

Information systems modeling

UDC 004.048

doi: <https://doi.org/10.20998/2522-9052.2026.1.01>Svitlana Krepych¹, Iryna Spivak¹, Serhii Spivak², Roman Krepych³¹ West Ukrainian National University, Ternopil, Ukraine² Ternopil Ivan Puluj National Technical University, Ternopil, Ukraine³ Kamianets-Podilskyi State Institute, Kamianets-Podilskyi, Ukraine

MATHEMATICAL MODEL FOR CALCULATING THE EXPERT'S COMPETENCY LEVEL

Abstract. In the context of the rapid development of information technologies, software quality is becoming critical for the successful operation of organizations in various industries. The growing complexity of modern software solutions requires the involvement of highly qualified specialists in software testing and quality assessment, capable of effectively identifying shortcomings and ensuring that the product meets established standards. At the same time, assessing the level of competence of such experts remains a difficult task, which is often based on subjective criteria and methods. The relevance of the study is due to the acute need of the modern IT market for objective tools for assessing the professional level of specialists, especially in the field of software quality assurance. Traditional approaches to qualification assessment, such as interviews, test tasks or resume analysis, often do not provide a complete and objective picture of the expert's competence. This problem becomes especially acute in the conditions of the global labor market, when companies are forced to evaluate specialists remotely, relying only on a limited set of data on their experience and skills. Today, software has become an integral part of many areas of our everyday life - from automation and optimization of production processes to creating comfort for an individual. **The object of the study** is the process of determining the level of competence of experts in software quality assessment. **The subject of the study** is a mathematical model for calculating the level of competence of an expert. The practical value of the results of the work is determined by the possibility of using the developed system by HR managers for effective selection of specialists, by heads of QA departments for the formation of balanced testing teams, by certification centers for objective assessment of competence, as well as by the experts themselves for planning their own professional development. **Conclusion** the developed mathematical model for calculating the level of competence of an expert allows you to reduce the time for assessing the competence of specialists, minimize the influence of subjective factors when making personnel decisions, and optimize the distribution of human resources in software development and testing projects.

Keywords: expert evaluation; testing; software quality; mathematical evaluation methods; model.

Introduction

Software quality assessment is a complex process that includes the analysis of functional compliance, performance, security, usability, maintainability and other characteristics [1]. The results of such an assessment significantly depend on the qualifications of the experts involved. However, traditional methods for determining the level of expertise of specialists are often based on subjective criteria or formal indicators that do not always reflect the real level of competence [2–4]. The research problem is the need to develop an objective, mathematically based system for assessing experts, which would allow formalizing the process of analyzing professional qualities and minimizing the influence of subjective factors. Modern research on this topic often focuses on individual aspects of assessment, ignoring the complex nature of the examination and the relationship between various professional indicators [5].

The publication [6] considers methods and means of expert assessment of software systems, and also provides their comparative characteristics. Special attention is paid to the modified method of expert assessment of software systems based on interval data analysis. The proposed method allows obtaining an interval assessment of a software product by experts, which is guaranteed to satisfy the requirements of

software developers.

The study [7] considers the task of constructing a criterion for the compatibility of expert assessments in a group. The feasibility of using interval data analysis methods to construct the specified criterion is shown. The example demonstrates the effectiveness of using this criterion for selecting experts for their further involvement in the project assessment process. The work focuses on the compatibility of opinions between experts, which is an important aspect in the formation of expert groups. The use of interval data is effective for taking into account uncertainty in expert assessments, which is relevant for increasing the reliability of the results.

The article [8] investigates the problem of assessing the level of objectivity and qualification of experts in project assessments presented by them according to specified criteria. The feasibility of constructing quality assessment criteria for each expert based on a modified interval method of expert assessment is determined. The formulated criterion is based on the consistency of the interval assessment of the expert with two important indicators: the initial specified interval assessment of the customer and the interval assessment of the end user. The use of interval assessments allows to take into account uncertainty, but can complicate the comparison and aggregation of assessments.

In the study [9], the assessment is considered as a procedure for fixing quantitative and qualitative characteristics, which plays a significant role in the analysis of the final result of any activity. Studies of software systems assessment have shown that none of the methods and approaches to the expert assessment procedure is ideal. The quality of the final assessments is significantly affected by such criteria as the subjectivity of the expert or group of experts, as well as their competence and qualifications. The article shows how the proposed criterion for assessing the level of expert competence can minimize the influence of the subjectivity of opinions on the quality of assessments. In practice, the effectiveness of using the modified interval method of expert assessment when selecting experts who will carry out the procedure of expert assessment of software.

