Logistic Regression Analysis Of Predictors Of Loan Defaults By Customers Of Non-Traditional Banks In Ghana

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Abstract

The objective of this research is to identify the risk factors that influence loan defaults by customers in the microfinance sector and to develop a model that links these factors to credit default by customers in the sector. Data from a microfinance institution based in Accra Ghana was used. A binomial logistic regression analysis was fitted to a data of 548 customers who were granted credit from January 2013 to December 2014. The results of the study revealed that six factors: X3 (Marital Status); X7 (Dependents); X11 (Type of Collateral or Security); X13(Assessment); X15 (Duration); and X16 (Loan Type) were statistically significant in the prediction of loan default payment with a predicted default rate of 86.67%. It is therefore suggested that microfinance institutions adopt among others, the default risk model to ascertain the level of risk since it's relatively efficient and cost effective. There should also be up to date training for loan officers of microfinance institutions in order to improve on their assessment skills and methodology. The supervising body of microfinance institutions (Bank of Ghana) should also consider enacting laws that will ensure that all such institutions in Ghana are roped into centralized database to check multiple borrowing and also serve as an internal control measure for the sustainability of these institutions.

Keywords: Loan Default, Repayment, Microfinance Institutions, Logistic Regression

Introduction

Introduction With almost thirty per cent of Ghanaians living below the poverty line, microfinance has been identified as an important means of providing financial services to the population. It is therefore not surprising that the country's present and past governments have perceived microfinance as central to achieving the greater goal of poverty reduction. Through microfinance, the various governments have aimed to provide the poor, who do not have access to the formal financial sector, with greater access to customized financial services. Robinson (2001) defined microfinance as small-scale financial services provided to people into petty businesses in both rural and urban communities. He further indicated that microfinance institutions are primarily actablished to provide business and consumer credit

small-scale financial services provided to people into petty businesses in both rural and urban communities. He further indicated that microfinance institutions are primarily established to provide business and consumer credit or bring financial services to poor and vulnerable groups in society with the ultimate goal of improving living standards or eradicating poverty. Micro financial institutions have had immense effects on real economic activity augmenting the operations of traditional financial houses and banks. Crockett (1996) asserts that disruption to loan supply might cause great changes in economic activity. The dynamics of economics is such that once interest rate begins to decline it becomes attractive to borrow from the banks for trading purposes as the returns will far outweigh the interest rate payments. Both sole traders and corporate bodies are encouraged to borrow from the bank. However should interests rise astronomically, people are rather encourage to save in the banks or buy fixed interest bonds whose returns are higher as compared to borrowing for capital investment. Despite the risks associated with giving out loans, financial institutions are increasingly expanding their area of services. Currently there are institutions that employ full time personnel with sole duty of selling loans. They move from office to office appealing to the several public/private customers to subscribe to different loan packages. As a result of this, some private entrepreneurs have been able to expand their businesses and earn substantial profit. This process has ripple effect on the entire economy. If individuals establish businesses, they employ other people to help them in the day to day operations. According to Fama (1980) the banking institutions have advance loans to a wide category of customers. While some is given directly to corporate institutions, large portions to access the credit worthiness of the customers. This area is very critical if the banks are to be able to recover their loans when they are

defined as the inability of a borrower to fulfill his or her loan obligation when due. High default rates in small and medium enterprises (SMEs) lending should be of major concern to policy makers in developing countries, because of its unintended negative impacts on SMEs financing. Microfinance institutions all over the world are faced with the challenge of loan default/delinquency. This situation has been so alarming to the extent that some financial institution have adopted some unconventional means of retrieving these loans. The sustainability of microfinance institutions depends largely on their ability to collect their loans as efficiently and effectively as possible. In other words to be financially viable or sustainable, microfinance institutions must ensure high portfolio quality based on 100% repayment, or at worst low delinquency/default, cost recovery and efficient lending. lending.

Objective(s) of the Study According to Boateng and Ampratwum (2011) the need for micro-credit today is at its highest owing to the structure of the Ghanaian economy where majority (80%) of the workforce is in the informal sector. The industry is considered inherently risky mostly due to the sector it serves. In most Microfinance Institutions (MFIs) the methods of deciding whether to grant a loan to an individual are usually mostly speculative and mostly based on the experience of previous decisions other than objective analysis. Over the years there has been a debate as to which method works best. The consensus among the experts is that no one method stands out the choice is consensus among the experts is that no one method stands out, the choice is independent on other factors such as economic stability and the effectiveness and the dependability of the national database. The aim of this study is, therefore to determine the risk factors that influence loan default of customers in the microfinance sector and to develop model that links these factors to credit default for any customer in the sector.

