METHODICAL APPROACH TO ESTIMATION OF CORPORATE BONDS ISSUER DEFAULT RISK

O. Slutska

The article is devoted to the development of methodical approach to corporate bonds issuers default risk estimation. Fuzzy logic rules have been laid down generated to determine default risk level, focusing on four financial indicator values and of market interest rate level. In addition, probabilities of company transition to another risk level and bonds issuer default after a certain time period have been calculated.

Key words: corporate bonds, issue, default risk, fuzzy logic, the classification tree, Markov chain.

Due to the crisis in the Ukrainian economy in 2008 – 2010, there was a significant deterioration in corporate bonds issuers’ creditworthiness. This resulted in lower liquidity and significantly reduced amounts of primary and secondary market debt securities. One of the main reasons for this status was lack of attention, and sometimes ignorance, of risk management procedures by a number of borrowers and investors. Given this Ukrainian corporate bonds market restoration requires appropriate

issuer’s default risk assessment methods development. That will enable to assess properly corporate borrower reliability.

Research in bond issuers’ default probability estimation has been conducted by worldwide renowned scholars: E. Altman, F. Black, D. Duffy, R. Jerome, R. Merton, A. Peresetsky, R. Singleton, S. Turnbull, F. Fabozzi, W. Hickman, M. Scholes etc. Since the Ukrainian bond market is characterized by slightly different dimensions, existing models cannot be correctly applied to assess local issuers default risk, because they are based on data received from overseas financial markets. Problem of debt securities

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issuers reliability evaluation has also been studied by Ukrainian scientists: I. Britchenko, G. Velykoivanenko, V. Vitlinski, L. Dolinski, A. Kaminski, B. Kyshakevych etc. However, the problem of corporate bonds issuer’s default risk requires more detailed research and further development.

Risk assessment science-based methods development for corporate bonds and in particular issuer default risk as one of the most dangerous ones will heighten the interest of potential investors in these financial instruments. The tasks to be solved are based on the real defaults history in the domestic bond market in the past, with account taken of Ukrainian debt securities issuers peculiar features.

Bond issuers’ default risk assessment must include decision-making mechanism that will predict the risk. Each of credit risk default is unique. Relating to previous research finding [1], the most appropriate approach in terms of Ukrainian economy conditions implies the use of bond issuers financial statements. Higher quality results will be reached with account taken of macroeconomic indicators in the model, which determine default risk systematic component.

The correlation ratio has been calculated in order to analyse the relation between the observed number of defaults and macroeconomic indicators as well as determine the key exogenous indicators to forecast borrowers’ default [2]. The result has revealed the most significant correlation to be observed between the quantity of defaults and bank interest rates.

Subsequently the factor analysis and the "gravity centre" method of were used to select basic financial indices for corporate bonds issuer’s default risk estimation [3]. Financial leverage ratio, liquidity ratio, EBITDA interest coverage and EBITDA to assets ratio have been determined as the most significant financial indicators in determining a company’s financial position. In order to determine the issuer’s financial position classification criteria the three-level scale scores for previously selected parameters assessment have been elaborated [4]. The assessment is due to be conducted basing on the fuzzy logic tools.

The database of bond issuers, classified as reliable and unreliable ones in terms of their debt obligations, has been formed in order to identify certain relations between selected financial and macroeconomic indicators and the fact of a borrower’s default in the past. Selected bond issuers’ financial ratios that were outstanding within 2007 – 2011 have been transferred into fuzzy scale to form the initial database for further research. Each credit redemption event characterizes the issuer’s reliability, or the last default (coupon payments violation and principal debt repayment, or moratorium on obligations repayment), which characterizes the borrower’s unreliability have been considered as the research object. As a result, the sample of 112 objects, comprising of 69 event commitments and 43 defaults has been received.

A combination of StatSoft Statistica® Data Mining classification tree methods have been used to solve the bond issuers distribution problem in terms of default risk with regard to the selected ratios. Firstly, the Boosted Trees method was used. The basic principle of this method is to build a sequence of very simple binary trees [5]. As a result, 86 % of actually unreliable issuers, i.e. those who defaulted, have been classified correctly. As for reliable borrowers, the percentage of correctly classified ones was somewhat lower and amounted to almost 58 %.

Thus, the use of the boosted trees for company classification has resulted in failure to distribute bond issuers into two classes according to default risk. An intermediate class of issuers has been established taking into account a certain degree of belonging to reliable or unreliable issuers’ classes. Bond issuers have been divided into three classes according to risk, namely low, medium and high classes, on the basis of borrowers’ default probability represented using the boosted trees classification method. Furthermore, the CART method has been used to determine the classification rules for issuers’ default risk. As a result, the classification tree structure represented in Fig. 1, has been obtained.

![Fig. 1. Bond issuers classification at default risk](image)

The comparison of the actual credit events and projected default risk levels identified by the classification tree (Fig. 1) enabled to detect minor differences. Thus, only three issuers that were classified at low forecast default risk levels actually violated its debt obligation terms. However, it ought to be noted that the violations were recorded within technical default, because the obligations on the bonds were still being fulfilled, which confirmed the appropriate issuers financial condition level. Late payments in these cases were due to other reasons rather than borrower’s financial inability, which is difficult to predict.

