

Rural Migrants and Urban Informality: Evidence from Brazil

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Motivation

Urban population in developing countries grew by **12.5%** between 2015-2020; it is projected to grow by **64.7%** until 2050 ([UNCTAD, 2021](#)).

Rural-urban migration is a major contributor of urban growth ([Jedwab et al., 2017](#)), will increase with climate change ([Rigaud et al., 2018](#)).

Key question: Can developing country cities generate enough good jobs to accommodate this fast growing workforce?

Enough reasons for pessimism:

- Traditional "[Harris-Todaro-Fields](#)" view predicts that rural-urban migration leads to higher unemployment and informality in urban destinations.
- These predictions are supported by empirical evidence on the short-run effects of rural-urban migration (e.g. [Kleemans and Magruder, 2018](#)).
- Consistent with ample evidence on frictions facing firms and workers, low firm growth, high informality and unemployment.

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This paper

Question: What are the economic effects of rural-urban migration on local urban economies in Brazil?

- ① Shift-share IV design to identify the causal effects of immigration at destination in Brazil using decadal changes:
 - (i) Local labor markets: \uparrow formality, \downarrow informality and \downarrow wages (formal and informal); no effects on non-employment
 - (ii) Formal firms' dynamics: \uparrow formal firms, \uparrow jobs, \uparrow entry and \uparrow exit.
 - (iii) Results are due to the long time horizon: short-run specification gives the informality-increasing effects documented in the literature.
- ② Develop and estimate a model of firm dynamics and informality.
 - (i) Model replicates the IV results qualitatively and quantitatively
 - (ii) Transition dynamics with sluggish formal wage adjustment: \uparrow informality in the short run following a migration shock.
 - (iii) Informality serves as “stepping-stone” for firms, but reduces the overall dividends from immigration.

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Informality

Definitions:

- (i) **Extensive margin**: whether entrepreneurs register or not their *business*.
- (ii) **Intensive margin**: whether firms that are formally registered hire their workers with or without a formal contract.

$$\begin{aligned}\text{Informal firms} &= (1) \approx 70\% \text{ of firms in Brazil} \\ \text{Informal workers} &= (1) + (2) \quad 30\text{-}80\% \text{ of workers in LAC}\end{aligned}$$

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Contributions

Rural-urban migration and urban labor markets: theory (Fields, 1975; Harris and Todaro, 1970) and evidence (e.g. Corbi et al., 2021; Kleemans and Magruder, 2018).

- We show that rural-urban migration can lead to **lower informality**.

Labor market frictions in developing countries (e.g Abebe et al., 2021; Alfonsi et al., 2020; Carranza et al., 2022; Donovan et al., 2020; Franklin, 2018).

- We show that labor supply shocks can create formal jobs in equilibrium.

Population growth and firm dynamics in the US (Karahan et al., 2019; Pugsley and Sahin, 2019).

- Our findings are the mirror image + new model of formal and informal firm dynamics + first empirical evidence in a developing country context.

Immigration and firms: developed (e.g. Dustmann and Glitz, 2015; Kerr et al., 2015; Lewis, 2011; Peri, 2012) and developing countries (Albert et al., 2021; Imbert et al., 2022)

- We focus on aggregate effects, the role of informality and firm dynamics.

Outline

- 1 Empirical Analysis
- 2 Model
- 3 Model Estimation and Counterfactuals
- 4 Final remarks

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Data sources

- **Migration and Labor Market Outcomes:** Decennial Population Census, 1991-2010 [▶ Demo. Census](#) [▶ Desc. Stats](#)
 - Migrants = in their current location ≤ 10 years; we use the accumulated immigration rate 2000-2010.
 - Focus migration to urban areas (88% of all migration), and cross-state borders (40% of migration to urban areas). [▶ Map](#)
- **Firms:**
 - Matched employer-employee, admin data set from the Ministry of Labour → universe of formal firms and workers ([RAIS](#))
 - Matched employer-employee, survey data on small (up to 5 employees) formal and informal firms (ECINF)
- **Push Shocks:**
 - International Agricultural Commodity Price Shocks \times crop shares at the municipality level (in 1980). [▶ Shocks-Construction](#) [▶ Shocks-Maps](#)
 - Alternatively: drought shocks (SPEI drought index).

Empirical Design

$$\Delta y_d = \beta_0 + \beta_1 Mig_d + \beta_2' X_d + u_d$$

where

- $\Delta y_d = y_{d,2010} - y_{d,2000}$ (Dem. Census)
- $\Delta y_d = y_{d,2011-12} - y_{d,1999-00}$ (RAIS)
- $Mig_d = \sum_o \sum_{t=2001}^{2010} \frac{Mig_{o,d,t}}{Pop_{d,2000}}$;
- X_d : share of male, young and high skill (completed HS) in 2000.

Instrument: $Z_d = \sum_o \underbrace{\lambda_{o,d}}_{\text{mig.matrix 90's}} \times \underbrace{s_o}_{\text{price shifters}}$

► First Stage

Dynamic effects:

$$\Delta y_{dt} = y_{d,t} - y_{d,1999-00},$$

for $t = 1997-98$ and $t = 2011-12, \dots, 2017-18$.

Effects of Immigration on Workers

	Wage employment			Log monthly wage		
	Overall (1)	Formal (2)	Informal (3)	Overall (4)	Formal (5)	Informal (6)
Panel A: OLS						
Immigration	0.037 (0.019)	0.105 (0.023)	-0.068 (0.014)	0.062 (0.076)	0.031 (0.068)	0.034 (0.092)
Panel B: IV-Price						
Immigration	0.102 (0.101)	0.397 (0.147)	-0.294 (0.099)	-1.575 (0.568)	-2.149 (0.667)	-1.864 (0.726)
F Statistic (IV)	16.87	16.87	16.87	16.87	16.87	16.87
Baseline Mean	0.332	0.229	0.103	-	-	-
Observations	3545	3545	3545	3545	3545	3545

- 1p.p. $\uparrow\uparrow$ in $Mig_d(18.5\%SD) \rightarrow \uparrow\uparrow$ 0.4 p.p. in share of formal workers (avg. employment share of 23%) \approx 1.7% increase.

