Credit Allocation and Macroeconomic Fluctuations

Karsten Müller
Princeton University

Emil Verner
MIT

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Motivation

Rapid credit expansions are often, *but not always*, followed by economic downturns (Schularick-Taylor, 2012; Mian et al. 2017; Greenwood et al., 2020)

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But how credit interacts with business cycles remains poorly understood

- Why do some credit expansions end badly, while others are linked to growth spurts?
- How can we tell apart “good” from “bad” booms (Gorton & Ordoñez, 2020)?
- Does it matter who gets the borrowed money during credit booms?

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This paper: role of sectoral allocation of credit for understanding linkages between credit booms, macroeconomic fluctuations, and financial crises

Why focus on the allocation of credit across sectors?

Motivated by models of credit cycles with sectoral heterogeneity (e.g. Schneider-Tornell, 2004)

- Main distinction: tradable (T) vs. non-tradable (NT) and household sectors
- Key frictions: (1) sensitivity to credit supply shocks; (2) sensitivity to household demand

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Channels linking NT and HH credit to economic downturns

- Fueling unsustainable demand booms (e.g. Schmitt-Grohé-Uribe, 2016; Mian-Sufi-Verner, 2020)
- Contributing to financial fragility (e.g. Schneider-Tornell, 2004; Kalantzis, 2015)
- Contributing to intersectoral misallocation (e.g. Reis, 2013; Benigno-Fornaro, 2014)

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Yet prominent theories of credit cycles do not emphasize borrower heterogeneity (e.g. Brunnermeier-Sannikov, 2014; Bordalo-Gennaioli-Shleifer, 2016)

- Whether the allocation of credit matters empirically is an open question

This paper

To test for a role of sectoral credit allocation, we construct a new cross-country panel database from more than 600 individual sources, many newly digitized

Comparison with Existing Data Sources on Private Credit

Dataset	Start	Countries	Sectors
BIS	1940	43	2
IMF GDD	1950	83	2
Jordà et al. (2016)	1870	17	3
Müller and Verner (2020)	**1940**	**116**	**2–60 (mean=16)**

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We use these data to study the link between sectoral credit, business cycles, and crises

Main results

1. **Stark differences in macro outcomes across sectoral credit expansions**

- Credit to non-tradable and household sectors predict slower medium-run growth
- Credit to tradable sector predicts stable or even stronger growth

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- NT and HH credit predict demand booms and busts
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Takeaway: whether credit booms are “good” or “bad” depends on what credit is used for

- Distinguishing varieties of firm credit expansions is important

Related literature

1. **Macro-financial linkages**

- Credit and crises: e.g. Borio and Lowe (2002); Reinhart and Rogoff (2009); Gourinchas and Obstfeld (2012); Schularick and Taylor (2012); Jordà, Schularick, and Taylor (2016); Baron and Xiong (2017); López-Salido, Stein, Zakrajšek (2017); Krishnamurthy and Muir (2017); Mian, Sufi, and Verner (2017, 2020); Gorton and Ordoñez (2019); Brunnermeier, Palia, Karthik, and Sims (2020); Greenwood, Hanson, Shleifer, and Sørensen (2020); Giroud and Mueller (2020), Richter and Diebold (2021)
- Credit and growth: e.g. Goldsmith (1969); King and Levine (1993); Rajan and Zingales (1998); Levine, Loyaza, and Beck (2000); Beck et al. (2012)

2. **International macroeconomics**

- e.g. Mendoza (2002); Schneider and Tornell (2004); Tornell and Westermann (2005); Mendoza and Terrones (2008); Benigno and Fornaro (2014); Kalantzis (2015); Schmitt-Grohé and Uribe (2016); Bleck and Liu (2018)

→ **Whether credit expansions end badly depends on what credit is used for, along lines emphasized by open economy models**

A new database on sectoral credit

$>$ 600 sources, $^1/_3$ newly digitized
Mainly: statistical yearbooks, central banks

Previously unpublished data
provided by central banks and regulators

Systematic coding of classification changes
help from 150 employees of national authorities

Extensive documentation
data appendix, spreadsheets, code routines

Sectoral credit database

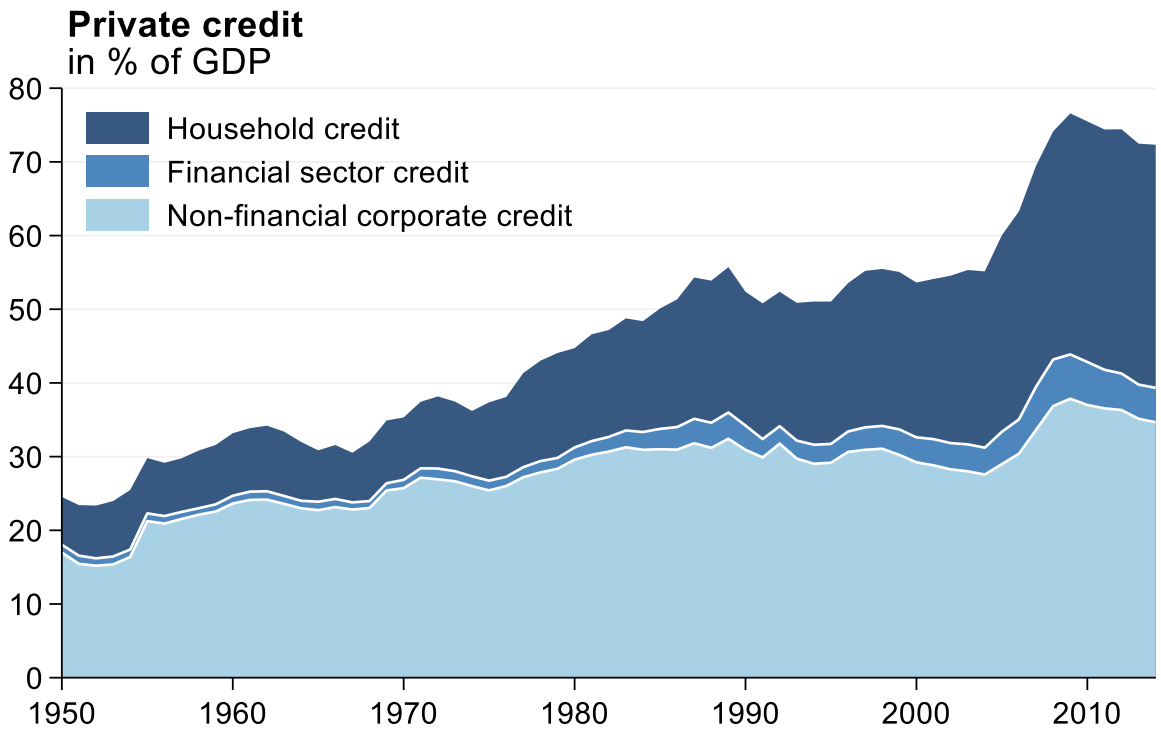
116 countries
1940-2014
Sector classification: ISIC Rev. 4
Covers all domestic credit

Forthcoming

More countries
Update until 2021
Website to explore data
Data and code

New facts about allocation of credit

(a) Booming household, stalling firm credit



Sample: 51 advanced and 46 emerging economies.

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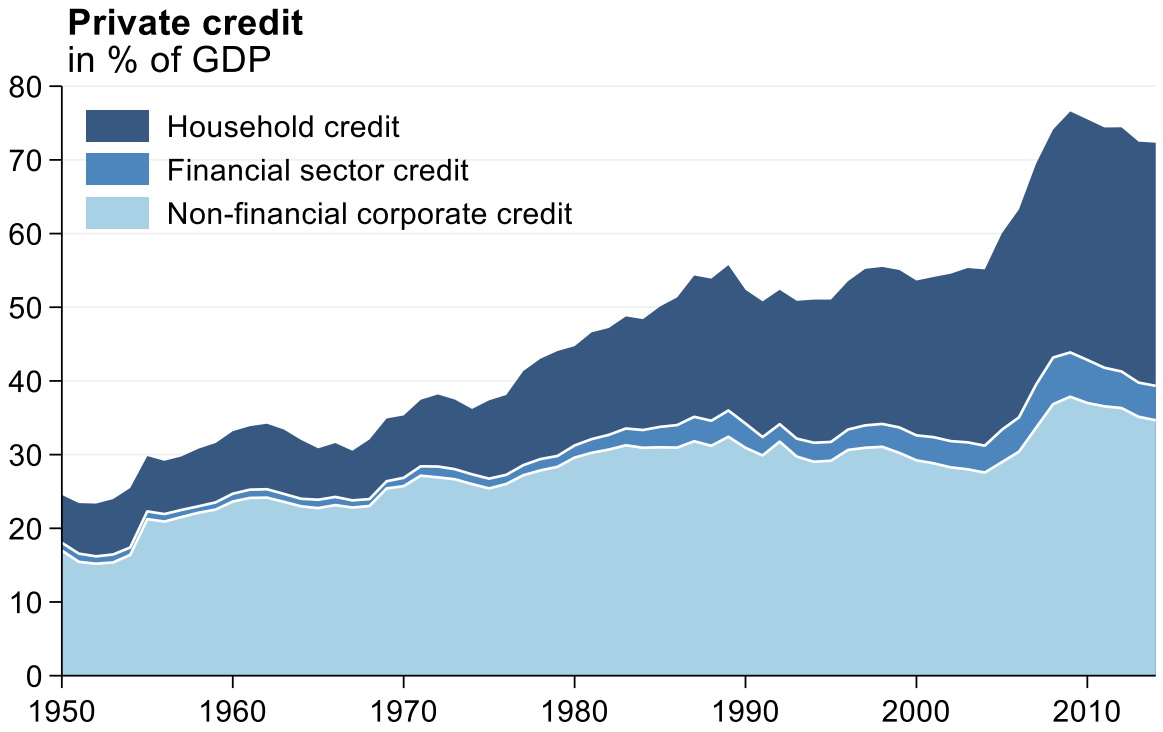

