The Network Origins of Firm Dynamics

¹Sciences Po ²Princeton ³Manchester

Contracting Frictions and Dynamism with Long-Term Relationships

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Weak Contract Enforcement and Long Term Relationships

Systematic differences in firm dynamics across countries (Hsieh-Klenow 2014)

Long term relationships can substitute for formal contract enforcement

- \cdot static benefit: helps incentives o lower transaction costs
- potential cost: less likely to switch to better supplier

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Johnson, McMillan, Woodruff (JLEO 2002):

- · Survey of firms in Eastern Europe
- Belief in quality of courts varies across countries
- "If another firm you have never purchased from offered to supply this input for a price 10% lower than this supplier, would you purchase from the new firm instead of this supplier?"
 - \cdot Custom inputs: less confidence in courts \implies more likely to reject new offer
 - · Standard inputs: little difference

Monarch (2020): US imports from China

Firms in more contract intensive industries stay with suppliers for longer

What is the role of relationships in firm dynamics and growth?

This paper

Evidence on firm dynamics facts, relationships, and their relationship to contracting frictions from India and Pakistan

Quantitative model with firm dynamics built on firm-to-firm trade and relational contracts

- · Arrival of potential suppliers (for buyers), potential buyers (for suppliers)
- Assess calibration of model to Pakistani data in its ability to explain firm dynamics facts

Application to understand quantitative role of contracting frictions in reducing dynamism

- \cdot Weak enforcement: lower arrival of new draws \implies less switching
 - ⇒ Lower aggregate productivity
 - ⇒ Slower firm dynamics
- Calibrated to Indian/Pakistani context, suggests substantial losses from misallocation of demand (~ 3x static losses)

Literature

Firm Dynamics:

- Customer Capital: Luttmer (2011), Gourio Rudanko (2014), Afrouzi Drenik Kim, Argente Fitzgerald
 Moreira Priolo, Einav Klenow Levin Murciano-Goroff, Foster Haltiwanger Syverson (2016)
- Input-Switching: Gopinath Neiman (2014), Lu Mariscal Mejia (2024), Damijan Konings Polanec (2014), Monarch (2022) Baqaee Burstein Duprez Farhi (2023)
- · Kortum-Klette: Lentz Mortenson (2008), Akcigit Kerr (2018), Garcia-Macia Hseih Klenow (2019)

· Firm-to-firm trade

- Firm heterogeneity, static: Oberfield (2018), Bernard Moxnes Ultveit-Moe (2018), Eaton Kortum Kramarz (2024), Bernard Dhyne Magerman Manova Moxnes (2022)
- Deterministic Life Cycle: Chaney (2014) and Aekka Khanna
- Dynamics with Frictions: Huneeus, Miyauchi, Martin Mejean Parenti (2023) and Fontaine Martin Mejean (2023)
- Frictions and Dynamism: Hopenhayn, Rogerson (1993), Hsieh, Klenow (2014), Akcigit Alp Peters (2021)
- Contracting frictions: Boehm (2022), Amirapu (2021), Boehm Oberfield (2020)
- · Relational contracts: Kranton (1996), Hemous, Olsen (2018), Macchiavello Morjaria (2015,2021)

Outline

- · Some motivational empirical results from India & Pakistan
- Single Industry Model
- Quantitative model with multiple Industries
 - Assess firm dynamics
- · Introduce weak enforcement, relational contracts
 - · Predictions using numerical Simulation
 - · Assess predictions with microdata on Indian & Pakistani data
- Implications for Aggregate Productivity

Data

- Indian Annual Survey of Industries, 1989/90-2014/15 (with gaps)
 - Plant-level panel survey of formal manufacturing plants
 - · All plants that have 100+ employees
 - \cdot 1/5 of all plants between 20 (10 if using power) and 100 employees
 - $\cdot \sim$ 25,000 plants per year
 - Great info on 5-digit outputs and inputs
 - $\boldsymbol{\cdot}$ Do not observe plants every year, do not see firm-to-firm transactions
- · Supplement with Pakistan Value Added Tax data 2011-2019
 - Monthly Firm-to-Firm sales transactions
 - $\cdot\,\sim$ 150k firms, 21m monthly transactions, aggregated to 5.4m annual transactions
 - Firms register if annual sales above threshold (\$2-15k)
 - · Only have 2-digit industry of firm, do not see products traded