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- Formalization effect driven by workers moving from informal to formal jobs; if anything, a slight increase in total wage employment.

Effects on Firms

	# firms (1)	Entry (2)	Exit (3)	Nb jobs (4)	Firm wage (5)
Panel A: OLS					
Immigration	1.344 (0.109)	0.746 (0.263)	0.864 (0.434)	1.071 (0.269)	0.370 (0.101)
Panel B: IV - Price					
Immigration	2.395 (0.615)	7.205 (2.402)	6.563 (3.118)	2.178 (0.843)	-3.403 (1.147)
F Statistic (IV)	16.87	16.87	16.87	16.87	16.87
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- 1p.p. $\uparrow\uparrow$ in $Mig_d \rightarrow \uparrow\uparrow$ 2.4% in the number of firms, 2.2% in the number of formal jobs and $\downarrow\downarrow$ 3.4% in wages.

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- There is greater churn, effect on entry is slightly higher.

▶ Are migrants creating firms?

▶ Dynamic Effects

Further implications for local economies

- Firm composition shifts towards retail, services and construction, and away from manufacturing (similar effects on jobs). [▶ Composition-Industries](#)
 - New entrants are small in size: $\uparrow\uparrow$ share of firms ≤ 5 employees. [▶ Composition-Firm Size](#)
 - Expansion occurs in the middle of the productivity distribution, with a relative decline in the share of top-productivity firms. [▶ Composition-Firm Quality](#)
- ⇒ Similar to the "urbanization without industrialization" in [Gollin et al. \(2016\)](#).
- Effects driven by municipalities in the bottom and, to a lesser extent, middle terciles of the GDP per capita distribution.
 - Top tercile's avg. income is $2\times$ higher than middle's and $4.5\times$ larger than bottom tercile's.

Robustness

- **Pre-trends**: dynamic effects + Include lagged changes in outcomes as a control
- **Potential confounders**: Control for population, industry shares and log GDP per capita at baseline + driving distance to capital
- **Persistence of migration** (the shares): Control for lagged migration rates
[▶ Results](#)
- **Demand Channel**: Control for price shocks at destination and shocks to neighboring regions weighted by distance.
- **Capital Channel**: Control for exposure through bank network. [▶ Results](#)
- Estimate all results using Borusyak, Hull and Jaravel (2021)

Discussion: Short vs Long Run

For a subset of destination municipalities (705), we can construct a yearly panel using data from the National Household Survey (Corbi et al., 2023).

Short-run, y-o-y specification: $\Delta y_{dt} = \beta_0 + \beta_1 Mig_{dt} + \beta_2' X_d + \gamma_t + u_d$

We need a higher frequency shifter: **drought shocks using SPEI** (Albert et al., 2023; Corbi et al., 2023). [▶ Drought Shock](#) [▶ Shocks-Maps](#)

Results using drought shocks:

- **Benchmark** specification (decadal changes), full sample: **results unchanged** [▶ results](#)
- **Benchmark** specification, restricted sample: **results unchanged** [▶ results](#)
- **Short-run** specification (y-o-y): standard "Harris-Todaro-Fields" results
 ⇓⇓ wage employment and formality [▶ short run](#)

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Model: Overview

- Dynamic setting that differentiates firms' initial conditions and transitory shocks to productivity → heterogeneous growth profiles across firms
- Selection in/out of two margins of informality can occur both upon entry and over the life cycle of firms → informality can be a stepping-stone.
- All firms have the same technology, use labor as their only input, operate in the same industry, produce a homogeneous good and are price takers.
- **Frictions:** regulations that are imperfectly enforced, giving rise to informality
 - **Informal firms:** lower entry costs and no regulatory costs (e.g. taxes); **but** cost of operation is increasing in firm's size.
 - **Formal firms:** face all regulatory costs, but constant marginal costs; can evade labor regulations by hiring informal workers.
- No aggregate shocks, **homogeneous labor**, and labor supply is fixed.

Key feature: Firms' productivity process

- **Dynamics** are driven by the evolution of firms' **productivity**.
- Firms differ in terms of their **current productivity**, θ_{jt} , and their **long-run productivity** $\nu \sim H$, which is observed before entry occurs.
- The expected value of entry depends on ν : $E[V_s(\theta, w) | \nu]$, $s = i, f$.
- After entry in either sector occurs, the productivity process is given by:

$$\ln \theta_{j,1} = \ln \nu_j + \ln \epsilon_{j,1}$$

$$\ln \theta_{j,t} = \rho_s \ln \theta_{j,t-1} + (1 - \rho_s) \ln \nu_j + \ln \epsilon_{j,t}, \quad t \geq 2$$

where j indexes firms, $s = i, f$ denotes the sector, and $\ln \epsilon \sim \ln \mathcal{N}(0, \sigma_\epsilon^2)$.

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Estimation and counterfactuals

Estimation:

- We use a two-step Simulated Method of Moments (SMM) procedure.
 - 1 First step: System GMM and panel data for productivity process parameter + statutory values of taxes.
 - 2 Second step: 12 remaining parameter estimated using SMM

[▸ Estimates](#)[▸ Model Fit](#)

Counterfactuals:

- 1 **Immigration**: once and for all 10% labor supply shock (\approx 80th percentile immigration rate).
- 2 **Transition dynamics**: sluggish (downward) wage adjustment in the formal sector.
- 3 **Immigration + enforcement**: government intensifies enforcement.

Aggregate effects of a labor supply shock

Once-and-for-all increase in labor supply of 10% (\approx 80th percentile immigration rate).

