present the results in Appendix Table A2. The robust standard errors are much smaller than the clustered standard errors, and a positive difference between the two types of errors would point to the presence of cluster correlation in our sample. It is also important to note that the clustered standard errors are larger because the number of clusters is smaller than our sample size. We

present the marginal effects of the estimated coefficients at the mean values of the explanatory variables ( $X_{it}$ ).

However, QMLE estimators may be inconsistent, since the asymptotic properties of fractional logit models require a balanced panel (Papke and Wooldridge, 2008; Wooldridge, 2010). We could potentially also treat the data as a pooled cross-section (with appropriate controls for time), but a Hausman test showed that this approach would give inefficient results. Nevertheless, for completion Section 5 also discusses the results when fixed effects are not included, but we note that this is not our preferred specification, because of the Hausman test results.

## 5. Estimation

This section discusses the estimation results of the specifications explained in the previous section. Table 3 shows the results for the model that explains the variation in the fraction of female full-time permanent workers. The results suggest that medium-sized and large firms tend to employ larger fractions of female permanent workers. This finding is consistent with our previous interpretation of the firm size variable, but runs counter to the results of Fasih and Ghazalian (2015), who find that, in MENA's manufacturing sector, full-time female workers prefer to work in smaller rather than larger firms. This could be due to factors such as more complex technologies and more unpleasant working conditions in larger firms.

Interestingly, firms with websites are found to employ a smaller fraction of women (6.3%). A similar result was found by Fasih and Ghazalian (2015) for manufacturing firms in the MENA region. One potential explanation is that firms in Bangladesh that adopt IT, such as using websites, tend to use more skilled labour, and therefore employ a greater proportion of men, who are more likely to be skilled than their female counterparts.<sup>12</sup>

There is also evidence suggesting that firms that export a large proportion of their output employ higher proportions of women. This is likely to be attributable to exports being very unskilled-labour intensive in Bangladesh. If women are more likely to be less educated and less skilled, they are more likely to accept lower wages and to find employment with export-intensive firms. This result is consistent with the findings of previous empirical studies (e.g., Bussmann, 2009; Gaddis and Pieters, 2012; Kabeer and Mahmud, 2004; Moghadam, 2005). Specifically, we find that manufacturing firms in Bangladesh that are engaged in exporting activities have 14% more female permanent workers than non-exporting firms.

Domestically-owned firms are found to employ higher proportions of women than either government-owned or foreign-owned firms. This result runs counter to the popularly-held belief that government-owned and foreign-owned firms tend to employ more women than domestic firms in developing countries (e.g., Curd *et al.*, 2007; Fasih and Ghazalian, 2015; Siegel *et al.*, 2014). It suggests that the Bangladeshi government may have a preference for employing men rather than women, and that foreign-owned firms are more likely to employ more skilled, better educated workers, which are more likely to be men. For government-owned firms, the impact of economic liberalisation may be biased towards skilled workers, which, in Bangladesh, are predominantly men.<sup>13</sup> In the case of foreign-owned firms, the findings might reflect the fact that foreign firms generally adopt skill-biased or labour-saving technologies in developing countries, which raises the relative demand for skilled labour, which is predominantly men in this case (see Feenstra and Hanson, 1995, for details). This result is consistent with the trends observed in middle-income economies within Latin America and East Asia (Robbins, 1996).

There is no evidence to suggest that firms owned by women employ greater proportions of women. This finding does not match our *a priori* expectations. If women constitute a very small proportion of firm ownership, this does not seem to contribute to a higher probability of female employment.<sup>14</sup> The coefficients on the industrial sector dummy variables suggest that women are less likely to be employed in some manufacturing industries than others, with work in the textiles/garment industry likely to be more common. For instance, women might prefer to stay away from industries that require night time or physically strenuous work and industries that make it difficult (due to location or working hours) for women to combine their responsibilities as primary caregivers in the family with paid employment.

The results from the alternative QMLE and GEE specifications that include firm fixed effects are presented in Columns (2) and (3) of Table 3. They are generally comparable to those presented in Column (1), though with a few differences. Specifically, the marginal effects of the medium and large firm variables on female employment rates remain positive, though they are smaller in magnitude and not statistically significant for the QMLE estimation (Column (2)). Similarly, the relationship between exporting firms and female employment rates does not hold after controlling for firm fixed effects in the QMLE estimation. These findings might suggest that there is a substantial degree of heterogeneity across manufacturing firms, resulting in lower marginal effects on female employment rates.

[Table 3 about here]

The results relating to the fraction of female full-time production workers are presented in Table 4, and are broadly consistent with those in Table 3. While fewer coefficient estimates are found

to be statistically significant in the model, which explains the fraction of female full-time non-production workers (Table 5), the results confirm that larger and domestically-owned firms employ larger proportions of women. This result is in contrast to the findings of Fasih and Ghazalian (2015) for the MENA region, where foreign-owned manufacturing firms are more likely to employ non-production female workers than domestic manufacturing firms. Furthermore, firms located in Chittagong, the commercial capital of Bangladesh, are found to employ a greater proportion of female non-production workers, which may reflect the structure of the manufacturing sector in this location, although this result is not statistically significantly different from zero in the QMLE specification that includes firm fixed effects (Column (2), Table 5).

[Tables 4 and 5 about here]

Overall, the results suggest that, as expected, unskilled-labour intensive industries hire relatively larger proportions of women. The labour market implications of these findings are that women in Bangladesh are likely to earn less than men and to work under worse conditions (see Kabeeer and Mahmud, 2004, for details). One possible way to begin to bridge this gap is to promote a greater equity in educational attainments between men and women. Indeed, while the proportion of children that finish school has increased significantly, and even though the gap between men and women has narrowed since the 1950s, progress is still slower in Bangladesh than in other developing countries (see Figure 1).

[Figure 1 about here]

We have conducted a number of robustness tests on the results obtained. The tests considered included: (a) dropping the industry and city dummy variables from the analysis; (b) focussing our analysis on the textiles and garments industry; (c) conducting the analysis separately for the city of Dhaka; and (d) re-running fractional logit models using firm observations collected in 2013.<sup>15</sup> Tables 6, 7, 8 and 9 present the results of the robustness check.

[Table 6 about here]

The marginal effects from estimating an alternative empirical specification that excludes industry and city dummy variables are displayed in Table 6. They are generally similar to the corresponding benchmark results presented in Tables 3, 4 and 5, but are mostly larger in magnitude (i.e., in absolute terms). As with Fasih and Ghazalian (2015), this appears to suggest that the inclusion of industry and city fixed effects would likely to encapsulate some variations between industries and cities, leading to lower estimates.<sup>16</sup> For example, the marginal effects still suggest that female workers (i.e., permanent, production and non-production workers) have more of a tendency to be employed in larger firms; however, the magnitudes of these effects have increased moderately relative to previous

estimates. The other notable finding is that the marginal effect of the manager's experience remains negative, but becomes statistically significant, unlike the benchmark results.

