

ISSN 2708-2032
e-ISSN 2708-2040



**INTERNATIONAL
UNIVERSITY**

**INTERNATIONAL
JOURNAL OF INFORMATION
& COMMUNICATION TECHNOLOGIES**

**Volume 2, Issue 1
March 2021**

ҚАЗАҚСТАН РЕСПУБЛИКАСЫНЫҢ БІЛІМ ЖӘНЕ ҒЫЛЫМ МИНИСТРЛІГІ
МИНИСТЕРСТВО ОБРАЗОВАНИЯ И НАУКИ РЕСПУБЛИКИ КАЗАХСТАН
MINISTRY OF EDUCATION AND SCIENCE OF THE REPUBLIC OF KAZAKHSTAN



**INTERNATIONAL JOURNAL OF
INFORMATION AND COMMUNICATION
TECHNOLOGIES**

**МЕЖДУНАРОДНЫЙ ЖУРНАЛ
ИНФОРМАЦИОННЫХ И
КОММУНИКАЦИОННЫХ ТЕХНОЛОГИЙ**

**ХАЛЫҚАРАЛЫҚ АҚПАРАТТЫҚ ЖӘНЕ
КОММУНИКАЦИЯЛЫҚ
ТЕХНОЛОГИЯЛАР ЖУРНАЛЫ**

Том 2, Выпуск 1
Март 2021

Главный редактор – Ректор АО МУИТ,
профессор, д.т.н.
Ускенбаева Р.К.

Заместитель главного редактора –
проректор по НиМД, PhD, ассоц.профессор
Дайнеко Е

Отв. Секретарь – PhD, ассоц.профессор, директор департамента по науке
Кальпеева Ж.Б.

ЧЛЕНЫ РЕДКОЛЛЕГИИ:

Отельбаев М. д.т.н., профессор, АО «МУИТ», Рысбайулы Б., д.т.н., профессор, АО «МУИТ», Куандыков А.А., д.т.н., профессор, АО «МУИТ», Синчев Б.К., д.т.н., профессор, АО «МУИТ», Дузбаев Н.Т., PhD, проректор по ЦИИ, АО «МУИТ», Ыдырыс А., PhD, заведующая кафедрой «МКМ», АО «МУИТ», Касымова А.Б., PhD, заведующая кафедрой «ИС», АО «МУИТ», Шильдибеков Е.Ж., PhD, заведующий кафедрой «ЭиБ», АО «МУИТ», Ипалакова М.Т., к.т.н., ассоц. профессор, заведующая кафедрой «КИИБ», АО «МУИТ», Айтмагамбетов А.З., к.т.н., профессор, АО «МУИТ», Амиргалиева С.Н., д.т.н., профессор, АО «МУИТ», Ниязгулова А.А., к.ф.н., заведующая кафедрой «МиИК», АО «МУИТ», Молдагулова А.Н., к.т.н., ассоциированный профессор, АО «МУИТ», Джоламанова Б.Д., ассоциированный профессор, АО «МУИТ», Prof. Young Im Cho, PhD, Gachon University, South Korea, Prof. Michele Pagano, PhD, University of Pisa, Italy, Tadeusz Wallas, Ph.D., D.Litt., Adam Mickiewicz University in Poznań, Тихвинский В.О., д.э.н., профессор, МТУСИ, Россия, Масалович А., к.ф.-м.н., Президент Консорциума Инфорус, Россия, Lucio Tommaso De Paolis is the Research Director of the Augmented and Virtual Laboratory (AVR Lab) of the Department of Engineering for Innovation, University of Salento and the Responsible of the research group on “Advanced Virtual Reality Application in Medicine” of the DREAM, a multidisciplinary research laboratory of the Hospital of Lecce (Italy), Liz Bacon, Professor, Deputy Principal and Deputy Vice-Chancellor, Abertay University (Great Britain).

Издание зарегистрировано Министерством информации и общественного развития Республики Казахстан. Свидетельство о постановке на учет № KZ82VPY00020475 от 20.02.2020 г.

