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FINTECH AND BIG TECH CREDIT: A NEW DATABASE

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*The views expressed here are those of the authors and not necessarily the Bank for International Settlements.

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Roadmap

- Fintech and big tech credit: selected trends
- Research questions and contribution
- Data and empirical methodology
- Results
- Conclusions



Fintech and Big Tech Credit: Selected Trends





Fintech and big tech credit: some definitions

Fintech

Fintech credit

Big tech

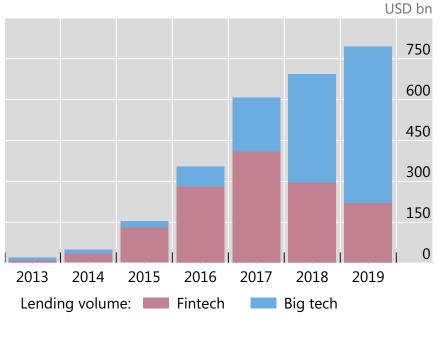
Technology-enabled Definition Credit activity Large companies whose innovation in financial facilitated by electronic primary activity is digital (online) platforms that services that could result services, rather than are not operated by in new business models, financial services (Frost commercial banks, e.g. applications, processes et al., 2019; BIS, 2019) P2P / marketplace or products ... (FSB, lending (Claessens et 2017) <u>al.</u>, 2018) stripe toss Examples Funding Circle ΖΟΡΑ 蚂蚁集团 amazon adyen paytm **自人**金科 YIREN DIGITAL **Lending**Club WeBank Grab 陆金所 SoFi 🗱 微众银行 m-pesa robinhood Lufax.com mintos Lemonade Harmoney mercado libre GOÅJEK Cambridge Centre

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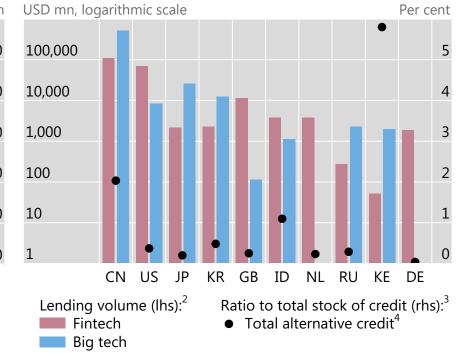
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Big tech credit is booming – reaching USD 572 bn in 2019

Big tech credit is overtaking fintech credit¹



These alternative forms of lending are becoming a significant portion of total credit in a few economies

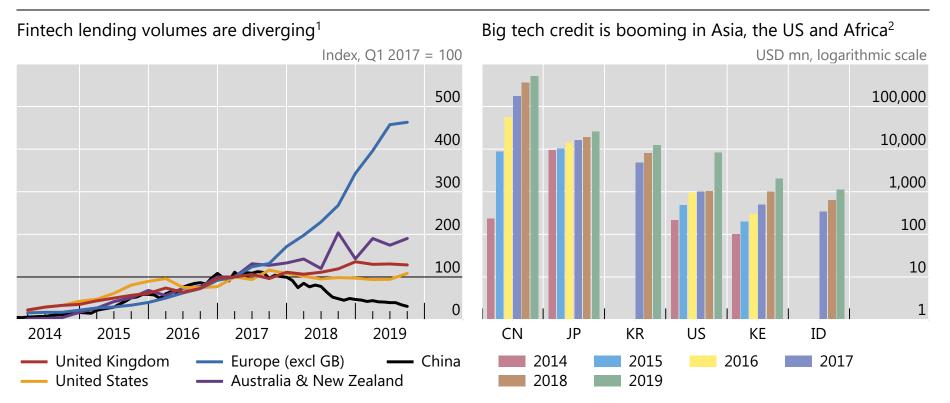


Figures include estimates. CN = China, US = United States, JP = Japan, KR = Korea, GB = Great Britain, ID = Indonesia, NL = Netherlands, RU = Russia, KE = Kenya, DE = Germany.

¹ 2019 fintech lending volume figures are estimated on AU, CN, EU, GB, NZ and US. ² Data for 2019. ³ Domestic credit provided by the financial sector. Data for 2018. ⁴ Total alternative credit is defined as the sum of fintech and big tech credit. Data for 2019.



