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WHO PAYS FOR AND WHO BENEFITS FROM MINIMUM WAGE INCREASES? EVIDENCE FROM ISRAELI TAX DATA ON BUSINESS OWNERS AND WORKERS

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ABSTRACT

A key goal of a higher minimum wage is income redistribution towards low-income families. Existing research on the minimum wage focuses on the impact on affected workers, but is silent on the incomes of the owners of businesses who pay for a higher minimum wage. Higher minimum wages will do more to redistribute income if the owners of businesses who pay the minimum wage are nearer the top of the income distribution, and vice versa. We study evidence on the incidence of the minimum wage on the incomes of business owners using a unique administrative dataset on the universe of tax records for Israel, in the period surrounding a large minimum wage increase. We find that the minimum wage hike reduced profits of companies, with minimum-wage intensive companies bearing the bulk of the cost and adjusting their workforces more aggressively. Notably, profits declined more for lower-income business owners. Moreover, owners of businesses with higher shares of minimum-wage workers ranked at the bottom of the income distribution of business owners. In addition, spouses of business owners earn less than the owners while minimum wage.

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1. Introduction

At its core, the minimum wage is a redistributive policy, meant to increase incomes of low-income families. In the U.S. context, Senator Edward Kennedy, a perennial sponsor of minimum wage increases, argued that "The minimum wage was one of the first – and is still one of the best – anti-poverty programs we have" (Clymer, 1999, p. 449).

Existing research on the minimum wage focuses on the impact on affected workers. Not surprisingly, minimum wages raise wages of affected workers. A good deal of evidence concludes that higher minimum wages reduce employment of the least-skilled (Neumark and Shirley, 2021), but this conclusion is contested, with some recent evidence to the contrary (e.g., Cengiz et al., 2019); and recent evidence indicates that far more workers may experience wage gains than job losses, in part because estimated job losses are small (Harasztosi and Lindner, 2019). In terms of redistribution, researchers have focused on whether minimum wages reduce poverty or "near-poverty" (being below one-half of the poverty line). Most research tends to find no statistically significant evidence of poverty reductions (e.g., Sabia and Nielsen, 2015), although the point estimates tend to point in this direction (see, e.g., Neumark, 2016; Dube, 2018), and one recent study finds evidence of substantial poverty reductions (Dube, 2019).

Another potentially important dimension of the relationship between the distributional effects of minimum wages and the impact of minimum wages is not who *benefits* from the minimum wage, but who *pays* for it. However, the research literature is silent on the incomes of the owners of businesses who pay for a higher minimum wage. Higher minimum wages will do more to redistribute income if the owners of businesses who pay the higher minimum wage are nearer the top of the income distribution. Conversely, if minimum wage employers have relatively low incomes, the redistributional effects are weakened, and – on this dimension at least – minimum wages would be less effective at redistributing income from high-income individuals or families to low-income individuals or families.¹

Casual evidence on the distributional effects of the incidence of the minimum wage on businesses

¹ In this regard, the minimum wage could be quite different from redistributional policies financed by taxes – such as the U.S. Earned Income Tax Credit – which, by construction, redistribute income *from* those who pay the most taxes. How progressive this financing of the redistribution is depends, of course, on the progressivity of the tax system.

appears to point in a number of possible directions. There are likely many relatively small, lower-income business owners ("mom and pop" shops) who use low-skill workers and hence will have to pay for higher minimum wages. But there are also some large corporations (e.g., Wal-Mart) that also pay relatively low wages. On the other hand, there are clearly many higher-income business owners, investors, and salaried workers who will not pay directly for a higher minimum wage, and it seems likely that this is true of many of those who have had the highest income growth in recent years.²

Researchers have lacked the data to study the distributional effects of the incidence of the minimum wage on business owners. In this paper, we study this question using matched administrative data from the universe of tax records for workers, firms, and firm owners in Israel. We study a large minimum wage increase that was an exogenous event driven by political events, to provide evidence on the incidence of the costs of higher minimum wages with respect to the family incomes of business owners, and the family income distribution more generally. We believe this is the first evidence on the incomes of business owners who pay for higher minimum wages and their position in the income distribution.

Distinguishing companies by the fraction of full-year workers earning at or below the new minimum wage prior to its increase (which we denote *FMW*), we find that the minimum wage hike reduced profits of companies, with minimum-wage intensive companies bearing the bulk of the cost and adjusting their workforces more aggressively. Notably, profits declined more for lower-income business owners. Moreover, owners of businesses with higher shares of minimum-wage workers ranked at the bottom of the income distribution of business owners. In addition, spouses of business owners earn less than the owners while minimum wage workers (i.e., those working at companies with high *FMW*) have higher earning spouses, further reducing the redistributive effect of the minimum wage.

2. Minimum Wages in Israel

The first minimum wage in Israel resulted from a 1972 collective labor agreement setting the minimum wage as a percentage of the average wage in the economy. The minimum wage is set in hourly and monthly terms, based on full-time work. Since 1987, the minimum wage level is set by legislation,

² In the United States, the surge in top incomes has been mostly associated with capital income (Piketty et al., 2018).

initially constrained to no less than 39% of the average wage (based on full-time and part-time jobs in the entire economy), which rose to 45% in April 1988 and 47.5% in April 1997. In setting the nominal minimum wage, this mandated percentage was used as guidance rather than for consistent and explicit indexation.

The minimum wage "event" we study is the set of increases that occurred from 2006 through 2008. As shown in Figure 1, there were no increases in the nominal minimum wage in the few years preceding 2006. There were four increments in the 2006-2008 period, leading to a sizable 15.4% increase in the nominal minimum wage.³ There were then no changes to the nominal minimum wage from 2008 until 2011. In addition to the minimum wage history in this period providing stable pre- and post-periods, the 2006-2008 increases were large and plausibly exogenous (as discussed below).⁴ The 2006-2008 minimum wage increases led to a large increase in the minimum wage relative to average wages in this period. In the three periods 2003-2005 (pre-period), 2006-2008 (treatment period), and 2009-2010 (post-period), the real changes in the minimum wage were -1.3%, +8.3%, and -6.4%, respectively, while the corresponding changes in the real average wage were +5.6%, +1.1%, and -3.4%.⁵

The 2006-2008 minimum wage increases resulted from a lengthy campaign led by the Labor Party in the elections for the 17th Knesset (Israel's parliament). After the 2006 elections, the Labor Party became a key party in the governing coalition. Increasing the minimum wage and supporting "fair" earnings was one of Labor's pre-election commitments.⁶ There were no substantive changes in labor market

³ The nominal increase is small relative to some increases in the U.S. minimum wage. For example, three increases from 2007-2009 raised the U.S. federal minimum wage by a cumulative 41% (nominal). However, U.S. minimum wage increases are much less frequent, with increases, in recent decades, only in 1990-91, 1996-97, and 2007-2009. Moreover, most of the U.S. evidence focuses on state minimum wage increases, which are often much smaller, with change of a few percent very common. For example, see the state minimum wage database at:

http://www.economics.uci.edu/~dneumark/MW_LW%20dataset%20updated%20through%202019%20-%201-01-20%20Update.csv.

⁴ There were frequent increases beginning in 2011. Our data only extend through 2016, but as Figure 1 shows, even a couple more years of data would not give us a post-period with a stable nominal minimum wage with which to study these later increases in the minimum wage.

⁵ The real minimum wage and the real average wage is in NIS per month (deflated by the CPI), based on full-time work. The values measure changes within periods; for example the pre-period value is the change in the CPI deflated minimum wage between 2003Q1 and 2006Q1.

⁶ The Labor Party's pre-election platform stated: "The challenge facing the Israeli society is to accelerate the processes of economic development and growth, and to ensure fair distribution to the public... The government, led

developments in the period leading up to the election; Figure 2 shows the time-series of the unemployment and participation rates and the output gap. Moreover, Google trends data show that the minimum wage did not get much attention prior to the election or implementation of the increases (see Figure A1 in the online appendix). Hence, the increases in the minimum wage in this period can be viewed as stemming from political developments exogenous to labor market changes, and unlikely to have generated substantial changes in behavior prior to implementation. Nonetheless, in our analysis we are careful to compare developments in businesses more affected vs. businesses less affected by the minimum wage increase, even within industry, to net out any aggregate or sector-specific influences, and to rule out pre-trends associated with subsequent minimum impacts.

3. Research on the Effects of Minimum Wages on Firms

The literature on the effects of minimum wages on labor markets is voluminous, although most of it focuses on employment effects. Reviews covering employment, but also outcomes including poverty, the distribution of wages and income, skill accumulation, and prices, are provided in the books by Card and Krueger (1995), Neumark and Wascher (2008), and Belman and Wolfson (2014).

Although we know of no work that studies the incidence of minimum wages on business owners who might pay for higher minimum wages, and their position in the income distribution, the very limited literature on the effects of minimum wages on firms might provide some clues as to how minimum wages affect business income or profits, and for which businesses.

Card and Krueger (1995) do an event study of the effects of the 1990-1991 U.S. federal minimum wage increases and another proposed increase – and news stories related to them – on the daily excess returns of stock prices. The results of these analyses do not provide a clear indication that profits fall, although the authors note that the news events studied may not have provided much relevant information to market participants. This research applies to publicly-held firms, and does not provide direct evidence on the incomes of those affected by paying for higher minimum wages. Stocks are on average held by high-

by the Labor Party, will set a goal of reducing unemployment, raising minimum wage, effectively enforcing labor laws, reducing social gaps in income, education and housing quality."

income people, but they are also held by pension funds.⁷ Our analysis also faces the challenge of not being able to interpret data on the effects of minimum wages on the profits of publicly-held firms in terms of the incomes of owners, but we have data on both publicly-held and non-publicly-held firms.

Draca et al. (2011) provide more compelling evidence on the effects on firms, from an analysis of the 1999 introduction of the minimum wage in the United Kingdom, using firm-level profit data. They use pre-1999 information on the distribution of wages to measure differences in how firms are impacted, and find significant reductions in profitability. One of their data sets covers a wide range of firms, including smaller and medium-sized firms (and of course many firms not listed on the stock market), although some small firms are excluded.⁸ The negative effect of the minimum wage on profits is more evident for firms with more market power, which the authors argue is consistent with more competitive firms being more likely to pass on minimum wage increases to prices. However, this evidence does not map into how effects are distributed across the income distribution of business owners.⁹

Research on impact on small businesses is potentially informative about the distribution of minimum wage impacts across the income distribution of business owners, because the incomes of small business owners are likely to be considerably lower than those of other business owners. There appears to be a presumption that small businesses may be hit hardest by minimum wage increases.¹⁰ However, the empirical evidence is not so clear. The existing studies – Draca et al.'s (2011) analysis of residential care homes (which tend to be small), U.K. survey evidence in Mason et al. (2006), and Orazem and Mattila's

⁷ In the United States, direct stock ownership is concentrated in the top 1% of the wealth distribution (53.4% for 2016, vs. 40% in the next 9% and 6.8% in the bottom 90%), while pension accounts are more equitably distributed, but still skewed to higher-wealth households (13.7%, 51.2%, and 35.2%); see Wolff (2017, Table 10).

⁸ A related analysis of data on U.K. residential care homes – a very low-wage sector – finds corroborating evidence. Machin and Wilson (2004) study the effects of the introduction of the U.K. minimum wage on closures of residential care homes, based on their own survey. They find some evidence of employment declines, but do not detect more closures among homes more impacted by the minimum wage.

⁹ Belman and Wolfson (2014) review five additional studies of effects of minimum wages on profitability (or exits or failures). Waltman et al. (1998) and Orazem and Mattila (2002) report conflicting evidence on firms exits or failures ¹⁰ For example, the U.S. Chamber of Commerce, in criticizing a proposed New Orleans living wage law in 2002, argued that "the mandated wage increases … will hit small businesses the hardest"

⁽https://www.uschamber.com/press-release/us-chamber-challenges-living-wage-law). And a website published by the Employment Policies Institute includes videos of small business owners claiming difficulties in adjusting to high minimum wages (https://www.facesof15.com/). U.S. minimum wage laws have long had a minimum sales or revenue threshold (https://www.dol.gov/whd/minwage/q-a.htm). Similarly, in Korea a large minimum wage hike in 2018 was accompanied by compensation to firms employing less than 30 workers (OECD, 2018).

(2002) study of firms in services and retail newly covered by Iowa's minimum wage law – do not ask whether minimum wage effects fell more heavily on small businesses. Luca and Luca (2019) use Yelp data to study the effects of minimum wages on restaurant closures. They find that more marginal restaurants (with lower ratings) disproportionately exit because of minimum wage increases, with no impact for the top-rated restaurants.

An exception is Chava et al. (2019), who use the National Establishment Time Series (NETS) to estimate the effects of minimum wage increases on establishment credit (*Paydex*) scores. They find that federal minimum wage increases in states where the federal minimum wage binds reduced credit scores and increased firm exit more for small establishments (below median sales in the 4-digit NAICS code). However, one can only draw an indirect inference from this evidence that lower-*income* business owners are more adversely affected by minimum wages.

4. Data

Sources and key variables

Our empirical analysis relies on administrative records of the Israeli Tax Authority. We match data from three sources – company records, employee data, and business owners – and measure family income for business owners and workers using these data. Our key variables are listed and defined in Table 1.

