

**Ministry of Education and Science of Ukraine**  
**Ukrainian-American Concordia University**  
Faculty of Management and Business  
*Department of International Economic Relations, Business &  
Management*

MASTER'S QUALIFICATION WORK

**ARTIFICIAL INTELLIGENCE IMPACT  
ON RISK MANAGEMENT  
(based on “COR-Medical” case)**

Master student of the 2<sup>nd</sup> year of study

Field of Study 07 – Management

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**Oleksandr Dubei**

Research supervisor

**Natalya Amalian**

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## **Abstract**

This work embarks on a journey to explore the multifaceted impact of AI on risk management, delving into its implications across various domains, from strategic planning and organizational behavior to leadership dynamics and ethical considerations. Through a synthesis of theoretical frameworks, empirical studies, and real-world case analyses, this thesis endeavors to illuminate the transformative potential of AI in reshaping risk management paradigms in the pharmaceutical field while critically examining the implications for organizational dynamics, managerial roles, and the future of work.

When doing the research, the materials of LLC "COR-Medical" were used, namely, the overall analysis of the company, its nature, structure, and business model. The work draws conclusions about the overall positive impact of AI implementation in risk management, enhancing and optimizing already existing strategies or inventing new ones.

In essence, by anchoring our exploration in the intersection of risk management, innovation, and AI, this work endeavors to not only deepen our understanding of these interconnected domains but also empower practitioners and scholars alike to embrace AI as a catalyst for transformative change in risk management practices.

**Keywords:** risk management, artificial intelligence, pharmaceutical field.

## **Анотація**

Ця робота розпочинає подорож для дослідження багатогранного впливу штучного інтелекту на управління ризиками, поглиблюючи його наслідки у різних сферах, від стратегічного планування і організаційної поведінки до динаміки лідерства та етичних вагань. Через синтез теоретичних каркасів, емпіричних досліджень та аналізу реальних випадків, ця дисертація прагне висвітлити трансформаційний потенціал штучного інтелекту у переформатуванні парадигм управління ризиками в фармацевтичній галузі, критично досліджуючи наслідки для організаційних динамік, менеджерських ролей та майбутньої роботи.

Під час проведення дослідження використовувалися матеріали ТОВ "КОР-Медікал", зокрема, загальний аналіз компанії, її природа, структура, бізнес-модель.

Робота робить висновки про загальний позитивний вплив впровадження штучного інтелекту в управління ризиками, підвищення і оптимізація вже існуючих стратегій або винаходження нових.

У сутності, закріпивши наше дослідження в перетині управління ризиками, інновації та штучного інтелекту, ця робота прагне не тільки поглибити наше розуміння цих взаємопов'язаних областей, але й надати практикам та вченим можливість прийняти штучний інтелект як каталізатор трансформаційних змін у практиках управління ризиками.

***Ключові слова:*** управління ризиками, штучний інтелект, фармацевтична сфера.

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**APPROVED**

Head of Department



**Prof. Zharova L.V.**

“

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**TASK**

**FOR MASTER’S QUALIFICATION WORK OF STUDENT**

**Oleksandr Dubei**

1. Topic of the master’s qualification work

**ARTIFICIAL INTELLIGENCE IMPACT  
ON RISK MANAGEMENT**

(based on “COR-Medical” case)

**Consultant of the master’s qualification work Natalya Amalian, Ph.D. in Economics,**

Which approved by Order of University from **“10” February 2024 № № 10-02/2024-1к**

2. Deadline for master thesis submission **“25” April 2024.**

3. Data-out to the master thesis

*Materials from internship received during consultation with representatives of the company. Information from open resources in the Internet, official reporting of financial and economic activities of the enterprise.*

4. Contents of the explanatory note (list of issues to be developed)

***There are three main topics a student should develop in this work:***

- 1. Theoretical framework of risk management within innovation projects***
- 2. Integration of artificial intelligence in risk management for innovation projects in pharmaceutical field***

### ***3. Recommendations for implementation of AI in risk management for innovation projects in pharmaceutical field***

5. List of graphic material (with exact indication of any mandatory drawings)

*Graphs and figures for analysis of economical and statistical information on the company and its development, visualization of mechanism of development, etc.*

6. Date of issue of the assignment December 4, 2023

#### **Time Schedule**

№	The title of the parts of the qualification paper (work)	Deadlines	Notes
1.	I part of master thesis	01.03.2024	<i>In time</i>
2.	II part of master thesis	20.03.2024	<i>In time</i>
3.	III part of master thesis	20.04.2024	<i>In time</i>
4.	Introduction, conclusions, summary	25.04.2024	<i>In time</i>
5.	Pre-defense of the thesis	30.04.2024	<i>In time</i>

Student \_\_\_\_\_

  
(signature)

Consultant 

**Conclusions.** *The master qualification work was designed according to the requirements: it contains all necessary parts of scientific research with the practical recommendations. The paper was written on the basis of the analysis of particular aspects of the risk management at the pharmaceutical companies in general, and in COR Medical – in particular. The study provides a meticulous analysis of methodology, technology, and tools of risk management and the possibility of integration of artificial intelligence in risk management. On the basis of the examination of COR Medical company the author focuses on the problem of alignment of AI impact on risk management with risk management strategies in innovation projects in pharmaceutical field. The practical recommendations aimed at Improvement of Risk Management in Innovation Projects in pharmaceutical field using AI Integration are formulated correctly and focused on the main goal and tasks of the work. Student takes active part in scientific life of the University, preparing for publication article for BESLI. In general, if successful defense, the thesis can claim to be "excellent".*

Consultant 

**Natalya Amalian**

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## INTRODUCTION

In an era where technology continuously reshapes the landscape of industries, the advent of Artificial Intelligence (AI) stands as a monumental stride toward innovation and efficiency. As organizations navigate the complexities of the modern world, the integration of AI into management practices has emerged as a pivotal phenomenon, promising to revolutionize the way businesses operate, strategize, and lead.

At its core, management encapsulates the art of orchestrating resources, people, and processes to achieve organizational objectives. Historically, this realm relied heavily on human intuition, experience, and foresight. However, the dawn of AI introduces a paradigm shift, augmenting traditional managerial approaches with computational prowess and data-driven insights. The allure of AI lies in its ability to process vast amounts of data at unprecedented speeds, unveiling patterns, correlations, and trends that elude human cognition. From predictive analytics to machine learning algorithms, AI empowers managers with foresight, enabling them to anticipate market shifts, customer preferences, and operational bottlenecks with unparalleled precision.

Beyond predictive capabilities, AI catalyzes automation, liberating managers from mundane tasks and empowering them to focus on strategic initiatives. Tasks that once demanded tedious manual labor are now delegated to algorithms, freeing human capital to engage in higher-order thinking, creativity, and innovation.

Yet, the integration of AI into management practices is not without its challenges and ethical considerations. As algorithms dictate decisions, concerns regarding transparency, accountability, and bias loom large. Moreover, the human element remains indispensable, as effective management encompasses empathy, judgment, and interpersonal skills and qualities that algorithms, no matter how advanced, struggle to replicate.

Thus, the intersection of AI and management presents a nuanced landscape, fraught with opportunities and dilemmas alike. This thesis embarks on a journey to explore the multifaceted impact of AI on risk management, delving into its implications across various domains, from strategic planning and organizational behavior to leadership dynamics and ethical considerations.

Through a synthesis of theoretical frameworks, empirical studies, and real-world case analyses, this thesis endeavors to illuminate the transformative potential of AI in reshaping risk management paradigms in pharmaceutical field while critically examining the implications for organizational dynamics, managerial roles, and the future of work.

As we embark on this exploration, it becomes evident that the fusion of AI and management heralds a new chapter in the evolution of organizational practices, one characterized by innovation, adaptability, and the relentless pursuit of excellence. By understanding and harnessing the power of AI, managers can navigate the complexities of the digital age with agility and insight, steering their organizations toward sustained growth, resilience, and prosperity.

First and foremost, before starting the journey of learning about the impact of Artificial Intelligence, let's outline certain objectives to make the road more fluid and controlled. the **object** of this master's qualification work is the Impact of Artificial Intelligence on Risk Management. Secondly, we have a **subject**, which is the "COR-Medical" company. The research will be completed by using different **research methods** depending on the topic, as some parts of the work may require qualitative research method, some may need quantitative instead. During the work, several others methods will be used, for example, survey, literature review, case study. Moving forward, we need to outline several tasks for this work:

The **first task** is to focus on the theoretical framework of risk management of innovation projects in pharmaceutical field. This task will be achieved via Chapter 1, by providing the necessary foundation for understanding the topic of risk management of the innovation projects, most whats and ifs, different ways of improvement and gradually flow into the topic of how Artificial Intelligence is one of the most prominent ways of improvement in our current digital age.

Let's not stop here and continue outlining the next tasks. The **second task** is to delve into common information on the topic of Risk Assessment and Identification to help the reader better understand further chapters by gaining the necessary terms and their definitions. And this task should also be focused on a component of risk management

known as proactive risk mitigation which will be completed by exploring the traditional strategies of enhancing risk management to understand the already existing ways of managing risks in innovative projects.

The **third task** is to understand the concept of “Stakeholder Engagement” and define why it is considered to be one of the crucial components of risk management.

The **fourth task** is to analyze and understand the most common methodology of Risk Management, different technologies and tools that are used in Risk Management, and also how they were successfully used in existing companies.

Moving on, we have the **fifth task**, which is to outline different approaches on how AI solutions are integrated into companies that will be useful further.

The **sixth task** is to conduct a research on “COR-Medical” company, to gain a better understanding of the current state of the company and use gathered information in the next chapter, using it as a basis for recommendations and potential improvements.

The **seventh task** is to get a broader understanding of such concept as “AI in Risk Management”. This task can be achieved by conducting a survey and a research using Ukrainian enterprises as sources of information and other perspectives from our researched subject.

The **eighth task** is to explain and provide direct examples that can be used as recommendations, and also to give clearly defined and substantiated recommendations to effectively address the challenges and harness the opportunities presented by AI integration in risk management for innovation projects in pharmaceutical field. It is worth to mention, that using already existing and commonly used AI solutions in Risk Management and explaining how they are used will give a clear picture on how said solutions can improve different parts of Risk Management in pharmaceutical field.

The last, **ninth task** of this work is to give direct recommendations for organization’s activities while aiming at achieving optimality, target efficiency, and practical implementation. This task can be achieved by using both gained knowledge during the learned and successfully completed courses, and everything that have been

gathered, researched, and analyzed in this work to reach its culmination.

Now, it is important to establish why the choice of uncovering the topic of the master's qualification work was decided to be conducted using the framework of risk management within innovation projects. The decision to focus on the theoretical framework of risk management within innovation projects serves as a strategic starting point for delving into the profound impact of Artificial Intelligence (AI) on management practices. By laying the groundwork in Chapter 1, we establish a solid understanding of the intricacies of risk management, exploring various methodologies, challenges, and opportunities for improvement. Moreover, the choice to examine AI's role in enhancing risk management practices within organizations stems from its undeniable prominence as a transformative force in the digital age. As it will be explained in Chapter 2, AI presents unparalleled opportunities to revolutionize risk management processes, leveraging advanced analytics, predictive modeling, and automation to mitigate uncertainties and optimize decision-making. By anchoring our exploration in the realm of risk management within innovation projects, we not only address a pressing organizational need but also unlock broader insights into the evolving landscape of management practices in the era of AI. This deliberate focus allows us to uncover practical strategies, empirical evidence, and theoretical frameworks that illuminate the symbiotic relationship between AI adoption and effective management. Going on, Chapter 3 will serve as a pivotal juncture where theoretical insights converge with practical recommendations, offering actionable proposals and models for managing the development of AI-driven risk management initiatives. Through a synthesis of gathered information, conducted research, and real-world case studies, this recommendatory section provides a roadmap for organizations seeking to harness the power of AI to navigate the complexities of risk management in innovation projects.

In essence, by anchoring our exploration in the intersection of risk management, innovation, and AI, this thesis endeavors to not only deepen our understanding of these interconnected domains but also empower practitioners and scholars alike to embrace AI as a catalyst for transformative change in management practices. Through meticulous analysis

and strategic inquiry, we embark on a journey to uncover the profound implications of AI's impact on management and chart a course toward a future where innovation thrives and risks are met with resilience and foresight. So let's embark on the journey.