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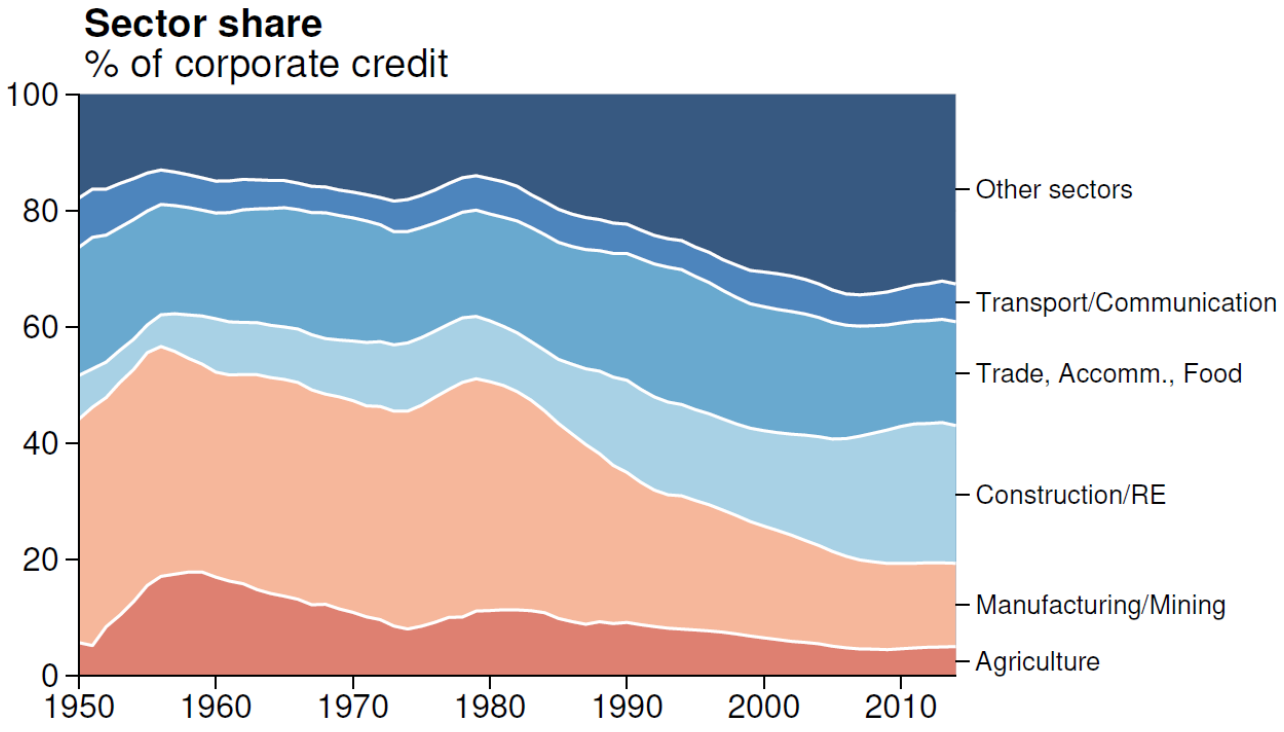
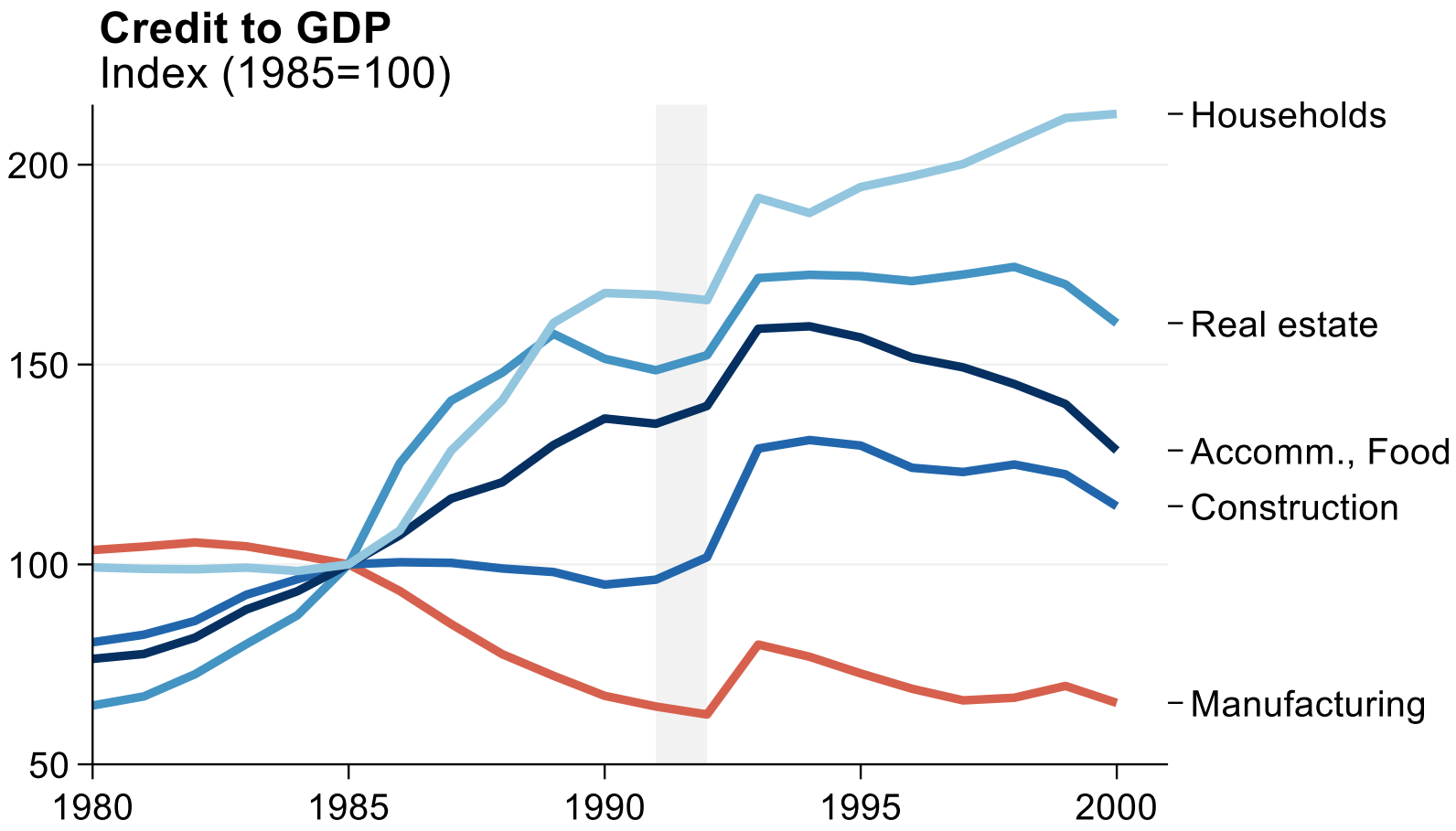
(b) Structural change in corporate credit



Sample: 35 advanced economies.

The 1980s credit boom in Japan



Similar pattern across most credit booms and crises in advanced and emerging economies

Empirical framework

Credit variables

- Tradable sector: agriculture; mining; manufacturing
- Non-tradable sector: construction/real estate; retail and wholesale trade/accom./food; transport/comm.
- Households

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What are key differences between T and NT sectors?

	Tradable	Non-tradable
1) Sensitivity to demand:		
Proximity to final demand	0.15	0.36
Exports/value added	0.78	0.11

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2) Financing constraints:		
Small firm share	0.79	0.90
Mortgage share	0.45	0.61

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Mortgage share	0.45	0.61
3) Productivity:		
Labor productivity	$56,263	$43,406
Labor productivity growth	3.2%	1.0%

Sources: WIOT, Eurostat, various central banks, Mano & Castillo (2015)

Empirical framework

Impulse responses from Jordà (2005) local projections:

$$\Delta_h y_{it+h} = \alpha_i^h + \sum_{j=0}^{J} \boldsymbol{\beta_{h,j}^{NT}} \Delta d_{it-j}^{NT} + \sum_{j=0}^{J} \boldsymbol{\beta_{h,j}^{T}} \Delta d_{it-j}^{T} + \sum_{j=0}^{J} \boldsymbol{\beta_{h,j}^{HH}} \Delta d_{it-j}^{HH}$$

$$+ \sum_{j=0}^{J} \gamma_{h,j} \Delta y_{it-j} + \epsilon_{it+h}\,, \qquad h = 1, \dots, 10 \qquad J = 5$$

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$\boldsymbol{y}$ = Log(real GDP)

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Country fixed effects

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$\boldsymbol{d^{NT}}$ = Credit to the non-tradable sector / GDP

Empirical framework

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$\boldsymbol{d^T}$ = Credit to the tradable sector / GDP

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$\boldsymbol{d^{HH}}$ = Credit to households / GDP

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Prediction horizon: 10 years

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Lag length: 5 years

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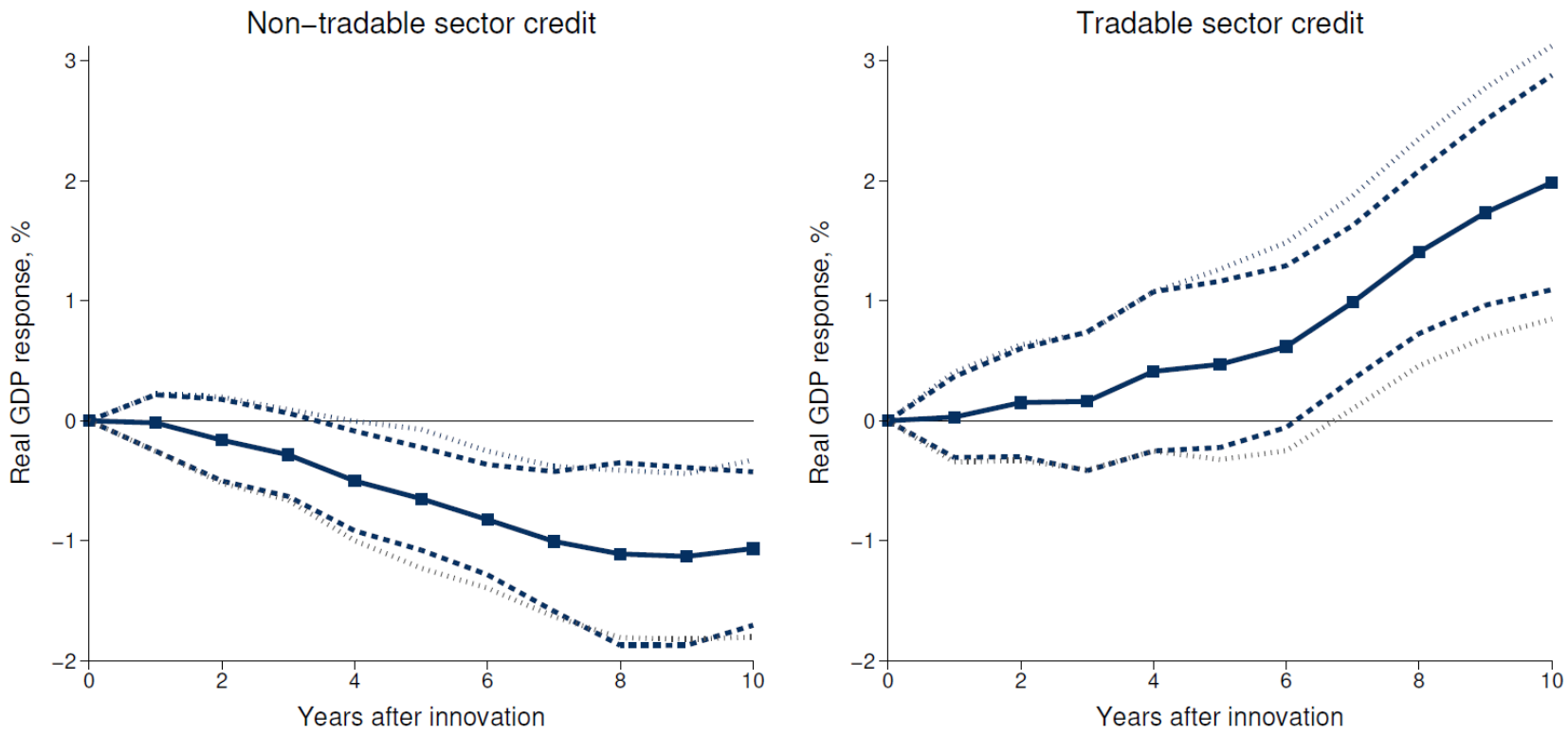
Inference: Driscoll-Kraay or two-way clustered standard errors (country and year)

Note on interpretation: Impulse responses ≠ causal effects

- Conditional on seeing a credit expansion, what happens to GDP (on average)?