Журнал зарегистрирован в Международном центре по регистрации сериальных изданий ISSN (ЮНЕСКО, г. Париж, Франция)

Выходит 4 раза в год.

УЧРЕДИТЕЛЬ:

АО «Международный университет информационных технологий»

ISSN 2708-2032 (print)
ISSN 2708-2040 (online)

СОДЕРЖАНИЕ

РАЗРАБОТКА ПРОГРАММНОГО ОБЕСПЕЧЕНИЯ И ИНЖЕНЕРИЯ ЗНАНИЙ

<i>Серікбай Е.Е., Таир Р.Т., Куандыков А.А, Нальгожина Н.Ж.</i>	
Разработка бизнес-процесса для оцифровки автомобильных дорог	10
<i>Жуманбаева С.К., Пащенко Г.Н.</i>	
Проектирование и разработка информационной системы для обработки научных трудов	18
<i>Айтбекова М.Б., Пащенко Г.Н.</i>	
Разработка информационной системы для расчетов рейтингов успеваемости студентов	24
<i>Алимжанова Л.М., Омарова А.Ш., Таштамышева А.Э.</i>	
Исследование актуальных проблем при переходе обучения в онлайн-формат.....	34
<i>Найзабаева Л., Аринова М. С.</i>	
Интеллектуальный анализ и прогнозирование токсичных элементов в почве	39
<i>Имангалиева А.А., Пащенко Г.Н.</i>	
Проектирование и разработка информационной системы для управления научно - образовательной деятельностью Университета	46

ИНФОКОММУНИКАЦИОННЫЕ СЕТИ И КИБЕРБЕЗОПАСНОСТЬ

<i>Базарбеков И.М., Шарипов Б.Ж.</i>	
Система smart кампус в университете: требования, преимущества и недостатки	53
<i>Кенескызы К., Ескермес С.Б.</i>	
Метод машинного обучения для обратных задач теплопроводности	59
<i>Алимжанова Л.М., Панарина А.В.</i>	
Ценность IT-аутсорсинга для клиента	65
<i>Мамен Е.К., Айтим А.К., Аднабеков А.Х., Абиев А.Б., Мустафина А.К.</i>	
Разработка умного холодильника без продавца	71
<i>Ауезова А.С., Муратова К.Н., Синчев Б.</i>	
Методы информационного поиска неструктурированных данных	79
<i>Надіров Н.Қ., Дүйсебекова К.С.</i>	
Разработка системы формирования профиля клиента на основе bigdata с использованием семантического анализа	85
<i>Бердыкулова Г.М.</i>	
Что такое научная статья и как не надо ее писать?	96

ЦИФРОВЫЕ ТЕХНОЛОГИИ В ЭКОНОМИКЕ И МЕНЕДЖМЕНТЕ

Мукан Б., Саркамбаева Ш.Г.

Развитие проектной деятельности в малом и среднем бизнесе в Казахстане 104

Мейрманова Л.Е.

Эффективность применения управления проектами в управлении человеческими ресурсам 113

Алдибекова Н.Б., Тяп А.В., Омаров И.Г., Мохамед А.Х., Алимжанова Л.М.

Использование математического моделирования и программного обеспечения в управлении проектными рисками 119

Омарова А.Ш., Махбаева Ә.Н.

Проблемы и влияние коммуникаций в управлении проектами 129

Момбекова А., Бекболова М.Б.

Влияние аналитики больших данных на эффективность процессов управления проектами ..137

Булантаев А.М., Мусахан Х.Б., Молдагулова А.Н., Сембина Г.К.

Прогноз ожидаемых убытков банка при предоставлении кредита 145

Омарова Е.Г.

