

Financial Literacy and Participation in the Financial Markets in Tanzania: An Application of Logit Regression Model

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Abstract: This paper hypothesizes that financial literacy and awareness have a major impact on individuals' participation in the financial markets. The paper empirically examines the factors determining the participation of individuals in the financial markets using qualitative data collected through questionnaire covering a sample of 484. Categorical data were collected from individuals residing in Arusha, Tanzania. The paper applies logistic regression model and maximum likelihood technique of estimation to analyse the variables that influence the participation of the financial markets. Findings show that there is a positive and statistically significant relationship between financial literacy, awareness and people's participation in the financial markets. The implication here is that people with financial knowledge have a better chance of investing in financial market despite their education level. Indeed surprisingly, results reveal that the level of education has no effect on the participation of the financial markets. Moreover, other variables such as gender, marital status, risk attitude and level of income play a significant role in influencing individuals' participation in the financial markets. It is revealed that the male and married individuals are more likely to invest in the financial markets. The policy implication of these results is that increasing training, awareness of the benefits, and operations of the financial markets, will result in people opting to participate in the financial participation which will, in turn, lead to increased trading of financial assets and hence create a ripple effect to the country's economy.

Keywords: Financial Literacy, Financial Markets, Logit Modeling

JEL Classification: C35, D81, G10.

1. INTRODUCTION

Limited level of financial market participation is a long-standing puzzle in empirical finance. The phenomenon is observed in many countries across the world, and indeed, participation rates vary widely even among developed countries. Many studies explain the phenomenon from the viewpoint of financial literacy, awareness, investors' risk aversion, and cost of stock market participation but could not be successful. (Almenberg & Dreber (2015), Arts (2018), Dillen & De Lille (2018), Gumbo & Maxwell

(2018), van Rooij, *et al.* (2011) and Sindambiwe (2014). As a result, households' lack of participation in the stock market remains a puzzle across the world (Kadoya *et al.*, 2017).

The Bank of Tanzania Act of 1995 paved the way for the development of financial markets in the country. The main purpose was to promote and facilitate the developments of an orderly, fair, and efficient capital market and securities in Tanzania. According to Bank of Tanzania (2018), financial markets in Tanzania consist of money, bonds, equities, foreign exchanges, and collective investment schemes. Also, according to Financial Inclusion Report (2017), 56 percent of adults in Tanzania are financially included through the use of mobile banking services. However, despite the fact that in 2006 the Banking and Financial Institutions' Act was formed to ensure that there are stable, safe, and sound financial systems; banks, and pension funds are preeminent in Tanzania's small financial sector (IMF, 2018). Financial sector assets are 36 percent of GDP, with banks (72 percent of system assets) and pension funds (26 percent of system assets) being systemic components. In Kenya for example, the share of financial sector assets in GDP is 76 percent (Financial Sector Regulators, 2019). More significantly, the financial market in Tanzania fails to reach its objectives due to a number of challenges including low financial education and literacy as well as low access to financial services to the majority of Tanzanians (Macha, 2009). Although, Tanzania decided to embark on financial liberalisation in 1991 to sustain its growth, as it considers the capital market development an important component of the economic recovery programme, its financial market is still described as an emerging market with only 28 companies listed in Dar es Salaam Stock Exchange (DSE) with a total market capitalization of TZS 20,186.48 billion, equivalent to US\$ 8.7 billion, as of 18th December 2015. Out of 44.9 million Tanzanian citizens (NBS, 2012), 52 percent are estimated to be a working-class in which only 200,000 are individual investors participating in the financial markets in Tanzania. This approximates 0.9 percent of total working-class participants.

Understandably, as Ziorklui (2001), Shahabadi, & Jafari (2017), and Babu (2018) argue, financial market institutions such as the stock market can assist better in the mobilization of domestic capital. This is due to the fact that they offer a simple mechanism for fund transfer which encourages investment. They also permit companies to have access to a larger number of investors in comparison to private-owned companies. Owing to this, the growth and development of the financial markets are of vital importance to the country as well as individuals' economy. However, for any market to progress and people to participate effectively there is a need for awareness and education on how the market operates (Balloch, *et al.*, 2015). Financial

decision making is affected by an individual's level of financial literacy since low literate individuals are less likely to invest in stocks and therefore are less likely to participate in the stock market (van Rooij *et al.*, 2011; Aubert *et al.*, 2018; and Arts, 2018).

Although many studies discuss the relevance of financial literacy to financial market participation and the importance of the latter, participation rates still are considerably low (van Rooij *et al.*, 2011; Andersen & Nielsen, 2011, and Arts, 2018). Additionally, substantial differences exist in stock market participation rates between countries (Guiso *et al.*, 2003). Nonetheless, evidence regarding the impact of financial illiteracy on financial behavior has been both scarce and mixed (Martin, 2007; Agarwal *et al.*, 2011). One reason for these limitations is that a substantial fraction of existing studies that address this question are based on the evaluation of specific financial education programs and policies (Yoong, 2011). Indeed, according to Bayer *et al.* (1996) & Bernheim *et al.* (2001), employer-based financial education increases participation in saving plans, while financial education significantly increases the adult propensity to save. However, other studies such as Duflo & Saez (2003) and Cole & Shastry (2009) have found small impacts of financial education on financial decision-making. Notably, financial education programmes may fail to affect literacy (Yoong, 2011). By contrast, Hilgert *et al.* (2003) find that individuals with more financial knowledge are more likely to engage in recommended financial practices. Similarly, Lusardi & Mitchell (2006, 2007) demonstrate that individuals with better financial knowledge are more likely to plan and to invest in complex assets.

The relationship between financial literacy and participation in the financial market is generally complex. Studies such as Hong, *et al.* (2004); Georgarakos & Pasini (2011); Almenberg & Dreber (2015) show that financial literacy and education apart, other individuals' characteristics namely, age, gender, wealth, and risk aversion, may influence financial market participation. Nevertheless, like financial literacy, the effect of other individuals' characteristics on financial market participation is controversial. For example, according to Georgarakos & Pasini (2011), households with a similar level of average wealth holdings are likely to have different rates of financial market participation rates. They differ even to such extent that households with lower average net wealth holdings have higher financial market participation rates than households with higher average net wealth holdings (Georgarakos & Pasini, 2011). This revelation indicates that more investigations and analyses of the effect of financial literacy, awareness, and other individual characteristics on participation in the financial market is of paramount importance.