Statement for the task

Existing approaches to assessing the level of experts in the field of software engineering are often based on qualitative indicators, such as reputation, peer recommendations, past projects. Although these factors are important, they are difficult to formalize and quantify. At the same time, the growth of the popularity of professional networks, in particular LinkedIn, has created a unique source of quantitative data on the professional path and achievements of specialists [10–12]. Key problems that exist in the subject area include:

- The subjectivity of traditional methods of assessing experts, which leads to inconsistency and potential bias in determining their qualifications;
- The lack of a standardized methodology for determining the weight of different professional indicators in assessing the overall level of expertise.
- The difficulty of establishing a correlation between formal indicators (experience, education, certificates) and the real ability of the expert to provide substantiated assessments of software quality.

- The dynamic nature of the industry, which requires constant updating of evaluation criteria and their relative importance.

- The need for a balance between the simplicity of data collection for evaluation and the complexity of analysis to ensure its accuracy.

An important aspect of the study is to establish a connection between the level of expertise of a specialist and the reliability of his judgments about the quality of software. Qualitative software evaluation requires not only technical knowledge, but also analytical skills, critical thinking and the ability to take into account various factors [13].

The developed mathematical model should take into account these nuances, providing a multidimensional assessment of expertise, which correlates with the ability of a specialist to provide informed judgments about the quality of software products [14–17].

Thus, the development of methods for objectively assessing the level of experts in the field of software engineering is a necessary step to improve the quality and reliability of software evaluation processes [18–20].

The creation of a mathematical model that allows you to quantitatively determine the level of trust in expert assessments is essential for improving software quality assurance practices and making informed decisions about the development and implementation of software products [21–27].

Main part

To build any mathematical model, the first step is to identify key indicators, the formalization of which most accurately affects the final result. To build a mathematical model for calculating the level of expert competence, a set of quantitative parameters was obtained from professional LinkedIn profiles (Fig. 1), each of which represented a separate aspect in the field of software engineering.

The screenshot shows a LinkedIn profile for Roman Krepch. The profile includes the following sections:

- Experience:** Lists four positions: Project Manager at Yalantis (Full-time, Jun 2023 - Present, 4 mos, Remote), Project Manager at GlobalDev Group (Full-time, Aug 2024 - Jan 2025, 6 mos, Remote), Project Manager at ELEKS (Full-time, Apr 2023 - Jun 2024, 1 yr 3 mos), and Project Manager at SoftServe (Full-time, Dec 2021 - Jan 2023, 1 yr 2 mos). A note says "helped me get this job".
- Education:** Lists a Master's degree in Computer Science from Ternopil Academy of National Economy (2000 - 2005, Grade: Red Diploma) and a Gold Medal from Terebovlya gymnasium (1994 - 2000).
- Licenses & certifications:** Lists a Project Management Professional (PMP) certification from Project Management Institute, issued Nov 2017, expired Oct 2020. A "Show credential" button is present.
- Skills:** Lists Project Planning and Software Development Life Cycle (SDLC). A "Show all 11 skills" button is present.
- Recommendations:** Shows two recommendations:
 - Sergii Kutuzov:** Head of Branch Office at ELEKS (May 19, 2025). Note: Sergii worked with Roman but on different teams. Text: "I highly recommend Roman Krepch as a Project Manager. He has a strong technical foundation combined with excellent skills in process and quality improvement. Roman is a dedicated PM with solid abilities in planning and team leadership and consistently delivers great results. He would be a valuable asset to any team."
 - Denys Teslenko:** Senior Software Engineer | ASP.NET Core | Angular at Intaker (May 16, 2025). Note: Denys worked with Roman on the same team. Text: "I enjoyed my cooperation with Roman very much. He is an outstanding Project Manager, who is capable of managing very sophisticated projects and teams of different sizes, managing resources and releases at the highest level, being professional and friendly at the same time, facilitating high productivity of the team and easy communication with the client. Roman is always very enthusiastic about product success and team comfort equally."