Setting

- · Court Quality: Average age of pending cases Correlation with GDP/capita
 - India: High courts. Range of avg age: 1-4.5 years
 - · Pakistan: District courts. Range: 1-3 years
- Standardized vs. Relationship-specific (Rauch)
 - Standardized pprox sold on an organized exchange, ref. price in trade pub.
 - · Relationship-specific pprox everything else
 - \cdot India: \sim 70% of sales are rel-spec
 - Pakistan: \sim 60% of sales is rel-spec
- · Approach
 - Differential effect of court quality on firms that produce relationship-specific vs. standardized
 - Always include district and industry fixed effects
 - IV for court quality in India (Boehm & Oberfield, 2020) Oberfield, 2020)

Duration of Relationships (Pakistan)

	Dependent variable: Age of Relationship (in Years)					
	(1)	(2)	(3)	(4)	(5)	
Age of pending cases (S) \times RelSpec $_{ m S}$	0.225** (0.045)					
Age of pending cases (B) \times RelSpec $_S$	0.0638 (0.045)					
Age of pending cases (Min(B,S)) \times RelSpec $_S$		0.281** (0.032)	0.264** (0.041)			
Age of pending cases (Min(B,S)) \times EnforcementIntensity _{b,s}				0.0228* (0.011)	0.0258 (0.013	
B × S Industry FE	Yes	Yes	Yes	Yes	Yes	
B District FE	Yes	Yes		Yes		
S District FE	Yes	Yes		Yes		
S District × S Industry FE			Yes		Yes	
B District \times B Industry FE			Yes		Yes	
R^2	0.0630	0.0636	0.0929	0.0625	0.0922	
Observations	2140189	2142616	2141943	2142616	214194	

 $^{^{+}}$ p < 0.10. * p < 0.05. ** p < 0.01

Contracting frictions in output markets ⇒ lower variance of plant growth

	Dependent variable: $\sigma(\Delta \log {\sf Sales})_{d\omega}$						
	(1)	(2)	(3)	(4)			
Avg age of civil cases \times Rel. spec.	-0.0177* (0.0089)	-0.0187* (0.0088)	-0.0401* (0.016)	-0.0385* (0.016)			
$(\Delta \log Sales)_{d\omega}$		-0.273** (0.024)		-0.273** (0.024)			
State FE 5-digit Industry FE	Yes Yes	Yes Yes	Yes Yes	Yes Yes			
Estimator	OLS	OLS	IV	IV			
R ² Observations	0.287 7574	0.302 7574	-0.000369 7574	0.0207 7574			

Regression at the state \times industry level. Only state-industry cells with more than 5 observations used.

Dependent variable: standard deviation of residualized* annualized sales growth in each state-industry cell

^{*}residualized by age, year, state and industry

Exit Rates by Size (Pakistan)

	Dependent variable: P(exit)					
	(1)	(2)	(3)	(4)		
Q1 Dummy	0.0738*** (0.0023)	0.0717*** (0.0057)				
Q2 Dummy	0.0255*** (0.0018)	0.0208*** (0.0033)	-0.0460*** (0.0013)	-0.0469*** (0.0042)		
Q3 Dummy	0.0131*** (0.00099)	0.00979*** (0.0016)	-0.0576*** (0.0016)	-0.0567*** (0.0043)		
Q4 Dummy	0.00800*** (0.00071)	0.00677*** (0.0011)	-0.0611*** (0.0018)	-0.0586*** (0.0044)		
$Q1 \times Relspec \times AvgAgeCourts$		0.00129 (0.0026)		-0.00539* (0.0025)		
Q2 × Relspec × AvgAgeCourts		0.00299* (0.0014)		-0.00501** (0.0019)		
Q3 × Relspec × AvgAgeCourts		0.00221* (0.00099)		-0.00627*** (0.0016)		
Q4 × Relspec × AvgAgeCourts		0.000871 (0.00087)		-0.00755*** (0.0016)		
Industry × Year FE			Yes	Yes		
R ² Observations	0.0525 417711	0.0526 411541	0.0460 417698	0.0462 411528		