Real GDP and T vs. NT sector firm credit expansions

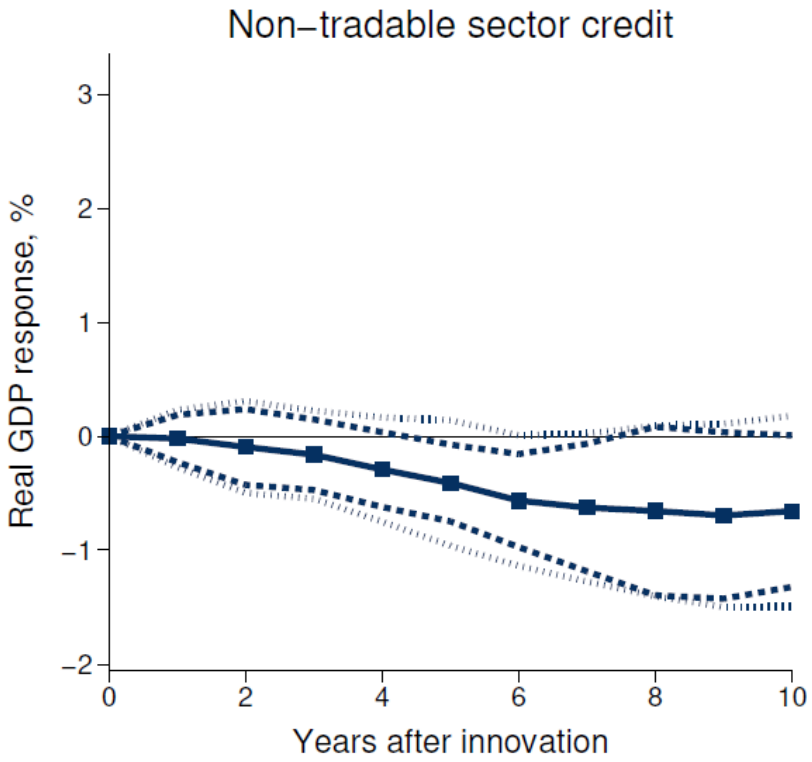
$$\Delta_h y_{it+h} = \alpha_i^h + \sum_{j=0}^{5} \beta_{h,j}^{NT} \Delta d_{it-j}^{NT} + \sum_{j=0}^{5} \beta_{h,j}^{T} \Delta d_{it-j}^{T} + \sum_{j=0}^{5} \gamma_{h,j} \Delta y_{it-j} + \epsilon_{it+h}$$

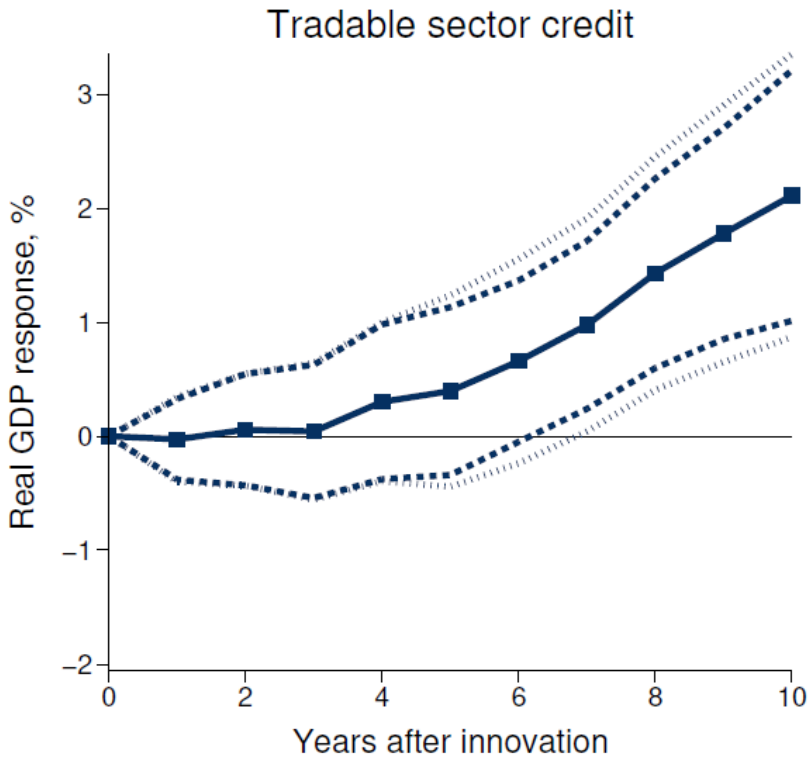


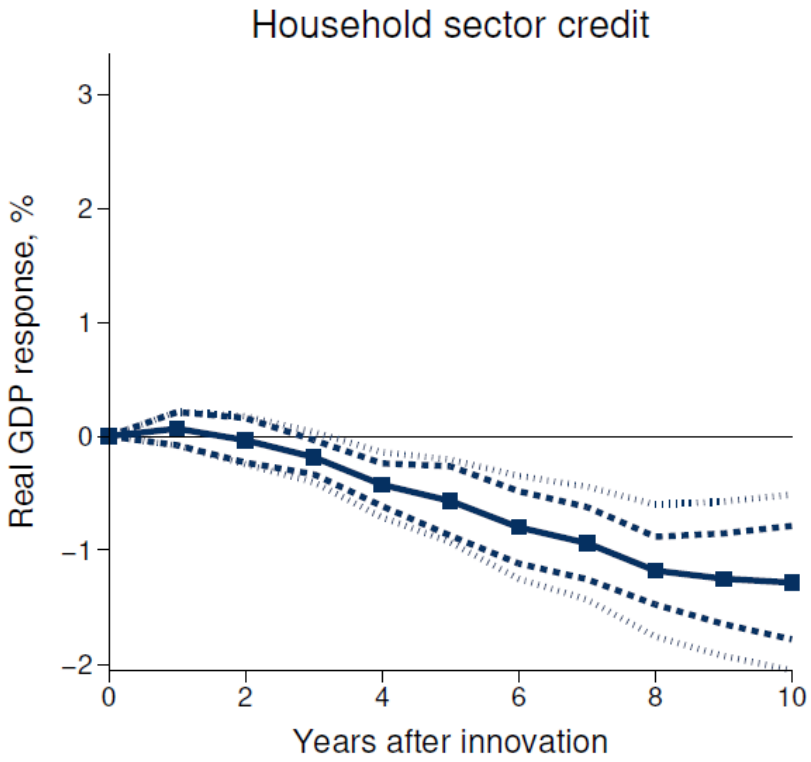
In the paper, we show these patterns are robust and hold when controlling for output shares

Similar when controlling for household debt expansion

$$\Delta_h y_{it+h} = \alpha_i^h + \sum_{j=0}^{5} \boldsymbol{\beta_{h,j}^{NT}} \Delta d_{it-j}^{NT} + \sum_{j=0}^{5} \boldsymbol{\beta_{h,j}^{T}} \Delta d_{it-j}^{T} + \sum_{j=0}^{5} \boldsymbol{\beta_{h,j}^{HH}} \Delta d_{it-j}^{HH} + \sum_{j=0}^{5} \gamma_{h,j} \Delta y_{it-j} + \epsilon_{it+h}$$







Splitting firm credit along sector characteristics

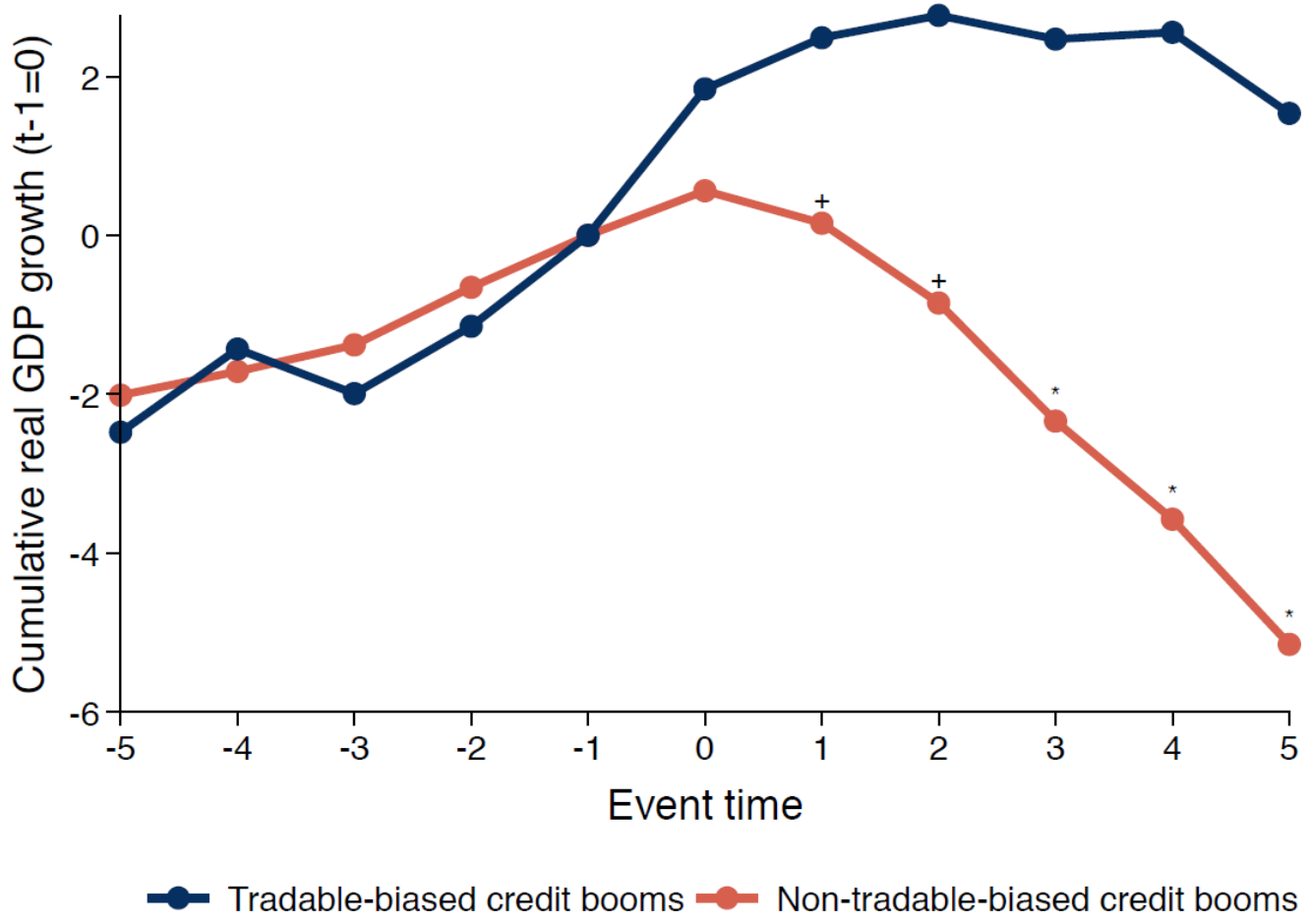
$$\Delta_3 y_{it+h} = \alpha_i^h + \boldsymbol{\beta_h^{HIGH}} \Delta_3 d_{it}^{HIGH} + \boldsymbol{\beta_h^{LOW}} \Delta_3 d_{it}^{LOW} + \epsilon_{it+h}, \qquad h = 0, 1, \ldots, 5$$