Алгоритм автоматизации классификации финансовых активов при розничном кредитовании 150

CONTENTS

SOFTWARE DEVELOPMENT AND KNOWLEDGE ENGINEERING

<i>Alin G.T., Rakhymzhanova N.K.</i> Software development project management: project risk management	9
<i>Zhumanbaeva S.K., Pachshenko G.N.</i> Desining and development of information system for the processing scientific works	19
<i>Aitbekova M.B., Pachshenko G.N.</i> Development of the information system for calculating students' performance rating.....	25
<i>Alimzhanova L.M., Omarova A.S., Tashtamysheva A.E.</i> Investigation of topical problems during the transition of learning in online format.....	35
<i>Naizabayeva L., Arinova M.S.</i> Intellectual analysis and prediction of toxic elements in the soil	40
<i>Imangalieva A., Pachshenko G.N.</i> Development of an information system for managing research and educational activi-ties of the university.....	47
<i>Serikbay Y.Y., Tair R.T., Kuandykov A.A., Nalgozhina N.Zh.</i> Development of a business process for digitizing roads.....	54

INFOCOMMUNICATION NETWORKS AND CYBERSECURITY

<i>Bazarbekov I.M., Sharipov B.Zh.</i> Smart campus system in the university: requirements, advantages and disadvantages.....	62
<i>Keneskyzy K., Yeskermes S.B.</i> Machine learning method for inverse heat conduction problems	68
<i>Alimzhanova L.M., Panarina A.V.</i> The value of IT outsourcing for the client	74
<i>Mamen Y.K., Aitim A.K., Adnabekov A.H., Abiyev A.B., Mustafina A.K.</i> Development of a smart refrigerator without a seller	80
<i>Auyezova A.S., Muratova K.N., Sinchev B.</i> Methods of information search for unstructured data	88
<i>Nadirov N. To., Duisenbekova K. S.</i> Development of a client profile formation system based on Big data using semantic analysis	94
<i>Berdykulova G.M.</i> What is an academic article or how not to write it?.....	105

DIGITAL TECHNOLOGIES IN ECONOMICS AND MANAGEMENT

<i>Mukan B., Sarkambaeva S.G.</i> Development of project activities in small and medium-sized businesses in Kazakhstan	113
<i>Meirmanova L.Ye.</i> Efficiency application project management in human resource management.....	122

<i>Aldibekova N.B., Tyan A.V., Omarov I.G., Mohamed A. Hamada., Alimzhanova L.M.</i>	
Using mathematical modelling and software programming in project risk management	128
<i>Omarova A., Makhbayeva A.</i>	
Challenges and impact of communication in project management	138
<i>Mombekova A., Bekbolova M.B.</i>	
The impact of big data analytics on the effectiveness of project management processes	146
<i>Bulantayev A.M., Musakhan K.B., Moldagulova A.N., Sembina G.K.</i>	
Forecasting expected bank losses at granting a loan.....	154
<i>Omarova Ye.G.</i>	
Algorithm for automating the classification of financial assets in retail lending	159

МАЗМҰНЫ

БАҒДАРЛАМАЛЫҚ ҚАМТАМА ҚҰРУ ЖӘНЕ БІЛІМ ТЕХНИКАСЫ

<i>Алин Г.Т., Рахимжанова Н.К.</i>	
Бағдарламалық даму жобасын басқару: жобаның тәуекелділігін басқару	9
<i>Жуманбаева С.К., Пащенко Г.Н.</i>	
Ғылыми еңбектерді өңдеуге арналған ақпараттық жүйені жобалау және зерттеу	19
<i>Айтбекова М.Б., Пащенко Г.Н.</i>	
Оқушылардың үлгерімін бағалауға арналған ақпараттық жүйе құру	25
<i>Алимжанова Л.М., Омарова А.Ш., Таштамышева А.Э.</i>	
Білім берудің онлайн форматқа көшуіндегі өзекті мәселелерді зерттеу	35
<i>Найзабаева Л., Аринова М. С.</i>	
Жер қыртысындағы улы элементтерді интеллектуалды талдау және болжау	40
<i>Иманғалиева А.А., Пащенко Г.Н.</i>	
Университеттің ғылым және білім беру қызметін басқару ақпараттық жүйесін жобалау және әзірлеу	47
<i>Серікбай Е.Е., Таур Р.Т., Қуандықов А.А., Нальгожина Н.Ж.</i>	
Жолдарды цифрландырудың бизнес процесін дамыту	54