	IV Estimation	Model
Share Informal Workers ($\Delta\%$)	-3.9	-4.1
Wages ($\Delta\%$)	-5.7	-3.4
Number Formal Firms ($\Delta\%$)	14.7	16.3
Newly created firms	–	9.9
Previously informal firms	–	6.4
Share Informal Firms ($\Delta\%$)	–	-5.3
Average Firm Productivity ($\Delta\%$)	–	-1.4
Output ($\Delta\%$)	–	7.1
Taxes ($\Delta\%$)	–	8.7

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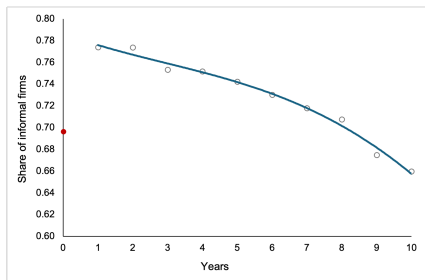
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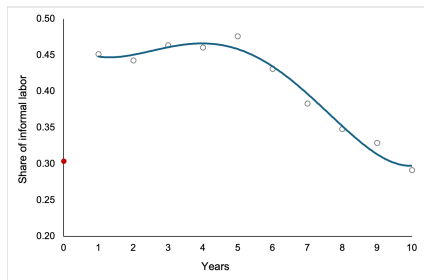
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Transition Dynamics

- Treat labor supply shock as a "MIT shock" that hits in $t = 1$. Focus on the equilibrium along a perfect-foresight path.
- Downward formal wage rigidity: $w_{f,t} = \gamma w_{t-1}$, $\gamma = 0.996$ following (Schmitt-Grohé and Uribe, 2016). Informal wages are fully adjustable.



(a) Firm informality



(b) Labor informality

What if we (nearly) shut down the informal sector?

	Baseline	LS Shock	LS Shock + Enforcement
Share Informal Labor	0.304	0.291	0.188
Share Informal Firms	0.696	0.660	0.221
Wages	1.000	0.966	0.979
# of Firms	1.000	1.038	0.889
# of Formal Firms	1.000	1.163	2.280
Avg. Firm Productivity	1.000	0.986	1.025
Output	1.000	1.071	1.083
Taxes	1.000	1.087	1.309

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Final remarks

- Immigration leads to a decrease in wages, and an increase in entry of formal firms, number of formal firms, jobs and formality share at destination.
- These contrast with the common narrative that rural-urban migration increases informality or under-employment in developing country cities.
 - Negative results are observed in the short- but not the the long-run.
- Our results thus suggest that developing countries might experience long-run demographic dividends, in particular from internal migration.
- However, empirical results and counterfactuals indicate that these gains do not accrue to the most productive firms → output per worker falls.
- Increasing enforcement could lead to higher dividends from immigration, but at the expense of a potentially large displacement of informal firms.
 - Broader lesson for the role of frictions?

SUPPORT SLIDES

Model: Set Up

- Continuum of firms indexed by their individual productivity, θ . Formal and informal firms have the same technology and use labor as their only input:

$$f(\ell) = \theta q(\ell), \quad q' > 0, q'' < 0$$

- Formal and informal firms operate in the same industry, produce an homogeneous good and are price takers.
- *Incumbents* pay a per-period fixed cost to operate, \bar{c}_s , $s = i, f$. *Entrants* pay a cost of entry into both sectors: $c_f^e > c_i^e$.
- In addition to endogenous exit, firms in both sectors face a death shock every period, denoted by δ_s .
- No industry-wide shocks + continuum of firms \rightarrow all aggregate vars. are deterministic.
- Labor supply is fixed.

Incumbents: Profit functions

- Informal firms:

$$\Pi_i(\theta, w) = \max_{\ell} \{\theta q(\ell) - \tau_i(\ell) w\}$$

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$$\Pi_i(\theta, w) = \max_{\ell} \{\theta q(\ell) - \tau_i(\ell) w\}$$

where $\tau_i', \tau_i'' > 0$ and $\tau_i(0) = 0$.

Incumbents: Profit functions

- **Formal firms:** $\Pi_f(\theta, w) = \max_{\ell} \{(1 - \tau_y) \theta q(\ell) - C(\ell)\}$

where

$$C(\ell) = \begin{cases} \tau_{fi}(\ell) w, & \ell \leq \tilde{\ell} \\ \tau_{fi}(\tilde{\ell}) w + (1 + \tau_w) w (\ell - \tilde{\ell}), & \ell > \tilde{\ell} \end{cases}$$

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and

$$\tau'_{fi}, \tau''_{fi} > 0 \text{ and } \tau_{fi}(0) = 0.$$

$$\tilde{\ell} \text{ is such that } \tau'_{fi}(\tilde{\ell}) = 1 + \tau_w.$$

Dynamics

- Dynamics are driven by the evolution of firms' idiosyncratic productivity, θ .
- Firms differ in terms of their current productivity, θ_{jt} , and their long-run productivity $\nu \sim H$, which is observed before entry occurs and drawn from:

$$H(\nu \geq x) = \begin{cases} \left(\frac{\nu_0}{x}\right)^\xi & \text{for } x \geq \nu_0 \\ 1 & \text{for } x < \nu_0 \end{cases}$$

- After entry in either sector occurs, the productivity process is given by:

$$\ln \theta_{j,1} = \ln \nu_j + \ln \epsilon_{j,1}$$

$$\ln \theta_{j,t} = \rho_s \ln \theta_{j,t-1} + (1 - \rho_s) \ln \nu_j + \ln \epsilon_{j,t}, \quad t \geq 2$$

where j indexes firms, $s = i, f$ denotes the sector, and $\ln \epsilon \sim \ln \mathcal{N}(0, \sigma_s^2)$.

- This structure implies that firms' first productivity draw – given by $\theta_1 = \nu \epsilon_1$ – has a Pareto-Lognormal distribution.

- Formal firms cannot become informal. Informal firms can pay the difference between formal and informal entry costs, $\tilde{c}^e = c_f^e - c_i^e$, and formalize.

- The value functions of formal and informal *incumbents*, respectively:

$$V_f(\theta, w) = \pi_f(\theta, w) + (1 - \delta_f) \beta \max \{0, E_\nu [V_f(\theta', w) | \theta]\}$$

$$V_i(\theta, w) = \pi_i(\theta, w) + \beta \max \{0, (1 - \delta_i) E_\nu [V_i(\theta', w) | \theta], (1 - \delta_f) E_\nu [V_f(\theta', w) | \theta] - \tilde{c}^e\}$$

where β is the discount factor, δ_s the exogenous exit.

