Can Minimum Wages Cause a Big Push? Evidence from Indonesia

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Abstract

Big Push models suggest that local product demand can create multiple labor market equilibria: one featuring high wages, formalization, and high demand and one with low wages, informality, and low demand. I demonstrate that minimum wages may coordinate development at the high wage equilibrium. Using data from 1990s Indonesia, where minimum wages increased in a varied way, I develop a difference in spatial differences estimator which weakens the common trend assumption of difference in differences. Estimation reveals strong trends in support of a big push: formal employment increases and informal employment decreases in response to the minimum wage. Local product demand also increases, and this formalization occurs only in the non-tradable, industrialize industries suggested by the model (while employment in tradable and non-industrialize industries also conforms to model predictions).

KEYWORDS: Minimum Wage, Big Push, Spatial Regression Discontinuity
JEL CODES: O1, J8

1 Introduction

Wage floors traditionally play a simple role in partial equilibrium models of firm behavior: higher wages imply higher marginal costs, lower labor demand, and lower profits. This simple but powerful intuition has guided a great deal of economic analysis and intuition regarding the likely effects of externally imposing higher wages, for example

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through legal minima. However, if local consumption is an important component of product demand, then the story quickly becomes much more nuanced. In that event, higher wages still remain a cost borne by firms; however, they may also increase the size of the local market which, in turn, can increase either quantities sold or prices and positively influence profitability and labor demand.

This insight is at the core of classical big push modeling. For example, Rosenstein-Rodan (1943) suggests that "if... one million unemployed workers were taken from the land and put... into a whole series of industries which produce the bulk of the goods on which the workers would spend their wages.... it would create its own additional market, thus realizing an expansion of world output." Rosenstein-Rodan clarifies: because of this potential of market formation, wages paid by local firms have the externality of boosting local product demand. The same intuition is formalized in more recent big-push modeling, such as Murphy, Shleifer, and Vishny (1989; henceforth MSV). MSV demonstrate that, if workers require a wage premium to supply labor to the formal sector, than multiple equilibria are possible with industrializing and non-industrializing equilibria coexisting for some parameter values. In other words, if firm profits are tied to local consumption, then firms create an externality by paying high wages: the size of the market for other firms increases with worker wages and wealth.

In popular culture, this insight is usually attributed to Henry Ford, who famously doubled wages at his auto plants to five dollars a day in January 1914. Ford wrote "our own sales depend on the wages we pay. If we can distribute high wages, then that money is going to be spent and it will serve to make... workers in other lines more prosperous and their prosperity is going to be reflected in our sales" (Ford 1922 p. 124-127, as qtd. in Taylor 2003). The idea that a firm could unilaterally use wages to increase demand for its own product enough to offset wage cost seems unlikely and is dismissed by Rosenstein-Rodan (1943), and Ford’s own statements on the matter were inconsistent (Taylor 2003). As a result, economists have traditionally interpreted Ford’s wage innovations as an effi-
ciency wage (e.g. Raff and Summers 1989). However, while it may be not be in a firm’s private interest to pay higher wages, a key intuition of coordinated development is that a firm’s profits may depend on the wages paid by local firms more generally. This insight was dubbed the "high wage doctrine" and endorsed by policy makers and economists during the great depression, who saw stimulating consumer demand through wages as an important policy vector to combat the economic downturn (Taylor 2003 extensively documents the policy discussion). That is, this high wage doctrine reemphasizes this wage externality. While profits may be decreasing in a firm’s own wages, they may in fact be increasing in the wages of other nearby firms, which can happen so long as prices do not fully adjust\(^1\). In the MSV modeling, this is achieved through a bisected market of industrialized high productivity monopolists with prices bounded by a low-productivity informal fringe, though one can also describe insufficient price adjustment if some goods are tradable and hence have prices which are unaffected by innovations in local wages\(^2\).

In this paper, I add a minimum wage to the MSV big push model, and show that it can change a multiple-equilibrium setting to one where only the industrialized equilibrium remains as long as some firms are constrained by the minimum wage. To my knowledge, this is the first paper to explore the consequences of minimum wages in a big push framework\(^3\). I then discuss minimum wages in 1990s Indonesia, where real minimum wages rose rapidly in a varied way and then dropped quickly with the inflation rate in the South East Asian financial crash. These changes combine with flexible government policy to create substantial spatial and intertemporal variation in the wage rate. I demonstrate that minimum wages did disturb wage distributions, and then present some suggestive evidence that the employment repercussions of these minimum wages do not

\(^1\)A critique of the high wage doctrine involving the quantity theory of money is presented in Taylor and Selgin (1999). This critique is not relevant to the subnational analysis below.
\(^2\)Though not pursued here, if consumers have CES utility over tradables and non-tradables, and firms are competitive with Cobb-Douglas production, then one can show that so long as tradable and non-tradable goods are sufficiently complementary, the price of non-tradable goods rises enough in the presence of minimum wages to motivate increased employment.
\(^3\)A search of the keywords "minimum wage" and "big push" on econlit turns up zero hits as of 3/1/2011.
accord well with predictions from neoclassical modeling. In particular, examining districts on alternate sides of a minimum wage policy border, areas with higher minimum wages see a greater proportion of the population employed in full time wage work and fewer employed as relatively uncovered entrepreneurs. To allow more robust estimates of the effects of minimum wages in Indonesia, I develop a difference in spatial differences estimator based on the spatial fixed effects in Conley and Udry (2010) and Goldstein and Udry (2009) which weakens the standard difference-in-difference identification assumption of similar trends to allow trends to be non-parametric over time but spatially local. This estimation procedure reveals several surprising trends which are congruent with a big push in Indonesia. First, they support the earlier evidence in documenting that formal sector work increases with the minimum wage, while informal sector work decreases, which is predicted by the MSV model but is the opposite of the prediction a partial equilibrium analysis would achieve. I also document that local real demand increases, particularly for non-food items, suggesting that minimum wages are boosting demand, the causal channel highlighted in the model. The big push model also has strongly heterogeneous predictions across industries, depending on tradability and industrialization potential. I look at tradable and non-tradable, formal and informal manufacturing firms to document that while the expansion of the formal sector and retraction of the informal sector are powerful enough trends to be statistically significant for the labor force on average, they are observed in non-tradable manufacturing but not tradable manufacturing, consistent with model predictions. In contrast, in services (where industrialization potential is minimal) there is an expansion of informal work, while retail consolidates in response to rising minimum wages. Finally, I document that trends seem unlikely to be created through reverse causality, migration, or monopsonistic markets, and I close by suggesting some explanations for why these policies may have been successful in 1990s Indonesia, but were widely considered failures in depression-era America.
2 Model

The model is a straightforward adaptation of Murphy, Shleifer, and Vishny (1989). There are Q goods, each produced by a distinct industry. A representative consumer has Cobb-Douglass Utility over these goods, \( U(x) = \sum_{x=1}^{Q} \ln(x) \). In MSV’s original case, each good could be produced by two production technologies: an informal fringe, which produces and remunerates at unity, and an industrializing, increasing returns to scale technology which requires an input of \( F \) units of labor to access but then can produce \( \alpha > 1 \) units with each additional unit of labor. Because the informal fringe will compete away any prices above unity, prices are fixed at that level.

Finally, MSV suppose that industries feed local demand. With Cobb-Douglass utility, the consumer will spend \( y/Q \) on each good; at unit prices this is the production amount which requires either \( y/Q \) units of fringe labor or \( y/Q + F \) units of industrialized labor. Therefore, if the monopolist pays wage \( w \), she earns \( \pi \) profits

\[
\pi = \frac{y}{Q} \left( \alpha - \frac{w}{\alpha} \right) - Fw \tag{1}
\]

For our purposes, suppose that \( y \) is wage income (very similar results are achieved if \( y \) is the sum of wage income and profits). Then, if all firms industrialize,

\[
y = \frac{wy}{\alpha} + QFw + L - \frac{y}{\alpha} - QF \tag{2}
\]

or, solving for \( y \),

\[
y = \frac{L + QF (w - 1)}{1 - \left( \frac{w-1}{\alpha} \right)}
\]

Thus, industrialization is profitable for the last firm, if

\[
\frac{L + QF (w - 1)}{Q \left( 1 - \left( \frac{w-1}{\alpha} \right) \right)} \left( \frac{\alpha - w}{\alpha} \right) - Fw > 0 \tag{3}
\]
If no one industrializes, \( y = L \); thus industrialization is unprofitable for the first firm if

\[
\frac{L}{Q} \left( \frac{\alpha - w}{\alpha} \right) < F
\]

Which suggests that for certain values of \( F \), there can be multiple equilibria whenever \( w > 1 \).

2.1 Minimum wages and the Big Push

MSV motivate the potential for multiple equilibria from the possibility that workers will require a wage premium to sell labor to industrializing firms, perhaps because of worse working conditions or perhaps because of needs related to urbanization. A key point is that in this model, minimum wages can encourage development. There are two obvious channels for this to take place:

First, if the equilibrium with no minimum wage is selected, the presence of a minimum wage may allow coordination at the superior industrializing equilibria. The idea behind this is simple, and motivated by MSV’s observation that expectations of development become self-fulfilling in a model like this. Both equilibrium in this model are ex ante equally sensible. The minimum wage may make formal the expectation that other firms will pay higher wages, which would push firms towards the industrialized equilibrium. While I don’t formally develop this equilibrium selection rule here, one might think it would be particularly credible in a time of internationally financed economic growth like 1990s Indonesia.