ИНФОКОММУНИКАЦИЯЛЫҚ ЖЕЛІЛЕР ЖӘНЕ КИБЕРҚАУІПСІЗДІК

<i>Базарбеков И.М., Шарипов Б.Ж.</i>	
Университеттегі Smart кампус системасы: талаптар, артықшылықтары мен кемшіліктері ...	62
<i>Кенесқызы К., Ескермес С.Б.</i>	
Кері жылу өткізгіштік есептеріне арналған машиналық оқыту әдісі	68
<i>Алимжанова Л.М., Панарина А.В.</i>	
Клиент үшін ІТ аутсорсингтің мәні.....	74
<i>Мамен Е.К., Әйтім Ә.Қ., Аднабеков А.Х., Абиев А.Б., Мустафина А.Қ.</i>	
Сатушысыз ақылды тоңазытқышты құру	80
<i>Ауезова Ә.С., Муратова К.Н., Синчев Б.</i>	
Құрылымданбаған деректерді ақпараттық іздеу әдістері	88
<i>Нәдіров Н.Қ., Дуйсебекова К.С.</i>	
Семантикалық талдауды қолдану арқылы Bigdata негізінде клиент профилін құру жүйесін қалыптастыру	94
<i>Бердыкулова Г.М.</i>	
Ғылыми мақала дегеніміз не немесе оны қалай жазбауға болады?	105

ЭКОНОМИКА ЖӘНЕ МЕНЕДЖМЕНТТЕГІ ЦИФРЛЫҚ ТЕХНОЛОГИЯЛАР

Мұқан Б., Саркамбаева Ш.Г.

Қазақстандағы шағын және орта бизнесте жобалау қызметін дамыту 113

Мейрманова Л.Е.

Адам ресурстарын басқаруда жобаларды басқаруды қолданудың тиімділігі 122

Алдибекова Н.Б., Тяп А.В., Омаров И.Г., Мохамед А.Х., Алимжанова Л.М.

Жобалық тәуекелдерді басқаруда математикалық модельдеу мен бағдарламалық жасақтаманы қолдану 128

Омарова А.Ш., Махбаева Ә.Н.

Жобаларды басқарудағы коммуникацияның мәселелері мен әсері 138

Момбекова А., Бекболова М.Б.

Үлкен деректер аналитикасының жобаны басқару процестерінің тиімділігіне әсері 146

Булантаев А.М., Мұсахан Х.Б., Молдагулова А.Н., Сембина Г.К.

Несие беру үшін банктен күтілетін шығындардың болжауы 154

Омарова Е. Ғ.

Бөлшек несиеелеудегі қаржы активтерін жіктеуді автоматтандыру алгоритмі 159

Bulantayev A.M.*, Musakhan K.B., Moldagulova A.N., Sembina G.K.

International Information Technology University, Almaty, Kazakhstan

FORECASTING EXPECTED BANK LOSSES AT GRANTING A LOAN

Abstract. This article uses the sample data of the SAS platform as an example to introduce the statistical analysis and prediction of the expected loss of loans issued by banks. The original data for this study comes from a Kaggle source, which provides information about the credit history of bank customers. The technology is based on logistic regression, graphical data analysis, and the basis of building a model on the SAS platform. The model can be used to predict credit risk and describe credit risk in the banking system.

Keywords: component, data analysis, credit risk, Loss Given Default, Expected Loss, Probability of Default, Exposure at Default, logistic regression, model, non-performing loans, Special Air Service platform, forecasting

Introduction

This kind of credit risk is one of the most common types of risks, and it has a vital impact on the robustness of a single bank's exercise and the entire framework of the currency of custody. One of the main reasons for emergencies is to prompt financial analysts to underestimate the risks associated with the use of unused currency violations and the subsequent emergencies. In combination with the potentially unsafe consequences of credit opportunities, it is essential to fully evaluate them.