## **CHAPTER 1. THEORETICAL FRAMEWORK OF RISK MANAGEMENT WITHIN INNOVATION PROJECTS**

Innovation, the engine of progress and growth, has become the lifeblood of modern economics. Organizations across various sectors are continually challenged to develop and implement innovative projects to stay competitive and meet the ever-evolving needs of their markets. However, the pursuit of innovation is not without its perils. Innovation projects, by their very nature, are fraught with risks and uncertainties that can jeopardize their success. Effectively managing these risks is a pivotal concern for organizations, and the evolution of risk management in innovation projects has garnered significant attention in recent years.

Risk management in the context of innovation projects extends beyond the conventional paradigms of project management. It encompasses a dynamic interplay of technical, market, financial, and strategic risks, necessitating a multifaceted approach. Innovation, while promising in its potential for groundbreaking success, also poses complex challenges due to the high degree of novelty and unpredictability associated with it.

This theoretical section embarks on a comprehensive exploration of the ways to enhance risk management in the context of innovation projects. It seeks to unravel the strategies, methodologies, and best practices that organizations can employ to mitigate the risks inherent in innovation and to promote a culture of informed and proactive risk management.

The section is structured in a way to provide insights into various dimensions of risk management in innovation projects, encompassing risk assessment and identification, proactive risk mitigation, stakeholder engagement, agile project management, the role of technology and tools, knowledge management, and real-world case studies. By synthesizing the body of knowledge and research in these areas, we aim to equip innovators and project managers with the knowledge and tools they need to navigate the challenging terrain of innovation while safeguarding their investments and maximizing the potential for success.

In the following sections, we delve into each of these aspects in detail, drawing from

a comprehensive range of scholarly works and industry practices. By doing so, we aspire to contribute to a deeper understanding of how risk management can be enhanced within the context of innovation projects, thereby assisting organizations and innovators in realizing their ambitions with greater confidence and clarity.

### ***1.1 Crucial Components of Risk Management.***

Risk assessment and identification are foundational steps in the effective management of risk within innovation projects. To mitigate the diverse uncertainties inherent in these projects, organizations must employ robust methodologies and tools for risk analysis. These approaches encompass both qualitative and quantitative techniques and help organizations understand and categorize potential risks (Hillson & Simon, 2007).

*Qualitative Risk Assessment.* One commonly employed method for risk assessment in innovation projects is the qualitative approach. This method is a fundamental component of risk management in innovation projects and relies on subjective evaluation and expert judgment to identify and evaluate potential risks. This method offers valuable insights into potential challenges and uncertainties early in the project lifecycle, enabling proactive risk management. Qualitative risk assessment is resource-efficient, making it particularly useful for swiftly prioritizing risks and allocating resources for further analysis (Project Management Institute, 2013). It involves methodologies such as risk probability and impact assessment, where risks are categorized based on their likelihood and potential impact (Hillson & Simon, 2007). Expert judgment, a crucial component of qualitative assessment, taps into the knowledge and expertise of experienced team members within the organization (Chapman & Ward, 2003). Additionally, risk rating scales are employed to assign numerical values to risks, facilitating ranking and prioritization. This approach empowers project managers to make informed, strategic decisions by providing an initial understanding of risks and their potential implications (Hillson & Simon, 2007). Ultimately, qualitative risk assessment serves as the cornerstone for effective risk management, aiding in the early identification and assessment of risks to ensure innovation projects are well-prepared for potential challenges and uncertainties.

*Quantitative Risk Assessment.* Quantitative risk assessment, on the other hand,

involves a more systematic and data-driven analysis. It is a pivotal element of innovation project risk management, employs numerical data and probabilistic models to evaluate potential risks quantitatively. This approach provides a deeper and more precise understanding of risk exposure, aiding in data-driven decision-making. Quantitative risk assessment is founded on robust methodologies such as Monte Carlo simulations, which generate multiple scenarios based on input variables to predict project outcomes and quantify risks (Hillson & Simon, 2007). Additionally, decision trees are employed to model the sequential impact of decisions and uncertainties, offering insights into optimal choices within a risk context (Project Management Institute, 2013). Sensitivity analysis assesses the impact of variations in input parameters on project outcomes, identifying which variables significantly influence risk (Heldman, 2018). By leveraging historical data, expert opinions, and statistical models, quantitative risk assessment offers a quantitative estimate of risk exposure and serves as a powerful tool for optimizing resource allocation and risk response planning (Kendrick, 2015). In summary, quantitative risk assessment enhances risk management in innovation projects by providing a data-driven approach to identifying, quantifying, and mitigating risks, ultimately contributing to better-informed decision-making and project success.

*Failure Mode and Effects Analysis (FMEA).* Another valuable tool for risk assessment in innovation projects is Failure Mode and Effects Analysis (FMEA). FMEA is a comprehensive and systematic methodology used for risk assessment and identification in innovation projects. It entails identifying potential failure modes in a process, product, or system, evaluating their effects, and prioritizing them based on severity, occurrence, and detection (Besterfield, 2016). FMEA is grounded in the principle of prevention rather than correction, emphasizing early risk mitigation. This approach is invaluable in innovation projects as it allows teams to identify vulnerabilities, assess their potential impacts, and develop risk response strategies. Moreover, it aids in optimizing resource allocation by focusing efforts on mitigating the most critical risks. FMEA's structured nature makes it particularly effective for addressing complex systems, where identifying risks and their potential effects can be challenging (Project Management Institute, 2013). By systematically examining potential failure modes and their consequences, FMEA plays

a critical role in ensuring that innovation projects are well-prepared to address uncertainties and challenges.

*Continuous Risk Management.* In innovation projects, risk assessment is not a one-time task; it is an ongoing process. Continuous risk monitoring is essential to ensure that new risks are identified and assessed as the project progresses. CRM is an evolving and dynamic approach to risk assessment and identification that acknowledges the fluid nature of innovation projects. Unlike traditional risk management, which is often confined to initial project phases, CRM is a proactive strategy that involves ongoing risk assessment and adaptation throughout the project's lifecycle. CRM aims to enhance decision-making by continuously identifying, assessing, and mitigating risks in real time. It leverages approaches like risk radar, which is an ongoing monitoring system that tracks potential risks as they emerge (Kendrick, 2015). This approach helps project teams stay vigilant, ensuring that they respond promptly to changing risk landscapes. Additionally, CRM integrates tools like risk trigger identification, where specific triggers are set to initiate predefined risk responses (Schmidt & Lyytinen, 2002). The flexible and adaptive nature of CRM is particularly well-suited for innovation projects, which are inherently uncertain and subject to evolving conditions and external influences. By embracing CRM, project teams can proactively manage risks as they evolve, ensuring that innovation projects remain on course and adaptive to changing circumstances.

*Integration into Project Management.* Integration of risk assessment and identification into project management is a critical aspect of ensuring the success of innovation projects. It involves the seamless inclusion of risk management processes throughout the project lifecycle, aligning them with project objectives, scope, and schedule. Integration enhances project management by offering a systematic approach to risk identification and proactive risk response planning. Project managers can employ various methodologies, including Monte Carlo simulations and decision trees, to assess risks and make data-driven decisions to optimize resource allocation (Chapman & Ward, 2003). The adoption of a Risk Breakdown Structure (RBS), an essential component of integration, allows project teams to categorize and hierarchically structure risks, making them easier to manage (Heldman, 2018). Furthermore, integration requires regular risk

monitoring and reporting mechanisms to keep stakeholders informed and provide the project team with real-time data on risk exposure (Kendrick, 2015). It also involves linking risk management with other project management processes, such as scope change control, quality management, and procurement management, ensuring that risks are comprehensively addressed. Successful integration of risk assessment and identification into project management is fundamental to innovation project success, enabling teams to make informed decisions, allocate resources efficiently, and mitigate risks proactively and systematically.

To sum up this part, risk assessment and identification form the basis of risk management in innovation projects. Employing both qualitative and quantitative methods, as well as tools like FMEA, organizations can proactively identify and assess potential risks. Additionally, continuous risk monitoring ensures that evolving risks are addressed promptly. This comprehensive approach to risk assessment sets the stage for informed decision-making and targeted risk mitigation strategies throughout the innovation project's lifecycle.

Proactive risk mitigation is a critical aspect of innovation project risk management. Unlike a reactive approach, which addresses risks as they emerge, proactive mitigation strategies aim to anticipate potential issues and take measures to minimize their impact. This section explores key strategies for proactively mitigating risks in innovation projects, drawing from a range of scholarly sources and industry practices.

*Risk Response Plans.* One fundamental component of proactive risk mitigation in innovation projects is the development of comprehensive risk response plans. Risk response plans play a pivotal role in proactive risk mitigation within innovation projects. These plans are instrumental in addressing identified risks and ensuring that strategies are in place to manage potential challenges. The process involves leveraging methodologies such as risk avoidance, risk transfer, risk mitigation, and risk acceptance (Kerzner, 2017). Risk avoidance strategies aim to eliminate the risk by altering the project's scope or approach, while risk transfer may involve insurance, warranties, or contracts to shift the risk to third parties (Heldman, 2018). Risk mitigation strategies focus on reducing the probability or impact of identified risks through various means, including process changes

or technical solutions (Schmidt & Lyytinen, 2002). Risk acceptance, the conscious decision to not take action against a risk, can be a valid strategy when the cost of mitigation outweighs the potential impact. An effective risk response plan, informed by historical data, expert input, and quantitative analysis, ensures that innovation projects are well-prepared to navigate uncertainties and challenges, ultimately contributing to their success.

*Risk Transfer.* In some cases, it may be advantageous for organizations to transfer risk to external entities. This can be achieved through mechanisms such as insurance policies and contractual agreements (Schwartz & Lavis, 2017). By transferring risk to third parties, organizations can reduce their exposure and potential financial losses in the event of a risk materializing. Now about the term. Risk transfer is a fundamental strategy in proactive risk mitigation for innovation projects, serving as a means to shift identified risks to third parties or insurance mechanisms. This approach can significantly reduce the financial and operational impact of potential challenges. One widely recognized method of risk transfer is through the use of insurance (Heagney, 2016). Insurance policies can cover various types of risks, from property damage to liability claims, and are particularly valuable in innovation projects where uncertainties are prevalent. In addition to insurance, contracts and agreements can be structured to transfer specific risks to suppliers, partners, or subcontractors (Smith, 2013). The allocation of risks to third parties can include penalty clauses, warranties, or indemnification agreements, which delineate the responsibilities and liabilities of each party involved. By employing risk transfer strategies, innovation projects can minimize the financial and operational impact of potential risks and ensure that the organization's assets and objectives remain protected.

*Contingency Measures.* Contingency measures are an integral part of proactive risk mitigation. These measures constitute an essential aspect of proactive risk mitigation in innovation projects. These measures are designed to prepare project teams for the potential occurrence of identified risks and to establish predefined responses to manage those risks effectively. In the context of innovation projects, the adoption of contingency measures can take various forms. Risk contingency reserves are established to allocate additional budget or resources for addressing unforeseen events (Schwalbe, 2018). Similarly,

contingency plans outline the actions that will be taken if a specific risk materializes (Marchewka, 2015). For instance, in software development projects, contingency measures might include alternative development approaches or strategies for fast-tracking project timelines (Marchewka, 2015). The application of contingency measures ensures that innovation projects remain adaptable and resilient when faced with the unpredictabilities inherent in such endeavors, ultimately enhancing the project's chances of success.

*Regular Risk Reviews.* Proactive risk mitigation necessitates ongoing risk reviews and assessments throughout the project lifecycle (Schwartz & Lavis, 2017). Regular risk reviews, a critical element of proactive risk mitigation, are instrumental in maintaining project success by continuously evaluating and adapting to potential risks. These reviews provide a structured framework for the systematic examination of identified risks, their current status, and the effectiveness of mitigation strategies. They are essential for remaining vigilant in the face of evolving project conditions and uncertainties, particularly in the context of innovation projects where unforeseen challenges are common. Regular risk reviews involve assessing risk triggers and warning signs, which act as indicators of potential risk materialization (Kendrick, 2015). By closely monitoring these triggers, project teams can take timely, pre-defined actions to mitigate risks or adapt to changing conditions. Furthermore, risk reviews support informed decision-making by providing up-to-date information on the risk landscape, allowing project managers to allocate resources effectively and make strategic adjustments (Schwalbe, 2018). These reviews foster a culture of vigilance, preparedness, and adaptability, crucial attributes for innovation project success.