The first data source is annual company tax records covering 2003-2010, which provide data on each company's characteristics and indicators of business activity. The company files are used to construct annual profit measures, or corporate income from all sources, including reimbursement of the owner's salary. We use the inverse hyperbolic sine (IHS) of profits to accommodate zeroes and negative numbers (Ravallion, 2017).¹¹ We measure the profit rate as corporate income divided by sales. Finally, we use the Central Bureau of Statistics (CBS) classification of economic sectors to define 23 sector dummy variables.

Our second data source is tax records of all employees for 2003-2010. The employee files provide data on each worker, including employment type, earnings, and number of months the employee worked at

¹¹ The interpretation of estimated coefficients of right-hand side variables when the dependent variable is expressed in terms of the IHS are nearly identical to the interpretation when the dependent variable is expressed in natural logs.

the same job. Our tax records cover employees for whom pay is reported to tax authorities in Israel (i.e., for companies situated in Israel), and temporary employment abroad that is included in the payroll of an Israeli company.¹² For each company, we match employees based on unique personal and company identifiers. We created the database at the employee level, matching workers to firms over time, and then calculated employee-related aggregates for each firm, including total employment.

Our research does not cover self-employed workers or employees related to them (in unincorporated businesses). In 2005, there were approximately 287,000 self-employed workers, representing 11.5% of employment, and 61,460 unincorporated businesses that employed 126,688 workers. The majority of these unincorporated businesses were very small, averaging around 2 employees per establishment.¹³ The minimum wage does not bind on wages or earnings of the self-employed themselves. And adding profits of unincorporated business is problematic since such businesses report income to the tax authority in a different manner than companies do; this, we study only incorporated companies.¹⁴

We use the data on employees to construct a critical variable for our analysis: the fraction affected (*FMW*), which is the proportion of workers at a company earning below or equal to the minimum wage. We define *FMW* as the fraction of full-year employees working at their main job, in 2005 (the last year of the pre-treatment period), who were paid less than or equal to the monthly minimum wage that prevailed in the post-hike period (2009-2010).¹⁵ The data do not distinguish part-time vs. full-time workers, and earnings are reported monthly, not hourly. Thus, there is the potential for some measurement error in *FMW*, which would presumably attenuate our estimated effects of *FMW*.¹⁶ However, we have taken a number of

¹² However, people who work abroad and their "center of life" is not in Israel (based on a given set of criteria such as presence in Israel, family location, etc.) are not liable for taxes in Israel, and hence are not included in the tax files. ¹³ This is based on CBS data (https://old.cbs.gov.il/publications18/1673/pdf/t02_01.pdf); the figures are for 2010, as prior data regarding the self-employed are less reliable. There is a small number of larger unincorporated businesses of this kind (with tax records tying workers to the individuals who own the companies, rather than company identifiers).

¹⁴ However, we compared monthly earnings of employees in unincorporated businesses to earnings in companies; excluding extreme values, earnings appear to be similarly distributed, suggesting that our results could be applicable to unincorporated businesses as well. See Figure A2 in the online appendix.

¹⁵ The main job is the one that the worker chooses for the application of tax credits and benefits. In general, if an individual works in more than one job, defining the higher-paying job as the main job minimizes tax payments. ¹⁶ Data from the 2005 Labor Force Survey (LFS) indicate that there is a 0.26 correlation between full-time work and employment above 10 months (the relevant variable available in the LFS), suggesting, as we would expect, that full-

steps to mitigate this measurement error, and to document why we think it is second order. We base *FMW* on full-year employees working at their main job because when we compared the Tax Authority earnings reports with those of the Central Bureau of Statistics (CBS), by industry, we found that the earnings of employees that worked full-year at the main job were closest to the wages that CBS reports; in other words, although not surprising, earnings in the tax data are much more similar to those of full-year workers. Moreover, we compared average wages and *FMW* derived from the tax data we use to those reported in the Income Survey (the equivalent of the U.S. Current Population Survey), by economic sector. Monthly wages and *FMW* rates are very similar by sector (see Figures A3 and A4 in the online appendix).

The third data source is a registry of company owners for 2003-2010, which includes ownership structure for each company and allows the identification of individual business owners.¹⁷ We identify owners (and their spouses, when present) of the companies in the sample for the year 2005. This is the most recent year before the minimum wage hike, and the year for which *FMW* was computed. Of the 57,520 companies in that year, we could identify data on 47,275 sole owners.¹⁸

We measure family incomes for both business owners and workers. We assembled data on owners' salary from the company, and from other sources, age, marital status, and spouse earnings. We also measure the percentile of each business owner's and worker's income in the overall distribution of income.¹⁹

year workers are more likely to work full-time. By basing *FMW* on full-year employees at their primary job (if applicable), we minimize the role of part-time workers who could create some measurement issues for *FMW*. ¹⁷ Some of the companies are owned by other companies or foreign residents.

¹⁸ This is out of a total of 63,674 owners, because of foreign ownership, firms that are owned by other local companies, and co-ownership. There were 2,010 companies that were fully owned by foreigners, and 2,414 that were fully owned by other companies. According to tax records, 60.5% of companies have a single owner, 29.1% have two owners, and the remainder have three or more owners.

¹⁹ We study labor income rather than asset income. One problem with asset income is that wealthy individuals can hold large companies, which could post significant losses in given year (in the form of undistributed profits), resulting in low income ranks that would be misleading. Another potential measurement issue with business income is that firm owners could take money out of firms not just via profits but via expenses on depreciation. Depreciation expenditures in Israel follow a schedule of rates in the Income Tax Ordinance (Section 21). The depreciation rate is linear over the year and starts from the operation date of the property. The rate depends on the type of the property: e.g., 7% for machines and general equipment, 6% for furniture, and 1-2% for real estate. The low rates and limited period imply that depreciation is not an effective "tax avoidance" instrument. Regardless, "income" from this source would likely only make our results on incidence stronger since depreciation would depress measured profits of more-profitable

There are two age-related issues regarding the measurement of family income. First, teens could not be matched to the parents, as tax records do not provide information beyond number of children in the household. Thus, family income includes incomes of the spouses, and teenagers' labor incomes are not included. However, teen employment is relatively low in Israel; the participation rate of youths (aged 15-17) is 9.6%, of whom only 1.5% are not in school. Correspondingly, according to the Income Survey only 7.2% of households had more than two providers in year 2005; and according to tax records, only 3.7% of employees (including the self-employed) were younger than 19 in 2005.²⁰ Among business owners, the tax records show that there was a single owner (i.e., one person) aged 18 in the year 2005, and owners below the age of 25 were fewer than 1% of owners. Second, in the sample period military service in Israel was obligatory for men aged (18-21) and women aged (18-20, with some exceptions). Soldiers are counted as part of the labor force, but the army was excluded because it is in the government sector. (However, there is no restriction on ownership of companies by soldiers.)

Sample

To define our sample, we begin with data on all companies that were active in Israel from 2003 through 2010.²¹ The pre-period for our analysis of minimum wage effects begins in 2004; although our data start in 2003, the calculation of profits requires lagged data for carryover losses. The treatment period (2006-2008) was excluded from the estimation sample for most of our analyses, so we evaluate the evidence on changes from the 2004-2005 pre-period to the 2009-2010 post-period. We impose several restrictions to arrive at our analysis sample.²²

We exclude the government sector and non-profit organizations, which we would expect to respond differently to minimum wages than do private-sector firms. Moreover, our interest is in the

firms. The data do not contain detailed profit and loss statements (including information on depreciation) for the vast majority of the companies in the sample (only for the largest ones).

²⁰ For the 2005 Income Survey, see

 $https://old.cbs.gov.il/webpub/pub/text_page_eng.html?publ=11\&CYear=2005\&CMonth=1. For the tax records, see https://mof.gov.il/ChiefEcon/StateRevenues/StateRevenuesReport/DocLib/2005/Report2005_05.pdf.$

²¹ Firms that exit in 2010 are still included in the sample since the post-treatment period covers 2009-2010. Firms that exited in 2009 or earlier were excluded from the sample except for some robustness analyses described below. ²² Table A1 in the online appendix shows how the sample size changes with these restrictions (and also reports some

descriptive statistics on key measures, to show how these change with the sample restrictions.)

incidence of minimum wages in relation to the income distribution of owners of firms, and the incomes of firm owners are most simply conceptualized and defined for private-sector firms. We exclude companies that were in the liquidation process at any point during our sample period. In principle, these liquidation events could be related to the minimum wage, leading us to potentially understate adverse effects of the minimum wage. However, such events could entail large employment or profit declines, and liquidation could be caused by quite different factors. We do not exclude publicly held companies, except where necessitated by the data. Our sample includes 144 publicly held companies in 2005, with 135 surviving through 2010.

Companies that are likely to be holding financial companies, based on extreme values of the profit rate, are also excluded. These companies can show extraordinarily high profits relative to sales (or employment) because of the nature of such companies, and because of ambiguities regarding where profits are assigned relative to where they are generated.²³ We cannot directly identify holding companies, but when we explored extreme profit rate values for specific companies, we found that they were frequently of this type.²⁴ Thus, we excluded observations with profit rates outside the range -100% to +100%.

For companies established after 2005, we cannot measure *FMW* independently of employment or wage adjustments after minimum wage increases begin in 2006. Hence, we exclude these companies. Similarly, we exclude companies with no earnings data in 2005, for which we cannot measure *FMW*.²⁵

We believe our restrictions provide a sample that is informative about the effects of the minimum wage on private-sector firms that employ at least some minimum wage workers. Our sample period begins in 2004 with 55,159 companies (with 1,802,782 employees) and ends in 2010 with 43,553 companies (with 1,769,029 employees). The number declines because companies had to have been established by 2005 to be

²³ For example, the data include a company with three employees that reports revenues of a large construction company fully owned by it.

²⁴ We examined extreme values of profit rates to try to understand if these represent real activity or some recording technique, and found that the majority of companies with these extremes were holding companies – companies fully owned by business group owners without any economic activity beyond holding a portfolio of companies.

²⁵ As Table A1 in the online appendix shows, this excludes a fairly large number of companies. There are two reasons. First, there are some very small companies in terms of number of employees, with high labor turnover during the year. Second, until 2017 many companies were set up for tax purposes, which had earned profits but had not distributed them (effectively used as a tax shelter for various professionals due to higher personal income tax rates).

included in our sample, and we lose companies through deaths but do not gain them through births. *Descriptive statistics*

Table 1 provides descriptive statistics on the key variables for our analysis sample. On average, in 2005 the average share of minimum wage workers across companies was 18%. These workers were concentrated in fewer than half of all companies (the median value of *FMW* is 0); 56% of companies had no minimum wage workers.²⁶

Average profits of the firms in the sample were around 1.3 million NIS, with a much smaller median value (0.2 million NIS). Profits can be negative (or zero). Average profit rates of companies, calculated as profits divided by sales, were 15%.

Average employment per firm was 36 workers. Large share of companies with *FMW* equal to 0% or 100% were firms with only one worker (the only values *FMW* can take on for these firms); 16.6% of firms with *FMW* equal to zero had only one worker, and 33.3% of firms with *FMW* equal to 100% had only one worker.²⁷ Companies with *FMW* in the range 80-90% had the highest employment, averaging around 250 workers per firm (Figure 3). The companies in this group were, naturally, in economic sectors with large shares of minimum-wage workers (e.g., cleaning, security, and health services and social work).

5. Effects on Employment and Profits

In this section we first report some standard types of evidence on the effects of minimum wages on employment and earnings, and then move on to evidence on how the minimum wage affects company profits. The following section turns to the evidence on how the cost of minimum wage increases is distributed across the income distribution of business owners.

Minimum wage effects on earnings and employment

We want to be clear that countries with a national minimum wage – like Israel – pose more of a challenge for estimating minimum wage effects than countries with regional variation. We are using the

²⁶ Note that *FMW* would likely be higher (and higher at more firms) if we could directly measure wages. There is no way to verify this directly, because if we expand the sample to workers other than full-year workers in their main job, we cannot map as reliably from earnings to wages (and hence *FMW*).

 $^{^{27}}$ Figure A5 in the online appendix shows the distribution of company-year observations in the estimation sample by *FMW* bins.

Israeli data not to get a better estimate of the employment effect of the minimum wage, but rather to obtain evidence – non-existent in the minimum wage literature – on the incidence of the costs of minimum wages across the income distribution of business owners. That said, we try to obtain compelling evidence on effects on employment (and earnings), and our ability to do so is important because much of our evidence on effects on business owners uses the same identification strategy. Our identification exploits variation in effects for firms with different fractions of workers paid at or below the new, higher minimum in the period just before the minimum wage increases, denoted *FMW*. We do many analyses and alterations to our specifications to bolster the case that we are estimating a causal effect.

Figure 4 shows trends in average annual earnings,²⁸ covering the pre-treatment, treatment, and post-treatment years, for workers in firms with very few minimum wage workers ($FMW \le 10\%$), firms in the midrange ($40\% < FMW \le 75\%$), and firms with a high fraction of affected workers (FMW > 75%). We normalize to 2005 values. In the pre-treatment period, there is not much difference in the trends in average earnings at the three groups of firms; if anything earnings are growing slower at the firms with higher *FMW*. After the minimum wage increase, average earnings in lower-wage (higher *FMW*) firms grew faster, consistent with a direct effect of minimum wages on earnings. The initial relative increases are largest for the highest *FMW* firms, as we would expect. There is no evidence of a change in the trend for the low *FMW* firms.