Under this circumstance, the state-of-the-art bank must not determine the importance of its image in the same way that it assesses the opportunity level of a credit asset portfolio when conducting lending activities. In the unstable budget market and currency-related emergencies, it is vital to be able to foresee the pattern of bank marks.

However, the current problem is the need for mandatory tools to predict credit risk. We studied a modern method based on the use of evaluable strategies to estimate the expected misfortunes of banks when issuing advances, taking into account different variables that affect the risk of advances portfolio, and using information provided by Kaggle.

The biggest purpose of consideration is to estimate the unfortunate circumstances of banks when issuing credit for explanatory considerations of suitability. The strange thing to consider is the improvement of the program using the fact-finding data of the banking framework obtained by Kaggle. The emergence of term papers is a demonstration of the strategy of using historical bank data and the use of SAS equipment to analyze the credit risk within the scope of banking services.

New impetus for adjusting credit risk management

From the perspective of cash-related instructions, the conditions for the activation of the COVID-19 emergency have special propositions for monitoring and coordinating credit risk. In the past three months, banks have been adapting to the evolving trend and researching the most cutting-edge methods to deal with challenges. These inspections assess the overall impact of the emergency on the national or regional economy, and the impact of various departments and sub-sectors, especially credit risk issues that require real-time inspections. The examination started when it was directly discovered that this emergency situation had five curious effects on credit risk.

At the time of crisis, the change in budget conservatism is more critical than the differences between departments and the past divestments. Certain companies, such as nutrition wholesalers, dominate the crisis and are fighting to meet the increasing demand. Other factors, such as broadcast communications and medicines, were hardly affected. But as we all know, certain sectors (such as travel, transportation, tourism and neighborhoods) are indeed challenged [1].

Nonetheless, to properly outline cash-related robustness within this emergency, banks must go through inspections of parts or sub-sectors and investigate the borrowers. The business model from one company to another in the same sub-industry may be absolutely correct, and along these lines of thinking, it is more or less suitable for survival and faster recovery in the current environment. For example, many companies have very close online businesses, while others do not. Therefore, banks cannot draw conclusions from sub-industry inspections alone, that is, whether a particular borrower is annoyed.

In addition, the standard data sources used in credit risk assessment must become obsolete overnight. This crisis has manifested itself as a powerful external shock in most cases of the global credit cycle. Both supply and asking prices were accidentally stifled. In addition, suddenly, the data of the past six or twelve months of data is not important for the quality of the survey of individual borrowers. The innovative method of acquiring and using high-frequency data is today's organization [2].

In the process of developing the COVID-19 crisis, due to the prejudice of customers, their banks eventually developed naturally, and the benefits of habitual installment payment collection technologies (phones, emails, letters) are getting smaller and smaller. The huge wave of undesirable progress that is currently forming will occupy a long exhaust direction resource in the near future.

Towards Data-Driven analysis and Real-Time decision-making

In order to respond to emergency situations, teaching related to driving money has begun to adopt unused market segment surveys, borrower versatility and high-frequency analysis methods to recognize and observe. One of the key changes we have seen is that pioneers are rapidly shifting from the perspective of one department to a sub-sector, and finally using real-time information and analysis techniques of debt to quickly transfer to the perspective of debt [3]. Most banks have conducted precise speculations in specific sub-sectors and are getting closer (or up to now) to the debtor's view of risk assessment.

Crossing this skyline is the method of investigating forms of credit underwriting by using real-time trade information in decision-making and schedule analysis [4]. Switching to these unused strategies will help banks adapt to emergencies, but in addition, as a practice, we believe that credit risk management should be changed in the coming months and a long time. The best banks will indeed continue to maintain and extend these conditions after emergencies, in order to more effectively monitor credit risk, to serve senior customers and enable them to recover faster.

SAS application style for predicting expected bank losses when issuing loans

The software market continues to grow. A new software package was developed based on the rapidly developing modern computer technology. The functions of the program are being expanded so that people can use them not only as simple tools to simplify work, but also as mature assistants that can solve complex problems in banks. This study shows the use of statistical software package to analyze and predict bank data.

