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Fiscal Stimulus and Household Debt through the Lens of 80-million Loan Accounts

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Discliamer: The views expressed in this study are our own and do not represent those of the Bank of Thailand or the National Credit Bureau. Data provided in this study is from statistical database of the National Credit Bureau.



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National Credit Bureau (NCB) Statistical Database

	Mar 2016
Loan outstanding (trillion baht)	8.7
Number of borrowers (million)	16.05
Number of active accounts (million)	49.1
NCB member	89

Source: Statistical database of the National Credit Bureau (authors' calculation)

Note: 1. Data from comsumer statistical database

- 2. BAAC not yet include in this study as it becomes member of NCB in Q2,16
- 3. Members comprise of 19 commercial banks, 5 SFIs, 65 non-banks
- (i.e. credit card firms, hire purchase firms, insurance companies, etc.)

Unique Features

1) Coverage

- Wide population (covering 74% of total household debt)
- Consolidated view of financial institutions relations for each borrower

2) Granularity

• Account level information



Roadmap

1) New perspectives of household debt through the lens of credit bureau data

2) Using credit bureau data to understand potential impacts of fiscal stimulus



1) New perspectives of household debt through the lens of credit bureau data





Helicopter view of household debt (Mar 2016)

Only 24% of population borrow

Debt burden is highly concentrated





Geographical distribution of household debt

- Distribution of debt intensity much different than those of headcount
- High debt intensity for borrowers in the Northeast







Geographical distribution of household debt

- Mean and median debt per borrower high in the Northeast
- Concentration is high across regions, especially Bangkok and vicinity





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Distribution of household debt by age

- Access to debt low for working age while debt burden is high and concentrated
- Duration of peak debt is long and debt run-down stunted at old age





Helicopter view of delinquent loan (Mar 2016)

✤ 16.4% of borrowers have delinquent loans. Delinquency severity is high.



- 1. Delinquent loans: loans that is over 90 days past due.
- 2. Defaulter: if borrower is default in one account , the person will be marked as defaulter.
- 3. Delinquent loan rate per defaulter can be overestimated given that no recovery rate is taken into account.



Geographical distribution of delinquent loan

Highest delinquency in the South, Bangkok and vicinity and lowest in the North







Distribution of delinquent loan by age

Delinquent loan headcount is high for young adult in early 30's





Borrower debt portfolio (Mar 2016): What financial institutions do people use?

Substantial segmentation across institutions...



64.3% of borrowers use only ONE type of financial institution

52.7% of borrowers borrow from only ONE institution

* Others include non-banks (i.e. credit card firms, hire purchase firms, etc.) and insurance company, co-operative



Borrower debt portfolio (Mar 2016): What types of loan do people have?

Again substantial segmentation across products



54% of borrowers only use unsecured loans

3.5% of borrowers with all account types make up 19.7% of total loan

13.7% of borrowers with housing loan account for 50.1% of total loan

High delinquency among those with car and unsecured loans

1. Delinquent loan: loans that is over 90 days past due

Note:

2. Unsecured loan comprises of p-loan, OD, credit card and others



Helicopter view of debt growth (2009-2016)

Debt growth from 2009-2016 comes from rise in both headcount as well as debt per borrower



Note: total population in Dec09 and Dec15 is 63,525,062 and 65,729,098, respectively.



Geographical distribution of debt growth (2009-2016)

Catch-up growth (both access and debt per borrower) in credit in the Northeast



Debt Headcount

2016 (Q1)



Geographical distribution of debt growth (2009-2016)

Catch-up growth (both access and debt per borrower) in credit in the Northeast



Debt per borrower 2016(Q1)



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Composition of debt growth (2009-2016)

Debt growth by product



Debt growth by age



Growth of auto loan: by borrower type



- Highest growth during 2010-2011
- ✤ 2013-2015 growth mainly from home loan
- 2011-2013 growth dominated by auto loan and age 25-35
- Unusual high contribution of new entry borrower to 2011-2013 auto loan growth



2) Using credit bureau data to understand potential impacts of fiscal stimulus

Thailand has seen a wide array of fiscal stimulus programs over the past decade

Thailand's major fiscal stimulus programs since 2008 Real estate fee cut First house First car Village funds First house Real estate fee cut Rice pledging Flood soft loans Tax deduction for TKK check Village funds restaurant/hotel 2008 2010 2012 2014 2016 Village funds Rice pledging Rice pledging Tax deduction for shopping/hotel Farmer payouts: 1000 per rai 5 million per Tambon