Figure 5 presents a more granular disaggregation by *FMW*. In this figure, for earnings and employment, we report estimates from a regression of the percent change from 2004-2005 to 2009-2010 on dummy variables for *FMW* bins and firm fixed effects; the figure reports the *FMW* estimates (relative to FMW = 0%). Average real earnings rose more from 2003-2005 to 2009-2010 in high *FMW* firms (the grey line), and the relationship with *FMW* is approximately monotonic. Figure 5 also presents evidence on employment changes (the black line), showing that employment fell among firms with a higher *FMW* (above around 20%), but not at very low *FMW* firms. Moreover, among the higher *FMW* firms, the

²⁸ These are earnings associated with the main job position, but since we cover full-year employees on their main jobs, it is not unreasonable to think of these as wages. Still, to be as clear as possible we refer to earnings instead of wages.

employment decline was generally sharper the higher *FMW*. The exception is at the very highest *FMW* firms, which we discuss below. The line labeled "combined" effect is the average earnings-weighted employment change. Given that earnings rose at more-affected (higher *FMW*) firms, this series declined by less than employment. With regard to effects on firm income, the employment declines suggested by Figure 5 will mitigate the costs of the higher minimum wage for more-affected firms; but this will also impact firm revenue and income.

To estimate the effects of minimum wages in a regression framework that allows for the variety of different analyses we perform, we begin with a simple difference-in-differences specification. Indexing firms by i, periods by t, and denoting firm-level controls (such as sector dummy variables) by X, we first estimate:

$$Log(Employment)_{it} = \alpha + \beta FMW_i + \gamma POST_t + \delta FMW_i \times POST_t + X_i\lambda + \varepsilon_{it} .$$
(1)

FMW has only an *i* subscript, since it is defined at the firm level for the pre-treatment period. *POST* has only a period subscript, since it is a dummy variable for the post-treatment period. While we refer to *FMW* as a percent, in the regressions we define it on a 0-1 scale. The estimate of δ from equation (1) is a difference-in-differences estimate of the effect of *FMW* – the fraction affected by the minimum wage – on employment. We add sector dummy variables and their interactions with *POST*, to allow for differential changes in outcomes over time by sector that could be correlated with *FMW*, as well as firm fixed effects. The model is also augmented in other ways described below.

The results for employment, reported in Table 2, indicate a negative impact of the minimum wage increase on employment. The estimated coefficient of δ (on *FMW* × *POST*) in the first three columns, which exclude exiters, is negative and significant. The estimate is robust across the columns, which add firm fixed effects and then sector × *POST* interactions. The estimated elasticity from column (3) of Table 2 implies that for a company with *FMW* larger by 10 percentage points, the employment decline is 1.1% larger. Given that the mean of *FMW* is 0.18, the implied elasticity is around -0.02 (approximately 1.1/[(10/18) ×100].) This is smaller than the central tendency of employment estimates (with respect to changes in the minimum wage) in the literature, which is in the range of -0.1 to -0.2 (Belman and

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Wolfson, 2019; Neumark and Shirley, 2021). However, that literature is nearly exclusively based on elasticities computed for workers, typically with a focus on workers likely to be strongly affected (like teens). Moreover, we do not weight in our analysis, since our goal is to study effects on business owners. For both of these reasons, we do not think there is a clear comparison to most estimates in the literature.

We explored whether there are changes in the pre-treatment period that look similar to changes in the treatment period, which would be evidence of spurious trends. We estimated an event-study regression for strongly treated (FMW > 0.4) and weakly treated (FMW < 0.1) firms, and found no differential in the pre-period, while the effect emerges over time in the post period; see Table 3, column (1). If anything, employment was growing faster at high FMW firms, although the estimated coefficient (*Treatment* × 2005) is not significantly different from zero.²⁹

Another potential concern is that sectors with relatively high *FMW* were more adversely impacted by the global financial crisis, potentially leading to spurious evidence of adverse effects of minimum wages. This is not a concern. First, the financial crisis had a relatively modest effect in Israel, with the unemployment rate going up only 1.5 percentage points in 2009, and declining the next year (as Figure 2 shows). Second, the companies most affected were in the financial sector, and those that are export oriented. These are sectors where firms are generally not in the high *FMW* range, so if there was a bias it would be against finding adverse minimum wage effects. Moreover, the financial sector was already partially omitted based on our profit rate criterion. In addition, our model (in columns (2) and (3) of Table 2, and in most other specifications we report) includes sector $\times POST$ interactions, to control for differential changes by sector that could otherwise be confounded with minimum wage increases. In addition, we have run the models in Table 2 (and the tables that follow) excluding the financial and export-oriented sectors, and the results/estimates barely change; see Tables A2-A6 in the online appendix.

In Table 4, we report estimates from a less constrained model, where instead of using a linear measure of *FMW*, we break *FMW* into a number of categories. We show the estimates with and without the

 $^{^{29}}$ The same conclusion is evident from a plot of 2004 to 2005 employment growth against *FMW*, as shown in Panel A of Figure A6 in the online appendix.

sector × *POST* interactions, but all columns include firm fixed effects. In columns (1) and (2) – column (2) includes the sector-*POST* interactions – the negative minimum wage effect on employment is increasing with *FMW*, except in the top range of *FMW* > 80%.

Companies in the upper range of *FMW* are quite heterogeneous in terms of size and activity, with companies with very high *FMW* concentrated in different industries and much smaller (as noted earlier). In column (3) of Table 4 we therefore split the *FMW* > 80% range into two groups: $80\% < FMW \le 94\%$ and *FMW* > 94%. Companies with $80\% < FMW \le 94\%$ are characterized by large employment, with 196.4 workers on average, while companies with *FMW* > 94% are much smaller, with average employment of 16.8. Moreover, among companies with *FMW* > 94%, 32.9% have only one worker; this percentage is 0.4% among companies with $80\% < FMW \le 94\%$. The sectors also differ substantially.³⁰ When, in column (3) of Table 4, we break up the *FMW* > 80% range this way, we find that only in the very top range (*FMW* > 94%) is the employment effect not negative (and near zero). We believe this is because companies with such a high *FMW* to adjust employment is consistent with two other pieces of evidence discussed below: that the highest *FMW* firms are more likely to exit as a result of the minimum wage increase; and that the profits of companies with very high *FMW* are reduced more strongly by the minimum wage increase; 32,33

We did one other potentially important robustness check, re-estimating the models for employment

³⁰ Companies in the lower range ($80\% < FMW \le 94\%$) are concentrated in wholesale and retail trade (16%), cleaning and security (16%), and health services and social work (12%), while companies in the upper range (FMW > 94%) are highly concentrated in the latter industry (46%), and also manpower (8%) and business services (7%).

³¹ For companies with only one worker, which is common in this top *FMW* range, the inability to adjust employment is clear. For larger companies, the lack of employment response may reflect one of the Marshall-Hicks-Allen Laws of Labor Demand, in which under some conditions the elasticity of labor demand is smaller when labor's share is larger – in particular, when the elasticity of substitution is less than the elasticity of product demand (see, e.g., Hoffman, 2009). This carries over to the consideration of low-skilled labor as a separate input, and we might think that when the firm uses largely only one type of labor, the elasticity of substitution away from (low-skilled) labor is low.

 $^{^{32}}$ Table 4 does not provide an indication of relative employment growth at low *FMW* firms stemming from the minimum wage increase, which suggests there is not a large reallocation of labor from lower-wage to higher-wage firms. Longitudinally linked employee-employer data, which we do not have at this time, could provide more decisive evidence on whether some reallocation occurs, and where (by sector or by *FMW*).

 $^{^{33}}$ Tables A7 and A8 in the online appendix show the regressions corresponding to Tables 2 and 4 (for non-exiters) for earnings. Table A8 shows a clear increase (nearly monotonic) in the positive effect on earnings the higher is *FMW*.

retaining the firms that exited before the post-period, setting their employment to zero, and using the IHS function instead of logs to accommodate the zero values. As shown in column (4) of Tables 2 and 4, we found stronger evidence of negative employment effects for higher values of *FMW*, consistent with the higher minimum wage doing more to increase firm exit. The estimates were a good deal more negative, which is not surprising given that the employment declines to zero can be very influential. In addition, in these specification we found a strong and significant negative effect for the highest *FMW* firms (*FMW* > 94%), with an estimated coefficient of -0.212 (significant at the 1% level) – consistent with an inability to adjust employment, as discussed above, which reduces the ability of these firms to survive.³⁴

Minimum wage effects on profits

We next turn to estimation of the effects of minimum wages on owners. We begin with the effects of minimum wages on profits. We begin with a similar specification to equation (1), but for profits:

$$IHS(\pi)_{it} = \alpha + \beta FMW_i + \gamma POST_t + \delta FMW_i \times POST_t + X_i\lambda + D_{\pi,b}\theta + D_{\pi,b} \times POST_t\eta + \varepsilon_{it} \quad (2)$$

Aside from the different dependent variable, we also add a vector of dummy variables for initial profit quartiles ($D_{\pi,b}$, see Table 1) and their interactions with *POST*, to allow for differential changes associated with initial profits. We begin, in column (1) of Table 5, with the simplest specification, and in columns (2) and (3) we add sector dummy variables and their interactions with *POST* and firm fixed effects (as for employment in equation (1)). The estimate of δ from equation (2) is a difference-in-differences estimate of the effect of *FMW* – the fraction affected by the minimum wage – on profits.

The estimates in columns (1) and (2) indicate that profits decline in relative terms for firms with a higher value of FMW in the pre-treatment period. The negative estimates imply that a higher minimum wage reduces profits at firms with a high share of minimum wage workers, relative to firms with a low share. Since our sample does not include failed firms, the adverse effect of the minimum wage on higher

³⁴ We do not feature these estimates because, as discussed below, it is not clear how to define profits for firms that fail, and we wanted to focus on similar samples for the employment and profit results. See the next footnote for additional detail on the results for profits including exiters.

FMW firms could be stronger than this estimate implies.³⁵

Like for the analysis of employment, we explored whether we see changes in the pre-treatment period that look the same as the changes in the treatment period, which would be evidence of spurious trends. Turning back to Table 3, column (3), we actually find that profits were growing faster at high *FMW* firms (slope not significantly different from zero), which rules out the treatment effects on profits reflecting declining profits at high *FMW* firms in the pre-treatment period.³⁶

To interpret the magnitude, given that in the regression *FMW* is defined on a scale of zero to one, the estimated coefficient of -0.749 on *FMW* × *POST* in Table 5, column (1) implies that profits declined by about 7.5% for firms where the fraction affected by the minimum wage, based on the pre-treatment period data, was about 10 percentage points higher. With the sector dummies and "trends" added, in column (2), this estimate becomes -0.795, and with firm fixed effects the estimate is similar (-0.720). These estimates appear to us to be sizable. For example, the latter estimate implies that profits of firms at the 90th percentile of *FMW* decline by 21% more than profits of firms at the 75th percentile of *FMW*.³⁷ (We return to the other

³⁵ It is unclear how to treat exiters when studying profits. Assigning zero values for profits can be misleading since profits could have been negative before exiting, in which case exiters could appear more profitable relative to surviving firms than they actually are. In Table A9 in the online appendix, we therefore present the profit regressions in two alternative ways: including the exiters with profits equal to zero; and also setting the negative profits for survivors to zero, to avoid a making the exiters more profitable. We show that regardless of how we treat the data, there is a negative, sizable, and significant estimated coefficient of *FMW* × *POST*.

A related point is that, especially if there is some exit of the most-affected firms, surviving firms could do better. Indeed, in a recent study of Germany's new minimum wage, Dustmann et al. (2019) find that affected workers have some tendency to move to higher-wage firms, consistent with reallocation towards more-productive firms, and they also find evidence that there is some firm exit (among small businesses) and that surviving firms became more productive. We estimated models for firm survival to the post-treatment period, and found that the most-affected firms were significantly less likely to survive (see Table A10 and Figure A7 in the online appendix). This implies that our evidence of declines in profits for affected firms understates the negative impact of the minimum wage on firms. ³⁶ Panel B in Figure A6 in the online appendix illustrates the same point. Looking at the three columns of Table 3, earnings dip more at the affected firms prior to the treatment. Typically we would be worried about the opposite – that if the post-treatment changes were in the same direction as the pre-treatment changes, the post-treatment changes may not be causal. Here, the opposite occurs, and then earnings rise at the most-affected firms as the minimum wage increase takes effect. We do not think there is a simple Ashenfelter dip based on random, mean-reverting declines in earnings at the most-affected firms in 2005 (relative to 2004), because earnings do not recover in the next year; rather, they begin to rise slowly, and monotonically, out through 2010. Moreover, the "proof is in the pudding," in the sense that if we just had an Ashenfelter dip/reversion to the mean in earnings, there would be no effect on employment and profits. As additional evidence, in Table A11 of the online appendix we show that the employment results are robust to dropping 2005, the year with the earnings dip in more-affected firms.

³⁷ The 75th percentile of *FMW* is 0.308, and the 90th percentile is 0.600. Multiplying this difference by the estimated effect of -0.720 predicts a decline in profits of 21%.

columns of Table 5 later.)³⁸

While the estimated profit effect is large, we do not think it is extreme based on comparisons to other studies (although admittedly we have very little evidence to compare). Consider a 0% vs. a 100% *FMW* firm. Then labor costs go up at the latter by about 15% because of the minimum wage increase. Table 5 (column (3)) implies that the effect of the 100 percentage point change in *FMW* reduces profits by 72%. Draca et al. (2011) estimate roughly a 2.7 percentage point decline in the profit rate (about 27%), from an induced wage increase of about 11%. So our estimated effect scaled by the induced wage increase is in the same ballpark (72/15 vs. 27/11). Harasztosi and Linder (2019) estimate that where 100% of the workforce is affected, wages rose by 58% (p. 2705), employment fell by about 8% (p. 2702), and the profit rate fell by around 30% (p. 2711). So their effect on profits is smaller but not completely out of line with ours. Moreover, the comparison may not be apt because we do not weight, following most of the literature, while Harasztosi and Linder do (see their footnote 12).