Fintech credit is growing fast in Europe and many other jurisdictions

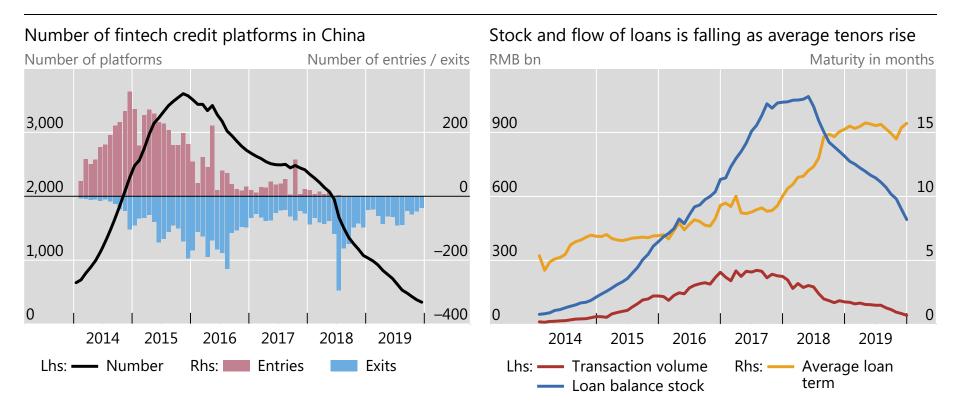


CN = China, JP = Japan, KR = Korea, US = United States, KE = Kenya, ID = Indonesia.

¹ Data are based on five platforms for Australia and New Zealand, all platforms covered by WDZJ.com for China, 49 platforms for Europe, 34 for the United Kingdom and five for the United States. Volumes are reported in local currency. ² Figures include estimates.



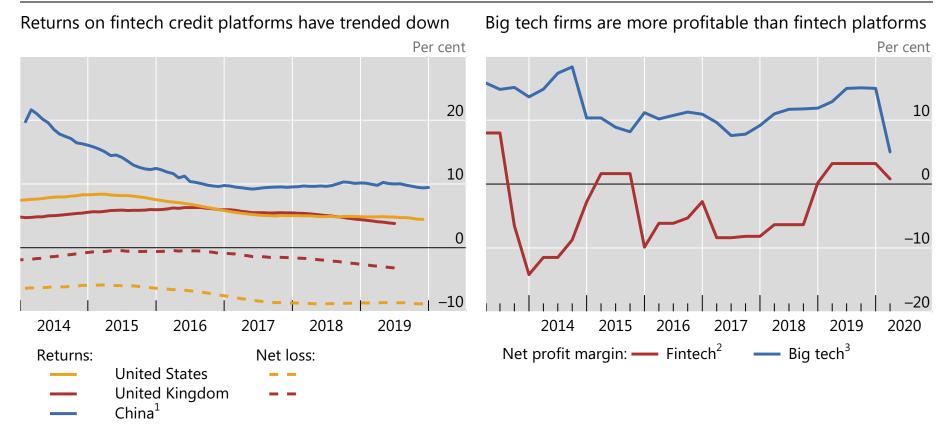
Fintech credit platforms continue to exit in China, as loans stocks and flows fall



Data based on all platforms covered by WDZJ.com for China.



Big tech firms are highly profitable, while fintech platforms have often struggled



¹ Average interest rate. ² Simple average of Black Knight Financial Services, Elevate, Enova International, Fellow Finance, Funding Circle, LendingClub, Lendingtree, Nelnet, OnDeck and Synchrony. ³ Simple average of Alibaba, Amazon, Apple, Baidu / Du Xiaoman, Facebook, Google, JD.com, Kakao, LINE, Microsoft, MTS bank, Orange, Rakuten, Samsung, Tencent, Uber, Vodacom, Vodafone and Yandex.



Research Questions and Contribution

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Potential Drivers of Fintech and Big Tech Credit

On the demand side:

Economic development (Claessens et al., 2018, for fintech credit) Lack of competition (as alluded to in Philippon, 2015) Unmet customer demand (Hau et al., 2018 for China; De Roure et al., 2016 for Germany; Tang, 2018 and Jagtiani and Lemieux, 2018 for the United States)

On the supply side:

<u>Access to data</u> (Jagtiani and Lemieux, 2019; Fuster et al., 2018; Berg et al., 2018)

<u>Technological advantage</u> (van Liebergen, 2017; Gambacorta et al., 2020) <u>Regulatory stance</u> (Buchak et al., 2017 and Barba Navaretti et al., 2017 for fintech)

Institutional characteristics (World Bank, 2020; Rau, 2020)



Research Questions and Contribution

Research questions

How large are fintech and big tech credit markets, in absolute terms and relative to overall credit markets? What economic and institutional factors are driving their growth and adoption?