	Dependent var.: GDP growth over...					
$\Delta_3 d_{it}^k$	(1) (t-3,t)	(2) (t-2,t+1)	(3) (t-1,t+2)	(4) (t,t+3)	(5) (t+1,t+4)	(6) (t+2,t+5)
Panel A: Sorting by proximity to household demand						
High proximity to HH	0.23*	-0.0097	-0.23*	-0.35**	-0.39**	-0.33**
	(0.100)	(0.11)	(0.10)	(0.083)	(0.075)	(0.077)
Low proximity to HH	0.39**	0.30**	0.20	0.19	0.22	0.26*
	(0.094)	(0.11)	(0.13)	(0.14)	(0.15)	(0.12)
Panel B: Sorting by small firm share						
High small firm share	0.21*	-0.048	-0.27*	-0.40**	-0.43**	-0.38*
	(0.087)	(0.099)	(0.11)	(0.13)	(0.15)	(0.15)
Low small firm share	0.38**	0.29*	0.17	0.16	0.15	0.17
	(0.083)	(0.11)	(0.15)	(0.17)	(0.19)	(0.19)

Similar patterns when splitting along: export/VA, housing input share, or mortgage debt share

Real GDP around major credit booms

1. **Identify credit booms:** based on detrended total credit/GDP
2. **Split by composition of boom:** NT/HH-biased or T-biased boom



Mechanisms

Recap: potential channels linking NT and HH credit to lower medium-run growth

1. **Credit-driven demand boom and bust (e.g. Schmitt-Grohé-Uribe, 2016)**

→ NT/HH credit predict reallocation toward NT sector, real exchange rate appreciation

2. **Differences in financial fragility across sectors (e.g. Schneider-Tornell, 2004)**

→ NT/HH credit predict financial crises, sectoral losses

3. **Lower productivity growth through misallocation across sectors (e.g. Reis, 2013)**

→ NT/HH credit predict sluggish productivity growth

→ T credit predicts higher productivity growth

1. Sectoral credit and demand booms

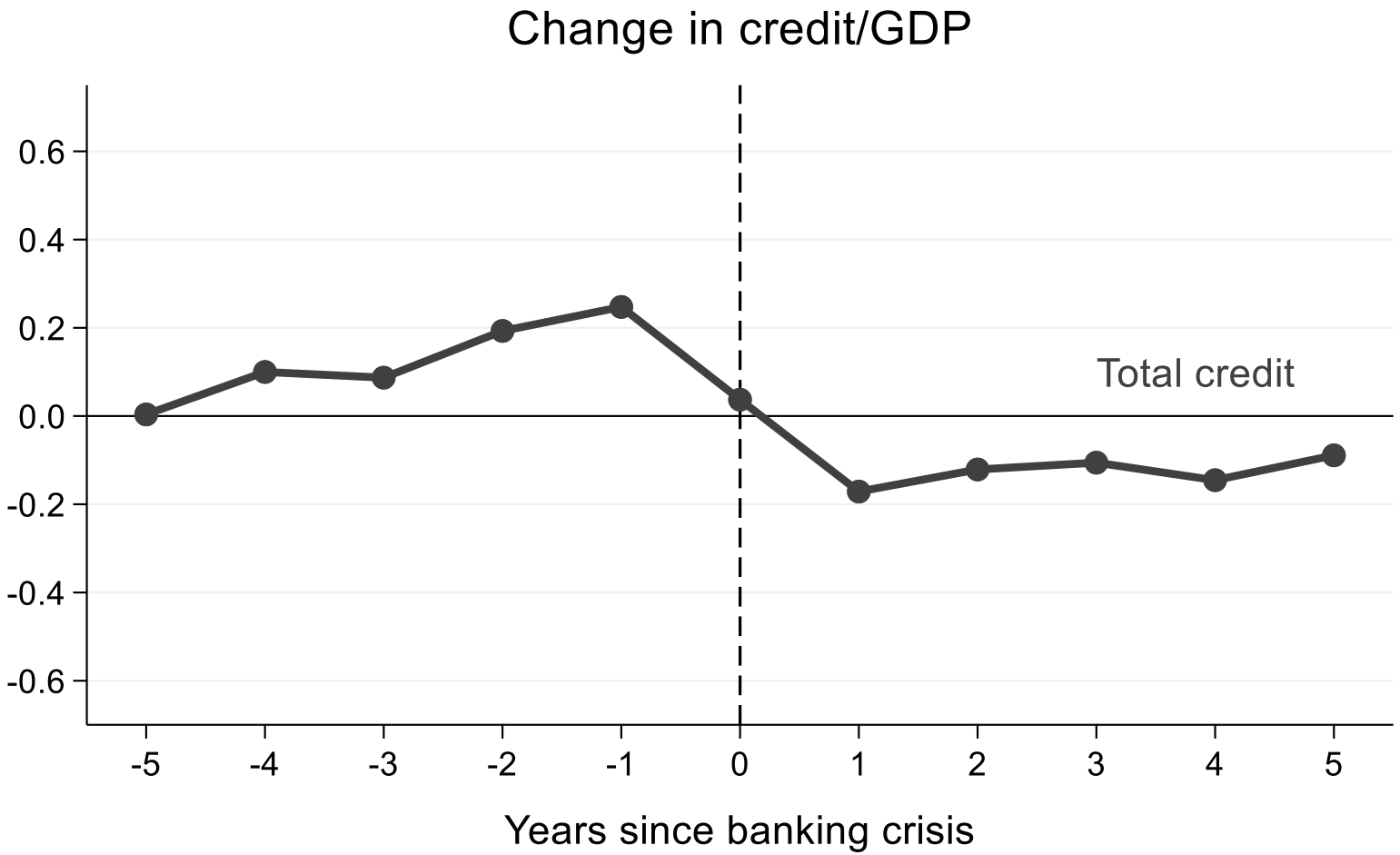
$$\Delta_3 y_{it} = \alpha_i^h + \boldsymbol{\beta_h^{NT}} \Delta_3 d_{it}^{NT} + \boldsymbol{\beta_h^{T}} \Delta_3 d_{it}^{T} + \boldsymbol{\beta_h^{HH}} \Delta_3 d_{it}^{HH} + \epsilon_{it}$$

	$\Delta_3 \ln\left(\frac{E^{NT}}{E^{T}}\right)$	$\Delta_3 \ln(RER)$
$\Delta_3 d_{it}^k$	(1)	(2)
Tradables	-0.18	-0.27
	(0.16)	(0.30)
Non-tradables	0.44**	0.43+
	(0.073)	(0.22)
Households	0.44**	0.30*
	(0.048)	(0.12)
Observations	992	1,755
# Countries	45	73
R^2	0.14	0.03

- NT and HH sector credit associated with reallocation of real activity towards NT, real appreciation
- Consistent with credit expansion boosting demand (Mian-Sufi-Verner, 2020)

2. Differences in financial fragility across sectors

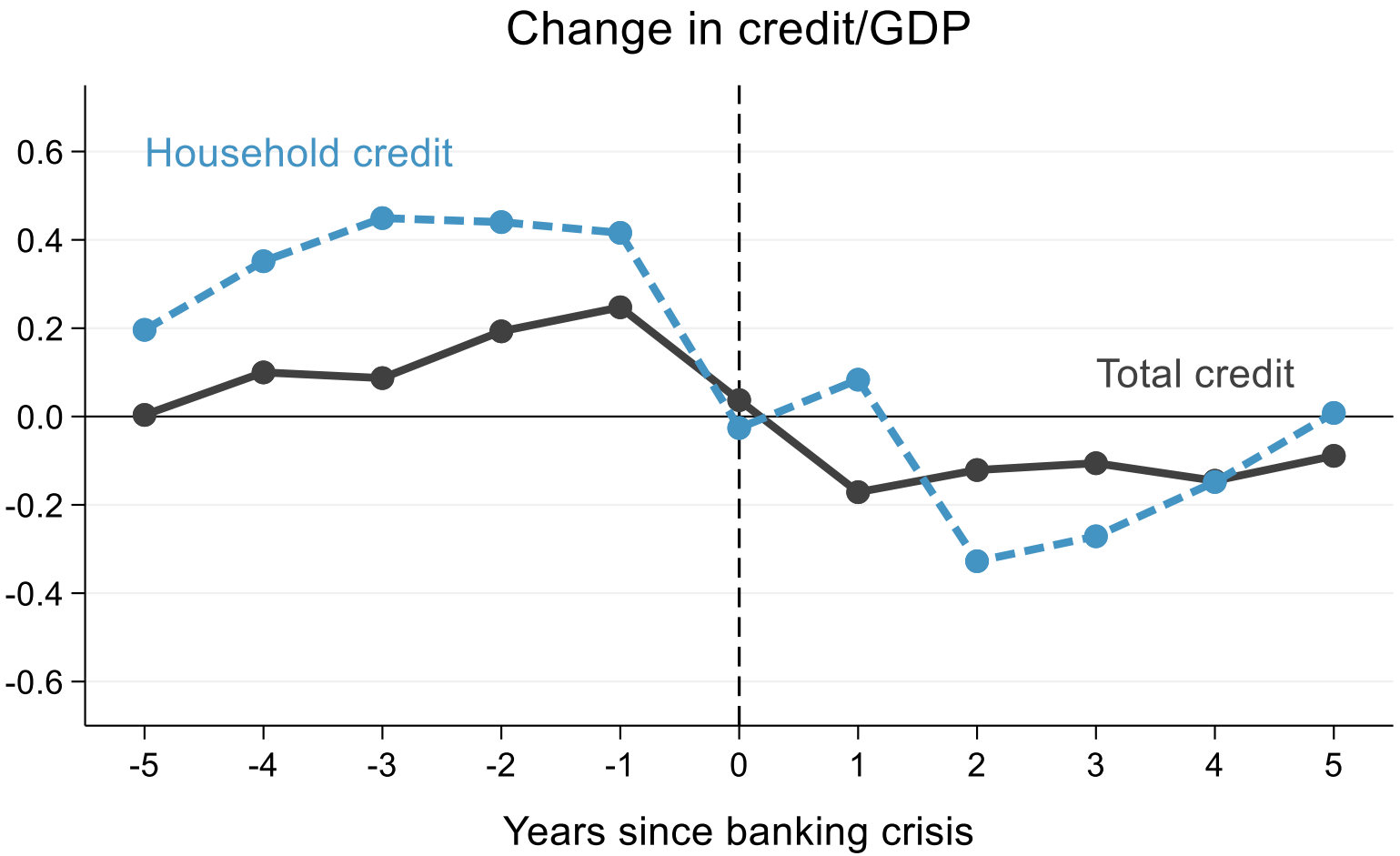
Established finding: total credit/GDP expands before crises



Note: Crisis dates from BVX (2020) and LV (2018).