*Cross-Functional Teams.* These teams are crucial elements of proactive risk mitigation in innovation projects and promote a multidisciplinary approach to identifying, assessing, and addressing potential risks. These teams consist of members from various departments or expertise areas, each contributing their unique perspectives and knowledge to the risk management process. This collaborative approach can be particularly effective in addressing complex and multifaceted risks that often arise in innovation projects (Stark, 2015). Cross-functional teams foster creative problem-solving by integrating diverse

viewpoints, enabling a comprehensive analysis of potential risks and their mitigation strategies (Schmidt & Lyytinen, 2002). Additionally, this approach enhances risk identification by encouraging open communication and information sharing across different areas of expertise (Marchewka, 2015). By bringing together a diverse group of individuals, cross-functional teams enhance the project's ability to proactively address risks, contributing to its overall success and adaptability in the face of uncertainties.

After analyzing different studies, we can say that proactive risk mitigation is an indispensable component of risk management in innovation projects. Strategies such as the development of risk response plans, risk transfer, contingency measures, regular risk reviews, and cross-functional teams contribute to organizations' ability to anticipate, address, and minimize potential risks. These proactive measures not only enhance risk management but also improve the overall resilience of innovation projects, thereby increasing the likelihood of successful outcomes.

Stakeholder engagement is a pivotal aspect of effective risk management in innovation projects. Engaging with project stakeholders not only enriches the decision-making process but also plays a fundamental role in risk identification and mitigation. In this section, we explore the significance of stakeholder engagement in innovation projects and draw upon scholarly sources to elucidate its impact.

*Diverse Perspectives.* Innovation projects, by their very nature, are multifaceted and often involve a myriad of technical, market, and strategic risks. Diverse perspectives within stakeholder engagement are a cornerstone of successful innovation projects, as they foster a rich exchange of ideas and insights that can drive project success. Diverse perspectives are derived from stakeholders with varying backgrounds, experiences, and expertise, and they provide a holistic view of potential risks and opportunities (Heldman, 2018). This diversity includes internal stakeholders such as project team members from different departments, external stakeholders like customers, vendors, and regulatory bodies, and individuals from diverse demographic and cultural backgrounds (Turner & Müller, 2005). Such diversity ensures that innovation projects consider a wide array of factors, leading to more robust risk assessments and response plans. It also enhances stakeholder buy-in and commitment, as individuals feel their perspectives are valued and

integrated into project decisions, ultimately contributing to project success (Kerzner, 2017). In essence, the incorporation of diverse perspectives within stakeholder engagement is an essential component of innovation project risk management, enriching the decision-making process and helping project teams proactively address risks.

*Risk Identification.* Effective stakeholder engagement is a linchpin in innovation project risk management, facilitating the identification of potential risks and contributing to proactive risk mitigation. It involves gathering input from a wide array of stakeholders, such as project team members, customers, suppliers, regulatory bodies, and experts from relevant domains (Turner & Müller, 2005). This multi-dimensional engagement approach encourages the sharing of diverse perspectives and insights, which is instrumental in identifying a broader spectrum of risks and potential challenges. Through active dialogue and information exchange, stakeholders can uncover hidden risks, emerging issues, and unanticipated consequences, providing a comprehensive foundation for risk identification (Marchewka, 2015). Moreover, engaging stakeholders fosters a culture of vigilance, making it more likely for risks to be brought to the forefront in a timely manner, enabling project teams to develop risk response plans before challenges become critical (Heldman, 2018). By embracing stakeholder engagement as a risk identification strategy, innovation projects can leverage the collective wisdom and expertise of various stakeholders, ensuring a more holistic approach to risk assessment and ultimately enhancing the project's ability to proactively manage risks.

*Risk Prioritization.* Risk prioritization within stakeholder engagement is a pivotal step in innovation project risk management, enabling project teams to focus their resources and efforts on the most critical risks. It involves collaboratively assessing identified risks with stakeholders and assigning priorities based on their potential impact and likelihood (Kloppenborg, 2015). By involving a diverse group of stakeholders, including project team members, customers, and subject matter experts, a more comprehensive understanding of risks and their relative significance is achieved (Pinto & Slevin, 1998). This collaborative approach ensures that the prioritization process benefits from the collective wisdom and expertise of those with a stake in the project's outcome. In addition to expert opinions, quantitative methodologies like the Risk Priority Number (RPN) or decision matrices can

be employed to assign numerical values to risks and facilitate objective prioritization (Hillson & Simon, 2007). Through stakeholder engagement, project teams gain valuable insights into the potential consequences and strategic importance of risks, allowing them to allocate resources effectively and develop targeted risk response plans that enhance the project's ability to proactively manage risks and achieve its objectives.

*Risk Mitigation Strategies.* Effective stakeholder engagement in innovation projects is closely tied to the development and implementation of robust risk mitigation strategies. These strategies are pivotal in addressing identified risks and ensuring that the project remains on course. Engaging stakeholders with diverse perspectives and expertise is essential for shaping and validating these mitigation plans. Furthermore, incorporating their feedback can enhance the quality and relevance of risk response strategies. A combination of strategies is often used, including risk avoidance (Heagney, 2016), which involves altering project plans or approaches to steer clear of high-impact risks. Risk mitigation strategies such as risk reduction (Marchewka, 2015) entail taking proactive measures to minimize the likelihood or impact of risks. Additionally, risk transfer (Smith, 2013) can involve contractual arrangements to shift specific risks to third parties, and contingency plans that outline predefined responses to address unforeseen risks (Schwalbe, 2018). The involvement of stakeholders in shaping these strategies enhances their effectiveness and fosters a sense of shared responsibility for risk management, ultimately contributing to project success.

*Risk Awareness.* Risk awareness, as an integral component of stakeholder engagement in innovation projects, emphasizes the importance of ensuring that stakeholders are well-informed about potential risks and their implications. This aspect of engagement not only encourages transparency but also empowers stakeholders to contribute their insights and expertise effectively. When stakeholders are made aware of potential risks and their consequences, they can offer valuable input, identify risks that may have been overlooked, and actively participate in the risk identification and mitigation process (Schmidt & Lyytinen, 2002). This collaborative approach ensures that the project benefits from the collective vigilance and knowledge of various stakeholders, enhancing the project's ability to proactively manage risks. Furthermore, risk awareness

fosters a sense of shared responsibility and commitment among stakeholders, as they understand the potential impact of risks and the importance of mitigating them (Turner & Müller, 2005). By making risk awareness an integral part of stakeholder engagement, innovation projects create a culture of vigilance and preparedness, ultimately contributing to project success.

*Effective Communication.* Effective communication within stakeholder engagement is a fundamental element of innovation project risk management, fostering understanding, collaboration, and the alignment of project goals. It enables project teams to convey risk-related information clearly, ensuring that stakeholders are aware of potential risks and their implications (Verzuh, 2015). Effective communication strategies such as risk reporting mechanisms (Heagney, 2016) provide stakeholders with regular updates on the project's risk landscape, facilitating a transparent and informed decision-making process. Furthermore, active listening and feedback mechanisms (Kloppenborg, 2015) encourage stakeholders to share their concerns, insights, and recommendations regarding identified risks. A collaborative exchange of information ensures that diverse perspectives are taken into account and that the risk management process is enriched by the collective wisdom of stakeholders (Pinto & Slevin, 1998). Additionally, the use of visual aids and data visualization techniques (Schwalbe, 2018) can enhance the clarity and impact of risk-related messages, making it easier for stakeholders to grasp the implications of potential risks. Effective communication not only enhances risk awareness but also fosters a culture of transparency, trust, and commitment among stakeholders, ultimately contributing to the project's ability to proactively manage risks and achieve its objectives.

We can say that stakeholder engagement is one of the critical components regarding risk management in innovation projects. Engaging with diverse stakeholders enriches risk identification, prioritization, and mitigation strategies. It fosters a culture of risk awareness and transparency, ultimately enhancing the organization's ability to manage the multifaceted and dynamic risks inherent in innovation projects.

## ***1.2 Common Methodology, Technology, and Tools of Risk Management.***

Agile project management methodologies have emerged as effective approaches for

handling innovation projects. Agile methodologies, such as Scrum and Lean, have gained prominence due to their inherent flexibility, iterative nature, and adaptability to changing project dynamics (Schwabe, 2004). They promote continuous reassessment of project priorities, frequent reevaluation of project objectives, and ongoing adaptation to evolving requirements. This approach aligns closely with the dynamic nature of innovation projects, making Agile project management an attractive choice for managing risks in these ventures. One key aspect of Agile project management is its focus on dynamic risk assessment and management. Agile teams regularly assess and reassess risks, responding to emerging challenges quickly and efficiently. This proactive approach enables teams to address uncertainties and pivot when necessary, minimizing the impact of potential risks. On another note, they also advocate for close collaboration among team members and stakeholders, fostering a culture of transparency and continuous feedback (Schwabe, 2004). This collaborative approach supports the identification and mitigation of risks in real time. Agile teams are more adaptive and responsive, enabling them to make informed decisions as risks manifest or evolve.

After evaluating agile project management, we can say that these methodologies offer innovation projects a dynamic and adaptable approach to risk management. By encouraging ongoing risk assessment, rapid response, and close collaboration among team members and stakeholders, Agile project management enhances an organization's ability to handle the dynamic and multifaceted risks associated with innovation projects.

The use of technology and specialized tools has revolutionized risk management in innovation projects. These innovations provide organizations with the means to enhance risk assessment, analysis, and monitoring. Now, we will explore the role of technology and tools in risk management and the impact on innovation projects.

*Risk Management Information Systems (RMIS).* One of the notable advancements in risk management is the utilization of Risk Management Information Systems (RMIS). RMIS is software designed to support the risk management process by facilitating the identification, assessment, and monitoring of risks (Hillson & Murray-Webster, 2017). These systems often integrate databases, analytical tools, and reporting capabilities, enabling organizations to automate risk assessment processes, improve efficiency and

accuracy. RMIS allows organizations to maintain a centralized repository of risk information, making it easier to track and manage identified risks throughout the project lifecycle. RMIS can generate comprehensive reports, helping project teams make data-driven decisions and allocate resources for risk mitigation (Hillson & Murray-Webster, 2017).

*Data Analytics and Predictive Modeling.* In addition to RMIS, organizations are increasingly leveraging data analytics and predictive modeling to enhance risk management practices. These techniques harness the power of data to provide valuable insights into potential risks within innovation projects. Data analytics involves the systematic analysis of large datasets to identify patterns, trends, and potential risk factors (Mathews & Russel, 2020). It can uncover hidden correlations and dependencies that may not be apparent through traditional risk assessment methods. For innovation projects, this means a more comprehensive understanding of the various facets of risk, ranging from technical challenges to market dynamics. Predictive modeling takes data analytics a step further, enabling organizations to anticipate risks before they materialize. Machine learning algorithms are employed to build models that can forecast potential risks and their impacts (Chen, Brem, & Viardot, 2019). By analyzing historical project data, these models can detect patterns and anomalies, providing project teams with early warning signals and actionable insights for proactive risk management. Moreover, predictive modeling can simulate different risk scenarios to assess their potential consequences. By considering various factors, such as resource availability, cost, and impact, organizations can develop and optimize risk response plans that are finely tuned to the specific characteristics of the project (Chen, Brem, & Viardot, 2019, 2019).

*Artificial Intelligence (AI).* AI technologies are also making inroads in risk management. AI algorithms can be used to identify emerging risks and provide recommendations for risk mitigation (Mathews & Russel, 2020). By continuously analyzing data and monitoring project parameters, AI systems can alert project teams to potential risks in real-time, allowing for quick response and resolution. AI can automate the process of identifying and assessing risks. Machine learning models can analyze historical project data to detect patterns, anomalies, and potential risk factors (Chen, Brem,

& Viardot, 2019). By continuously processing real time data from various sources, AI systems can identify deviations from expected project performance, triggering proactive risk management actions. Moreover, the natural language processing (NLP) capabilities of AI can be harnessed for sentiment analysis and text mining. This can help in extracting valuable insights from unstructured data sources such as project reports, customer feedback, or even social media to identify potential emerging risks (Chandrasekaran & Vinodhini, 2012). The integration of AI into risk management tools can also aid in optimizing risk response plans. AI algorithms can assess different response scenarios, considering various factors such as resource availability, cost, and impact, to recommend the most effective risk response strategies (Chen, Brem, & Viardot, 2019).

Knowledge management is a foundational element for enhancing risk management practices in innovation projects. The systematic collection, organization, and dissemination of knowledge play a crucial role in improving the identification and mitigation of risks. This section delves into the significance of knowledge management and its multifaceted role in innovation project risk management, drawing insights from scholarly sources.