To provide a different way to convey the magnitude of the effect, we estimate a specification that measures the effect on profits of an estimate of the mechanical increase in labor costs from the minimum wage increase, based on the number of minimum wage workers in 2005 (the end of the pre-period). We modify equation (2) to be, instead:

$$IHS(\pi)_{it} = \alpha + \beta log(WBMW_i) + \gamma log(RMW_t) + \delta log(WBMW_i) \times log(RMW_t) + X_i \lambda$$
$$+ D_{\pi,b}\theta + D_{\pi,b} \times POST_t \eta + \varepsilon_{it} .$$
(3)

In this specification, *log(WBMI)* is the log of the wage bill for minimum wage workers, computed as the number of minimum wage workers in 2005 multiplied by the statutory 2005 minimum wage, where minimum wage workers are defined as those below the deflated minimum wage in the post-period.

³⁸ Table 3 provides comparable estimates of the effects of the minimum wage on employment, earnings, and profits. In these estimates, and based on the last pre-treatment year (2005) and the last treatment year (2010), the own-wage elasticity for employment appears to be close to -1.3, and the own-wage profit elasticity about -5. However, these are not comparable to what is often estimated in the minimum wage literature, where the wage effect is calculated for wages of workers who remain employed; here, in contrast, we are estimating an earnings effect, which includes the employment effect. As a rough guide, if the "lost" employment were added back to earnings, with no other changes, the resulting wage elasticity would be roughly double, cutting these own-wage elasticities roughly in half.

Log(RMW) is the log of the relative minimum wage in the pre- and post-periods, defined as zero (log(1)) for the pre-period and log(1.154) for the post-period, given the 15.4% increase in the minimum wage. The interaction therefore measures the percent change in profits for a one-percent higher effect of the minimum wage increase on the wage bill (based on minimum wage workers in 2005).^{39,40} Note that the increase in the wage bill we consider is a "mechanical" one, assuming no other adjustments in behavior. We expect the estimate to be negative but less than one in absolute value, as firms can adjust to the wage increase by varying employment, prices, and other variables.

The specification is otherwise the same as in Table 5 (and we estimated the model corresponding to column (3), including the sector dummies and their interactions with *POST*, and firm fixed effects). The estimated coefficient of the interaction is -0.269 (statistically significant at the 1% level), indicating that, for example, if the minimum wage increase implied a 10% larger mechanical increase in the wage bill for minimum wage workers for a firm, its profits fell by 2.7%.⁴¹ The implied effect is far less than one-for-one, consistent with the average *FMW* of 0.18 and with firms making other adjustments to offset the effects of the higher minimum wage (including, but not limited to, the reductions in employment among the lowest-earning workers that we have already estimated). For example, there is evidence of some pass-through of minimum wages into prices (e.g., Aaronson, 2001, Lemos, 2008, and Ashenfelter and Jurajda, 2021), reductions in benefits – in particular, health insurance (Marks, 2011; Clemens et al., 2018a), potentially productivity increases (Coviello et al., 2020), and substitution towards higher-skilled labor (e.g., Clemens et al., 2018b).

A richer breakdown of firms into six bins based on *FMW*, paralleling our employment specifications, provides results that reinforce the conclusion that profits fell more at firms with larger shares of minimum wage workers (see Table 6). Compared with the zero *FMW* firms in the sample, the estimates

³⁹ This is not quite the same as the implied cost of topping off below-minimum wage workers to the minimum wage, and hence allows for some increases in wages to a bit above the minimum wage; on the other hand, it does not include potential spillovers to those initially above the new minimum wage, and hence should provide a reasonable back-of-the envelope estimate of the cost of the minimum wage increase.

⁴⁰ While the inclusion of $log(WBMW_i)$ in equation (3) helps explain the strategy, when we include firm fixed effects it is fully absorbed by them.

⁴¹ The full regression results are in Table A12 in the online appendix.

show that the minimum wage hike had little or no statistically significant effect on profits for the second and third bins ($0\% < FMW \le 40\%$). For companies with *FMW* above 40%, profits decline – for the fourth bin (40-60%) by 20%, for the fifth bin (60-80%) by 44-45%, and for the highest *FMW* firms (80-100%) by 87-92%. Hence, the relationship is monotonic. The results are robust to allowing for different changes over time by economic sector (column (2)). As noted in the previous discussion of the employment results in Table 4, low labor adjustment in the top *FMW* firms can explain the larger negative impact on profits coupled with weaker evidence of employment declines at the firms with the very highest *FMW*.

Returning to Table 5, the remaining columns (4)-(7) report estimates for the effects of minimum wages on profits for the four quartiles of the profit distribution (based on the data from the period prior to the minimum wage increase) – from lowest to highest. We find the strongest negative impact of minimum wage on profits in the 1st quartile of profits (including companies with losses or near-zero profits). The effect weakens significantly for firms in the 2nd to 4th quartiles. This evidence suggests that it is the low-profit firms that bear most of the cost associated with higher minimum wages.

Effects by year

Our empirical results thus far compare the post- and pre-periods. To provide more information on the evolution of outcomes by year, Figure 6 graphs estimates of effects for earnings, employment, and IHS profits. The specifications (except for variation in the dependent variable) are the same as those in Tables 4 and 6, column (1), with the firm fixed effects included, although interacting the minimum wage variables with dummy variables for each year. We graph the interactions of the minimum wage variables with the dummy variable for each year from 2006-2010. For earnings, we see that the increases for the more-affected firms began early and reached their maximum in the post-period (2009-2010). Employment at more-affected firms began to fall with the minimum wage increases, also reaching the lowest levels in the post-period for all groups except the highest *FMW* group (as discussed earlier). In the profits graph, the decline is persistent and clear for the three most-affected groups of firms (*FMW* between 40% and 60%,

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between 60% and 80%, and between 80% and 100%).⁴²

6. Evidence on the Distribution of the Costs of Minimum Wage Increases

We now turn to the evidence exploring how the incidence of the cost of minimum wage increases varies across the income distribution of business owners.

Minimum wage effects across the income distribution

We noted earlier that a more adverse effect of minimum wages on profits at relatively low-profit firms would likely also predict that a higher minimum wage has more adverse effects on business owners with relatively low incomes ("mom and pop" shops, among businesses). To explore this question directly, we alter our regression model to estimate the effects of minimum wages on profit not by *FMW*, but by the owners' income percentile in the overall income distribution (see Table 1 for variable definitions). Denoting income percentile by *IP*,⁴³ our regression model now becomes:

$$IHS(\pi)_{it} = \alpha + \beta IP_i + \gamma POST_t + \delta IP_i \times POST_t + X_i\lambda + \varepsilon_{it} .$$
⁽⁴⁾

We have substituted *IP* for *FMW*, to detect post-treatment changes in profits across the distribution of incomes of business owners. Note that *IP* has only an *i* subscript, since it is defined at the owner level, but only for the pre-treatment period.

The estimates of equation (4) are reported in Table 7, columns (1) and (2). Recall that a higher income percentile (*IP*) implies higher income of business owners. Thus, the positive estimated coefficients on the $IP \times POST$ interaction imply that profits of lower-income business owners declined more in relative terms in the post-treatment period. For example, the estimated coefficient of 2.351 in column (2) implies that firms with owners earning median income suffered a 7.6% larger decline in profits than those at the

 $^{^{42}}$ It would be too messy to include confidence intervals in these figures. In the earnings graph, the standard errors for the estimates shown reach a maximum of about 0.013, so nearly all of the estimates are significant (using, for this footnote, the 5% significance level). For the employment regressions, the standard errors for the estimates shown reach a maximum of about 0.022, so again most of the estimates are significant. For the profit regressions, the standard errors are larger, reaching a maximum of about 0.123. All of the estimates for the groups with *FMW* between 60% and 80%, and between 80% and 100%, are statistically significant, as are the estimates for firms with *FMW* between 40% and 60% beginning in 2007.

⁴³ The income percentiles are calculated based on universe of business owners and workers in 2005. Owners' income includes wages from owned business and other labor income.

75th percentile.⁴⁴ In columns (3) and (4) we report estimates of an expanded model disaggregating *IP* by the 10 deciles of the distribution, with the top decile as the reference category. The estimates indicate that the negative effect on profits is larger for the lower income business owners and the relationship is quite monotonic.

Minimum wage incidence relative to the income distribution of business owners

The implication of the preceding estimates is that profits of lower-income business owners were more adversely affected by the 2006-08 minimum wage increases. We already saw that profits (and employment) of higher *FMW* firms were also more adversely affected. Column (1) of Table 8 shows why these results coincide. Based on pre-treatment data – as is all the analysis in this subsection – the owners of businesses with higher *FMW* are lower in the income distribution. The relationship between *FMW* and the income percentile of business owners is nearly monotonically negative.⁴⁵

Note, though, that even among the high *FMW* firms, business owners are relatively high in the income distribution – for individuals, for example (column (1)), ranging from the 74th to the 94th percentile across different *FMW* bins. Thus, this evidence suggests that the impact of minimum wage increases is regressive in terms of the income distribution of business owners, but still relatively progressive with regard to the overall distribution of income (and this would be more the case if, as seems likely, business owners have more assets because of the businesses they own).

To provide a richer perspective on distributional effects, column (2) shows the average income percentiles of workers, based on FMW – again, defined based on full-year workers – at the firms at which they were employed prior to the minimum wage increase. As we would expect, workers in higher FMW firms have lower incomes on average; the income percentile declines nearly monotonically with FMW.

Although our principle focus is on the incidence across the income distribution of business owners, the comparison between columns (1) and (2) is potentially of interest, showing that in the lowest *FMW*

⁴⁴ The median of *IP* is 0.9516, and the 75th percentile is 0.9842. Multiplying this difference by the estimated effect of 2.351 predicts a decline in profits of 7.6%.

⁴⁵ We use pre-treatment ownership data to abstract from ownership and income changes that could result from the minimum wage hike.

firms, workers' average income percentiles (77) are about the same as those of owners in the higher *FMW* firms (the percentiles range from 74 to 79 for firms with *FMW* above 60%). In this sense, too, the redistributional impacts of the minimum wage seem, if not strictly regressive, than at least sharing the burden of the minimum wage very unequally between economic agents at similar points in the income distribution, with owners paying directly for the higher minimum wage, while workers with similar incomes do not. However, this evidence has to be interpreted cautiously, because we cannot measure the asset value of the firm, and owners may invest in their businesses in ways that lower reported income but increase the value of their business. On the other hand, incomes of workers in the high *FMW* firms are clearly lower than incomes of the owners of these firms, and in that sense the minimum wage does redistribute income downward.

In columns (3) and (4) of Table 8 we extend the analysis to family income – the usual lens through which we view redistribution, poverty, etc. Doing so amplifies the regressivity of the incidence of the minimum wage with respect to the distribution of incomes of business owners. Families with ownership of the highest *FMW* firms earn slightly less than the median family income in 2005 (column (3)), and the income percentiles for firms with *FMW* > 60% are not much above the median. Moreover, the income percentile declines significantly more, as *FMW* increases, in column (3) compared to column (1).⁴⁶

Table 9 provides more details on the income distribution by *FMW* that includes family incomes (adding owners'/workers' spouses). Panel A shows the details for married business owners.⁴⁷ Owners of high *FMW* businesses are younger, slightly more likely to be female, and have more children. The first two, at least, are almost certainly associated with lower income. As shown in column (5), average incomes of owners drop dramatically with *FMW* – and the contrast between lower *FMW* firms (e.g., *FMW* below 20%)

⁴⁶ In addition, workers in the lowest *FMW* firms are at higher income percentiles than business owners at the highest *FMW* firms. On the other hand, it is still very much the case that workers at high *FMW* firms are considerably lower in the family income distribution than workers at low *FMW* firms (column (4)). Thus, we might conclude that, overall, the effect of the minimum wage hike is fairly regressive for business owners (and even more so for families of business owners), but is progressive for workers (and even more so for families including minimum wage workers). ⁴⁷ Disaggregating by married and single provides income comparisons that do not depend on marital status. Among

business owners, 85% are married, while the share of married workers is lower at 61%. The comparison of single business owners/workers is shown in Table 10.

and higher *FMW* firms (e.g., *FMW* above 30%) is quite striking. Columns (9) and (10) show that incomes of spouses exacerbate the relationship between *FMW* and income, as incomes of spouses are higher for owners of low *FMW* firms, and hence the gaps in family income across the distribution of *FMW* are generally larger than for owner income alone (column (5) vs. (10)). The numbers in column (11) provide a summary measure showing this conclusion, indicating that family income declines quite sharply as *FMW* increases. These data reinforce the conclusion that, among business owners, the burden of the minimum wage falls on lower-income families.