2. Differences in financial fragility across sectors

Household debt expands earlier than firm debt



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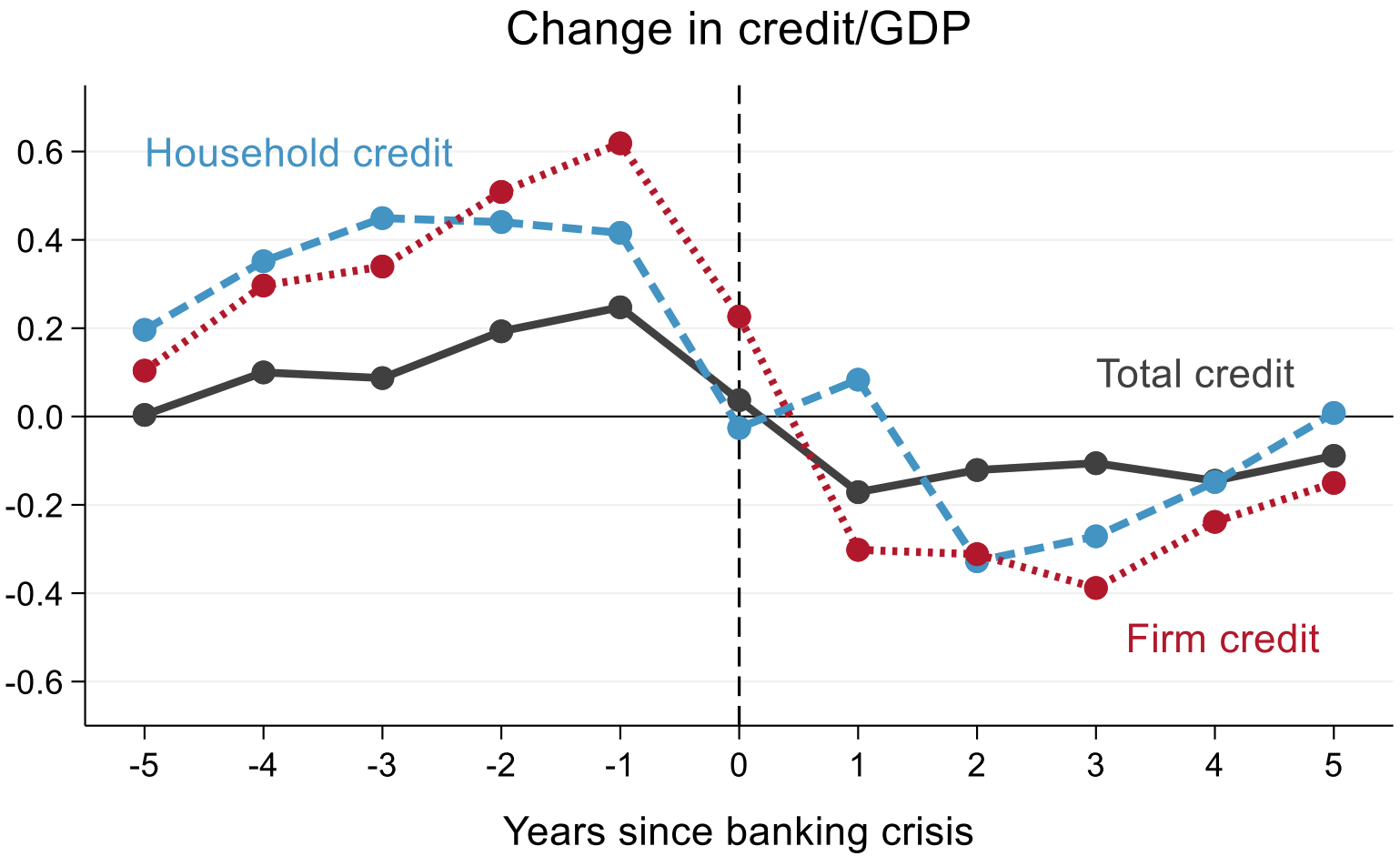

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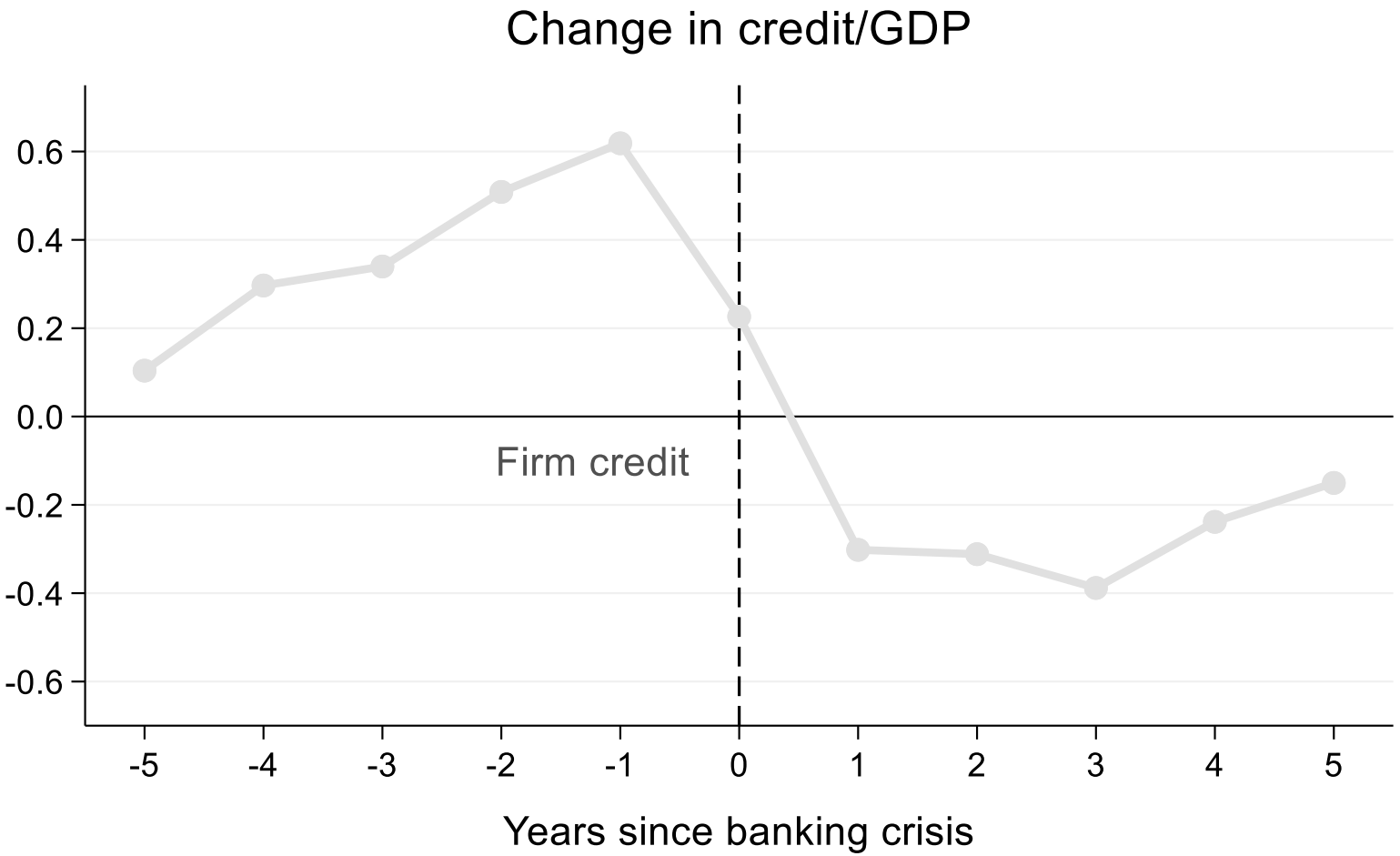
Firm credit expansions mainly driven by NT sector



