

Commercialization and the Decline of Joint Liability Microcredit

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Joint Liability

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 - ▶ Costly state verification
 - ▶ Limited enforcement

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 - ▶ Adverse selection
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 - ▶ Costly state verification
 - ▶ Limited enforcement
- ▶ Long thought of as foundational to MFIs' very low default rates on unsecured credit to the poor.

- ▶ But over the last 10 years, numerous authors have pointed to a decline in the use of joint liability contracts.
 - ▶ Hermes and Lensink (2007), Armendáriz de Aghion and Morduch (2010), Breza (2011), Giné, Krishnaswamy and Ponce (2011), Feigenberg, Field and Pande (2013), Carpena et al. (2013), Giné and Karlan (2014)

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- ▶ **Essentially anecdotal**, centering around the decision of Grameen and BancoSol to switch to Individual Liability lending in early 2000s.
- ▶ **This paper:**
 - ▶ Is it true?
 - ▶ Why?

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 1. Growth of for-profit microcredit
 2. Increased competitiveness & improvement in borrowers' outside options

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- ▶ Our claim: two parallel forces we term **commercialization** predict decline of JL
 1. Growth of for-profit microcredit
 2. Increased competitiveness & improvement in borrowers' outside options
- ▶ No claims about other hard-to-measure forces (technological progress, preference change, “borrowers never liked JL”).
- ▶ **No normative claims**
 - ▶ Companion paper (de Quidt et al., 2016) - restricted model to analyze specific market structures & implications for welfare.

Outline

1. Document a set of facts about the microcredit industry.
2. Simple contracting model predicts a causal link from commercialization to decline of JL.
3. Three testable predictions.

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 - ▶ Broad move toward commercialization
 - ▶ Modest decline in JL lending in a 4-year panel
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3. Three testable predictions.

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1. Document a set of facts about the microcredit industry.
2. Simple contracting model predicts a causal link from commercialization to decline of JL.
 - ▶ For-profit and non-profit lenders.
 - ▶ Weak enforcement environment with dynamic repayment incentives.
 - ▶ JL improves repayment through social pressure.
 - ▶ Improved borrower outside options undermine repayment.
3. Three testable predictions.

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2. Simple contracting model predicts a causal link from commercialization to decline of JL.
3. Three testable predictions.
 - ▶ For-profits use JL less than non-profits
 - ▶ Non-profits decrease use of JL when borrowers' outside options improve
 - ▶ For-profits *increase* use of JL when borrowers' outside options improve

Not the first to identify an association between commercialization and contract type

- ▶ Cull et al. (2009): For-profit lenders less likely to use JL than non-profits.
- ▶ Karlan & Zinman (2009):

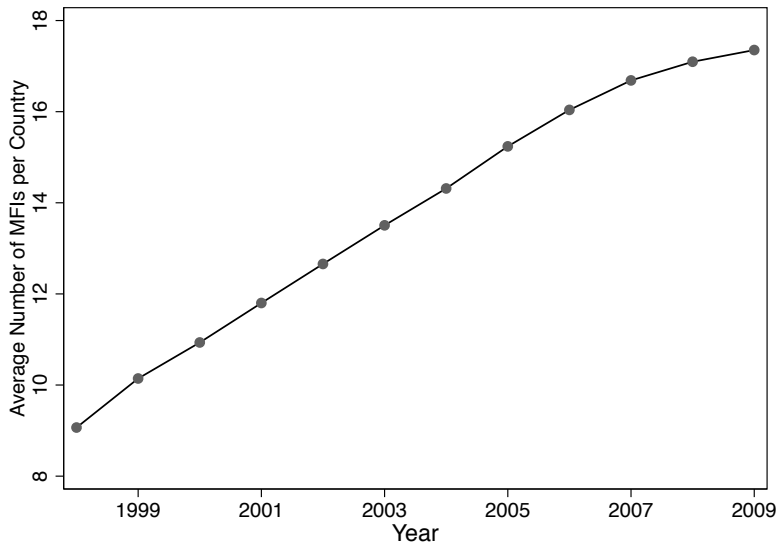
[T]he industrial organization of microcredit is trending toward something that looks more like the cash loan market: for-profit, more competitive delivery of untargeted, individual liability loans... This evolution is happening from both the bottom-up (non-profits converting to for-profits) and the top-down (for-profits expanding into subprime and consumer segments).