This paper, therefore, empirically examines the effect of financial literacy and other factors, namely age group, marital status, level of education, level of income, risk attitude, and self-assessed financial knowledge on participation in the financial markets in Tanzania, taking a case study of Arusha city, using categorical data and logistic regression analysis. The paper is confined itself to Dar es Salaam Stock Exchange (DSE) market and mutual fund market: Unit Trust of Tanzania (UTT), with the aim of providing policy recommendations that help to improve marketing strategy and individual financial market participation and investment. The rationale behind investigating the relationship between financial literacy and other individual characteristics on the financial market participation is that previous studies show mixed findings. More significantly so, the improvement of public awareness about financial market investment would increase the rate of participation, which would in turn lead to economic growth and development. This is very important because financial markets play an important role in the economy by creating the open and regulatory system for the firms to acquire amounts of capital resulting in companies issuing shares to the public. This helps the companies to raise enough capital for businesses through selling company shares and bonds. The important benefit of this is reflected in the individual's savings along with investments in the securities which then promote commerce and the industry as opposed to idle bank deposits (Saeed, 1971; Fohlin, 1999; Antwi-Asare & Addison, 2000).

2. LITERATURE REVIEW

The degree of involvement in the financial markets is greatly influenced by the awareness and knowledge of such market's operations, risks, and benefits involved (Lusardi & Mitchell, 2014; Lusardi & Mitchell, 2007a; van Rooij *et al.*, 2011; Andersen & Nielsen, 2011; Arts, 2018). Several studies have attempted to find the degree of association between participation, awareness, and literacy in the financial market. For example, Japelli & Guiso (2005) find that lack of awareness as well as ignorance on investment opportunities are one of the major reasons that many people do not invest in mutual funds, investment accounts, stocks, and corporate bonds. The implication is that if all investors are aware of the risk involved in the participation of the investment activities, then many will invest and thus increase in the investment activities. In addition, if consumers are greatly aware of the available financial assets then they will be more willing to participate in the markets. Indeed, Japelli & Guiso (2005) findings suggest that forms of information dissemination such as training, advertisement, and incentives may make individuals aware of financial markets.

As has been stated, the relationship between individual characteristics and participation in the financial markets is inclusive. According to Sindambiwe (2014), individuals in managerial positions are more skilled and aware of the operations of the financial markets. They measure high their financial literacy in terms of investment in regards to interest rate calculations, exchange rate calculations, business diversifications, portfolio management, and stock market location but they have little awareness of the existence of brokerage services (Sindambiwe, 2014). Notably, despite the knowledge of the financial market, the level of participation in the financial market is low mainly because companies do not have important information and opportunity of investing mainly due to a lack of financial advisors (Sindambiwe, 2014). This, however, suggests that education and other factors such as demographics and level of income play a role in the decision to participate in the financial market. For example, education is likely to influence the level of awareness and participation in which those with higher education are observed to be more aware compared to the ones with lesser education. Also, Japelli & Guiso (2004) demonstrate, based on the household income, that individuals with a surplus are at a high level to invest compared to those without or with a lesser surplus. Indeed, individuals with a higher level of income have a higher literacy level compared to those with a lower level of income (Klapper *et al.*, 2015). However, according to Sindambiwe (2014), knowledge of the financial market is a major factor in comparison to demographics and income level because it leads to the belief in the return obtained as a result of investing in the markets. In a similar study, Gumbo & Sandada (2018) argue that education is not the main factor influencing financial market participation. Indeed, Gumbo & Sandada (2018) show that individuals with higher education levels participate less in comparison to individuals with lower education levels. According to Gumbo & Sandada (2018), awareness, cognitive skills, and transaction costs are a significant factor in participation in the financial market. Similarly, Omole & Falokun (1999) reveal that a few firms participate in the financial market because individuals' awareness of financial markets is restricted in the Nigeria stock exchange. The government favours the money market rather than the capital markets (Omole & Falokun, 1999). In Tanzania, the effort which is made through the media by DSE and UTT is still not enough, and therefore there is a need for further investigation and hence proper suggestions (Massele *et al.*, 2013). Despite that many studies show a lack of awareness is likely to affect participation in the financial markets, there is still a requirement of further investigation on what level of awareness is required and if at all this is a major factor of influence in the investing selection in Tanzania.

Financial literacy could range from fairly knowledgeable to highly knowledgeable of the undertakings of the financial markets. It is not a necessity to be very knowledgeable of the financial markets and operations but some studies show that just a basic knowledge can cause individuals to participate in the financial markets (van Rooij *et al.*, 2011). For one to be financially literate, they should be able to understand the concepts of financial workings including workings of interest rate, compounding interest, inflation, and risk diversification (Klapper *et al.*, 2015). It is worth noting that education is not a good proxy for financial literacy. That is, according to Lusard & Mitchell (2011), when education and financial literacy are included in multivariate regression models of financial market participation, both tend to be statistically significant, indicating that financial literacy has an effect above and beyond education. García & Tessada (2013) analyse the influence of education on the participation in the financial markets. Their findings show that the increase in the level of education increases the probability of participation by 3 percent. They conclude that education has a major influence on the individual's participation in the financial markets.

Financial illiteracy is widespread even when financial markets are well developed as in Germany, the Netherlands, Sweden, Italy, Japan, and New Zealand (Lusard & Mitchell, 2011). Thus observed low levels of financial literacy in developing countries such as Tanzania, are prevalent elsewhere, rather than specific to any given country or stage of economic development. Furthermore, financial literacy rates differ in important ways when it comes to characteristics such as gender, education level, income, and age. Previous studies show that men are more financially literate than women (Lusardi & Mitchell, 2014). Worldwide, 35 percent of men are financially literate, compared with 30 percent of women (Klapper *et al.*, 2015). Women have weaker financial skills than men even considering variations in age, country, education, and income (Klapper *et al.*, 2015). According to (Klapper *et al.*, 2015) there is a slight variation in the average gender gap in financial literacy between developed and emerging markets. Although women tend to be better decision-makers when it comes to money management and investment, the limited knowledge of the stock market plays a big factor in lacking investment thereof (Kulkarni & Deepa, 2019). The psychological and cultural norms are also at play when gender is concerned with the thought that the male dominates the economy and investment decisions. Despite the fact that females making 40 percent of the overall workforce, they are most unlikely to be educated on financial education (Isaac, 2014).