Standard errors in parentheses, clustered at the industry-region level. $^+$ p < 0.10. * p < 0.05. * p < 0.01

Model: Single Industry

- · Growing industry with many firms.
- · Each firm produces using labor and one input

$$y_b = A(z_{bs}x_s)^{\alpha}l^{1-\alpha}, \qquad A \equiv \alpha^{-\alpha}(1-\alpha)^{-(1-\alpha)}$$

- New potential suppliers, potential customers arrive via Poisson process
 - Each new potential match: random supplier s, random match-specific productivity z_{bs}
 - Key decision: switch or not
- Large number of retailers
 - Same production function as firms
 - · Same arrival rate of new techniques as firms
 - · Sell output to household (but not to other firms or retailers)
 - · Non-retailer firms sell to other firms and to retailers, but not to household

Static Equilibrium

- · Representative Household
 - \cdot Dixit-stiglitz preferences across varieties sold by retailers (elast. arepsilon)
 - · Households inelastically supplies a growing quantity of labor L (growth rate γ)
 - Labor used for production or to create new firms
- · Market structure
 - Monopolistic Competition across retailers
 - · Bilateral contracts in firm-to-firm trade (quantity, transfer)
 - · Countably stable: no countable coalition wants to alter/drop contracts
 - ⇒ Efficient production within supply chains (quantities)

$$c_b = \left(\frac{c_s}{z_{bs}}\right)^{\alpha} w^{1-\alpha}$$

- Many ways to split surplus
 - $\boldsymbol{\cdot}$ Focus on equilibrium in which surplus split according to cost shares

Keeping the model tractable

- · State variable for a firm is, in principle, very large
- · We focus on one economic decision:
 - · New supplier comes along: switch or not
 - Easy if each supplier's cost follows random walk with the same distribution of increments: lower cost now \implies better distribution of future cost (FOSD)
- · No mean reversion in cost
 - · Productivity of new potential match inspired by current supply chain
 - · no option to go back to old supplier
 - · No supplier death

Productivity of new potential match inspired by current supply chain

· Productivity delivered by current chain is

$$q \equiv z_0 z_1^{\alpha} z_2^{\alpha^2} \dots$$

where $z_0, z_1, z_2, ...$ are firm's own, its supplier's, its supplier's supplier's...

· match-specific prod. with new potential supplier:

$$z = \underbrace{b}_{\begin{subarray}{c} \begin{subarray}{c} \begin{subarra$$

 \cdot The arrival rate of new suppliers with original component larger than b is

$$\kappa b^{-\beta}$$

 \implies Arrival rate of supplier that delivers cost reduction larger than x is

$$\phi X^{-\beta}$$
, $\phi \equiv \kappa \int (c_s/w)^{-\beta} dF(c_s)$

Entry and Exit

- \cdot Potential problem: Random walk for cost \implies no stationary distribution
 - · Usual: Reflecting barrier (Gabaix) or endog exit (Hopenhayn/Luttmer) \implies mean reversion
 - · Solution: Mass of entrants grows over time
- Population grows at rate γ , $L_t = L_0 e^{\gamma t}$
- Entry
 - Free entry: unit of labor \implies flow χ of firms and χ_R of retailers
 - $\cdot \implies$ Along BGP, flow of entrants grows at population growth rate, γ
 - Each entrant draws potential suppliers: The number of draws of techniques with match-specific component larger than z is Poisson with mean $\kappa_0 z^{-\beta}$
- Exit
 - · Firms never die. But if no customers, output is zero
 - · A firm "exits" when it loses its last customer
 - · May gain customers later, still draws new suppliers, etc