A more detailed analysis of eight issuers, who eventually fulfilled their obligations on the bonds, but had a high estimated risk level, led to the following summaries. Four issuers are subsidiaries of foreign or domestic financial groups. As a result, it is important for the mentioned debt issuers to maintain high reputation and positive credit history. Therefore, most likely, even with poor borrowing enterprises’ financial state, parent companies did not permit to default on other group members bonds. As for other issuers with high default risk, which had still fulfilled their obligations, it ought to be noted that they made a breach in other payment periods on those or other series of bonds, and some of them were eventually declared bankrupt. All the above confirms high predicted borrower’s default risk and possibility of their debt failure.
Subsequently, certain future default risk evaluation rules, represented as fuzzy linguistic variables shown in Tab. 1, have been formed on the basis of generalized bond issuers’ distribution with regard to classification trees with fuzzy financial performance scales. Note that each rule weight reflects its degree of importance and adequacy. In order to determine optimal weight values for each of the proposed rules neural-fuzzy networks method with back propagation algorithm [6], which is implemented in the Matlab® ANFIS editor has been applied. Following up the procedure described, of model parameters optimization has been performed. Initial weight values for each rule have remained unchanged and equal to 1.

<table>
<thead>
<tr>
<th>No</th>
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</table>

**Note:** L(-L) – low (not low) level; M(-M) – medium (not medium) level; H(-H) – high (not high) level

The borrower’s default risk has been determined, using the operations with fuzzy sets based on the proposed fuzzy scales of four financial indicators, the interest rates level in the economy and prescribed rules. It ought to be mentioned that, as a result of calculation, the ambiguous issuer identity at one of default risk three levels has been obtained, providing more valid risk assessment, given the impact of numerous external and internal factors for each individual borrower.

Although at previous stages bonds issuers’ classification problem was solved in terms of default risk, the next step is to solve the problem of borrower’s risk class forecasting. To achieve this, the bond issuer’s dynamic default risk level has been analysed using Markov chains method. Markov chain is a sequence of dependent trials, in which conditional events occurrence probability in each trial depends solely on the outcome of the previous trial [7].

In order to elaborate a transition probability matrix for each of the abovementioned 112 issues, specified issuers’ default risk level has been assessed for the first-, second-, third and fourth-quarter period prior to a credit event (execution or defaults on bond issue). A number of causes for risk level conversions in any quarter in comparison with the previous quarter have been calculated. For the last quarter prior to credit events the transition to one of two states – default or debt payment – have been traced. Markov chain of bonds issuers’ default risk levels has been schematically represented in Fig. 2.

![Fig. 2. The probabilities of changing the bonds issuers’ default risk level](image)

**Table 1**

**Rules for determining corporate bonds issuers’ default risk level**

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**Table 2**

**Bonds issuers’ default risk probability**

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<th>The current bonds issuer’s risk class</th>
<th>The issuer’s default probability</th>
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<td>Low level</td>
<td>3% within 1 quarter, 9% within 2 quarters, 14% within 3 quarters, 18% within 4 quarters</td>
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<tr>
<td>Medium level</td>
<td>7% within 1 quarter, 14% within 2 quarters, 20% within 3 quarters, 24% within 4 quarters</td>
</tr>
<tr>
<td>High level</td>
<td>24% within 1 quarter, 33% within 2 quarters, 39% within 3 quarters, 43% within 4 quarters</td>
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[1] The low default risk level has been schematically represented in Fig. 2. The probabilities of changing the bonds issuers’ default risk level

In Fig. 2 each risk level is defined by the probability of company’s bond obligation default of to be averted in the next quarter. Thus, if the issuer is characterized by low risk in the current quarter, within the next period it will remain in this class with the probability of 46%. It will shift to the middle level with the probability of 35%, transfer to the high grade accounts for the probability of 17%. The remaining 3% characterises the default probability on bonds.

Thus, the use of Markov chains has enabled to predict the bonds issuer’s default risk over time, relating to the current risk level and probability values transfer within risk classification in the next quarter. Bonds issuers’ default probability after a certain period of time has been defined. The results have been presented in Tab. 2.
Thus, in case the issuer is currently at the low risk level the issuer default probability within a quarter will total 3 %, and within 4 quarters – 18 % (see Table 2). The borrower’s default possibility, being in the middle or high risk class in the current quarter also increases gradually along with broadening forecast horizon. These results are quite logical, because a longer time period always implies higher risk, due to the influence of various factors, both internal and external, which are difficult to predict.

Thus, the results of the study have formed grounds for a common methodological approach to corporate bonds issuers default risk evaluation. The sequence of main steps within this approach and implementation methods have been presented schematically in Fig. 3.

Thus, correlation between the values of bond issuers selected financial indicators (a reflection of the borrower internal condition) and the market interest rates level (a reflection of the environment) on the one hand, and fulfillment of obligations or default on bonds, on the other, have been ascertained.

The research findings have enabled to classify bonds issuers into three default risk levels on the basis of the generated and optimized fuzzy rules of attribution to a particular risk level. Probability values of transfer to other level in the forthcoming quarter have been determined and the issuer’s default probability after a certain period of time has been estimated on the basis of current risk class.

Fig. 3. The sequence of steps for methodical approach to bonds issuers’ default risk assessment

Conclusions. The issuer classification highly depends on two major indices: EBITDA interest coverage ratio and other bond market risks estimates. These factors on the borrower, excluded from the model.

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