Moreover, for the emerging markets, youth are more informed and knowledgeable in financial education compared to those with above 35

years of age (Klapper *et al.*, 2015). However, this is contrary to developing economy whereby individuals with above 35 years of age are more financially literate than the youth (Klapper *et al.*, 2015).

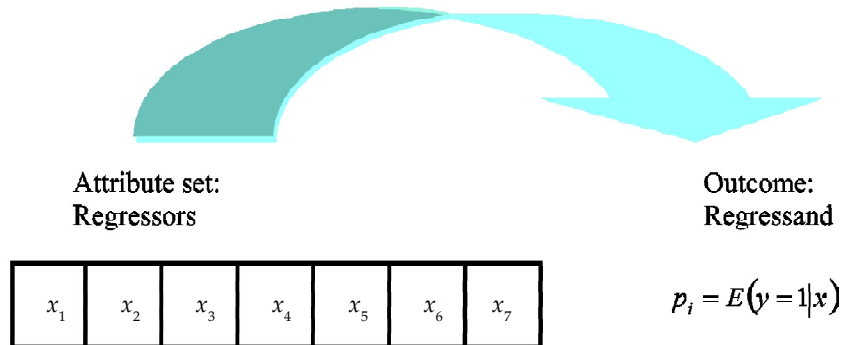
Studies on people's participation in the financial markets show that trust in reference to the financial market is how a person can perceive the involved risk and benefits and decide that the perceived benefits outweigh the risk and hence safe for the individual's investment (see for example Balloch, *et al.*, 2015). Balloch *et al.* (2015) reveal that stock market literacy and trust in the stock markets became the resulting factors for the households to invest in the stock markets. Different individuals have different perceptions of risk. The knowledge of the financial market place may guarantee even the risk-averse individual a position to take the risk. Cheng *et al.*, (2018) argue that if a household can identify and understand the stocks better, they can choose the amount of risk they are willing to sustain. According to Yao *et al.* (2003), despite that females are of higher life expectancy, they are keener and risk-averse. Males are more prone to taking higher risks. Also, as mentioned earlier, age seems to be a factor of interest in the participation of the financial markets. Financial Engines (2018) demonstrate that older people are more careful considering they have higher experience and have picked up habits along the way. In addition, as individuals' age increase, they are more likely to accumulate wealth and hence engage in financial market investments (Kartik, *et al.*, 2016).

The financial market participation puzzle is relatively incomplete. Despite its proven importance, previous studies still could not completely rationalize the rather low financial market participation rates and the factors determining market participation rates. By and large, previous studies, for example, Cheng *et al.* (2018), Sindambiwe (2014), Gumbo & Sandada (2018), and García & Tessada (2013) examine the factors determining participation in the financial markets, specifically stock markets. They focus more on the households and organisations while identifying the factors that influence participation in the stock markets and capital markets. Factors such as age, education level, gender, risk attitude, marital status and self-assessed financial knowledge are widely used in previous studies. It is worth noting, however, that individual participation in the financial markets in Tanzania has been rarely analysed and hence needs more investigation. Theories indicate that individual awareness in the financial market and consequential participation in the financial markets is of great importance, and therefore there is a need for empirical justification in a country such as Tanzania, in which its financial market is still developing. Overall, this paper focuses mainly on the factors influencing the individual investor's decisions to participate in the financial markets taking multiple variables, namely age

group, gender, marital status, education level, risk attitude, income level, and self-assessment of the financial literacy into account.

3. CONCEPTUAL FRAMEWORK

This paper analyses financial literacy and individuals' participation in the financial markets, based on the conceptual framework reported in Figure 1. This framework presents the key variables that are analysed using the logistic regression model and maximum likelihood estimation technique as discussed in the methodology section, section 4.



Variables

x_1 = Gender

x_2 = Age group

x_3 = Marital status

x_4 = Education level

x_5 = Monthly income

x_6 = Risk assessment

x_7 = Self-assessed financial knowledge

p_i = Probability of financial market participation

Figure 1: Conceptual Framework

All regressors and regressand are categorical variables and their measurements are presented and discussed in detail in methodology section. The causal relationship between individual regressors and regressand can be explained as follows: First, as far as the level of education is concerned, several studies have revealed that a person with a higher level of education has come with better financial knowledge (Sindambiwe, 2014). Second, as previous studies have revealed, women are more risk-averse and also good planners (Kulkarni & Deepa, 2019 and Yao *et al.*, 2003). Hence, gender plays a significant role in financial investment. Third, it is believed that the more people have to spare as surplus income the more they will be willing to

invest in different instruments including financial sectors (Japelli & Guiso, 2004; Cheng *et al.*, 2018). Thus, individuals with higher level of income are more likely to participate in financial market. Fourth, age plays a key role in risk-taking. Studies have shown that the younger the person the more willing they are to take riskier options (Kartik *et al.*, 2016). For this reason, age group plays a significant role in determining whether a person can or is willing to invest in several financial instruments. Fifth, researches have found that a married individual is less likely to invest in riskier markets and make the riskier financial decisions in comparison to the single individual. Thus, it is important to analyse if marital status influences the participation of the financial markets. Sixth, there are three types of individual's attitudes to risk; the risk-averse, risk-takers and risk-neutral. The risk attitude can majorly influence one to invest in the financial market if all factors are left constant (Balloch *et al.*, 2014). Therefore, risk attitude is one of the factors examined to establish if an individual can participate in financial markets based on riskiness. As shown in the section of methodology, the risk attitude is being given a score so as to identify the risk character of the participants. Lastly, the paper assesses the perceived financial literacy of the individual to see how individuals evaluate their financial knowledge and participation in the financial market. To assess financial literacy, individuals state how they think they know their level of knowledge in relations to finance by rating their knowledge between scales of 1 to 10. This also is discussed in the methodology section. In summary, the coefficients of the level of education, gender, level of income, age group, marital status, and self-assessed financial knowledge are expected to be positive whereas the coefficient of risk attitude is expected to be negative.