Fig. 1. LinkedIn forms

These criteria were carefully selected to cover both technical competence and professional recognition within the industry ecosystem. The measurable nature of these parameters ensures objectivity in the assessment process, and their diversity creates a multidimensional assessment model. Each criterion serves as an indicator of different aspects of professional development - from formal education to practical experience and recognition by colleagues. So, the following indicators will serve as input data for the mathematical model:

E – experience (calculated as the sum of working months);

C – number of companies where the expert worked (on the one hand, the higher this number, the higher the probability that the expert may be unstable and unprofessional, but on the other hand, a large number of companies means multifaceted experience);

Ed – (although nowadays in most cases for software developers, education may not be relevant to their competence, however, completed higher education is still an important criterion that shows at least the ability to complete long-term processes);

P – projects (the higher this number, the wider the range of projects with which the expert worked, which may indicate higher competence, at the same time, a small number of projects, but their long-term duration, may indicate the level of the expert in terms of ideology, high technical expertise, sustainability, etc.);

S – services offered (the scope of the expert's skills and their relevance to specific domains of software evaluation);

CL – number of certificates, licenses, and courses completed;

Table 1 – Meaning of expert selection indicators

| Експерт | <i>E</i> | <i>C</i> | <i>Ed</i> | <i>P</i> | <i>S</i> | <i>CL</i> | <i>L</i> | <i>RR</i> | <i>GR</i> | <i>Sk</i> | <i>SkE</i> | <i>Ec</i> | <i>R</i> | |
|---------|----------|----------|-----------|----------|----------|-----------|----------|-----------|-----------|-----------|------------|-----------|----------|---|
| 1 | 70 | 2 | 1 | 0 | 1 | 1 | 0 | 0 | 2 | 29 | 27 | 238 | 7 | |
| 2 | 173 | 6 | 3 | 0 | 1 | 2 | 5 | 0 | 1 | 50 | 33 | 213 | 10 | |
| 3 | 196 | 6 | 3 | 0 | 26 | 1 | 0 | 1 | 4 | 13 | 13 | 94 | 9 | |
| 4 | 119 | 3 | 3 | 0 | 2 | 2 | 0 | 0 | 0 | 21 | 21 | 87 | 8 | |
| 5 | 166 | 3 | 3 | 10 | 0 | 1 | 0 | 0 | 0 | 9 | 9 | 241 | 10 | |
| 6 | 78 | 1 | 0 | 3 | 0 | 1 | 0 | 0 | 0 | 10 | 10 | 30 | 6 | |
| 7 | 64 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 17 | 11 | 14 | 4 | |
| 8 | 71 | 2 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 18 | 13 | 37 | 5 | |
| 9 | 124 | 5 | 3 | 0 | 1 | 1 | 0 | 1 | 0 | 19 | 15 | 40 | 8 | |
| 10 | 151 | 7 | 1 | 5 | 12 | 1 | 0 | 0 | 1 | 53 | 39 | 140 | 9 | |
| 11 | 188 | 4 | 1 | 10 | 2 | 1 | 0 | 1 | 1 | 12 | 12 | 250 | 10 | |
| 12 | 129 | 5 | 2 | 0 | 3 | 1 | 0 | 0 | 0 | 21 | 21 | 47 | 9 | |
| 13 | 113 | 3 | 0 | 0 | 0 | 1 | 0 | 0 | 2 | 13 | 9 | 35 | 5 | |
| 14 | 206 | 4 | 1 | 13 | 0 | 1 | 0 | 10 | 6 | 31 | 31 | 582 | 10 | |
| 15 | 154 | 4 | 1 | 0 | 4 | 1 | 0 | 3 | 1 | 59 | 20 | 53 | 9 | |
| 16 | 121 | 6 | 1 | 0 | 0 | 1 | 0 | 2 | 1 | 7 | 7 | 321 | 8 | |
| 17 | 230 | 4 | 2 | 0 | 1 | 1 | 0 | 1 | 2 | 9 | 9 | 19 | 9 | |
| 18 | 179 | 3 | 1 | 2 | 7 | 2 | 0 | 12 | 9 | 77 | 18 | 527 | 5 | |
| 19 | 153 | 3 | 2 | 0 | 1 | 1 | 0 | 0 | 0 | 23 | 9 | 24 | 7 | |
| 20 | 130 | 1 | 5 | 0 | 2 | 0 | 0 | 0 | 0 | 15 | 9 | 35 | 6 | |
| 21 | 118 | 4 | 1 | 0 | 1 | 0 | 0 | 1 | 3 | 5 | 5 | 43 | 5 | |
| 22 | 190 | 9 | 3 | 10 | 2 | 2 | 0 | 5 | 12 | 18 | 18 | 210 | 6 | |
| 23 | 209 | 12 | 1 | 0 | 3 | 1 | 0 | 3 | 4 | 25 | 25 | 529 | 7 | |
| 24 | 143 | 4 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 8 | 8 | 16 | 6 |