Theoretical framework The Concept of Microfinance

The Concept of Microfinance Microfinance is defined by Steel and Andah (2002) as small financial transactions with low income households and micro enterprises using non-standard methodologies like character-based lending, group guarantee and short term repeated loans. Microfinance is provided by formal, semi-formal and informal financial agents. According to Otero (1994), microfinance creates access to productive capital for the poor, which together with human capital, addressed through education and training, and social capital, achieved through local organization building, enables people to move out of poverty. By providing material capital to a poor person, their sense of dignity is strengthened and this can help to empower them to participate in the

economy and society (Otero, 1994). Credit granting non-governmental organizations, credit cooperatives and to some extent, a few rural banks have utilized microfinance as a sustainable mechanism to provide basic financial services to small-scale borrowers (Llanto, 2001). The wide network of low-income clients of microfinance institutions proves that there is a great demand for credit by the poor and that they can successfully use these small loans to earn income (Llanto, 2001). The institutions involved in microfinance are either formal, semi-formal or informal. The formal microfinance institutions are larger businesses licensed to do financial intermediation through deposit mobilization and credit giving. They include the rural banks and the savings and loans companies. The semi-formal microcredit institutions include the non-governmental institutions involved in giving microcredit to the poorer members of the community. These include, susu firms, credit unions, and the like.

The underlying business principles and characteristics of these institutions influence their client bases and affect the repayment of the facilities they grant. The informal financial intermediaries have a cost advantage over the formal intermediaries in addressing the information asymmetry problem because they have access to devices and mechanisms to collect the necessary information about their clients. They frequently have more detailed knowledge of clients and their communities, and the local conditions in which they operate. As a result, they often face lower transaction cost in their service delivery. According to Aleem (1990), informal lenders mainly use the established relationship with borrowers as a screening and credit rationing mechanism. The formal intermediaries on the other hand enjoy economy of scale and can mobilize large amounts of deposits for lending. Institutions have typically been burdened with severe agency problems in dealing with peculiar risks, that is, the problems caused by costly and imperfect information such as adverse selection, moral hazard, and contract enforcement (Nissanke & Aryeetey, 1998). Karlan and Zinman (2004) used a randomized intervention to identify the extent of adverse selection and moral hazard in a South African credit

Karlan and Zinman (2004) used a randomized intervention to identify the extent of adverse selection and moral hazard in a South African credit market. They found out that about 40% of defaults in this market can be attributed to asymmetric information. Meyer and Nagarajan (2000) recommend that formal financial institutions must design their products and services according to the expected demand in rural areas, taking into consideration the presence of informal credit sources, and according to how costs could be recovered and profits could be generated. Such would then lead to a widened outreach. Essel (1996) has also noted that group-lending schemes (informal methodology) have been gradually gaining ground among the Rural Banks in Ghana and that the benefits reaped by the banks and their customers to date have been significant. He concludes by arguing that the banks should intensify efforts to establish group-based lending programmes.

Loan Default Repayment

Loan repayment default is the oldest and most prominent form of risk in the financial markets. If credit can be defined as "nothing but the expectation of a sum of money within some limited time", then loan repayment default is the chance that this expectation will not be met. Caouette, Altman, & Narayanan (1998) states that loan repayment default is as old as lending itself, which means that it dates back at least as far as 1800 BC.

A loan is delinquent when a payment is late (CGAP, 1999). A delinquent loan becomes a defaulted loan when the chance of recovery becomes minimal. Delinquency is measured because it indicates an increased risk of loss, warnings of operational problems, and may help to predict how much of the portfolio will eventually be lost because it never gets repaid. There are three broad types of delinquency indicators: collection rates, which measures amounts actually paid against amounts that have fallen due; arrears rates, measures overdue amounts against total loan amounts; and portfolio at risk rates which measures the outstanding balance of loans that are not being paid on time against the outstanding balance of total loans (CGAP, 1999). Default occurs when a debtor has not met his or her legal obligations according to the debt contract. For example a debtor has not made a scheduled payment, or has violated a loan covenant (condition) of the debt contract (Ameyaw-Amankwah, 2011). A default is the failure to pay back a loan. Default may occur if the debtor is either unwilling or unable to pay their debt. A loan default occurs when the borrower does not make required payments or in some other way does not comply with the terms of a loan