[Table 7 about here]

Table 7 presents the marginal effects obtained when we restrict ourselves to the textile/garments industry sample, since it covers the majority of the observations. In line with the benchmark results, the medium and large textile firms have a greater propensity to employ female workers (i.e., permanent, production and non-production workers). However, the magnitudes of these effects are considerably larger than those for the full sample. For example, the marginal effects of medium and larger firms are positively significant when the outcome variable is female permanent workers, sitting at about 61%–66% (Column (1)). The corresponding estimates are 63%–65% for female production workers and 31%–86% for female non-production workers (Columns (4) and (7)). The marginal effects from the alternative QMLE and GEE empirical specifications that include firm fixed effects are also reported in Table 7, and remain generally consistent with the benchmark results.

[Table 8 about here]

Table 8 presents the marginal effects obtained when restricting the sample to Dhaka city, the capital of Bangladesh.<sup>17</sup> The marginal effects are found to be qualitatively similar to the benchmark results, with the one notable change that the co-efficient for female workers becomes statistically insignificant for medium-sized firms (with one exception). It also decreases in magnitude, but remains positive in most cases. We also find that, unlike the benchmark estimates, foreign-owned firms do not have any statistically significant effect on the fractions of female full-time permanent and production workers.

[Table 9 about here]

Finally, we also check the robustness of the results by restricting the data set to data collected between April and September, 2013.<sup>18</sup> Note that this particular survey covers 1442 business establishments, of which more than 80% are manufacturing firms. After dropping missing values, we end up with 987 manufacturing firms. The marginal effects of the QMLE estimates are presented in Table 9. They are generally reminiscent of some earlier findings reported in Column (1) of Tables 3–5, but show relatively small effects on female employment. We also measure QMLE estimates that exclude industry and city dummy variables, and report the results in Columns (4)–(6). They are found to be comparable to the corresponding benchmark results reported in Table 6.

## **6. Conclusion and policy considerations**

In recent years, Bangladesh, a country characterised by the practice of extreme patriarchy, has exhibited an impressive rate of growth in the participation of women in employment, particularly manufacturing. However, the recent tragedy in Rana Plaza, together with some emerging academic evidence, suggests that women are generally employed in low-skilled and low-paid industries within the manufacturing sector. This study sheds light on the demand-side determinants of a greater female employment in such industries, which existing studies have largely neglected.

The findings presented suggest that firm size, whether medium or large, and firms' export-oriented activities, have an important impact on women's employment in the manufacturing sector in Bangladesh. However, the results indicate that foreign-owned firms have a lower propensity to employ female workers. Similarly, we find that manufacturing firms that have a website are also less likely to employ women. We interpret these findings as suggesting that firms that hire more women are relatively less skilled-labour intensive than firms that hire more men. These findings imply that women are more likely to obtain lower wages and to work in poorer conditions, such as sweatshops.

This underlying disparity between men and women stems from the fact that women have less access to education and skilling-opportunities than men. That is, our analysis of the demand-side determinants of female employment corroborates previous findings that suggest that supply-side constraints remain a major problem in Bangladesh. Therefore, the policy implications are that the promotion of equitable access to schooling for men and women should be continued, in the hope that this can eventually bridge the gap between labour market outcomes in the country. Social programs that promote the importance of educational attainments will also be crucial, particularly those geared at the household-level. Similarly, adult education programs that focus on skilling working-age women in major industrial areas, such as Dhaka and Chittagong, could also be useful. Finally, affirmative action programs within government firms, which were found to be less likely to hire women, could also help to begin promoting the necessary cultural changes. Indeed, our finding that larger firms hire more women is interpreted as suggesting that firms can respond to greater government scrutiny regarding equitable employment, meaning that there is no reason to expect that government-owned firms could not engage in similar practices and still remain profitable.

## Notes

[1] In general, GEE is an extension of the GLM method. However, unlike the GLM method, which is based on the maximum likelihood theory of independent observations, the GEE method is based on the quasi-likelihood theory, and no assumptions are made about the distribution of response observations. As far as the authors are aware, there is no test available for ascertaining whether one method should be preferred over the other.

[2] While the temporal gaps between time periods are not uniform, wave (time) effects are included in the model to account for any business cycle effects.

[3] As noted by Fakhri and Ghazalian (2015), sample attrition is typical in firm-level panel data, and can often lead to bias results if firms those are dropped-out are systematically different from those that remain in the sample. In order to test whether this is a problem here, we estimated a model on a balanced panel of 117 manufacturing firms, consisting of 346 observations. The results, which are available upon request, are qualitatively similar to those presented in this paper.

[4] The following quote from the 'Enterprise Survey' questionnaire defines permanent and non-production workers: 'Permanent, full-time employees are defined as all paid employees that are contracted for a term of one or more fiscal years with a guaranteed renewal of their employment contract and that work eight or more hours per day. Non-production workers refers to support staff, such as administrative, technical, IT, cleaning/security etc., who are not directly employed in the production line of the enterprise' (ES, 2007, p. 24).

[5] As an anonymous referee of this paper noted, female ownership may potentially be an endogenous variable. We simply verify this by using an instrument for whether the firm manager is female or not, based on the argument that female owners would be more likely to employ female manager. The relevant test gives strong credence to our use of the instrumental variable ( $p = 0.04$ ). However, the exogeneity of the female ownership variable is not rejected at any reasonable level of significance in all cases. The Hausman test yields test statistics of 1.44 ( $p = 0.23$ ) for female permanent workers; 2.49 ( $p = 0.16$ ) for female production workers; and 2.04 ( $p = 0.15$ ) for female non-production workers. When a dummy variable is included for female managers as an additional explanatory variable, it was statistically insignificant, further justifying its use as an instrument.

[6] Chittagong is home to Bangladesh's export-oriented manufacturing sector, meaning that women have a greater chance of participating in wage employment, reflecting manufacturing firms' demand for women's labour.

[7] As noted by Fakhri and Ghazalian (2015), the inclusion of industry and city dummy variables in a regression model does not produce identification problems (e.g., the incidental parameters problem) because the sample size is determined by the number of firms.

[8] The Variance Inflation Factor (VIF) measures the degree of multicollinearity among explanatory variables in a regression model. A higher VIF score (e.g., over 10) indicates multicollinearity problems.

[9] See Papke and Wooldridge (1996) for more details regarding this approach.

[10] The link test does not reject the logit specification. Furthermore, the use of the normal c.d.f. leads to similar results.