- Exit decisions and informal-to-formal transitions follow cutoff rules:

$$E_\nu [V_s(\theta', w) | \underline{\theta}_s] = 0, \quad s = i, f$$

$$E_\nu [V_f(\theta', w) - V_i(\theta', w) | \bar{\theta}_i] = \tilde{c}^e$$

Entry

- Entrants in both sectors must pay a fixed cost of entry, denoted by c_s^e , $s = f, i$.
- These parameters will be estimated, but we expect that $c_f^e > c_i^e$.
- The expected value of entry for a firm with long-run productivity ν :
 $E [V_s (\theta, w) | \nu]$.
- Entry is characterized by the following threshold rule:

$$\begin{aligned} E [V_i (\theta, w) | \underline{\nu}_i] &= c_i^e \\ E [V_f (\theta, w) - V_i (\theta, w) | \underline{\nu}_f] &= c_f^e - c_i^e \end{aligned}$$

where $\underline{\nu}_s$ characterizes the last firm to enter sector $s = i, f$.

Migration and Labor Market Outcomes

- ◇ Unit of analysis: Brazilian municipalities
- ◇ Data source: [Decennial Population Census](#), 1991-2010
- ◇ Definitions:
 - We restrict the sample to 15-64 years old.
 - Migrants = those who came to their current location ≤ 10 years.
 - We compute the accumulated immigration rate between 2000 and 2010 and obtain a squared migration matrix between 3,658 municipalities.
 - Focus on flows to urban areas (88% of all migration), and across state borders (40% of migration to urban areas). [▶ Map](#)
 - We define formal workers as private sector employees with a formal contract, and informal ones are those without a formal contract. [▶ back](#)

Formal firms' outcomes

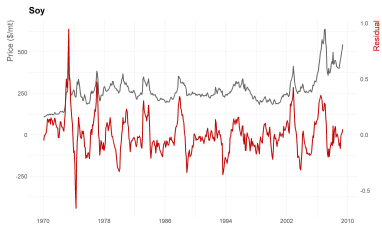
- ◇ Data source: *Relação Anual de Informações Sociais* (RAIS)
 - Matched employer-employee, admin data set from the Ministry of Labour in Brazil → universe of formal firms and workers.
 - Moments at the municipality level: (i) entry and exit; (ii) avg. firm size (as # employees); (iii) total number of establishments and formal workers; and (iv) the firm-level average wage.

Push Shocks: Crop Prices

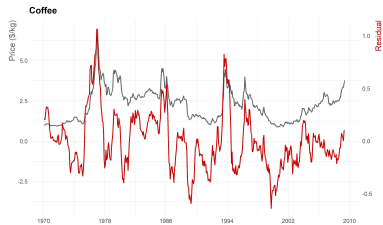
Price shocks:

- Source: World Bank Commodity Price Data (The Pink Sheet) 1972-2020.
- 12 Crops: bananas, cocoa, coffee, cotton, maize, orange, rice, soybeans, sugar, tobacco, wheat and wood.
- Crop×month-level price shock, ε_{cm} : residual from AR(1) process.
- Municipality×year level shock: sum of crop-level shocks weighted by the share of each crop in value of production in 1980 Agricultural Census, π_{oc} .

$$s_o^{prices} = \sum_m \sum_c (\pi_{oc} \times \varepsilon_{cm})$$



(c) Soy



(d) Coffee

[▶ Maps](#)[▶ Back](#)

Push Shocks: Droughts

- SPEI (Standardized Precipitation and Evapotranspiration Index), geo-localised measures of water balance linked to rainfall and temperature (Vicente-Serrano et al., 2010).
- Municipality×month-level shock, D_{om} : indicator for a drought if $\text{SPEI} < 0$.
- Municipality×year-level shock: sum of month-level shock weighted by the share of agricultural production, π_{oc} , that is in its growing season, g_{ocm} (1980 Agricultural Census).

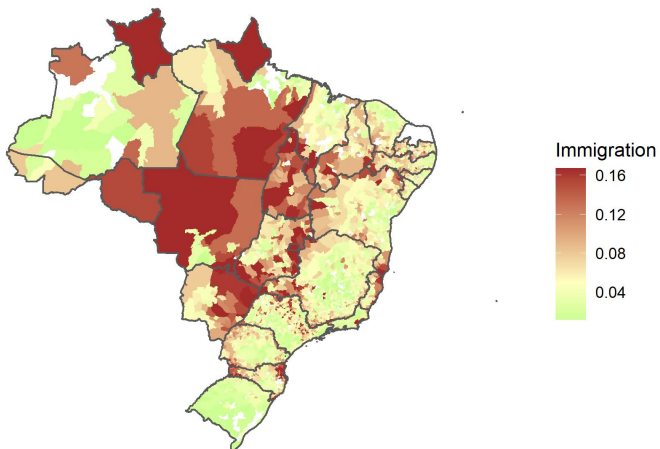
$$s_{ot}^{drought} = \sum_m \sum_c (\pi_{oc} \times g_{ocm} \times D_{om})$$

Table 1: Descriptive Stats - Census

	2010				2000			
	Mean	SD	Med.	N	Mean	SD	Med.	N
Population	24,380	141,257	4,890	3,548	18,064	167,356	3,743	3,453
% High Skill	0.253	0.079	0.241	3,548	0.172	0.062	0.171	3,453
Out-mig.	0.319	0.249	0.281	3,548	0.254	0.173	0.213	3,453
Out-mig. S-to-S	0.112	0.155	0.074	3,548	0.095	0.104	0.059	3,453
% Formal	0.194	0.120	0.165	3,548	0.138	0.118	0.106	3,453
% Informal	0.150	0.064	0.146	3,548	0.074	0.042	0.068	3,453
% non-employed	0.446	0.091	0.438	3,548	0.432	0.072	0.426	3,453
Formal wage	4.356	1.466	4.094	3,548	3.181	1.373	3.006	3,453
Informal wage	2.753	1.013	2.493	3,548	2.328	1.146	2.069	3,453

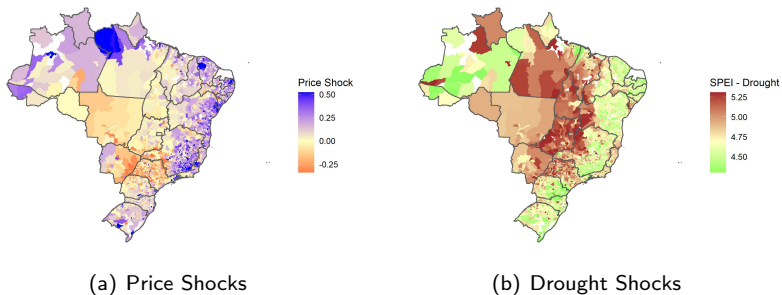
Notes: Weighted by the population at the destination municipality in the previous census; we compute the share of formal and informal as a proportion of total hours worked.

