

Firms' Debt: Currency Choice and Exchange Rate Pass-Through

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Financial Dollarization: Causes and Consequences

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- Positive and normative questions:
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 - Q2: is the resulting risk-sharing efficient?
- Our approach:
 - **theoretical framework** \Rightarrow **sufficient statistic** for currency choice
 - **credit register + accounting data** \Rightarrow strong empirical support

- **Methodological approach:** Gopinath, Itskhoki & Rigobon (2010)
- **Theoretical models:** Froot, Scharfstein & Stein (1993), Bocola & Lorenzoni (2020), Gopinath & Stein (2020), Drenik, Kirpalani & Perez (2019), Aoki, Benigno & Kiyotaki (2016)
- **Empirical evidence:** Maggiori, Neiman & Schreger (2020), Christiano, Dalgic & Nurbekyan (2020), Salomao & Varela (2020), Baskaya, Giovanni, Kalemli-Ozcan, Peydro & Ulu (2018), Verner and Gyongyosi (2020)

THEORETICAL FRAMEWORK

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- A firm decides how to finance its investment project:

$$\max_{W, B^h, B^f} \mathbb{E}[W] - \frac{\bar{\gamma}}{2} \mathbb{V}[W]$$

$$\text{s.t. } W = \Pi - R^h B^h - \mathcal{E} R^f B^f$$

$$B^h + B^f = B$$

- B is total debt/investment
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- Normalize by the size of the debt:

$$\max_{w, x} \mathbb{E}[w] - \frac{\gamma}{2} \mathbb{V}[w]$$

$$\text{s.t. } w = (\pi - R^h) + (R^h - \mathcal{E} R^f) x,$$

- where $w = \frac{W}{B}$, $\pi = \frac{\Pi}{B}$, $x = \frac{B^f}{B}$, $\gamma = \bar{\gamma} B$

Sufficient Statistic

- Optimal debt composition:

$$x = \frac{\frac{1}{\gamma} \mathbb{E}[\psi] - \text{cov}(\pi, \psi)}{\mathbb{V}[\psi]}$$

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- ① increasing in **UIP deviations** $\mathbb{E}[\psi] \Rightarrow$ common to all firms
- ② decreasing in **riskiness** $\text{cov}(\pi, \psi) \Rightarrow$ depends on firm's cash flow

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- **Sufficient statistic** for FC debt:

$$x = \frac{\text{cov}(\pi, \Delta e)}{\mathbb{V}[\Delta e]} + \text{const}$$

- Can be directly estimated via OLS regression:

$$\pi = \alpha + \beta \Delta e + \varepsilon$$

- Advantages of the sufficient statistic:
 - independent of fundamentals behind π and Δe
 - can exploit both cross-firm and time variations
 - applies to any pair of currencies
 - extends to binary choice

$$x = 1 \quad \Leftrightarrow \quad \frac{\text{cov}(\pi, \Delta e)}{\mathbb{V}[\Delta e]} \geq \frac{1}{2} - \frac{\mathbb{E}\psi}{\gamma}$$

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- Limitations:
 - ψ may vary across firms (e.g. because of defaults)
 - maturity structure of debt
 - ...
- Going to control for these confounding factors

DATA

- Credit register
 - universe of bank loans taken by firms in Russia
 - 2016-19, ~147,000 firms each year
 - variables: loan size, maturity, interest rate, currency...
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- Match firms by taxpayer id
 - successful matching for ~80% of firms (~90% of debt)

Summary Statistics

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- Firms with multi-currency debt are extremely
 - scarce:
 - only 0.7% of firms (~800 out of >117,000) have any dollar loans
 - only 26% of these have loans of more than one currency
 - large: their median revenue is 214 times higher

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 - scarce:
 - only 0.7% of firms (~800 out of >117,000) have any dollar loans
 - only 26% of these have loans of more than one currency
 - large: their median revenue is 214 times higher
- Dollar debt is cheaper and more long-term

PRELIMINARY RESULTS

Preliminary Steps

- Want to estimate pass-through β

$$\Delta\pi_{it} = \alpha + \beta\Delta e_t + \varepsilon_{it}$$

- π_{it} is log of profits for firm i in year t
- e_t is log of ruble-dollar exchange rate
- Do it separately for firms with no \$ and only \$ bank loans
 - too few observations with non-trivial share of \$ loans
- Focus on long-term (>1 year) loans only
 - limited by annual frequency of profits
 - 82% of all debt is long-term

$$\Delta\pi_{it} = \alpha + \beta\Delta e_t + \varepsilon_{it}$$

	no \$ loans	only \$ loans
Baseline β	-0.02 (0.03)	3.26*** (0.71)
# of obs	386,169	810
# of firms	136,654	362

Notes. Clustered st. err. in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

$$\Delta\pi_{it} = \alpha + \beta\Delta e_t + \varepsilon_{it}$$

- Exclude observations with top 1% and bottom 1% of $\Delta\pi_{it}$

	no \$ loans	only \$ loans
Baseline β	-0.02 (0.03)	3.26*** (0.71)
Without outliers	-0.02 (0.03)	3.25*** (0.51)
# of obs	386,169	810
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- Weigh each observation with log of firm's revenue (averaged over the period)

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Baseline β	-0.02 (0.03)	3.26*** (0.71)
Without outliers	-0.02 (0.03)	3.25*** (0.51)
Weighted by revenue	0.05* (0.02)	3.43*** (0.70)
# of obs	386,169	810
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$$\Delta\pi_{it} = \alpha_i + \beta\Delta\mathbf{e}_t + \varepsilon_{it}$$

- Allow for firm-level fixed effects, α_i

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Baseline β	-0.02 (0.03)	3.26*** (0.71)
Without outliers	-0.02 (0.03)	3.25*** (0.51)
Weighted by revenue	0.05* (0.02)	3.43*** (0.70)
With firm-level trends	-0.01 (0.03)	2.04** (0.83)
# of obs	386,169	810
# of firms	136,654	362

Notes. Clustered st. err. in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

$$\Delta\pi_{it}^e = \alpha + \beta\Delta e_t^{e/d} + \varepsilon_{it}$$

- Euro-dollar pass-through for firms with either euro or dollar loans

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Without outliers	-0.02 (0.03)	3.25*** (0.51)
Weighted by revenue	0.05* (0.02)	3.43*** (0.70)
With firm-level trends	-0.01 (0.03)	2.04** (0.83)
# of obs	386,169	810
# of firms	136,654	362
Euro vs dollar	1.28 (1.41)	5.32*** (1.14)
# of obs	1,135	1,836

Notes. Clustered st. err. in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Intensive Margin

- So far, have enough variation only for extensive margin
 - firms that either never or always borrow in \$
- Get more observations for intensive margin
 - include short-term debt
 - allow for gaps in the data (fixed effects instead of first differences)
 - allow firms to switch groups over time

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	no \$	few \$	some \$	many \$	only \$
Fixed effects spec.	0.36*** (0.04)	2.96*** (1.02)	3.04** (1.48)	4.10*** (1.07)	5.67*** (0.75)
# of obs	298,624	360	344	345	1,072
# of firms	148,201	191	176	202	596

Notes. Clustered st. err. in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

- Reasons and implications of wide use of the dollar
- First look at evidence based on cross-firm variation
- Preliminary answers
 - Q1: how do firms choose between LC and FC debt?
 - A1: support for model with tractable sufficient statistic
 - Q2: is the resulting risk-sharing efficient?
 - A2: more FC debt is held by firms that are better hedged against this risk
- Major challenge is to find enough variation despite high concentration