

# Risky Assets and Household Investment Behaviors: The Causal Effect of Income on Risk Preferences

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## Intro.

Finance theory predicts that rational households are expected to allocate their risky assets in a well-diversified portfolio (Markowits, 1952; Rubinstein, 2002). However, there is a research gap to explore how households’ socio-economic characteristics influence risky asset investment behavior.

## Research Question

This research aims to explore what socio-economic factors drive household risky asset investment behavior, and take China as an example. In particular, **this research would like to study the causal effect of households’ annual income on the households risky investment behavior.**

## Data

- 2019 Chinese Household Financial Survey (CHFS) dataset, which explores income and expenditure, social programs and insurance involvement, preferred financial tools, and investment behaviors with socio-demographic (CnOpenData, 2019).
- dataset consists of 17,149 households
- Provide different geo-spatial information across provinces, districts and counties, and villages (neighborhoods) in China

	hhid	RB	RS
Min.	:2.019e+09	Min. :0.0000	Min. : 0
1st Qu.:	:2.019e+09	1st Qu.:0.0000	1st Qu.: 0
Median :	:2.019e+09	Median :0.0000	Median : 0
Mean :	:2.019e+09	Mean :0.4757	Mean : 9131
3rd Qu.:	:2.019e+09	3rd Qu.:1.0000	3rd Qu.: 1000
Max.	:2.019e+09	Max. :1.0000	Max. :700000019
	household	lnincome	city.level
Min.	:0.0000	Min. : 0.000	Min. :1.000
1st Qu.:	:0.0000	1st Qu.: 9.776	1st Qu.:2.000
Median :	:0.0000	Median :10.714	Median :2.000
Mean :	:0.4067	Mean :10.433	Mean :2.214
3rd Qu.:	:1.0000	3rd Qu.:11.290	3rd Qu.:3.000
Max.	:1.0000	Max. :17.910	Max. :3.000

	age	knowledge	edu	risk
Min.	: 10.00	Min. :0.000	Min. :1.000	Min. :0.0000
1st Qu.:	34.00	1st Qu.:3.000	1st Qu.:3.000	1st Qu.:0.0000
Median :	45.00	Median :3.250	Median :4.000	Median :0.0000
Mean :	47.56	Mean :3.332	Mean :4.365	Mean :0.1903
3rd Qu.:	62.00	3rd Qu.:4.000	3rd Qu.:6.000	3rd Qu.:0.0000
Max.	:101.00	Max. :4.500	Max. :9.000	Max. :1.0000

Variables	Description
hhid	Household ID
RB	RB could see a household’s willingness to participate in stocks, bonds, financial derivatives and non-RMB market. Binary variable: If the sum is larger than 0, then RB=1. otherwise RB=0
Age	Age is defined as the median age in the household. Continuous variable: we calculate the median of the age in each household.
Knowledge	Knowledge means the level of a household’s financial knowledge. An index using the responses for each household, range from 0 to 4.5.
Education	Education is defined as the highest level of education that one household ever reached. Categorical variable: primary school and below = 1, junior high school =2, senior high school =3, and college and above =4.
Risk	Risk means the household representative’s risk preference. Binary variable: setting the responses below 3 equals 1(prefer risk), and answers 3 and above as 0(avoid risk).
Household type	Household type represents where the household is located, urban or rural. Binary variable: set rural type=0, and non-rural type(city) =1.
City level	City level tries to categorize what level of city a household lives in. Categorical variable: level 1 city =1, level 2 city = 2, level 3 city = 3.
income	Income means a household annual income. Continuous variable: sum up all individual values within one household.

## Method

### 3.1 Causal effect identification

To estimate the local average treatment effect (LATE) in a fuzzy regression discontinuity design (RDD).