Immigration, 2000-2010



Notes: Computed using the Decennial Population Census. Darker areas denote higher immigration rates.

Figure 1: Migration Push Shocks

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First Stage

	Immigration	
	(1)	(2)
Price	-0.053 (0.006)	
Drought		0.077 (0.004)
Observations	3545	3545

[▶ Back](#)

Composition effects at destination

IV-Price			
Dep. var. (share of):	Female	Low Skill	Young
	(1)	(2)	(3)
Immigration	-0.083 (0.030)	0.029 (0.210)	0.257 (0.058)
Baseline Mean	0.483	0.720	0.128
Observations	3,548	3,548	3,548

Labor Market Effects by Skill

	Wage employment			Log monthly wage		
	Overall (1)	Formal (2)	Informal (3)	Overall (4)	Formal (5)	Informal (6)
Panel A: High-Skilled Workers						
Immigration	0.130 (0.125)	0.364 (0.172)	-0.235 (0.104)	-1.313 (0.455)	-1.610 (0.524)	-1.076 (0.670)
Observations	3,545	3,545	3,545	3,545	3,524	3,511
Panel B: Low-Skilled Workers						
Immigration	0.039 (0.107)	0.325 (0.109)	-0.286 (0.096)	-1.876 (0.746)	-2.268 (0.791)	-2.462 (0.994)
Observations	3,545	3,545	3,545	3,545	3,543	3,545

Wage Effects by Skill

[▶ Back](#)

Occupational composition

	Formal (1)	Informal (2)	Non-emp (3)	Self-emp (4)
Immigration	0.394 (0.149)	-0.294 (0.100)	0.092 (0.129)	-0.041 (0.047)
Baseline Mean	0.229	0.103	0.435	0.119
Observations	3545	3545	3545	3545
	Employer (5)	Domestic (6)	Public (7)	Non-remun (8)
Immigration	-0.040 (0.020)	0.020 (0.031)	-0.102 (0.061)	-0.029 (0.022)
Baseline Mean	0.019	0.046	0.038	0.012
Observations	3,545	3,545	3,545	3,545

Share of migrants by firm ownership

Share of Migrants:	Self-emp. (1)	Firm Owner (2)	Small Firm Owner (3)	Big Firm Owner (4)
Immigration	-0.257 (0.195)	0.173 (0.288)	0.160 (0.330)	-0.062 (0.459)
Observations	3,547	3,076	2,969	2,061

Note: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

▶ Back

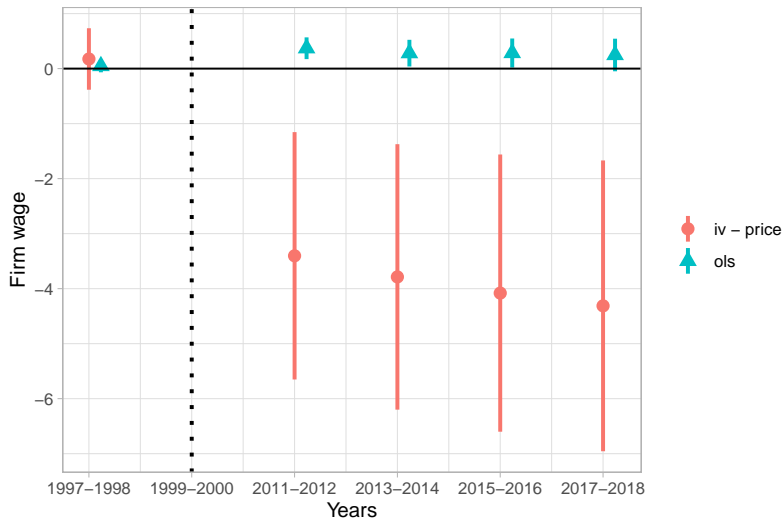
Sectoral composition

Industries:	Retail and Services (1)	Construction (2)	Manufacturing (3)	Other Sectors (4)
<i>Panel A: Shares of Firms</i>				
Immigration	0.226 (0.150)	0.176 (0.053)	-0.303 (0.136)	-0.099 (0.136)
Baseline Mean	0.738	0.033	0.111	0.118
<i>Panel B: Shares of Jobs</i>				
Immigration	0.373 (0.402)	-0.114 (0.109)	-0.341 (0.257)	0.082 (0.408)
Baseline Mean	0.465	0.041	0.185	0.309
Observations	3,545	3,545	3,545	3,545

Sizes of firms:	≤ 5 (1)	6 to 10 (2)	11 to 20 (3)	21 to 50 (4)	> 50 (5)
<i>Panel A: Shares of Firms</i>					
Immigration	0.367 (0.138)	-0.161 (0.068)	-0.122 (0.048)	-0.079 (0.041)	-0.005 (0.033)
Baseline Mean	0.706	0.131	0.078	0.048	0.036
<i>Panel B: Shares of Jobs</i>					
Immigration	0.050 (0.095)	-0.067 (0.062)	-0.136 (0.072)	-0.087 (0.103)	0.240 (0.258)
Baseline share	0.129	0.079	0.086	0.112	0.594
Observations	3,545	3,545	3,545	3,545	3,545