We build a mathematical model that includes all twelve indicators of the "quality" of an expert and for

this we choose a linear structure of the model of the following form:

L – number of languages (specified (this criterion only takes into account languages that are foreign to the expert);

RR – recommendations received (an indicator of the recognition of the expert's competence by other LinkedIn members and, at the same time, communication and reputational assessments from colleagues in the industry);

GR – recommendations given (a reflection of the expert's ability to evaluate and support others in their professional network);

Sk – skills (the total number of skills listed in the LinkedIn profile. This criterion indicates the breadth of the person's expertise, demonstrating their versatility and knowledge in various fields);

SkE – confirmed skills (the number of confirmed skills reflects the recognition of specific professional skills by other members of the person's network of contacts, emphasizing their reliability and competence in these areas);

Ec – number of endorsements (the total number of endorsements (skills) demonstrates the degree of recognition of the person's skills by colleagues, highlighting their influence and authority within their professional community);

R – assessment (expert level assessment. It is based on personal knowledge and professional interaction with each expert in the sample. The assessment reflects a reasoned judgment about the experts' competence, reliability, and overall level of expertise in the field of software evaluation).

Table 1 presents the summary indicators of the expert sample.

$$R_i = k_1 \cdot E_i + k_2 \cdot C_i + k_3 \cdot Ed_i + k_4 \cdot P_i + k_5 \cdot S_i + k_6 \cdot CL_i + k_7 \cdot L_i + k_8 \cdot RR_i + k_9 \cdot GR_i + k_{10} \cdot SK_i + k_{11} \cdot SkE_i + k_{12} \cdot Ec_i + k_{13}, \quad (1)$$

where $k_j, j = 1, 2, \dots, 13$ – unknown coefficients, the values of which must be calculated based on the analysis of the collected data by a group of experts; i – the variable in the model, depends on the amount of experimental data.

Let us compose a system of linear algebraic equations of the form (2).

$$\left\{ \begin{array}{l} R_1 = k_1 \cdot E_1 + k_2 \cdot C_1 + k_3 \cdot Ed_1 + k_4 \cdot P_1 + k_5 \cdot S_1 + k_6 \cdot CL_1 + k_7 \cdot L_1 + k_8 \cdot RR_1 + k_9 \cdot GR_1 + k_{10} \cdot SK_1 + k_{11} \cdot SkE_1 + k_{12} \cdot Ec_1 + k_{13}; \\ \dots \\ R_i = k_1 \cdot E_i + k_2 \cdot C_i + k_3 \cdot Ed_i + k_4 \cdot P_i + k_5 \cdot S_i + k_6 \cdot CL_i + k_7 \cdot L_i + k_8 \cdot RR_i + k_9 \cdot GR_i + k_{10} \cdot SK_i + k_{11} \cdot SkE_i + k_{12} \cdot Ec_i + k_{13}; \\ \dots \\ R_{24} = k_1 \cdot E_{24} + k_2 \cdot C_{24} + k_3 \cdot Ed_{24} + k_4 \cdot P_{24} + k_5 \cdot S_{24} + k_6 \cdot CL_{24} + k_7 \cdot L_{24} + k_8 \cdot RR_{24} + k_9 \cdot GR_{24} + k_{10} \cdot SK_{24} + k_{11} \cdot SkE_{24} + k_{12} \cdot Ec_{24} + k_{13}. \end{array} \right. \quad (2)$$

The solution of system (2) is the domain of model coefficients. We use the least squares method to find estimates of the coefficients of the SLAR (2) model, and we obtain, accordingly, the following model (3): the variable in the model depends on the amount of experimental data.

We compose a system of linear algebraic equations of the form (2).

$$R_i = 0.019 \cdot E_i + 0.057 \cdot C_i + 0.09 \cdot Ed_i + 0.163 \cdot P_i + 0.068 \cdot S_i + 0.984 \cdot CL_i + 0.147 \cdot L_i + 0.219 \cdot RR_i - 0.6 \cdot GR_i - 0.029 \cdot SK_i + 0.066 \cdot SkE_i - 0.001 \cdot Ec_i + 3.289. \quad (3)$$

Fig. 2 presents the results of checking the adequacy of the mathematical model (3) taking into account a 15% deviation corridor of the expert ratings, so that the element of subjectivity of the assessment is taken into account.