An arbitrary low-income cutoff, set at 6,450 yuan by The State Council of the People’s Republic of China (2023)

## Results

Table 2. Model estimates of the logistic regression

	Model 1:	Model 2:	Model 3:	Model 4:
Treatment(cutoff)	1.813*** (0.582)	1892*** (0.590)	1.896*** (0.602)	1.911*** (0.598)
Annual income	-0.0005 (0.0005)	-0.0004 (0.0006)	-0.0004 (0.0006)	-0.0004 (0.0005)
Age of the household		0.002 (0.010)	0.002 (0.010)	
Household’s type		-0.398 (0.401)	-0.400 (0.400)	-0.400 (0.400)
Risk preference		0.073 (0.795)		
Education level		0.396*** (0.128)	0.396*** (0.128)	0.383*** (0.111)
Financial knowledge level		0.333* (0.186)	0.333* (0.186)	0.332* (0.186)
City level		0.017 (0.170)	0.016 (0.170)	
Note:	*p<0.1; **p<0.05; ***p<0.01			

### 3.2 Regression model of the variables

The following six models are considered:

- $ln(\frac{P(RB=1|D,X,Z)}{P(RB=0|D,X,Z)}) = \beta_0 + \beta_1 D + \beta_2 X,$   
Where we do not control for any covariates.
- $ln(\frac{P(RB=1|D,X,Z)}{P(RB=0|D,X,Z)}) = \beta_0 + \beta_1 D + \beta_2 X + \beta_3 Z,$   
where  $Z = \{age, knowledge, edu, household, city.level, risk\}$  (full model)
- $ln(\frac{P(RB=1|D,X,Z)}{P(RB=0|D,X,Z)}) = \beta_0 + \beta_1 D + \beta_2 X + \beta_3 Z,$   
Where  $Z = \{age, knowledge, edu, household, city.level\}$
- $ln(\frac{P(RB=1|D,X,Z)}{P(RB=0|D,X,Z)}) = \beta_0 + \beta_1 D + \beta_2 X + \beta_3 Z,$   
Where  $Z = \{knowledge, edu, household\}$
- $ln(\frac{P(RB=1|D,X,Z)}{P(RB=0|D,X,Z)}) = \beta_0 + \beta_1 D + \sum_{i=1}^4 \beta_i B_i(X) + \beta_4 Z$   
, fitted with B-splines with order 4, Where  $Z = \{knowledge, edu, household\}$ .
- $ln(\frac{P(RB=1|D,X,Z)}{P(RB=0|D,X,Z)}) = \beta_0 + \beta_1 D + \beta_2 X + \beta_3 X^2 + \beta_4 Z$   
,Where  $Z = \{knowledge, edu, household\}$ .

Table 3. Model estimates of different functional form

	Model 4:	Model 5:	Model 6:
Treatment(cutoff)	1.911** (0.598)	1.691* (0.936)	2.42*** (0.676)
Annual income	-0.0004 (0.0005)	1 2 3 4 2.1 -0.6 2.6 0.4	1 2 -7.442 -4.826* (5.5) (2.6)
Age of the household		(2) (1.2) (2.2) (1.4)	(5.5) (2.6)
Household’s type	-0.400 (0.400)	-0.453 (0.407)	-0.462 (0.406)
Risk preference			
Education level	0.383** (0.111)	0.382*** (0.112)	0.390*** (0.112)
Financial knowledge level	0.332* (0.186)	0.331* (0.189)	0.349* (0.187)
City level			
Note:	*p<0.1; **p<0.05; ***p<0.01;"1,2,3,4" is the order of corresponding functional form		

Table 4. Model estimates of alternative bandwidth

	Narrow:	middle:	Wider:
Treatment(cutoff)	2.440*** (0.946)	2.872*** (0.634)	3.149*** (0.509)
Annual income	1 2 -1.74 -2.79	1 2 -16.2*** -12.1***	1 2 -32.5*** -22.6***
	(4.80) (2.6)	(6.01) (2.98)	(7.31) (3.76)
Age of the household			
Household’s type	-0.088 (0.450)	-0.370 (0.390)	0.726*** (0.133)
Risk preference			
Education level	0.282* (0.145)	0.416*** (0.110)	0.482*** (0.035)
Financial knowledge level	0.496** (0.238)	0.334*** (0.181)	0.459*** (0.079)
City level			
Note:	*p<0.1; **p<0.05; ***p<0.01;"1,2,3,4" is the order of corresponding functional form		

## Conclusion

### 5.1 Discussion

- Future research could consider explore how economic fluctuations and public health emergencies(e.g. Covid 19) using 2022 CHFS data influence financial decision-making over the long term.
- Latent variables, such as household expectations of future income, confidence in financial markets, and attitudes toward risk, could be considered in the future research.

### 5.2 Limitation

- The sample may not fully represent all households’ situations in China, thus limiting the external validity.
- key macroeconomic indicators, such as GDP and interest rates, were not included in this study.

## Reference

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