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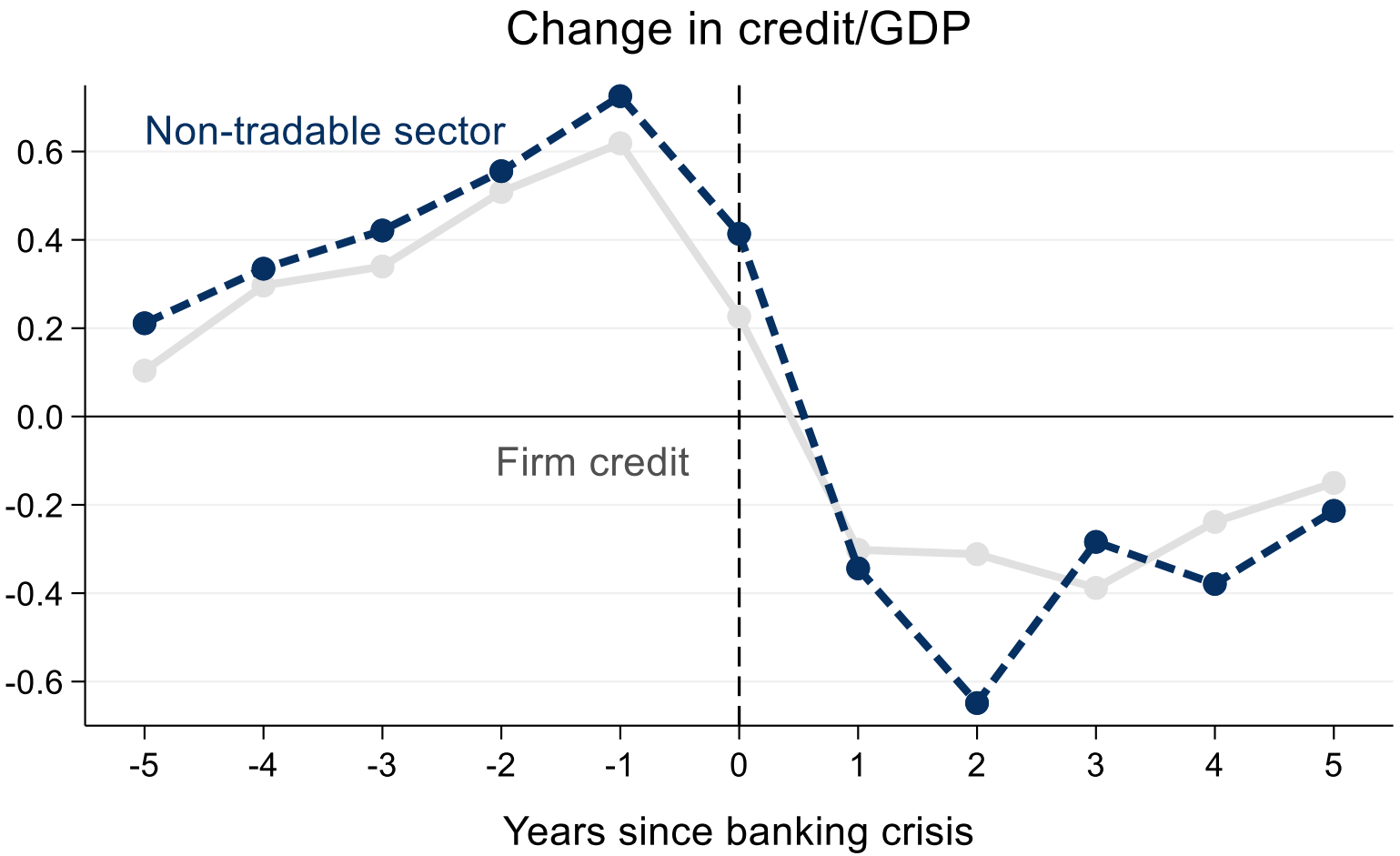

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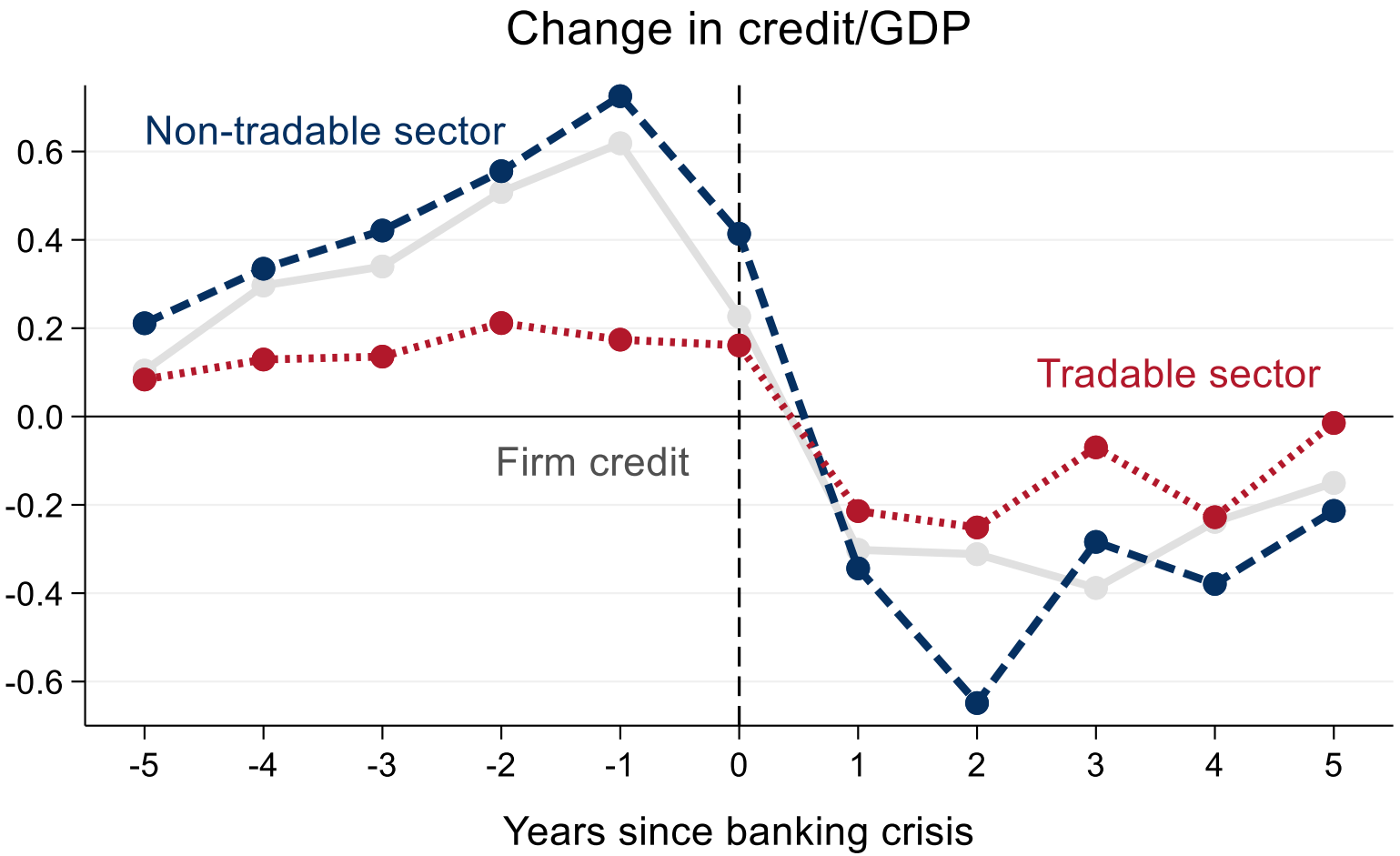

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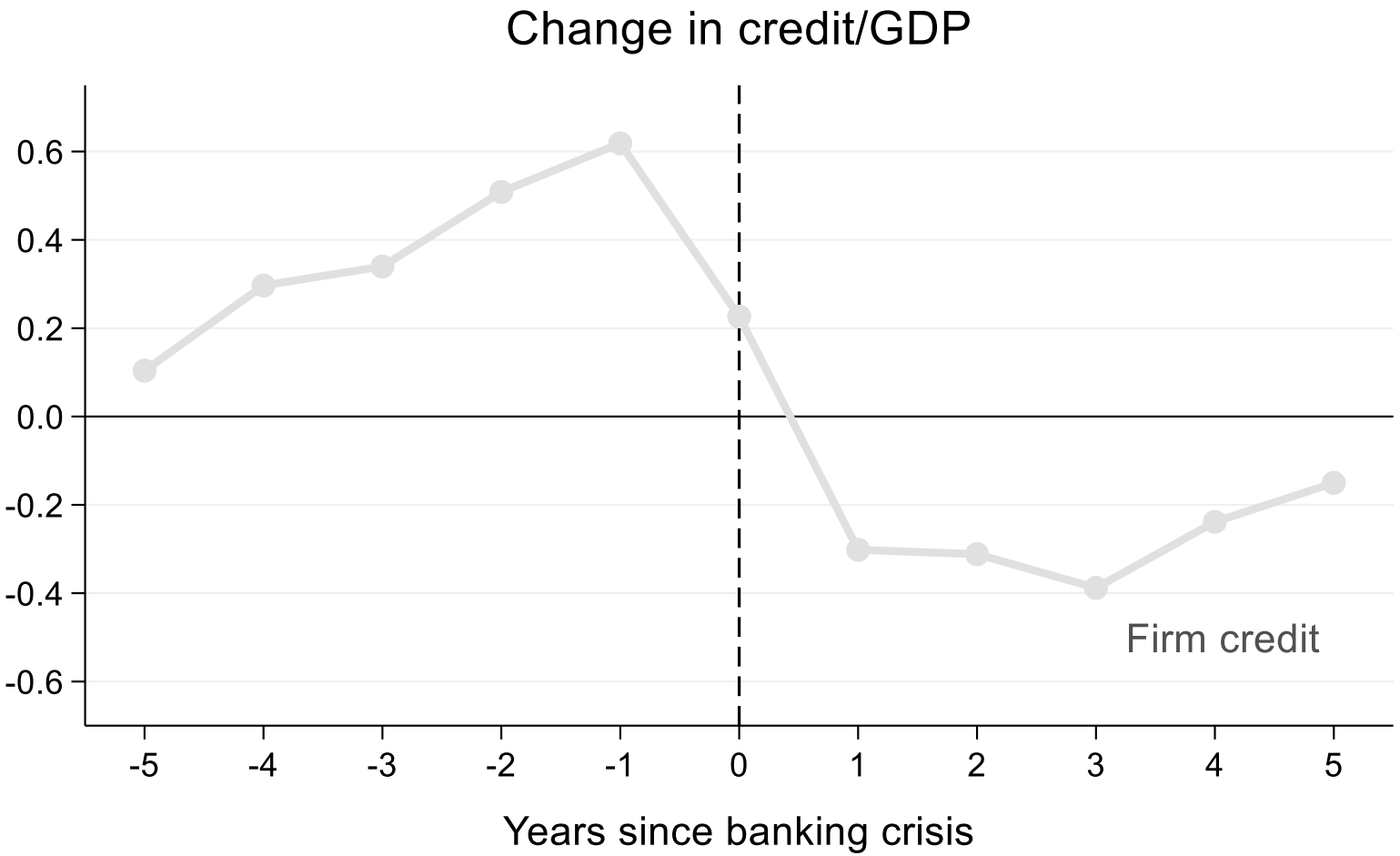
NT sector expansions not only driven by housing