*Knowledge Capture and Retention.* The systematic capture and retention of knowledge in innovation projects constitute a fundamental practice within knowledge management that profoundly impacts risk management. This process involves creating repositories and databases that serve as repositories for various project documents, reports, and the valuable insights gained from past projects (Jugdev & Müller, 2005). These repositories are not merely archives but dynamic resources for future innovation projects. They encompass several key elements, as highlighted by Jugdev and Müller (2005), one of the critical components of knowledge capture is the documentation of lessons learned from past innovation projects. This involves a detailed account of the challenges encountered, the risks faced, and the strategies employed to mitigate them. The collection of lessons learned provides a comprehensive narrative that helps project teams understand the nuances of risk management through real-world examples and scenarios. Furthermore, risk registers, which categorize and detail various types of risks, their potential impact, and the responses used to address them, are vital components of knowledge capture (Kerzner,

2017). These registers serve as living documents that not only record historical risk data but also offer insights into the evolution of risks over time. Effective knowledge retention necessitates the organized and accessible storage of data, often achieved through digital databases, cloud storage, or knowledge management systems (Jugdev & Müller, 2005). The role of metadata and tagging in categorizing documents and insights cannot be underestimated. Properly tagged information allows for easy search and retrieval, ensuring that project teams can quickly access relevant information (Kerzner, 2017). Moreover, the knowledge must be readily accessible to project teams to inform their risk management strategies. Continuous updating of the knowledge repository is essential to keep the information relevant and aligned with evolving risk management practices and project dynamics. Practical benefits of knowledge capture and retention include avoiding the repetition of mistakes, proactive risk identification, and informed decision-making (Jugdev & Müller, 2005; Kerzner, 2017). Ultimately, these practices contribute to the enhanced efficiency and success of innovation projects.

*Expertise Networks.* Building and nurturing expertise networks in the context of innovation projects is a pivotal aspect of knowledge management and risk management enhancement. As emphasized by Bai, Wei, Zhang, and Zheng (2022), these networks serve as interconnected communities of individuals with specialized knowledge and experiences in specific areas, particularly related to risk management. They facilitate a dynamic platform for knowledge sharing and collaboration, creating an ecosystem where expertise is readily accessible to project teams. These networks are invaluable because they bridge the gap between theoretical risk management concepts and real-world application, ensuring that risks are not only identified but also comprehensively assessed and effectively mitigated. In the words of Bai et al. (2022), the collaborative aspect of expertise networks is particularly impactful. These networks foster open communication and interactive exchanges between experts and team members. This collaborative approach enhances the depth and accuracy of risk assessments, as it allows multiple perspectives to be considered. The input of experts, who have faced similar challenges in previous projects, provides real-world context to risk management decisions. Moreover, these networks encourage peer reviews and expert evaluations, adding an additional layer

of scrutiny to risk scenarios. By facilitating the sharing of insights and experiences, expertise networks ensure that risk management becomes a proactive, collaborative, and highly informed process, ultimately contributing to the success of innovation projects.

*Innovation Project Learning Culture.* Fostering an innovation project learning culture within an organization is a fundamental pillar of knowledge management and risk management enhancement. This cultural approach prioritizes open communication, collaboration, and knowledge sharing among project teams, with a specific emphasis on the experiences and insights related to risk management. As Jugdev and Müller (2005) emphasize, a learning culture encourages teams to learn from both the successes and failures of past innovation projects. It serves as the breeding ground for the documentation and dissemination of valuable risk management experiences, fostering a proactive approach to risk identification and response. Kerzner (2017) underscores the role of a learning culture in ensuring that knowledge is not only captured but also actively applied in subsequent projects. Teams are encouraged to document and share their experiences with risk identification, assessment, and response, which leads to the development of best practices. Through this iterative learning process, organizations refine their risk management strategies and continuously adapt them to evolving project dynamics. A learning culture also contributes to a sense of collective responsibility, where teams take ownership of risk management and collectively work toward project success. This cultural shift results in more informed decision-making, better risk assessment, and the proactive application of risk response strategies. Ultimately, it aligns the entire organization with the objective of continuous improvement in innovation projects, ensuring that knowledge is not a passive asset but an active driver of project success.

To sum it up, we can say that knowledge management is central to improving risk management in innovation projects. By systematically capturing and retaining knowledge, building expertise networks, and fostering a learning culture, organizations can enhance their risk management capabilities. The insights gained from previous projects and the expertise of team members and experts create a robust foundation for proactive risk assessment and response, ultimately contributing to the success of innovation projects.

### ***1.3 Case Studies.***

Real-world case studies provide valuable insights into how organizations have successfully improved risk management within innovation projects. By examining specific instances, we can glean practical lessons and strategies. Two illustrative case studies highlight the successful implementation of knowledge management and advanced risk management practices:

#### *1) NASA's Risk Management during the Apollo Program.*

The Apollo program undertaken by NASA stands as a remarkable exemplar of effective risk management within innovation projects. This groundbreaking endeavor, which successfully placed humans on the Moon, was characterized by its rigorous approach to risk assessment and mitigation (Dick & Launis, 2007). The Apollo missions confronted an array of intricate challenges, including the daunting task of navigating through the hazards of space, ensuring the safety of astronauts, and executing precision lunar landings. In this context, knowledge management emerged as a pivotal element in risk management within the program.

NASA's approach to risk management during the Apollo program was underpinned by meticulous knowledge capture and retention practices. The organization recognized the profound value of lessons learned and leveraged them for future missions. As noted by Dick & Launis (2007), the knowledge gleaned from one mission, such as Apollo 11, was diligently captured and retained to inform subsequent missions. Lessons learned encompassed an exhaustive account of challenges faced, risks encountered, and the strategies employed for risk mitigation. Additionally, risk registers were a crucial component of this approach. These registers systematically categorized various types of risks, detailed their potential impact, and recorded the strategies and countermeasures used to address them.

The significance of knowledge capture and retention is highlighted by the fact that the risk landscape for each Apollo mission was unique. As a result, the lessons and insights from earlier missions, such as Apollo 8 or 10, played a crucial role in preparing for the unforeseen challenges of subsequent missions. NASA's ability to proactively identify, assess, and respond to risks was heavily reliant on the comprehensive

documentation of past experiences. This approach was instrumental in ensuring mission success, ultimately leading to the historic Apollo 11 lunar landing.

Moreover, the meticulous record-keeping of lessons learned and the creation of a dynamic knowledge repository served as a powerful foundation for informed decision-making in the face of uncertainties. It allowed project teams to anticipate and address potential risks, which is particularly critical in innovation projects where unforeseen challenges are inherent. In essence, NASA's approach to knowledge management and risk management during the Apollo program was an essential factor in achieving the audacious goal of lunar exploration and stands as a testament to the value of capturing and retaining knowledge for innovative and risky endeavors.

## 2) *Toyota's Lean Product Development.*

Toyota's Lean Product Development (LPD) is a compelling case study that showcases the profound impact of a learning culture and expertise networks on risk management in innovation projects. LPD is deeply rooted in Toyota's organizational culture, emphasizing continuous learning, collaboration, and knowledge sharing (Morgan & Liker, 2006). The LPD system actively fosters a culture where teams are not only encouraged to learn from past experiences but are expected to do so. This culture of learning ensures that insights related to risk management are captured, retained, and proactively applied to subsequent projects.

Knowledge capture and retention are integral to the LPD process. In a culture where learning from both successes and failures is celebrated, lessons learned from each project are diligently documented. As noted by Morgan and Liker (2006), these lessons are not merely archived but are actively shared and discussed across teams. They serve as a source of valuable insights for future projects, especially in terms of risk management. This approach ensures that past risks, whether they resulted in success or setbacks, contribute to a culture of continuous improvement.

Expertise networks within Toyota's LPD system create a dynamic platform for collaborative risk management. Engineers, managers, and workers across various teams are connected in a web of expertise sharing. This collaborative network allows for the rapid dissemination of knowledge and experiences, particularly those related to risk

management (Morgan & Liker, 2006). The involvement of experts and team members from diverse backgrounds in risk assessment and response planning adds depth and robustness to the process. By engaging in discussions and leveraging collective expertise, LPD teams collectively identify, assess, and mitigate risks, enhancing the overall quality and reliability of product development.

Additionally, knowledge retention within LPD ensures that insights are not lost but are perpetually applied and improved upon. The iterative nature of Toyota's approach to risk management means that lessons learned from one project are woven into the fabric of subsequent projects. As risks and challenges evolve, LPD teams are better equipped to anticipate and address them. The LPD system, with its learning culture and expertise networks, underscores that knowledge management is a dynamic, ongoing process that contributes to informed decision-making, proactive risk management, and the achievement of innovation project goals. In this way, it aligns the entire organization with the objective of continuous improvement and ensures that knowledge is not merely a passive asset but an active driver of project success.

### *3) SpaceX's Falcon 9 Reusable Rocket Program.*

SpaceX, under the visionary leadership of Elon Musk, provides an exemplary case study of effective risk management within the realm of innovation projects. One of the most remarkable endeavors in SpaceX's portfolio is the Falcon 9 Reusable Rocket Program. The mission to develop reusable rocket technology represented a radical departure from traditional aerospace practices, presenting substantial technical and financial risks. However, through a strategic blend of knowledge management practices and a relentless focus on risk mitigation, SpaceX achieved groundbreaking success.

SpaceX's approach to risk management in the Falcon 9 program is marked by its unwavering commitment to knowledge capture and retention. Lessons learned from every Falcon 9 launch, whether successful or encountering setbacks, are meticulously documented (Foust, 2018). These lessons contribute to a dynamic knowledge base that informs subsequent launches. As a result, the organization can proactively address potential risks based on the data and insights gleaned from previous missions. This knowledge-driven approach enables SpaceX to continuously improve its risk management

strategies, emphasizing safety and cost-effectiveness.

The formation of expertise networks is another noteworthy element of SpaceX's risk management strategy. The organization comprises a diverse team of engineers, scientists, and experts who collaborate closely on each Falcon 9 launch (Foust, 2018). This collaboration not only ensures the technical integrity of each mission but also enhances risk assessment and response planning. By engaging in interactive discussions and sharing their collective expertise, SpaceX's teams collectively identify, assess, and mitigate risks, contributing to the program's unprecedented success in landing and reusing rocket components.

The commitment to learning is a core aspect of SpaceX's culture, echoing Toyota's Lean Product Development system. The organization actively encourages teams to learn from past experiences and continuously refine their risk management practices (Foust, 2018). SpaceX's learning culture reinforces the importance of knowledge capture and retention, particularly within the context of innovation projects like the Falcon 9 program. By consistently applying lessons learned, engaging in expert collaboration, and fostering a culture of learning, SpaceX achieved its vision of revolutionizing space access with reusable rockets, exemplifying the power of knowledge management and proactive risk management in the pursuit of innovation.

#### *4) Samsung's Rise in the Electronics Industry*

Samsung, a South Korean conglomerate, has become a global leader in the electronics industry. This case study examines Samsung's remarkable journey and the strategic projects that have contributed to its success. In the 1960s, South Korea's economy was struggling, and Samsung initiated a project to manufacture and export electronics, particularly televisions. Samsung's founder, Lee Byung-Chul, invested heavily in research and development (R&D) and created the foundation for Samsung Electronics (Heshmati, Kim & Kim, 2012). The project aimed to produce high-quality products at competitive prices, gaining an edge in the global market.

One of the critical projects that catalyzed Samsung's growth was the "New Management" initiative launched in the 1990s. Under the leadership of Lee Kun-hee, the project sought to transform Samsung into a premium brand through innovation and

quality. Samsung invested heavily in R&D, driving the development of cutting-edge technologies, including semiconductors, flat-panel displays, and mobile devices (Song & Lee, 2016).

Samsung's "Galaxy Project" was another strategic endeavor. It focused on the development of smartphones and aimed to challenge Apple's dominance. This project led to the introduction of the Samsung Galaxy series, which quickly became a global success (PROECHO SOLUTIONS, 2021). The project involved aggressive marketing, extensive R&D, and a commitment to continuous improvement, making Samsung a leading player in the mobile device industry.

The success of Samsung's projects can be attributed to a combination of factors, including substantial R&D investments, a focus on quality, and a willingness to take risks. Samsung's projects transformed it from a regional player into a global giant in the electronics industry, showcasing the power of effective project management.

These case studies showcase how knowledge management practices, including knowledge capture and retention, expertise networks, and learning culture, significantly enhance risk management in innovation projects. And to start the process of switching towards the topic, let's spend a bit of time and symbols to think of how AI could have dealt with those case studies instead or improved them.

Drawing from the insights gleaned from the preceding case studies and discussions on risk management for innovation projects, it's evident that there are various effective strategies and practices in place to mitigate uncertainties and enhance project success. However, one can't help but ponder how Artificial Intelligence (AI) could have been harnessed to augment these solutions and further bolster risk management efforts.