Second, if the minimum wage can be enforced on some workers in the undeveloped equilibrium (perhaps because some industries are owned by monopolists but able to op-
erate at low wages, for example, if the wage premium that workers need is heterogeneous, then a minimum wage can rule out the undeveloped equilibrium in some parameter spaces. Specifically, suppose δL earn the minimum wage, w, w ≥ w. As a result, when the next firm (after the first δ) is considering paying a higher wage, it observes wage income

\[ \delta \bar{w}L + (1 - \delta) L \]  

and is willing to invest if

\[ \frac{\delta \bar{w}L + (1 - \delta) L \left( \frac{\alpha - \bar{w}}{\alpha \bar{w}} \right)}{Q} > F \]  

It is easily verified that there are levels of F such that there would be undeveloped equilibria without a minimum wage but that there is none with the minimum wage if

\[ \delta \bar{w} + (1 - \delta) > \left( \frac{\alpha - \bar{w}}{\alpha \bar{w}} \right) \left( \frac{\bar{w}}{w} \right) \]  

So that if the minimum wage successfully binds on at least some workers, then it is possible for the economy to only allow industrializing equilibria. This is certainly true locally at \( \bar{w} \approx w \), and it is easy to verify that a similar result holds for \( \bar{w} < w \).

2.2 Industry Heterogeneity

The MSV model, above, assumes that industries are homogeneous. They have two attributes which are important to demand-driven development. First, they are untradable and their product is consumed locally (that is, they are "articulated" in the language of de Janvry and Sadoulet (1983)). Second, they all have the potential for industrialization. Homogeneity in these assumptions makes the model transparent and tractable; however, to develop a broader number of empirical restrictions it is necessary to loosen them. Suppose, therefore, that the MSV modeling assumptions of untradability and potential for
industrialization are relevant for fraction $\eta$ of industries. Fraction $\gamma$ are tradable (and all industrializable). For these firms, demand is not related to local income, thus, they receive some return $R$ from using $\bar{\gamma}L$ units of labor which they demand inelastically so long as they are profitable. $1 - \gamma - \eta$ are neither industrializable nor tradable (for example, services). These industries can only be produced through the informal fringe, and thus their labor demand is always equal to $y/Q$.

In this scenario, non-tradable, industrializable firms behave very similarly: we now have two equilibria if

$$\frac{L (1 + \bar{\gamma} (w - 1))}{Qw} \left( \frac{\alpha - w}{\alpha} \right) < F < \frac{L (1 + \bar{\gamma} (w - 1))}{Q (w - \eta (w - 1))} \left( \frac{\alpha - w}{\alpha} \right)$$

(9)

where the non-industrializable equilibrium can again be ruled out by minimum wages if there is some compliance. Tradable firms, whose product demand is not subject to local demand, behave differently: minimum wages evoke either a zero or a negative employment response. Since the industrialized equilibrium is associated with a higher $y$ than the non-industrialized firms, this model would suggest that untradable industries which cannot industrialize increase their labor demand.

### 2.3 Empirical Predictions

The modeling above highlights the potential of a labor standard to produce demand-driven development. It also highlights that a broad number of conditions are necessary for this potential to be realized, including the presence of unutilized technological potential, the articulation of local firms with consumer demand, and that minimum wages be appropriately set (one can easily verify that very high minimum wages can serve to eliminate the industrializing equilibrium, as well). In the empirical analysis, I will take several predictions to the data. First, if minimum wages are inducing big-push-style development motivated by local demand, we should observe that formal sector employment in
untradable industries with the potential for industrialization increases while informal sec-
tor employment in these industries decrease in response to an increased minimum wage. Examining untradable industries with no potential for industrialization, such as services, we should observe informal sector employment to increase when minimum wages increase. Third, examining tradable industries, we should observe no increase, and poten-
tially a decrease, in formal sector employment in response to minimum wages. Finally, an additional prediction of this modeling is that minimum wages are associated with growth in local product demand: indeed, this is precisely the mechanism which drives this model, and one that I test below.

3 Minimum Wages in Indonesia

The first half of the 1990s was a time of rapid economic expansion in Indonesia. Along-
side this economic expansion, minimum wages grew very quickly in 1990s Indonesia in a varied way. Commentators suggest that there were two primary pressures which caused this rise in minimum wages: pressure from the US government, associated with concur-
rent anti-sweatshop activism; and a desire to enforce a national minimum wage which could purchase a better-than-subsistence consumption bundle (see Rama 2001 for a more extensive discussion). The schedule and timing of these minimum wage increases was determined by regional tripartite local councils, including representatives of the ministry of manpower as well as local employers and employees; this procedure led to a varied growth in minimum wages with 32 different minimum wage regimes countrywide. The regimes are either provinces or collections of districts within a province. Real minimum wages (averaged across the country) are presented in figure 1; this graph makes clear that minimum wages doubled in real terms between 1990 and 1997 before falling as prices rose alongside the financial crisis. Other studies of these minimum wages (Alatas and Cameron 2008; Harrison and Scorce 2010; Rama 2001) have used difference-in-differences
or control function approaches and found that these minimum wages reduced employment, though sometimes only for a subset of firms, and all using data prior to the crisis. Most similar to this study, Alatas and Cameron (2008) completed a matched difference-in-differences on firms inside or outside of Jakarta, and found that small firms reduced employment in response to minimum wages but that there was no effect on large firms.

If these minimum wages had labor market effects, it seems necessary that they affected the wage distribution. Given the high level of informality, we may be concerned about whether minimum wages were in fact enforceable in 1990s Indonesia. Here, I construct wage histograms to verify both that the minimum wage did distort the wage distribution, and that this minimum wage was sufficiently high to evoke credible demand responses. Figure 2 presents the log wage histogram across years and Indonesia, where wages are normalized so that 0 indicates a log wage equal to the log minimum wage in that wage group. Clearly, enforcement is far from perfect, as about a third of full-time wage workers earn below the minimum in their district. However, there is also a clear jump in wage densities at the minimum wage, indicating that for some jobs at least, the minimum wage does affect wages. Analysis below will focus on differences between nearby districts. Figures 3a and 3b present minimum wage histograms for individuals who live within 25 miles of a minimum wage boundary. In Figure 3a, wages are normalized relative to the minimum in the higher wage regime at each border. In Figure 3b, they are normalized relative to the minimum in the lower wage regime at each border. As the reader can see, there is a clear jump in density at the minimum in the own-regime in all cases which is not present in the alternate wage regime. In other words, people who live nearby each other but under different minimum wage laws have wages distorted according to their own laws, so that laws are enforced discontinuously at the border.

A first order analysis would ask what happens to employment locally at the border. Figure 4(A) plots mean full time wage work against distance to the border of a minimum wage regime, where a positive distance indicates that the district is located at the side of
the border with the higher minimum wage and water distances are presumed to be infinite. In 1993, there appears to be a strong spatial trend in employment on the high wage side of the border, making it difficult to draw clear conclusions. However, by 1997 and 2000, there is a clearly discernible and statistically significant increase in full time wage work on the side of the border with higher minimum wages. Figure 4b plots mean entrepreneurship against distance to the border. In this figure, there is a clearly discernible and statistically significant drop in entrepreneurship associated with minimum wages by 1997. Taken at face value, these figures suggest that minimum wages are related to an increase in full time wage work, which is likely to be formal sector employment, and a decrease in entrepreneurship, which is likely to be informal and uncovered by the minimum wage. These findings are striking, and suggest that a neoclassical partial equilibrium model may not be appropriate for 1990s Indonesia. However, it is also possible that there are persistent differences correlated with minimum wage policy, and so further analysis will take this possibility seriously.

4 Empirical Strategy

Employment effects of labor regulation are a traditional topic in economics, and are surveyed in Blau and Kahn (1999); Nickell and Layard (1999); and Freeman (2009) in developing countries. The majority of these estimates are constructed through either difference-in-difference style estimators (e.g. Besley and Burgess 2004, Bertrand and Kramarz 2002; Neumark and Wascher 1992), comparisons across small regions of space (e.g. Dube, Lester, and Reich 2010), or some combination of these two approaches (e.g. Card and Krueger 1994; Magruder 2011). Dube, Lester, and Reich (2010) summarize the distinctions between these two approaches, which in minimum wage studies in the US have

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4 These figures use a bandwidth of 40 miles as that is approximately the median distance at which observations which border any alternate minimum wage regimes only border a single alternate minimum wage regime. That is, at higher bandwidths the majority of observations which border alternate wage regimes border more than one of these regimes.
tended to find small or positive employment effects of minimum wages under close spatial comparisons and negative employment effects using the difference-in-differences strategy.

Identification by spatial discontinuity and difference in differences posit two alternate forms for endogeneity. In either case, the concern is that minimum wage laws are different in different districts because of some underlying characteristics of the labor market. In the case of Indonesia, we may be most concerned that the tripartite councils succeeded in their explicit goals of maintaining similar real wages and that the differences in nominal minimum wages focus on underlying differences in the price paths of different districts. Given that local inflation rates seem likely to be related to economic activity, this could lead to bias in estimation which presumes common trends across treatment and control areas. A spatial regression discontinuity approach would take advantage of economic theory suggesting that the capability to trade should render prices and other economic conditions similar across nearby districts. In this case, nearby labor markets would make good comparison groups if policy makers target minimum wage policy in response to aggregate labor market characteristics rather than specific characteristics of labor markets at the borders. The assumption on targeting at the aggregate (policy group) level rather than the level of individual districts seems appealing and will be tested in the robustness section. That said, a spatial discontinuity strategy is vulnerable to any differences between nearby districts, including differences in legislative environments. Difference-in-difference identification relies on the familiar assumption that trends would remain the same in the absence of changes in minimum wage law. This assumption seems likely to

Moreover, since the hypothesized difference is one of local product markets and consumption demand, such an approach also requires that consumption demand is more localized than the confounding market characteristics which minimum wages may be targeted on. Of course, product markets will vary in their localization depending on their tradability, and below empirical restrictions based on this fact are developed. Nonetheless, there is good reason to expect that product markets are sufficiently local, at least for some goods. The literature has documented that the markets for consumer goods and restaurant are tremendously local (with large market shares within five miles) even in the US, where infrastructure is strong and access to transport is good (e.g. Anderson and Matsa 2010 for restaurants, Holmes 2009 and Jia 2005 for consumer goods).
hold in Indonesia if the changing time paths of minimum wages are due more to external pressure or the somewhat arbitrary exercise of bureaucratic power, but is less tenable if these time paths of policy are indeed the result of careful planning and targeting.