But (we think) first to formalize the association and claim causality

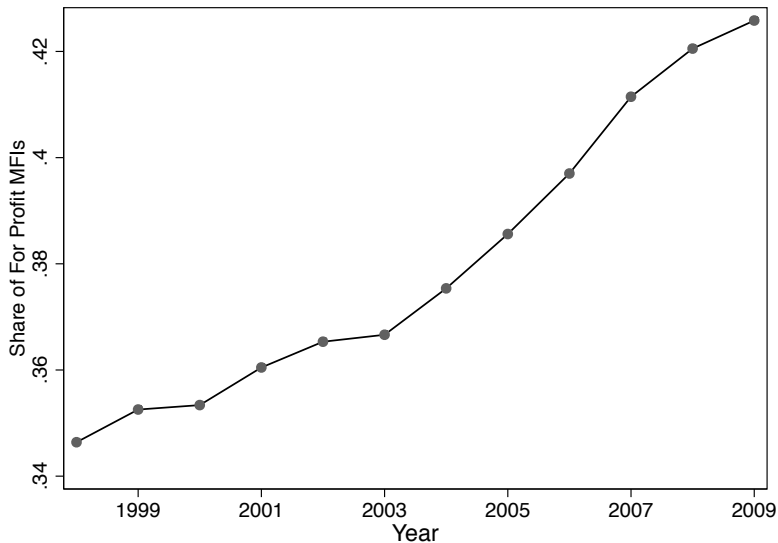
Other literature

- ▶ Microcredit contracts
 - ▶ Besley & Coate (1995), Ghatak & Guinnane (1999), many more
- ▶ Joint vs individual liability (empirical)
 - ▶ Giné & Karlan (2014), Carpena et al. (2013), Mahmud (2015), Attanasio et al. (2015)
- ▶ Importance of social capital
 - ▶ Besley & Coate (1995), Karlan (2005, 2007), Ahlin & Townsend (2007), Cassar & Wydick (2010), de Quidt et al. (forthcoming, 2016)
- ▶ Competition
 - ▶ Spillovers on other lenders: Hoff & Stiglitz (1997), McIntosh et al. (2005), de Quidt et al. (2016)
 - ▶ Spillovers on poor borrowers: McIntosh & Wydick (2005)
 - ▶ Spillovers on informal credit market: Demont (forthcoming)
- ▶ Impact of microcredit
 - ▶ AEJ: Applied 7(1) 2015: Banerjee et al., Tarozzi et al., Attanasio et al., Crépon et al., Angelucci et al., Augsburg et al.
 - ▶ Meta-analysis: Meager (2016)

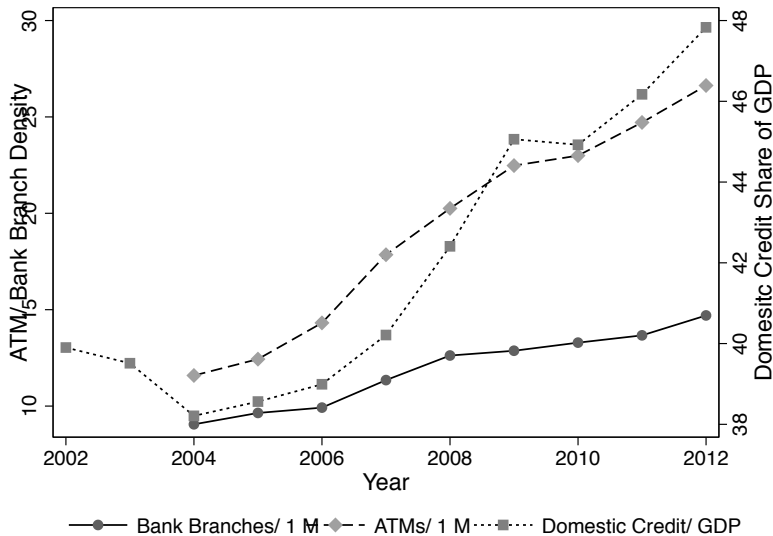
1. Steady growth of market scale



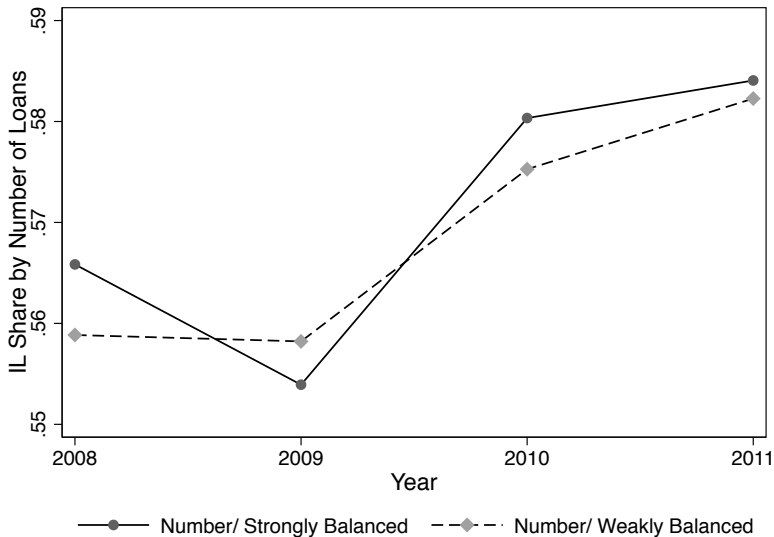
2. Increasing importance of for-profit lenders



3. Steady growth in other measures of financial penetration



4. Modest decline of joint liability lending



Many challenges to documenting trends in lending methodology

- ▶ Short time window
- ▶ Global trend driven by
 - ▶ Shift to IL within MFI portfolios (our figures)
 - ▶ Captures e.g. a JL lender switching to IL (like Grameen)
 - ▶ Relative growth of IL lenders (difficult to capture)
 - ▶ More entry/less exit of IL lenders (not observable)
- ▶ Selective reporting leads us to focus on within-MFI changes

- ▶ Study choice of lending methodology within a weak enforcement/strategic default/ex-post moral hazard model.
- ▶ Basic framework used in other work
 - ▶ Original form: Besley & Coate (1995)
 - ▶ This form: de Quidt et al. (forthcoming), de Quidt et al. (2016)
 - ▶ Also Allen (2015) same basic setup
- ▶ We focus the analysis on
 - ▶ (Exogenous) changes in for-profit/non-profit composition
 - ▶ (Exogenous) changes in competition or borrower outside options