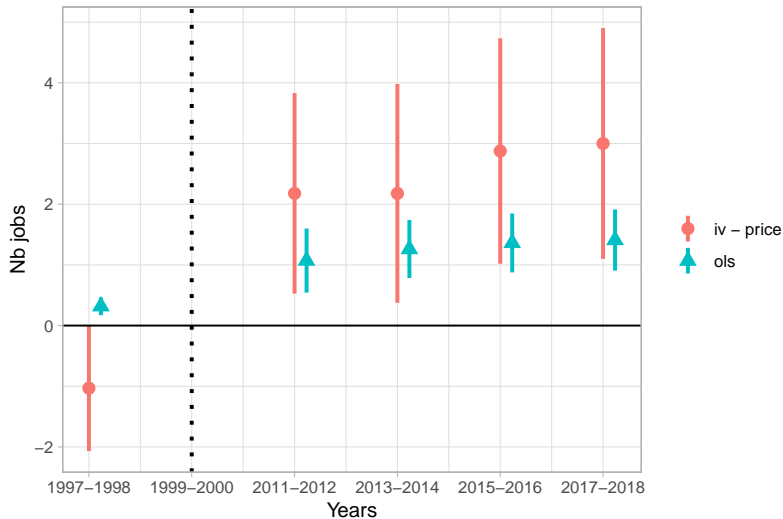
Dynamic effects

Wages



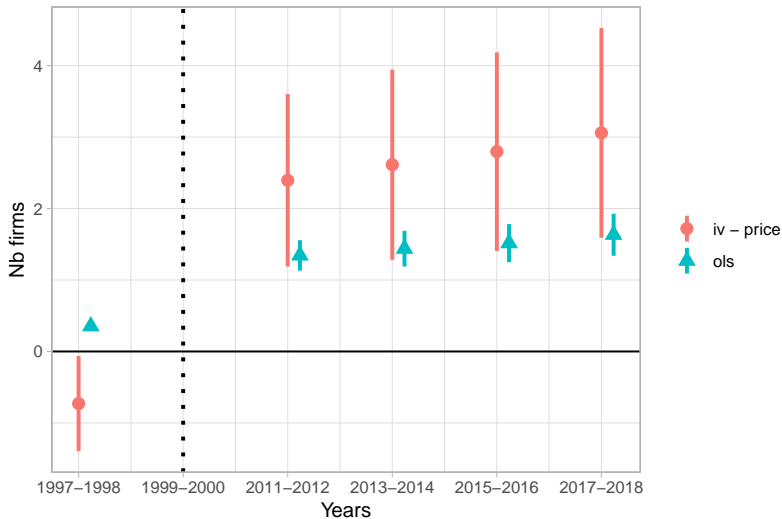
Dynamic effects

Number of Formal Jobs



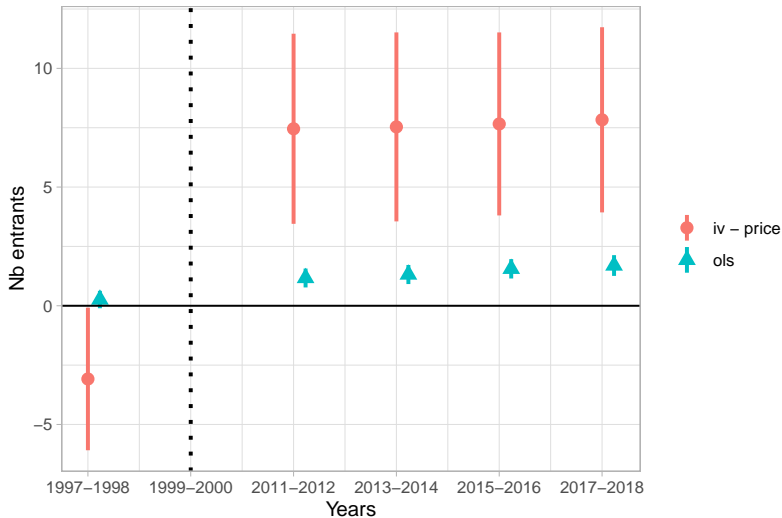
Dynamic effects

Number of Formal Firms

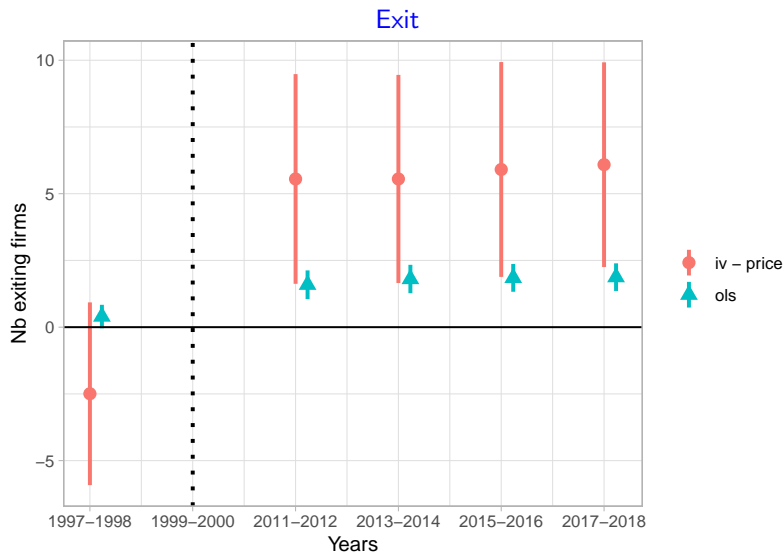


Dynamic effects

Entry



Dynamic effects



Effects on the composition of firms

We compute firm-level average wages and remove year and region fixed effects.

We compute the quartiles of firms' residual average wage distribution at baseline (1996-1999).