This paper will estimate the effects of minimum wages on employment in three ways. First, I consider a standard difference-in-difference specification,

\[ y_{it} = \alpha_i + \delta_t + \beta (\text{minwage}_{it}) + \gamma X_{it} + \epsilon_{it} \] (10)

where \( y_{it} \) is employment in district \( i \) in year \( t \), \( \text{minwage}_{it} \) is the log real minimum wage, and \( X_{it} \) is a vector of controls. As discussed above, this approach will yield consistent estimates of the effects of minimum wages on employment if changes in the minimum wage are unrelated to changes in local labor market conditions. Following, I use spatial-temporal fixed effects to estimate a spatial discontinuity, following Conley and Udry (2010), Goldstein and Udry (2009), and Magruder (2011). This procedure allows minimum wages to be related endogenously to observations, but requires that the endogeneity is similar among spatially proximate districts. More specifically, it specifies that the "true" underlying structural equation is

\[ y_{it} = \delta_t + \beta \text{minwage}_{it} + \gamma X_{it} + \nu_{it} + \epsilon_{it} \] (11)

where \( \nu_{it} \) is unobserved but related to \( \text{minwage}_{it} \) and \( \epsilon_{it} \) is exogenous. The idea behind this specification is that labor market characteristics at time \( t \) are related to the level of the minimum wage. However, because of the potential for local trade, those characteristics are similar at time \( t \) for other districts within some radius \( R \), so that if we call this set of districts \( R(i) \), then \( E[\nu_{it} | i' \in R(i), X_{it}, \text{minwage}] = E[\nu_{it} | X_{it}, \text{minwage}] \). Since every district-year will have a unique radius and unique labor market effect, it is not possible to represent these effects as a matrix of dummy variables as is conventionally done in fixed effects analysis. However, we can still treat the spatial effects as nuisance parameters and
estimate the within estimator,

\[ y_{it} - \sum_{i' \in R(i)} \frac{y_{i't}}{n_{R(i)}} = \beta \left( \text{minwage}_{it} - \sum_{i' \in R(i)} \frac{\text{minwage}_{i't}}{n_{R(i)}} \right) + \gamma \left( X_{it} - \sum_{i' \in R(i)} \frac{X_{i't}}{n_{R(i)}} \right) \]

where \( n_{R(i)} \) is the number of district-year observations within radius \( R \) of district \( i \).

Since \( E[v_{i't} | i' \in R(i), X_{it}, \text{minwage}] = E[v_{it} | X_{it}, \text{minwage}] \), the endogenous component of the error term disappears in expectation and so if we make an assumption of strict exogeneity similar to those used elsewhere in fixed effects analyses \( E[\varepsilon_{i't} | \text{minwage}_{i't}, X_{i't}] = 0 \forall i', i'' \in R(i) \), then equation (12) will consistently estimate the effects of minimum wage law.

Of course, there may be some circumstances in which spatial discontinuity may not hold. For example, provincial boundaries may be correlated with other legal differences which affect local labor markets in a discontinuous way, or infrastructure may imperfectly link two (physically proximate) differences. For this reason, I also estimate a difference in spatial differences (hence DSD), which loosens the assumptions of both difference-in-differences and spatial discontinuity estimation. Specifically, I estimate

\[ y_{it} - \sum_{i' \in R(i)} \frac{y_{i't}}{n_{R(i)}} = \left( \alpha_i - \sum_{i' \in R(i)} \frac{\alpha_{i'}}{n_{R(i)}} \right) + \beta \left( \text{minwage}_{it} - \sum_{i' \in R(i)} \frac{\text{minwage}_{i't}}{n_{R(i)}} \right) \]

\[ + \gamma \left( X_{it} - \sum_{i' \in R(i)} \frac{X_{i't}}{n_{R(i)}} \right) + \varepsilon_{it} - \sum_{i' \in R(i)} \frac{\varepsilon_{i't}}{n_{R(i)}} \]

This approach will be consistent under either the assumptions of difference-in-difference estimation or the assumptions in spatial discontinuity estimation, and can be understood as a generalization of either. From the perspective of spatial discontinuity, this approach asks how labor market characteristics are discontinuously different between nearby districts in year \( t \) as a function of sharp differences in the minimum wage in that year, but
controls nonparametrically for differences between nearby districts which persist over
the length of the panel. From the perspective of difference-in-differences, this approach
loosens the assumption on symmetric trends: rather than requiring districts to have labor
market trends which are unrelated to the presence of minimum wage law (except through
causal mechanisms), this approach allows observations to have non-parametric trends, so
long as those trends are shared among nearby districts (as theory would suggest). Since
this estimation strategy will be consistent whenever either a spatial discontinuity or a dif-
ference in differences produces consistent estimates, these will be my preferred results.

4.1 Standard Errors

The discussion of estimation strategies emphasizes two dimensions of correlation within
observations. First, observations of districts seem likely to be characterized by a seri-
ally correlated component of the error term, which suggests the need for clustering at the
observation level (e.g. Bertrand, Duflo, and Mullainathan 2002). In practice, conven-
tion normally clusters at the level of policy groups. Secondly, the theoretical discussion
suggests that labor markets are similar across small regions of space, suggesting the desir-
ability of spatial clustering (Conley 1999). Moreover, the spatial demeaning will generate
such a correlation even in the absence of an underlying one. Thus, it is important to
cluster over space as well. In practice, I allow observations which are physically close
(within 0.5 degrees latitude and longitude) or in the same policy group in any year to
be related. This is a special case of the Conley (1999) errors, where observations are de-
finite as economically close if they are either physically close or in the same policy group,
and also can be understood as the more computationally intensive procedure outlined in
Cameron, Gelbach, and Miller (2009). Small cluster numbers may be an issue as well,
particularly in one of the data sets; I return to this possibility in the Data section.
5 Data

Two data sets are used in this exercise, and summary statistics of the main variables from each are provided in Table 1. The first data set is waves 1, 2, and 3 of the Indonesia Family Life Survey (IFLS), a panel data set which began in 1993 with 7,224 households and grew to include all a large number of split-off households so that 10,435 households were interviewed in 2000. Sample selection was completed so that baseline households are representative of 83% of Indonesia’s population. Waves 1, 2, and 3 take place in 1993, 1997, and 2000. The IFLS is further notable for its low attrition rate, with 95% of original households recontacted in 2000, and over 90% surveyed in all three years. Individual level data allowed the construction of wage histograms used above, and population-level data will be useful in allowing measurement of informal economic activity as well as controlling for demographic changes. Moreover, it will allow control for endogenous migration effects; thus, we will be able to answer whether individual outcomes change as a result of changes in minimum wage policy. The second dataset, Statistics Industry (SI), is an annual census of all manufacturing firms in Indonesia with at least 20 employees collected by the Indonesian government statistical agency. Data are aggregated to the district (Kabupaten) level; coverage includes 209 districts annually for 10 years (1990-2000, with 1997 omitted due to data unavailability). The manufacturing census has two primary strengths: first, it fully characterizes formal employment in manufactures over the ten years of coverage. Thus, we can directly examine formal sector employment using that data. Second, it allows a great deal of firm-specific data, allowing us to disentangle, for example, employment in exporting firms from those which serve domestic markets. Employment numbers of firms in a particular category from these data are therefore the sum of employment across firms in that category in that district in that year, with zeros imputed for districts where zero firms register with the census. Given that manufacturing firms with greater than 20 workers are legally obligated to register with the census, this imputation seems reasonable; results which use only the (endogenously changing)
sample of districts with positive employment numbers are available from the author. Interestingly, spatial discontinuity approaches remain broadly similar using this approach, but difference-in-difference approaches reveal different estimates, as one would expect if the intertemporal attrition of districts is representing economically meaningful information. The IFLS sampling procedure excludes some minimum wage groups, so rather than 32 groups of coverage, 21 are represented in these data. Thirty-two is probably sufficient for cluster asymptotics to represent reasonable approximations; however, twenty one is small enough that we may be somewhat concerned. There is an active debate on the robustness of a variety of measures to prevent overrejection of the null hypothesis in the event of small numbers of clusters, with Cameron, Gelbach, and Miller (2009) finding that an ad-hoc approach of using critical values from a $t_{G-2}$ distribution (where $G$ is the number of clusters) works reasonably well in the range of 20 clusters. The 1%, 5%, and 10% critical values for a $t_{19}$ are about 2.87, 2.1, and 1.73; as the reader can verify, the main results of this paper are robust to using these more conservative critical values. Moreover, the SI, a full coverage census, is not subject to this concern, and reports similar results.

IFLS and SI geographical locations were determined from actual coordinates of IFLS sample communities and internet resources; these are averaged to find district level coordinates. District codes which changed over time are mapped using the IFLS documentation supplemented with Olken (2009).

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6This distribution recalls the results of Donald and Lang (2007), who derive that the $t_{G-2}$ distribution is in fact the correct distribution under a few specific assumptions.

Cameron, Gelbach, and Miller also report simulation evidence that a wild cluster bootstrap-t test performs best in the presence of low numbers of clusters. However, these methods haven’t been adapted for multiple non-nested clusters which I argue above are necessary. In particular, this application has a small number of clusters in the number of policy groups and a large number of spatial clusters. The Wild cluster bootstrap-t test derives a Wald test statistic where the observed coefficient divided by unclustered standard errors is compared to bootstrapped realizations. An intuitively appealing approach would allow a second (spatial) dimension to the cluster to generate a test statistic where the observed coefficient is divided by spatially clustered standard errors and compared to analogous bootstrapped realizations. Under this approach, the main results of the paper are slightly noisier but generally maintain statistical significance, with p-values being particularly consistent at the 25-mile bandwidth, possibly lending support to that bandwidth choice. However, it is beyond the purpose of the paper to derive the asymptotic properties of this test, and so preference is given to the $t_{G-2}$ approximation.
6 Results

The theoretical modeling suggested a number of strong predictions for expected results. First, and potentially most striking, is that formal sector employment could increase in response to minimum wage increases, at least for firms which create products which are not tradable, and which have the potential for industrialization. Nearly as striking, the big push modeling suggests that these firms should see a reduction in informal employment, as the formal sector crowds out lower productivity informal work. Here, I undertake several forms of analysis to test these hypotheses. First, I examine average trends in formalization and employment; next, I verify that the causal channel of increased local expenditures does respond to minimum wages; and then I explore industry heterogeneity suggested by the model.