- ▶ Each period, risk-neutral, infinitely lived borrowers have access to a productive investment opportunity that
 - ▶ costs 1
 - ▶ yields R with probability p
 - ▶ yields 0 with probability $1 - p$
- ▶ Investment payoffs independently distributed across borrowers.
- ▶ No assets and no saving
 - ▶ Must borrow to finance investment.
 - ▶ Consume all income each period.
- ▶ Can take *at most one* loan per period.
- ▶ Discount future payoffs by per-period discount factor δ

- ▶ Lenders are either **for-profit** or **non-profit**.
 - ▶ For-profit lenders maximize *per-period* profits from a given borrower.
 - ▶ Rationale: capacity constraints + costless replacement
 - ▶ Non-profit lenders maximize borrower welfare, subject to break-even
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- ▶ **Weak enforcement**: project returns are non-contractible
 - ▶ e.g. because state verification is prohibitively costly
- ▶ Dynamic repayment incentives: defaulting borrowers' contracts are terminated.
- ▶ **Limited liability**: unsuccessful borrowers *cannot* repay and are inefficiently terminated.

- ▶ Lenders offer take-it-or-leave-it individual liability (IL) or joint liability (JL) contracts.
- ▶ Loan of 1, gross repayment of r at period end.
- ▶ IL: defaulting borrower is terminated.
- ▶ JL: groups of two borrowers jointly liable. **Both** contracts terminated unless **both** loans repaid.
 - ▶ JL incentivizes successful borrowers to assist their unsuccessful partners with repayment.

Theory

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- ▶ de Quidt et al. (2016): model U explicitly as “waiting for credit.”
 - ▶ Defaulters enter a pool of “unmatched” borrowers, waiting for an available slot at a lender.
 - ▶ Competitive equilibrium, analogous to Shapiro & Stiglitz (1984) (also Ghosh & Ray, forthcoming).
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 - ▶ But U is an equilibrium object - hard to study *ceteris paribus* changes in competitiveness.
- ▶ This paper, fully **reduced form** approach: increased competitiveness reflected in increased U

Individual Liability

- ▶ Consider a borrower offered an IL loan at interest rate r .
- ▶ If she repays when successful, value function is:

$$\begin{aligned}V^{IL} &= p(R - r) + \delta p V^{IL} + \delta(1 - p)U \\ &= \frac{p(R - r)}{1 - \delta p} + \frac{\delta(1 - p)}{1 - \delta p}U\end{aligned}$$

- ▶ Repayment is incentive-compatible iff

$$\begin{aligned}\delta U &\leq \delta V^{IL} - r \\ r &\leq \delta p R - \delta(1 - p)U \equiv r_{IC1}(U)\end{aligned}\tag{IC1}$$

- ▶ IC1 implies $V^{IL} \geq U$ (participation/individual rationality constraint)

Since we are interested in the move from JL to IL, we assume:

Assumption

IL is always feasible: $\rho \leq p_{IC1}(U)$

Theory

Joint liability groups

Assumption

Within JL groups, borrowers can observe and contract on output realizations.

- ▶ Typical assumption in the microcredit literature
- ▶ They write contingent repayment contracts, “repayment rules,” specifying who repays what and when.
- ▶ Penalty for violating the repayment rule is a social sanction, S .
 - ▶ Not required for this paper, but simplifies comparative statics
 - ▶ Again, fully reduced-form treatment.
 - ▶ Many microfoundations - real punishment, loss of reputation, breakdown of social ties, collapse of other informal contracts, ...

Theory

Joint liability groups

- ▶ Focus on *efficient, stationary, symmetric* repayment rules.
 - ▶ Efficiency \Rightarrow max. borrower welfare + no social sanctions enacted in equilibrium.
 - ▶ Stationarity \Rightarrow stationary value function
 - ▶ Symmetry \Rightarrow representative borrower
- ▶ Can restrict attention to three rules:
 1. Always default.
 2. Repay when both are successful, default otherwise.
 3. Repay own loan when successful, and also repay partner's if she is unsuccessful.
- ▶ For simplicity, assume borrowers can always *afford* to repay partner's loan when successful. Sufficient condition:

Assumption

$$\delta p \leq \frac{1}{2}$$

Joint liability

- ▶ Now consider a pair of borrowers offered a JL contract at rate r .
- ▶ Efficient, symmetric repayment rule \Rightarrow if expected repayment is πr , contract renewal probability is π .

Intuition:

- ▶ Groups always repay both loans, or neither (efficiency).
 - ▶ \Rightarrow contract is renewed whenever my loan is repaid.
 - ▶ I am as likely to repay partner's loan as she is to repay mine (symmetry).
- ▶ Value function:

$$\begin{aligned}V^{JL} &= pR - \pi r + \delta\pi V^{JL} + \delta(1 - \pi)U \\ &= \frac{pR - \pi r}{1 - \delta\pi} + \frac{\delta(1 - \pi)}{1 - \delta\pi}U\end{aligned}$$

Incentive compatibility

- ▶ Step 1: repay own loan, when partner is repaying?

$$\begin{aligned}\delta U &\leq \delta V^{JL} - r \\ r &\leq \delta pR - \delta(1 - \delta)U \equiv r_{IC1}(U)\end{aligned}\tag{IC1}$$

- ▶ **Same condition as for IL**
- ▶ Note: did not invoke social sanction S , why?
If IC1 does not hold efficient repayment rule is “always default”
 $\Rightarrow S$ never needed to enforce individual repayment.
- ▶ If IC1 holds, welfare is increasing in π , so efficient rule will maximize repayment.