4. METHODOLOGY

4.1. Model Specification

The logit regression model is adopted appropriately to examine a cause-effect relationship between the variables. Logit or Logistic regression analysis studies the relationship between a categorical dependent variable and a set of independent variables. The logistic regression is very applicable here since the dependent variable has only two values, that is 0 and 1 or Yes and No. Logit regression is a nonlinear regression model designed for binary dependent variables. Because regressions with a binary dependent y model the probability $y = 1$, it makes sense to adopt a nonlinear formulation that forces the predicted value to be between 0 and 1 (Stock & Watson, 2007). Indeed, the logit model is a significant improvement on the linear probability model. It is based on the cumulative distribution function (CDF)

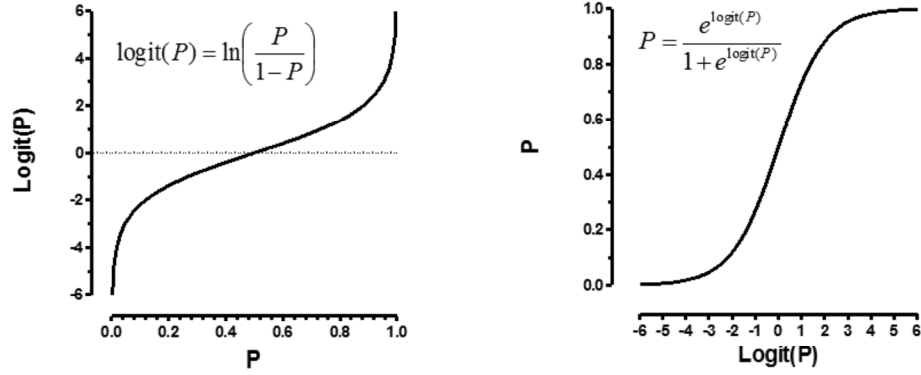


Figure 2: The Logit Model

of a random variable, with the logit model following the logistic CDF, giving the relationship as shown in Figure 2. As it is shown in Figure 2, the regression line is non-linear, giving a more realistic description of the data, with very little change in the probability at the extreme value that the explanatory variable can take. As has been stated, in the logit model, the probability of the dependent variable takes the value of 1 for a given value of the explanatory variable. This method is better suited for modeling and analysing categorical-response variables because it does not assume that the independent variables are normally distributed.

Consider a linear regression model (1) and associated probability function (2). Where y is continuous while the x s can be purely continuous, dichotomous (dummy variables) or both. Recall that a continuous variable is one that can take any value between two numbers.

$$y_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \beta_3 x_{3i} + \beta_4 x_{4i} + \beta_5 x_{5i} + \beta_6 x_{6i} + \beta_7 x_{7i} + \varepsilon_i \quad (1)$$

$$p_i = E[y = 1 | x_i] = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \beta_3 x_{3i} + \beta_4 x_{4i} + \beta_5 x_{5i} + \beta_6 x_{6i} + \beta_7 x_{7i} \quad (2)$$

Where p_i = Probability of financial market participation

x_1 = Gender

x_2 = Age group

x_3 = Marital status

x_4 = Education level

x_5 = Monthly income

x_6 = Risk assessment

x_7 = Self-assessed financial knowledge

$\beta_0, \beta_1, \beta_2, \dots, \beta_7 =$ parameters

$\varepsilon_i =$ The disturbance term

The cumulative Logistic distributive function can then be written as:

$$p_i = \frac{1}{1 + e^{-z_i}} \quad (3)$$

$$z_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \beta_3 x_{3i} + \beta_4 x_{4i} + \beta_5 x_{5i} + \beta_6 x_{6i} + \beta_7 x_{7i} + \varepsilon_i \quad (4)$$

where the probability of financial market participation, P_i , is equal to 1. e , is the exponential under the logit approach. Then, the logistic model estimated would be specified as

$$p_i = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \beta_3 x_{3i} + \beta_4 x_{4i} + \beta_5 x_{5i} + \beta_6 x_{6i} + \beta_7 x_{7i} + \varepsilon_i)}} \quad (5)$$

The model is then, expressed as the odds ratio, which is simply the probability of financial market participation relative to the probability of non-participation in the financial market. Given that

$$1 - p_i = \frac{1}{1 + e^{z_i}} \quad (6)$$

$$\frac{p_i}{1 - p_i} = \frac{1 + e^{z_i}}{1 + e^{-z_i}} = e^{z_i} \quad (7)$$

The ratio $p_i / (1 - p_i)$ is the odds ratio of financial market participation. For example, if the probability of financial market participation, p_i , is 0.75, then the probability of non-participation in the financial market, $1 - p_i$, is 0.25. The odds of participation in the financial market are defined as the ratio of the probability of financial market participation over the probability of non-participation in the financial market. Thus, the odds of financial market participants will be $0.75 / 0.25 = 3$. This will imply that the odds of financial market participants are 3 to 1. Then by taking the natural log of the odds ratio, we produce the logit (L_i), as follows:

$$L_i = \ln\left(\frac{p_i}{1 - p_i}\right) = z_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \beta_3 x_{3i} + \beta_4 x_{4i} + \beta_5 x_{5i} + \beta_6 x_{6i} + \beta_7 x_{7i} + \varepsilon_i \quad (8)$$

Model (8) suggests that L is linear in x , the probabilities (p) are not linear. The logit model can be used in multiple regression tests. The slope coefficient measures the change in the log-odds ratio for a unit change in the explanatory variable. Once the logit has been estimated, hypothesis

testing and econometric analysis can be undertaken in much the same way as for linear equations. If L is positive, as the value of the explanatory variables increases, the odds that the dependent variable equals 1 increase. The model estimates for each respondent the logarithm of the probability of financial market participants to the probability of not participating in the financial market. Table 1 presents the definitions, description and measurements of key variables that are included in the logit model.