[11] We compute the Hausman test using the Stata module **xtreg**.

[12] As with Fakhri and Ghazalian (2015), the regressions also conducted with the addition of a binary variable that equals one for firms that use e-mail as a form of business communication and zero otherwise. The corresponding results are consistent with the benchmark regression.

[13] The explanation for this depends partly on public sector retrenchment. As a part of broader economic reform strategies, retrenchment in public firms has been taking place in Bangladesh since the 1990s, which is more likely to affect women than men because of the reduction in low-skilled and low-paid public sector jobs, many of which have traditionally been held by women.

[14] Unfortunately, we do not have any data on the proportion of firms owned by females to enable us to confirm this hypothesis.

[15] We would like to thank an anonymous referee for mentioning these robustness tests.

[16] Indeed, the likelihood ratio (LR) test results suggest that adding industry and city dummy variables to the model improves the fit of the model significantly compared to a model without these dummies, with the exception of female non-production workers. In the case of female permanent workers,  $\chi^2(6) = 16.19$  ( $p = 0.012$ ), while in the case of female production and non-production workers, the corresponding values are  $\chi^2(6) = 19.86$  ( $p = 0.003$ ) and  $\chi^2(6) = 8.37$  ( $p = 0.212$ ), respectively.

[17] We have not conducted a separate empirical analysis for Chittagong city, due to the limited number of observations.

[18] The 2013 data are the most recent available at the time of writing.

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Table 1. Summary statistics

Variable	Mean	Std. Dev	Min	Max
<b>Dependent variables</b>				
Female full-time permanent workers	0.357	0.273	0	0.923
Female full-time production workers	0.383	0.296	0	0.938
Female full-time non-production workers	0.191	0.237	0	1
<b>Firm-related variables</b>				
Small firm*	0.092	0.290	0	1
Medium firm	0.139	0.346	0	1
Large firm	0.769	0.422	0	1
Woman is an owner	0.413	0.493	0	1
Manager's experience	17.472	8.805	0	55
Firm age	2.847	0.626	0.693	5.176
Firm has website	0.660	0.474	0	1
Profit ratio	0.803	0.224	-1.203	1.000
Exporter	0.545	0.499	0	1
Government ownership	0.010	0.099	0	1
Private foreign ownership	0.036	0.187	0	1
Private domestic ownership*	0.964	0.187	0	1
Skilled production workers	0.504	0.415	0	1
Non-production workers	0.172	0.160	0	0.935
Food	0.102	0.304	0	1
Textiles/Garments*	0.634	0.483	0	1
Leather	0.059	0.237	0	1
Chemical	0.092	0.290	0	1
Electronics	0.063	0.243	0	1
Other manufacturing industry	0.050	0.217	0	1
Number of observations	303			

Notes: \* implies reference categories in the estimated equations.

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.

Table 2. Variance inflation factor (VIF)

	VIF	Tolerance (1/VIF)	VIF	Tolerance (1/VIF)
Medium firm	2.37	0.42	2.52	0.39
Large firm	3.14	0.32	3.84	0.26
Woman is an owner	1.16	0.86	1.24	0.8
Manager's experience	1.21	0.82	1.26	0.79
Firm age	1.20	0.83	1.26	0.8
Firm has website	1.17	0.85	1.26	0.79
Exporter	1.37	0.73	1.55	0.64
Government ownership	1.05	0.96	1.06	0.94
Private foreign ownership	1.06	0.94	1.17	0.85
Skilled production workers	1.11	0.90	1.17	0.86
Non-production workers	1.26	0.79	1.62	0.62
Profit ratio	1.05	0.95	1.06	0.94
Chittagong			1.26	0.79
Food			1.5	0.67
Leather			1.29	0.77
Chemical			1.65	0.61
Electronics			1.38	0.72
Other manufacturing industry			1.44	0.69
Mean	1.43		1.53	

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.

Table 3. Fraction of female full-time permanent workers, 2007–2013  
 Estimation of equation (1)

	Fractional logit <sup>a</sup>		
	(1)	(2)	(3)
	QMLE	QMLE Fixed effects	GEE Fixed effects
Medium firm	0.260* (0.136)	0.030 (0.094)	0.253* (0.135)
Large firm	0.222** (0.081)	0.041 (0.073)	0.218** (0.081)
Woman is an owner	0.018 (0.027)	0.003 (0.019)	0.018 (0.027)
Manager's experience	-0.002 (0.002)	-0.001 (0.001)	-0.002 (0.002)
Firm age	0.019 (0.019)	0.015 (0.017)	0.019 (0.019)
Firm has website	-0.0631* (0.027)	-0.0417* (0.019)	-0.0638* (0.027)
Exporter	0.135*** (0.027)	0.018 (0.027)	0.133*** (0.028)
Government ownership	-0.129*** (0.027)	-0.0787*** (0.020)	-0.129*** (0.027)
Private foreign ownership	-0.155* (0.066)	-0.104*** (0.014)	-0.154* (0.065)
Skilled production workers	-0.013 (0.109)	0.001 (0.084)	-0.011 (0.110)
Non-production workers	-0.153 (0.154)	-0.124 (0.089)	-0.156 (0.152)
Profit ratio	0.011 (0.064)	0.012 (0.040)	0.011 (0.064)
Chittagong	0.005 (0.028)	-0.101*** (0.021)	0.005 (0.028)
Food	-0.199*** (0.032)	-0.461*** (0.042)	-0.200*** (0.031)
Leather	-0.274*** (0.022)	0.0066 (0.060)	-0.275*** (0.022)
Chemical	-0.219*** (0.025)	-0.405*** (0.026)	-0.219*** (0.025)
Electronics	-0.163* (0.069)	-0.302*** (0.020)	-0.163* (0.069)
Other manufacturing industry	-0.305*** (0.026)	-0.260*** (0.012)	-0.305*** (0.026)
Time dummies	Yes	Yes	Yes
$\chi^2$ statistic	327.43	315.75	318.92
<i>p</i> -value	(0.000)	(0.000)	(0.000)
Number of observations	303	303	303

Notes: <sup>a</sup> Marginal effects. The marginal effect of discrete variables is given by the discrete change in the outcome for each observation and their average over the sample. Robust standard errors are in parentheses and are clustered at the firm-level. The  $\chi^2$  statistic is for testing the null hypothesis that all the coefficients are zero. The omitted categories are small firm, private domestic ownership, and Dhaka and textiles/garments industry. \*, \*\*, and \*\*\* indicate significance at 10%, 5%, and 1%, respectively.

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.