As we can see, the mathematical model for calculating the expert's competence level demonstrates high performance when applied to a calibration sample. It also successfully determines the assessment of the competence level of an individual representative from the LinkedIn system. Additionally, experimental studies were conducted to reduce the number of input parameters from 12 to a

smaller number, but all potential simplified models showed significantly worse predictive results. Models built using only seven or fewer criteria demonstrated significant deviations from the initially defined ones, with errors often exceeding 30-40% of the estimates obtained from real data.

This rapid decrease in accuracy shows that there is a critical threshold below which the assessment structure cannot adequately reflect the multidimensional nature of the indicators of the "quality" of an expert in the field of software development. This study is only an initial stage in the study of expert competence levels, and further work will only develop and improve the results obtained.

Conclusions

As a result of the analysis of existing methods for assessing the professional competence of software evaluation specialists, it was found that most of them are characterized by a high level of subjectivity and insufficient formalization, which complicates their application in the conditions of the global labor market. It was found that professional social networks, in particular LinkedIn, contain a significant amount of quantitative data that can be used to objectively assess the level of competence of specialists, but existing solutions do not provide the proper level of their integration and analysis. A mathematical model for assessing the level of competence of software quality assessment experts has been developed, which takes into account a set of quantitative indicators from professional profiles, including work experience, implemented projects, technical skills, education, certifications and professional connections. Each indicator has an individual weight, which is determined on the basis of an expert analysis of its significance for a specific specialization in the field of software evaluation. The proposed model provides a multifactorial assessment of a specialist's competence, which allows you to obtain an objective idea of his professional level.

The practical value of the developed mathematical model is confirmed by the possibility of its use for solving various tasks in the field of human resources management in the field of software development. The model allows you to reduce the time for assessing the competence of specialists, minimize the influence of

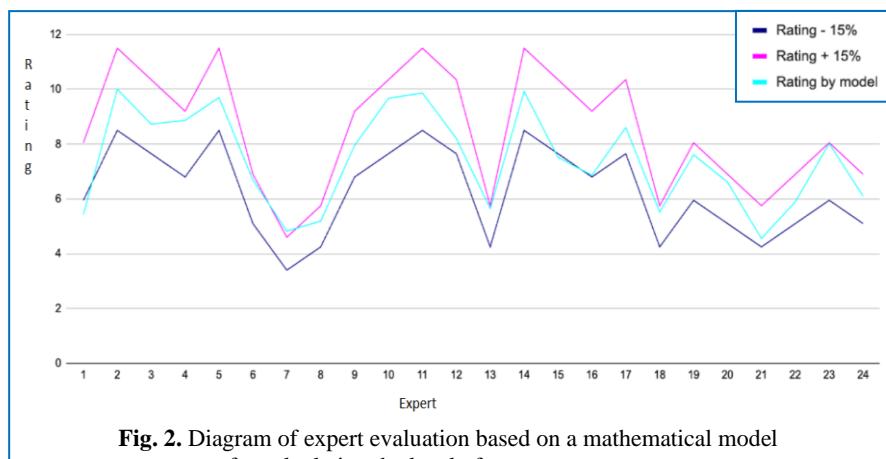


Fig. 2. Diagram of expert evaluation based on a mathematical model for calculating the level of expert competence

subjective factors when making personnel decisions, and optimize the distribution of human resources in software development and testing projects.

Conflicts of interest

The authors declare that they have no conflicts of interest in relation to the current study, including

financial, personal, authorship, or any other, that could affect the study, as well as the results reported in this paper.

Use of artificial intelligence

The authors confirm that they did not use artificial intelligence technologies when creating the current work.