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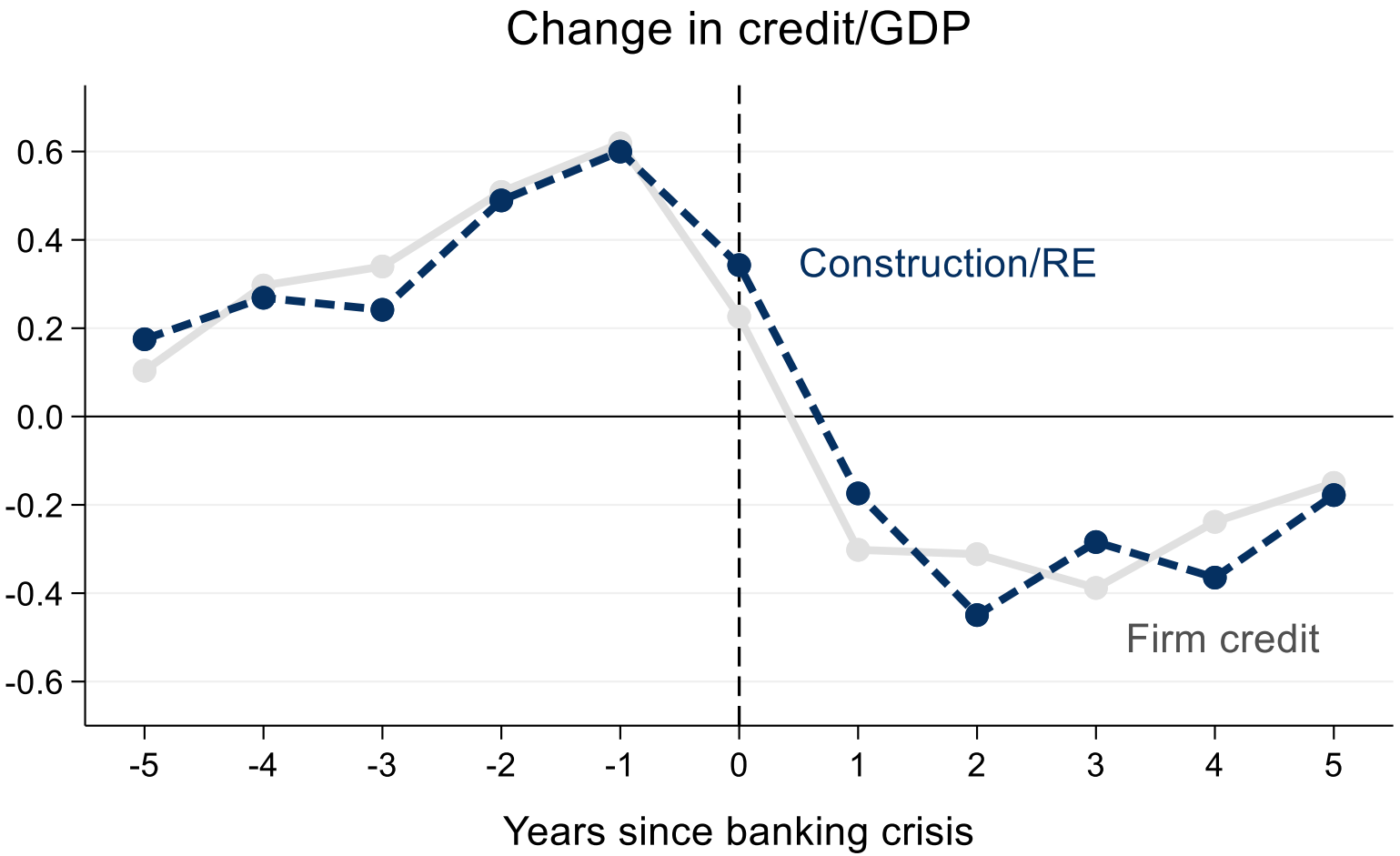
2. Differences in financial fragility across sectors

NT sector expansions not only driven by housing



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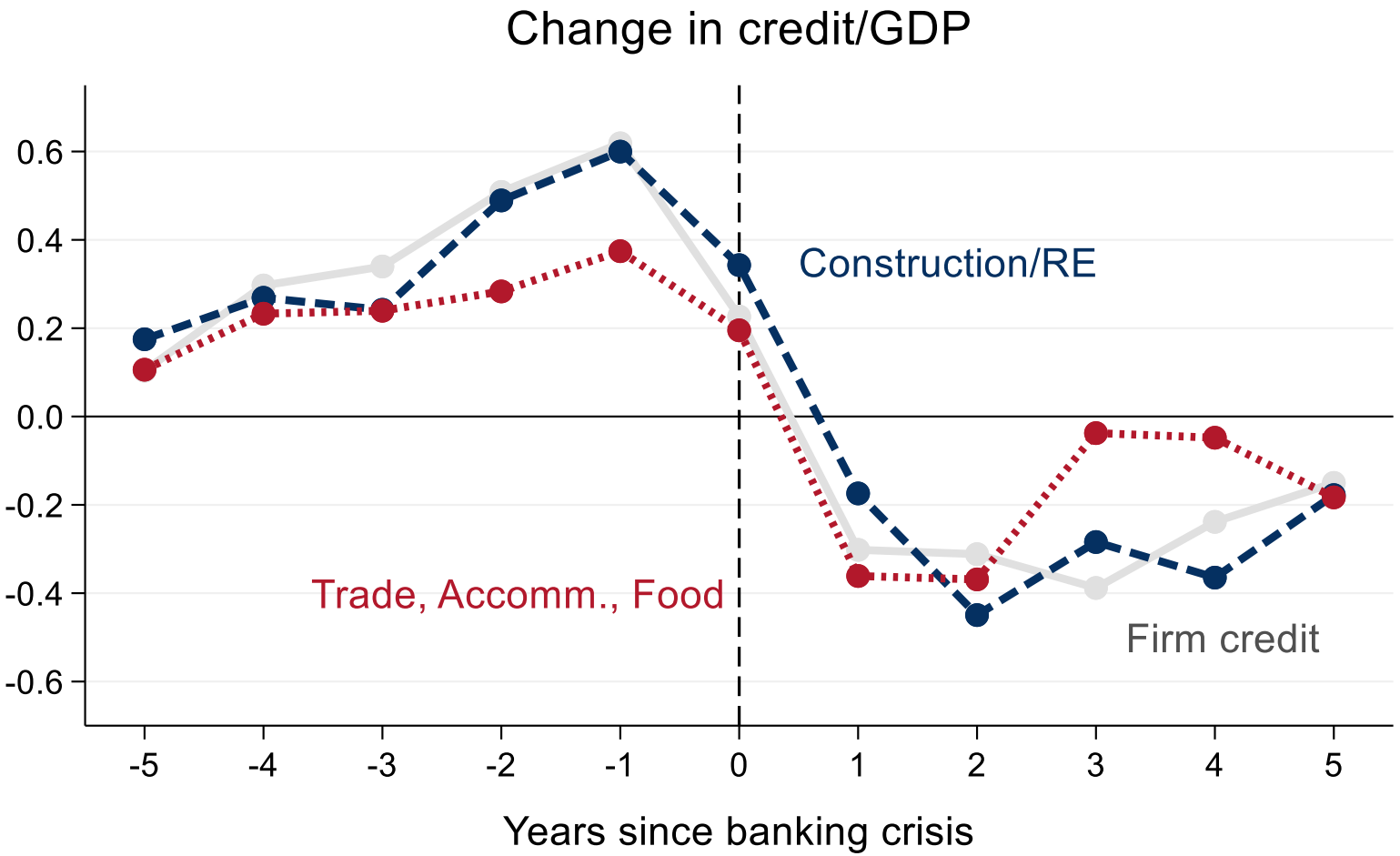

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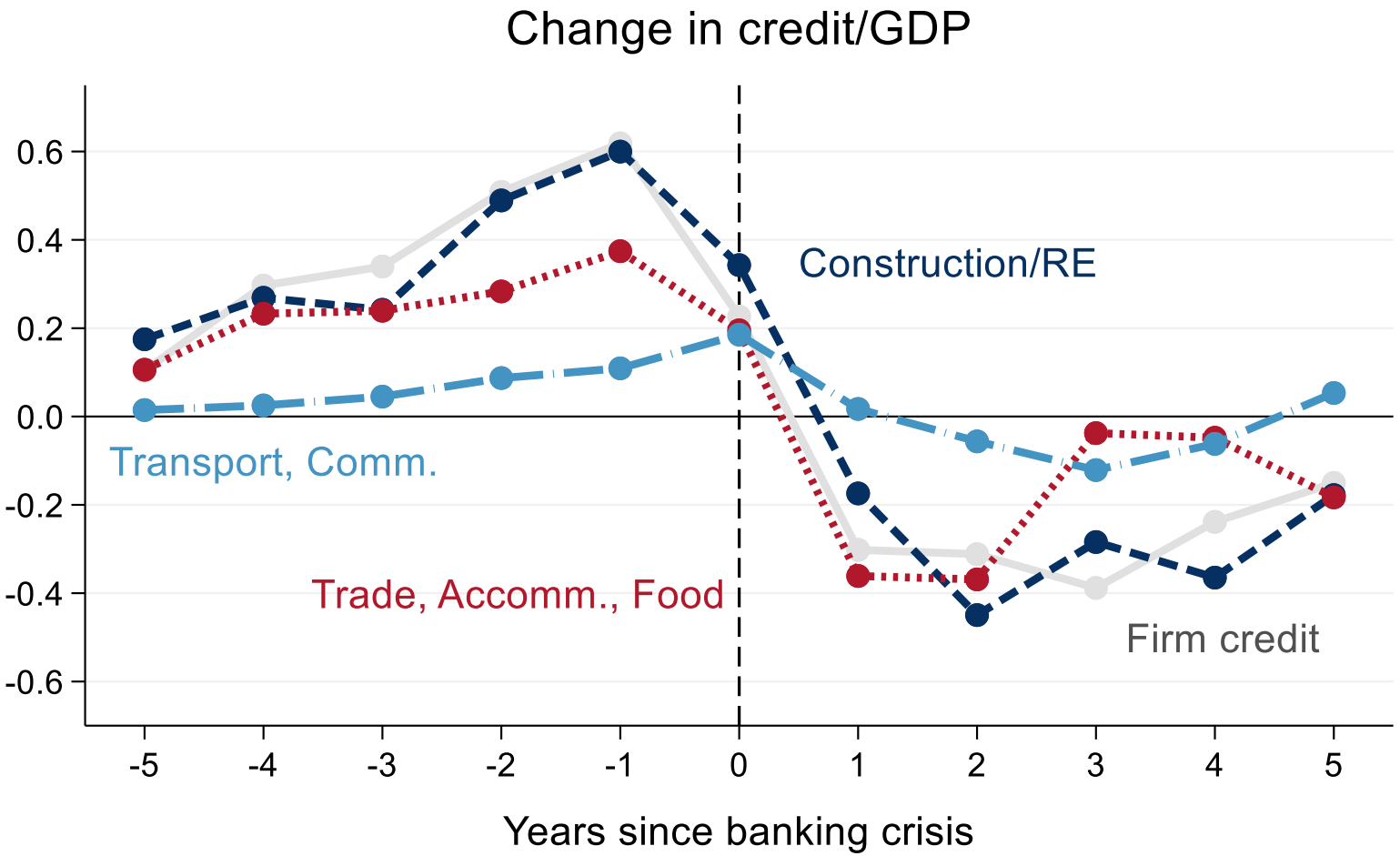