Table 4. Fraction of female full-time production workers, 2007–2013  
 Estimation of equation (1)

	Fractional logit <sup>a</sup>		
	(1)	(2)	(3)
	QMLE	QMLE Fixed effects	GEE Fixed effects
Medium firm	0.308* (0.154)	0.033 (0.107)	0.331* (0.154)
Large firm	0.263** (0.092)	0.032 (0.079)	0.275** (0.092)
Woman is an owner	0.017 (0.031)	0.002 (0.019)	0.018 (0.031)
Manager's experience	-0.002 (0.002)	-0.001 (0.001)	-0.002 (0.002)
Firm age	0.029 (0.022)	0.017 (0.018)	0.027 (0.022)
Firm has website	-0.0715* (0.031)	-0.0447* (0.021)	-0.0681* (0.031)
Exporter	0.150*** (0.031)	0.0187 (0.028)	0.155*** (0.031)
Government ownership	-0.158*** (0.043)	-0.0801*** (0.021)	-0.152*** (0.042)
Private foreign ownership	-0.117 (0.105)	-0.0819** (0.027)	-0.121 (0.105)
Skilled production workers	-0.073 (0.158)	-0.049 (0.093)	-0.081 (0.152)
Non-production workers	0.123 (0.131)	0.058 (0.102)	0.137 (0.129)
Profit ratio	-0.024 (0.072)	-0.009 (0.039)	-0.022 (0.073)
Chittagong	-0.0197 (0.032)	-0.107*** (0.022)	-0.021 (0.032)
Food	-0.238*** (0.036)	-0.442*** (0.043)	-0.233*** (0.037)
Leather	-0.308*** (0.025)	-0.035 (0.057)	-0.304*** (0.025)
Chemical	-0.253*** (0.031)	-0.387*** (0.027)	-0.250*** (0.030)
Electronics	-0.205** (0.076)	-0.244*** (0.019)	-0.203** (0.076)
Other manufacturing industry	-0.333*** (0.037)	-0.215*** (0.012)	-0.334*** (0.036)
Time dummies	Yes	Yes	Yes
$\chi^2$ statistic	243.61	292.60	252.67
$p$ -value	(0.000)	(0.000)	(0.000)
Number of observations	303	303	303

Notes: <sup>a</sup> Marginal effects. The marginal effect of discrete variables is given by the discrete change in the outcome for each observation and their average over the sample. Robust standard errors are in parentheses and are clustered at the firm-level. The  $\chi^2$  statistic is for testing the null hypothesis that all the coefficients are zero.

The omitted categories are small firm, private domestic ownership, and Dhaka and textiles/garments industry.

\*, \*\*, and \*\*\* indicate significance at 10%, 5%, and 1%, respectively.

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.

Table 5. Fraction of female full-time non-production workers, 2007–2013  
 Estimation of equation (1)

	Fractional logit <sup>a</sup>		
	(1)	(2)	(3)
	QMLE	QMLE Fixed effects	GEE Fixed effects
Medium firm	0.124 (0.101)	-0.003 (0.013)	0.123 (0.101)
Large firm	0.111** (0.043)	0.004 (0.022)	0.110** (0.044)
Woman is an owner	0.038 (0.023)	0.001 (0.007)	0.036 (0.024)
Manager's experience	-0.002 (0.001)	0.000 (0.000)	-0.003 (0.002)
Firm age	-0.008 (0.019)	0.012 (0.007)	-0.009 (0.020)
Firm has website	-0.003 (0.024)	0.004 (0.006)	-0.004 (0.026)
Exporter	0.026 (0.020)	0.0149 (0.009)	0.028 (0.021)
Government ownership	-0.021 (0.072)	-0.003 (0.013)	-0.023 (0.074)
Private foreign ownership	-0.0849* (0.042)	-0.0183* (0.008)	-0.0847* (0.041)
Skilled production workers	-0.112 (0.088)	-0.032 (0.025)	-0.11 (0.089)
Non-production workers	-0.175 (0.137)	-0.0431 (0.034)	-0.173 (0.139)
Profit ratio	0.089 (0.095)	0.032 (0.028)	0.087 (0.096)
Chittagong	0.0795** (0.029)	0.049 (0.033)	0.0796** (0.030)
Food	0.001 (0.044)	-0.112*** (0.020)	0.002 (0.045)
Leather	-0.126*** (0.018)	-0.0538*** (0.005)	-0.125*** (0.019)
Chemical	-0.040 (0.033)	-0.0850*** (0.012)	-0.038 (0.034)
Electronics	-0.010 (0.040)	-0.177*** (0.022)	-0.012 (0.039)
Other manufacturing industry	-0.145*** (0.014)	-0.115*** (0.011)	-0.146*** (0.015)
Time dummies	Yes	Yes	Yes
$\chi^2$ statistic	154.61	989.47	153.62
<i>p</i> -value	(0.000)	(0.000)	(0.000)
Number of observations	303	303	303

Notes: <sup>a</sup> Marginal effects. The marginal effect of discrete variables is given by the discrete change in the outcome for each observation and their average over the sample. Robust standard errors are in parentheses and are clustered at the firm-level. The  $\chi^2$  statistic is for testing the null hypothesis that all the coefficients are zero.

The omitted categories are small firm, private domestic ownership, and Dhaka and textiles/garments industry.

\*, \*\*, and \*\*\* indicate significance at 10%, 5%, and 1%, respectively.

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.



Table 6. Robustness check for female full-time workers without industry and city dummies