REFERENCES

1. Yakovyna, V. and Symets, I. (2021), "Reliability assessment of CubeSat nanosatellites flight software by high-order Markov chains", *Procedia Computer Science*, vol.192, pp. 447–456, doi: <https://doi.org/10.1016/j.procs.2021.08.046>
2. , Scanda, V.C., Prasad, S.S., Dheemanth, G.R. and Kumar, N.S. (2019), "Assessment of quality of program based on static analysis", *IEEE 10th Int. Conf. on Techn. for Education (T4E)*, pp. 276–279, doi: <https://doi.org/10.1109/T4E.2019.00072>
3. Lu, S., Li, H. and Jiang, Z. (2020), "Comparative study of open source software reliability assessment tools", *IEEE Int. Conf. on Art. Intelligence and Inf. Systems (ICAIIS)*, China, pp. 49–55, doi: <https://doi.org/10.1109/ICAIIS49377.2020.9194946>
4. Yakovyna, V., Seniv, M., Symets, I. and Sambir, N. (2020), "Algorithms and software suite for reliability assessment of complex technical systems", *Radio Electronics, Computer Science, Control*, vol. 4, pp. 163–177, doi: <https://doi.org/10.15588/1607-3274-2020-4-16>
5. San, K.K., Washizaki, H., Fukazawa, Y., Honda, K., Taga, M. and Matsuzaki, A. (2021), "Deep cross-project software reliability growth model using project similarity-based clustering", *Mathematics*, vol. 9, no. 22, article number 2945, doi: <https://doi.org/10.3390/math9222945>
6. Spivak, I., Krepoch, S. and Budenchuk, S. (2018), "Methods and Means of Expert Evaluation of Software Systems on the Basis of Interval Data Analysis", 14th Int. Conference on Advanced Trends in Radioelectronics, Telecommunications and Computer Engineering, pp. 164–167, doi: <https://doi.org/10.1109/TCSET.2018.8336178>
7. Spivak, I., Krepoch, S. and Krepoch, R. (2018), "Construction of the Criterion for the Agree of Expert Groups Estimates Based on Analysis of Interval Data", International Scientific-Practical Conference Problems of Infocommunications Science and Technology, Kharkiv, pp. 261–264, doi: <https://doi.org/10.1109/INFOCOMMST.2018.8632053>
8. Spivak, I., Krepoch, S., Krepoch, R. and Bayurskii, A. (2019), "Construction of a Criterion for Assessing the Level of Objectivity of Experts Based on a Modified Interval Expert Appraisal Method", IEEE International Scientific-Practical Conference: Problems of Infocommunications Science and Technology, PIC S and T 2019 – Proceedings, Kyiv, pp. 311–314, doi: <https://doi.org/10.1109/PICST47496.2019.9061486>
9. Krepoch, S., Spivak, I. and Krepoch, R. (2018), "Research of the agree of experts' evaluations in the estimation of software systems", CEUR Workshop Proceedings 2300, pp. 203–206, available at: <https://ceur-ws.org/Vol-2300/Paper49.pdf>
10. King, G., Church, K., Schmidt, P. and Harnowo, A. (2024), "Using LinkedIn for Career Building", AIS Educator Journal, vol. 18, is. 1, pp. 65–76, doi: <https://doi.org/10.3194/1935-8156-18.1.65>
11. Ododo, E. P., Inyang, A. B. and Billy, D. (2024), "Impact of LinkedIn marketing on recruitment and selection processes in the IT industry", *The American Journal of Management and Economics Innovations*, no. 6. pp. 31–40, doi: <https://doi.org/10.37547/tajmei/Volume06Issue09-06>
12. Wu, C.-Y. and Huang, C.-Y. (2021), "A study of incorporation of deep learning into software reliability modeling and assessment", *IEEE Transactions on Reliability*, vol. 70, no. 4, pp. 1621–1640, doi: <https://doi.org/10.1109/TR.2021.3105531>
13. Jagtap, M., Katragadda, P. and Satelkar, P. (2022), "Software reliability: development of software defect prediction models using advanced techniques", *Annual Reliability and Maintainability Symposium (RAMS)*, pp. 1–7, doi: <https://doi.org/10.1109/RAMS51457.2022.9893986>
14. Nafreen, M., Luperon, M., Fiondella, L., Nagaraju, V., Shi, Y. and Wandji, T. (2020), "Connecting software reliability growth Models to software defect tracking", *IEEE 31st International Symposium on Software Reliability Engineering (ISSRE)*, pp. 138–147, doi: <https://doi.org/10.1109/ISSRE5003.2020.00022>
15. Rezanov, B., and Kuchuk, H. (2023), "Model of elemental data flow distribution in the Internet of Things supporting Fog platform", *Innovative Technologies and Scientific Solutions for Industries*, vol. 2023 (3), pp. 88–97, doi: <https://doi.org/10.30837/ITSSI.2023.25.088>
16. Jain, R. and Sharma, A. (2019), "Assessing software reliability using genetic algorithms", *The Journal of Engineering Research [TJER]*, vol. 16(1), pp. 11–17, doi: <https://doi.org/10.24200/tjer.vol16iss1>
17. Micro, R., Chren, S. and Rossi, B. (2022), "Applicability of software reliability growth models to open source software", *48th Euromicro Conference on Software Engineering and Advanced Applications (SEAA)*, pp. 255–262, doi: <https://doi.org/10.1109/SEAA56994.2022.00047>
18. Lu, S., Li, H. and Jiang, Z. (2020), "Comparative Study of Open Source Software Reliability Assessment Tools", 2020 IEEE International Conference on Artificial Intelligence and Information Systems (ICAIIS), Dalian, China, pp. 49–55, doi: <https://doi.org/10.1109/ICAIIS49377.2020.9194946>
19. Kim, T., Ryu, D. and Baik, J. (2024), "Enhancing software reliability growth modeling: a comprehensive analysis of historical datasets and optimal model selections", *IEEE 24th International Conference on Software Quality, Reliability and Security (QRS)*, pp. 147–158, doi: <https://doi.org/10.1109/QRS62785.2024.00024>
20. Solovei, O. (2025), "A weighted sensitivity metric for predicting latency in a KAFKA cluster", *Innovative technologies and scientific solutions for industries*, vol. 3(33), pp. 152–165, doi: <https://doi.org/10.30837/2522-9818.2025.3.152>
21. Nafreen, M., Luperon, M., Fiondella, L., Nagaraju, V., Shi, Y. and Wandji, T. (2020), "Connecting Software Reliability Growth Models to Software Defect Tracking", *IEEE 31st International Symposium on Software Reliability Engineering (ISSRE)*, Coimbra, Portugal, pp. 138–147, doi: <https://doi.org/10.1109/ISSRE5003.2020.00022>
22. Cao, W., Kosenko, V. and Semenov, S. (2022), "Study of the efficiency of the software security improving method and substantiation of practical recommendations for its use", *Innovative Technologies and Scientific Solutions for Industries*, vol. 1(19), pp. 55–64, doi: <https://doi.org/10.30837/ITSSI.2022.19.055>