Table 1
Definitions and Measurement of Key Variables

	<i>Proxy</i>	<i>Variable</i>	<i>Measurements</i>	
1	p_i	Probability of financial market participation	Participating = 1	Otherwise = 0
	$1 - p_i$	Probability of nonparticipation in the financial market		
2	x_1	Gender	Male = 1	Female = 0
3	x_2	Age group	25+ Years = 1	Below 25 Years = 0
4	x_3	Marital status	Married = 1	Single/divorced = 0
5	x_4	Education level	Above primary = 1	Primary/informal = 0
6	x_5	Monthly income	Above TZS 300,000 = 1	Less than/equal to TZS 300,000 = 0
7	x_6	Risk assessment	6-10 Scores = 1	1-5 Scores = 0
8	x_7	Financial knowledge	5-10 Scores = 1	1-5 Scores = 0

Source: Authors' construction (2019)

4.2. Sample Size, Nature of Data and Data Collection Method

The paper targeted the adult population within Arusha city who were randomly selected. The adult population was selected because of the likelihood to have been heard of, or participated in the market and hence would be able to provide viable feedback. The paper, therefore, targeted 130,000 individuals in Arusha city (NBS, 2012). The sample size was computed using Yamane (1967) probability sampling technique. In common practice, Yamane (1967) provides a simplified formula to calculate the sample size. This simplified population proportion method that corrects the weaknesses found in Cochran and the finite population correction methods, and assuming a 95% confidence level and $\rho = 0.5$ can be expressed as follows:

$$n = \frac{N}{1 + N(e)^2} \quad (9)$$

where n = The sample size

N = The population size

e = The acceptable sampling error or the error precision level

Applying expression (9), the sample size (n) of 130,000 population (N) was obtained as follows:

$$n = \frac{N}{1 + N(e^2)} = \frac{130000}{1 + 130000(0.05^2)} = \frac{130000}{326} = 399 \quad (10)$$

However, the fact that the logistic model requires a big sample size, the study managed to obtain 484 respondents; hence a sample size of 484 individuals was used for analysis. Indeed, the sample size of 484 is appropriate for logistic regression (Hsieh, 1989).

Data were collected using the online questionnaire. Questionnaires were distributed through the internet and once complete were returned via email. Questions were randomly administered to several people selected randomly given a period of two weeks to respond and resend the questionnaires. This approach could quickly and easily capture the relevant information from a large number of respondents in a short period of time. The questionnaire captured relevant information such as general profile of the respondent including location, age, gender, income, marital status as well as education level. It also captured significant information such as risk assessment, financial literacy assessment, as well as financial market participation of the respondent.

To ensure there is no biasedness of data, the randomisation technique was used. This technique helps to eliminate selection and confounding bias and thus guaranteeing the groups are comparable regardless of the occurrence of factors other than those being examined (Kang, 2013). A pilot test was administered so as to ensure the applicability and suitability of the content, clarity and sufficiency of the instruments, which would in turn ensure validity and reliability of data. Equally important, to ensure that data are reliable and that they can be used in testing other models, Cronbach's alpha test was applied. The Cronbach's alpha test measures the internal consistency of data; it is the measure of scale reliability (Gliem & Gliem, 2003). It was computed by correlating score for each question in the questionnaire and results are reported in Table 2. Since the scale reliability coefficient is 0.7, which is between 0.65 and 0.8, data have internal reliability. Usually the coefficients below 0.5 are unacceptable. Equally noteworthy, chi-square goodness of fit test is used to check for the consistence of random distribution of the participants. Results are reported

Table 2
Cronbach's Alpha tests for Reliability

Test scale = mean (unstandardized items)	
Reversed item: education level	
Average inter item covariance:	0.0262888
Number of items in the scale:	18
Scale reliability coefficient:	0.7332

Source: Authors' estimates

Table 3
Chi-Square Goodness-of-Fit Test

Number of observations	=	484
Number of covariate patterns	=	78
Pearson χ^2 (70)	=	100.90
Prob > χ^2	=	0.0092

H_0 : All data are consistent with a random distribution

H_1 : At least one of the data is inconsistent with random distribution

Source: Authors' estimates

in Table 3. Since the probability of $\chi^2 = 0.009$ is less than the 0.05 level of significance, we fail to reject the null hypothesis that data are consistent with random distribution.

Table 4 provides a summary of categorical responses of all 484 individuals who participated in the process of data collection. Qualitative data are based on the number or percentage of financial market participations. Responses also are itemized into gender, age group, and marital status. Other categories of respondents include education level, monthly income, self-assessed financial knowledge, and attitude towards risk. Presumably, since financial markets rise and fall at any given time, resulting into gains and losses, individuals more inclined to take risks are more willing to invest in financial trading. As reported in the Table, 69.6 percent of the respondents have positive risk attitude. It is also revealed that about 93.2 percent of the respondents seem to have either fair, or moderate, or advanced knowledge of financial market. Surprisingly however, only 17 percent of the respondents with diploma or degree level of education, participate in the financial market while 59 percent of respondents with primary or informal education participate in the financial knowledge.

Table 4
Descriptive Analysis of the Variables

<i>Categorical variable</i>	<i>Response</i>	<i>Frequency</i>	<i>Percent</i>
1 Financial market participation (DSE or UTT)	Yes	134	27.7
	No	350	72.3
2.1 Gender responses	Male	301	61.8
	Female	183	38.1
2.2 Male participation (of 301 males)	Yes	101	33.6
	No	200	66.4
2.3 Female participation (of 183 females)	Yes	33	18.0
	No	150	82.0
3.1 Marital status	Married	205	42.3
	Single	279	57.7
3.2 Financial market participation (of 205 married)	Yes	88	43.0
	No	117	57.0
3.3 Financial market participation (of 279 single)	Yes	110	39.4
	No	169	60.6
4.1 Age group, 18-24 years		155	32.0
4.2 Age group, 25-35 years		256	53.0
4.3 Age group, above 35 years		73	15.0
4.4 Age group, 18-24 years, participation	Yes	23	15.0
	No	132	85.0
4.5 Age group, 25-35 years, participation	Yes	67	26.1
	No	189	73.9
4.6 Age group, above 35 years, participation	Yes	40	54.8
	No	33	45.2
5.1 Education level, primary & informal		80	16.5
5.2 Education level, secondary education		84	17.3
5.3 Education level, diploma/degree		264	54.6
5.4 Education level, postgraduate		56	11.6
5.5 Education level, primary & informal, participation	Yes	47	59.0
	No	33	41.0
5.6 Education level, secondary education, participation	Yes	22	26.0
	No	62	74.0
5.7 Education level, diploma/degree, participation	Yes	45	17.0
	No	219	83.0
5.8 Education level, postgraduate, participation	Yes	18	32.0
	No	38	68.0