Of the two datasets, the IFLS, as a population survey, presents better evidence on what most people are doing in all sectors, including informal work. While the data do not include identifiers for the legal classification of work, we can divide workers into full-time wage-earners, part-time wage-earners, and the self-employed. It seems reasonable to expect that the self-employed are primarily engaged in informal work, while a greater proportion of full-time wage-workers should be engaged in formal sector work. Part-time wage-workers may be formally or informally employed. Thus, we can get some good evidence of the effects of the minimum wage on the informal sector by examining the effects of minimum wages on entrepreneurship, and some less precise estimates of the effects of minimum wages on formal employment by examining the effects of minimum wages on full-time wage work, though we will be unsure to the extent to which this wage work is truly in the formal sector. In contrast, the SI presents a complete picture of large (greater than 20 employees) manufacturing firms. This includes some firms which are not legally registered, and thus can provide some evidence on the informal sector as well. SI results will only be relevant to the manufacturing industry, however, and so that dataset can’t be used to summarize overall employment trends.
6.1 Employment Trends

Table 2 presents regressions of various categories of employment on real minimum wages (normalized by a national price index) using the IFLS and all three estimation strategies, difference in difference (DD), spatial difference (SD), and difference in spatial differences (DSD). The bandwidth utilized in the spatial estimation is ultimately a choice variable. In these tables estimates are constructed using a 15, 25, and 50 mile bandwidth to demonstrate robustness to bandwidth choice. The DSD estimator, which was the most robust to underlying heterogeneity, shows large positive effects of minimum wages on full time wage work. The DSD estimator suggests that a 100% increase in real minimum wages is associated with approximately a 10% increase in full time waged employment, which is statistically significant as long as bandwidths are sufficiently large\(^7\). The SD estimator finds effects which are similar in sign, though often larger in magnitude as the DSD estimator, a trend which continues throughout most of the analysis in this paper.

Self Employment provides an opposite trend. The DSD estimator reveals that there is 10-20% less self employment in response to a 100% increase in the minimum wage. Once again, the SD estimates are similar. In contrast, the DD estimates are the opposite sign and statistically significant. Given that the difference in spatial differences loosens the difference-in-difference identification assumption, we can infer that the common trends assumption is inappropriate in this case and that a difference in difference approach could lead to mistaken inference.

The divergence of results between a conventional difference-in-difference strategy and the difference-in-spatial-differences strategy warrants additional comments. First, it highlights why results in this study are different from those of previous studies of the Indonesian minimum wage. Under a common trends assumption, it appears that minimum wages are either unchanging or are shrinking formal economic activity, while in-
formal activity is increased (in fact, if samples are restricted to the SI data in the years considered in Harrison and Scorce (2010), Rama (2001) and Alatas and Cameron (2009), then the DD estimation reveals negative significant results in formal employment, as it does in those studies, while DSD estimates are similar to those here). In contrast, when trends are allowed to differ non-parametrically subject to the restriction that they are spatially local as in the DSD estimation, there is a clear positive effect of the minimum wage on formal employment and negative effect on informal employment. Given that the DSD trends assumption is strictly weaker than the DD assumption, so that whenever the DD estimation is valid, the DSD is as well, we must prefer the DSD estimates. Second, this divergence speaks directly to whether reverse causality may be important. Reverse causality in this case would happen if the minimum wage was carefully targeted at regions with higher secular trends in economic growth. Obviously, both the DD and the DSD estimates would be robust to targeting on the level of economic development. Given that the same wage profile is applied to a large regions, if regional trends were affecting the regional evolution of the minimum wage, we would expect the positive relationship to show up in the DD estimates most strongly since the DD estimates reflect the average trends in the purview of the tripartite council. The fact that they show up only in the DSD estimates suggests that if reverse causality is important, it means that minimum wage profiles are being established for entire regions based on labor market trends in border districts and not the trends elsewhere in the region, which requires some specific and curious incentives of politicians. This is strong prima facia evidence that the effects of the minimum wage documented in this paper are not being driven by reverse causality, which will be developed further below.

Table 2 suggests that minimum wages are inducing a shift towards formalization of labor contracts, and potentially a shift towards formalization of economic activity. There are some limitations to using the IFLS to directly infer whether these new full time workers are employed in the large, formal firms modelled above. Most notably, firm size data
is only available in two of the three survey years, and only coarse (two-digit) occupation data has been provided to this point (with no industry data provided). Below I’ll use the fact that some of these occupations do correspond closely to one-digit industries; however it would be nice to observe whether the new full time wage jobs which are being created are the sort of job which takes place in formal firms. Here, I use the 1997 and 2000 firm size data to predict the distribution of firm sizes that employ people who work in each occupation. I then divide these into occupations which tend to work in small firm occupations (where less than 10% are employed in firms with more than 20 workers), and large firm occupations with greater than 10% employed in these larger firms. According to this definition, 57% of workers work in small firm occupations and 43% work in large firm occupations, and these estimates are robust to alternate sensible definitions. Table 3 presents the results of estimating the DD, SD, and DSD approaches, where the dependent variable is working in a small firm or large firm occupation. The SD and DSD estimates are consistent in demonstrating that employment is dropping in the types of occupations which work in small firms and increasing rapidly in the types of occupations which work in large firms in response to higher minimum wages, consistent with a formalization response to minimum wages.

6.2 Market Expansion

The analysis above suggested that employment in the formal sector is increasing in response to minimum wage growth. The model suggests that one mechanism for this shift could be local product demand. Under the big push story, firms internalize local demand, so that if nearby firms are paying higher wages, then local product markets expand which provides the impetus for industrial growth. This mechanism is testable. That is, if local demand is the sponsor for growth, we would expect local consumption to be increasing along with minimum wages, as well. Here, I use real consumption data from the IFLS. In particular, I test whether total real expenditures and expenditures on foods or non-
food products increases in response to minimum wages, where non-food products are defined as household durables and clothing, products that would be consumed even by relatively low income wage workers. As a placebo test, I also examine whether food production changes in response to minimum wages, as local food production is presumably not associated with demand for the product of local industries. All expenditure data are normalized relative to a national price index.

Table 4 performs this analysis. Here, and going forward, analysis is restricted to the robust DSD approach with the 25 mile bandwidth for space considerations.\footnote{Results at alternate bandwidths are available from the author. Most results maintain the same sign and significance pattern at all three bandwidths and all results presented here maintain the same sign pattern at all bandwidths and the same significance pattern at at least two of the three bandwidths.} One issue is that it’s not immediately obvious whether the inclusion of population-level demographic covariates is desirable in estimating the effects of minimum wages on product demand. On the one hand, the model emphasized the total size of local markets, and not the conditional correlations given demographic characteristics. On the other hand, incomplete sampling and characteristics of panel surveys means that the sample may evolve over time which may be related to consumption trends. To the extent that evolutions in consumption are due to the evolution of the panel sample, controlling for these changes would be desirable. I present both sets of estimates in Table 4. Row 1 reveals that local expenditures (and hence, local product markets) are growing in response to minimum wages. They’re growing substantially: these estimates indicate that the minimum wage is being passed on more or less fully to expenditures, so a 1% increase in the minimum wage is associated with 1% increased total expenditures. Row 2 presents some weaker evidence that food expenditures are growing. Without controlling for covariates, we do see a large increase in food expenditures, and this increase is attenuated somewhat but still marginally significant once we control for the population, age, gender distribution, and mean education levels in the district. Row 3 reveals that there is a large increase in non-food consumption which is robust to the inclusion of covariates. This speaks to the
plausibility of a big-push: Most industrializable, non-tradable industries produce non-food goods, and these estimates suggest that their market is indeed expanding. Finally, Row 4 presents a placebo test, examining food production. Food production may be reduced in response to minimum wages if workers leave agriculture, but it seems unlikely that food production would increase with industrialization and better formal sector labor market opportunities. Reassuringly, the data passes this test, and there is no statistical relationship between food production and minimum wages.

6.3 Industry Heterogeneity

The model suggested a number of dimensions by which different industries would respond differently to minimum wage law. The primary intuition of the big push model, that demand creates an externality which can increase formal sector employment which in turn crowds out the informal sector, is applicable only to industries which were both untradable (or at least with high costs of trade) and which have the potential for formalization. In contrast, industries which are tradable should experience zero or negative employment effects from minimum wages. For these firms, there is no externality produced by other workers increasing the market size, as their product does not specifically feed the local market. Employment effects may be small if fixed costs are substantial, but they should not be positive. The second condition, that the potential for formalization exists, was necessary for formal employment to "crowd out" informal employment. Otherwise, if the product is untradable, the externality of increased demand remains, suggesting an increase in labor demand. In industries with low potential for formalization, that growth will be consigned to the informal sector. These differences amount to strong empirical restrictions on the data. In principle, comparing across industries is similar to adding another level of differencing across industries within the data, except that the-

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9This trend in non-food consumption recalls the multiplier effects described in Rosenstein-Rodan (1943) and Murphy, Shleifer, and Vishny (1989).
ory provides stronger predictions on employment patterns than simply that one industry is a control for another. Here I consider three sectors in turn, for which the model has different empirical predictions: Manufacturing, Retail, and Services.

