# FinTech Adoption and Household Risk-Taking NYU China Initiative

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Based on joint work with Claire Yurong Hong (SAIF) and Xiaomeng Lu (FISF)

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  - ▶ **Super Apps:** Financial services delivered directly via mobile apps.
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- In China, activities central to household finance taking place on FinTech platforms:
  - ▶ **Consumption:** online consumption accounts for 25% of the total.
  - ▶ Investments: 30% of mutual fund purchases takes place on FinTech platforms.
  - Payments: digital payments everywhere.



## Alipay as a One-Stop FinTech App

#### Imagine if

- Main-street banks
- Wall Street's brokers
- Boston's asset managers
- Connecticut's insurers

#### all shrunk to fit into

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#### Motivations and Research Questions

"The study of household finance is challenging because household behavior is difficult to measure, and households face constraints not captured by textbook models."

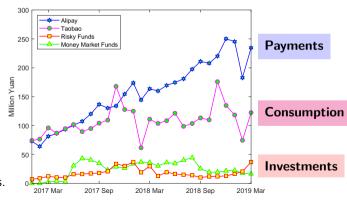
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- A random sample of 50,000 individuals from Ant Group.
- Consumption:
  - Basic
  - Development
  - Enjoy
- Investments:
  - Risky funds: Bond, Mixed, Equity, Index, QDII, Gold.
  - Riskfree: Money market funds.



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#### Can FinTech lower investment barrier and improve household risk-taking?

- ▶ Physical costs: convenience, transaction costs, and access to information.
- Psychological costs: familiarity, trust, and financial literacy.

#### • Who benefits more from FinTech Inclusion?

- ▶ The otherwise more constrained investors prior to the arrival of FinTech.
- ▶ Individuals who are more risk-tolerant.
- Individuals living in areas under-served by financial institutions.

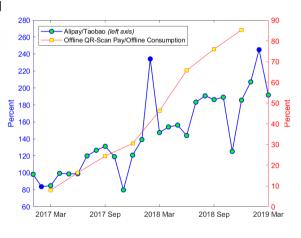
#### Offline Digital Payments via QR-Code Scan

#### 买菜也能扫码支付了 绍兴首家智慧农贸市场下月使用

2017-12-22 17:16 | 绍兴晚报



"You can use QR-Code Scan payment at local farmer's markets in Shaoxing."



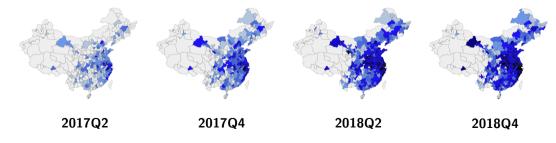
#### Individual-Level Measures of FinTech Adoption

ullet Individual i's consumption on Alipay and Taobao during month t:

$$\mathsf{AliFrac}_t^i = rac{\mathsf{Alipay}_t^i}{\mathsf{Alipay}_t^i + \mathsf{Taobao}_t^i}$$

(0.60, 0.65) (0.55, 0.60) (0.55, 0.55) (0.45, 0.50) (0.40, 0.45) (0.35, 0.40) (0.30, 0.35) (0.00, 0.30)

Aggregated to the city level using individuals' location information:



#### Determinants of FinTech Adoption

		Alil	- rac		$\Delta$ AliFrac				
	All U	Jsers	Active users		All Users		Active	users	
$\sigma_C$	0.032***	0.034***	0.033***	0.035***		-0.010***	-0.010***	-0.008***	-0.009***
	(11.13)	(12.31)	(9.47)	(10.52)		(-3.87)	(-3.77)	(-2.43)	(-2.55)
Log(C)	-0.104***	-0.107***	-0.107***	-0.109***		-0.026***	-0.026***	-0.023***	-0.022***
	(-81.59)	(-94.95)	(-72.19)	(-80.70)		(-18.29)	(-18.19)	(-12.74)	(-12.57)
Female	-0.054***	-0.050***	-0.055***	-0.051***		-0.016***	-0 016***	-0.017***	-0.017***
	(-16.95)	(-16.18)	(-16.00)	(-15.21)		(-6.79)	(-6.86)	(-5-41)	(-5.61)
Log(Age)	0.000	-0.002	-0.015	-0.017*		0.102***	0.101***	0.103***	0.102***
	(0.03)	(-0.27)	(-1.61)	(-1.86)		(16.32)	(16.26)	(14.90)	(14.78)
Log(GDP)	0.023**		0.022**			-0.004		-0.009**	
	(2.50)		(2.18)			(-1.61)		(-2.54)	
Log(Income)	0.029***		0.029***			-0.005***		-0.005**	
	(4.32)		(4.45)			(-3.24)		(-2.55)	
Log(Population)	0.006		0.005			0.002		0.001	
	(0.90)		(0.71)			(1.10)		(0.61)	
Log(#Branch)	-0.003		-0.004			-0.006**		0.002	
	(-0.35)		(-0.34)			(-2.10)		(0.45)	
Citylevel=1	-0.059**		-0.059**			0.005		0.003	
	(-2.50)		(-2.65)			(1.31)		(0.68)	
City FE	N	Υ	N	Υ		N	Υ	N	Υ
Adjusted R <sup>2</sup>	0.210	0.208	0.230	0.230		0.021	0.021	0.019	0.019
N	49,087	50,000	27,886	28,393		49,087	50,000	27,886	28,393

## Summary of Main Findings

- Use **AliFrac** and  $\Delta$ **AliFrac** to capture the speed and intensity of FinTech adoption.
- Higher FinTech adoption results in increased risk-taking.
  - One-std increase of AliFrac leads to increases of 2.8% in participation (avg=38%); 2.9% in risky share (avg=45%).
  - ▶ One-std increase of  $\triangle$ AliFrac leads to increases of  $\triangle$ participate=0.3% and  $\triangle$ risky share=1.5%.
  - ▶ Instrumented with Distance to Hangzhou, one-std increase of AliFrac leads to increases of **2.6%** in participation and **4.1%** in risky share.
- Who benefits more from FinTech inclusion?
  - Individuals with higher risk tolerance.
  - Cities under-served by traditional banks.
  - ▶ Mature and high risk tolerance individuals living in under-banked cities.

#### Related Literature

- Portfolio Choice: Markowitz (1952), Tobin (1958), and Merton (1969, 1971).
- Household Finance: Campbell (2006).
- Risk-Taking and
  - ► Consumption Volatility: Mankiw and Zeldes (1991).
  - ▶ Familiarity: Hong, Kubic and Stein (2004).
  - ► **Trust:** Guiso, Sapienza, and Zingales (2008).
- Technology and Investor Behavior:
  - ▶ Internet and stock trading: Barber and Odean (2002).
  - ► FinTech platforms and mutual fund flows: Hong, Lu, and Pan (2020).
  - Mobile money in developing economies: Suri (2017).
  - ▶ Digital loans in Kenya: Suri, Bharadwaj, and Jack (2021).

## Data and Key Measures

- Unique FinTech data:
  - ► Account-level for a random sample of 50,000 individuals (age, gender, location).
  - Monthly consumption, investment, and payments from Jan 2017 to Mar 2019.
  - ▶ Out of all users, 28,393 active users with at least 100 RMB fund purchases.
- Risk-taking measures:
  - ▶ Participate: 1 for active users with a positive investment in risky funds
  - Risky Share: Portfolio weight on the risky funds.
  - **Portfolio Volatility** ( $\sigma_W$ ): Estimated from monthly returns.
- Risk tolerance:
  - ▶ Proxied by **consumption volatility**  $(\sigma_{C})$ .
  - ▶ Individuals with higher  $\sigma_{C}$  are more risk tolerant.
    - \* Explicit under the complete-market setting of Merton:  $\sigma_{\rm C} = \sigma_{\rm W} = \frac{1}{\gamma} \frac{\mu r}{\sigma_{\rm R}}$ .
    - ★ Valid under more general settings with certain assumptions.