Source: Authors' computations

Table 4
Descriptive Analysis of the Variables

<i>Categorical variable</i>	<i>Response</i>	<i>Frequency</i>	<i>Percent</i>
6.1 Monthly income, below TZS 300,000		135	27.9
6.2 Monthly income, TZS 300,000 – TZS 1Million		222	45.8
6.3 Monthly income, TZS 1Million– TZS 2Million		98	20.3
6.4 Monthly income, above TZS 2Million		29	6.0
6.5 Income, below TZS 300,000, participation	Yes	22	16.0
	No	113	84.0
6.6 Income, TZS 300,000 – TZS 1M, participation	Yes	70	31.4
	No	152	68.6
6.7 Income, TZS 1– TZS 2M, participation	Yes	32	32.7
	No	66	67.3
6.8 Income, above TZS 2M, participation	Yes	10	34.5
	No	19	65.5
7.1 Self-assessment, not knowledgeable		81	16.8
7.2 Self-assessment, fairly knowledgeable		203	41.9
7.3 Self-assessment, moderately knowledgeable		148	30.6
7.4 Self-assessment, very knowledgeable		52	10.7
7.5 Not knowledgeable, participation	Yes	11	13.4
	No	70	86.6
7.6 Fairly knowledgeable, participation	Yes	52	25.5
	No	151	74.5
7.7 Moderately knowledgeable, participation	Yes	31	20.8
	No	117	79.1
7.8 Very knowledgeable, participation	Yes	38	73.1
	No	14	26.9
8.1 Riskiness, risk taker		337	69.6
8.2 Riskiness, risk averse		147	30.4
8.3 Risk taker, participation	Yes	106	31.5
	No	231	68.5
8.4 Risk averse, participation	Yes	26	18.0
	No	121	82.0

Source: Authors' computations

4.3. Estimation Method: Maximum Likelihood

As has been presented, Logistic regression is commonly applied to model the outcomes of a categorical dependent variable. It is inappropriate to use linear regression for categorical variables because the response values are not measured on a ratio scale and the error terms are not normally distributed. In addition, a categorical variable such as participation in the

financial market can only take on a limited number of discrete values within a specified range whereas the linear regression model can be generated as predicted values of any real number ranging from negative infinity to positive infinity.

Logistic regression predicts probabilities, thus one can fit it using likelihood. For each data-point, there is a vector of features, x_i , and an observed class y_i . The probability of that class, as discussed above, is p , if $y_i = 1$, or $(1-p)$ if $Y_i = 0$. The likelihood function for logistic regression is expressed as

$$L(\beta_0, \dots, \beta_7) = \prod_{i=1}^n P(X_i)^{y_i} [1 - p(x_i)]^{1-y_i} \quad (11)$$

Then, the log-likelihood function is given as

$$\text{Log}(L) = \sum_{i=1}^n y_i \log p(x_i) + (1 - y_i) \log [1 - p(x_i)] \quad (12)$$

$$= \sum_{i=1}^n -\log(1 + e^{(\beta_0 + x_i \beta_i)}) + \sum_{i=1}^n y_i (\beta_0 + x_i \beta_i) \quad (13)$$

To find the maximum likelihood (ML) estimates, we differentiate the log-likelihood with respect to the parameters and set the derivatives equal to zero to get the estimates. Since this equation is nonlinear in β , some special methods should be used in order to obtain the estimated parameters. The iteratively re-weighted least squares (IRLS) method can be applied to get the solutions. The ML estimator of β can be obtained by using the IRLS algorithm as follows:

$$\hat{\beta}_{MLE} = (X' \hat{W} X)^{-1} X' \hat{W} \hat{Z} \quad (14)$$

Where $\hat{W} = \text{diag} \{ \hat{p}_i (1 - \hat{p}_i) \}$

$$\hat{z}_i = \log(\hat{p}_i) + \frac{y_i - \hat{p}_i}{\hat{p}_i (1 - \hat{p}_i)}$$

This is the i^{th} element of the vector \hat{Z} . The hats in the equations show the iterative process.

5. EMPIRICAL RESULTS

Table 5 presents the results of the logit regression using stata. The regressand is the probability of financial market participation (*participate*), whereas

regressors are gender, age group (*age*), marital status (*marital*), level of education (*education*), monthly income (*income*), risk assessment (*risk*), and self-assessed financial knowledge (*knowledge*). It should be noted that when a binary outcome variable is modeled using logistic regression, it is assumed that the logit transformation of the outcome variable has a linear relationship with the predictor variables. The *Log likelihood* is the log likelihood of the final model. The value -234.8114 has no meaning in and of itself; rather, this number can be used to help compare nested models. The Number of obs is the number of observations that is used in the analysis. Since Stata uses a list-wise deletion by default, if there is a missing value for any variable in the logistic regression, the entire case will be excluded from the analysis as a result this number may be smaller than the total number of observations in the data set.

LR chi2(7) is the likelihood ratio (LR) chi-square test. This is minus two (i.e., -2) times the difference between the starting and ending log likelihood. The number in the parenthesis indicates the number of degrees of freedom. In this model, there are 7 predictors, so there are 7 degrees of freedom. *Prob > chi2* is the probability of obtaining the chi-square statistic given that the null hypothesis is true. In other words, this is the probability of obtaining the chi-square statistic, 99.53, if there is in fact no joint effect of the regressors on the regressand. It is the p-value, which is compared to a critical value, say 0.05 or 0.01 to determine if the overall model is statistically significant. In this case, the model is statistically significant because the p-value is 0.000. *Pseudo R²* is the pseudo R-squared. It should be noted that this statistic does not mean what R-square means in OLS regression.

The constant, in this model, is the expected value of the log-odds of financial market participation when all of the independent variables equal zero. In most cases, this is not interesting. Similar to linear regression, the standard error associated with the coefficient (*Std. Err.*) is used for testing whether the parameter is significantly different from 0. The standard errors can also be used to form a confidence interval for the parameter. *z and P > |z|* provide the z-value and 2-tailed p-value used in testing the null hypothesis that the coefficient (parameter) is 0. Coefficients having a p-value of 0.05 or less are statistically significant. That is, we can reject the null hypothesis and say that the coefficient is significantly different from 0. The *[95% Conf. Interval]* shows a 95% confidence interval for the coefficient. This is very useful as it helps us understand how high and how low the actual population value of the parameter might be. The confidence intervals are related to the p-values such that the coefficient will not be statistically significant if the confidence interval includes 0.