Theory

Incentive-compatibility constraints

- ▶ Step 2: repay partner's loan, when partner is not repaying?

$$\delta(U - S) \leq \delta V^{JL} - 2r \quad (\text{IC2})$$

- ▶ Larger values of S relax IC2, enhancing borrowers' ability to side-contract.
- ▶ If IC2 does not hold, efficient rule is “repay own loan when partner is successful”
- ▶ But this achieves lower **repayment**, **welfare** and **lender profit** than IL.
- ▶ So JL never offered if IC2 does not hold.

Theory

Incentive-compatibility constraints

- ▶ If IC2 holds, efficient rule is “Repay own loan when successful, and also repay partner’s if she is unsuccessful.”
- ▶ Expected repayment: $pr + p(1 - p)r = p(2 - p)r$ so

$$\pi = p(2 - p) \equiv q$$
$$V^JL = \frac{pR - qr}{1 - \delta q} + \frac{\delta(1 - q)U}{1 - \delta q}$$

and substituting into IC2:

$$r \leq \frac{\delta pR - \delta(1 - \delta)U + \delta(1 - \delta q)S}{2 - \delta q}$$
$$= \frac{r_{IC1}(U) + \delta(1 - \delta q)S}{2 - \delta q} \equiv r_{IC2}(U, S) \quad (IC2)$$

Recap

- ▶ IL contracts must satisfy:

$$r^{IL} \leq r_{IC1}(U)$$

achieving repayment rate p

- ▶ JL contracts must satisfy:

$$r^{JL} \leq \min\{r_{IC1}(U), r_{IC2}(U, S)\}$$

achieving repayment rate $q > p$

Non-profit lender

- ▶ Non-profit lender maximizes borrower welfare, subject to break-even.
- ▶ Break-even interest rates:

$$\hat{r}^{IL} = \frac{\rho}{p}$$

$$\hat{r}^{JL} = \frac{\rho}{q}$$

- ▶ Borrower welfare is higher under JL, so JL offered whenever possible. Why?
 - ▶ Higher repayment \Rightarrow less inefficient termination & lower r
- ▶ Non-profit offers JL iff:

$$\rho \leq q \min\{r_{IC1}(U), r_{IC2}(U, S)\}$$

or

$$S \geq \hat{S}(U)$$

For-profit lender

- ▶ For-profit lender maximizes (per-period) profits

$$\Pi = \pi r - \rho$$

- ▶ Charges the highest possible interest rate, subject to incentive-compatibility:

$$\tilde{r}^{LL} = r_{IC1}(U)$$

$$\tilde{r}^{JL} = \min\{r_{IC1}(U), r_{IC2}(U, S)\}$$

- ▶ Offers JL iff

$$p\tilde{r}^{LL} \leq q\tilde{r}^{JL}$$

or

$$S \geq \tilde{S}(U)$$

- ▶ Suppose that S is distributed in the population according to $F(S)$
 - ▶ e.g. different villages have stronger/weaker social ties
- ▶ Non-profit lender's IL share: $F(\hat{S}(U))$
- ▶ For-profit lender's IL share: $F(\tilde{S}(U))$
- ▶ IL shares are weakly monotone increasing in \hat{S}, \tilde{S} .

Prediction 1. For-profits vs non-profits

Proposition

Non-profits are weakly less likely to offer IL than for-profits:

$$\hat{S}(U) \leq \tilde{S}(U)$$

The inequality is strict if IL earns positive profits

Intuition

- ▶ Non-profits offer JL whenever it breaks even:
 $q \min\{r_{IC1}(U), r_{IC2}(U, S)\} - \rho \geq 0$
- ▶ For-profits offer JL whenever it breaks even **and** is more profitable than IL
 $q \min\{r_{IC1}(U), r_{IC2}(U, S)\} - \rho \geq pr_{IC1}(U) - \rho \geq 0.$

Prediction 2. Competition and non-profits

Proposition

Increases to U increase IL lending by non-profits

$$\hat{S}'(U) \geq 0$$

The inequality is strict if IL is offered for some U ($p > \delta q$)

Intuition

- ▶ Increases to U tighten incentive-compatibility constraints - decreasing the maximum interest rate the lender can charge.
- ▶ If $r_{IC2}(U, S) < r_{IC1}(U)$, tightening r_{IC2} may render JL loss-making.

Prediction 3. Competition and for-profits

Proposition

Increases to U decrease IL lending by for-profits

$$\tilde{S}'(U) \leq 0$$

The inequality is strict if IL is offered for some U ($p > \delta q$)

Intuition

- ▶ Increases in U tighten both IC2 and IC1, decreasing profits under both IL and JL.
- ▶ But JL interest rate is less sensitive.
- ▶ Heuristically, IC2 bounds $2r$, IC1 bounds r , so a ΔU tightening has a larger effect on IL than JL.

Useful discriminating prediction - unlikely to be generated by other correlates of commercialization and methodology change.

Taking stock

We identify commercialization with

- ▶ Growth in the market share of for-profits

- ▶ Increasing competition in the market

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Net effect in principle ambiguous, **but** for sufficiently low initial share of for-profits, commercialization induces a trend toward IL.

Several strong assumptions, useful to explore how crucial they are.