# Data and Key Measures

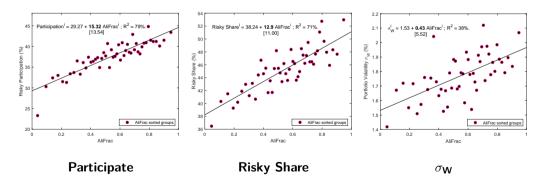
28,393 Active Users ( $> 100$ RMB Fund Purchases)											
			Consum	Consumption		FinTech		Risk-Taking			
	Female	Age	C (¥)	$\sigma_{C}$	AliFrac	$\Delta$ AliFrac	Participate	Risky Share	$\sigma_{W}$ (%)		
Mean	0.61	31.1	2,292	1.21	0.55	0.08	0.66	0.45	1.77		
Median	1.00	30.0	1,396	1.16	0.57	0.07	1.00	0.15	0.18		
Std	0.49	7.8	4,732	0.40	0.22	0.17	0.47	0.47	2.97		
					AII 50,00	0 Users					
Mean	0.61	30.4	2,155	1.21	0.54	0.08	0.38				
Median	1.00	29.0	1,259	1.16	0.56	0.07	0.00				
Std	0.49	7.8	17,063	0.40	0.22	0.22	0.48				

- FinTech adoption from zero to one corresponds to an increase of
  - ▶ 12.7% in risky participation (average=38% among 50,000 individuals)
  - ▶ 13.1% in risky share (average=45% among 28,393 individuals)
  - ▶ **0.43%** in portfolio volatility (average=1.77% among 28,393 individuals)

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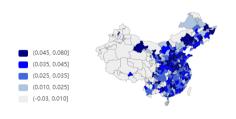
	Partio	cipate	Risky	Share	$\sigma_{W}$ (%)		
AliFrac	0.127***	0.239***	0.131***	0.146***	0.431***	0.446***	
	(10.47)	(17.94)	(7.65)	(7.80)	(4.76)	(4.59)	
$\sigma_{C}$	0.037***	0.019***	0.052***	0.018***	0.345***	0.163***	
	(7.37)	(3.69)	(7.87)	(2.72)	(8.43)	(4.07)	
Log(C)		0.076***		0.031***		0.128***	
		(30.06)		(9.03)		(5.46)	
Female		-0.067***		-0.102***		-0.542***	
		(-12.24)		(-15.12)		(-15.52)	
Log(Age)		0.007		-0.171***		-0.861***	
-, -,		(0.57)		(-11.11)		(-10.50)	
City FE	Υ	Ϋ́	Υ	Y	Υ	Y	
Adjusted R2	0.004	0.024	0.006	0.025	0.004	0.016	
N	50,000	50,000	28,393	28,393	28,393	28,393	

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 $\Delta$ AliFrac: 2018 minus 2017

	$\Delta$ Participate	$\Delta$ Risky Share
$\Delta$ AliFrac	0.014**	0.087***
	(2.08)	(5.30)
$\sigma_c$	0.009**	-0.010
	(2.23)	(-1.32)
Log(C)	0.013***	0.000
	(8.25)	(0.10)
Female	-0.025***	-0.004
	(-8.31)	(-0.68)
Log(Age)	-0.041***	0.012
	(-5.98)	(0.98)
City FE	Υ	Υ
Adjusted R2	0.004	0.154
N	50,000	28,393

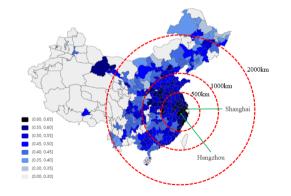
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- Monthly panel regressions with high-dimensional fixed effects:

Fixed Effect	Participate	Risky Share
none individual	12.6% 9.53%	11.1% 3.90%
$month \times city \\ individual + month \times city$	6.95% 0.57%	9.17% 1.95%

- FinTech adoption at city level: less affected by individual's self-selection.
  - ▶ Results consistent with our individual-level findings, both in levels and changes.