In order to determine the initial predicted value of the credit risk level, we recommend using SAS statistical analysis methods. To predict the risk of the loan portfolio, it is necessary to determine the criteria for changing the level of this indicator. We suggest that the main criterion for credit risk indicators is a model built on the bank's historical database, which shows the possible risks of a specific customer, that is, the possible loss of the bank due to a specific customer. We believe that this model most accurately characterizes the quality of bank loan portfolios.

All customer credit records are stored in the bank database. Credit history is information about your loan obligations. It shows the bank, the microfinance organization (MFI) or the credit consumer cooperative (CCP) from which you are applying for loans and borrowing. When was that and how much you spent. Have you ever been a co-borrower or guarantor of loans from others [5]? Do they pay carefully or delay payment?

The amount of loss that the bank may face can be predicted based on the credit history data. The average level of bank losses is estimated by calculating the expected loss index (EL). From the perspective of the bank, the indicator of expected loss is an element of business value. Use basic risk parameters to estimate the amount of expected loss: PD gives the average annual probability of default of the borrower, and $EAD \times LGD$ - the level of losses [6]. Thus, the expected losses can be estimated as follows:

$$EL = PD * EAD * LGD. \quad (1)$$

Where EL— Expected Loss; PD - Probability of Default; EAD - Exposure at Default; LGD - Loss Given Default. symbols in your equation have been defined before or immediately following the equation.

The Basel Committee determined the following [12]:

- Average annual default probability (default rate) and borrower rating. PD is the probability that the loan cannot be repaid. The probability of default will be calculated for each employee. There are many different models that allow you to calculate PD based on available information. Three main categories can be distinguished: structural models, simplified models, and credit scoring models. The first two methods are based on market data (stock prices, bond yields). Therefore, the credit scoring model has the greatest practical significance. All value ranges are evaluated by the rating team. In addition, with the help of a special calibration of the rated value, the possibility of a default can be determined. PD, a similar rating group, has been elected president for one year [7];

- According to requirements exposure (EAD). EAD is an estimate of the amount of risk (that is, part of the loan). In this case, the following factors must be considered: First, the debt under the system (especially complex products with a limit system) may fluctuate over time, so it is necessary to evaluate its value in the event of default. [8]. Secondly, the existence of highly liquid collateral allows you to reduce EAD, so its implementation makes it possible to quickly repay part of the lost loan. However, the left part of the loan is unlikely to be repaid in full;

- The average expected loss of capital share (loss at default, LGD) at the time of default is usually calculated as a percentage of EAD. LGD is exactly the estimated loss of EAD. It is necessary to provide additional guarantees for the loan, assess the importance of the collateral to the customer and the current financial situation of the borrower, namely its rating. When calculating LGD and EAD, it is very important to correctly determine the issues of cost, liquidity and return probability.

Collect data on indicator evaluation and processing

The basis of the research is data on bank credit records. As the initial information, data describing credit history at the beginning of 2020 was selected based on monthly records of credit transaction cases. The following categories of credit history are introduced:

When analyzing a customer's credit history, three main types of tasks are solved: comparing credit history indicators, analyzing the dynamics of credit operations, and analyzing the level of credit operations.

Table 1 - EL, PD, LGD, EAD results

Funded_amnt	PD	LGD	EAD	EL
5000	0.1647909316466200	0.9137287259284660	2949.6084487904	444.13495613990800
2500	0.2828104744734790	0.9154816249384510	1944.4333780507400	503.42895400336700
2400	0.2307362354090770	0.9194838219680120	1579.9343016863600	335.19607381040500
10000	0.20874299947855700	0.9049241978540790	6606.559611654160	1247.9565912633200
3000	0.1292812709740640	0.9114527959410170	2124.631666989780	250.3533717013720
5000	0.13580650461121300	0.9188613978157500	3174.1650557315200	396.095660557456
7000	0.29133725401367500	0.9165305685864820	5052.384586825450	1349.0852014910900
3000	0.2975299298493430	0.91369595859268	2056.324222637450	559.0156355556010
5600	0.41801183398990700	0.9002332778846770	4810.650248440880	1810.2869602668200
5375	0.1882600619740530	0.919839777105950	3901.7992509115700	675.6710395256220
6500	0.180805842929102	0.9063090874655570	4788.668721435330	784.6998858150640