23. Bayurskii, A. and Krepich, S. (2018), "Intelligent System Analyzing Quality of Land Plots", CEUR Workshop Proceedings 2300, pp. 166–169, available at: <https://ceur-ws.org/Vol-2300/Paper40.pdf>

24. Saini, G.L., Panwar, D. and Singh, V. (2021), "Software reliability prediction of open source software using soft computing technique", *Recent Advances in Computer Science and Communications (Formerly: Recent Patents on Computer Science)*, vol. 14, no. 2, pp. 612–621, doi: <https://doi.org/10.2174/221327591266190307165332>

25. Kuchuk, H., Mozhaiev, O., Kuchuk, N., Tiulieniev, S., Mozhaiev, M., Gnusov, Y., Tsuranov, M., Bykova, T., Klivets, S., and Kuleshov, A. (2024), "Devising a method for the virtual clustering of the Internet of Things edge environment", *Eastern-European Journal of Enterprise Technologies*, vol. 1, no. 9 (127), pp. 60–71, doi: <https://doi.org/10.15587/1729-4061.2024.298431>

26. Hrytsiuk, Yu. and Dalyavskyy, V. (2020), "Use of petal diagrams to visualize the results of expert evaluation of software quality", *Scientific Bulletin of NLTU of Ukraine*, vol. n28, issue 9, pp. 95–104, doi: <https://doi.org/10.15421/40280919>

27. Teuber, S. and Weigl, A. (2021), "Quantifying Software Reliability via Model-Counting", *Lecture Notes in Computer Science*, vol 12846, Springer, Cham, doi: https://doi.org/10.1007/978-3-030-85172-9_4.

Received (Надійшла) 11.09.2025

Accepted for publication (Прийнята до друку) 07.01.2026

ВІДОМОСТІ ПРО АВТОРІВ/ ABOUT THE AUTHORS

Крепич Світлана Ярославівна – кандидат технічних наук, доцент, доцент кафедри комп’ютерних наук, Західноукраїнський національний університет, Тернопіль, Україна;
Svitlana Krepich – Candidate of Technical Sciences, Associate Professor, Associate Professor of Computer Science Department, Western Ukrainian National University, Ternopil, Ukraine;
e-mail: msva220189@gmail.com; ORCID Author ID: <https://orcid.org/0000-0001-7700-8367>;
Scopus Author ID: <https://www.scopus.com/authid/detail.uri?authorId=55225606100>.