6.3.1 Manufacturing

Manufacturing is a sector which includes some tradable and untradable industries (or, at least, industries where the extent to which markets are local are highly variable). We also have more complete data about employment trends in manufacturing available in the SI. The model had few predictions for overall employment trends in an industry like manufacturing which is almost fully industrializable but very heterogeneous in tradability. Table 5 examines the effects on overall employment in manufacturing and on employment in registered firms and employment in unregistered firms, using the SI. Though not shown here, the SD estimator finds large differences in manufacturing employment. That is, districts with higher minimum wages than their neighbors tend to have much more manufacturing employment than their neighbors. However, these effects are not sufficiently robust to survive the DSD procedure, presented in row 1, suggesting that this difference is fixed over time. Overall, there is little evidence of on an overall effect of minimum wages on manufacturing in the SI. Similar results are presented in row 2, which examines the probability that there is zero manufacturing employment in a given district.

The SI however contains detail about industrial production and output which can help examine whether these fairly weak estimates are masking heterogeneous impacts. Theory would suggest that heterogeneity should be quite important in an industry like manufacturing: we should see positive employment effects of minimum wages on non-tradable formal firms, negative employment effects on non-tradable informal firms, and zero or negative employment effects on tradable firms. Of course, it is difficult to precisely capture the importance of a local market in a firm’s output, and as a census of large
manufacturing firms, the SI is bound to undersample informal firms. Nonetheless, we can develop proxies for each of these. Here, I suggest that firms who do not export any of their product are likely to be less tradable and that firms who export at least some of their production are more likely to be tradable. Moreover, the SI does include a number of firms (about 25%) who report having no legal status. While there is likely some misreporting here, we can at least suspect that firms that report no legal status are more likely to be informal, and that firms who report some legal status are more likely to be formal. Nearly all firms without legal status do not export, so by these definitions we would conclude that they are not tradable, echoing the modeling assumptions.

Rows 3 through 5 of Table 5 analyzes the effects of minimum wages on regulated firms which produce for domestic production, unregulated firms which produce for domestic production, and firms which export (who are nearly all regulated; very similar estimates are obtained excluding the few unregulated firms who export). The difference in spatial differences estimator reveals a clear pattern which lines up with theory: regulated, domestic firms see an increase in employment in response to the minimum wage; unregulated, domestic firms decrease employment; and there is no effect on exporting firms. The manufacturing census is revealing a clear pattern, which recalls the patterns in Tables 2 and 3: in response to the minimum wage, formal firms are advantaged relative to informal firms, but only in the case where their product is consumed locally.

Column 1 of Table 6 conducts a similar analysis using the IFLS data. Row 1 finds weak positive overall effects, and rows 2 and 4 demonstrate similar patterns to the SI, though they are less precisely estimated. Minimum wages are increasing the likelihood that individuals are manufacturing wage workers, which is significant. They are also less likely to be self employed in manufacturing, though point estimates are small in magnitude and not distinguishable from zero. Taken together these results are again supportive of an increase in the formal sector relative to the informal. Rows 3 and 5 examine wages and profits in manufacturing. Mean wages are increasing in response to the minimum wage,
as one might anticipate from an industry where much of wage work would be formal (though significance is only marginal). However, profits are not changing among the individuals who remain entrepreneurs in manufacturing.

### 6.3.2 Services

Services are an industry for which industrialization potential is minimal at best. Services are also fully untradable. The model suggests therefore that there should be an increase in labor demand in services. However, that increase in demand should be felt across the board, not just in the formal sector. That is, since services cannot industrialize, we shouldn’t observe the crowding out of informal employment that was demonstrated overall and in manufacturing. Column 2 of Table 6 examines sensitivities of several aspects of services to minimum wages using data from the IFLS. Row 1 reveals that employment in services expands when minimum wages grow, so that a 100% increase in minimum wages causes approximately 10% greater employment in services. Rows 2 and 4 reveal that this increase in employment is felt similarly in waged employment and self employment. In services, both of these are likely informal, though the IFLS does not allow us to determine carefully what fraction of wage workers and entrepreneurs are formal versus informal. However, Row 3 shows that wages aren’t increasing in services. This is consistent with the idea that services are fundamentally an informal industry, with little potential for industrialization and ultimate enforcement (and hence little wage gains in response to the law). Moreover, row 5 reveals that service profits are growing with the minimum wage. Along with the growth in service entrepreneurs, this is further evidence that the overall service sector is expanding as entrepreneurs earn greater income.

### 6.3.3 Retail

It is somewhat less clear what kinds of effects we should anticipate for retail. Clearly, retail is untradable, which means we have a clear prediction that the overall sector should
be growing. However, the extent to which it can be industrialized is much less clear and with that employment predictions are ambiguous. On the one hand, the retail sector is clearly different from some manufacturing industries, where one can easily envision the returns to establishing a large factory. On the other hand, there are clear gains to consolidation in the retail sector, as it is quite clear that the very micro retail establishments which are commonplace in developing countries are less profitable than the large, horizontally and vertically integrated retail sector which characterizes much retail in developed countries. Given this ambiguity, the results presents in Column 3 of Table 6 are quite striking. Increased minimum wages are causing retail employment to contract sharply overall in Indonesia. Rows 2 and 4 demonstrate that the contraction is largest for self employment, but still evident for wage work. Of course, the IFLS data can allow only imperfect separation of formal and informal wage work, and so while it’s fairly clear that informal retail is contracting (since the self employment should be virtually all informal), it is less clear what is happening to formal retail employment. At the same time, row 5 reveals that at the same time that entrepreneurship in retail is contracting, the profits of retail entrepreneurs are increasing	extsuperscript{10}. In other words, the retail sector is consolidating in response to minimum wages, which is consistent with the formalization results in the other sectors if there are available returns to scale in retail. Consistent with formalization within retail, row 3 documents a growth in wages per worker in response to minimum wages, as well.

7 Robustness

In this section, I evaluate whether the results presented above could be driven by four candidate confounding explanations. The first three could create results which resemble these without a large increase in employment due to the minimum wages: reverse causality...

\textsuperscript{10}In fact, the point estimate suggests that the total retail profits are increasing, as the average profits are increasing by more than employment is contracting.
tion, differential inflation, and migration. The last, monoponistic markets, is an alternate interpretation and motivation for the mechanism of the model.

### 7.1 Reverse Causation

The primary endogeneity concern in this study is that minimum wages were set according to secular trends in labor markets. Of course, the DSD estimation is robust to the possibility that more developed labor markets received higher minimum wages, but may be affected if areas are targeted for high minimum wage growth because there is knowledge that they are on a positive growth trend. The fact that a standard difference-in-difference estimator does not yield the same patterns is a first piece of evidence against this hypothesis as discussed above: if the tripartite councils are setting minimum wages according to labor market trends in their districts, the trends should be strongest on the raw difference-in-differences which identifies average trends across minimum wage regions rather than the local spatial identification since the councils should be concerned with average outcomes within their purview. However, it remains possible that border regions are somehow different from the interior of minimum wages, possibly due to some spurious correlation or some endogeneity in the chosen minimum wage boundaries, and therefore analysis which focuses on local comparisons at the border could be affected by reverse causation without a similar affect being yielded in the raw difference in differences. If this hypothesis were true, then we would expect to see that minimum wages are closely and differentially related to employment trends in the border regions. This suggests a simple specification:

\[ y_{it} = \alpha_i + \beta_1 (\text{minwage}_{it}) + \beta_2 (\text{minwage}_{it} \times \text{border}_i) + \gamma X_{it} + \delta_1 t + \delta_2 t \times \text{border}_i + \epsilon_{it} \]

(14)

where \( \text{border}_i \) is an indicator for being within 25 miles of a minimum wage group border. Given that there are strong time trends in real minimum wages, when we stratify on
a variable like border status it is important to stratify time trends as well, contained here in the $\delta_{1t}$ and $\delta_{2t}$ (the DSD procedure, of course, adds additional flexibility along these lines). If the preceding trends of formalization and growth are being driven by reverse causation, then we should expect that $\beta_2$ is different from zero and has the same sign as the DSD estimates above. Table 7 performs this analysis for the key results of tables 2 through 4: the effects of minimum wages on full time wage work, self employment, firm size, and expenditures. This specification tests whether trends in minimum wages are correlated with trends in labor market characteristics differentially in border regions.

Column (1) of table 7 reveals that there is no statistical difference in minimum wages based on either general trends in wage work or trends specific to the border region. As in table 3, Column 2 reveals that difference-in-difference estimates of the relationship between self employment and minimum wages are on average the opposite sign of the more robust DSD estimates. Once again, there is no differential relationship at the border. Columns 3-6 show no significance in $\beta_1$ or $\beta_2$ - in other words, minimum wages increases are not related to aggregate trends in firm size or expenditures, nor are they differentially related at the border. There is therefore no evidence to suggest that minimum wages are being set in response to increased formalization and even less evidence that they are being set differentially at borders in a way which could yield the estimates contained in this paper. Therefore, if reverse causation is important in the results presented in this paper, minimum wages are being targeted in a specific way which is tough to reconcile with sensible policy making: they must be being targeted at places which have advancing formalization trends relative to their neighbors, but not places with advancing overall trends. This both requires a great deal of sophistication on the part of policy makers, and a confusing objective function.
7.2 Prices

As discussed in section 3, one of the key motivations for the variation in minimum wages was to maintain a constant real wage. This suggests that a concern in analyzing these minimum wages is whether the estimation strategy does not identify systematic differences in prices. While it’s not immediately clear what relationship local inflation rates should have with employment growth in particular sectors, it is clear that if there were differential inflation associated with the minimum wage, this could lead to misestimates in the local product demand specification. Since this was the prime support of the model’s main mechanism, it is important to verify that the DSD technique is not picking up systematic differences in prices.