Changes in Cost

- 'Get a better supplier' or 'supplier gets a better supplier', or 'supplier's supplier gets...
 - · Jump process with infinite activity
 - · Along any interval with finite length, infinite number of jumps
- · MGF of change in $\log \frac{w}{cost}$ over interval with length au

$$\mathbb{E}\left[\left(\frac{cost_{j,t}}{cost_{j,t+\tau}}\right)^{s}\right] = e^{-\tau\phi\sum_{k=1}^{\infty}\frac{s}{\beta\alpha^{-k}+s}}$$

· Along BGP, distribution of cost has a power law left tail

$$\lim_{c \to 0} \frac{\log \text{ Fraction with cost } \le c}{\log c} = \nu$$

where ν is unique solution to $\gamma = \phi \sum_{k=1}^{\infty} \frac{\nu}{\beta \alpha^{-k} - \nu}$

Aggregate Output along BGP

Aggregate output is

$$Y_t = \left(|R_t| \int_0^\infty c^{1-\varepsilon} dF(c)\right)^{\frac{1}{\varepsilon-1}} (1-\eta) L_t$$

In special case where $\beta=\varepsilon-1$, output per capita is

$$\frac{Y_t}{L_t} = (1 - \eta) \left(\frac{\eta \chi_R}{\gamma} L_0\right)^{\frac{1}{\beta}} \left[\frac{\kappa_0^{\alpha} \Gamma \left(1 - \alpha\right)}{1 + \frac{\phi}{\gamma} \sum_{k=1}^{\infty} \frac{1}{1 - \alpha^{-k}}}\right]^{\frac{1}{1 - \alpha} \frac{1}{\beta}} e^{\frac{\gamma}{\beta} t}$$

Semi-endogenous growth

- · Distribution of cost in cross section is constant over time
- \cdot Growth from gains from variety
- · Firm-level dynamics matter for level of output along BGP

Quantitative Model with Multiple Industries

Multiple Industries

• Firm b in industry ω

$$y_b = A_\omega l^{\alpha_{\omega l}} \prod_{\omega'} (z_{bs'} x_{s'})^{\alpha_{\omega \omega'}}$$
 with
$$A_\omega \equiv \alpha_{\omega l}^{-\alpha_{\omega l}} \prod_{\omega'} \alpha_{\omega \omega'}^{-\alpha_{\omega \omega'}}$$

- For each input, match-specific productivity of new potential suppliers inspired by current supply chain for that input
- · Some industries produce relationship specific goods
- · Cobb-Douglas keeps it tractable:
 - \cdot log cost is weighted sum of random walks
 - \cdot Cobb Douglas \implies weights are fixed within industry

Weak Enforcement and Relational Contracts

- Less efficient courts \implies switch suppliers of relationship-specific goods less frequently
 - $\kappa \downarrow$ uniformly for relationship-specific inputs
- · For today: Behavioral assumption
- Potential microfoundation: relational contracting as substitute for courts more



- · Repeated game, many equilibria
- · We can show equilibrium for some special cases of model
- · Working on proof for full model

Numerical Simulation

Parameter	Value	Target	Target value	Data source
Population growth (γ)	0.04	Employment share by age		Hsieh & Klenow (2014)
New technique shape (β)	3.52	Δ cost from new suppliers	-0.284	Baqaee et al. (2023)
New supplier arrival rate (ϕ)	0.58	Mean relationship length	1.72 years	Pakistan data
Observation threshold	varies	Median sales above threshold Threshold	6.36	Pakistan data
Number of retailer firms ratio	60	Annual exit probability	0.05	
Household EoS $(arepsilon)$	4.52	$\beta + 1$		

Table 1: Parameterization

- Firms per industry, Industry cost shares from Indian ASI data (coming later)
- Add positive drift to cost to center distribution of $\Delta \log$ cost at zero

1 additional year of average age of pending cases \Rightarrow relationships with rel.spec. inputs last \sim 0.25 year longer

 \Rightarrow calibrate κ for products with frictions to match that (in the worst congested state)