- ▶ No saving.
 - ▶ **Important.** Unbounded saving causes dynamic incentives to unravel and undermines repeat borrowing.
 - ▶ Evidence for saving constraints (e.g. Dupas and Robinson, 2013).
 - ▶ Repeat borrowing is common.

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- ▶ IL always feasible
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 - ▶ **Unimportant for the empirical exercise** as we only analyze within-MFI portfolio shifts.
- ▶ Borrowers can only take one loan at a time
 - ▶ **Unclear.** No simple way to model multiple borrowing.

- ▶ Competition influence only through borrowers' outside option.
 - ▶ **Not important.** Modeling “ex-ante” competition weakly reinforces our qualitative findings.
- ▶ Borrowers can always afford their JL payment.
 - ▶ **Not important.** Dropping this condition introduces an additional constraint but does not otherwise alter our qualitative conclusions.
- ▶ Bernoulli output distribution
 - ▶ **Somewhat important.** Other contracts become attractive for richer distributions (de Quidt et al., forthcoming)
- ▶ For-profit lenders are myopic.
 - ▶ **Not important.** Same qualitative results with forward-looking lenders (quantitatively weaker).
- ▶ Risk-neutral borrowers
 - ▶ **Not important.**
- ▶ Social sanctions
 - ▶ **Not important.** S permits continuous comparative statics.

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- ▶ Key observables: founding dates, for-profit/non-profit status, lending methodology.
 - ▶ Portfolios divided into “individual,” “solidarity group,” “village bank.”
 - ▶ We (and many others) treat “individual” as individual liability, the rest as joint liability.
 - ▶ Matches with what we know about practices for specific MFIs.
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 - ▶ Concern: not all “solidarity group” loans are JL. Identifying assumption: IL/JL breakdown not changing in confounding direction.
- ▶ Observe portfolio composition in 2008-2011, measured by value and by number of loans.

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- ▶ Construct two panels: *weakly balanced* and *strongly balanced* .
 - ▶ **Weakly balanced** (MFIs that report lending methodology at least twice): ~ 930 institutions, 100 countries
 - ▶ **Strongly balanced** (MFIs that report every year): ~ 380 institutions, 64 countries

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 - ▶ **Strongly balanced** (MFIs that report every year): ~ 380 institutions, 64 countries
- ▶ Caution: imperfectly representative of the MIX Market population (which may be imperfectly representative of the world).

Sample frame: MFIs

Table: MFI Characteristics for MFIs reporting IL share by Number of Loans

	Full Sample		Weakly Balanced			Strongly Balanced		
	Mean	N	Mean	N	p	Mean	N	p
IL Share by Number of Loans	0.60	1538	0.58	932	0.35	0.58	378	0.72
IL Share by Loan Value	0.64	1476	0.64	894	0.87	0.64	365	0.99
Non Profit	0.60	1408	0.60	932	0.94	0.66	378	0.19
Non-Regulated	0.33	1768	0.39	932	<0.01	0.46	378	<0.01
NGO	0.32	1898	0.36	932	0.01	0.44	378	<0.01
Portfolio at Risk 90 days	6.43	1732	5.71	930	0.26	4.86	378	<0.01
Return on Assets	-0.25	1657	0.62	930	<0.01	1.56	378	0.01
Profit Margin	-4.88	1741	0.45	931	<0.01	4.85	378	<0.01
MFI Risk Rating (1-5)	2.65	1920	2.95	932	<0.01	3.57	378	<0.01
Capital to Asset Ratio	36.77	1813	31.90	931	0.11	29.98	378	0.78
Debt to Equity Ratio	8.47	1772	4.84	931	0.16	7.10	378	0.08
Average Loan Balance	6405.76	1906	1448.17	932	0.66	1273.97	378	0.20
Cost per Borrower	304.37	1514	241.57	923	0.10	197.31	378	<0.01
Write Offs/ Assets	2.36	1623	2.21	929	0.31	2.21	378	0.58

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Table: Country characteristics

	Full Sample		Weakly Balanced			Strongly Balanced		
	Mean	N	Mean	N	p	Mean	N	p
Urban population share	0.47	113	0.47	100	0.57	0.51	64	0.03
Mobile Phones/100 people	74.16	112	73.13	99	0.39	82.21	63	0.01
Agriculture share in GDP	18.18	103	18.52	92	0.63	15.64	61	0.02
Industrial sector share in GDP	29.06	103	28.27	92	0.27	28.93	61	0.97
Service sector share in GDP	53.21	104	53.71	93	0.25	56.14	62	<0.01
Development Aid as share of GDP	6.72	107	6.19	95	0.51	5.31	61	0.17
GDP Growth Rate	3.87	111	3.99	98	0.17	3.82	64	0.99
GDP per capita	3.68	111	3.33	98	0.06	3.78	64	0.72
Domestic Credit / GDP	4.52	105	4.34	93	0.60	4.70	61	0.37
Commercial bank density	1.30	112	1.29	100	0.82	1.65	64	<0.01
ATM Density	2.26	110	2.16	98	0.21	2.61	63	0.07

Final caveat

- ▶ Our results turn out to be highly sensitive to inclusion of one country: Peru.
 - ▶ Not sensitive to any other country or any individual MFI.
- ▶ Our interpretation:
 - ▶ Peru experienced very rapid growth in our competition proxies over the period.
 - ▶ Stretches their interpretation as valid proxies.
- ▶ Results shown today exclude Peru
- ▶ Qualitative (sign) results largely hold up to inclusion, but point estimates shrink toward zero.