Panel A: Shares of Firms				
	Q1	Q2	Q3	Q4
Immigration	-0.407 (0.617)	2.664 (0.958)	1.153 (1.223)	-3.410 (1.439)
Panel B: Shares of Jobs				
	Q1	Q2	Q3	Q4
Immigration	-0.159 (0.304)	0.922 (0.427)	1.052 (0.721)	-1.814 (1.046)
Observations	3545	3545	3545	3545

Effects by firm quartile

Quartiles:	Bottom (1)	Mid-bottom (2)	Mid-top (3)	Top (4)
<i>Panel A: Shares of Firms</i>				
Immigration	-0.393 (0.626)	2.686 (0.976)	1.250 (1.239)	-3.543 (1.487)
<i>Panel B: Shares of Jobs</i>				
Immigration	-0.165 (0.310)	0.951 (0.440)	1.083 (0.733)	-1.869 (1.072)
Observations	3,545	3,545	3,545	3,545

Results with drought shock: benchmark specification

	Wage employment			Log monthly wage		
	Overall (1)	Formal (2)	Informal (3)	Overall (4)	Formal (5)	Informal (6)
IV-Drought Immigration	-0.014 (0.060)	0.271 (0.089)	-0.284 (0.072)	-0.126 (0.284)	-0.671 (0.336)	-0.200 (0.352)
F Statistic (IV)	18.11	18.11	18.11	18.11	18.11	18.11
Observations	3,545	3,545	3,545	3,545	3,545	3,545

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Results with drought shock: benchmark specification

	Nb firms (1)	Entry (2)	Exit (3)	Nb jobs (4)	Firm wage (5)
IV - Drought Immigration	1.625 (0.306)	2.555 (0.877)	2.758 (1.123)	2.031 (0.624)	-0.747 (0.554)
F Statistic (IV)	18.11	18.11	18.11	18.11	18.11
Observations	3,545	3,545	3,545	3,545	3,545

▶ Back

Drought shock: benchmark specification w/restricted sample

	Wage employment			Log monthly wage		
	Overall	Formal	Informal	Overall	Formal	Informal
	(1)	(2)	(3)	(4)	(5)	(6)
Panel A: OLS						
Immigration	0.004 (0.022)	0.067*** (0.023)	-0.062*** (0.018)	0.059 (0.099)	0.024 (0.086)	0.044 (0.124)
Panel B: IV-Price						
Immigration	-0.091 (0.103)	0.148 (0.119)	-0.239** (0.101)	-1.663** (0.682)	-2.102*** (0.792)	-1.906** (0.831)
F Statistic (IV)	11.16	11.16	11.16	11.16	11.16	11.16
Baseline Mean	0.338	0.243	0.095	-	-	-
Observations	700	700	700	700	700	700

Note:

*p<0.1; **p<0.05; ***p<0.01

► Back

Drought shock: benchmark specification w/restricted sample

	Nb firms	Entry	Exit	Nb jobs	Firm wage
	(1)	(2)	(3)	(4)	(5)
Panel A: OLS					
Immigration	1.215*** (0.119)	1.296*** (0.283)	1.870*** (0.365)	0.922*** (0.319)	0.498*** (0.134)
Panel B: IV-Price					
Immigration	2.310*** (0.644)	6.632*** (2.234)	5.126** (2.223)	2.089** (0.910)	-2.609** (1.203)
F Statistic (IV)	11.16	11.16	11.16	11.16	11.16
Observations	700	700	700	700	700

Note:

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Drought shock: short run specification – workers

	Wage employment			Log monthly wage		
	Overall	Formal	Informal	Overall	Formal	Informal
	(1)	(2)	(3)	(4)	(5)	(6)
Panel A: OLS						
Immigration	0.099** (0.040)	0.087** (0.040)	0.011 (0.022)	0.162 (0.175)	0.285 (0.188)	-0.176 (0.343)
Panel B: IV-Drought						
Immigration	-1.233** (0.624)	-1.199** (0.576)	-0.034 (0.324)	0.512 (2.080)	1.226 (2.507)	-1.890 (3.256)
F Statistic (IV)	21.53	21.53	21.53	21.53	21.53	21.61
Baseline Mean	0.335	0.242	0.092	-	-	-
Observations	6,407	6,407	6,407	6,407	6,381	6,377

Note:

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Drought shock: short run specification – firms

	Nb firms (1)	Entry (2)	Exit (3)	Nb jobs (4)	Firm wage (5)
Panel A: OLS					
Immigration	0.106 (0.024)	-0.081 (0.105)	-0.246 (0.063)	-0.011 (0.076)	0.077 (0.042)
Panel B: IV-Drought					
Immigration	0.704 (0.332)	2.921 (4.115)	3.079 (0.996)	-14.425 (5.211)	-0.609 (0.528)
F Statistic (IV)	21.52	21.52	21.52	21.52	21.52
Observations	6,382	6,382	6,382	6,382	6,382

Robustness: Control for Omitted Variables

	Nb firms (1)	Entry (2)	Exit (3)	Nb jobs (4)	Firm wage (5)
Panel A: Controlling for Outcome Lag					
Immigration	2.439 (0.593)	7.412 (1.991)	3.838 (0.933)	2.045 (0.792)	-3.410 (1.153)
Panel B: Controlling for Population Lag					
Immigration	2.435 (0.646)	7.345 (2.086)	5.916 (2.065)	2.089 (0.882)	-2.984 (1.020)
Panel C: Controlling for Immigration Lag					
Immigration	5.303 (3.212)	29.224 (13.644)	20.135 (10.284)	6.163 (4.112)	-15.124 (8.432)
Panel D: Controlling for log(GDP) Lag					
Immigration	2.486 (0.683)	8.472 (2.443)	6.908 (2.476)	2.263 (0.907)	-2.887 (0.971)
Panel E: IV-Price - Controlling for Industries Lag					
Immigration	1.945 (0.531)	6.595 (1.851)	5.167 (1.860)	2.155 (0.811)	-2.549 (0.932)
Observations	3545	3545	3545	3545	3545

Robustness: Control for Alternative Channels

	Nb firms (1)	Entry (2)	Exit (3)	Nb jobs (4)	Firm wage (5)
<i>Panel A: Controlling for local and neighboring municipalities' price shocks</i>					
Immigration	2.158 (0.530)	4.697 (1.370)	2.667 (1.411)	2.219 (0.739)	-3.920 (1.181)
Observations	3,545	3,545	3,545	3,545	3,545
<i>Panel B: Controlling for capital reallocation channel</i>					
Immigration	2.415 (0.692)	7.391 (2.266)	4.979 (2.223)	2.527 (0.941)	-3.494 (1.336)
Observations	2,627	2,627	2,627	2,627	2,627
<i>Panel C: Excluding firms that produce agricultural goods</i>					
Immigration	2.785 (0.648)	7.513 (2.054)	5.542 (2.039)	2.395 (0.885)	-3.823 (1.247)
Observations	3,545	3,545	3,545	3,545	3,545