	Fractional logit <sup>a</sup>								
	Permanent workers			Production workers			Non-production workers		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	QMLE	QMLE	GEE	QMLE	QMLE	GEE	QMLE	QMLE	GEE
		Fixed effects	Fixed effects		Fixed effects	Fixed effects	Fixed effects	Fixed effects	Fixed effects
Medium firm	0.271** (0.135)	0.036 (0.098)	0.257* (0.132)	0.309** (0.146)	0.031 (0.103)	0.305** (0.146)	0.117 (0.113)	-0.001 (0.015)	0.144 (0.123)
Large firm	0.333*** (0.059)	0.038 (0.074)	0.324*** (0.059)	0.379*** (0.065)	0.024 (0.078)	0.377*** (0.065)	0.140*** (0.046)	0.000 (0.024)	0.149*** (0.047)
Woman is an owner	0.038 (0.030)	0.005 (0.019)	0.037 (0.029)	0.048 (0.034)	0.005 (0.018)	0.047 (0.034)	0.021 (0.024)	0.001 (0.007)	0.023 (0.023)
Manager's experience	-0.004** (0.002)	-0.001 (0.001)	-0.005*** (0.002)	-0.004** (0.002)	-0.001 (0.001)	-0.004** (0.002)	-0.003** (0.002)	-0.001* (0.000)	-0.003* (0.002)
Firm age	-0.013 (0.021)	0.016 (0.017)	-0.012 (0.020)	-0.010 (0.024)	0.018 (0.017)	-0.010 (0.024)	-0.007 (0.019)	0.012 (0.007)	-0.014 (0.018)
Firm has website	-0.037 (0.032)	-0.040** (0.019)	-0.040 (0.031)	-0.043 (0.036)	-0.045** (0.020)	-0.044 (0.036)	0.019 (0.027)	0.006 (0.006)	0.015 (0.027)
Exporter	0.172*** (0.031)	0.026 (0.026)	0.167*** (0.032)	0.189*** (0.036)	0.021 (0.026)	0.187*** (0.036)	0.029 (0.025)	0.016* (0.009)	0.025 (0.024)
Government ownership	-0.054 (0.051)	-0.074*** (0.024)	-0.062 (0.051)	-0.068 (0.080)	-0.078*** (0.018)	-0.072 (0.080)	-0.026 (0.070)	0.000 (0.013)	-0.024 (0.072)
Private foreign ownership	-0.172*** (0.065)	-0.107*** (0.012)	-0.170*** (0.064)	-0.144 (0.101)	-0.078*** (0.024)	-0.143 (0.101)	-0.071 (0.057)	-0.020*** (0.007)	-0.066 (0.054)
Skilled production workers	0.085 (0.140)	-0.005 (0.082)	0.076 (0.138)	0.034 (0.186)	-0.030 (0.089)	0.032 (0.186)	-0.075 (0.095)	-0.043* (0.025)	-0.054 (0.093)
Non-production workers	-0.445*** (0.160)	-0.093 (0.082)	-0.433*** (0.157)	-0.201 (0.140)	0.031 (0.089)	-0.199 (0.140)	-0.281* (0.149)	-0.018 (0.033)	-0.301** (0.150)
Profit ratio	0.009 (0.065)	0.020 (0.042)	0.011 (0.064)	-0.021 (0.071)	-0.010 (0.037)	-0.022 (0.070)	0.076 (0.077)	0.036 (0.030)	0.058 (0.066)
Time dummies	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Number of observations	303	303	303	303	303	303	303	303	303

Notes: <sup>a</sup> Marginal effects. The marginal effect of discrete variables is given by the discrete change in the outcome for each observation and their average over the sample. Robust standard errors are in parentheses and are clustered at the firm-level.

\*, \*\*, and \*\*\* indicate significance at 10%, 5%, and 1%, respectively. Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.

Table 7. Robustness check for female full-time workers in textiles/garments industry

	Fractional logit <sup>a</sup>								
	Permanent workers			Production workers			Non-production workers		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	QMLE	QMLE	GEE	QMLE	QMLE	GEE	QMLE	QMLE	GEE
		Fixed effects	Fixed effects		Fixed effects	Fixed effects		Fixed effects	Fixed effects
Medium firm	0.656*** (0.001)	0.647*** (0.008)	0.0001 (0.376)	0.629*** (0.006)	0.620*** (0.008)	0.491*** (0.026)	0.860*** (0.007)	0.890*** (0.006)	0.798*** (0.039)
Large firm	0.613*** (0.005)	0.594*** (0.007)	-0.155 (0.383)	0.647*** (0.007)	0.627*** (0.007)	0.513*** (0.036)	0.305*** (0.007)	0.249*** (0.015)	0.254*** (0.018)
Woman is an owner	-0.005 (0.032)	-0.055 (0.045)	0.008 (0.033)	-0.004 (0.036)	-0.055 (0.052)	0.008 (0.035)	0.005 (0.027)	-0.080*** (0.030)	0.048* (0.028)
Manager's experience	-0.003* (0.002)	-0.003 (0.003)	-0.004* (0.002)	-0.003 (0.002)	-0.003 (0.003)	-0.004* (0.002)	-0.004** (0.002)	-0.005** (0.002)	-0.004* (0.002)
Firm age	0.018 (0.024)	0.022 (0.044)	0.017 (0.029)	0.022 (0.027)	0.015 (0.050)	0.027 (0.026)	0.008 (0.024)	0.100*** (0.039)	-0.024 (0.025)
Firm has website	-0.121*** (0.035)	-0.120*** (0.042)	-0.112*** (0.034)	-0.133*** (0.038)	-0.127** (0.049)	-0.125*** (0.037)	-0.041 (0.038)	-0.035 (0.040)	-0.042 (0.041)
Exporter	0.144*** (0.037)	0.040 (0.065)	0.137*** (0.036)	0.151*** (0.042)	0.040 (0.074)	0.135*** (0.039)	0.040 (0.030)	0.095** (0.041)	0.025 (0.028)
Government ownership	-0.204*** (0.034)	-0.216** (0.090)	-0.221*** (0.041)	-0.241*** (0.051)	-0.280** (0.109)	-0.269*** (0.050)	-0.045 (0.107)	0.082 (0.104)	-0.053 (0.106)
Private foreign ownership	-0.279*** (0.100)	-0.408*** (0.050)	-0.279** (0.109)	-0.319*** (0.109)	-0.461*** (0.048)	-0.315*** (0.115)	-0.120 (0.079)	-0.130** (0.060)	-0.140** (0.067)
Skilled production workers	0.120 (0.132)	0.044 (0.237)	0.062 (0.133)	0.164 (0.147)	0.089 (0.263)	0.116 (0.145)	-0.063 (0.140)	-0.267 (0.168)	0.099 (0.141)
Non-production workers	-0.311* (0.182)	-0.170 (0.273)	-0.237 (0.184)	0.069 (0.216)	0.214 (0.326)	0.120 (0.202)	-0.225 (0.181)	-0.057 (0.251)	-0.360* (0.197)
Profit ratio	0.008 (0.068)	0.036 (0.088)	-0.007 (0.065)	-0.002 (0.075)	0.023 (0.102)	-0.017 (0.071)	0.041 (0.083)	0.112 (0.109)	-0.009 (0.053)

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Table 7. Continued

	Fractional logit <sup>a</sup>								
	Permanent workers			Production workers			Non-production workers		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	QMLE	QMLE	GEE	QMLE	QMLE	GEE	QMLE	QMLE	GEE
		Fixed effects	Fixed effects		Fixed effects	Fixed effects		Fixed effects	Fixed effects
Chittagong	-0.027	-0.315***	-0.030	-0.043	-0.344***	-0.044	0.086**	0.104	0.108***
	(0.031)	(0.061)	(0.033)	(0.033)	(0.069)	(0.034)	(0.037)	(0.096)	(0.035)
Time dummies	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Number of observations	192	192	192	192	192	192	192	192	192

Notes: <sup>a</sup> Marginal effects. The marginal effect of discrete variables is given by the discrete change in the outcome for each observation and their average over the sample. Robust standard errors are in parentheses and are clustered at the firm-level.