Basic empirical specification

$$IL_{icrt} = \alpha NP_i + \eta C_{ct} + \gamma NP_i \times C_{ct} + \mathbf{X}'_{ict} \beta + a_{icr} + b_t + \epsilon_{icrt}$$

- ▶ i : MFI, c : country, r : region, t : year (2008-2011)
- ▶ IL : IL share in portfolio
- ▶ NP : non-profit dummy
- ▶ C : competition proxy measure
- ▶ a : MFI/country/region fixed-effect
- ▶ b : time fixed-effect
- ▶ Standard errors clustered at country level

Basic empirical specification

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Predictions

1. $\alpha < 0$: non-profits use IL less than for-profits.
2. $\eta < 0$: for-profits decrease IL use when competition increases
3. $\eta + \gamma > 0$: non-profits increase IL use when competition increases
4. $\gamma > 0$: *non-profit IL respond more positively to competition*

No IV or natural experiment: instead rely on increasingly stringent fixed-effects specifications to (we hope) soak up spurious variation.

- ▶ Prediction 1: for-profits use IL more than non-profits
 - ▶ No within-MFI variation in profit status
 - ▶ Exploit within-region or within-country variation

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- ▶ Predictions 2 & 3: non-profit/for-profit competition response
 - ▶ Within-region, -country or -MFI variation over time in competition proxies
- ▶ Prediction 4: non-profits respond relatively more positively
 - ▶ All the above, + within-country-year variation between MFIs with different for-profit status.

Main results

Table: Non Profit Status, Competition and IL Lending

	Strongly Balanced			Weakly Balanced		
	(1)	(2)	(3)	(4)	(5)	(6)
	<i>Panel A: IL Share by Number of Loans</i>					
Commercial bank density	-0.059 (0.065)	-0.088* (0.052)	-0.023 (0.017)	-0.058 (0.036)	-0.047 (0.029)	-0.021 (0.014)
Non Profit	-0.139** (0.059)	-0.179** (0.077)		-0.098* (0.050)	-0.169*** (0.046)	
Non-Profit x Bank Branch Density	0.067 (0.052)	0.113* (0.060)	0.031 (0.020)	0.069** (0.029)	0.067** (0.033)	0.024* (0.014)
<i>Joint test:</i>						
Comp + Non-Profit x Comp = 0?	.008 (.0415)	.024* (.0132)	.008 (.0116)	.011 (.0253)	.02** (.00864)	.003 (.00621)
MFIs	348	348	348	878	878	878
Countries	64	64	64	94	94	94
Observations	1392	1392	1392	2756	2756	2756
Year FE	X	X	X	X	X	X
Region FE	X			X		
Country FE		X			X	
MFI FE			X			X

Main results

Table: Non Profit Status, Competition and IL Lending

	Strongly Balanced			Weakly Balanced		
	(1)	(2)	(3)	(4)	(5)	(6)
<i>Panel B: IL Share by Gross Loan Portfolio</i>						
Commercial bank density	-0.090 (0.064)	-0.100** (0.043)	-0.017 (0.012)	-0.075** (0.033)	-0.066** (0.026)	-0.018 (0.011)
Non Profit	-0.151*** (0.051)	-0.182*** (0.062)		-0.119*** (0.043)	-0.170*** (0.041)	
Non-Profit x Bank Branch Density	0.086* (0.049)	0.140** (0.054)	0.032 (0.023)	0.080*** (0.028)	0.092*** (0.032)	0.029 (0.018)
<i>Joint test:</i>						
Comp + Non-Profit x Comp = 0?	-.005 (.04)	.04** (.0156)	.015 (.017)	.005 (.024)	.026** (.0122)	.011 (.0116)
MFIs	340	340	340	831	831	831
Countries	60	60	60	92	92	92
Observations	1360	1360	1360	2605	2605	2605
Year FE	X	X	X	X	X	X
Region FE	X			X		
Country FE		X			X	
MFI FE			X			X

Summary of main results

1. Robust finding: non-profits use IL less than for-profits
2. Negative association between bank branch density and use of IL by for-profits
3. Positive (in most specifications) association for non-profits
4. *Competition effect relatively more positive for non-profits*

- ▶ Alternative proxies [▶ Go](#)
- ▶ Additional country-level controls [▶ Go](#)
- ▶ Additional MFI-level controls [▶ Go](#)
- ▶ Country×year fixed effects [▶ Go](#)
- ▶ Dropping village banks and MFIs with data discrepancies [▶ Go](#)
- ▶ Including Peru [▶ Go](#)
- ▶ Regulatory shocks [▶ Go](#)

Conclusion

- ▶ A modest decline in (within-MFI) JL usage over 2008-2011, alongside a long-run trend toward commercialized microcredit.
- ▶ Simple contracting model to capture main features of the environment:
 - ▶ Variation in lender motivation
 - ▶ Variation in borrower outside options
- ▶ Three testable predictions, broadly consistent with the data.
- ▶ Provides an explanation for changes in lending patterns through observable changes in the market environment.
- ▶ Future work:
 - ▶ Exploit credit bureau data & within-country variation in competitive environment
 - ▶ Natural experiments?