Firstly, AI presents a formidable opportunity to revolutionize the process of holistic risk assessment. While traditional methods such as Failure Mode and Effects Analysis (FMEA) and Probabilistic Risk Assessment (PRA) have proven effective, AI-powered predictive analytics and machine learning algorithms could offer a more dynamic and nuanced approach. For instance, AI algorithms could analyze vast datasets encompassing historical project data, industry trends, and external factors to identify potential risks with unprecedented accuracy and foresight. Monte Carlo simulations could be automated and

optimized through AI, allowing for more sophisticated risk modeling and scenario analysis.

Moreover, continuous monitoring and adaptation, vital components of effective risk management, could be significantly enhanced through AI-driven real-time analytics. Rather than relying solely on human oversight, AI algorithms could continuously monitor project parameters, detect anomalies, and predict potential risks before they escalate. Drawing inspiration from NASA's Apollo program, where real-time risk assessment played a pivotal role, AI-powered monitoring systems could provide project managers with actionable insights, enabling timely intervention and course correction.

Cultivating a risk-aware culture within organizations, as exemplified by Toyota's product development system, could also benefit from AI integration. AI-powered risk detection tools could empower employees at all levels to identify and report risks more effectively. Natural Language Processing (NLP) algorithms could sift through vast amounts of textual data, such as project reports and incident logs, to identify emerging risks and trends. Additionally, AI-driven collaboration platforms could facilitate cross-departmental communication and knowledge sharing, fostering a culture of proactive risk management.

Strategic project selection, a cornerstone of risk management, could be further optimized through AI-driven predictive analytics. Building on Samsung's success story, where strategic project selection propelled the company to industry leadership, AI algorithms could analyze market data, consumer trends, and competitive landscapes to identify projects with the highest growth potential and lowest inherent risks. By leveraging AI-powered decision support systems, organizations can make more informed and data-driven project selection decisions, aligning investments with strategic objectives and maximizing returns.

Furthermore, technology and data-driven tools, such as AI and machine learning, hold immense promise for augmenting risk management efforts in innovation projects. The Hyundai Motor Group case study underscores the potential of AI-driven tools for early risk identification and monitoring. AI algorithms could sift through vast amounts of structured and unstructured data, such as sensor readings, market reports, and social media

sentiment analysis, to detect early warning signs of potential risks. By harnessing AI-powered predictive analytics, organizations can proactively anticipate and mitigate risks before they escalate, thereby enhancing project resilience and success rates.

To sum up it up, while the traditional strategies and practices discussed in the case studies have proven effective in managing risks for innovation projects, the integration of Artificial Intelligence offers a paradigm shift in risk management capabilities. By harnessing the power of AI for holistic risk assessment, continuous monitoring and adaptation, cultivating a risk-aware culture, strategic project selection, and leveraging technology-driven tools, organizations can elevate their risk management efforts to unprecedented levels of effectiveness and agility. In an era defined by rapid technological innovation and disruptive change, embracing AI-enabled risk management is not just a competitive advantage; it's a strategic imperative for organizations seeking to thrive in the dynamic business landscape of the 21st century. But let's not stop here and move on to the next chapter which should be more interesting than a wall of theory.

## **CHAPTER 2. INTEGRATION OF ARTIFICIAL INTELLIGENCE IN RISK MANAGEMENT FOR INNOVATION PROJECTS IN PHARMACEUTICAL FIELD.**

Innovation projects stand as the lifeblood of organizational growth and competitiveness, serving as crucibles for transformative ideas and groundbreaking initiatives. Yet, they are also intrinsically fraught with uncertainties and risks, presenting formidable challenges to even the most visionary of enterprises. However, amidst this landscape of complexity and ambiguity, the advent of Artificial Intelligence (AI) emerges as a beacon of hope, a transformative opportunity to revolutionize risk management practices within these projects.

In this chapter, we embark on a journey to explore the integration of AI-driven solutions in risk management for innovation projects in pharmaceutical field, recognizing the profound implications of harnessing AI technologies to navigate the intricacies of uncertainty and volatility. As we delve into the realms of AI-powered analytics, predictive modeling, and automation, our aim is twofold: to elucidate how AI can enhance decision-making processes and mitigate project risks, and to unveil the transformative potential of AI in driving organizational resilience and competitive advantage. At the heart of this exploration lies a fundamental question: How can AI technologies be leveraged to augment traditional risk management approaches and empower organizations to thrive in the face of uncertainty? Through a comprehensive analysis of case studies, empirical research, and theoretical frameworks, we endeavor to provide insights into the strategic imperatives and practical considerations underlying the integration of AI in risk management for innovation projects.

Central to our inquiry is the exploration of the impact of integrating AI in risk management in pharmaceutical field, a paradigm shift that promises to redefine the contours of organizational resilience and strategic agility. By harnessing the analytical prowess of AI algorithms, organizations can gain unprecedented insights into potential risks, anticipate emerging threats, and proactively mitigate adverse outcomes. Moreover, AI-powered decision support systems enable real-time risk assessment, empowering project teams to adapt swiftly to evolving challenges and capitalize on emerging

opportunities. As we navigate the intricate interplay between AI technologies and risk management practices, we are confronted with a myriad of possibilities and challenges. From data governance and ethical considerations to organizational readiness and change management, the journey towards AI-enabled risk management is fraught with complexities and uncertainties. Yet, it is also imbued with immense promise of enhanced decision-making capabilities, optimized resource allocation, and sustained innovation in the pursuit of organizational excellence. As a matter of requirements and guidelines, this chapter will use gathered information during the pre-diploma internship that was conducted in “COR-Medical” company.

So let’s delve deeper into the integration of AI-driven solutions in risk management for innovation projects in pharmaceutical field, like divers delve into the depths of the ocean, to unravel the intricacies of AI adoption, and explore the transformative potential of AI in shaping the future of organizational resilience and strategic foresight. Through interdisciplinary collaboration and strategic inquiry, we seek to illuminate the path toward a future where AI serves as a catalyst for innovation, driving sustainable growth and competitive advantage in the dynamic landscape of organizational management.

### ***2.1 Alignment of AI Impact on Risk Management with Risk Management Strategies in Innovation Projects in Pharmaceutical Field.***

The intersection of risk management strategies in innovation projects in pharmaceutical field with AI adoption and management practices represents a pivotal part for organizational advancement. By harnessing AI technologies, organizations can augment traditional risk management approaches with advanced analytics, predictive modeling, and automation capabilities (Lee & Kim, 2021). AI enables a proactive and data-driven approach to risk identification, assessment, and mitigation, thereby enhancing project outcomes and fortifying organizational resilience in the face of uncertainty (Lee & Kim, 2021).

*Augmenting Traditional Approaches.* Traditionally, risk management in innovation projects relied on subjective assessments and historical data analysis, often leading to suboptimal outcomes and missed opportunities (Cecere & Corrocher, 2021). However, with the advent of AI, organizations can augment these traditional approaches by

leveraging advanced analytics techniques such as machine learning and natural language processing (NLP) (Schwartz & Teerling, 2021). By analyzing vast and disparate datasets, AI algorithms can uncover hidden patterns, correlations, and trends, enabling more accurate risk assessment and informed decision-making (Schwartz & Teerling, 2021).

*Proactive and Data-Driven Approach.* One of the key advantages of integrating AI into risk management is its ability to enable a proactive and data-driven approach to risk identification and mitigation (Cecere & Corrocher, 2021). Unlike traditional methods that rely on historical data and intuition, AI-powered risk management systems can continuously monitor project parameters, detect anomalies, and predict potential risks before they escalate (Lee & Kim, 2021). By leveraging real-time data analytics and predictive modeling, organizations can anticipate emerging threats and implement timely interventions to mitigate their impact (Lee & Kim, 2021).

*Enhancing Project Outcomes.* The integration of AI in risk management for innovation projects holds immense promise for enhancing project outcomes and driving organizational success (Schwartz & Teerling, 2021). By optimizing risk identification, assessment, and mitigation processes, AI enables project teams to allocate resources more efficiently, minimize project delays, and enhance overall project performance (Cecere & Corrocher, 2021). Moreover, by enabling organizations to make data-driven decisions, AI empowers project managers to navigate uncertainties with confidence and achieve project objectives more effectively (Schwartz & Teerling, 2021).

*Fortifying Organizational Resilience.* In the face of increasing uncertainty and volatility, organizational resilience has become a critical determinant of long-term success (Lee & Kim, 2021). By fortifying risk management practices with AI-driven capabilities, organizations can build a resilient foundation that enables them to adapt and thrive in rapidly changing environments (Cecere & Corrocher, 2021). From early risk identification to proactive mitigation strategies, AI empowers organizations to anticipate and respond to emerging threats with agility and foresight (Lee & Kim, 2021).

To sum this chapter up, the alignment of AI impact on management with risk management strategies in innovation projects represents a paradigm shift in organizational practices. By harnessing AI technologies to augment traditional approaches, organizations

can enable a proactive and data-driven approach to risk management, enhance project outcomes, and fortify organizational resilience in the face of uncertainty. Through interdisciplinary collaboration and strategic implementation, organizations can unlock the transformative potential of AI in reshaping the future of risk management and organizational management as a whole.

## ***2.2 “COR-Medical” Company Analysis.***

Here we will provide an analysis on the “COR-Medical” company, then conduct a survey, and complete this section with a research of how the concept of AI is inscribed in the heads of employees that work in management field.

First, let’s start with the analysis of the “COR-Medical” company. COR-Medical stands as a prominent trademark under which leading distributors of medical goods operate in Ukraine, including companies such as "Iridium" and "Optimatrading." Each of these entities specializes in the supply of specific types of modern medical equipment, instruments, implants, suture materials, and other medical products sourced from global manufacturing giants. With a legacy spanning 16 years, COR-Medical's distributorship firms have maintained a reputation as reliable and progressive partners. They have achieved this distinction by consistently offering verified and dependable innovative products within the realm of healthcare. The commitment to providing proven solutions in the health sector has solidified COR-Medical's status as a trusted player in the distribution of medical goods. Covering the entire territory of Ukraine, COR-Medical ensures the distribution of products that not only save lives but also contribute to the extension and enhancement of overall well-being. This internship report delves into a comprehensive analysis of COR-Medical's operations, encompassing areas such as economic planning, human resources management, marketing and logistics, IT management, accounting and finance, strategic management, and innovative activities.

*Characteristic of the Status of the Enterprise.* COR-Medical, registered as a Limited Liability Company (LLC) on February 15, 2022, boasts its legal presence in Ukraine with its registered address at 7 Anton Tsyedika Street, Kyiv. The organizational status as of November, 2023, indicates that the company is not undergoing any cessation process, reflecting its stable operational standing.

*Characteristic of Economic Activity.* Operating under the organizational-legal form of a Limited Liability Company, COR-Medical engages in various activities. The primary focus lies in wholesale trade of pharmaceutical products (code 46.46), while also extending its reach into activities such as repairing electronic and optical equipment, installation and assembly of machinery, wholesale trade of other machinery and equipment, non-specialized wholesale trade, retail trade of pharmaceutical products in specialized stores, and retail trade of medical and orthopedic goods in specialized stores, among others.

*Nature of State Regulation and Control over the Activity of the Enterprise.* As a legal entity, COR-Medical operates within the regulatory framework defined by the Ukrainian government. The enterprise adheres to the legal requirements and standards applicable to its industry, ensuring compliance with relevant laws and regulations. State supervision and control mechanisms play a role in overseeing and enforcing compliance within the healthcare distribution sector

*Nature of Owners' Management of the Enterprise.* The company is led by Mr. Tigran Vilyamsovich Amalyan, who serves as the director, steering the organization towards its strategic goals. The ownership structure is reflected in the authorized capital of 750,000.00 UAH, signifying the financial commitment and stake of the owners in the enterprise.

*Nature of the Management Organization of the Enterprise.* COR-Medical operates within a Limited Liability Company (LLC) framework, employing a management structure designed to facilitate effective decision-making, coordination, and communication. The management organization aims to align with the strategic objectives of the company, ensuring a cohesive approach to daily operations and long-term planning.

*Characteristic of Instructional, Normative, and Methodical Materials.* The company utilizes a range of instructional, normative, and methodical materials to guide its operations. The quality and completeness of these materials are essential for ensuring that employees have access to clear guidelines and procedures. Regular updates and reviews of these materials contribute to the efficiency and effectiveness of the company's processes. The overall form of purpose structure of companies can be presented using an image from

one of the provided materials (Figure 2.2).