Panel B of Table 9 shows that, as expected, average worker earnings decline with *FMW* (column (5)) – reflecting the fact that at high *FMW* firms most workers earn the minimum wage. This is reinforced by spouse's earnings, which also decline with *FMW*, indicating that incomes of spouses are positively correlated. However, spouses' earnings do not decline nearly as steeply with *FMW*. Moreover, earnings of spouses of workers in high *FMW* firms are generally much higher, as reflected in the fact that average earnings of spouses of those in high *FMW* firms are much higher than those of the workers in these firms (column (8) vs. column (5)). This indicates that a large share of minimum wage workers are the secondary earners in families, which reduces the redistributional effects of minimum wages towards families with the lowest incomes.⁴⁸ At the same time, a comparison of the family incomes of owners and workers in higher *FMW* firms – in column (10) of Panels A and B – indicates that the family incomes of minimum wage workers are much lower than those of the owners of high *FMW* firms.⁴⁹

Analysis for singles, who constitute a much smaller share of the sample, provides similar results (see Table 10). As before, incomes of business owners and average earnings of workers decline with *FMW*. The main and obvious difference is that single owners and workers do not have the additional incomes of

⁴⁸ This parallels evidence for the United States, for example, that the high share of teens and young adults among minimum wage workers implies that a large share of the benefits of higher minimum wages do not go to poor families (Lundstrom, 2017).

⁴⁹ For example, in firms where *FMW* exceeds 90%, average family income of owners is 198,334 NIS, vs. average family income of workers in those firms of 74,137 NIS. And the average family incomes of workers in low *FMW* firms are similar to the average family incomes of the owners of high *FMW* firms, implying that a large share of the cost of the minimum wage – for high *FMW* firms – falls heavily on one among two different types of agents (owners vs. workers), despite them having relative similar incomes, which is inconsistent with horizontal equity.

spouses.

Thus, the conclusions from Table 9 (and 10) are similar to those from Table 8. The higher minimum wage does redistribute income towards lower-income families. But the redistribution is not from the top of the income distribution, but rather from owners whose incomes are higher than those of minimum wage workers, but whose incomes are lower than those of many other business owners who do not bear the burden of the minimum wage (and whose incomes are more similar to those of many higher-earning workers).⁵⁰ Moreover, note that this distributional analysis does not account for profits of publicly-held companies paid out as dividends (or reflected in appreciation of stocks). Since these companies are large and likely least affected by minimum wage increases,⁵¹ and stocks are held by higher-income individuals and families, our distributional analysis almost certainly understates the degree to which the burden of the minimum wage falls on lower-income business owners – making our evidence of the regressive nature of the burden of minimum wages, with respect to the income distribution among business owners, more striking.

7. Conclusions

In this paper, we study an unexplored dimension of the distributional effects of the minimum wage, turning attention from the question of the effects on workers, who are potentially helped by a higher minimum wage, to the question of who *pays* for it. A key goal of a higher minimum wage is income redistribution towards low-income families. Existing research on the minimum wage focuses on the impact on affected workers, but is largely silent on the incomes of the owners of businesses who pay for a higher minimum wage, and especially silent on the incidence of the costs of a higher minimum wage across the income distribution of business owners.

 $^{^{50}}$ A related issue, which we do not address here, is how minimum wages affect prices, since price changes affect the purchasing power of income changes. Indeed recent U.S. evidence in Renkin et al. (forthcoming) points to significant (actually, *full*) price pass-throughs of minimum wage increases, with the negative effects larger for poor families. Effects like these work against any progressive redistribution from the minimum wage. Related to this point, we would of course ideally like to measure effects on consumption (or utility) rather than income. We have no consumption data for business owners, or information on loans or sources of credit that might help smooth consumption.

⁵¹ In the online appendix, Figure A8 shows the equivalent of Figure A5, but for publicly-traded companies. There are much lower shares of these companies in bins with high values of *FMW*.

We study evidence on this dimension of the distributional effects of minimum wages using a unique administrative dataset on the universe of tax records for Israel, in the period surrounding a large and plausibly exogenous minimum wage increase. The evidence indicates that the minimum wage hike reduced profits of companies, with minimum-wage intensive companies bearing the bulk of the cost and adjusting their workforces more aggressively, and profits declining more for lower-income business owners. Moreover, owners of businesses with higher shares of minimum wage workers ranked at the bottom of the income distribution of business owners. In most cases, spouses of business owners earn less than the owners while spouses of minimum wage workers earn more, further reducing the redistributive effect of the minimum wage.

Overall, then, the evidence indicates that the higher minimum wage does redistribute income towards lower-income families. However, the redistribution is not from the top of the income distribution. Rather, it is from owners whose incomes are higher than those of minimum wage workers, but whose incomes are lower compared to other owners who do not bear the burden of the minimum wage.

Moreover, although not our main focus (in part because of data limitations), the incomes of owners most affected by minimum wage increases are similar to the incomes of many higher-earning workers. This implies that a large share of the cost of the minimum wage falls heavily on one among two different types of agents (owners of high *FMW* firms vs. workers at low *FMW* firms), despite them having relative similar incomes – inconsistent with horizontal equity.

Thus, minimum wages, while somewhat redistributive, are not effectively redistributing from the highest-income individuals or families. In that sense, the minimum wage is quite different from other redistributional policies financed by taxes – such as the U.S. Earned Income Tax Credit.⁵² We have no way of doing a direct comparison of the minimum wage with other redistributive tax policies, including simply more progressive taxation, in the absence of a structural model, so we do not, in this paper, quantify the

⁵² There is also an EITC in Israel (see Brender and Strawczynski, 2019).

differences in the distributional effects of the minimum wage and alternative policies.⁵³ But our results clearly suggest that such an inquiry could be very useful (for Israel as well as other countries). Another policy that could be considered to improve the targeting of who *pays* for a higher minimum wage is to create a tax credit that offsets part of the cost of higher minimum wages borne disproportionately by the businesses that employ a relatively higher share of minimum wage workers (Neumark, 2019) – businesses that, as we have shown, have lower profits and are owned by those with lower incomes than the owners of other businesses.

⁵³ And of course our evidence on the regressive nature of the burden of higher minimum wages for business owners should be viewed alongside evidence, addressed in prior research, that minimum wages do not do a very good job of targeting benefits to low-income individuals or families.

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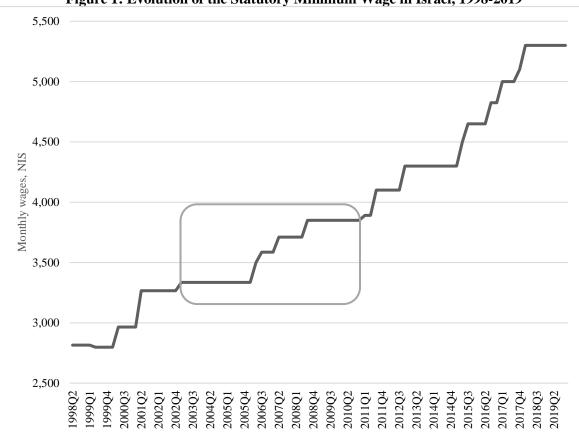


Figure 1: Evolution of the Statutory Minimum Wage in Israel, 1998-2019

Notes: The minimum wage is in New Israeli Shekels (NIS) per month, at current prices, based on full-time work. The box shows the period covered by our analysis.

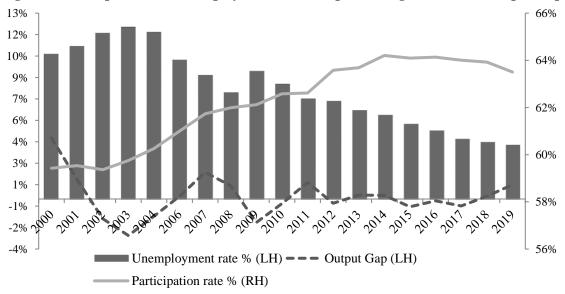


Figure 2: Participation and Unemployment Rates (Population aged 15+), and Output Gap

Source: For unemployment rate and participation rate: Central Bureau of Statistics (CBS), Labor Force Survey (LFS). We calculated the output gap by applying Hodrick-Prescott filter to quarterly national accounts series for GDP. Output Gap is calculated as part of the macroeconomic forecasting process in the Israeli Ministry of Finance and serves as a measure of slack in the economy. OECD publishes similar figures for the Israeli economy as part of the OECD Economic Outlook.

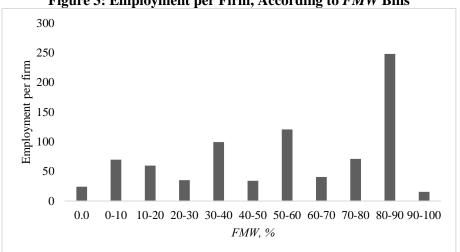


Figure 3: Employment per Firm, According to *FMW* Bins

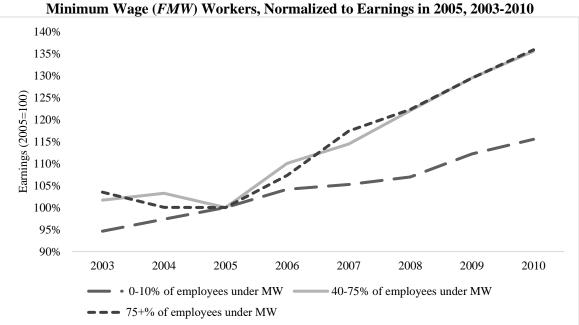
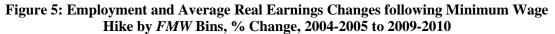
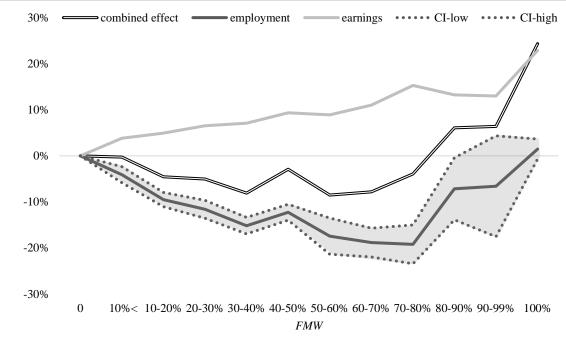


Figure 4: Earnings Trends for Workers in Firms with Low and High Initial Fractions of Minimum Wage (*FMW*) Workers, Normalized to Earnings in 2005, 2003-2010

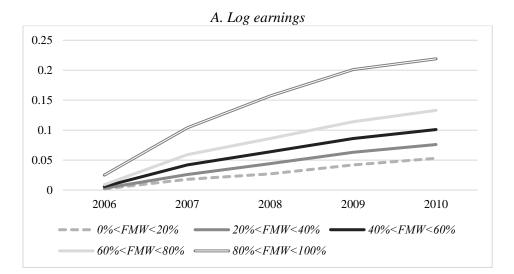
Note: Annual earnings in NIS normalized to earnings in 2005 for each group of workers based on *FMW* at their firms.

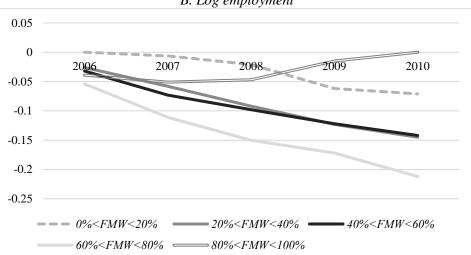




Notes: The estimates in this figure are based on regressions on the *FMW* bins and firm fixed effects, with estimates measured relative to the FMW = 0 group. Earnings growth and the combined effect are in real terms. The combined effect is the sum of (generally negative) employment effect and (positive) earnings effect calculated for each bin. For employment effects, "CI-low" and "CI-high" are the bottom and top of the 95% confidence interval.

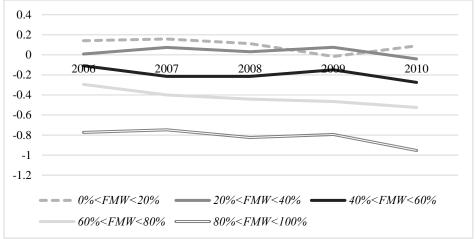
Figure 6: Estimated Effects on Earnings, Employment, and Profits by Year, 2006-2010





B. Log employment

C. IHS Profits



Note: Specifications correspond to Tables 4 and 6, column (1), except that we include dummy variables for each year rather than a simple *POST* variable.

Variable	Description	Ν	Median	Mean	Std. Dev.	Minimum	Maximum
Profits, thousands NIS	Taxable corporate income from all sources including reimbursement of owner's salary (1,000's NIS/year)	202,187	195	1,282	40,859	-1,610,081	8,840,595
IHS(π)	Inverse hyperbolic sine (IHS) transformation of corporate income variable, defined as:	202,187	5.95	4.11	4.80	-14.98	16.69
	$\sinh^{-1}(\operatorname{income}) = \log[\operatorname{income} + (\sqrt{\operatorname{income}^2 + 1})]$	202.105	0.00	0.1.7	0.05	1.00	1.00
Profit rate	Corporate income/sales (used for sample restriction)	202,187	0.09	0.15	0.27	-1.00	1.00
Initial profits quartiles:	Dummy variables for quartiles of companies' profits in the pre-period (2004-2005)						
Low		202,187	0	0.23	0.42	0	1
Medium-low		202,187	0	0.24	0.43	0	1
Medium		202,187	0	0.26	0.44	0	1
High		202,187	0	0.27	0.44	0	1
Employment	Employee headcount at the company level (main position)	201,742	8.00	35.97	254.51	1	29,694
Log(employment)	Natural logarithm of employee headcount (main position)	201,742	2.08	2.15	1.44	0.00	10.30
FMW	Fraction of employees paid monthly wages at or below the post-hike minimum wage for 2009-2010 based only on 2005 data on employees who worked continuously throughout the full year on their main job; 2005 is the last year before treatment; <i>FMW</i> is measured on a 0-1 scale (although the charts present percentages for clarity)	202,187	0	0.18	0.28	0	1
POST	Post-treatment dummy variable for years 2009-2010 (vs. the 2004-2005 pre-treatment period)	202,187	0	0.44	0.50	0	1
Owners' income percentile (<i>IP</i>) in 2005	Percentiles of business owners' incomes in the total income distribution of business owners and workers, in 2005	58,384	95	90	15	0	100
Workers income percentile (<i>IP</i>) in 2005	Percentiles of workers' incomes in the total income distribution of business owners and workers, in 2005	1,269,078	48	48	28	0	99
Weighted owners income percentile by company (<i>IP</i>) in 2005	After calculating owners' income percentile (<i>IP</i>) in 2005 at the individual-level, we calculated a weighted average, by percentage of ownership, of <i>IP</i> for each company	44,866	94	90	14	0	99

Table 1: Variables Used in Study and Descriptive Statistics of the Estimation Sample, 2004-2005 and 2009-2010

NIS = New Israeli Shekels. We also define 23 sector dummy variables, for the following sectors: manufacturing; agriculture; high tech; electricity and water supply; construction; wholesale and retail trade; accommodation services and restaurants; transport and communications; financial institutions; real estate; computer and related activities; R&D; manpower; security and cleaning; education; health services and social work; communities and other social and personal services; and other business activities.