The estimates

$$\ln\left(\frac{p_i}{1-p_i}\right) = \hat{\beta}_0 + \hat{\beta}_1 x_{1i} + \hat{\beta}_2 x_{2i} + \hat{\beta}_3 x_{3i} + \hat{\beta}_4 x_{4i} + \hat{\beta}_5 x_{5i} + \hat{\beta}_6 x_{6i} + \hat{\beta}_7 x_{7i} \quad (12.34)$$

where the coefficients and variables are described in Table 5, tell us about the relationship between the independent variables and the dependent variable, where the dependent variable is on the logit scale. These estimates tell the amount of increase in the predicted log odds of financial market participation = 1 that would be predicted by a 1 unit increase in the predictor, holding all other predictors constant. It is worth noting that for the regressors which are not significant, the coefficients are not significantly different from 0, which should be taken into account when interpreting the coefficients. Table 6 reports the decisions on the significance of individual coefficients. Because these regression coefficients are in log-odds units, their interpretation is somewhat tricky, so they are often converted into odds ratios. This means that the coefficients in a simple logistic regression are in terms of the log odds, that is, the coefficient of gender of 0.6189526 implies that a one unit change in gender results in a 0.6189526 unit change in the log of the odds. Odds ratios are obtained by exponentiating the coefficient, or by using the logistic command. Odds ratio results are reported in Table 7.

Table 5
Logit Regression Analysis

Logit estimates				Number of obs	=	484
				LR chi2(7)	=	99.53
Log likelihood	=	-234.8114		Prob > chi2	=	0.0000
				Pseudo R ²	=	0.1749
<i>Participate</i>	<i>Coef.</i>	<i>Std. Err.</i>	<i>z</i>	<i>p > z </i>	<i>[95% Conf. Interval]</i>	
Gender	0.6189526	0.2474617	2.50	0.012	0.1339367	1.103969
Age	0.3084204	0.2729490	1.13	0.258	-0.226549	0.843390
Marital	1.0712810	0.2437708	4.39	0.000	0.5934991	1.549063
Education	-1.316985	0.2912257	-4.52	0.000	-1.887777	-0.74619
Income	0.5137532	0.2868568	1.79	0.073	-0.048475	1.075982
Risk	0.4306466	0.2679872	1.61	0.108	-0.094598	0.955892
Knowledge	1.0490520	0.3734711	2.81	0.005	0.3270617	1.781041
_cons	-2.673201	0.5800079	-4.61	0.000	-3.809996	-1.53641

Source: Authors' estimates

Table 6
Decisions on the Significance of Individual Coefficients

<i>Independent Variable</i>	<i>Coef</i>	<i>z value</i>	<i>P value</i>	
Gender (x_1)	$\beta_1 = 0.62$	2.50	0.01	Reject H_0 at 1 percent
Age group (x_2)	$\beta_2 = 0.31$	1.13	0.26	Accept H_0
Marital status (x_3)	$\beta_3 = 1.07$	4.39	0.00	Reject H_0 at 1 percent
Education level (x_4)	$\beta_4 = -0.1.31$	-4.52	0.00	Reject H_0 at 1 percent
Monthly income (x_5)	$\beta_5 = 0.51$	1.79	0.07	Reject H_0 at 10 percent
Risk assessment (x_6)	$\beta_6 = 0.43$	1.61	0.11	Reject H_0 at 10 percent
Financial knowledge (x_7)	$\beta_7 = 1.05$	2.81	0.01	Reject H_0 at 1 percent

$H_0 : \beta_i = 0$, i.e. individual coefficient is not different from zero

Source: Authors' estimates

Table 7
Logistic Regression: Odds Ratio

Logistic regression		Number of obs	=	484
		LR chi2(7)	=	99.53
Log likelihood = -234.8114		Prob > chi2	=	0.0000
		Pseudo R^2	=	0.1749

<i>participate</i>	<i>Odds Ratio</i>	<i>Std. Err.</i>	<i>z</i>	<i>p > z </i>	<i>[95% Conf. Interval]</i>	
gender	1.856982	0.4595318	2.50	0.012	1.143320	3.016112
age	1.361273	0.3715581	1.13	0.258	0.797279	2.324234
marital	2.919117	0.7115956	4.39	0.000	1.810312	4.707059
education	0.267942	0.0780316	-4.52	0.000	0.151408	0.474168
income	1.671553	0.4794966	1.79	0.073	0.952681	2.932872
risk	1.538252	0.4122319	1.61	0.108	0.909738	2.600989
knowledge	2.854942	1.0662380	2.81	0.005	1.373087	5.936035

Source: Authors' estimates

Odds of being participated to financial market are about 1.86 greater for males than females while odds of financial market participation for age group of 25+ years and married individuals are 1.36 and 2.92 times that of age group below 25 years and single/divorced participants respectively. Likewise, the odds of financial market participation for individuals with monthly income above TZS 300,000, and individuals with fair, or moderate, or advanced knowledge of financial market are 1.67 and 2.85 times that of individuals with monthly income below TZS 300,000, and individuals with poor knowledge of financial knowledge respectively. Moreover, the odds of being participated to financial market for risk taker are about 1.54 times

that of risk averse. Surprisingly, the odds of being participated to the financial market for individuals with education level above primary are 0.27 times that of individuals with primary or informal education. It should be noted that odds ratio can vary between 0 to positive infinity while log odds ratio varies between $[-\infty, \infty]$. Specifically, when odds ratio lies between $[0,1]$, log odds ratio is negative.

For more intuition, it is also possible to compute the marginal effect of an independent variable on the probability. With binary independent variables, marginal effects measure discrete change, i.e. how do predicted probabilities change as the binary independent variable changes from 0 to 1? The marginal effect for categorical variables therefore shows how $P(Y=1)$ changes as the categorical variable changes from 0 to 1, holding all other variables at their means. Results for marginal effects are reported in Table 8. In this case, the marginal effect for gender of 0.10 suggests that the predicted probability of financial market participation is 0.10 greater for male individuals than for female individuals while marginal effects for age and marital status of 0.05 and 0.17 imply that the predicted probability of financial market participation is 0.05 and 0.17 greater for age group of 25+ years and married individuals than for age group below 25 years and single or divorced individuals respectively. Marginal effect also suggests that the predicted probability of financial market participation is 1 lower for individuals with level of education above primary while the predicted probability of financial market participation for individuals with monthly