One limitation of the data is that we do not observe the full universe of prices and so cannot ask how prices of services or non-food items change. However, the IFLS does have data on food prices for a subset of districts, though these price data are quite variable. In the appendix table, I normalized price data into z-scores and test whether rice, beef, fish, sugar, salt, and cooking oil prices are associated with real minimum wages. None of these prices have a statistically significant association with the minimum wage, and in most DSD specifications point estimates are actually negative (though standard errors are sometimes large, which makes inference difficult). This is consistent with the retail sector becoming more efficient as discussed above.

7.3 Migration

A third concern is that migration could skew statistics. As a panel study, the IFLS will reflect trends in the population who were initially surveyed. If individuals who are well-suited for formal sector work respond to the presence of higher minimum wages by systematically migrating to high minimum-wage districts, that could skew employment results (though the presence of jobs for them is still an interesting phenomenon). The first
column of table 8 illustrates that individuals do indeed respond to changes in minimum wage rates by migrating. The dependent variable in this column is the fraction of individuals in that district-year who entered the panel in a different district. At least at the larger bandwidths, there is strong evidence that a larger fraction of surveyed individuals are migrants when wages are higher than in nearby districts. Immediately, this suggests the potential for bias in estimates due to migration. Fortunately, it is possible to perform a robustness analysis. In particular, rather than using a person’s present location and minimum wage in constructing employment statistics we can use the samples’ original districts. That is, we can ignore the information on their current location and test whether individuals experience changes in type of employment or overall expenditures based on the minimum wages in the district that they began the sample in, rather than the district that they currently are in. Column 2 repeats the migration analysis, but asks this time whether people are more likely to migrate from their home district based on the home district’s minimum wages. If minimum wages are producing economic opportunities as the analysis above has suggested, we would not expect an increase in out-migration due to higher minimum wages. Indeed, column 2 finds zero effects on outmigration as a result of minimum wage growth, and taken together, these two columns suggest that individuals are migrating to places with higher minimum wages.

The remaining columns of table 8 repeat the analysis of tables 2 and 3, using the origin-province minimum wages rather than those of the current province. As the reader can verify, no results change. That is, even though there is a systematic migration response to higher minimum wages, this response isn’t the variation that drove analysis in the previous section. The origin analysis shows that if an individual begins the sample in a district which will have higher minimum wage growth than a neighboring district, that individual can expect to be employed in a more formalized labor market over the coming 7 years than his neighbors.
7.4 Monopsonistic Markets

A somewhat more classical explanation of the trends presented in this paper could exist if Indonesian firms have monopsonistic power. In that case, formal sector monopsonistic firms would keep employment down in the absence of minimum wages with the goal of keeping wages down, as well. When minimum wages come in, this motivation for artificially low employment disappears. Firms expand, competing away some fraction of informal employment. With more workers employed at higher wages, local product demand increases. While it is unclear whether the degree of monopsony would line up with low levels of tradability and the inter-industry restrictions developed above, it is at least possible that these industrial characteristics are related to monopsonistic power. If so, a monopsonistic model would line up as well as the big push model with the trends presented thus far.

Identifying the potential of monopsony to create employment gains in response to minimum wages requires firm level data to understand how concentrated local labor markets are. Using the SI census, I can create a Herfindahl index of how concentrated employment in the manufacturing sector is in each district on average over the period 1990-2000.\textsuperscript{11} If monopsony is important, we should anticipate the identified trends to be strongest in the districts which have the highest Herfindahl index.

For a simple test, I split the sample into quartiles of Herfindahl index. The lowest quartile ranges from an Herfindahl level of 0.007 to 0.094, while the highest quartile ranges from 0.488 to 1. I then interact these two indicators with minimum wages, and

\textsuperscript{11}Specifically, I estimate Herfindahl indices in number of workers for each year (1990-2000) and then average them over the length of the panel in a given district. Unfortunately, monopsony indices cannot be constructed for other sectors due to an absence of firm-level data, so this exercise cannot rule out the presence of monopsony in other sectors contributing to the estimation. However, manufacturing seems likely to be a main suspect for monopsony due to the likelihood of larger scale than other sectors such as retail or services.
reestimate equation [13] using these interactions, that is

\[
y_{it} - \sum_{i' \in R(i)} \frac{y_{i't}}{n_{R(i)}} = \left( \alpha_i - \sum_{i' \in R(i)} \frac{\alpha_{i'}}{n_{R(i)}} \right) + \beta_1 \left( \text{minwage}_{it} - \sum_{i' \in R(i)} \frac{\text{minwage}_{i't}}{n_{R(i)}} \right) + \beta_k \sum_{k \in \{1,4\}} \left( H_{ik} \ast \text{minwage}_{it} - \sum_{i' \in R(i)} \frac{H_{ik} \ast \text{minwage}_{i't}}{n_{R(i)}} \right)
\]

\[
+ \gamma \left( X_{it} - \sum_{i' \in R(i)} \frac{X_{i't}}{n_{R(i)}} \right) + u_{it} - \sum_{i' \in R(i)} \frac{u_{i't}}{n_{R(i)}}
\]

where \( H_{ik} \) is an indicator for being in Herfindahl quartile \( k \) and \( u_{it} \) is a composite error term including both \( \nu_{it} \) and \( \varepsilon_{it} \) from equation [13]. As \( H_{ik} \) are collinear with \( \alpha_i \), any differences across space in the average density of monopsony is subsumed in the district fixed effects.

Table 9 reports the results of this estimation. Across the board, we find stable results consistent with the trends presented earlier - in the middle 2 quartiles, a doubling of the real minimum wage inspires a 13% growth in full time wage work, a 24% contraction of self employment, a 40% contraction of small firm employment and a 21% expansion of large firm employment, and about a doubling of total and non-food expenditures. If anything, effects are even stronger in the lowest quartile of the Herfindahl index, as the interaction term is always the same sign as the main effect. Moreover, for most dependent variables there is some evidence that these trends are weakest in the most concentrated markets. This suggests that, to the extent that employment growth as a response to minimum wages is heterogeneous across market structures, it is concentrated among the the least monopsonistic markets, which is consistent with a big push explanation but not a monopsony one.
8 Discussion and Conclusion

The early 1990s were a time of massive foreign investment and rapid economic growth for Indonesia. During that time period, the government quickly raised minimum wages. According to a big push model, if 1990s Indonesia was characterized by the potential for higher productivity industrial structures, and if the minimum wages were designed appropriately, then they could achieve a big push - a movement from a low wage, low consumption, informal labor market to a high wage, high consumption, formal labor market. This paper has presented evidence that such a shift did indeed take place. When minimum wages rose in one district relative to their neighbors, that district observed an increase in formal sector employment and a decrease in informal employment. It also observed an increase in local expenditures, which is consistent with the hypothesized mechanism of the big push: that local product demand increases labor demand. Moreover, this increase was only observed in local industries which can be industrialized and do supply local demand, supporting the model further. Tradable manufacturing firms saw no growth in employment, and untradable, but non-industrializable services saw an increase in informal employment.

These trends were identified through a difference in spatial differences analysis, an approach which weakens the identification assumptions of both difference-in-differences and spatial discontinuities. This weakening of assumptions was important: a raw difference-in-differences would have missed these connections. This suggests two things: first, in a context where labor law is varied, the underlying identification assumptions may be quite important. In this case, assuming average trends to be the same across districts with different minimum wage profiles would have underestimated employment effects. Second, it suggests that reverse causality is unlikely to be the explanation for these controversial findings. Minimum wage laws were established for entire regions. If governments were establishing these varied laws due to variation in the secular trends present in these regions, then they should have been responding to the average trend present in their region
rather than local trends at the borders. Since the raw difference-in-differences suggests that average trends in employment are not strongly related to the minimum wage profiles, the potential for reverse causality to be creating these positive estimates is minimized, and I confirm also that trends at the border are not differentially associated with minimum wage trends.

This big push discussion strongly recalls much older economic thought which has been widely discredited within the profession. Few economists today argue as 1920s and 1930s economists did, that increasing wages and local demand could be a motor for economic growth. One reason is the limited (and potentially negative) effect these policies had on depression-era America. There are of course many differences between 1990s Indonesia and 1930s America. One, as a less-developed country receiving substantial foreign investment, Indonesia may have had new access to potential, unadopted, and profitable technologies that simply needed a market. A second is that much of the 1990s were a time of growth in Indonesia, when sticky wages may have limited wage growth (the opposite of conditions in the depression). Finally, Harrison and Scorse (2010) show that anti-sweatshop activism also raised labor standards in foreign firms without an accompanying drop in employment. This indicates that wages may have indeed been below marginal products in the 1990s, reducing coordination and creating an opening for policy. Of course, the analysis employed in this paper cannot determine whether any of these conditions were important for these results. Further research, both empirical and theoretical is needed in considering the role of labor standards throughout the business cycle in modern less developed countries.