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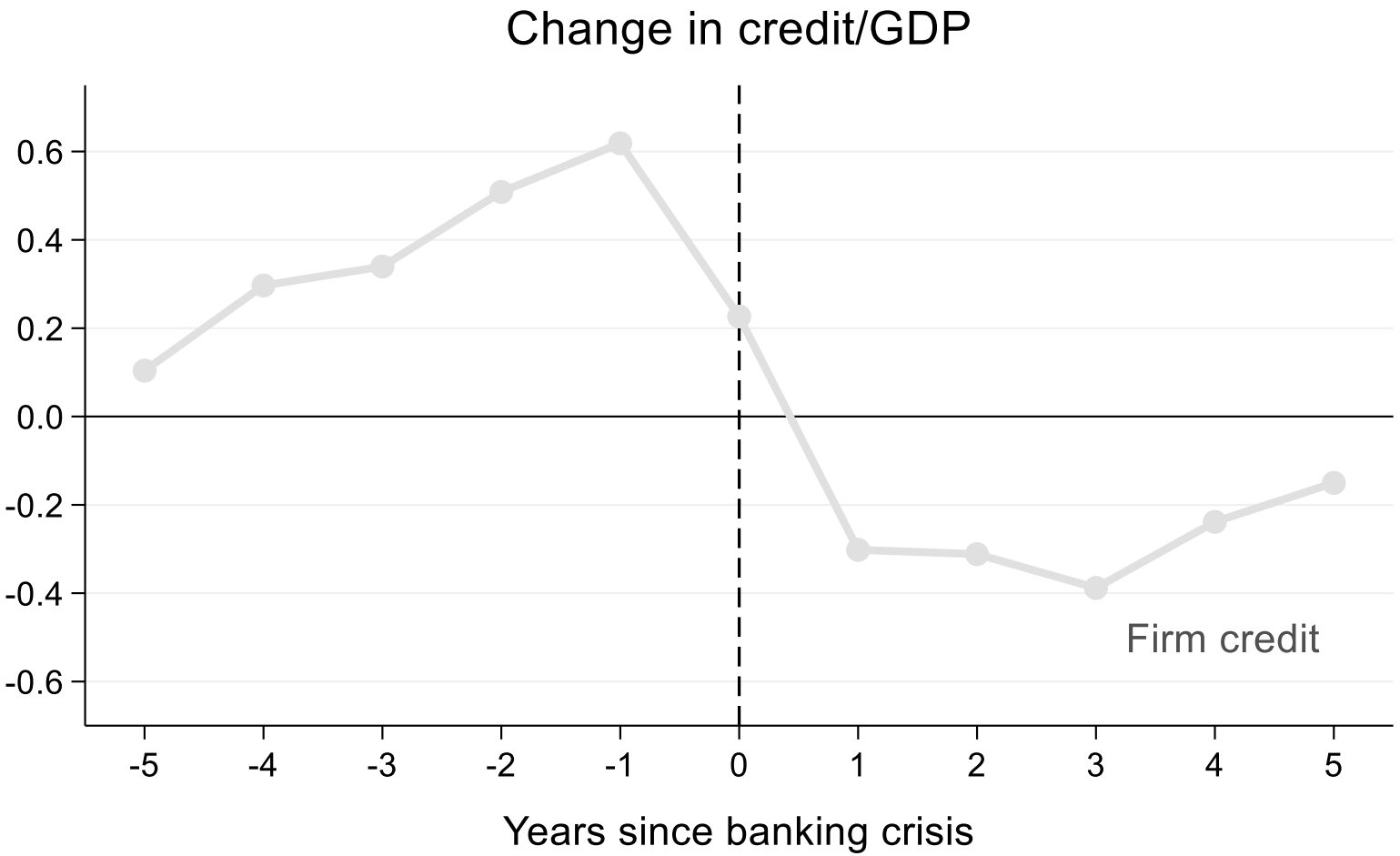
T sector credit growth muted before crises



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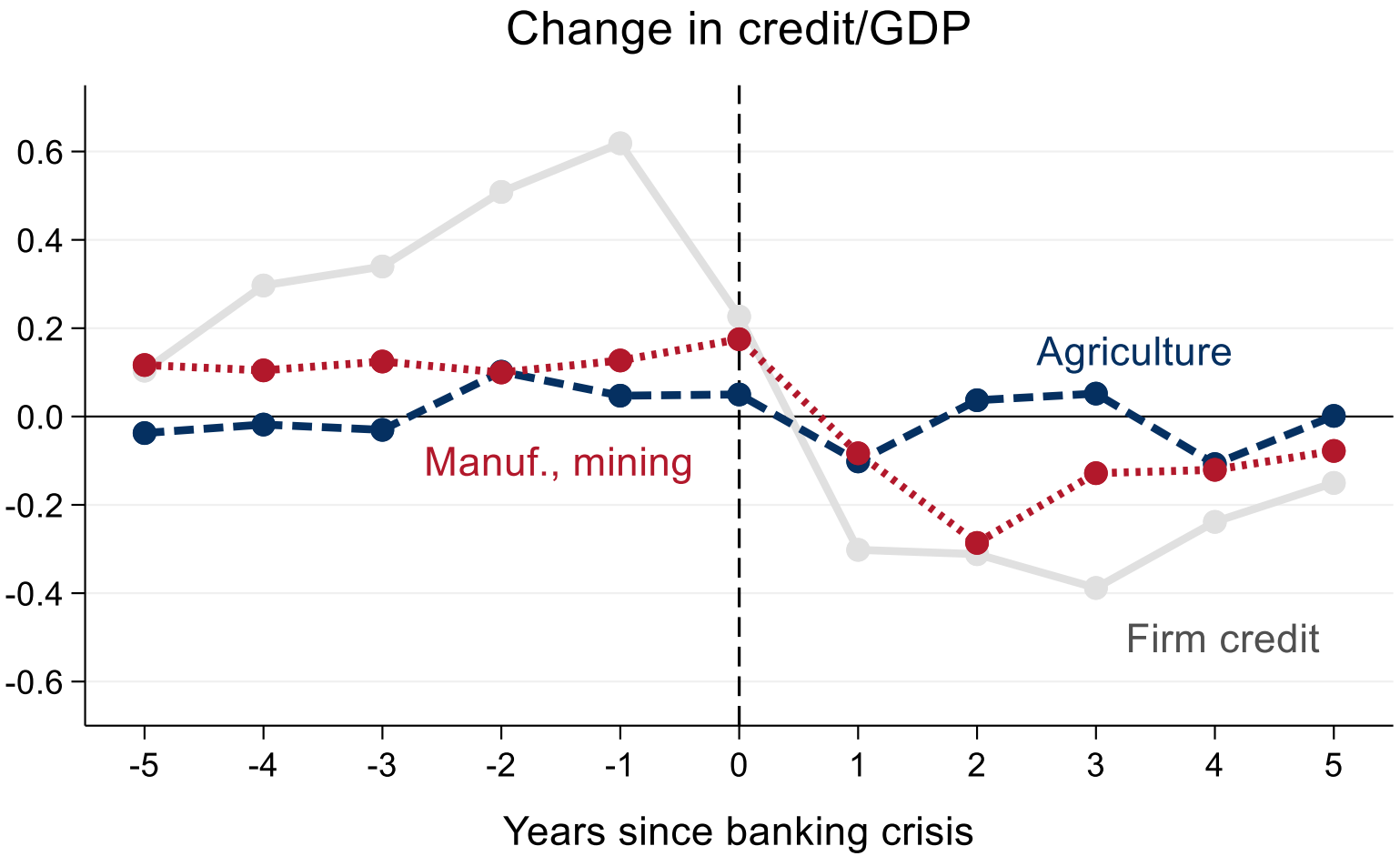

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2. Differences in financial fragility across sectors

$$Crisis_{it\ to\ it+h} = \alpha_i^h + \boldsymbol{\beta_h^{NT}}\Delta_3 d_{it}^{NT} + \boldsymbol{\beta_h^{T}}\Delta_3 d_{it}^{T} + \boldsymbol{\beta_h^{HH}}\Delta_3 d_{it}^{HH} + \epsilon_{it+h}, \qquad h = 1, \ldots, 4$$

	Dependent variable: Crisis within...			
	1 year	2 years	3 years	4 years
Tradables	-0.006	-0.009	-0.008	-0.005
	(0.004)	(0.005)	(0.005)	(0.005)
Non-tradables	0.013**	0.017**	0.017**	0.015**
	(0.003)	(0.002)	(0.003)	(0.004)
Households	0.006*	0.009**	0.011**	0.013**
	(0.003)	(0.003)	(0.003)	(0.003)
Observations	1,527	1,531	1,534	1,536
# Countries	70	70	70	70
# Crises	46	45	45	44
AUC	0.74	0.72	0.70	0.68
SE of AUC	0.03	0.03	0.02	0.02

1 SD higher non-tradable sector credit → crisis probability 0.063 pp higher (baseline: ≈0.03)

2. Differences in financial fragility across sectors

Financial crisis frequency by credit expansion bin

Frequency of financial crisis within 4 years

	Tradables	
Non-tradables and households	Bottom 75%	Top 25%
Bottom 75%	0.07	0.10
Top 25%	0.22	0.23

3. Lower productivity growth

$$\Delta_3 Labor\ Productivity_{it+h} = \alpha_i + \boldsymbol{\beta^{NT}} \Delta_3 d_{it}^{NT} + \boldsymbol{\beta^{T}} \Delta_3 d_{it}^{T} + \boldsymbol{\beta^{HH}} \Delta_3 d_{it}^{HH} + \epsilon_{it}, \qquad h = 0, \ldots, 5$$

	Dependent variable: Labor productivity growth over...					
$\Delta_3 d_{it}^k$	(1) (t-3,t)	(2) (t-2,t+1)	(3) (t-1,t+2)	(4) (t,t+3)	(5) (t+1,t+4)	(6) (t+2,t+5)
Tradables	0.188^{+}	0.177*	0.216*	0.219^{+}	0.183	0.141
	(0.094)	(0.075)	(0.088)	(0.119)	(0.148)	(0.169)
Non-tradables	0.098	-0.049	-0.162^{+}	-0.146^{+}	-0.073	0.002
	(0.141)	(0.127)	(0.090)	(0.075)	(0.057)	(0.059)
Households	-0.137*	-0.158*	-0.191**	-0.229**	-0.291**	-0.302**
	(0.064)	(0.066)	(0.055)	(0.061)	(0.074)	(0.067)
Observations	1,423	1,423	1,423	1,423	1,423	1,423
# Countries	67	67	67	67	67	67
R^2	0.01	0.01	0.02	0.03	0.03	0.03

- 1 SD higher NT credit growth → 0.5% lower productivity growth, similar for estimated TFP growth
- Could reflect misallocation of resources across sectors (e.g. Reis, 2013; Benigno-Fornaro, 2014)

Conclusion

Sectoral allocation of credit matters for understanding macro-financial linkages

- Credit to non-tradable/household sector → lower growth
- Credit to tradable sectors → stable/higher growth
- Channels: (1) credit-driven demand boom and bust; (2) financial fragility; (3) lower productivity

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New perspective on "finance-growth" and "credit booms gone bust" views

- What credit is used for matters for whether booms end badly

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Implications

- Heterogeneity in firm credit matters for understanding credit cycles
- Housing and household debt important but not the entire story; other firm sectors also important

Credit Allocation and Macroeconomic Fluctuations

Karsten Müller
Princeton University

Emil Verner
MIT

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