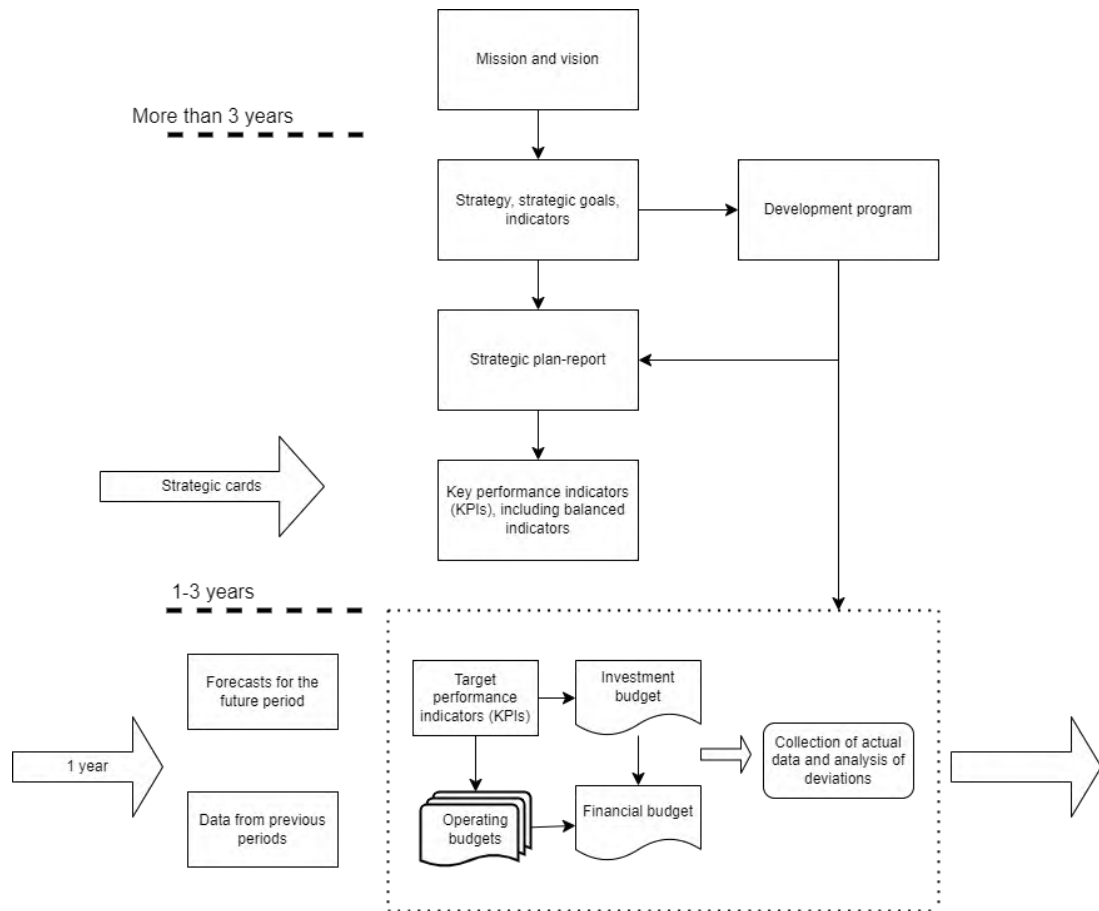


Fig. 2.2. Form of purpose structure of companies.

Source: Developed by author.

Next, we will delve deeper into specific spheres of COR-Medical's activities, including economic planning, human resources management, marketing, IT management, accounting and finance, strategic management, and innovative initiatives, providing a comprehensive overview of the enterprise's operations.

*Main Spheres of Financial and Economic Activity.* The company's main focus is on wholesale trade in pharmaceutical products (code: 46.46). Additionally, it engages in diverse activities such as the repair and maintenance of electronic and optical equipment, installation and assembly of machinery and equipment, wholesale trade in other machinery and equipment, non-specialized wholesale trade, and various forms of retail trade.

*Assortment Structure.* The company's assortment structure encompasses a range of activities, indicating diversification in its offerings. A detailed breakdown of the

assortment structure within each economic activity would provide a clearer understanding.

*Competitive Advantages.* Identifying competitive advantages requires an in-depth analysis of the company's strengths, weaknesses, opportunities, and threats (SWOT analysis). Factors such as market positioning, product quality, innovation, and customer relations can contribute to competitive advantages.

*Marketing Research.* A thorough marketing research analysis would involve evaluating the competitiveness of goods, market trends, customer preferences, and potential areas for improvement. It would also assess the effectiveness of current marketing strategies.

*Competitor Analysis.* There are 2 635 active companies that work in the same field of wholesale of pharmaceutical products. List of the biggest competitors located in Kyiv (Figure 2.2.1):

- "Optima-Pharm, Ltd". Business type: joint Ukrainian-Estonian Enterprise. Revenue: UAH 44,346,843,000.
- "Asino Ukraine" LLC. Business type: limited liability company. Revenue: UAH 3,644,424,000.
- "Medical Center "M. T. K." LLC. Business type: limited liability company. Revenue: UAH 3,101,524,000.
- "Arterium Corporation". Business type: corporation. Revenue: UAH 3,048,525,000.
- "Lekhim" PJSC. Business type: private joint-stock company. Revenue: UAH 2,898,741,000.
- "Teva Ukraine" LLC. Business type: limited liability company. Revenue: UAH 2,739,351,000.
- "HLED FARM LTD" LLC. Business type: limited liability company. Revenue: UAH 2,457,267,000.
- "SERVIE Ukraine" LLC. Business type: limited liability company. Revenue: UAH 1,722,054,000.
- "AMETRIN FK" LLC. Business type: limited liability company. Revenue: UAH 1,640,139,000.

- “SANOFI-AVENTIS Ukraine” LLC. Business type: limited liability company.  
Revenue: UAH 1,501,843,000.

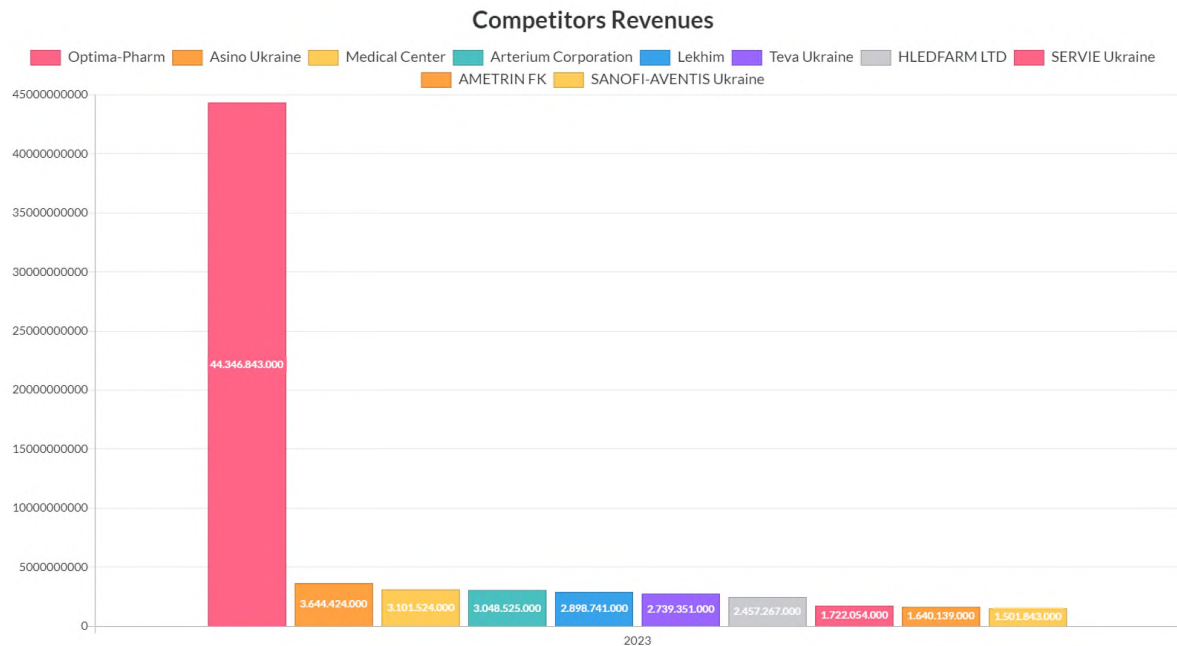


Figure 2.2.1. Competitors Revenues during year 2023.

Source: Developed by author.

Due to the fact that “COR-Medical” is a private company that is unwilling to share their financial statements, it was decided to use “Bayer” company and its public annual report (Table 2.2) to get a better understanding of the state of pharmaceutical companies in 2023. It is also worth to add the general model for pharmaceutical companies (Figure 2.2.2).

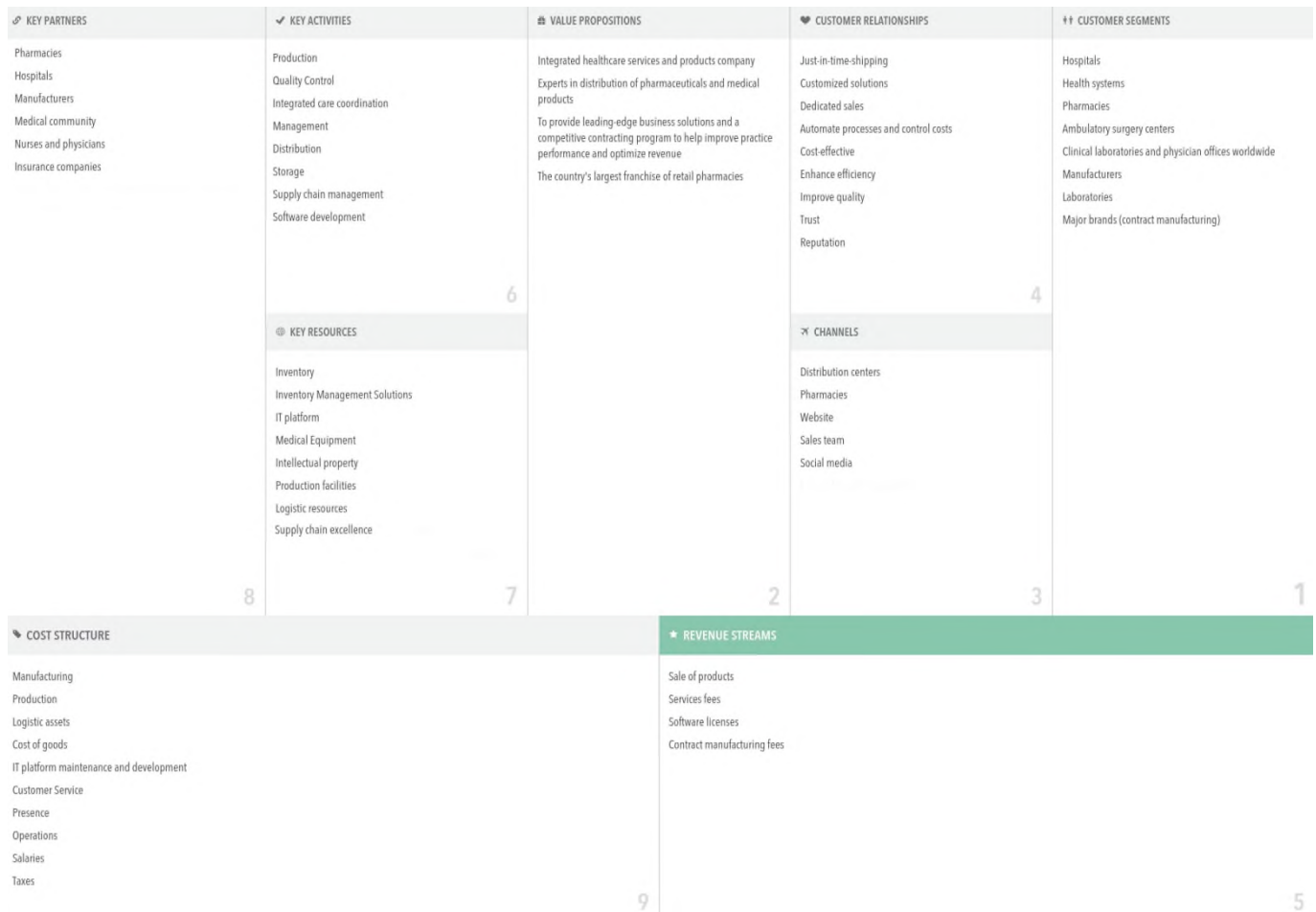


Figure 2.2.2. Business model for pharmaceutical companies.