References


Table 1: Summary Statistics

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<td>0.111</td>
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</tr>
<tr>
<td>Log Real Minimum Wage (in 1000 Rs)</td>
<td>447</td>
<td>1.196</td>
<td>0.210</td>
<td>0.702</td>
<td>1.672</td>
</tr>
<tr>
<td>Mean Workers in Small Firm Occupations</td>
<td>447</td>
<td>0.388</td>
<td>0.183</td>
<td>0.065</td>
<td>0.926</td>
</tr>
<tr>
<td>Mean Workers in Large Firm Occupations</td>
<td>447</td>
<td>0.270</td>
<td>0.120</td>
<td>0.011</td>
<td>0.722</td>
</tr>
<tr>
<td>Mean Workers in Service</td>
<td>447</td>
<td>0.059</td>
<td>0.053</td>
<td>0.000</td>
<td>0.271</td>
</tr>
<tr>
<td>Mean Wage Workers in Service</td>
<td>447</td>
<td>0.026</td>
<td>0.029</td>
<td>0.000</td>
<td>0.137</td>
</tr>
<tr>
<td>Mean Self Employed in Service</td>
<td>447</td>
<td>0.022</td>
<td>0.025</td>
<td>0.000</td>
<td>0.123</td>
</tr>
<tr>
<td>Mean Workers in Retail</td>
<td>447</td>
<td>0.123</td>
<td>0.068</td>
<td>0.000</td>
<td>0.362</td>
</tr>
<tr>
<td>Mean Wage Workers in Retail</td>
<td>447</td>
<td>0.029</td>
<td>0.038</td>
<td>0.000</td>
<td>0.286</td>
</tr>
<tr>
<td>Mean Self Employed in Retail</td>
<td>447</td>
<td>0.146</td>
<td>0.089</td>
<td>0.000</td>
<td>0.565</td>
</tr>
<tr>
<td>Mean Workers in Manufacturing</td>
<td>447</td>
<td>0.076</td>
<td>0.059</td>
<td>0.000</td>
<td>0.333</td>
</tr>
<tr>
<td>Mean Wage Workers in Manufacturing</td>
<td>447</td>
<td>0.043</td>
<td>0.046</td>
<td>0.000</td>
<td>0.242</td>
</tr>
<tr>
<td>Mean Self Employed in Manufacturing</td>
<td>447</td>
<td>0.026</td>
<td>0.028</td>
<td>0.000</td>
<td>0.200</td>
</tr>
<tr>
<td>Log Sum Total Expenditures</td>
<td>447</td>
<td>4.786</td>
<td>0.821</td>
<td>2.872</td>
<td>7.250</td>
</tr>
<tr>
<td>Log Sum Food Purchases</td>
<td>447</td>
<td>4.073</td>
<td>0.748</td>
<td>2.495</td>
<td>6.296</td>
</tr>
<tr>
<td>Log Sum Non-Food Purchases</td>
<td>447</td>
<td>4.125</td>
<td>0.892</td>
<td>2.021</td>
<td>6.766</td>
</tr>
<tr>
<td>Log Sum Food Produced</td>
<td>447</td>
<td>2.106</td>
<td>0.923</td>
<td>0.000</td>
<td>4.584</td>
</tr>
<tr>
<td>SI</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Log Production Workers</td>
<td>2090</td>
<td>6.860</td>
<td>2.947</td>
<td>0.000</td>
<td>12.300</td>
</tr>
<tr>
<td>Indicator for zero workers</td>
<td>2090</td>
<td>0.089</td>
<td>0.285</td>
<td>0.000</td>
<td>1.000</td>
</tr>
<tr>
<td>Log Registered, Domestic Prod. Workers</td>
<td>2090</td>
<td>6.171</td>
<td>2.942</td>
<td>0.000</td>
<td>11.961</td>
</tr>
<tr>
<td>Log Unregistered, Domestic Prod. Workers</td>
<td>2090</td>
<td>2.609</td>
<td>2.976</td>
<td>0.000</td>
<td>9.523</td>
</tr>
<tr>
<td>Log Exporting Workers</td>
<td>2090</td>
<td>1.282</td>
<td>1.353</td>
<td>0.000</td>
<td>5.236</td>
</tr>
</tbody>
</table>
Table 2: Employment Effects of Minimum Wages

<table>
<thead>
<tr>
<th>Bandwidth</th>
<th>Full Time Wage Work</th>
<th>Part Time Wage Work</th>
<th>Self Employment</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
</tr>
<tr>
<td>15 miles</td>
<td>0.031</td>
<td>0.346*</td>
<td>0.050</td>
</tr>
<tr>
<td></td>
<td>(0.043)</td>
<td>(0.198)</td>
<td>(0.079)</td>
</tr>
<tr>
<td>25 miles</td>
<td>0.274***</td>
<td>0.127**</td>
<td>-0.043**</td>
</tr>
<tr>
<td></td>
<td>(0.090)</td>
<td>(0.052)</td>
<td>(0.009)</td>
</tr>
<tr>
<td>50 miles</td>
<td>0.222***</td>
<td>0.104**</td>
<td>-0.020</td>
</tr>
<tr>
<td></td>
<td>(0.078)</td>
<td>(0.048)</td>
<td>(0.019)</td>
</tr>
</tbody>
</table>

Notes
1. All estimates are coefficients on log real minimum wages where the dependent variable is the proportion of adults who work in the category given in the column heading.
2. DD estimates present a difference-in-differences; SD estimates present a spatial discontinuity, and DSD estimates present a Difference in Spatial Differences.
3. Standard Errors are clustered over space and within policy groups.
4. Spatial distances are presumed larger than bandwidths across different major islands.
5. Estimates control for time trends, urbanization, population surveyed, and mean gender, age, and education.
Table 3: Minimum Wage Effects on Small and Large Firm Occupations

<table>
<thead>
<tr>
<th>Bandwidth</th>
<th>Small Firm Occupations</th>
<th>Large Firm Occupations</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
</tr>
<tr>
<td>15 miles</td>
<td>-0.026</td>
<td>-0.251</td>
</tr>
<tr>
<td></td>
<td>(0.069)</td>
<td>(0.163)</td>
</tr>
<tr>
<td>25 miles</td>
<td>-0.182</td>
<td>-0.438***</td>
</tr>
<tr>
<td></td>
<td>(0.170)</td>
<td>(0.059)</td>
</tr>
<tr>
<td>50 miles</td>
<td>-0.168</td>
<td>-0.360***</td>
</tr>
<tr>
<td></td>
<td>(0.137)</td>
<td>(0.061)</td>
</tr>
</tbody>
</table>

Method | DD | SD | DSD | DD | SD | DSD |
N      | 447 | 447 | 447 | 447 | 447 | 447 |
Unique Districts | 151 | 151 | 151 | 151 | 151 | 151 |

Notes
1. All estimates are coefficients on log real minimum wages where the dependent variable is the proportion of adults who work in the category in the column heading
2. DD estimates present a difference-in-differences; SD estimates present a spatial discontinuity, and DSD estimates present a difference in spatial differences
3. Standard Errors are clustered over space and within policy groups
4. Spatial distances are presumed larger than bandwidths for across different major islands
5. Small Firm occupations have fewer than 10% of employees working in firms with at least 20 employees, large firm occupations have more than 10%
6. Estimates control for time trends, urbanization, population surveyed, and mean gender age, and education
## Table 4: Minimum Wages and Expenditures

<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th>(1)</th>
<th>(2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Log Total Expenditure</td>
<td>1.149***</td>
<td>0.798***</td>
</tr>
<tr>
<td></td>
<td>(0.306)</td>
<td>(0.273)</td>
</tr>
<tr>
<td>Log Food Purchases</td>
<td>0.993**</td>
<td>0.689*</td>
</tr>
<tr>
<td></td>
<td>(0.452)</td>
<td>(0.385)</td>
</tr>
<tr>
<td>Log Non-Food Purchases</td>
<td>1.279***</td>
<td>0.950***</td>
</tr>
<tr>
<td></td>
<td>(0.222)</td>
<td>(0.224)</td>
</tr>
<tr>
<td>Log Food Production</td>
<td>-0.544</td>
<td>-0.207</td>
</tr>
<tr>
<td></td>
<td>(2.371)</td>
<td>(2.392)</td>
</tr>
</tbody>
</table>

Covariates?  | No | Yes  
Method       | DSD | DSD  
N             | 447 | 447  
Unique Districts | 151 | 151  

### Notes
1. All estimates are coefficients on log real minimum wages.
2. All estimates are identified by Difference in Spatial Differences (DSD) with a 25 mile bandwidth.
3. Standard Errors are clustered over space and within policy groups.
4. Spatial distances are presumed larger than bandwidth across different major islands.
5. All specifications include time trends; covariates include mean gender, age, education, and the population surveyed.
Table 5: SI Manufacturing

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Dependent Variable</strong></td>
<td></td>
</tr>
<tr>
<td>Log Workers</td>
<td>-0.116</td>
</tr>
<tr>
<td></td>
<td>(0.388)</td>
</tr>
<tr>
<td>Zero Employment</td>
<td>-0.340</td>
</tr>
<tr>
<td></td>
<td>(0.204)</td>
</tr>
<tr>
<td>Registered Firms</td>
<td>4.325***</td>
</tr>
<tr>
<td></td>
<td>(1.885)</td>
</tr>
<tr>
<td>Domestic Firm Employment</td>
<td></td>
</tr>
<tr>
<td>Unregistered Firms</td>
<td>-2.099*</td>
</tr>
<tr>
<td></td>
<td>(1.130)</td>
</tr>
<tr>
<td>Exporting Firm Employment</td>
<td>0.487</td>
</tr>
<tr>
<td></td>
<td>(1.015)</td>
</tr>
<tr>
<td><strong>Method</strong></td>
<td>DSD</td>
</tr>
<tr>
<td><strong>N</strong></td>
<td>2090</td>
</tr>
<tr>
<td><strong>N Unique Districts</strong></td>
<td>209</td>
</tr>
<tr>
<td><strong>Data</strong></td>
<td>SI</td>
</tr>
</tbody>
</table>

**Notes**

1. All estimates are coefficients on log real minimum wages where the dependent variable is the proportion working in the given category.
2. All estimates are identified by Difference in Spatial Differences (DSD).
3. Standard Errors are clustered over space and within policy groups.
4. Spatial distances are presumed larger than bandwidths across different major islands.
5. Estimates control for population surveyed, time trends, and mean gender, age, and education.
### Table 6: Industry Heterogeneity