Default occurs when a debtor has not met his or her legal obligations according to the debt contract. For example a debtor has not made a scheduled payment, or has violated a loan covenant (condition) of the debt contract (Ameyaw-Amankwah, 2011). A default is the failure to pay back a loan. Default may occur if the debtor is either unwilling or unable to pay their debt. A loan default occurs when the borrower does not make required payments or in some other way does not comply with the terms of a loan (Murray, 2011). Moreover, Pearson and Greef (2006) defined default as a risk threshold that describes the point in the borrower's repayment history where he or she missed at least three installments within a 24 month period. This represents a point in time and indicator of behaviour, wherein there is a demonstrable increase in the risk that the borrower eventually will truly default, by ceasing all repayments. The definition is consistent with international standards, and was necessary because consistent analysis required a common definition. This definition does not mean that the borrower had entirely stopped paying the loan and therefore been referred to collection or legal processes; or from an accounting perspective that the loan had been classified as bad or doubtful, or actually written-off. Loan default can be defined as the inability of a borrower to fulfill his or her loan obligation as at when due (Balogun and Alimi, 1990).

Methodology

The design of the study is strictly quantitative. The study makes use of quantitative data; a historical data that contains the profile of customers of a microfinance institution was collected. Due to company policy, the name of the microfinance company was restricted and will be referred to in this paper as the MFI.

The target population included all customers of microfinance institutions in Ghana. However, Out of five institutions contacted, only one made available files and loan profiles for analysis. For many of the institutions contacted either their operational data was not adequately captured to fit the purpose of the study or the required data were unavailable. The analysis was thus restricted to one data source. The data set comprises of all individual loans disbursed from January 2013 to December 2014.

Given the sampling frame all the files of loans disbursed under the period of study were grouped into three; on time, not on time, default. On time: clients who were able to discharge their obligation within the time given for payment. Not on time: clients finished payment within one month after contract expired. Default: clients who have neither finished payment nor finished payment one month beyond the contract expiry period. The profiles of such clients; personal and loan characteristics were then captured with an SPSS database. Only those whose contracts were within the study period were captured.

Definition of Variables

Below is the predictor variables were considered for this study.

Xl = AgeX2 = GenderX3= Marital Status X4= Type of Business X5 = Residential Status X6 = Number of years at residence X7= Dependents X8= Purpose of loan X9 = Amount disbursed *X10*=Date of disbursement *X11*=Type of collateral (security) X12=Guarantor X13=Assessment X14=Officer X15=Duration X16=Loan Type

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Dependent variable construction and pre-processing

The data preparation step deals with the choice and creation of the desired variables dependent and covariates. In this study a binary dependent variable default (Y) was created.

 $Y_i = \begin{cases} 1 \text{ default (if) was created.} \\ 1 \text{ defaulted; if delayed payment is } > 30 \text{ days} \\ 0 \text{ non } - \text{ defaulted; if delayed payment } \le 30 \text{ days} \\ \text{This criterion is consistent with general definition of loan default.} \end{cases}$ Since including all variables will make the model unnecessarily large, the principle of parsimony will justify small model. The researcher employed statistical procedures such as forward and backward selection processes to verify consistency of variables selected in the model.

Analytical Tools

The study makes use of the logistic regression model. Logistic regression is based on binomial probability theory. It is a mathematical modeling approach used in describing the relationship of several independent variables to a dichotomous dependent variable or a limited dependent variable. The logit function is employed because the dependent variable "default" is dichotomous, whereas the proposed covariates were mixture of continuous and categorical random variables. Thus the model was chosen over others due to the data structure and purpose. Also the independent variables need not be interval, nor normally distributed, nor linearly related, nor equal variance within each group. The logit model is a derivative of the odds function. The odd of a function is the ratio of the probability of success to that of failure. Thus

$$Odds(Y = 1) = \frac{P(Y = 1/X = x)}{P(Y = 0/X = x)}$$

Where Odds(Y = 1) is the odds of defaul t; P(Y = 1) is the probability that default occurs given a set of explanatory variables and P(Y = 0) is the probability of non-default given set of explanatory variables.

If the odds of default is greater than one it means there is a higher probability of default compared to that of non-default. A value less than one indicate a higher probability of non-default than that of default.