				Including companies that exit, IHS(employment)
Variables	(1)	(2)	(3)	(4)
FMW	0.109***	-0.084***		
	(0.022)	(0.020)		
POST	0.068^{***}	0.023	-0.057	-0.587***
	(0.005)	(0.063)	(0.058)	(0.073)
$FMW \times POST$	-0.122***	-0.098***	-0.114***	-0.368***
	(0.019)	(0.018)	(0.013)	(0.021)
Sector dummies and sector dummies $\times POST$	no	yes	yes	yes
Firm fixed effects	no	no	yes	yes
R ²	0.001	0.118	0.925	0.863
N	201,742	201,742	201,742	224,932

 Table 2: Regression Estimates for Minimum Wage Effect on Log Employment, Based on Fraction

 Affected by Minimum Wage (FMW), 2004-2005 and 2009-2010

Notes: Dependent variable is log(employment) per firm in each year (IHS(employment) in (4)). Levels of significance: $10\%^*$, $5\%^{**}$, $1\%^{***}$. Standard errors are clustered at the firm level. *FMW* is measured on a 0-1 scale in the regressions. We use IHS(employment) because of the zero values for exiting companies in (4). Three dots (...) indicates that the variable is absorbed by the firm fixed effects.

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	(1)	(2)	(3)
Variables	Ln(employment)	Ln(earnings)	$IHS(\pi)$
$Treatment \times 2005$	-0.006	-0.076***	0.130***
(pre-trend)	(0.006)	(0.003)	(0.028)
$Treatment \times 2006$	-0.042***	-0.027***	-0.267***
	(0.007)	(0.004)	(0.050)
$Treatment \times 2007$	-0.080***	0.022***	-0.313***
	(0.008)	(0.005)	(0.053)
$Treatment \times 2008$	-0.100***	0.052***	-0.357***
	(0.009)	(0.005)	(0.056)
$Treatment \times 2009$	-0.106***	0.079***	-0.319***
	(0.010)	(0.006)	(0.058)
$Treatment \times 2010$	-0.119***	0.094***	-0.456***
	(0.011)	(0.006)	(0.060)
Year fixed effects	yes	yes	yes
Firm fixed effects	yes	yes	yes
Initial profits dummies and	no	no	yes
initial profits $\times POST$			
R ²	0.924	0.880	0.625
N	282,579	282,579	283,355

 Table 3: Event-Study Regression Estimates for Minimum Wage Effects, Based on

 High vs. Low Fraction Affected by Minimum Wage (FMW), 2004-2010

Notes: Dependent variables are log(employment), log(earnings), and IHS(π) per firm in each year. Levels of significance: 10%^{*}, 5%^{**}, 1%^{***}. Standard errors are clustered at the firm level. *FMW* is measured on a 0-1 scale in the regressions. Treatment group defined as a dummy variable equal to 1 if *FMW*>40% (affected firms) and 0 if *FMW*<10% (unaffected firms). The sample is restricted to these *FMW* categories, but there are more observations than in Table 2 because 2006-2008 observations are included. The main treatment effect is absorbed by the firm fixed effects, and the treatment × year interactions are relative to 2004.

,				Including companies that exit, IHS(employment)
Variables	(1)	(2)	(3)	(4)
POST	0.029***	-0.035	-0.034	-0.569***
	(0.004)	(0.049)	(0.049)	(0.073)
<i>FMW</i> >0-20% ×	-0.072***	-0.072***	-0.072***	-0.048***
POST	(0.009)	(0.009)	(0.009)	(0.018)
<i>FMW</i> 20-40% ×	-0.135***	-0.132***	-0.132***	-0.169***
POST	(0.010)	(0.010)	(0.010)	(0.019)
<i>FMW</i> 40-60% ×	-0.130***	-0.126***	-0.127***	-0.252***
POST	(0.013)	(0.013)	(0.013)	(0.022)
FMW 60-80% $ imes$	-0.190***	-0.184***	-0.184***	-0.475***
POST	(0.021)	(0.021)	(0.021)	(0.038)
FMW 80-100% \times	0.005	0.006		
POST	(0.016)	(0.016)		
FMW 80-94% $ imes$			-0.062*	-0.691***
POST			(0.037)	(0.138)
FMW 94-100% \times			0.014	-0.212***
POST			(0.435)	(0.024)
Sector dummies and	no	yes	yes	yes
sector dummies \times				
POST				
Firm fixed effects	yes	yes	yes	yes
R ²	0.925	0.926	0.926	0.862
N	201,742	201,742	201,742	224,932

Table 4: Regression Estimates for Minimum Wage Effect on Log Employment, with SeparateFMW Bins, 2004-2005 and 2009-2010

Notes: Dependent variable is log(employment) per firm in each year (IHS(employment) in column (4)). Levels of significance: $10\%^*$, $5\%^{**}$, $1\%^{***}$. Standard errors are clustered at the firm level. *FMW* is measured on a 0-1 scale in the regressions. Specification (3) splits the top *FMW* 80-100% bin into two groups that are more homogenous, as explained in the text. The main effects of the *FMW* bins are absorbed by the firm fixed effects. The *FMW* × *POST* interactions are relative to *FMW* 0%. IHS(employment) is used in column (4) because of the zero values for exiting companies.

Variables	(1)	(2)	(3)	(4)	(5)	(6)	(7)
				Sample: low	Sample: medium	Sample: medium	Sample: high initial
	Full sample	Full sample	Full sample	initial profits	low initial profits	initial profits	profits
FMW	0.155***	0.179***					
	(0.022)	(0.023)					
POST	-1.416***	-2.022***	-2.017***	4.603***	-1.330***	-1.668***	-1.841***
	(0.034)	(0.224)	(0.227)	(0.503)	(0.354)	(0.387)	(0.553)
$FMW \times POST$	-0.749***	-0.795***	-0.720***	-1.805***	-0.172*	-0.362**	0.026
	(0.068)	(0.070)	(0.070)	(0.152)	(0.105)	(0.123)	(0.167)
Initial profits	yes	yes	yes	no	no	no	no
dummies							
Initial profits × POST	yes	yes	yes	no	no	no	no
Sector dummies and	no	yes	yes	yes	yes	yes	yes
sector dummies \times							
POST							
Firm fixed effects	no	no	yes	yes	yes	yes	yes
\mathbb{R}^2	0.551	0.554	0.732	0.641	0.471	0.465	0.483
Ν	202,187	202,187	202,187	47,146	48,082	52,332	54,627

Table 5: Regression Estimates of Minimum Wage Effect on Profits ($IHS(\pi)$), Based on Fraction Affected by Minimum Wage (FMW), 2004-2005 and 2009-2010

Notes: Dependent variable is $IHS(\pi)$ per firm in each year. Levels of significance: 10%^{*}, 5%^{**}, 1%^{***}. Standard errors are clustered at the firm level. *FMW* is measured on a 0-1 scale in the regressions. Three dots (...) indicates that the variable is absorbed by the firm fixed effects.

Variables	(1)	(2)
POST	-1.431***	-2.088***
	(0.037)	(0.227)
$FMW > 0-20\% \times POST$	0.015	0.098^{*}
	(0.059)	(0.059)
FMW 20-40% $\times POST$	-0.004	0.032
	(0.056)	(0.056)
FMW 40-60% $\times POST$	-0.201***	-0.197***
	(0.064)	(0.065)
FMW 60-80% $\times POST$	-0.438***	-0.451***
	(0.105)	(0.106)
FMW 80-100% $\times POST$	-0.874***	-0.924***
	(0.084)	(0.085)
Initial profit dummies	yes	yes
Initial profit dummies $\times POST$	yes	yes
Sector dummies and sector dummies $\times POST$	no	yes
Firm fixed effects	yes	yes
R ²	0.730	0.732
Ν	202,187	202,187

Table 6: Regression Estimates of Minimum Wage Effect on Profits $(IHS(\pi))$, Based on Fraction Affected by Minimum Wage (FMW), with Separate *FMW* Bins, 2004-2005 and 2009-2010

Notes: Dependent variable is $IHS(\pi)$ per firm in each year. Levels of significance: 10%^{*}, 5%^{**}, 1%^{***}. Standard errors are clustered at the firm level. The main effects of the *FMW* bins are absorbed by the firm fixed effects. The *FMW* × *POST* interactions are relative to *FMW* 0%. IHS(employment) because of the zero values for exiting companies.

Variables	(1)	(2)	(3)	(4)
POST	-3.431***	-3.909***	-1.075***	-1.602***
	(0.175)	(0.285)	(0.032)	(0.229)
Income percentile $(IP) \times POST$	2.395***	2.351***		
•	(0.178)	(0.178)		
Income percentile (IP) 0-10% ×			-1.803***	-1.775***
POST			(0.483)	(0.485)
Income percentile (IP) 10-20%			-1.315***	-1.342***
$\times POST$			(0.357)	(0.358)
Income percentile (IP) 20-30%			-1.399***	-1.379***
× POST			(0.342)	(0.336)
Income percentile (IP) 30-40%			-1.258***	-1.237***
× POST			(0.277)	(0.278)
Income percentile (IP) 40-50%			-1.535***	-1.526***
$\times POST$			(0.240)	(0.234)
Income percentile (IP) 50-60%			-0.859***	-0.833***
× POST			(0.200)	(0.200)
Income percentile (IP) 60-70%			-0.889***	-0.865***
$\times POST$			(0.127)	(0.127)
Income percentile (IP) 70-80%			-0.723***	-0.695***
× POST			(0.086)	(0.085)
Income percentile (IP) 80-90%			-0.382***	-0.369***
$\times POST$			(0.057)	(0.057)
Initial profit dummies	yes	yes	yes	yes
Initial profit dummies × POST	yes	yes	yes	yes
Sector dummies and sector	no	yes	no	yes
dummies $\times POST$		-		-
Firm fixed effects	yes	yes	yes	yes
R ²	0.724	0.725	0.724	0.725
N	155,647	155,647	155,647	155,647

Table 7: Regression Estimates of Minimum Wage Effect on Profits ($IHS(\pi)$), Based on Income Percentile for Incomes of Business Owners, 2004-2005 and 2009-2010

Notes: Dependent variable is $IHS(\pi)$ per firm in each year. Levels of significance: 10%^{*}, 5%^{**}, 1%^{***}. Standard errors are clustered at the firm level. (Clustering at the owner level is complicated because there can be multiple and overlapping owners.) *IP* is measured on a 0-1 scale in the regressions. The main effect of *IP* is absorbed by the firm fixed effects. In columns (3) and (4), the *IP* decile × *POST* interactions are relative to the top decile.

	(1)	(2)	(3)	(4)
	Business owners'	Workers' income	Business owners' family	Workers' family
FMW	income percentile	percentile	income percentile	income percentile
0	88	77	71	57
< 10%	94	77	82	57
10-20%	87	60	72	44
20-30%	85	54	69	40
30-40%	81	48	63	37
40-50%	79	44	60	35
50-60%	80	38	61	31
60-70%	78	38	56	30
70-80%	78	38	57	28
80-90%	79	34	53	25
90-100%	74	36	48	24
Total	87	68	68	49

Table 8: Income Percentiles of Business Owners and Workers, Individual and Family, by FMW, 2005

Note: Individual figures represent approximate rank of column (5) in Tables 9 and 10. Family income figures represent approximate rank of column (10) in Table 9. Family income represents combined income of both married spouses in 2005. Annual owner or worker earnings below 6,000 NIS were omitted from calculation, as these can reflect small components of income that are not labor income from the main job. The rank was calculated based on the family income distribution including both owners' and workers' families.