Source: Developed by author.

Table 2.2. Fiver-Year Summary of “Bayer” company.

€ million	2019	2020	2021	2022	2023
<b>Bayer Group Financial KPIs</b>					
Sales	43,545	41,400	44,081	50,739	47,637
EBITDA	9,529	(2,910)	6,409	13,515	10,632
EBITDA before special items	11,474	11,461	11,179	13,513	11,706
EBITDA margin before special items	26.3%	27.7%	25.4%	26.6%	24.6%
EBIT	4,162	(16,169)	3,353	7,012	612
EBIT before special items	6,975	7,095	7,295	9,257	7,589
Net income (from continuing and discontinued operations)	4,091	(10,495)	1,000	4,150	(2,941)
Earnings per share (from continuing and discontinued operations) (€)	4.17	(10.68)	1.02	4.22	(2.99)

Core earnings per share (from continuing operations) (€)	6.38	6.39	6.51	7.94	6.39
Free cash flow	4,214	1,343	1,415	3,111	1,311
Net financial debt	34,068	30,045	33,137	31,809	34,498
Return on capital employed (ROCE) (%)	3.7	-16.5	3.8	7.7	0.7
Research and development expenses	5,301	7,126	5,412	6,572	5,371
Dividend per share (€)	2.80	2.00	2.00	2.40	0.11
<b>Bayer Group nonfinancial KPIs</b>					
Number of smallholder farmers in low- and middle-income countries supported by products, services and partnerships (million)	42	45	49	52	53
Number of women in low- and middle-income countries who have their need for modern contraception satisfied due to interventions supported by Bayer (million)	38	40	41	44	46
Number of people in underserved communities whose self-care is supported by interventions from Bayer (million)	41	43	46	49	51
Scope 1 and 2 greenhouse gas emissions (million metric tons)	3.76	3.58	3.17	3.03	3.00
Scope 3 greenhouse gas emissions from relevant categories (million metric tons)	8.82	7.93	7.97	8.98	8.44
Offsetting of remaining Scope 1 and 2 greenhouse gas emissions (million metric tons)	0.00	0.20	0.30	0.45	0.60
<b>Employees</b>					
Number of employees (Dec. 31)	103,824	99,538	99,637	101,369	99,723
Personnel expenses (including pension expenses) (€ million)	11,788	9,769	11,798	12,619	10,691

Source: Bayer Annual Report (2023)

*Main Functions of the Planning and Economic Department.* The Planning and Economic Department at COR MEDICAL plays a pivotal role in the strategic and financial aspects of the enterprise. It is responsible for executing and ensuring the final outcome of key sub-processes, specifically financial analytics. The department takes charge of organizing and conducting the annual strategic session, contributing significantly to the strategic planning process.

*Methodical Backing of Planning and Analytical Work.* The economic planning and analytical work at COR MEDICAL are supported by a structured process outlined in regulations. The department utilizes a comprehensive approach, considering various perspectives such as finance, market and clients, potential (employees and infrastructure), and internal business processes. The methodical backing includes the development of a detailed strategic plan that encompasses financial and non-financial indicators.

*Methods Used in Economic Indicators Analysis.* Analysis of economic indicators involves a multifaceted approach. Financial indicators are scrutinized through the "Finance" perspective, evaluating factors crucial for meeting the company's financial goals. Similarly, market and client-related indicators are assessed to understand the company's ability to satisfy customer needs. Internal business processes are optimized, addressing organizational structures and procedures to ensure optimal customer service.

*System of Analysis and Planning of Economic Activity Indicators.* The system in place revolves around an annual strategic session. During this session, a thorough analysis of the past year's achievements and deviations from strategic goals is conducted. The outcomes influence strategic adjustments, with considerations given to external and internal threats, opportunities, stakeholder requirements, and key success factors. The subsequent development of the strategic plan integrates financial and non-financial perspectives.

*Analysis of Annual Financial and Statistical Reporting Forms.* The results of economic activity are assessed through an elaborate process. The annual strategic session provides a platform for reviewing the strategic report of the past year, identifying reasons for deviations from planned indicators, and proposing necessary adjustments to the

strategy. The strategic plan for the upcoming year is developed based on decisions from general meetings and input from department heads. Positive outcomes are reflected in an agreed and approved strategic plan.

*Assessment of Volumes and Dynamics of Important Indicators.* The assessment includes a thorough examination of the "Finance" perspective, considering financial and economic indicators crucial for participant satisfaction. Indicators related to market and clients, potential, and internal business processes are also scrutinized. The ultimate positive result is the agreed and director-approved strategic plan that integrates strategic goals and planned deferred indicators.

As we can see, COR MEDICAL employs a robust planning and economic analysis system, incorporating financial analytics and strategic sessions. The detailed process ensures a comprehensive evaluation of economic indicators, fostering adaptability and informed decision-making for sustainable business development.

*Analysis of Staff List and Structure.* The staff list is dynamic and depends on the needs identified by heads of structural units. The analysis involves assessing the gender, age, education, and category distribution of employees. It ensures that the workforce composition aligns with the organizational requirements and adheres to the regulations outlined in RBP C4.02. From given resources, we can get the current staff list (Appendix A).

Moving forward, we should focus on IT management maintaining the Corporate Information System (CIS) in optimal working condition. External service providers play a crucial role in ensuring the seamless functioning of the information system. This includes addressing technical needs, analyzing technical specifications, and collaborating closely with company employees. The external service provider must align with these standards, ensuring the programming systems are in compliance. This requires a deep understanding of programming system features and their relevance to the enterprise's operational requirements.

*Research of Information System Tools and Quality and Reliability Assessment.* The IT management team, working closely with external service providers, conducts research on the tools available for optimizing the information system. This involves identifying

tools that enhance efficiency, security, and overall functionality. They also conduct an assessment of the quality and reliability of the information systems which is another critical aspect. This includes evaluating the performance of the external service provider, identifying and rectifying errors, and ensuring the smooth operation of network software and hardware.

*Valuation of Information Systems.* Another essential step in the process is the process is the valuation of one of the information systems in use. This involves a comprehensive examination of its features, capabilities, and alignment with the company's needs.

*Positive Outcomes.* Positive outcomes in IT management involve maintaining a well-functioning corporate information system, ensuring it meets the production needs of the company. The focus is on system reliability, efficiency, and adherence to security protocols. Since the company is quite new, not a single significant modernization hasn't been conducted yet.

The next step in our analysis is the management of accounting and financial transactions of the enterprise. The accounting system within the enterprise encompasses organizational, technical, and methodological spheres. This includes the quantification and qualification of specialists within the financial service, critical for the accurate management of financial transactions. A thorough evaluation of the financial service is imperative, considering both quantitative and qualitative aspects of the specialists involved. This evaluation extends to the quality of service output products, covering analytical, planning, and control forms. A comprehensive analysis of the financial and property status of the enterprise includes evaluating solvency, financial stability, business activity, and profitability indices. This offers insights into the overall health and sustainability of the enterprise.

*Execution of the "Communication" Function.* Directors of TOV "Iridium" and TOV "Optimatrading" are responsible for executing the communication function. This involves fostering a unified information space, preventing information gaps, engaging employees in indirect management processes, and promoting high standards of business and personal ethics. Communication within the companies is grounded in principles such as constant

monitoring, transparency, personalization, multivector approach, and innovation. Positive outcomes of this subprocess include achieving communication function goals.

*Execution of the “Control and Regulation” Function.* Directors are accountable for executing the control and regulation function. Tasks involve collecting and systematizing information on company performance, assessing results, analyzing deviations, identifying destabilizing factors, and making informed decisions based on forecasts. Financial analysts play a crucial role in providing documented information on business results. Directors evaluate the financial and operational aspects using this information, making operational decisions based on identified deviations from planned indicators. Regulatory measures, communicated through orders, directives, and electronic messages, are implemented to rectify identified deviations. Positive outcomes include achieving the goals of the control and regulation function.

*Management of Company Development.* Directors assess the results of the annual development program, with responsible employees providing electronic reports on project outcomes. Directors analyze project effectiveness, calculate the percentage of actual project completion, and devise corrective actions if goals are not met. Then they utilize information on the state of the annual development program for the preparation and submission of annual strategic quality management reports. Positive outcomes involve having objective information on the comprehensive execution of the development program and the implementation of corrective actions.

At last, we reach the last part of this analysis, which is the innovative activity management, which will give us yet another perspective on the state of company that will be useful later when evaluating and proposing recommendations on the implementation of AI solutions in risk management side of the company. The responsibility for implementing innovation policy lies with the directors of TOV "Iridium" and TOV "Optimatrading." This unit plays a pivotal role in steering the enterprise toward a culture of continuous innovation and growth. The degree of innovation activity is assessed through various lenses. Firstly, we have *the Recent Innovations*, which includes the introduction of new technologies, products, and organizational changes, a forward-looking analysis ensures the enterprise stays at the forefront of industry advancements. Secondly, there is an *the*

*Employee Involvement*, the process of generating innovative proposals by employees which is a crucial indicator of a dynamic and creative organizational culture. In compliance with the Law of Ukraine "On Innovation Activity," the structural elements of innovations and the final results of enterprise activity are meticulously analyzed. This involves scrutinizing the alignment of enterprise developments with the legal definition of "innovative products.". The innovative projects are subjected to a rigorous evaluation process to assess their effectiveness. This involves examining the outcomes, impact on operations, and alignment with strategic goals. Another critical aspect is *Feasibility and Investment Analysis*, which involves assessing the feasibility and potential for designing an innovative strategy for the enterprise. The analysis delves into the need and feasibility of financial investments, ensuring a balanced approach to resource allocation. When developing innovative proposals, there should be a solid understanding that the proposals should be grounded in the assessed needs, potential benefits, and strategic objectives of the enterprise.

*Management of Companies' Development, Focused Planning for Success.* Directors of TOV "Iridium" and TOV "Optimatrading" shoulder the responsibility for planning the development of the companies. The annual development program and project plans are key components of this process. Towards the end of the year, they craft the "Companies' Development Program" for the upcoming year in alignment with the Companies' Strategy. The Strategic Plan, developed under the A1.01 "Strategic Management" framework, serves as the primary source of information for these programs. The Companies' Strategic Plan prioritizes projects for the upcoming year. The Development Program comprises individual development projects, and employees responsible for specific projects are actively involved in its creation. The directors approve the Program, and responsible employees develop, coordinate, and submit plans for individual projects. The "Tree of Goals" principle ensures the systematic and goal-oriented development of the Companies. The annual "Strategy Map" is developed by financial analysts based on the approved Strategic Plan. This map ensures a balanced representation of strategic goals, cause-and-effect relationships, and the interconnectedness of various perspectives. The positive outcomes of this meticulous planning and development process include a well-structured

Development Program and project plans for the Companies. These outcomes serve as the foundation for the strategic growth and success of the enterprise.

*Control and Analysis, Monitoring Progress for Informed Decision-Making.* Directors take charge of monitoring the execution of planned activities and the status of the Companies' development projects. Responsible employees receive allocated resources and budgets for the implementation of planned activities. Employees responsible for development projects organize the execution, ensuring the achievement of project goals within specified timelines. Intermediate reports are prepared, highlighting key milestones and results achieved. Based on these reports, directors assess the status of individual development projects and the overall effectiveness of the entire Program. Informed decisions and directives are then provided, ensuring alignment with strategic objectives. If justified by credible reasons, adjustments are made to the plans of individual development projects and the overall Program. Changes are proposed, coordinated, and implemented promptly, reflecting a commitment to agility and adaptability. The positive outcomes of this control and analysis process include objective information about the intermediate state of program execution and the implementation of development projects. These results, along with planned corrective and preventive actions, contribute to the continuous improvement and success of the companies.

*Analyzing Year-End Results, a Strategic Review for Future Success.* Here the directors of TOV "Iridium" and TOV "Optimatrading" play yet another crucial role in assessing the results of the annual Development Program. The first step is Project Review and Reporting. Responsible employees evaluate the results of project implementation at the end of the year. By January 20th, comprehensive reports on development projects are submitted electronically. Next, directors analyze the summarized reports, scrutinizing the achievements of individual projects and the overall effectiveness of the Program. Selective verification ensures the accuracy and reliability of the information. In case the set goals of projects or the Program are not achieved, directors collaborate with relevant stakeholders to formulate corrective or preventive actions. These actions are designed to enhance future planning and execution. The information gathered from the analysis of year-end results is instrumental in preparing the annual strategic quality management report. This report

provides a comprehensive overview of the Companies' development and the successful realization of the Development Program. The positive outcomes of this thorough analysis include a well-informed strategic report, highlighting the achievements and areas for improvement in the Companies' development. This cyclical process of planning, execution, control, and analysis ensures the continual enhancement and sustainability of the Companies' strategic objectives.