Alternative proxies

ATM Density [▶ Back](#)

Table: IL Share by Number of Loans: Robustness to Other Competition Proxy Variables

	Strongly Balanced			Weakly Balanced		
	(1)	(2)	(3)	(4)	(5)	(6)
ATM Density	-0.057 (0.055)	-0.050 (0.047)	-0.014 (0.031)	-0.055 (0.044)	-0.023 (0.037)	-0.010 (0.020)
Non Profit	-0.157** (0.059)	-0.209*** (0.076)		-0.109** (0.051)	-0.187*** (0.047)	
Non-Profit x ATM Density	0.042 (0.045)	0.057 (0.055)	0.006 (0.036)	0.028 (0.042)	0.026 (0.038)	0.013 (0.020)
<i>Joint test:</i>						
Comp + Non-Profit x Comp = 0?	-.015 (.0311)	.008 (.0286)	-.008 (.0289)	-.027 (.0223)	.003 (.0114)	.003 (.00823)
MFIs	346	346	346	864	864	864
Countries	63	63	63	91	91	91
Observations	1348	1348	1348	2667	2667	2667
Year FE	X	X	X	X	X	X
Region FE	X			X		
Country FE		X			X	
MFI FE			X			X

Alternative proxies

Credit/GDP [▶ Back](#)

Table: IL Share by Number of Loans: Robustness to Other Competition Proxy Variables

	Strongly Balanced			Weakly Balanced		
	(1)	(2)	(3)	(4)	(5)	(6)
Domestic Credit Share	-0.130** (0.054)	-0.189*** (0.058)	-0.112*** (0.037)	-0.097** (0.038)	-0.086* (0.049)	-0.110*** (0.038)
Non Profit	-0.153*** (0.051)	-0.227*** (0.061)		-0.113** (0.049)	-0.193*** (0.043)	
Non-Profit x Domestic Credit Share	0.109* (0.057)	0.174** (0.075)	0.066 (0.041)	0.077* (0.040)	0.086 (0.054)	0.094** (0.036)
<i>Joint test:</i>						
Comp + Non-Profit x Comp = 0?	-.021 (.0293)	-.015 (.0343)	-.045 (.0309)	-.02 (.0219)	.001 (.0367)	-.016 (.0208)
MFIs	338	338	338	833	833	833
Countries	61	61	61	88	88	88
Observations	1352	1352	1352	2640	2640	2640
Year FE	X	X	X	X	X	X
Region FE	X			X		
Country FE		X			X	
MFI FE			X			X

Country-level controls

▶ Back

Table: Additional country-level controls, IL shares by number of loans

	Strongly Balanced			Weakly Balanced		
	(1)	(2)	(3)	(4)	(5)	(6)
Commercial bank density	-0.110 (0.075)	-0.151*** (0.053)	-0.032 (0.020)	-0.073* (0.039)	-0.062* (0.034)	-0.026 (0.017)
Non Profit	-0.166*** (0.050)	-0.200*** (0.052)		-0.124*** (0.038)	-0.184*** (0.041)	
Non-Profit x Bank Branch Density	0.134** (0.065)	0.197*** (0.066)	0.048** (0.020)	0.103** (0.041)	0.102** (0.039)	0.038** (0.016)
<i>Further Interactions:</i>						
Urban population share	-0.026 (0.452)	-0.056 (1.283)	-0.990 (1.383)	-0.253 (0.266)	1.615 (1.418)	-0.322 (0.883)
Non Profit x Urban population share	0.129 (0.487)	-0.085 (0.526)	-0.207 (1.844)	0.094 (0.358)	-0.220 (0.386)	-0.288 (1.812)
Mobile Phones/100 people	-0.002 (0.001)	-0.001 (0.001)	-0.000 (0.001)	-0.000 (0.001)	-0.001 (0.001)	-0.000 (0.000)
Non Profit x Mobile Phones/100 people	-0.006*** (0.001)	-0.004** (0.002)	0.001 (0.001)	-0.007*** (0.001)	-0.004*** (0.001)	-0.000 (0.001)
GDP per capita	-0.006 (0.017)	-0.004 (0.008)	-0.013** (0.005)	-0.020* (0.012)	-0.006 (0.007)	-0.013** (0.005)
Non Profit x GDP per capita	0.018 (0.026)	0.007 (0.028)	-0.021** (0.009)	0.019 (0.022)	0.005 (0.023)	-0.008 (0.011)
<i>Joint test:</i>						
Comp + Non-Profit x Comp = 0?	.024 (.0524)	.046** (.0179)	.016 (.0111)	.03 (.0379)	.039*** (.0148)	.013* (.00711)
MFIs	334	334	334	792	792	792
Countries	58	58	58	82	82	82
Observations	1335	1335	1335	2517	2517	2517
Year FE	X	X	X	X	X	X
Region FE	X			X		
Country FE		X			X	
MFI FE			X			X

MFI-level controls and country×year FE

▶ Back

Table: Additional fixed effects and MFI-level controls

	Strongly Balanced				Weakly Balanced			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Non Profit	-0.179** (0.083)		-0.104 (0.074)		-0.170*** (0.049)		-0.133*** (0.042)	
Non-Profit x Bank Branch Density	0.112* (0.066)	0.016 (0.010)	0.117* (0.070)	0.020* (0.010)	0.067* (0.034)	0.017** (0.008)	0.076** (0.032)	0.015* (0.008)
MFIs	348	348	348	348	878	878	874	874
Countries	64	64	64	64	94	94	94	94
Observations	1392	1392	1347	1347	2756	2756	2611	2611
Country x Year FE	X	X	X	X	X	X	X	X
MFI FE		X		X		X		X
Controls			X	X			X	X