This paper

Constructs a new database of fintech and big tech credit volumes for 79 countries around the world over 2013-2019

Provides insights into economic and institutional drivers based on panel regressions

Gives empirical grounding for discussion of future growth and policy implications



Data and Empirical Methodology

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Database Construction

Fintech Credit

Cambridge Centre for Alternative Finance (CCAF) Global Alternative Finance database (Wardrop et al., 2015; Ziegler et al., 2020) Annual online questionnaire of alternative finance volumes and characteristics Available for 2013–18; 2019 volumes estimated based on other sources (Brismo, WDZJ, Korean P2P Lending Association, central banks' input)

Big tech credit

Hand-collected from public sources, company contacts and central banks In some cases, conversion from lending stocks to flows (and vice versa) When unavailable, 2019 numbers extrapolated based on user numbers or revenues



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Empirical Model

Panel regressions over 2013–18 for fintech credit, big tech credit and total alternative credit per capita:

 $\begin{aligned} &\ln(Credit_{it}) \\ &= \alpha + \beta_1 y_{i,t-1} + \beta_2 y_{i,t-1}^2 + \gamma L I_{i,t-1} + \delta R S_{i,t-1} + \mu B N_{i,t-1} + \sigma X_{i,t-1} + \vartheta D_k + \varepsilon_i \end{aligned}$

 $y_{i,t-1}$: GDP per capita in economy *i* at year *t-1*

 $LI_{i,t-1}$: Lerner index of banking sector mark-ups

 $RS_{i,t-1}$: regulatory stringency for the banking sector

 $BN_{i,t}$: the density of the bank branch network compared to the adult population

 $X_{i,t}$: control variables, e.g. growth in GDP and total credit, real interest rate, mobile phone subscriptions / adult population, dummy for a financial crisis, dummy for AE

 D_k : geographical area fixed effects

 $\varepsilon_{i,t}$: error term



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Descriptive Statistics – Part 1

Descriptive statistics

Table 1

Variable	Observations	Mean	Standard deviation	Min	Max
GDP per capita (in thousands of USD)	453	21.53	18.21	0.67	87.76
Lerner index ¹	453	0.30	0.15	-0.05	1.00
Bank branches per 100,000 adults	453	17.65	14.03	1.43	83.75
Normalised index of bank regulatory stringency ²	453	0.72	0.10	0.38	0.96
Score-Starting a business (overall)	425	82.55	11.46	23.04	99.96
Score-Time (days)	425	81.96	17.33	0.00	100.00
Score-Paid-in Minimum capital (% of income per capita)	425	94.85	15.45	0.00	100.00
Score-Cost (% of income per capita)	425	84.92	26.95	0.00	100.00
Extent of disclosure index (0-10)	425	64.44	23.80	0.00	100.00
Trial and judgment (days)	425	407.85	203.51	90.00	1095.00
Enforcement of judgment (days)	425	177.98	110.59	26.00	597.00
Enforcement fees (% of claim)	425	5.37	5.23	0.00	23.30
Bank credit to bank deposits (%)	212	105.39	80.48	27.73	702.09
Bank regulatory capital to risk-weighted assets (%)	197	17.20	3.81	10.59	35.65
Provisions to non-performing loans (%)	187	64.29	37.26	0.00	232.06





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Descriptive Statistics – Part 2

Descriptive statistics						
Variable	Observations	Mean	Mean Standard deviation		Max	
Loans from non-resident banks to GDP (%)	194	27.53	28.33	1.24	158.53	
Corporate bond average maturity (years)	135	10.26	5.75	3.54	34.09	
Corporate bond issuance volume to GDP (%)	137	2.14	1.81	0.05	13.83	
Total factoring volume to GDP (%)	145	5.16	4.90	0.07	16.29	
Global leasing volume to GDP (%)	78	1.32	0.95	0.01	4.81	
Stock market total value traded to GDP (%)	167	50.64	89.08	0.00	562.92	
Stock market turnover ratio (%)	161	53.61	67.37	0.84	556.91	
Ln(Total alternative credit per capita (in USD) ³)	453	0.93	1.43	-1.97	5.11	
Ln(Big tech credit per capita (in USD))	453	0.09	0.99	-3.57	4.55	
Ln(Fintech credit per capita (in USD))	453	-1.14	2.76	-7.20	4.81	

Ln = natural logarithm. The dependent variables have been winsorised at the 1% and 99% level.