Size and Cost

- Firm size depends on how many customers, how large those customers are, which depends on how many customers they have
- Prospects for growth depend on cost: Lower cost \implies attract customers more quickly
 - Potential customers more likely to switch
- · Changes in size:
 - · Downstream: Gain or lose customers, customers grow or shrink...
 - · Upstream: Get better supplier: household buys more from downstream retailers
- · One type of shock (arrival of potential relationship), but many subtle patterns

Predictions

When enforcement is worse:

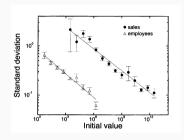
- \cdot Lower variance of firm growth \rightarrow see earlier results
- · Less mean reversion
- Less skewed size distribution
- Lower exit rate \rightarrow see earlier results

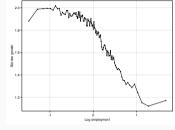
Predictions

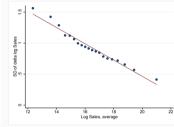
When enforcement is worse:

- Lower variance of firm growth \rightarrow see earlier results
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- Less skewed size distribution
- Lower exit rate → see earlier results

Standard Deviation of Growth Rates by Size

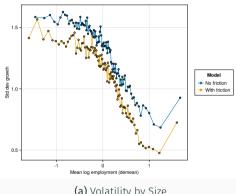




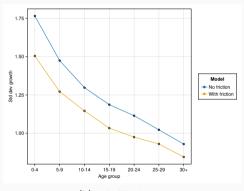


- · Larger ⇒ lower standard deviation of growth rates (Hymer and Pashigian, 1962)
 - \cdot Usual mechanism: Large firms composed of more subunits \implies diversification
 - $\boldsymbol{\cdot}$ Here: Large firm tends to have more customers
- · Declines more slowly than $\sqrt{\text{size}}$
 - · Usual mechanism: correlation across subunits, granular subunits
 - Here: granular customers (also some correlation from cost changes) Comparison

Standard Deviation of Growth Rates: Frictions vs No Frictions (Model)

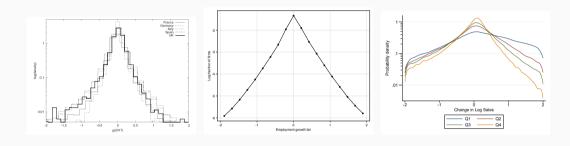


(a) Volatility by Size



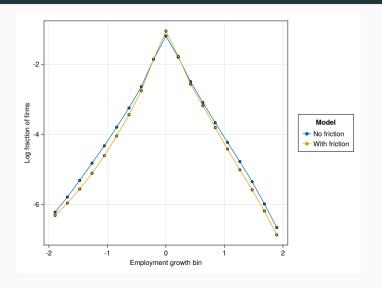
(b) Volatility by Age

Growth Rates

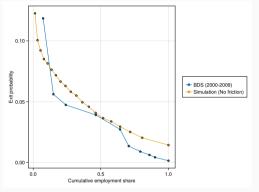


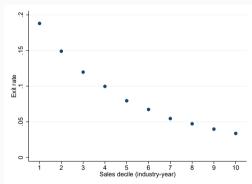
- Fat tails: Ashton, 1926, Laplace dist: Stanley, et al. (1996)
- Here: Mixture of getting one large customer, many small customers

Growth Rates: Frictions vs No Frictions (Model)



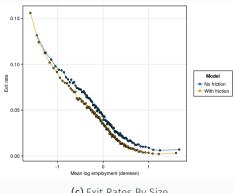
Exit Rates by Size



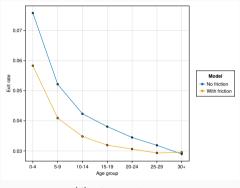


- · Firms exit when they lose last customer
- · Large firms can have one large customer
- · Number of buyers is a good predictor of exit

Exit Rates: Frictions vs No Frictions (Model)