	-		•	· · · ·		A. O	wners					
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
											Family income relative to	Owner/
				# children				Annual	% of		average in	spouse
		Average		under 18	Owners	Spouse	Spouse %	spouse	spouses	Family	$0 < FMW \le 0.1$	income
FMW	Ν	age	% female	in 2005	income	age	female	income	working	income	range	ratio
0	28,904	49.85	4.4%	2.33	235,192	46.9	95.7%	116,235	64.3%	312,083	68%	3.1
< 10%	3,419	52.66	2.7%	2.32	372,911	49.4	97.3%	140,150	59.4%	458,600	100%	4.5
10-20%	5,181	50.92	3.4%	2.34	230,751	47.7	96.6%	104,015	63.7%	298,226	65%	3.5
20-30%	3,086	49.50	4.5%	2.38	202,910	46.4	95.5%	101,814	66.1%	269,795	59%	3.0
30-40%	3,712	49.05	4.4%	2.44	168,659	45.8	95.7%	91,188	66.9%	229,877	50%	2.8
40-50%	3,962	49.26	5.1%	2.38	159,372	46.1	95.0%	88,750	67.4%	221,902	48%	2.7
50-60%	801	48.37	4.4%	2.45	166,414	45.0	95.7%	86,703	63.8%	223,853	49%	3.0
60-70%	1,149	48.63	5.0%	2.48	159,633	45.2	95.1%	82,982	64.2%	215,152	47%	3.0
70-80%	628	47.16	6.2%	2.64	149,760	44.1	93.9%	87,533	64.3%	206,732	45%	2.7
80-90%	258	45.96	5.8%	2.63	168,803	42.3	94.1%	92,481	53.9%	217,249	47%	3.4
90-100%	2,728	50.83	6.0%	2.42	134,650	47.7	94.2%	91,137	62.0%	198,334	43%	2.4
						B. W	orkers					
											Family income	Worker/
				# children				Annual	% of		relative to	spouse
		Average		under 18	Annual	Spouse	Spouse %	spouse	spouses	Family	$0 < FMW \le 0.1$	income
FMW	N	age	% female	in 2005	earnings	age	female	income	working	income	range	ratio
0	228,395	40.72	29.3%	1.44	131,069	39.1	66.1%	104,518	67.4%	207,498	102%	1.9
< 10%	263,140	40.06	34.5%	1.35	130,888	38.6	61.6%	98,712	70.8%	202,744	100%	1.9
10-20%	100,238	40.04	45.7%	1.35	73,505	39.4	47.5%	89,811	65.1%	138,431	68%	1.3
20-30%	60,511	38.54	46.5%	1.26	62,541	37.9	47.0%	85,500	65.1%	122,979	61%	1.1
30-40%	60,808	38.47	50.1%	1.22	53,218	38.0	42.6%	81,848	64.6%	112,959	56%	1.0
40-50%	37,324	38.42	48.7%	1.27	49,519	38.2	43.6%	83,757	63.0%	107,895	53%	0.9
50-60%	19,391	38.35	52.4%	1.22	41,239	38.3	39.4%	74,144	62.3%	93,238	46%	0.9
60-70%	23,041	39.63	59.0%	1.26	38,086	39.9	33.6%	74,344	63.7%	89,567	44%	0.8
70-80%	10,339	38.18	55.0%	1.32	37,129	38.4	34.8%	73,247	60.2%	85,074	42%	0.8
80-90%	9,759	40.66	66.7%	1.40	29,494	41.2	24.1%	71,446	59.8%	74,276	37%	0.7
90-100%	18,561	43.81	70.1%	1.32	28,652	44.1	21.9%	79,643	58.7%	74,137	37%	0.6

Table 9: Descriptive Statistics for Family Income by FMW: Married Business Owners and Workers, 2005

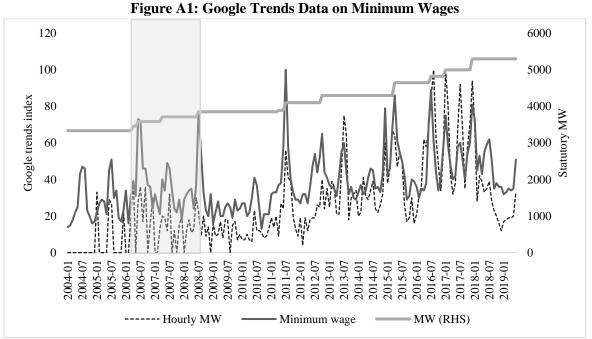
Notes: Annual owner or worker earnings below 6,000 NIS were omitted from calculation. Family income is calculated by combining spouses' incomes.

A. Owners								
	(1)	(2)	(3)	(4)	(5)			
				# children under	Annual owner			
FMW	Ν	Age	% female	18 in 2005	earnings			
0	5,347	47.38	21.5%	1.80	191,728			
< 10%	491	53.35	25.7%	1.79	393,442			
10-20%	766	49.20	23.5%	1.82	199,852			
20-30%	489	46.60	26.6%	1.89	160,847			
30-40%	667	47.07	25.8%	1.81	142,512			
40-50%	773	45.72	25.5%	1.86	124,418			
50-60%	150	44.31	19.3%	2.00	151,856			
60-70%	236	43.35	23.7%	1.79	105,619			
70-80%	123	43.11	28.5%	2.21	113,148			
80-90%	62	39.81	16.1%	2.58	108,605			
90-100%	742	45.48	24.4%	1.76	81,367			
			B. Workers					
				# children under				
FMW	Ν	Age	% female	18 in 2005	Annual earnings			
0	100,316	32.26	37.5%	0.36	72,937			
< 10%	111,393	32.37	44.3%	0.33	75,927			
10-20%	62,031	31.12	47.8%	0.31	43,042			
20-30%	48,414	28.90	44.9%	0.23	37,045			
30-40%	52,701	28.09	47.6%	0.22	30,853			
40-50%	35,770	27.51	46.6%	0.20	27,958			
50-60%	21,264	27.11	47.5%	0.21	24,578			
60-70%	22,865	28.37	55.2%	0.24	23,465			
70-80%	10,429	27.75	49.9%	0.22	23,289			
80-90%	8,215	31.61	60.4%	0.33	20,296			
90-100%	12,312	38.46	71.6%	0.44	21,396			

 Table 10: Descriptive Statistics by FMW: Single Business Owners and Workers, 2005

Notes: Annual owner or worker earnings below 6,000 NIS were omitted from calculation.

Online Appendix



Notes: Leading search terms related to minimum wage in Hebrew as follows: minimum wage – שכר מינימום, hourly minimum wage – שכר מינימום לשעה. Google trends index is available since 2004. Zero values in Google trends commonly represent a search query sample that is not large enough to be reported. The shaded area shows the period of the minimum wage increases covered by our analysis.

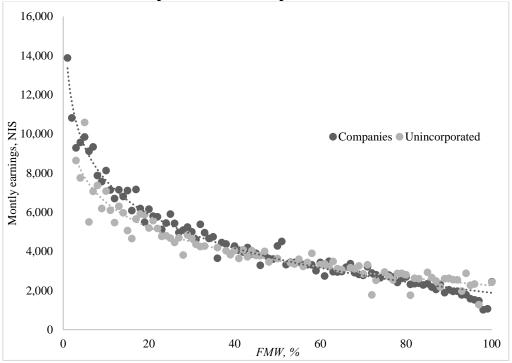


Figure A2: Comparisons of Monthly Earnings of Employees in Companies and Unincorporated Businesses

Note: Averages for 1% bins of *FMW* distribution in the year 2005. The chart uses the *FMW* distribution of unincorporated business that was calculated based on earnings of salaried workers employed by unincorporated business.

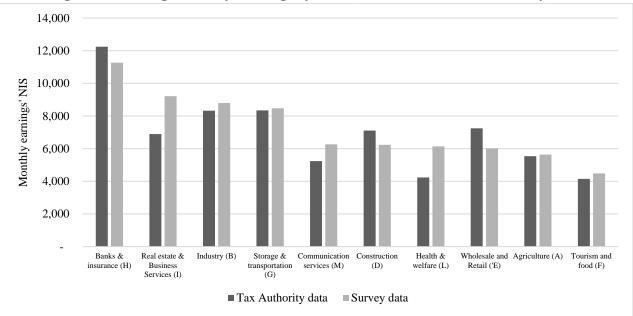


Figure A3: Average Monthly Earnings by Sector, Tax Data and Income Survey, 2005

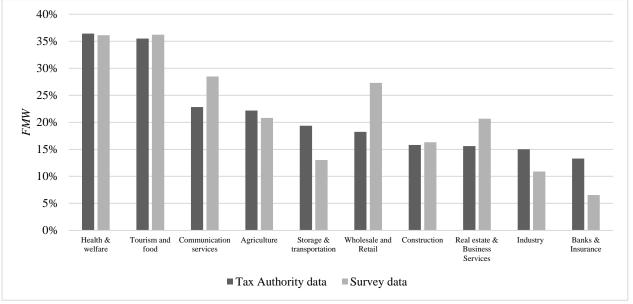
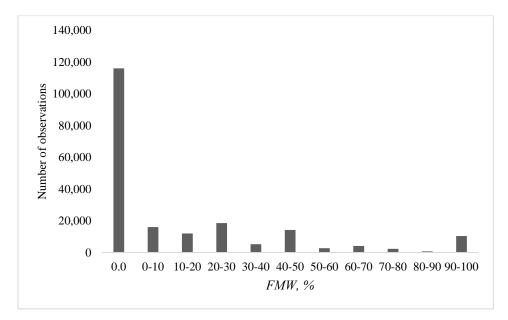


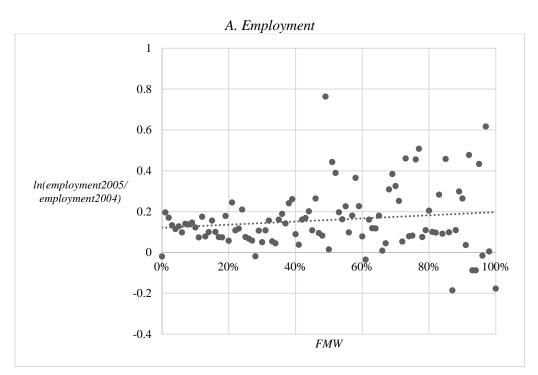
Figure A4: FMW by Sector, Tax Data and CBS Income Survey, 2005

Note: The Central Bureau of Statistics (CBS) income survey includes earnings data.

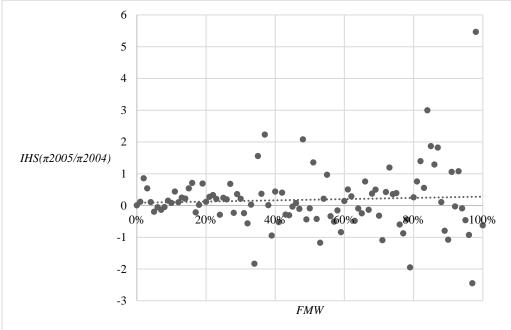
Figure A5: Distribution of Company-Year Observations in the Analysis Sample, by *FMW* Bins











Note: For employment, in a regression of the pre-treatment employment change on *FMW*, the estimated coefficient (standard error) was 0.126 (0.031). For *IHS*(π), the estimate (standard error) was 0.088 (0.200).

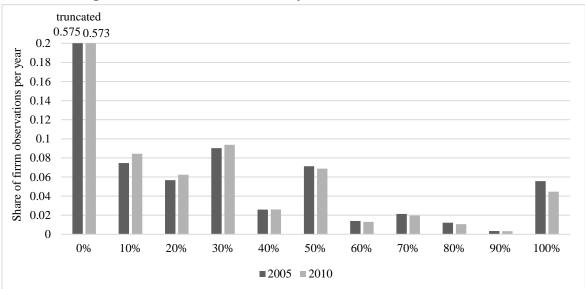


Figure A7: Distribution of Firms by FMW Bins in 2005 and 2010

Notes: For the FMW = 0 bin, we truncated the bars. The actual values are shown. Note that the 0% and 100% values refer to FMW = 0% or 100% exactly, while the other bars are for ranges (e.g., 10% for > 0% and $\le 10\%$).

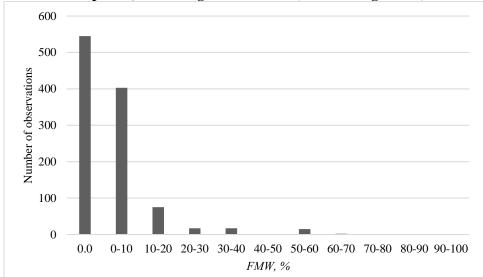


Figure A8: Distribution of Company-Year Observations on Publicly-Traded Companies, According to *FMW* Bins (Parallels Figure A5)

Note: This figure covers publicly-traded companies.

		Number of			
	Number of	employee			
	company	observations on	Employees per company	Profit rate	FMW
Sample	observations	the main job	(min, median, mean, max)	(min, median, mean, max)	(min, median, mean, max)
Full sample of companies	658,804	21,719,245	(1, 6, 33.48, 36, 387)	(-35,331,478, 0.08, -1,667.6, 9,892,492)	(0.00, 0.00, 0.20, 1.00)
matched to employees,					
2004-2010					
Exclude government sector	654,159	19,668,656	(1, 6, 30.54, 32,606)	(-35,331,478, 0.08, -1,676.2, 9,892,492)	(0.00, 0.00, 0.20, 1.00)
Exclude non-profits	593,865	17,185,309	(1, 6, 29.35, 32,606)	(-35,331,478, 0.09, -1,778.3, 9,892,492)	(0.00, 0.00, 0.18, 1.00)
Exclude companies in	588,484	17,128,975	(1, 6, 29.51, 32,606)	(-35,331,478, 0.09, -1,793.8, 9,892,492)	(0.00, 0.00, 0.18, 1.00)
liquidation					
Exclude companies likely to be	542,583	16,358,953	(1, 7, 30.53, 32,606)	(-1, 0.09, 0.15, 1)	(0.00, 0.00, 0.18, 1.00)
holding companies based on					
extreme profit rate measures					
Exclude companies established	519,050	16,018,885	(1, 7, 31.17, 32,606)	(-1, 0.09, 0.16, 1)	(0.00, 0.00, 0.18, 1.00)
after 2005					
Exclude companies with no	356,893	13,120,034	(1, 8, 36.85, 32,606)	(-1, 0.10, 0.16, 1)	(0.00, 0.00, 0.18, 1.00)
earnings data in 2005					
Exclude company observations	202,187	7,255,846	(1, 8, 35.97, 29,694)	(-1, 0.09, 0.15, 1)	(0.00, 0.00, 0.18, 1.00)
in 2006-2008 (treatment period)					

Table A1: Construction of Analysis Sample, and Descriptive Statistics

Note: Pre-treatment years are 2004-2005 and post-treatment years are 2009-2010.