<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th>(1) Industry</th>
<th>Manufacturing</th>
<th>(2) Services</th>
<th>(3) Retail</th>
</tr>
</thead>
<tbody>
<tr>
<td>Employment</td>
<td>0.073*</td>
<td>0.096***</td>
<td>-0.299***</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.040)</td>
<td>(0.033)</td>
<td>(0.030)</td>
<td></td>
</tr>
<tr>
<td>Wage Employment</td>
<td>0.085**</td>
<td>0.0460***</td>
<td>-0.087*</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.034)</td>
<td>(0.018)</td>
<td>(0.048)</td>
<td></td>
</tr>
<tr>
<td>Wages</td>
<td>0.091*</td>
<td>-0.080</td>
<td>0.280***</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.050)</td>
<td>(0.096)</td>
<td>(0.059)</td>
<td></td>
</tr>
<tr>
<td>Self Employment</td>
<td>-0.011</td>
<td>0.053*</td>
<td>-0.266***</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.021)</td>
<td>(0.032)</td>
<td>(0.022)</td>
<td></td>
</tr>
<tr>
<td>Profits</td>
<td>-0.194</td>
<td>0.878***</td>
<td>0.325***</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.142)</td>
<td>(0.214)</td>
<td>(0.116)</td>
<td></td>
</tr>
</tbody>
</table>

| Method             | DSD          | DSD           | DSD         |
| N Employment Data  | 447          | 447           | 447         |
| N Wage Data        | 341          | 325           | 277         |
| N Profits Data     | 351          | 279           | 442         |

**Notes**

1. All estimates are coefficients on log real minimum wages where the dependent variable is either the proportion working in the given category or log mean payments.
2. All estimates are identified by Difference in Spatial Differences (DSD).
3. Standard Errors are clustered over space and within policy groups.
4. Spatial distances are presumed larger than bandwidths across different major islands.
5. Estimates control for population surveyed, time trends, and mean gender, age, and education.
6. Wage and Profits data exclude 0 observations.
### Table 7: Reverse Causation at the Border

<table>
<thead>
<tr>
<th></th>
<th>Full Time Wage Work</th>
<th>Self Employment</th>
<th>Small Firm Employment</th>
<th>Large Firm Employment</th>
<th>Log Food Purchases</th>
<th>Log Non-Food Purchases</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Log Minimum Wage</strong></td>
<td>0.016 (0.040)</td>
<td>0.103** (0.051)</td>
<td>0.004 (0.069)</td>
<td>0.116 (0.071)</td>
<td>0.150 (0.143)</td>
<td>0.007 (0.251)</td>
</tr>
<tr>
<td><strong>Border*Log</strong></td>
<td>-0.014 (0.060)</td>
<td>-0.081 (0.095)</td>
<td>-0.148 (0.167)</td>
<td>-0.039 (0.111)</td>
<td>0.035 (0.289)</td>
<td>0.333 (0.362)</td>
</tr>
<tr>
<td><strong>Minimum Wage</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Method**
- DD

**Observations**
- 447

**Unique Districts**
- 151

---

**Notes**
1. All estimates are identified by Difference in Differences (DD)
2. Standard Errors are clustered over space and within policy groups
3. Spatial distances are presumed larger than bandwidths across different major islands
4. Estimates control for population surveyed, time trends, and mean gender, age, and education
5. Time trends are allowed to be different in border and non-border regions.
<table>
<thead>
<tr>
<th>Bandwidth</th>
<th>Migrant Percentage</th>
<th>Full Time</th>
<th>Part Time</th>
<th>Self Employment</th>
<th>Small Firm Occupations</th>
<th>Large Firm Occupations</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>In (1)</td>
<td>Out (2)</td>
<td>Wage Work (3)</td>
<td>Wage Work (4)</td>
<td>Employment (5)</td>
<td>Occupations (6)</td>
</tr>
<tr>
<td>15 miles</td>
<td>0.109 (0.119)</td>
<td>-0.137 (0.155)</td>
<td>0.135** (0.068)</td>
<td>0.004 (0.050)</td>
<td>-0.254*** (0.083)</td>
<td>-0.512*** (0.045)</td>
</tr>
<tr>
<td>25 miles</td>
<td>0.479*** (0.076)</td>
<td>-0.080 (0.107)</td>
<td>0.114** (0.048)</td>
<td>0.039 (0.040)</td>
<td>-0.196*** (0.038)</td>
<td>-0.307*** (0.052)</td>
</tr>
<tr>
<td>50 miles</td>
<td>0.354*** (0.090)</td>
<td>0.031 (0.101)</td>
<td>0.102** (0.041)</td>
<td>-0.011 (0.028)</td>
<td>-0.082*** (0.029)</td>
<td>-0.277*** (0.038)</td>
</tr>
</tbody>
</table>

District Method
- Current: DSD
- Origin: DSD

Observations
- 447
- 449
- 449
- 449
- 449
- 449

Unique Districts
- 151
- 152
- 152
- 152
- 152
- 152

Notes
1. All estimates are coefficients on log real minimum wages where the dependent variable is indicated in the column heading.
2. All estimates are identified by Difference in Spatial Differences (DSD).
3. Standard Errors are clustered over space and within policy groups.
4. Spatial distances are presumed larger than bandwidths for provinces on different islands.
5. Estimates control for population surveyed, time trends, and mean gender, age, and education.
6. Current districts are means over the current residence, while origin districts are means over the origin district.
### Table 9: A Big Push or Monopsony?

<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th>Mean Full Time Wage Work (1)</th>
<th>Mean Self Employment Work (2)</th>
<th>Mean Small Firm Employment (3)</th>
<th>Mean Large Firm Employment (4)</th>
<th>Log Total Expenditures (5)</th>
<th>Log Non-Food Expenditures (6)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Log Real Minimum Wage</td>
<td>0.131**</td>
<td>-0.242***</td>
<td>-0.437***</td>
<td>0.206***</td>
<td>0.836**</td>
<td>0.978***</td>
</tr>
<tr>
<td></td>
<td>(0.063)</td>
<td>(0.082)</td>
<td>(0.091)</td>
<td>(0.064)</td>
<td>(0.347)</td>
<td>(0.292)</td>
</tr>
<tr>
<td>Least Monopsonistic Quartile*</td>
<td>0.067*</td>
<td>-0.047</td>
<td>-0.034</td>
<td>0.025</td>
<td>0.110</td>
<td>0.384</td>
</tr>
<tr>
<td>Log Real Minimum Wage</td>
<td>(0.036)</td>
<td>(0.041)</td>
<td>(0.069)</td>
<td>(0.040)</td>
<td>(0.272)</td>
<td>(0.254)</td>
</tr>
<tr>
<td>Most Monopsonistic Quartile*</td>
<td>-0.009</td>
<td>0.173*</td>
<td>0.008</td>
<td>0.066</td>
<td>-0.582</td>
<td>-0.440</td>
</tr>
<tr>
<td>Log Real Minimum Wage</td>
<td>(0.032)</td>
<td>(0.096)</td>
<td>(0.141)</td>
<td>(0.045)</td>
<td>(0.591)</td>
<td>(0.717)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Method</th>
<th>DSD</th>
<th>DSD</th>
<th>DSD</th>
<th>DSD</th>
<th>DSD</th>
<th>DSD</th>
</tr>
</thead>
<tbody>
<tr>
<td>Observations</td>
<td>411</td>
<td>411</td>
<td>411</td>
<td>411</td>
<td>411</td>
<td>411</td>
</tr>
<tr>
<td>Unique Districts</td>
<td>140</td>
<td>140</td>
<td>140</td>
<td>140</td>
<td>140</td>
<td>140</td>
</tr>
</tbody>
</table>

**Notes**

1. Least and Most monopsonistic quartiles are determined by calculating average herfindahl indices in the SI Manufacturing Censuses from 1990-2000
2. All estimates are identified by Difference in Spatial Differences (DSD) with a 25 mile bandwidth
3. Standard Errors are clustered over space and within policy groups
4. Spatial distances are presumed larger than bandwidths across different major islands
5. Specifications include mean gender, ag, education, and the population surveyed
6. A few observations are lost compared to previous tables as some districts are not present in the SI from 1990-2000
Figure 3: Wage Densities in Nearby Districts

(a) Normalized to Lower Minimum Wage

(b) Normalized to Higher Minimum Wage

Presents histograms of log wages where the sample includes all sub-districts within 25 miles of a minimum wage border. Log wages are normalized to either the higher or lower minimum wage.
Figure 4: Spatial Regression Discontinuity on Minimum Wages

(a) Full Time Wage Work

(b) Self Employment

Presents kernel-weighted local polynomial regressions of minimum wages on distance to the border of a given minimum wage regime. Positive distances indicate that the observation is on the side of the border with the higher minimum wage.
### Appendix Table: Food Prices

<table>
<thead>
<tr>
<th>Food Item</th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Rice</td>
<td>-0.569</td>
<td>-1.468</td>
<td>0.107</td>
</tr>
<tr>
<td></td>
<td>(1.567)</td>
<td>(1.172)</td>
<td>(0.223)</td>
</tr>
<tr>
<td>Beef</td>
<td>-1.239</td>
<td>-1.020</td>
<td>-1.219</td>
</tr>
<tr>
<td></td>
<td>(0.791)</td>
<td>(1.437)</td>
<td>(1.662)</td>
</tr>
<tr>
<td>Fish</td>
<td>0.972</td>
<td>-0.137</td>
<td>-0.868</td>
</tr>
<tr>
<td></td>
<td>(1.021)</td>
<td>(0.932)</td>
<td>(0.559)</td>
</tr>
<tr>
<td>Sugar</td>
<td>1.898</td>
<td>0.363</td>
<td>1.712</td>
</tr>
<tr>
<td></td>
<td>(1.317)</td>
<td>(0.869)</td>
<td>(3.098)</td>
</tr>
<tr>
<td>Salt</td>
<td>1.311</td>
<td>-0.650</td>
<td>-1.616</td>
</tr>
<tr>
<td></td>
<td>(2.161)</td>
<td>(0.626)</td>
<td>(2.564)</td>
</tr>
<tr>
<td>Oil</td>
<td>0.759</td>
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<td>(1.360)</td>
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**Notes**

1. All estimates are coefficients on log real minimum wages where the dependent variable is a z-score in price of a good
2. Standard Errors are clustered over space and within policy groups
3. Spatial distances are presumed larger than bandwidths across major islands
4. Specifications include rime trends and an indicator for urbanization