

THE EFFECTIVENESS OF HOUSING COLLATERAL TIGHTENING POLICY

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INTRODUCTION

Effectiveness of macroprudential policy:

- ▶ Long-term: consensus on impact
- ▶ Challenges to identification of transmission channel:
 - ▶ Confounding macro factors
 - ▶ Measures usually proposed as *packages*
 - ▶ Supply vs. demand response
 - ▶ Selection vs. treatment effects
- ▶ Conventional view:
High LTV borrowers are riskier and more likely to become delinquent
- ▶ Key question:
Does collateral tightening attract the right type of borrowers?

- ① Exploit collateral tightening policy intervention in market for *investor loans*
- ② Consequence of the policy roll-out:
 - ▶ No change in loan take-up
 - ▶ Loans are *more* likely to become delinquent
- ③ Substantial shift in the composition of borrowers:
 - ▶ Lower credit quality and less liquid assets
 - ▶ Optimists: Take the risk of a liquidity crunch to bet on the housing market
- ④ Persistence of selection effect: 1 year.
- ⑤ External validation: aggregate effect on mortgage bankruptcy

- ▶ Collateral constraints: mortgage lending and credit card debt
 - ▶ Qi and Yang, 2009; Mian and Sufi, 2011; Fuster and Zafar, 2015; Agarwal et al., 2015, Corbae and Quintin, 2015; Agarwal and Qian, 2016
- ▶ Effectiveness of macroprudential policy
 - ▶ Akinci and Olmstead-Rumsey, 2015; Cerrutti, Claessens and Laeven, 2015; McDonald, 2015; Tressel and Zhang, 2016
- ▶ Liquidity channel of collateral policy:
 - ▶ Boz and Mendoza (2014); Wong, Ho and Tsang (2015)

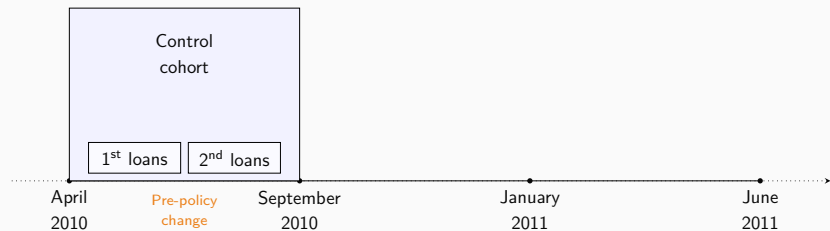
DATA AND METHODOLOGY

- ① Mortgage issuance and performance
 - ▶ Proprietary dataset from large Asian bank
 - ▶ LTV ratio, interest rate, penalties/delinquency
- ② Demographic information about Singapore residents
- ③ Credit card payment and spending histories
- ④ Checking account balances
 - ▶ Proprietary datasets from the same bank
 - ▶ Sample: mortgage borrowers
- ⑤ Bankruptcy cases in Singapore
- ⑥ Residential real estate transactions
 - ▶ Distinguish *owner-occupiers* from *investors*

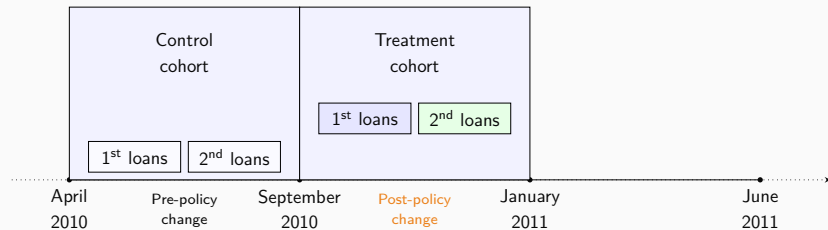
Policy change in Singapore

- ▶ *"For property buyers who already have one or more outstanding housing loans at the time of the new housing purchase:*
 - ① *Increase the minimum cash payment from 5% to 10%.*
 - ② *Decrease the Loan-to-Value (LTV) limit for housing loans granted by financial institutions from the current 80% to 70%."*
- ▶ *"The measures will take immediate effect on 30 August 2010."*
- ▶ **Note:** Second-loan market accounts for around 25% of outstanding loans in Singapore.

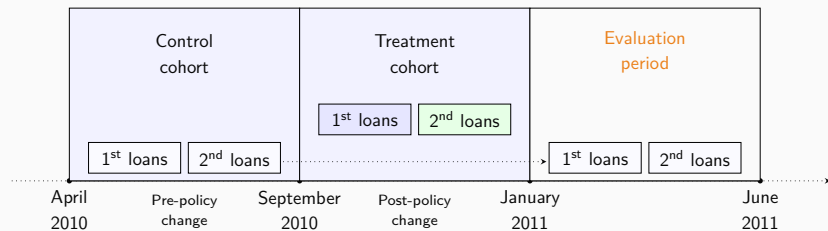
Identification approach



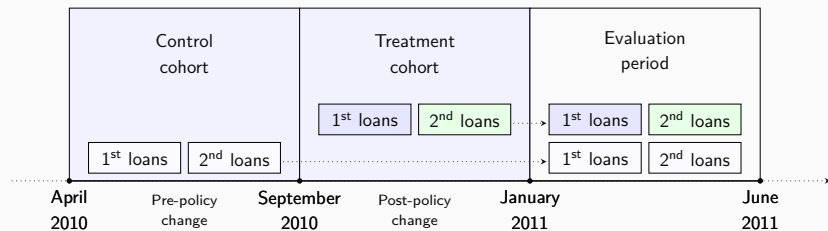
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- ▶ Benchmark estimated specification:

$$y_{i,t,n} = \tau_t + \underbrace{\xi X_{i,t,n}}_{\substack{\text{Loan and} \\ \text{borrower} \\ \text{characteristics}}} + \beta_1 \mathbf{1}_{n=2} + \beta_2 \mathbf{1}_{post} + \beta_4 \underbrace{\mathbf{1}_{n=2}}_{\substack{\text{Second loan}}} + \underbrace{\mathbf{1}_{post}}_{\substack{\text{Post-policy} \\ \text{borrower} \\ \text{cohort}}} + \varepsilon_{i,t,n}.$$

- ▶ Does the policy affect borrower behaviour?
- ▶ Are composition changes persistent?

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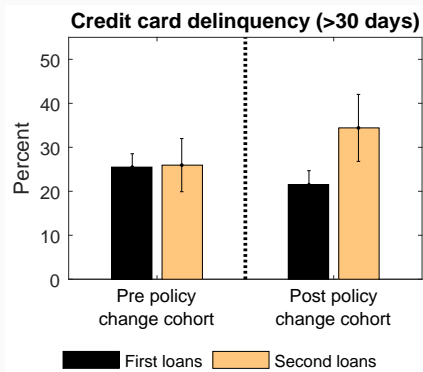
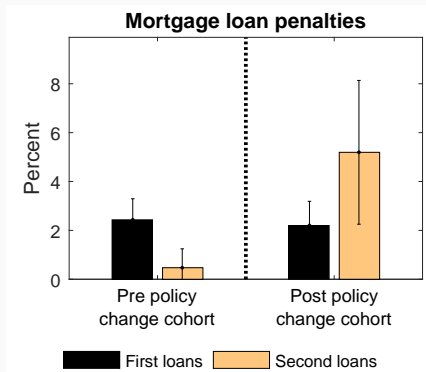
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RESULTS

Policy effects across cohorts



Loan characteristics across loan cohorts

	Pre-policy cohort	
	1 st loan	2 nd loan
Loan and property characteristics		
LTV ratio (<i>percent</i>)	68.14	66.97
Mortgage interest rate spread (<i>percent</i>)	1.70	1.69
Private property (<i>share</i>)	0.46	0.62
Property value (<i>'000s</i>)	\$1,021.15	\$1,489.06
Loan maturity (<i>years</i>)	25.06	24.44

Borrower characteristics

	Pre-policy cohort	
	1 st loan	2 nd loan
Borrower characteristics		
Average age (<i>years</i>)	41.19	44.48
Average income per year (<i>'000s</i>)	\$140.67	\$182.90
Length of tenure with the bank (<i>years</i>)	14.73	16.08
Foreign national (<i>share</i>)	0.30	0.23
Male (<i>share</i>)	0.76	0.83
Married (<i>share</i>)	0.58	0.70
Professional occupations (<i>share</i>)	0.52	0.51
Administrative occupations (<i>share</i>)	0.21	0.28
Graduate and postgraduate education (<i>share</i>)	0.72	0.83

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Borrower risk profile

	Pre-policy cohort	
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Borrower risk profile		
Credit card debt	\$469.60	\$589.19
Delinquency (<i>>30 days, frequency</i>)	0.25	0.26
Behavioural credit score (<i>units</i>)	752.24	762.61

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Estimation of selection effect

$$y_{i,n} = \beta_1 + \beta_2 \mathbf{1}_{n=2} + \beta_3 \mathbf{1}_{post} + \underbrace{\beta_4 \mathbf{1}_{n=2} \mathbf{1}_{post}}_{\substack{\text{Selection} \\ \text{effect}}} + \varepsilon_{i,n}$$

Loan and property characteristics

LTV ratio (<i>percent</i>)	-5.74***
Mortgage interest rate spread (<i>percent</i>)	0.13*
Private property (<i>share</i>)	0.13**
Property value (<i>'000s</i>)	-\$23.75

Borrower characteristics

Average age (<i>years</i>)	1.11
Average income per year (<i>'000s</i>)	-\$1.72
Length of tenure with the bank (<i>years</i>)	0.98
Foreign national (<i>share</i>)	-0.03
Administrative occupations (<i>share</i>)	0.00
Graduate and postgraduate education (<i>share</i>)	-0.03