\*, \*\*, and \*\*\* indicate significance at 10%, 5%, and 1%, respectively.

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.

Table 8. Robustness check for female full-time workers in Dhaka city

	Fractional logit <sup>a</sup>								
	Permanent workers			Production workers			Non-production workers		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	QMLE	QMLE	GEE	QMLE	QMLE	GEE	QMLE	QMLE	GEE
		Fixed effects	Fixed effects		Fixed effects	Fixed effects	Fixed effects	Fixed effects	
Medium firm	0.237 (0.182)	-0.030 (0.069)	0.249 (0.182)	0.276 (0.205)	-0.043 (0.056)	0.337* (0.203)	0.124 (0.118)	0.007 (0.025)	0.145 (0.124)
Large firm	0.217** (0.097)	-0.015 (0.093)	0.224** (0.096)	0.250** (0.113)	-0.042 (0.103)	0.284*** (0.107)	0.142*** (0.037)	0.018 (0.013)	0.150*** (0.037)
Woman is an owner	0.026 (0.028)	0.010 (0.016)	0.026 (0.028)	0.028 (0.033)	0.012 (0.015)	0.026 (0.033)	0.005 (0.018)	-0.008* (0.004)	0.012 (0.019)
Manager's experience	-0.003 (0.002)	-0.002 (0.001)	-0.003 (0.002)	-0.003 (0.003)	-0.001 (0.001)	-0.002 (0.003)	-0.002 (0.001)	-0.000 (0.000)	-0.002 (0.001)
Firm age	0.015 (0.021)	0.009 (0.015)	0.014 (0.021)	0.027 (0.024)	0.013 (0.015)	0.025 (0.025)	-0.006 (0.019)	0.009 (0.006)	-0.011 (0.019)
Firm has website	-0.057** (0.028)	-0.042** (0.019)	-0.055** (0.028)	-0.080** (0.033)	-0.050** (0.020)	-0.068** (0.034)	0.022 (0.023)	0.007 (0.005)	0.020 (0.024)
Exporter	0.111*** (0.030)	0.031 (0.023)	0.112*** (0.030)	0.131*** (0.035)	0.037 (0.023)	0.134*** (0.034)	0.013 (0.021)	0.013 (0.009)	0.010 (0.020)
Government ownership	-0.134*** (0.028)	-0.052** (0.020)	-0.134*** (0.028)	-0.167*** (0.041)	-0.049** (0.020)	-0.159*** (0.040)	-0.021 (0.062)	0.003 (0.015)	-0.019 (0.066)
Private foreign ownership	-0.012 (0.072)	-0.017 (0.059)	-0.012 (0.071)	0.286 (0.266)	0.143 (0.242)	0.271 (0.248)	-0.068*** (0.021)	-0.007 (0.008)	-0.070*** (0.021)
Skilled production workers	0.034 (0.109)	0.037 (0.068)	0.031 (0.108)	0.024 (0.141)	0.031 (0.064)	0.006 (0.134)	-0.115 (0.089)	-0.037* (0.019)	-0.103 (0.089)
Non-production workers	-0.095 (0.157)	-0.076 (0.078)	-0.094 (0.159)	0.149 (0.127)	0.043 (0.079)	0.164 (0.124)	-0.030 (0.121)	0.009 (0.022)	-0.049 (0.127)
Profit ratio	-0.006 (0.066)	0.014 (0.035)	-0.008 (0.066)	-0.038 (0.076)	0.002 (0.031)	-0.043 (0.074)	0.018 (0.062)	0.009 (0.012)	0.013 (0.059)
Food	-0.207*** (0.040)	-0.290*** (0.025)	-0.205*** (0.040)	-0.254*** (0.046)	-0.265*** (0.025)	-0.245*** (0.047)	0.039 (0.063)	-0.067*** (0.009)	0.048 (0.063)
Leather	-0.266*** (0.024)	-0.097*** (0.006)	-0.265*** (0.024)	-0.304*** (0.027)	-0.082*** (0.006)	-0.296*** (0.027)	-0.104*** (0.021)	0.000 (0.008)	-0.100*** (0.023)

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Table 8. Continued

	Fractional logit <sup>a</sup>								
	Permanent workers			Production workers			Non-production workers		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	QMLE	QMLE	GEE	QMLE	QMLE	GEE	QMLE	QMLE	GEE
	Fixed effects	Fixed effects		Fixed effects	Fixed effects		Fixed effects	Fixed effects	
Chemical	-0.225*** (0.026)	-0.391*** (0.028)	-0.225*** (0.026)	-0.259*** (0.033)	-0.365*** (0.029)	-0.257*** (0.031)	-0.080*** (0.020)	-0.092*** (0.013)	-0.082*** (0.019)
Electronics	-0.232*** (0.035)	-0.253*** (0.020)	-0.229*** (0.036)	-0.284*** (0.035)	-0.199*** (0.018)	-0.269*** (0.041)	-0.008 (0.046)	-0.134*** (0.019)	-0.005 (0.047)
Other manufacturing industry	-0.288*** (0.032)	-0.224*** (0.013)	-0.287*** (0.032)	-0.321*** (0.043)	-0.176*** (0.013)	-0.323*** (0.042)	-0.128*** (0.017)	-0.112*** (0.012)	-0.121*** (0.018)
Time dummies	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Number of observations	234	234	234	234	234	234	234	234	234

Notes: <sup>a</sup> Marginal effects. The marginal effect of discrete variables is given by the discrete change in the outcome for each observation and their average over the sample.

Robust standard errors are in parentheses and are clustered at the firm-level.

\*, \*\*, and \*\*\* indicate significance at 10%, 5%, and 1%, respectively.

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.