Model's parameters

Parameter	Description	Source	Value	SE
<i>First Step</i>				
τ_w	Payroll Tax	Statutory values	0.375	–
τ_y	Revenue Tax	Statutory values	0.293	–
ρ	Productivity Process: Persistence Parameter	GMM Estimation	0.92	–
ν_0	Pareto's Location Parameter	Calibrated	7.3	–
γ_f	Per-period fixed cost of operation (Formal)	Calibrated	0.7	–
<i>Second Step</i>				
φ_f	Intensive margin: $\tau_f = \left(1 + \frac{\ell}{\varphi_f}\right)\ell$	SMM Estimation	6.450	0.228
φ_i	Extensive margin: $\tau_i = \left(1 + \frac{\ell}{\varphi_i}\right)\ell$	SMM Estimation	5.427	0.303
δ_i	Informal death shock	SMM Estimation	0.148	0.015
δ_f	Formal death shock	SMM Estimation	0.066	0.011
γ_i	Informal, per-period fixed cost of operation	SMM Estimation	0.350	0.161
ξ	Pareto shape parameter	SMM Estimation	3.801	0.092
c_f^e [†]	Formal sector's entry cost	SMM Estimation	7,400	3,383
c_i^e [†]	Informal sector's entry cost	SMM Estimation	2,800	598
α	Span-of-control	SMM Estimation	0.643	0.218
σ_i	Informal productivity process: SD	SMM Estimation	0.144	0.053
σ_f	Formal productivity process: SD	SMM Estimation	0.148	0.032
ρ_i	Informal productivity process: persistence	SMM Estimation	0.935	0.091

[†] Estimates and SD expressed in R\$ of 2003.

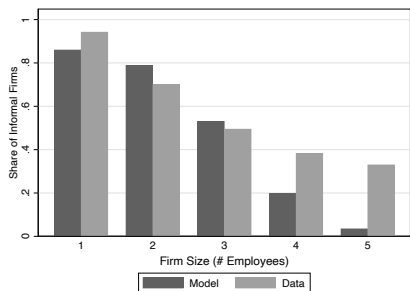
Model Fit (1/4): Targeted moments

Table 1: Model Fit – Targeted moments

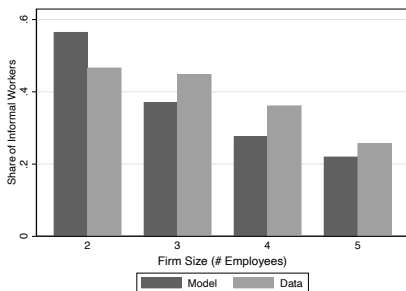
	Model	Data
Share Informal workers	0.304	0.298
Share Informal Firms	0.696	0.696
Informal Firms Size Distribution		
≤ 2 employees	0.933	0.957
≤ 5 employees	0.999	0.998
Formal Firms Size Distribution		
≤ 5 employees	0.658	0.697
6 to 10	0.136	0.144
11 to 20	0.092	0.083
21 to 50	0.053	0.048
> 50	0.023	0.028

Notes: Data moments computed using the RAIS, ECINF and PNAD data sets.

Model Fit (2/4): Extensive and intensive margins of informality

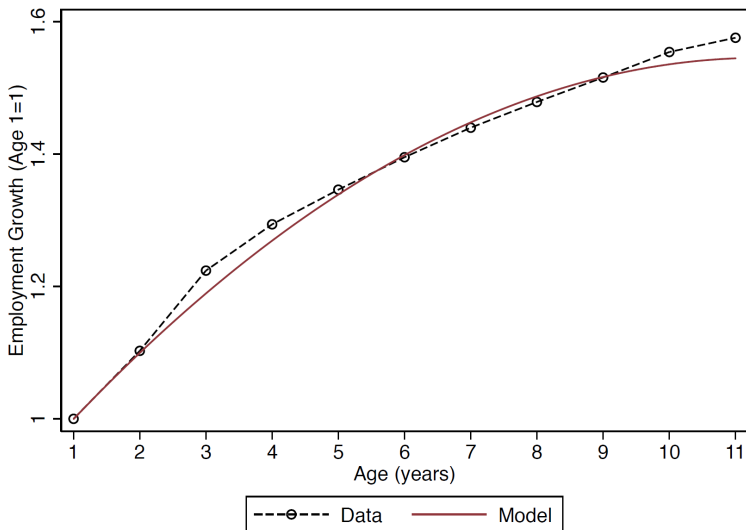


(c) Extensive Mg.

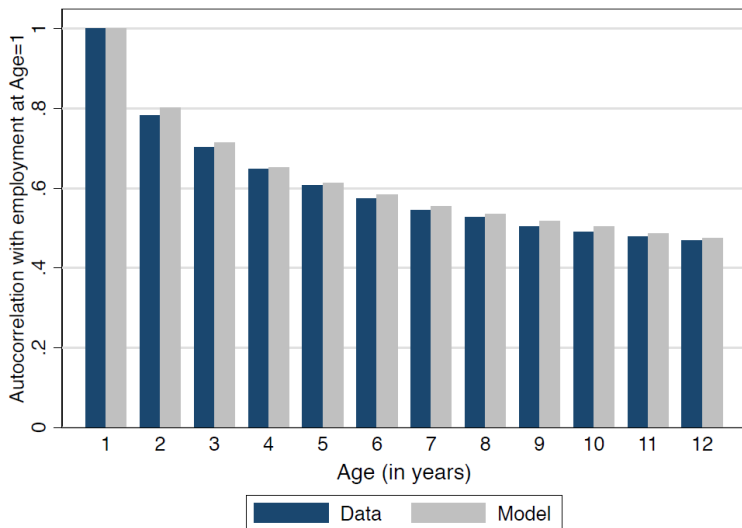


(d) Intensive Mg.

Model Fit (3/4): Firm Growth – Formal Sector



Model Fit (4/4): Autocorrelations – Formal Sector



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