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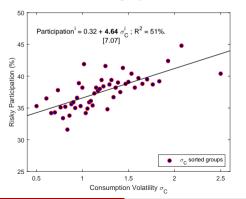
	First Sta	age: Y=Ali	Frac									
	≤200	≤500	≤1000	≤2000	All							
Log(Distance to HZ)	-0.392***	-0.437***	-1.096**	-1.955**	-1.995**							
_ ,	(-5.99)	(-3.99)	(-2.31)	(-2.14)	(-2.16)							
Controls	Y	Y	Y	Y	Y							
Time FE	Υ	Υ	Υ	Υ	Υ							
Observations	238	799	2,278	4,624	4,879							
R-squared	0.85	0.66	0.54	0.51	0.50							
First Stage: Y=AliFrac												
	First Sta	age: Y=Ali	Frac									
	First Sta ≤200	age: Y=Ali ≤500	Frac ≤1000	≤2000	All							
Log(Distance to SH)				≤2000 -1.731*	All -1.766*							
Log(Distance to SH)	≤200	≤500	≤1000									
Log(Distance to SH) Controls	≤200 0.124	≤500 0.129	≤1000 -0.936*	-1.731*	-1.766*							
,	≤200 0.124	≤500 0.129	≤1000 -0.936*	-1.731*	-1.766*							
Controls	≤200 0.124 (0.54) Y	≤500 0.129 (0.70) Y	≤1000 -0.936* (-1.84) Y	-1.731*	-1.766*							

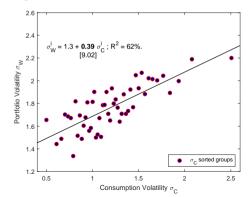
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- Comparing the economic significance of IV vs OLS tests:

	FinTech	Risky Share					
	(one-std)	All Cities	≤500km				
OLS	AliFrac	<b>1.17%</b> (3.04)	<b>2.34%</b> (2.21)				
IV	AliFrac	<b>1.16%</b> (2.32)	<b>4.10%</b> (5.26)				

#### Who Benefits More from FinTech Inclusion? Risk Tolerance

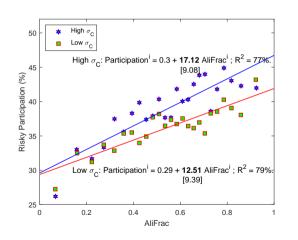
- Optimal portfolio weight  $w^* = \frac{\mu r}{\gamma \sigma_R^2}$ . Higher for more risk-tolerant individuals.
- ullet Proxy risk tolerance using consumption volatility  $\sigma_{\rm C}$ .
  - ▶ Male and young investors have higher  $\sigma_{C}$ .
  - More risk-taking by individuals with higher  $\sigma_{\rm C}$ .

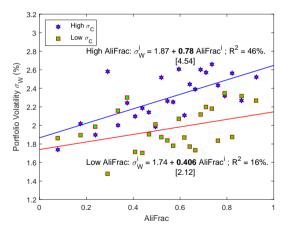




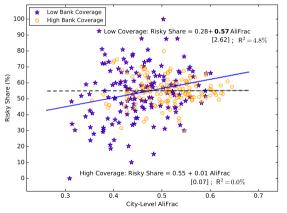
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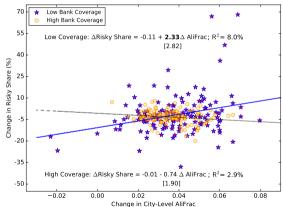
#### FinTech benefits stronger for more risk-tolerant individuals