Table 9. Robustness check for female full-time workers for the year 2013

	Fractional logit <sup>a</sup>					
	(1)	(2)	(3)	(4)	(5)	(6)
	Permanent workers QMLE	Production workers QMLE	Non-production workers QMLE	Permanent workers QMLE	Production workers QMLE	Non-production workers QMLE
Medium firm	0.027* (0.015)	0.034* (0.018)	-0.004 (0.015)	0.035** (0.016)	0.045** (0.019)	-0.004 (0.016)
Large firm	0.100*** (0.025)	0.129*** (0.031)	0.032 (0.023)	0.135*** (0.027)	0.164*** (0.032)	0.055** (0.025)
Woman is an owner	0.016 (0.014)	0.024 (0.017)	0.005 (0.013)	0.005 (0.013)	0.010 (0.016)	0.002 (0.014)
Manager's experience	0.001** (0.001)	0.001* (0.001)	0.000 (0.001)	0.001** (0.001)	0.001** (0.001)	0.000 (0.001)
Firm age	-0.051*** (0.007)	-0.063*** (0.009)	0.001 (0.008)	-0.059*** (0.008)	-0.072*** (0.010)	-0.000 (0.008)
Firm has website	0.031** (0.014)	0.041** (0.017)	0.006 (0.014)	0.029** (0.014)	0.038** (0.017)	0.009 (0.015)
Exporter	0.115*** (0.025)	0.128*** (0.030)	0.050** (0.022)	0.126*** (0.022)	0.141*** (0.025)	0.057*** (0.021)
Private foreign ownership	0.017 (0.024)	0.014 (0.034)	0.044 (0.035)	0.032 (0.033)	0.036 (0.043)	0.023 (0.030)
Skilled production workers	-0.080*** (0.026)	-0.083*** (0.032)	-0.029 (0.021)	-0.081*** (0.028)	-0.086** (0.034)	-0.030 (0.022)
Non-production workers	-0.134*** (0.033)	-0.023 (0.043)	-0.108*** (0.031)	-0.186*** (0.033)	-0.071* (0.042)	-0.094*** (0.029)
Profit ratio	0.023 (0.023)	0.012 (0.027)	0.050* (0.028)	0.010 (0.023)	0.000 (0.027)	0.033 (0.029)
Food	-0.081*** (0.009)	-0.085*** (0.012)	-0.051*** (0.010)			
Leather	-0.077*** (0.009)	-0.085*** (0.012)	-0.044*** (0.011)			
Chemical	-0.050*** (0.013)	-0.055*** (0.017)	0.023 (0.026)			

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Table 9. Continued

	Fractional logit <sup>a</sup>					
	(1)	(2)	(3)	(4)	(5)	(6)
	Permanent workers	Production workers	Non-production workers	Permanent workers	Production workers	Non-production workers
	QMLE	QMLE	QMLE	QMLE	QMLE	QMLE
Electronics	-0.075*** (0.024)	-0.081** (0.033)	-0.030 (0.026)			
Other manufacturing industry	-0.061*** (0.012)	-0.064*** (0.015)	-0.044*** (0.014)			
Chittagong	0.053*** (0.016)	0.076*** (0.021)	-0.039*** (0.009)			
Number of observations	987	987	987	987	987	987

Notes: <sup>a</sup> Marginal effects. The marginal effect of discrete variables is given by the discrete change in the outcome for each observation and their average over the sample. Robust standard errors are in parentheses and are clustered at the firm-level.

\*, \*\*, and \*\*\* indicate significance at 10%, 5%, and 1%, respectively.

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.

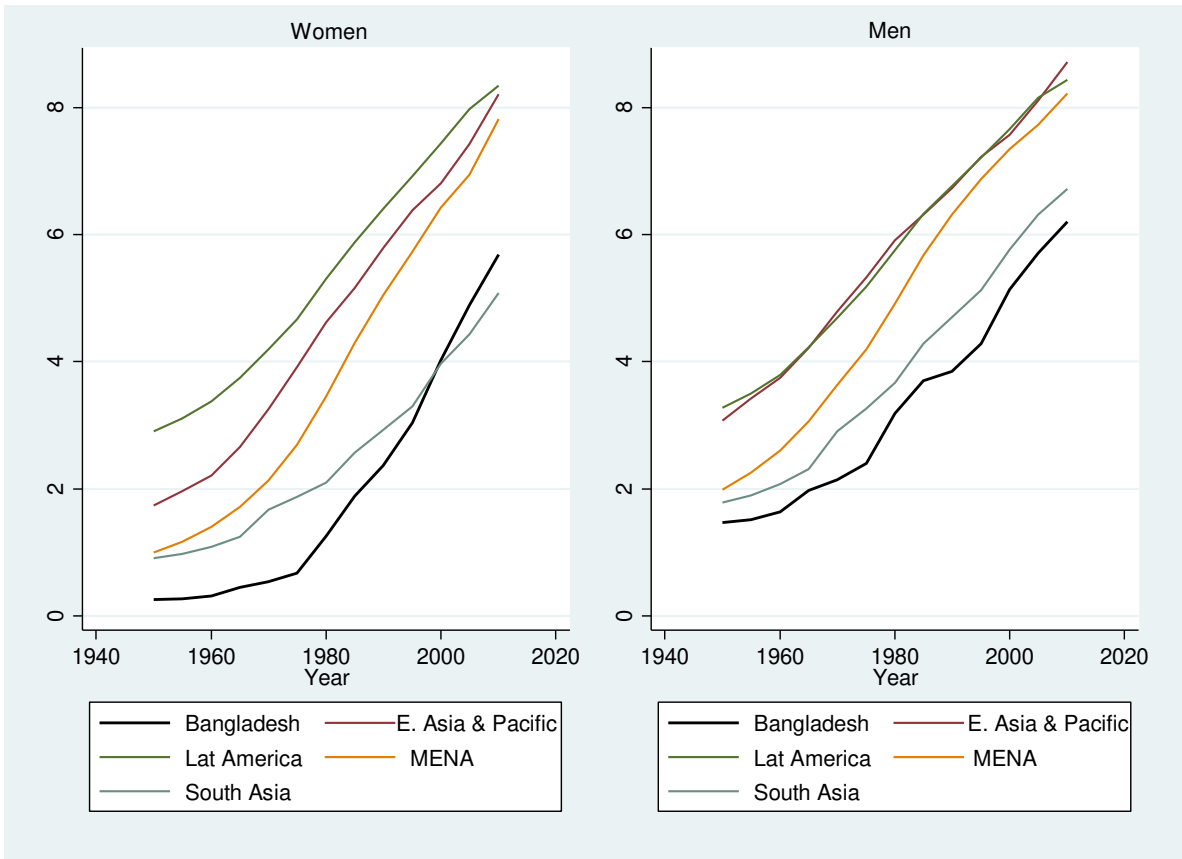


Figure 1. Years of schooling by gender, 1950–2010.  
 Source: Authors' calculation based on data from Barro and Lee (2013).