In order to collect data, it is necessary to select at least 50 customers belonging to the target industry group under consideration and collect the indicators that have been determined for them in the past few years. At the same time, data can be obtained from various sources (including those publicly available). In particular, data related to reports from the disclosure server, data from the Federal Office of State Statistics 2 etc. It is convenient to use related indicators (financial relations). In addition, financial indicators must be included in the same time range (quarter, month, year), and the duration of the production cycle must be considered in calculations.

Acknowledgment

Therefore, in modern banks, the role of risk management is significantly improving. When the decision to grant a loan is: accept/reject, the risk assessment allows you to give up the simple method. The credit risk management system becomes the basis for a reasonable dialogue with customers based on objectives including investment portfolio indicators. A properly structured business process will play an important role. The activities of the risk management department cannot be separated from other departments of the bank, because a correct and objective assessment requires all available information, including other departments and external resources. The role of risk management is not limited to the stage of considering the application - it is establishing the best portfolio structure, continuous monitoring, and the importance and understanding of existing risks.

It is impossible to completely eliminate the risk. There is no income without risk: the greater the risk, the higher the possible income. But you need to know the risks to prepare for possible consequences.

REFERENCES

1. V. Acharya. "Systematic Risk Theory and Prudent Banking Supervision Design", "Financial Stability Magazine", Volume 1. 5(3), 2007, 224-255 p.
2. E. Asarnow, D. Edwards. Measuring the loss of bank loans in arrears: a 24-year study. Journal of Business Loan, 2013, 11-23 p.
3. T. Bellotti, J. Crook. Model and predict the loss of credit card default settings. Working paper, Quantitative Financial Risk Management Center, 2007, 11-14 p.
4. S. Caselli, S. Gatti, F. Querci. The sensitivity of loss for a given default rate to systemic risk: new empirical evidence of bank loans. Journal of Financial Services Research, 34(1): 2008, 1-34 p.
M. Araten, M. Jacobs, P. Varshney. Measuring the default loss rate of commercial loans: an 18-year internal study. RMA Magazine, 2004, 96-103 p.
5. E. I. Altman. Default Recovery Rate and Default Loss Rate in Credit Risk Modeling and Practice: The latest review of literature and empirical evidence. Working paper, New York University Stern School of Business, 2006, - 123 p.
6. Schatz. The method of credit risk quantification: Probability of Default (PD). financial report."BBVA Magazine", 2011, - 133 p.
7. S.G. Ryan, D.Vyas. Bank credit risk model disclosures have an impact on their loan loss preparation timeliness and loan origination. Working paper of New York University, 2012, 24-27 p.

**Булантаев А.М.*, Мұсахан Х.Б., Молдагулова А.Н., Сембина Г.К.
Несие беру үшін банктен күтілетін шығындардың болжауы**

Булантаев Арыслан Муратович, 2 курс магистранты, Халықаралық ақпараттық тех-нологиялар университеті.

Мұсахан Хадиша Бақытжанқызы, 2 курс магистранты, Халықаралық ақпараттық технологиялар университеті.

Молдагулова Айман Николаевна, ф.-м.ғ.к., «Ақпараттық жүйелер» кафедрасының доценті, Халықаралық ақпараттық технологиялар университеті.

Сембина Гулбакыт Какеевна, т.ғ.к., «Ақпараттық жүйелер» кафедрасының доценті, Халықаралық ақпараттық технологиялар университеті.