¹ The Lerner index of banking sector mark-ups in economy *i* reflects market power by incumbent banks. World Bank data. For 2015-2017 data are estimated based on Igan et el (2020). ² The index is normalised between 0 (no regulation) and 1 (max regulation). The index is calculated from a survey conducted by the World Bank in given years, and therefore data are not available over the whole sample period, but proceed in steps. See https://datacatalog.worldbank.org/. ³ Defined as the sum of big tech and fintech credit.

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Sources: IMF, *World Economic Outlook*, World Bank; Cambridge Centre for Alternative Finance and research partners; company statements; authors' estimates.

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Results





Drivers of Fintech and Big Tech Credit Volumes

	Ln(total alternative credit per capita)	Ln(big tech credit per capita)	Ln(fintech credit per capita)	Difference b-a
		(a)	(b)	H ₀ : b-a <0
GDP per capita	0.123***	0.069***	0.171***	0.102***
	(0.022)	(0.020)	(0.038)	(0.043)
GDP per capita^2	-0.002***	-0.001***	-0.002***	0.001
	(0.000)	(0.000)	(0.001)	(0.001)
Lerner index	1.438***	0.867**	2.436***	1.569**
	(0.401)	(0.365)	(0.732)	(0.818)
Bank branches per 100,000	-0.017***	0.005	-0.028***	0.033***
adult population	(0.005)	(0.005)	(0.009)	(0.010)
Normalised regulation index	-4.665***	-1.735***	-8.427***	6.692***
	(0.560)	(0.544)	(1.068)	(1.199)
Other controls	Yes	Yes	Yes	
Geographic area fixed effects	Yes	Yes	Yes	
No. of observations	453	453	453	
Estimation method	OLS	OLS	OLS	
R ²	0.469	0.112	0.516	

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Estimation period 2013-2018. Robust standard errors in parentheses. ***/**/* denotes results significant at the 1/5/10% level. Ln = natural logarithm. The dependent variables have been winsorised at the 1% and 99% level.

Robustness Checks

- **Simple cross section:** 79 observations, but important to mitigate endogeneity concerns
- Inclusion of a complete set of time dummies: fixed effects tend to capture the common global trends in the evolution of these forms of credit, but overall results similar
- Log of the stock of bank credit: relationship between bank credit per capita and GDP per capita remains positive up to a level of GDP per capita of around USD 48,700. Negative correlation with the Lerner index and a more stringent bank regulation.
- Impact of explicit fintech regulation: dummy variable that takes a value of 1 if an explicit regulation of fintech credit was in place in a given country and year, and 0 elsewhere. Positive effects but endogeneity prevents to infer causality. All results remain the same.



Results for Country-specific Institutional Characteristics

- Ease of doing business: overall score (**), days to start a business (**), minimum paid-in capital (*), cost to start a business (*)
- Investor protection and judicial system: extent of disclosure (*), duration of trials and judgments (-***), duration of enforcement of judgments (-***), enforcement fees (-*)
- **Banking characteristics:** loan-to-deposit ratio (-**), regulatory capital to riskweighted assets (***), provisions to non-performing loans (***), loans from non-resident banks (-**), firms with a transaction account (***)
- **Financial market development:** corporate bond average maturity (**), corporate bond issuance volume to GDP (**), global leasing volume to GDP (***), stock market total value traded to GDP (***), stock market turnover ratio (**)

***/**/* denotes results significant at the 1/5/10% level.



Conclusions





Main Takeaways

- Fintech and big tech credit have grown fast to an estimated USD 223 and 572 bn in 2019
- Both are higher: (i) with higher GDP per capita, but at a declining rate; (ii) where banking mark-ups are higher and (iii) where banking regulation is less stringent
- Fintech credit is more prevalent where there are fewer bank branches per capita
- Total alternative credit is higher where ease of doing business is higher, with greater investor protection and judicial system quality, where bank capital, provisions and deposit funding are higher and where financial markets are more developed
- Overall, alternative forms of credit seems to complement more traditional credit markets, not to substitute for them