Table A2: Regression Estimates for Minimum Wage Effect onLog Employment, Based on Fraction Affected by MinimumWage (FMW), 2004-2005 and 2009-2010, Excluding FinancialSector and Export-Oriented Companies

Variables	(1)
POST	-0.058
	(0.049)
$FMW \times POST$	-0.108***
	(0.014)
Sector dummies and sector dummies $\times POST$	yes
Firm fixed effects	yes
\mathbb{R}^2	0.924
Ν	192,825

Notes: See notes to Table 2 in the paper. The specification corresponds to column (3) of Table 2 in the paper. The main effect of FMW is absorbed by the firm fixed effects.

Table A3: Regression Estimates for Minimum Wage Effect on Log Employment, with Separate *FMW* Bins, 2004-2005 and 2009-2010, Excluding Financial Sector and Export-Oriented Companies

Variables	(1)	(2)	(3)
POST	0.026^{***}	-0.036	-0.035
	(0.005)	(0.049)	(0.049)
<i>FMW</i> >0-20% × <i>POST</i>	-0.073***	-0.073***	-0.073***
	(0.009)	(0.009)	(0.009)
$FMW 20-40\% \times POST$	-0.133***	-0.130***	-0.130***
	(0.010)	(0.011)	(0.011)
$FMW 40-60\% \times POST$	-0.127***	-0.123***	-0.124***
	(0.013)	(0.013)	(0.013)
$FMW 60-80\% \times POST$	-0.182***	-0.176***	-0.176***
	(0.021)	(0.021)	(0.021)
<i>FMW</i> 80-100% × <i>POST</i>	0.006	0.007	
	(0.016)	(0.017)	
$FMW 80-94\% \times POST$			-0.061*
			(0.037)
<i>FMW</i> 94-100% × <i>POST</i>			0.015
			(0.018)
Sector dummies and	no	yes	yes
sector dummies $\times POST$		-	-
Firm fixed effects	yes	yes	yes
R ²	0.924	0.924	0.924
N	192,825	192,825	192,825

Notes: See notes to Table 4 in the paper. The specifications correspond to columns (1)-(3) of Table 4 in the paper. The main effects of the *FMW* bins are absorbed by the firm fixed effects. The *FMW* × *POST* interactions are relative to *FMW* 0%.

Table A4: Regression Estimates of Minimum Wage Effect on Profits $(IHS(\pi))$, Based on Fraction Affected by Minimum Wage (*FMW*), 2004-2005 and 2009-2010, Excluding Financial Sector and Export-Oriented Companies

	(1)	(2)	(2)	(4)	(5)
Variables	(1)	(2)	(3)	(4)	(5)
		Sample: low initial	Sample: medium	Sample: medium	Sample: high
	Full sample	profits	low initial profits	initial profits	initial profits
POST	-1.990***	4.589***	-1.323***	-1.669***	-1.835***
	(0.227)	(0.503)	(0.354)	(0.387)	(0.571)
$FMW \times POST$	-0.715***	-1.765***	-0.185*	-0.359***	-0.011
	(0.072)	(0.155)	(0.107)	(0.125)	(0.169)
Initial profits dummies	yes	no	no	no	no
Initial profits × POST	yes	no	no	no	no
Sector dummies and	yes	yes	yes	yes	yes
sector dummies ×					
POST					
Firm fixed effects	yes	yes	yes	yes	yes
R ²	0.732	0.643	0.470	0.464	0.481
N	193,237	45,124	46,355	50,365	51,393

Notes: See notes to Table 5 in the paper. The specifications correspond to columns (3)-(7) of Table 5 in the paper. The main effect of *FMW* is absorbed by the firm fixed effects.

Table A5: Regression Estimates of Minimum Wage Effect on Profits
$(IHS(\pi))$, Based on Fraction Affected by Minimum Wage (FMW), with
Separate FMW Bins, 2004-2005 and 2009-2010, Excluding Financial Sector
and Export-Oriented Companies

Variables	(1)	(2)
POST	-1.406***	-2.057***
	(0.038)	(0.227)
$FMW > 0-20\% \times POST$	-0.001	0.078
	(0.060)	(0.060)
$FMW 20-40\% \times POST$	-0.012	0.022
	(0.057)	(0.057)
$FMW 40-60\% \times POST$	-0.187***	-0.185***
	(0.065)	(0.066)
$FMW 60-80\% \times POST$	-0.438***	-0.454***
	(0.107)	(0.108)
FMW 80-100% × $POST$	-0.864***	-0.917***
	(0.086)	(0.086)
Initial profit dummies	yes	yes
Initial profit dummies $\times POST$	yes	yes
Sector dummies and sector dummies $\times POST$	no	yes
Firm fixed effects	yes	yes
R ²	0.730	0.732
N	193,237	193,237

Notes: See notes to Table 6 in the paper. The specifications correspond to Table 6 in the paper. The main effects of the *FMW* bins are absorbed by the firm fixed effects. The *FMW* \times *POST* interactions are relative to *FMW* 0%.

Table A6: Regression Estimates of Minimum Wage Effect on Profits $(IHS(\pi))$, Based on Income Percentile for Incomes of Business Owners, 2004-2005 and 2009-2010, Excluding Financial Sector and Export-Oriented Companies

Variables	(1)	(2)
POST	-3.419***	-3.891***
	(0.178)	(0.287)
Income percentile $(IP) \times POST$	2.391***	2.348***
	(0.181)	(0.181)
Initial profit dummies	yes	yes
Initial profit dummies × POST	yes	yes
Sector dummies and sector dummies $\times POST$	no	yes
Firm fixed effects	yes	yes
R ²	0.723	0.724
N	150,121	150,121

Notes: See notes to Table 7 in the paper. The specifications correspond to columns (1) and (2) of Table 7 in the paper. The main effect of *IP* is absorbed by the firm fixed effects.

Table A7: Regression Estimates for Minimum Wage Effect on Log Earnings, Based on Fraction Affected by Minimum Wage (*FMW*), 2004-2005 and 2009-2010 (Parallels Table 2)

Variables	(1)	(2)	(3)
FMW	-1.210***	-1.124***	•••
	(0.007)	(0.007)	
POST	0.142***	0.142***	0.141^{***}
	(0.002)	(0.025)	(0.024)
$FMW \times POST$	0.190***	0.188***	0.208***
	(0.009)	(0.009)	(0.008)
Sector dummies and sector	no	yes	yes
dummies $\times POST$			
Firm fixed effects	no	no	yes
R ²	0.279	0.338	0.879
N	201,742	201,742	201,742

Notes: See notes to Table 2. Dependent variable is average log(earnings) for workers in each firm and year. Three dots (...) indicates that the variable is absorbed by the firm fixed effects.

Separate FMW Bins, 2004-2005 and 2009-2010 Parallels Table 4				
Variables	(1)	(2)	(3)	
POST	0.117^{***}	0.139***	0.140***	
	(0.003)	(0.024)	(0.024)	
$FMW > 0-20\% \times POST$	0.045^{***}	0.044***	0.044***	
	(0.004)	(0.004)	(0.004)	
$FMW 20-40\% \times POST$	0.068^{***}	0.070***	0.070^{***}	
	(0.005)	(0.005)	(0.005)	
$FMW40-60\% \times POST$	0.093***	0.095***	0.095***	
	(0.006)	(0.007)	(0.007)	
FMW 60-80% × $POST$	0.125***	0.128***	0.128***	
	(0.009)	(0.009)	(0.009)	
FMW 80-100% × $POST$	0.217^{***}	0.220^{***}		
	(0.012)	(0.012)		
FMW 80-94% $\times POST$			0.123***	
			(0.016)	
FMW 94-100% × $POST$			0.231***	
			(0.013)	
Sector dummies and	no	yes	yes	
sector dummies $\times POST$				
Firm fixed effects	yes	yes	yes	
R ²	0.879	0.879	0.879	
N	201,742	201,742	201,742	

 Table A8: Regression Estimates for Minimum Wage Effect on Log Earnings, with

 Separate FMW Bins, 2004-2005 and 2009-2010 Parallels Table 4

Notes: See notes to Table 4. Dependent variable is log(earnings) per firm in each year.

Table A9: Regression Estimates of Minimum Wage Effect on Profits $(IHS(\pi))$, Based on Fraction Affected by Minimum Wage (FMW), 2004-2005 and 2009-2010, Alternative Treatment of Exiters

Variables	(1)	(2)	(3)
	Dropping exiters	Include	Include exiters, $\pi = 0$,
	(Table 5, column (3))	exiters, $\pi = 0$	set negative π to 0
POST	-2.017***	-1.612***	-2.188***
	(0.227)	(0.147)	(0.195)
$FMW \times POST$	-0.720***	-0.295***	-0.394***
	(0.070)	(0.043)	(0.058)
Initial profits	yes	yes	yes
dummies			
Initial profits $\times POST$	yes	yes	yes
Sector dummies and	yes	yes	yes
sector dummies \times			
POST			
Firm fixed effects	yes	yes	yes
\mathbb{R}^2	0.732	0.662	0.611
Ν	202,187	225,313	225,313

Notes: Dependent variable is $IHS(\pi)$ per firm in each year, including zero values for profits after exit in the post period and correcting to zero if profits had negative values. Levels of significance: $10\%^*$, $5\%^{**}$, $1\%^{***}$. Standard errors are clustered at the firm level. *FMW* is measured on a 0-1 scale in the regressions. The main effect of *FMW* is absorbed by the firm fixed effects.

Variables	(1)	(2)	(3)
FMW>0-20%	0.093***	0.074^{***}	0.075^{***}
	(0.005)	(0.005)	(0.005)
FMW 20-40%	0.045***	0.047***	0.050^{***}
	(0.005)	(0.005)	(0.005)
FMW 40-60%	-0.014***	-0.005	-0.011*
	(0.006)	(0.006)	(0.006)
FMW 60-80%	-0.048***	-0.021**	-0.013
	(0.009)	(0.018)	(0.009)
FMW 80-100%	-0.123***	-0.076***	-0.076***
	(0.007)	(0.007)	(0.007)
Initial profits dummies:	no	yes	yes
Low profits		-0.173***	-0.172***
		(0.005)	(0.005)
Mid-Low profits		-0.108***	-0.105***
		(0.005)	(0.005)
Mid profits		-0.033***	-0.033***
		(0.005)	(0.004)
Sector dummies and sector dummies \times	no	no	yes
POST			
R ²	0.016	0.044	0.052
N	57,520	57,520	57,520

Table A10: Regression Estimates of Linear Probability Model for Survival to the Post-period (2009-2010), Based on Fraction Affected by Minimum Wage (*FMW*), with Separate *FMW* Bins

Notes: Dependent variable is a dummy variable for survival, set equal to 1 for companies that appear in the post-period (2009-2010) and 0 otherwise. OLS regression with one observation per firm. Levels of significance: $10\%^*$, $5\%^{**}$, $1\%^{***}$. Effects of *FMW* bins are estimated relative to *FMW* 0%. Estimated effects of the profit quartile dummy variables are estimated relative to "high."

Table A11: Regression Estimates for Minimum Wage Effect on Log Employment, Based on Fraction Affected by Minimum Wage (*FMW*), 2004 only and 2009-2010, Parallels Table 2, column (3)

Variables	(3)
POST	-0.054
	(0.036)
$FMW \times POST$	-0.131***
	(0.011)
Sector dummies and sector dummies $\times POST$	yes
Firm fixed effects	yes
\mathbb{R}^2	0.932
Ν	144,222

Notes: Dependent variable is log(employment) per firm in each year. Levels of significance: $10\%^*$, $5\%^{**}$, $1\%^{***}$. Standard errors are clustered at the firm level. *FMW* is measured on a 0-1 scale in the regressions. The main effect of *FMW* is absorbed by the firm fixed effects.

Table A12: Regression Estimates of Minimum Wage Effect on Profits $(IHS(\pi))$, Based on Estimated Mechanical Cost of Minimum Wage Increase, 2004-2005 and 2009-2010

Variables	(1)
	Full sample
Log relative minimum wage (log RMW)	-13.065***
	(1.713)
Log wage bill for minimum wage workers × log relative	-0.269***
minimum wage	(0.093)
Initial profits dummies	yes
Initial profits $\times POST$	yes
Sector dummies and sector dummies $\times POST$	yes
Firm fixed effects	yes
\mathbb{R}^2	0.732
Ν	202,187

Notes: Dependent variable is $IHS(\pi)$ per firm in each year. Specification includes firm fixed effects. *Log relative minimum wage* is defined as zero (log(1)) for the pre-period and log(1.154) for the post-period, given the 15.4% increase in the minimum wage. The interaction measures the percent change in profits for a one-percent higher effect of the minimum wage increase of the wage bill (based on minimum wage workers in 2005). Levels of significance: 10%^{*}, 5%^{**}, 1%^{***}. Standard errors are clustered at the firm level. *FMW* is measured on a 0-1 scale in the regressions. The main effect of $log(WBMW_i)$ in equation (3) is fully absorbed by the firm fixed effects.