Аңдатпа. Бұл мақалада статистикалық талдау мен банктер шығарған несиелер бойынша күтілетін шығындар болжамын енгізу үшін SAS платформасының үлгі деректері келтірілген. Осы зерттеудің бастапқы деректері банк клиенттерінің несиелік тарихы туралы ақпарат беретін Kaggle көзінен алынған. Технология логистикалық регрессияға, графикалық деректерді талдауға және SAS платформасында модель құру негіздеріне негізделген. Үлгіні несиелік тәуекелді болжау және банк жүйесіндегі несиелік тәуекелді сипаттау үшін пайдалануға болады.

Түйін сөздер: компонент, деректерді талдау, несиелік тәуекел, берілген дефолттағы шығын, күтілетін шығын, дефолт ықтималдығы, дефолтқа ұшырау, логистикалық регрессия, модель, жұмыс істемейтін несиелер, арнайы әуе сервисінің платформасы, болжау

**Булантаев А.М.*, Мұсахан Х.Б., Молдагулова А.Н., Сембина Г.К.
Прогноз ожидаемых убытков банка при предоставлении кредита**

Булантаев Арыслан Муратович, магистрант 2 курса кафедры «Информационных си-стем» Международного университета информационных технологий.

Мұсахан Хадиша Бақытжанқызы, магистрант 2 курса кафедры «Информационных систем» Международного университета информационных технологий.

Молдагулова Айман Николаевна, к.ф.-м.н., доцент кафедры «Информационных си-стем» Международного университета информационных технологий.

Сембина Гулбакыт Какеевна, к.т.н., доцент кафедры «Информационных систем» Международного университета информационных технологий.

Аннотация. В этой статье используются образцы данных платформы SAS в качестве примера для представления статистического анализа и прогнозирования ожидаемых убытков по кредитам, выданным банками. Исходные данные для этого исследования взяты из источника Kaggle, который предоставляет информацию о кредитной истории клиентов банка. Технология основана на логистической регрессии, графическом анализе данных и на основе построения модели на платформе SAS. Модель может использоваться для прогнозирования кредитного риска и описания кредитного риска в банковской системе.

Ключевые слова: компонент, анализ данных, кредитный риск, убыток при дефолте, ожидаемый убыток, вероятность дефолта, риск дефолта, логистическая регрессия, модель, неработающие займы, платформа Special Air Service, прогнозирование.

About the authors:

Bulantayev Aryslan Muratovich, second year master's degree student, Department of In-formation Systems, International Information Technology University.

Musakhan Khadisha Bakytzhankyzy, second year master's degree student, Department of Information Systems, International Information Technology University.

Moldagulova Aiman Nikolaevna, PhD in Physics and Mathematics, Associate Professor, Department of Information Systems, International Information Technology University.

Sembina Gulbakyt Kakeevna, PhD in Engineering Science, Associate Professor, Depart-ment of Information Systems, International Information Technology University.

INTERNATIONAL JOURNAL OF INFORMATION AND
COMMUNICATION TECHNOLOGIES

МЕЖДУНАРОДНЫЙ ЖУРНАЛ ИНФОРМАЦИОННЫХ И
КОММУНИКАЦИОННЫХ ТЕХНОЛОГИЙ

ХАЛЫҚАРАЛЫҚ АҚПАРАТТЫҚ ЖӘНЕ
КОММУНИКАЦИЯЛЫҚ ТЕХНОЛОГИЯЛАР ЖУРНАЛЫ

Ответственный за выпуск	Есбергенов Досым Бектенович
Редакторы	Далабаева Айсара Касымбековна Садганова Эльмира Абуовна
Компьютерная верстка	Туратауова Айжаркын Ахметовна
Компьютерный дизайн	Туратауова Айжаркын Ахметовна

Редакция журнала не несет ответственности за
недостоверные сведения в статье и
неточную информацию по цитируемой литературе

Подписано в печать 26.03.2021 г.
Тираж 500 экз. Формат 60x84 1/16. Бумага тип.
Уч.-изд.л. 10.1. Заказ №158

Издание международный университет информационных технологий
Издательский центр КБТУ, Алматы, ул. Толе би, 59