

Chapter 3.20

Bankruptcy Prediction through Artificial Intelligence

Y. Goletsis

University of Ioannina, Greece

C. Papaloukas

University of Ioannina, Greece

Th. Exarhos

University of Ioannina, Greece

C. D. Katsis

University of Ioannina, Greece

INTRODUCTION

Bankruptcy prediction or corporate failure is considered a classic issue in both, academic and business communities. Bankruptcy risk is one of the most important factors (if not the most important one) to be considered when credit requests are screened or even existing debtors are evaluated. On the other hand, all potential stakeholders (shareholders, suppliers, customers, employees, creditors, auditors, etc.) have potential interest to identify if a company is on a trajectory that is tending towards failure. Commercial banks, public accounting firms and other institutional entities (e.g., bond rating agencies) appear to be the primary beneficiaries of accurate bankruptcy

prediction, since they can use research results to minimize exposure to potential client failures. In addition to avoiding potentially troubled obligors, the research can also benefit in other ways. It can help in accurately assessing the credit risk of bank loan portfolios. Credit risk has been the subject of much research activity, since the regulators are acknowledging the need and are urging the banks to assess the credit risk in their portfolios. Measuring the credit risk accurately also allows banks to engineer future lending transactions, so as to achieve targeted return/risk characteristics. The other benefit of the prediction of bankruptcies is for accounting firms. If an accounting firm audits a potentially troubled firm, and misses giving a warning signal then it faces costly lawsuits (Atiya, 2001).

DOI: 10.4018/978-1-60960-818-7.ch3.20

A series of techniques have been applied in literature. Econometric / statistical methods have first appeared in literature: In late 1960's (multiple) discriminant analysis (DA) was the dominant method; during the 1980's logistic analysis. In the 1990's artificial intelligence starts appearing in financial literature with neural networks (Odom & Sharda 1990) serving as an alternative to statistical methods demonstrating promising results.

The goal of this chapter is therefore two-fold: First, it intends to give an overview of the artificial intelligence techniques successfully applied to the problem, ranging from the first neural network applications to recent applications of biologically inspired algorithms, such as genetic algorithms. Then, two kernel based methods, namely the Radial Basis Function Neural Networks and the Support Vector Machines are applied to the bankruptcy problem.

BACKGROUND

Early statistical studies in bankruptcy prediction (e.g., Beaver, 1966 adopted a univariate methodology identifying the accounting ratios having the highest classification accuracy in separating failing and non-failing firms. Beaver investigated the predictability of 14 financial ratios. Altman (1968) examined simultaneously a series of financial ratios, enriching the single ratio approaches. A multiple discriminant function was calculated, the so-called Z-score composed of five financial ratios:

$$Z = 1.2X_1 + 1.4X_2 + 3.3X_3 + .6X_4 + .999X_5, \quad (1)$$

where

X_1 = Working Capital / Total Assets. (Measures liquidity)

X_2 = Retained Earnings / Total Assets. (Measures profitability)

X_3 = Earnings Before Interest and Taxes / Total Assets. (Measures operating efficiency)

X_4 = Market Value of Equity / Book Value of Total Liabilities. (Adds market dimension)

X_5 = Sales / Total Assets. (Standard measure for turnover)

Z-Score model was modified by Altman, Haldeman, and Narayanan (1977). Their ZETA model was composed from seven financial ratios. Since these early studies, a vast range of statistical methodologies have been applied for the purposes of corporate failure prediction including logistic regression (Martin, 1977), logit (Ohlson, 1980), Kolari, Glennon, Shin, and Caputo (2002), probit and maximum likelihood models (Zmijewski, 1984).

Literature review reveals that a series of financial ratios has been examined. Stability indicators, industry-specific indicators, macro-economic factors, firm's particular features have been examined. However, there is no agreement on the features that carry significant predictive power. According to Courtis (1978) and Dimitras (1995) the applied ratios should adequately cover three fields: profitability, management efficiency, and solvency.

AI FOR BANKRUPTCY PREDICTION

The statistical methods described above have some restrictive assumptions such as linearity, normality and independence among predictor or input variables. Considering that violation of these assumptions for independent variables frequently occurs with financial data (Deakin, 1976) these methods have limitations to obtain effectiveness and validity. Artificial intelligence (AI) methods have been proven to be less vulnerable to these assumptions.

In the following paragraphs we present a brief description of AI techniques employed and a short review of research attempts applying AI

8 more pages are available in the full version of this document, which may be purchased using the "Add to Cart" button on the publisher's webpage:

www.igi-global.com/chapter/bankruptcy-prediction-through-artificial-intelligence/56170

Related Content

R4 Model for Case-Based Reasoning and Its Application for Software Fault Prediction

Ekbal Rashid (2016). *International Journal of Software Science and Computational Intelligence* (pp. 19-38).

www.irma-international.org/article/r4-model-for-case-based-reasoning-and-its-application-for-software-fault-prediction/172125

Combining Ontology with Intelligent Agent to Provide Negotiation Service

Qiumei Pu, Yongcun Cao, Xiuqin Pan, Siyao Fuand Zengguang Hou (2010). *International Journal of Software Science and Computational Intelligence* (pp. 52-61).

www.irma-international.org/article/combining-ontology-intelligent-agent-provide/46146

A Novel Deep Federated Learning-Based Model to Enhance Privacy in Critical Infrastructure Systems

Akash Sharma, Sunil K. Singh, Anureet Chhabra, Sudhakar Kumar, Varsha Aryaand Massoud Moslehpour (2023). *International Journal of Software Science and Computational Intelligence* (pp. 1-23).

www.irma-international.org/article/a-novel-deep-federated-learning-based-model-to-enhance-privacy-in-critical-infrastructure-systems/334711

Intersection of Intellectual Property, Medicines, and Legal Concerns

Manvendra Singh, Vaishali Aroraand Kushagra Kulshreshta (2024). *Intersections of Law and Computational Intelligence in Health Governance* (pp. 117-141).

www.irma-international.org/chapter/intersection-of-intellectual-property-medicines-and-legal-concerns/354868

Modeling Agent Interactions using Common Ground Knowledge from a Joint Activity Theory Perspective

Divesh Lalaand Toyooki Nishida (2013). *International Journal of Software Science and Computational Intelligence* (pp. 1-19).

www.irma-international.org/article/modeling-agent-interactions-using-common-ground-knowledge-from-a-joint-activity-theory-perspective/108927