#### FinTech benefits stronger for cities with low bank coverage





		Low Bank		High Bank		Low-	High
	N	Mean	Std.	Mean	Std.	Mean	t-stat
$\sigma_{C}$	4,053	1.208	0.381	1.208	0.371	0.000	0.00
Log(C)	4,053	7.248	0.822	7.249	0.814	-0.001	-0.65
Female	4,053	0.592	0.491	0.592	0.491	0.000	N.A.
Log(Age)	4,053	3.425	0.237	3.426	0.237	0.000	-1.20
AliFrac	4,053	0.478	0.234	0.566	0.211	-0.089	-20.13
$\Delta$ Ali $Frac$	4,053	0.100	0.182	0.087	0.167	0.013	3.46
Risky Share	4,053	0.457	0.475	0.443	0.470	0.014	1.34
$\Delta$ Risky Share	4,053	0.067	0.456	0.055	0.445	0.011	1.14

	Risky Share					$\Delta$ Risky share			
	Low Bank	High Bank	Low-High			Low Bank	High Bank	Low-High	
AliFrac	0.183***	0.148***	0.035		$\Delta$ AliFrac	0.086**	-0.034	0.121**	
	(4.87)	(2.91)	(0.55)			(2.13)	(-0.79)	(2.03)	
$\sigma_{C}$	0.037*	0.008	0.028		$\sigma_{C}$	-0.015	-0.022	0.007	
	(1.71)	(0.37)	(0.90)			(-0.63)	(-1.04)	(0.23)	
Log(C)	0.039***	0.039***	0.000		Log(C)	0.003	-0.021**	0.024*	
	(3.75)	(4.20)	(0.01)			(0.28)	(-2.23)	(1.79)	
Female	-0.119***	-0.076***	-0.043*		Female	0.009	-0.016	0.026	
	(-7.23)	(-3.92)	(-1.70)			(0.48)	(-1.01)	(1.02)	
Log(Age)	-0.114***	-0.216***	0.102*		Log(Age)	-0.002	0.017	-0.019	
	(-3.46)	(-5.22)	(1.93)			(-0.07)	(0.48)	(-0.41)	
Constant	0.502***	0.849***	-0.676***		Constant	0.06	0.19	-0.125	
	(3.20)	(4.52)	(5.52)			(0.44)	(1.25)	(-1.23)	
City FE	Ϋ́	Ϋ́	` '		City FE	Ϋ́	Ϋ́	, ,	
Observations	4,053	4,053			Observations	4,053	4,053		
R-squared	0.083	0.064			R-squared	0.05	0.036		

Age	Low Bank	High Bank	Low-High		Gender	Low Bank	High Bank	Low-High
Young	0.132**	0.276***	-0.144	Ì	Female	0.171***	0.152**	0.019
	(2.05)	(2.97)	(-1.28)			(3.36)	(2.17)	(0.21)
Mature	0.192***	0.048	0.144*		Male	0.175***	0.104	0.071
	(3.85)	(0.79)	(1.83)			(3.11)	(1.36)	(0.74)
Consumption	Low Bank	High Bank	Low-High		$\sigma_{C}$	Low Bank	High Bank	Low-High
Low	0.194***	0.102	0.094	Ì	Low $\sigma_{C}$	0.136**	0.202***	-0.066
	(3.35)	(1.61)	(1.10)			(2.47)	(3.27)	(0.80)
High	0.186***	0.159**	0.025		High $\sigma_{C}$	0.242***	0.058	0.184**
	(3.07)	(2.07)	(0.26)			(4.64)	(0.80)	(2.06)

#### Conclusions

- We study how FinTech can help households lower barrier and improve risk-taking:
  - ▶ FinTech adoption improves risk-taking, more for risk-tolerant individuals.
  - ▶ Cities with low banking coverage benefit more from FinTech penetration.
  - ► Mature and high risk tolerance individuals under-served by banks benefit more from FinTech inclusion.
- Interpretations of our findings:
  - ► FinTech convenience reduces physical costs, increasing participation.
  - Repeated usage of Alipay builds familiarity and trust, reducing the psychological barriers against investing in risky assets.
- Future of FinTech:
  - ▶ Brighter for emerging economies lacking financial infrastructures.
  - ▶ From Tech to Fin, more content building.