## Appendix

Table A1. List of variable names and descriptions

Variable name	Description
<b>Dependent variables</b>	
Female full-time permanent workers	Fraction in total full-time permanent workers
Female full-time production workers	Fraction in total full-time production workers
Female full-time non-production workers	Fraction in total full-time non-production workers
<b>Firm-related variables</b>	
Small firm	1 firms with employees 19 or less
Medium firm	1 firms with between 20 and 99 employees
Large firm	1 firms with greater than or equal to 100 employees
Woman is an owner	1 firm owner is a woman
Manager's experience	Manager's experience in years
Firm age	log of firm age since operation
Firm has website	1 firms use of own internet website
Profit ratio	Ratio of total profit (total sales-total costs) to total sales
Exporter	1 firms with more than 10% annual sales derived from direct export
Government ownership	1 firms that are subsidiary of government-owned firms
Private foreign ownership	1 firms with more than 10% foreign ownership
Private domestic ownership	1 firms with less than 10% foreign ownership
Skilled production workers	Fraction in total full-time production workers
Non-production workers	Fraction in total employment
Food	1 if industry category is food
Textiles/Garments	1 if industry category is textiles/garments
Leather	1 if industry category is leather
Chemical	1 if industry category is chemical
Electronics	1 if industry category is electronics
Other manufacturing industry	1 if industry category is others

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.

Table A2. Robustness check for female full-time workers without clustered standard errors

	Fractional logit <sup>a</sup>								
	Permanent workers			Production workers			Non-production workers		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	QMLE	QMLE	GEE	QMLE	QMLE	GEE	QMLE	QMLE	GEE
		Fixed effects	Fixed effects		Fixed effects	Fixed effects		Fixed effects	Fixed effects
Medium firm	0.260*	0.030	0.253*	0.308**	0.033	0.331**	0.124	-0.003	0.165
	(0.134)	(0.069)	(0.135)	(0.147)	(0.078)	(0.154)	(0.105)	(0.010)	(0.116)
Large firm	0.222***	0.041	0.218***	0.263***	0.032	0.275***	0.111**	0.004	0.128***
	(0.077)	(0.055)	(0.081)	(0.088)	(0.059)	(0.092)	(0.046)	(0.017)	(0.044)
Woman is an owner	0.018	0.003	0.018	0.017	0.002	0.018	0.038	0.001	0.047**
	(0.025)	(0.015)	(0.027)	(0.030)	(0.016)	(0.031)	(0.024)	(0.005)	(0.023)
Manager's experience	-0.002	-0.001	-0.002	-0.002	-0.001	-0.002	-0.002	-0.000	-0.002*
	(0.002)	(0.001)	(0.002)	(0.002)	(0.001)	(0.002)	(0.001)	(0.000)	(0.001)
Firm age	0.018	0.015	0.019	0.029	0.017	0.027	-0.008	0.012*	-0.018
	(0.020)	(0.013)	(0.019)	(0.023)	(0.014)	(0.022)	(0.020)	(0.006)	(0.018)
Firm has website	-0.063**	-0.042**	-0.064**	-0.071**	-0.045**	-0.068**	-0.002	0.004	-0.007
	(0.028)	(0.017)	(0.027)	(0.032)	(0.019)	(0.031)	(0.024)	(0.005)	(0.024)
Exporter	0.135***	0.017	0.133***	0.150***	0.019	0.155***	0.026	0.015*	0.024
	(0.028)	(0.020)	(0.028)	(0.033)	(0.021)	(0.031)	(0.020)	(0.008)	(0.019)
Government ownership	-0.129***	-0.079***	-0.129***	-0.158***	-0.080***	-0.152***	-0.021	-0.003	-0.013
	(0.027)	(0.015)	(0.027)	(0.040)	(0.016)	(0.042)	(0.075)	(0.013)	(0.081)
Private foreign ownership	-0.155**	-0.104***	-0.154**	-0.117	-0.082***	-0.121	-0.085**	-0.018**	-0.087**
	(0.064)	(0.011)	(0.065)	(0.103)	(0.020)	(0.105)	(0.038)	(0.007)	(0.040)
Skilled production workers	-0.013	0.001	-0.011	-0.073	-0.049	-0.081	-0.112	-0.032	-0.094
	(0.114)	(0.070)	(0.110)	(0.159)	(0.076)	(0.152)	(0.090)	(0.023)	(0.087)
Non-production workers	-0.153	-0.124*	-0.156	0.123	0.058	0.137	-0.175	-0.043	-0.195
	(0.148)	(0.072)	(0.152)	(0.132)	(0.081)	(0.129)	(0.135)	(0.029)	(0.142)
Profit ratio	0.011	0.012	0.011	-0.024	-0.009	-0.022	0.089	0.032	0.066
	(0.062)	(0.032)	(0.064)	(0.069)	(0.032)	(0.073)	(0.093)	(0.024)	(0.081)
Chittagong	0.005	-0.101	0.005	-0.020	-0.107*	-0.021	0.080**	0.049	0.078***
	(0.030)	(0.067)	(0.028)	(0.034)	(0.057)	(0.031)	(0.034)	(0.036)	(0.028)
Food	-0.199***	-0.461***	-0.200***	-0.238***	-0.442***	-0.233***	0.001	-0.112***	0.012
	(0.033)	(0.045)	(0.031)	(0.036)	(0.045)	(0.037)	(0.051)	(0.028)	(0.043)

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Table A2. Continued

	Fractional logit <sup>a</sup>								
	Permanent workers			Production workers			Non-production workers		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	QMLE	QMLE	GEE	QMLE	QMLE	GEE	QMLE	QMLE	GEE
		Fixed effects	Fixed effects		Fixed effects	Fixed effects	Fixed effects	Fixed effects	Fixed effects
Leather	-0.274*** (0.025)	0.007 (0.048)	-0.275*** (0.022)	-0.308*** (0.029)	-0.035 (0.043)	-0.304*** (0.025)	-0.126*** (0.018)	-0.054*** (0.014)	-0.118*** (0.023)
Chemical	-0.219*** (0.033)	-0.405*** (0.034)	-0.219*** (0.025)	-0.253*** (0.037)	-0.387*** (0.034)	-0.250*** (0.030)	-0.040 (0.040)	-0.085*** (0.017)	-0.046 (0.032)
Electronics	-0.163*** (0.058)	-0.302*** (0.030)	-0.163** (0.069)	-0.205*** (0.065)	-0.244*** (0.027)	-0.203*** (0.076)	-0.010 (0.041)	-0.177*** (0.041)	-0.004 (0.042)
Other manufacturing industry	-0.305*** (0.028)	-0.260*** (0.016)	-0.305*** (0.026)	-0.333*** (0.037)	-0.215*** (0.014)	-0.334*** (0.036)	-0.145*** (0.015)	-0.115*** (0.022)	-0.138*** (0.015)
Time dummies	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Number of observations	303	303	303	303	303	303	303	303	303

Notes: <sup>a</sup> Marginal effects. The marginal effect of discrete variables is given by the discrete change in the outcome for each observation and their average over the sample. Robust standard errors are in parentheses.

\*, \*\*, and \*\*\* indicate significance at 10%, 5%, and 1%, respectively.

Source: Authors' calculation from the Enterprise Survey data sets for 2007, 2011 and 2013.