Controls include Diamonds, Capital to Asset Ratio, Debt to equity ratio, Average loan balance per borrower, Return on assets, Financial revenue/Assets, Yield on gross portfolio (nominal), Financial expense/assets, Operating expense/assets. Standard errors in parentheses are clustered at the country level, with stars indicating *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Drop observations

▶ Back

Table: Restricting the Analysis to Non-Village Banks, Institutions that did not switch Legal Status and have no Discrepancy in IL Shares reporting: Profit Status, Competition and IL Lending

	Strongly Balanced			Weakly Balanced		
	(1)	(2)	(3)	(4)	(5)	(6)
Commercial bank density	-0.055 (0.038)	-0.066 (0.042)	-0.019 (0.014)	-0.074** (0.032)	-0.042 (0.038)	-0.020 (0.013)
Non Profit	-0.178*** (0.062)	-0.199*** (0.074)		-0.090 (0.055)	-0.170*** (0.047)	
Non-Profit x Bank Branch Density	0.063 (0.047)	0.084 (0.052)	0.024 (0.016)	0.071** (0.035)	0.065 (0.041)	0.023* (0.013)
<i>Joint test:</i>						
Comp + Non-Profit x Comp = 0?	.009 (.032)	.017 (.0142)	.005 (.00935)	-.003 (.022)	.023** (.00922)	.004 (.00513)
MFIs	252	252	252	681	681	681
Countries	59	59	59	92	92	92
Observations	993	993	993	2060	2060	2060
Year FE	X	X	X	X	X	X
Region FE	X			X		
Country FE		X			X	
MFI FE			X			X

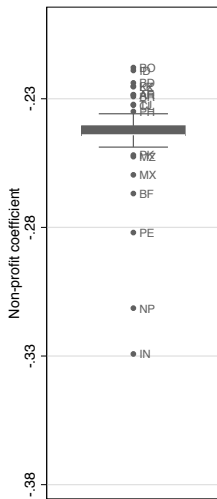
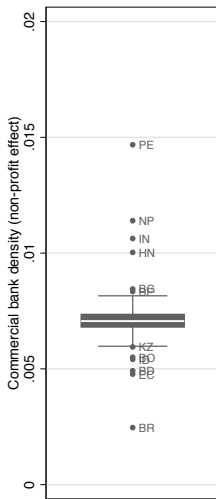
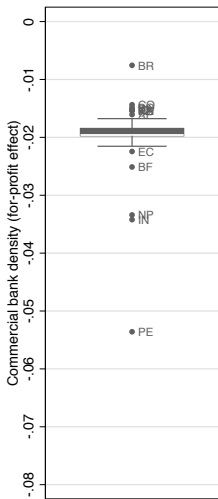
Including Peru

► Back

Table: Including Peru with IL Share by Number of Loans: Robustness to Other Competition Proxy Variables

	Strongly Balanced			Weakly Balanced		
	(1)	(2)	(3)	(4)	(5)	(6)
Commercial bank density	-0.023 (0.048)	-0.032 (0.052)	-0.009 (0.012)	-0.028 (0.037)	-0.029 (0.027)	-0.010 (0.009)
Non Profit	-0.156** (0.059)	-0.202** (0.077)		-0.112** (0.050)	-0.182*** (0.046)	
Non-Profit x Bank Branch Density	0.021 (0.051)	0.044 (0.066)	0.013 (0.013)	0.035 (0.036)	0.034 (0.037)	0.010 (0.010)
<i>Joint test:</i>						
Comp + Non-Profit x Comp = 0?	-.002 (.0342)	.012 (.0171)	.004 (.00987)	.006 (.0233)	.005 (.0157)	0 (.00574)
MFIs	378	378	378	932	932	932
Countries	65	65	65	95	95	95
Observations	1512	1512	1512	2934	2934	2934
Year FE	X	X	X	X	X	X
Region FE	X			X		
Country FE		X			X	
MFI FE			X			X

Dropping each country in turn



Regulatory Shocks

▶ Back

Table: Ruling out Time Varying Regulatory Shocks and Sensitivity to India: Profit Status, Competition and IL Lending

	Strongly Balanced			Weakly Balanced		
	(1)	(2)	(3)	(4)	(5)	(6)
Commercial bank density	-0.113** (0.054)	-0.138*** (0.043)	-0.149*** (0.045)	-0.059** (0.029)	-0.056 (0.037)	-0.059 (0.037)
Non Profit	-0.238*** (0.077)	-0.073 (0.059)	-0.093 (0.076)	-0.203*** (0.042)	-0.116** (0.046)	-0.125** (0.056)
Non-Profit x Bank Branch Density	0.142** (0.064)	0.198*** (0.062)	0.213*** (0.064)	0.081** (0.033)	0.091** (0.043)	0.095** (0.044)
<i>Joint test:</i>						
Comp + Non-Profit x Comp = 0?	.029** (.0144)	.06*** (.0226)	.064*** (.023)	.022** (.00835)	.036** (.0173)	.036** (.0174)
MFIs	310	348	310	803	878	803
Countries	63	64	63	93	94	93
Observations	1240	1392	1240	2506	2756	2506
India included?	No		No	No		No
Country FE	X	X	X	X	X	X
Year FE	X			X		
Region x Legal Status x Year FE		X	X		X	X