Estimation of selection effect

$$y_{i,n} = \beta_1 + \beta_2 \mathbf{1}_{n=2} + \beta_3 \mathbf{1}_{post} + \underbrace{\beta_4 \mathbf{1}_{n=2} \mathbf{1}_{post}}_{\substack{\text{Selection} \\ \text{effect}}} + \varepsilon_{i,n}$$

Borrower risk profile

Credit card debt	\$318.29
Delinquency (>30 days, frequency)	0.12**
Behavioural credit score (<i>units</i>)	-257.00*

Estimation of selection effect

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Borrower risk profile

Credit card debt	\$318.29
Delinquency (>30 days, frequency)	0.12**
Behavioural credit score (units)	-257.00*

ECONOMIC MECHANISM

Within-cohort behavioral response

$$y_{i,t,n} = \delta_t + \alpha 1_{post} + \underbrace{\beta 1_{n=2}}_{\text{Characteristics}} + \underbrace{\gamma 1_{post} 1_{n=2}}_{\text{Selection effect}} + \underbrace{\tau 1_{post} 1_{n=2} 1_{obs}}_{\text{Treatment effect}} + \varepsilon_{i,t,n}.$$

		Checking account	Total spending	Dining out	Services	Durable goods
Characteristics	β	0.38***	0.25***	0.12***	0.13***	0.08***
Selection effect	γ	-0.08***	-0.14***	-0.11***	-0.06***	-0.03
Treatment effect	τ	-0.06	0.00	0.10	-0.01	-0.02

Effect heterogeneity

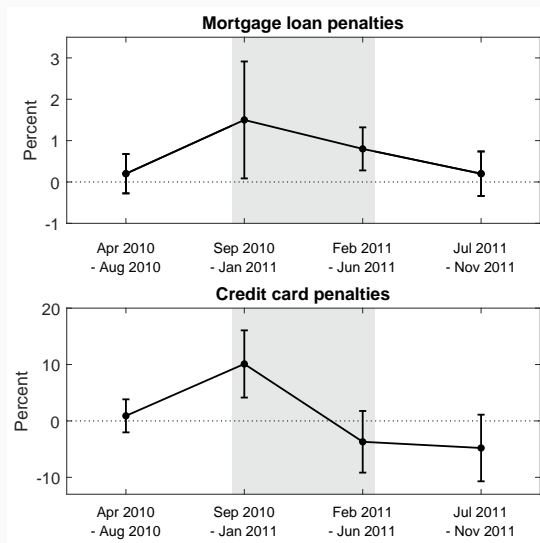
Dependent variable:	Unconditional	Conditional	
<u>Mortgage loan penalties</u>		Pre-policy house prices	Post-policy house prices
Pre-policy control cohort	-0.001	-0.005	-0.003
Post-policy treatment cohort	0.029***	0.003	0.014
District-level house prices		0.003	0.003
Pre-policy cohort (interaction)		0.011	-0.008
Post-policy (interaction)		0.049***	-0.026***
Behavioural score		-0.008***	-0.008***
Pre-policy (interaction)		0.003	0.002
Post-policy cohort (interaction)		-0.047***	-0.052***
Time fixed effects	Yes	Yes	Yes
Borrower characteristics	Yes	Yes	Yes
Cohort fixed effects	Yes	Yes	Yes
Number of observations	16,705	16,516	16,516
Adjusted R ²	0.005	0.037	0.037

Within-Borrower Spillovers

$$1_{p_{i,t,n}>0} = \mu_i + \beta X_{i,t,n} + \delta_1 1_{n=1,t \in [\text{Sep 2010, Jan 2011}]} + \delta_2 1_{n=2} + \varepsilon_{i,t},$$

	Mortgage loan penalties
First loan (Post-policy)	0.037* (0.020)
Second loan (Post-policy)	0.023* (0.013)
Borrower controls	Yes
Borrower fixed effects	Yes
Number of observations	276
Adjusted R ²	0.261

Persistence of cohort effects



EXTERNAL VALIDATION

Alternative dataset: mortgage bankruptcies

$$1_{investor,i,t} = \sum_{j=1}^5 \gamma_j 1_{cohort=j} + \varepsilon_{i,t}$$

	Mortgage bankruptcy
Apr 2010 - Aug 2010	0.07 (0.13)
Sep 2010 - Jan 2011	0.15 (0.12)
Feb 2011 - Jun 2011	0.19* (0.10)
Jul 2011 - Nov 2011	0.01 (0.15)
No. of obs.	94
Adjusted R ²	0.008

CONCLUSIONS

- ① Change in composition of borrowers suggests that the collateral tightening policy elicits a supply response towards relatively riskier individuals:
 - ▶ More optimistic and choose to take the risk of a liquidity crunch to bet on the housing market
 - ▶ Overestimate the possibility to repay and don't adjust consumption behaviour
- ② This phenomenon can alter the transmission mechanism that policy makers usually assume, delay and deteriorate the effectiveness of ad-hoc measures meant to deter speculation in the housing market

Loan origination volumes