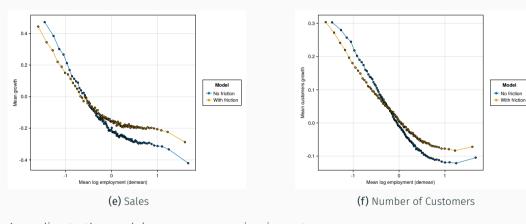


(c) Exit Rates By Size



(d) Exit Rates by Age

Mean Reversion: Frictions vs No Frictions (Model)



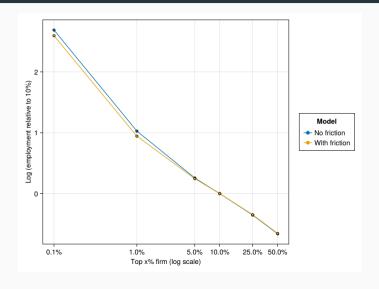
According to the model, no mean reversion in cost But: mean reversion in sales towards a long-run level commensurate with costs With fictions (\rightarrow less turnover) slower mean-reversion in sales

Mean reversion in firm size (slower with frictions)

	Dependent variable: Change in log Sales					
	(1)	(2)	(3)	(4)	(5)	(6)
$log Sales_{t-1}$	-0.403** (0.011)	-0.427** (0.025)	-0.555** (0.037)	-0.403** (0.012)	-0.436** (0.028)	-0.583** (0.038)
$\log \text{Sales}_{t-1} \times \text{Age civ. cases} \times \text{relspec}$	0.00709 ⁺ (0.0037)	0.0206* (0.0096)	0.0249 ⁺ (0.015)	0.00687 (0.0044)	0.0256* (0.012)	0.0405* (0.019)
Plant × 5-digit Industry FE	Yes	Yes	Yes	Yes	Yes	Yes
State FE	Yes			Yes		
Year × Previous Year FE	Yes			Yes		
Age FE		Yes	Yes		Yes	Yes
Industry × District × Year FE		Yes			Yes	
Industry \times District \times ($t, t - 1$) FE			Yes			Yes
Method	OLS	OLS	OLS	IV	IV	IV
R^2	0.457	0.636	0.671	0.256	0.250	0.278
Observations	204518	78053	51401	204518	78053	51401

Standard errors in parentheses, clustered at the state \times industry level.

Size Distribution



Contracting frictions in output markets ⇒ lower skewness in size distribution

	Dependent variable: Skewness of log Sales						
	(1)	(2)	(3)	(4)	(5)	(6)	
Relspec x Court Congestion	-0.360* (0.168)	-0.671* (0.287)	-0.799** (0.294)	-0.624 ⁺ (0.349)	-1.312* (0.598)	-0.905 (0.578)	
R^2	0.540	0.435	0.554	0.001	0.000	0.007	
State FE	Yes	Yes	Yes	Yes	Yes	Yes	
5-digit Industry FE	Yes	Yes	Yes	Yes	Yes	Yes	
Estimator	OLS	OLS	OLS	IV	IV	IV	
Statistic	25-75	50-75	50-90	25-75	50-75	50-90	
Observations	3008	3008	1448	3008	3008	1448	

$$Skewness_{s\omega} = \frac{log \, (Share \, of \, plants \, above \, S_1) - log \, (Share \, of \, plants \, above \, S_0)}{log \, S_1 - log \, S_0}$$

 S_0 and S_1 are different quantiles of overall plant size distribution (25th, 50th, and 75th, 90th)

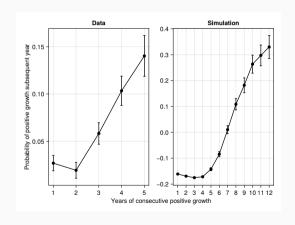
Similar with Pakistan data Pakistan

"Gazelles" / "rockets" / type dependence / ex ante heterogeneity

- Luttmer (2011): Need "rockets" that eventually slow to explain why largest firms are not so old
- Sedlecek, Sterk, Pugsley (2021): Hidden "ex ante heterogeneity" explains most of size dispersion at young ages, almost half of size dispersion at twenty
- Coad, Daunfeldt, Halvarsson (2018): autocorrelation of growth rates is positive for young firms and negative for older firms