Here we use NCB statistical data to enhance our understanding of their effects on individuals and economy

Application: First-car tax rebate program

Overview of the first car program

- An excise tax rebate for first-time car buyers
- Rebate of 50,000-100,000 baht
- Sep 2011-Dec 2012 (Over 90% bought cars by end of 2013)



Durable goods stimulus	 Many countries introduced such programs to revive auto demand during the global financial crisis Details vary but most also had the aim of removing inefficient vehicles from the road US (2009), France (2009), Japan (2009-2010)
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How could the first-car program help boost the economy?



Overview of the study



- Difference-in-Difference technique
- Exploit nationwide variation in exposure to the program

Does the first car scheme impact loan performance/portfolio of participants?

Individual-Level Empirical Strategy

- Treatment group = First-time buyers who bought qualified models <u>during</u> the program period
- Control group = First-time buyers who bought qualified models <u>outside</u> the program period
- Compare the responses before and after the purchase decision

New loan probability before and after the firstcar purchase (Treatment vs. Control)



Quarters relative to the first-car purchase

Source: Authors' estimate

Results indicate higher delinquency probability for the first car scheme participants

The delinquency probability is significantly higher on the first car..

Change in delinquency probability on the first car of the treatment group relative to the control group (%)

..and there is also an evidence of higher delinquency on other loans

Change in delinquency probability on unsecure loans of the treatment group relative to the control group (%)



Quarters relative to the first-car purchase

Quarters relative to the first-car purchase

The results suggest that the first car program might have encouraged individuals to prematurely originate auto loans

Notes: 1) The shaded band indicates 90% confidence level. 2) Unsecure loans include credit card, personal and OD loans Source: Authors' estimate

The first car scheme participants also have significantly lower new loan probability

The overall new loan probability is significantly lower for the treatment group..

Change in new loan probability on all loans of the treatment group relative to the control group (%)

..with the effect being more evident on secure loans

Change in new loan probability on secure loans of the treatment group relative to the control group (%)



Quarters relative to the first-car purchase



Quarters relative to the first-car purchase

Notes: 1) The shaded band indicates 90% confidence level. 2) Secure loans include mortgage and hire purchase loans. Source: Authors' estimate

The effects on new loan probability are evidently different between buyers of passenger cars and trucks

The fall in new loan probability concentrates only among passenger car buyers,

Change in new loan probability on all loans of the treatment group relative to the control group (%)



Quarters relative to the first-car purchase

.. with no impact among truck buyers

Change in new loan probability on all loans of the treatment group relative to the control group (%)



Quarters relative to the first-car purchase

While the program crowds out new loans when the stimulus was on consumer durables, there was no impact for subsidies to productive business durables

Notes: The shaded band indicates 90% confidence level. Source: Authors' estimate

Does the first car scheme produce any measurable effect on local economies?

Share of first-car scheme participants across postcodes

Darker shading indicates higher share of program participants



Postcode-Level Empirical Strategy

- Utilize postcode-level quarter data on loan growth and delinquencies
- Exposure variable = Share of the number of first-car program participants to the total number of borrowers
- Significant variation in the exposure to the program across the nation—the share ranges from 2% to 18%.

Source: Authors' estimate

The first car program produces significantly adverse impacts on local economies

Areas with higher exposure experience significantly lower loan growth..

Estimates of the effects of the first car program on loan growth (%)



.. and higher delinquency rates

on delinguency share (%)

Estimates of the effects of the first car program

These results could be driven by both participants themselves as well as spillover effects to other borrowers

Note: The figures show coefficients on the interaction between the share of program participants and the dummy indicating post-program periods, controlling for postcode, year and province-year fixed effects.

Key Takeaways

Significant adverse effects on program participants—undermining potential stimulus benefits

Important distinction between consumer and business durables more limited negative effects among truck buyers

Possible spillover effects—areas with more program exposure have experienced lower loan growth and higher delinquencies

Administrative micro-level data could help inform policymakers on the design of future fiscal stimulus programs