Співак Ірина Ярославівна – кандидат технічних наук, доцент, доцент кафедри комп’ютерних наук, Західноукраїнський національний університет, Тернопіль, Україна;
Iryna Spivak – Candidate of Technical Sciences, Associate Professor, Associate Professor of Computer Science Department, Western Ukrainian National University, Ternopil, Ukraine;
e-mail: spivak_iruna@gmail.com; ORCID Author ID: <https://orcid.org/0000-0003-4831-0780>;
Scopus Author ID: <https://www.scopus.com/authid/detail.uri?authorId=55226024100>.

Співак Сергій Михайлович – доктор економічних наук, професор, завідувач кафедри бухгалтерського обліку та аудиту, Тернопільський національний технічний університет імені Івана Пулюя, Тернопіль, Україна;
Serhii Spivak – Doctor of Economic Sciences, Professor, Head of the Accounting and Audit Department, Ternopil Ivan Puluj National Technical University, Ternopil, Ukraine;
e-mail: spivak_s@tnu.edu.ua; ORCID Author ID: <https://orcid.org/0000-0002-7160-2151>;
Scopus Author ID: <https://www.scopus.com/authid/detail.uri?authorId=57210559132>.

Крепич Роман Володимирович – викладач, ВСП Кам’янець-Подільський фаховий коледж НРЗВО "Кам’янець-Подільський державний інститут", Кам’янець-Подільський, Україна;
Roman Krepich – lector, SSD Kamianets-Podilskyi Vocational College ERIHE "Kamianets-Podilskyi State Institute", Kamianets-Podilskyi, Ukraine;
e-mail: jagmstar@gmail.com; ORCID Author ID: <https://orcid.org/0000-0003-4831-0780>;
Scopus Author ID: <https://www.scopus.com/authid/detail.uri?authorId=27368089600>.

Математична модель розрахунку рівня компетентності експерта

С. Я. Крепич, І. Я. Співак, С. М. Співак, Р. В. Крепич

Анотація. В умовах стрімкого розвитку інформаційних технологій якість програмного забезпечення набуває критичного значення для успішної діяльності організацій різних галузей. Зростаюча складність сучасних програмних рішень вимагає застосування фахівців з тестування та оцінки якості програмного забезпечення, здатних ефективно виявляти недоліки та забезпечувати відповідність продукту встановленим стандартам. Водночас оцінювання рівня компетентності таких експертів залишається складним завданням, яке часто ґрунтується на суб’єктивних критеріях та методах. Актуальність дослідження зумовлена гострою потребою сучасного IT-ринку в об’єктивних інструментах оцінювання професійного рівня фахівців, особливо у сфері забезпечення якості програмного забезпечення. Традиційні підходи до оцінки кваліфікації, як-от співбесіди, тестові завдання чи аналіз резюме, часто не дають повного та об’єктивного уявлення про компетентність експерта. Особливою гостроти ця проблема набуває в умовах глобального ринку праці, коли компанії змушені оцінювати фахівців дистанційно, спираючись лише на обмежений набір даних про їхній досвід та навички. На сьогодні програмне забезпечення перетворилося на невід’ємну складову багатьох сфер нашого повсякденного життя – від автоматизації і оптимізації процесів на виробництві до створення комфорту окремої людини. **Об’єктом дослідження** виступає процес визначення рівня компетентності експертів з оцінки якості програмного забезпечення. **Предметом дослідження** є математична модель розрахунку рівня компетентності експерта. Практична цінність результатів роботи визначається можливістю використання розробленої системи HR-менеджерами для ефективного підбору фахівців, керівниками QA-відділів для формування збалансованих команд тестування, сертифікаційними центрами для об’єктивної оцінки компетентності, а також самими експертами для планування власного професійного розвитку. **Висновок** розроблена математична модель розрахунку рівня компетентності експерта дозволяє скоротити час на оцінку компетентності фахівців, мінімізувати вплив суб’єктивних факторів при прийнятті кадрових рішень та оптимізувати розподіл людських ресурсів у проектах з розробки та тестування ПЗ.

Ключові слова: експертне оцінювання; тестування; якість програмного забезпечення; математичні методи оцінювання; модель.