Here: cost is hidden type

- · Cost determines inflow of customers
- Low cost at birth ⇒ persistent growth until inflows equals outflows
- Cost evolves over time

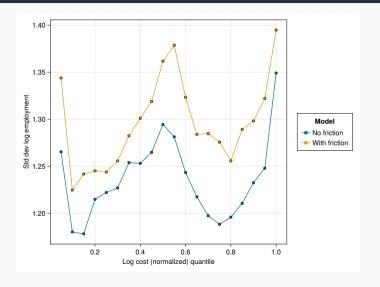


Implications for Aggregate Productivity

Productivity

- · Productivity growth is $\frac{\gamma}{\varepsilon-1}$
 - · Gains from variety/Population growth
- · Weak enforcement affects level of productivity
- Misallocation: Firms use worse suppliers than they would with better enforcement

Misallocation: Dispersion in Size



Misallocation: Correlation of Log Cost and Log Employment

Model	Correlation (demeaned)	Correlation (normalized)
No friction	-0.281	-0.370
With friction	-0.260	-0.340

Aggregate Productivity

	No friction	With friction
Mean income growth	0.015	0.015
Log real income difference	0.000	-0.162

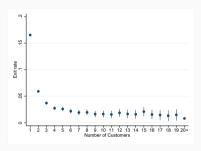
- · Note: In counterfactuals, entry rate held fixed
- \cdot More severe contracting frictions \implies lower entry (impact on welfare not obvious)

Conclusion

- · One response to weak contract enforcement is to use relational contracts
- · Static benefits, but less switching
 - ⇒ Slower firm dynamics
 - ⇒ Cost penalty builds up over time
 - \cdot Not switching in past \implies large impact on current aggregate productivity
- \cdot Dynamic costs of bad enforcement are \sim 3 times the size of static costs

Appendix

Number of Buyers is Good Predictor of Exit 🖼



	Dependent variable: P(exit)					
	(1)	(2)	(3)	(4)		
Constant	0.0878** (0.00039)	0.0879** (0.00038)	0.0878** (0.00038)	0.0879** (0.00038)		
Fixed Effects	Year	Year, #Buyers	Year, Sales vingtiles	Year, #Buyers, Sales vingtiles		
R ² Observations	0.0293 501828	0.0889 501431	0.0976 501828	0.112 501431		

Standard errors in parentheses, clustered at the industry-region level. + n < 0.10 * n < 0.05 * n < 0.01

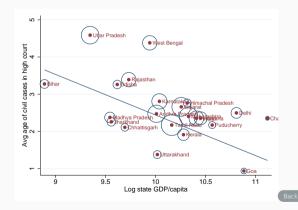
Determinants of Firm Growth Volatility

	Data (Pakistan)				Simulation					
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
log(Sales)	-0.138 (0.0018)		-0.092 (0.0025)	-0.105 (0.0022)	-0.103 (0.0022)	-0.3021 (0.0007)		-0.2424 (0.0009)	-0.2259 (0.0008)	-0.2256
$\overline{log(Buyers)}$		-0.217 (0.0031)	-0.111 (0.0042)				-0.4962 (0.0014)	-0.1845 (0.0018)		
$\overline{\log(\mathrm{HHI})}$				0.152 (0.0055)	0.202 (0.0067)				0.3179 (0.0017)	0.4224
log(HHI (weighted))					-0.051 (0.0037)					-0.1058 (0.0112
Fixed Effects Industry	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Ye
Statistics	0.262	0.244	0.206	0.207	0.200	0.7667	0.7202	0.7740	0.704	0.70
R ² R ² -within Observations	0.263 0.197 23,034	0.244 0.175 23,034	0.286 0.221 23,034	0.287 0.223 23,034	0.289 0.225 22,552	0.7667 0.2674 538,784	0.7393 0.1814 538,784	0.7713 0.282 538,784	0.781 0.3123 538,784	0.78: 0.312: 538,78

Standard errors in parentheses. The dependent variable is the log standard deviation of $\log sales_{t+1} - \log sales_t$.