Given the binary response variable (default or non-default), the probability distribution of the number of defaults in a given loan portfolio size, for given values of explanatory variables is binomial. Thus the probability that the number of default of a given portfolio size \mathbf{n} is exactly equal to size x is given by

$$P(X = x) = \frac{n!}{x! (n - x)!} p^{x} q^{n - x}$$

Where q = P y(0) =; probability of non-default). This means that given a portfolio size n and probability of default P(Y=1) from a financial institution, one can use the theorem to perform risk control analysis.

Justification of Methodology

The authors adopted the logistic regression analysis in the study in order to predict an outcome variable that is categorical from predictor variables that are categorical. This method was also used because having a categorical outcome variable violates the assumption of linearity in normal regression; since logistic regression deals with this problem by using a logarithmic transformation on the outcome variable which allows us to model a nonlinear association in a linear way. This methodology also expresses the linear regression equation in logarithmic terms.

Results and discussions

Table 1: Demographic Information of Survey Participants (n=548)							
Variables	Frequency	Percentages					
Gender							
Male	195	35.6					
Female	353	64.4					
Loan Repayment Status							
Paid	GHØ 2,021,984.9	74					
Arrears	GHØ 710,427.12	26					

In all there are 548 clients within the period under study. Table 1 summarizes the socio-demographic information of the clients. From the table, 195 respondents which represent 35.6% were males and 353 of them which represent 64.6% were female. Concerning the loan repayment status, in all as at December 2014, the company had disbursed GH 2,732,412. Out of this GH 2,021,984.9 which represents 74% had been paid as at the end of December, 2014 whilst the rest GH 710,427.12 representing 26% of the portfolio value was in arrears.



Figure 1: A Bar Graph Showing the Sum of Money Disbursed Against Type of Loan

Figure 1 above shows the distribution of the amount of money disbursed to customers by the type of loan accessed. It can be seen that among the type of loans, commercial loan was given the highest total amount approved to customers which is $GH \not C$ 917,000; followed by personal loan which also recorded $GH \not C$ 791,001. Transport loan was the third highest with a total amount approved as $GH \not C$ 623,412. Finally, agricultural loan recorded the least amount approved which was $GH \not C$ 401,000.

Count		Default Status		
		No	Yes	
I con tours	Agricultural	47	24	71
Loan type	Commercial	68	35	103
	Personal	196	71	277
	Transport	65	32	97
Total		376	172	548
Chi-Square Tes	t Value	df	Asymp. Sig. (2-sided)	
Pearson Chi-Squa	are 16.02	3	0.01	
N of Valid Case	s 548			

Table 2: Independence of Type of Loan versus Repayment Status

The count part of the table shows that, about 31.39% of the customers end up defaulting in the repayment of loans depending on the type of loan granted. The chi-square test of independence part also shows that the test is significant since the asymptotic significance of 0.01 is less than 0.05; and so the ability of a customer to default or otherwise of a loan depends on the Loan Type applied for.

	D	¢Е	Wold	Af	Sia	$E_{vm}(\mathbf{D})$	95.0% C.I	I.for EXP(B)
	В	5.E. wa	w alu	vald di	Sig.	Exp(B)	Lower	Upper
X1	0.171	0.218	0.616	1	0.433	1.187	0.774	1.821
X2	-0.274	0.170	2.601	1	0.107	0.760	0.545	1.061
X3	-0.843	0.200	17.693	1	0.000	0.430	0.291	0.638
X4	-0.238	0.266	0.801	1	0.371	0.788	0.468	1.327
X5	0.008	0.167	0.002	1	0.963	1.008	0.727	1.398
X6	0.293	0.223	1.727	1	0.189	1.341	0.866	2.076
X7	0.952	0.214	19.861	1	0.000	2.591	1.704	3.937
X8	-0.256	0.211	1.469	1	0.226	0.774	0.512	1.171
X9	-0.384	0.267	2.072	1	0.150	0.681	0.404	1.149
X10	0.149	0.270	0.307	1	0.580	1.161	0.685	1.969
X11	0.871	0.151	1.652	1	0.000	2.389	0.613	1.107
X12	0.146	0.140	1.096	1	0.295	1.158	0.880	1.523
X13	0.678	0.298	5.174	1	0.023	1.971	0.649	1.657
X14	0.036	0.239	0.023	1	0.880	1.037	1.098	3.535
X15	0.737	0.242	9.293	1	0.002	2.091	1.301	3.359
X16	-0.818	0.190	18.606	1	0.000	0.441	0.304	0.640
Constant	0.294	1.688	0.030	1	0.862	1.342		

Table 3: Logistics Regression Estimates of Factors Influencing Default Payment

Table 3 shows the result of logistic regression estimates of the various factors influencing loan payment defaults. The significant value of the Wald statistics for each independent variable indicates the contribution or importance of each predictor variable (P<0.05).

