ESTIMATION OF A CREDIT SCORING MODEL FOR LENDERS COMPANY

Felipe Alonso Arias Tutor: Juan Sebastián Bravo Co-tutor: Francisco Iván Zuluaga

Proposal Presentation Research Practice III

Mathematical Sciences Department, EAFIT University Linkvest Capital LLC, Miami, Fl August 28 2015



The Company

- Equity Firm
- Different Branches
- Mortgage Loans
- Midtown \neq Linkvest
- Loans Money
- High interest







Brief Explanation

- 1. People ask for a loan.
- 2. Linkvest analyze the background.
- 3. Make decisions based on experience.
- 4. Lend money.
- 5. Faith.



Justification

- Influent variables.
- Banks have their own models.
- Probability of default.
- Positive cash flows.



Three Big Problems

- Linkvest uses a loan application, but ...
- Decisions based on previous situations.
- How to know if the person is going to pay?





 Estimate a Credit Score Model to improve lending decisions on a person using historic data.



Specific Objective

- Analyze the best model to improve in our case based in the data.
- Identify the most influential variables for the company when lending money.
- Specify the model in mathematical terms.
- Estimate the probabilities of default.
- Analyze the results for decision making.



Schedule

Week	Task
4-5	Proposal Report
6-7	Specification of the model
8-9	Estimation of the model
10	Progress Presentation
11-14	More Estimations
15-16	Final Report
17	Extra Improvements
18	Final Presentation



Minimum Scope

The main goal is to estimate a credit score model to estimate the probability of default on personal loans using macroeconomic and personal data.

It is important to evaluate the model with real data and take into account the basic assumptions of the model chosen.



Literature Review

- Credit Scoring (Cantón, Rubio, y Blasco, 2010) (Hand y Henley, 1997)
- Discriminant Analysis (Altman, 1968).
- Linear Probability Models (Orgler, 1970).
- Logit Models (Wiginton, 1980).
- ▶ Neural Networks (Rosenberg y Gleit, 1994).



Methodology

The logistic regression model can be formulated as:

$$Z = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_k X_k + \mu$$

where μ is the disturbance and p is the probability of default and can be estimated as follows:

$$p = \frac{e^z}{1+e^z} = \frac{1}{1+e^z}$$

Then, our response is going to be

$$Default = \log\left(rac{p}{1-p}
ight)$$



Resources

- Altman, E. I. (1968, September). Financial Ratios, Discriminant Analysis and the Prediction of Corporate Bakruptcy. *The Journal of Finance*, XXIII(4), 589-609.
- Cantón, S. R., Rubio, J. L., y Blasco, D. C. (2010, June). A Credit Scoring Model for Institutions of Microfinance under the Basel II Normative. *Journal of Economics, Finance and Administrative Science*, 15(28).
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- Rosenberg, E., y Gleit, A. (1994, August). Quantitative Methods in Credit Management: A Survey. Journal of Operations Research, 42(4), 589-613.
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Thanks For Attending!