Slow Courts

- · Contract disputes between buyers and sellers
- District courts can de-facto be bypassed, cases would be filed in high courts
- Court quality measure: average age of pending civil cases in high court



Mean Reversion: Pakistan

	Dependent variable: Change in log Sales		
	(1)	(2)	(3)
$\log Sales_{\ell-1}$	-0.146** (0.0051)	-0.163** (0.010)	-0.163** (0.011)
$\log \text{Sales}_{t-1} \times \text{Age civ. cases} \times \text{relspec}$		0.0114 ⁺ (0.0060)	0.0128* (0.0062)
Firm × 2-digit Industry FE	Yes	Yes	Yes
District FE	Yes	Yes	
Year FE	Yes	Yes	
Age FE			Yes
${\sf Industry} \times {\sf District} \times {\sf Year} {\sf FE}$			Yes
R^2	0.218	0.218	0.249
Observations	205351	205254	201931

Standard errors in parentheses, clustered at the district \times industry level.

Skewness of Size Distribution: Pakistan

	Dependent variable: Skewness of log Sales			
	(1)	(2)	(3)	
Relspec x Court Congestion	-0.914	-1.053 ⁺	-1.465 ⁺	
	(0.593)	(0.562)	(0.831)	
District FE	Yes	Yes	Yes	
2-digit Industry FE	Yes	Yes	Yes	
Statistic	25-75	25-90	50-90	
R ²	0.424	0.598	0.547	
Observations	935	688	688	

$$^{+}$$
 $p < 0.10, * p < 0.05, ** p < 0.01$

Skewness<sub>S
$$\omega$$</sub> = $\frac{\log (\text{Share of plants above } S_1) - \log (\text{Share of plants above } S_0)}{\log S_1 - \log S_0}$

 S_0 and S_1 are different quantiles of overall plant size distribution (25th, 50th, and 75th, 90th)

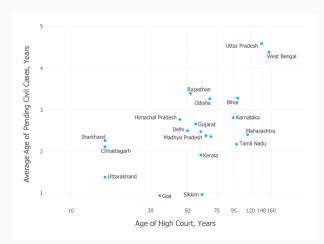


Notes on Pakistan

- \cdot 7 states, almost all economic activity is in two states, Sindh and Punjab
- · All of our data is in district courts
- VAT data: Size threshold: varies across years. 2-3k per year 15k per year
- · Can still register for VAT
- · Small firms effectively face sales tax
- Some sectors (notably agriculture, some services, companies owned by army) excluded from VAT
- For manufacturing, sum across firms of reported VA in data of firms represents 89% manufacturing VA as reported by National Accounts (for whole economy, much lower 30-40%)
- Currently use all transactions, whether reported by one or both parties. If parties disagree on value, use geometric mean of reported transactions
- Firms reports total sales separately from transactions For size, use declared sales of firm, not sum of transactions
- · Remove invoice mills
- · For firm: age (date registered), two digit industry codes (sometimes there is a

Endogeneity: IV

- · Since independence: # judges based on state population
- \Rightarrow backlogs have accumulated over time
 - But: new states have been created, with new high courts and clean slate



Weak Enforcement and Relational Contracts

- · Contract specifies level of defectiveness $\delta \in [0, 1]$. Surplus maximized at $\delta = 0$.
 - Supplier can produce defective input. Saves in cost, but possibility output will be defective.
 - · Claim can be enforced in court.
 - · But delay in court reduces value of payment
 - Cost proportional to value of transaction
- \cdot Static Nash: Supplier makes defective input, court. Priced in, but static surplus \downarrow
- Relational contract
 - Supplier chooses $\delta = 0$
 - Buyer chooses lower arrival rate of new suppliers (observable to supplier, not court)
 - · Backloads payoff, raises surplus of the relationship
 - · Enforcement: Trigger strategies
 - $\boldsymbol{\cdot}$ If supplier does not customize, buyer does not reduce arrival of new suppliers
 - · Punishment for defective inputs: Relationship ends faster + enforcement in court
 - · If buyer does not reduce arrival rate, supplier stops customizing