From the table, column six (6) determines the variables that contribute significantly to the predictive ability of the model at 0.05 level of significance. These variables are, X3 (Marital Status), X7 (Dependents), X11 (Type of collateral or security), X13(Assessment), X15 (Duration), and X16 (Loan type). This findings support the research results of Dinh and Kleimeier (2007) which indicated that the collateral value could also be a proxy for the borrowers' financial wealth since it is significantly positive correlated with the borrowers's income. It is also consistent with the observations of Dinh and Kleimeir (2007) who intimated that, marital status affects the borrower's level of responsibility, reliability, or maturity. The probability of default is higher for the married than single borrowers. They discover that the marital status is typically related to number of dependents which in turn reflects

financial pressure on the borrower and borrower's ability to repay a loan. Thus the logistic function is given by the equation (2) below:

 $P(Default) = \frac{1}{1 + e^{-(.294 - 0.843X3 + 0.952X7 + 0.871X11 + 0.678X13 + 0.737X15 - 0.818X16)}}$

Furthermore, the odd ratio $(Exp(\beta))$ for the significant factors, shows the increase (or decrease if the ratio is less than one) in odds of being in one outcome category (default or no default) when the value of the predictor increases by one unit. From table 2, the odds or risk of customer defaulting in repayment of loan, is 0.430 for X3 (Marital Status). This indicates that, the risk of a customer defaulting in a loan repayment is 0.430 times higher for a customer who is married and with children than for customers who are single, all other factors being equal. For X7 (Dependents), the odd ratio of 2.591 indicates that risk of customer with larger number of dependents, is 2.591 times more likely to default in repayment as compared with those with least number of dependents, all other factors being equal. For X11 (Type of collateral or security), the odd ratio of 2.389 indicates that the risk of a customer defaulting in repayment of loans is 2.389 times higher for a customer defaulting in repayment of loans is 2.389 times higher for a customer defaulting in repayment of loans is 2.389 times higher for a customer defaulting in repayment of loans is 2.389 times higher for a customer who used collateral than for a customer who used personal guarantee, all other factors being equal.

Also for X13 (Assessment) the odd ratio is 1.971 which means that for any wrong assessment of clients before loan are granted to them, the risk of defaulting increases by 1.971, all other factors being equal. Furthermore, for X15 (Duration), the odd ratio of 2.091 indicates that the risk of a customer defaulting in loan repayment is 2.091 times higher for a customer who has been given shorter duration to pay back than those who has been given longer duration, all other factors being equal. Finally, the odd ratio of 0.441 for X16 (Loan type) indicates that, for any type of loan granted, the risk of defaulting decreases by a factor of 0.441, all other factors being equal.

Conclusion

The study revealed that six (6) factors; X3 (Marital Status), X7 (Dependents), X11 (Type of collateral or security), X13(Assessment), X15 (Duration), and X16 (Loan type) were statistically significant in the prediction of loan default payment with a predicted default rate of 86.67%. This indicates that there is probability that 86.67% of customers, with the given characteristics are likely to default their loan repayment. Therefore, there is a need for Micro-financial Institutions to adopt

Therefore, there is a need for Micro-financial Institutions to adopt among others the default risk model to ascertain the level of risk since it's relatively efficient and cost effective. There should also be adequate training for Loan officers in the MFI in order to improve on their assessment skills and methodology. Finally, Bank of Ghana should consider enacting a law that will ensure that all MFIs in Ghana are put on a centralized database to check multiple borrowing and also serve as an internal control measure for sustainable MFIs in Ghana.

Limitations and recommendations

Limitations and recommendations An obvious limitation that had a significant impact on the results obtained from this study is the fact that only one organization was sampled from an industry of 468 players (microfinance institutions: https://www.bog.gov.gh; retrieved on January 4, 2016). This is coupled with the fact that very little information in terms of organized literature exists in Ghana on the operations of these non traditional banking institutions. It is therefore recommended that other researchers can delve deeper into this very important research area in order to scoop out further facts about loan defaults in the microfinance industry in Ghana.

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