Table 8
Logistic Regression: Marginal Effects

Average marginal effects		Number of obs		= 484		
Model VCE	: OIM					
Expression	: Pr (participate), predict ()					
Dy/dx w.r.t	: Gender age marital education income risk knowledge					
	<i>Margin (dy /dx)</i>	<i>Delta- method Std. Err.</i>	<i>z</i>	<i>p > z </i>	<i>[95% Conf. Interval]</i>	
Gender	0.0974801	0.0381557	2.55	0.011	0.0226963	0.172264
Age	0.0485738	0.0428392	1.13	0.257	-0.035389	0.132537
Marital	0.1687183	0.0358943	4.70	0.000	0.0983668	0.239070
Education	-1.316985	0.0421874	-4.92	0.000	-0.290101	-0.12473
Income	0.0809121	0.0447616	1.81	0.071	-0.006819	0.168643
Risk	0.0678234	0.0418816	1.62	0.105	-0.014263	0.149910
Knowledge	0.1652173	0.0574702	2.87	0.004	0.0525778	0.277857

Source: Authors' estimates

income above TZS 300,000 is 0.08 greater than individuals with monthly income below TZS 300,000. Furthermore, the marginal effects show that the predicted probability of financial market participation for risk takers and individuals with financial market knowledge is 0.07 and 0.17 higher than for risk averse and individual with no knowledge of financial market respectively.

In general, our empirical findings show that the coefficients on financial knowledge, gender, marital status, and the level of education are statistically significant at 1 percent while the coefficients on income level and risk taking are statistically significant at 10 percent. The coefficient on age group is insignificant suggesting that in our data sample age does not play a role in predicting financial market participation of individuals.

It was hypothesized that literacy of financial market would positively influence the participation of individuals in the financial market. It is worth noting that financial literacy characterises an individual's knowledge of basic concepts and it has been identified in this paper to explain the variation in an individual's tendency to participate in the financial markets. This findings suggest that individuals who are financially literate have the capacity to make more informed choices regarding financial matters such as whether and how much to save. These results are consistent to many previous studies including Gerhard *et al.* (2018); Klapper *et al.* (2015); van Rooij (2011); Stango & Zinman (2009) and Lusardi *et al.*, (2007). Indeed, the largest effects in terms of the magnitudes of the marginal effects, reported in Table 8, come from the financial knowledge.

Similarly, it was presumed that the level of education would have a positive effect on the participation of individuals in the financial market. In fact the coefficient of the level of education is significant, however, the effect is counterintuitive and not consistent with some other empirical findings (for example van Rooij, 2011, García & Tessada, 2013, and Campbell, 2006) but consistent with other studies (for example Gumbo & Sandada, 2018). Consistent with the findings of this paper Gumbo & Sandada (2018) reveal that individuals with higher education levels participate less in comparison to individuals with lower education levels. This results suggest that education is not the main factor influencing financial market participation. It should be noted that recent work on the impact of financial education on downstream financial behaviour suggests that traditional forms of education training have only limited effectiveness and points to the need of providing just-in-time financial training through for example coaching (Fernandes *et al.*, 2014). Likewise, results suggest that, consistent with Bayer *et al.* (1996) & Bernheim *et al.* (2001), employer-based

financial education that increases individuals financial knowledge is likely to increase participation rate in the financial market. Financial literate is likely to reduce information costs to a level that makes the financial market more attractive to individuals.

Empirical results also show that individual's income plays a significant role in financial market participation. The positive impact of the level of income on financial market participation is consistent with theories. Theoretically, labor income allows households to plan their consumption and follow the traditional consumption smoothing theory where households that have excess income have the tendency to save or invest in order to maintain their same level of consumption in the future. Intuitively, the higher the income of an individual, the more likely the individual is to participate on the stock market. Moreover, logit regressions show that gender and marital status explain the variations in individuals' participation in the financial market. The positive coefficients on gender and marital status imply that if the individual is a man or married, the more likely the individual is to participate in the financial market consistent with Dillen & De Lille (2018) and Isaac (2014). This also suggests that women are more risk-averse compared to men. They are reluctant to participate in the stock market as stocks are considered risky assets. Finally, in our logit model, risk taking is positively correlated with financial market participation. This implicates that the more risk taker an individual is, the more likely the individual is to participate in the stock market.

6. CONCLUSIONS

The limited level of participation in the financial markets by individuals and households in the real world conflicts with theoretical understanding about the benefits of financial markets. At the same time, this limited participation can have important effects on the behaviour of financial markets. Our findings reveal that lack of financial literacy prevents households from participating in the financial market, even for individuals whose wealth, education, and financial literacy are high relative to the general population. Financial knowledge, as a measure of financial literacy, is positive and highly significant, suggesting that financial literacy has a positive effect on individual and households participation in the financial market. Indeed, the role of financial literacy in financial market participation cannot be overstated because the loss of welfare from non-participation in the financial market may be substantial. As individuals decide how much to save for future consumption and how to invest their income, it is important to consider ways to enhance their level of financial knowledge or to guide them in their financial decisions.

Surprisingly, results reveal that individuals with a primary and informal education are more likely to participate in the financial markets than their counterparts with a secondary and tertiary education suggesting that the educational attainment does not increase sharply with financial literacy. Financial literacy rates and participation in the financial market differ in important ways when it comes to characteristics such as gender, marital status, the level of income, and risk taking. Men, married individuals, and individuals with higher levels of income are more likely to participate in the financial market than women, single/divorced, and individuals with lower monthly income. Similarly, individuals with high risk taking attitude are more likely to participate in the financial market. However, age seems to exert no statistical influence on financial market participation.

These results shed additional light on the debates over financial literacy and individual characteristics, and provide policymakers and practitioners with new evidence linking financial literacy to financial behaviour. From an economic viewpoint, the fact that financial literacy and financial market participation are positively correlated, the lack of financial literacy acts as a barrier to wealth-maximizing investment, which ultimately impair economic development. Proper financial education, seminars, and financial orientation in the workplace can improve financial literacy and awareness, which, in turn, can improve investment in the financial market.

Several potential avenues are available for future research. First, the model does not account for a number of variables such social interaction, compound interest, inflation, and time value of money, which may be relevant factors determining financial market participation of individuals and households. Second, the paper applies logit model and maximum likelihood techniques of estimation. Further studies may apply a two-stage least squares (2SLS) specification. Third, the study focuses on the general financial market without specifying the primary or secondary market and also stock market or mutual funds. Further studies could assess any of these markets separately.

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