### ***2.3. Survey, and research conducted using Ukrainian enterprises.***

Using another article as a source of information, to gain better understanding of the researched topic and gain a broader perspective of how the concept of AI is portrayed in the heads of Ukrainian companies. As it follows, the latest research from the international consulting agency PwC allows us to assess the state of the application and the speed of development of artificial intelligence. They estimate the contribution of such technologies to the global GDP at \$15.7 trillion and also project that this figure will increase by another 14% by 2030. This trend prompts company management to create necessary conditions for the use of artificial intelligence in their enterprises. The economic effect of implementing AI technologies primarily manifests in profit growth. This occurs through increased productivity in the production of goods and provision of services as a result of automating core business processes, as well as increased demand for goods and services due to the use of specialized analytical programs. Unfortunately, currently in our country, the managerial and social aspects of further developments in the field of artificial intelligence and the consequences of its practical use are not widely discussed among domestic company managers, sociologists, and the public. It is necessary to consider that the development of artificial intelligence requires a change in humanity's strategic thinking and further accumulation of knowledge in this field. This will help expand the possibilities of rational use of sixth technological level technologies. Managers who strive to increase the competitiveness of their enterprises must be confident in the feasibility and safety of implementing artificial intelligence.

Research conducted demonstrates that technological changes are inevitable in the long term. Innovations are constantly advancing, and the globalized world makes the

spread of new ideas much more efficient. At the same time, an important aspect is identifying the level of government, business, and ultimately, private individuals' readiness and ability to adapt to rapid technological changes. Analytical studies aimed at achieving this goal were carried out by experts from The Economist Group, who compiled the Technological Readiness ranking based on the analysis of the 82 largest world economies. This ranking was based on considerations of such groups of indicators as access to the Internet and mobile communication, economic infrastructure, and the level of innovation. Currently, Ukraine significantly lags behind European countries and ranked 38th, alongside Saudi Arabia and South Africa in 2018. At the current year of 2024, Ukraine is ranked 43th.

As we can see from the data provided, Ukraine is currently insufficiently adapted for the rapid implementation of innovative technologies. However, this does not exclude the possibility of ensuring rapid progress through the implementation of necessary scientific, organizational, investment, and other measures. At the moment, a significant number of developments in the field of artificial intelligence have been created and implemented, but experts believe that the level of development of such technologies is still insufficient.

In connection with the prospects of implementing achievements of the sixth technological order in various spheres of human activity, we conducted a survey of managers from a number of enterprises in Ukraine regarding the readiness to apply modern technologies, including the use of artificial intelligence in the field of production and service provision. Respondents included managers from middle and senior management levels who were asked to respond to questions from a specially designed questionnaire.

Let's analyze the results of the conducted research. Responses to the question "Have departments (divisions) responsible for technological innovations been created at your enterprise?" showed the following results (Figure 2.3). As we can see from (Figure 2.5.3), only 24% of respondents noted the presence of departments at their enterprises whose activities are focused on ensuring innovative technological development. 32% of respondents emphasized existing plans for their creation, and a significant portion (44%) indicated the absence of such departments in their responses. Another question of the

survey aimed to identify how deeply digital technologies have penetrated the management process of enterprises. Research shows that some managers currently do not use modern digital technologies in their professional activities, but their share among the total number of respondents is not very significant - 12%. Thus, we conclude that the majority of managers actively strive to work with digital tools, which has become a trend in their professional lives. The obtained results are illustrated (Figure 2.3.1).

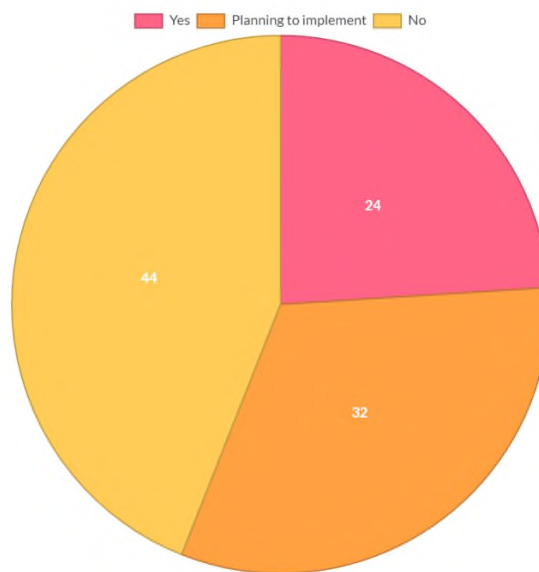


Figure 2.3 The results of the conducted survey on the existence at the enterprises of the units (divisions) responsible for technological innovations.

Source: Developed by author.

If enterprises are using digital technologies, it means that managers are guided by certain considerations in the process of their implementation. As responses to the corresponding question of the survey showed, the motives for use are diverse. Most often, these include adaptation to market demands and increasing the competitiveness of goods, which can be achieved through improved product (service) quality, reduced time costs for their design and manufacturing, and increased labor productivity.

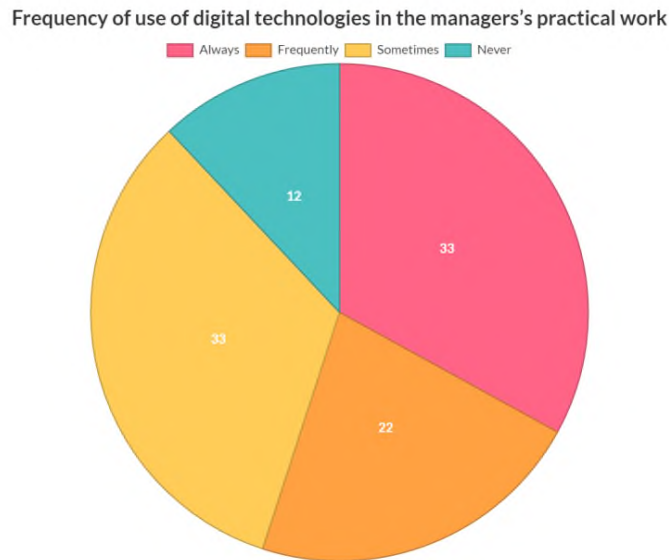


Figure 2.3.1. Frequency of use of digital technologies in the managers's practical work.

Source: Developed by author.

During technical innovations, almost every manager encounters challenges that hinder their implementation. In this case, it is important to identify the reasons for such a situation. The survey showed that most often, companies feel a lack of necessary financial resources for the implementation of advanced technologies due to their high cost, as indicated in the responses of 65% of respondents. Additionally, obstacles include insufficiently high levels of personnel training (12% of respondents) and other reasons stemming from the company's unpreparedness for changes. This once again proves that employee teams need to be gradually prepared for new changes in advance, even if their implementation is planned for the future. Using existing innovative practices, conducting qualified open professional dialogue with personnel, constructive cooperation with scientific institutions, and implementing special training programs are important steps toward facilitating the process of implementing artificial intelligence. Moreover, training and skill development at all stages of implementing scientific developments into life are crucial for ensuring comprehensive and effective practical application of technological innovations. It should be noted that this can be achieved through organizing training for personnel both by the company itself and by involving scientific institutions and specialized structures.

To prepare Ukrainian society for future implementation of artificial intelligence, it is

necessary to improve the existing education system. Currently, the concept of artificial intelligence is either superficially considered or not included at all in the curriculum of higher education institutions preparing future managers. Therefore, it is necessary to focus on training specialists who will work with AI systems in the future. Additionally, emphasis should be placed on fostering creativity and developing critical and systemic thinking in students. Effective training will be crucial for employees who will work in new types of activities and perform various tasks together with machines in the future.

During the survey, the aim was also to ascertain respondents' views on the place and prospects of applying artificial intelligence in enterprises. The survey showed the following results: 66% of respondents are confident that artificial intelligence will be used in conjunction with managers to balance their workload, while 34% believe it will free workers from routine tasks and work independently. None of the respondents agreed that implementing artificial intelligence is inappropriate as it poses a threat to humanity.

To determine the potential consequences of implementing artificial intelligence in the field of possible changes in the enterprise's workforce needs, the following question was included in the questionnaire: "Do you believe that the development of artificial intelligence will enable the creation of new jobs?" 56% of respondents responded positively to this question, while 44% believe it will lead to mass unemployment. Thus, according to the majority of respondents, the perspective of reducing the company's workforce due to the implementation of artificial intelligence is not likely; on the contrary, it will provide an opportunity to develop the company's activities and create new jobs.

This study also aimed to ascertain respondents' opinions regarding whether the development of artificial intelligence poses a certain threat to humanity. 59% of respondents answered that there is no such threat; 34% indicated that the threat is insignificant because humans control the use of artificial intelligence; 7% see a significant threat in artificial intelligence as it is capable of self-improvement and development, which could ultimately harm humanity. Thus, as the results of responses to the previous question showed, some managers perceive greater or lesser threats to humanity from artificial intelligence. So, do they believe that the development of artificial intelligence should be transparent and openly discussed with the public? All 100% of respondents agreed with

this statement. They noted that humans should control both the development and use of artificial intelligence, and the public should be aware of both its benefits and threats. Such a result is predictable, as proper protection of human interests should be ensured when applying technologies based on artificial intelligence. Therefore, regulation of artificial intelligence development is necessary both by the government and society, as confirmed by conducted research. Only developments that are safe for humans should be acceptable for implementation. Even at this stage of AI development, it is necessary to establish a legislative framework to control the process of AI development and implementation and hold accountable those who do not comply with necessary safety measures.

It is also worth mentioning, that the survey sought to understand managers' opinions on the priority areas for the use of artificial intelligence. The results showed that, according to surveyed enterprise managers, artificial intelligence should be an integral part of activities in some sectors and should find broad application there (financial services, industry, medicine, etc.), while in other sectors, artificial intelligence cannot replace humans, although it can somewhat streamline operational processes (for example, in the field of arts, education, and upbringing). Thus, the results of the survey demonstrate that artificial intelligence technologies can serve as extremely important levers for changing the way business is conducted and many other areas of human activity. They are intended to significantly streamline the process of decision-making and optimize management activities overall.

To conclude this particular analysis, we want to emphasize that the use and further development of artificial intelligence is a growing phenomenon that is irreversible. But account must be taken to the threats that arise as a result of the growth of the pace of improvement of artificial intelligence and its widespread implementation, really does exist. At the present stage of development, there is a risk that the latest technology, if used improperly, can lead to a systemic failure, which cannot always be corrected and eliminated by a person. In our opinion, developers of artificial intelligence must carefully analyze the risks that it brings, in order to avoid the possibility of global negative effects of its application. It follows that there is no need to give the technology excessive intellectual capabilities. Managing higher levels of activity should always rest with the man. In order

to prevent an unfavorable scenario, developers and employees of enterprises must study in detail all the nuances of the use of robotics and monitor the achievement of identified tasks. It is also worth noting those industries where artificial intelligence should be widely implemented, and where the use of robotics should be very limited. Studies have confirmed that our country, possessing considerable intellectual human potential and other resources, has real opportunities for the implementation of artificial intelligence. We believe that, provided that the proper and controlled use of artificial intelligence will certainly benefit humanity.

### **CHAPTER 3 RECOMMENDATIONS FOR IMPLEMENTATIONS OF AI IN RISK MANAGEMENT IN INNOVATION PROJECTS IN PHARMACEUTICAL FIELD.**

In this section, we will develop specific recommendations and proposals for managing the impact of Artificial Intelligence (AI) on risk management within innovation projects in pharmaceutical field. These recommendations will be based on the theoretical positions and methodological approaches outlined in the first section of the study, as well as the conclusions drawn from the analysis conducted in the second section.

Our primary focus will be on improving risk management practices within innovation projects through the strategic integration of AI technologies. We will explore ways to enhance risk identification, assessment, and mitigation processes to optimize project outcomes and increase organizational resilience.

To address the challenges and opportunities presented by AI integration in risk management in pharmaceutical field, we will propose several key strategies:

- Enhanced Data Governance.;
- Advanced Analytics Capabilities;
- Cross-functional Collaboration;
- Continuous Improvement;
- Ethical and Responsible AI Use.

Each recommendation will be accompanied by a detailed justification, outlining the rationale behind the proposed enhancement and its potential impact on risk management effectiveness. We will provide quantitative justification where possible, including cost-benefit analyses and return on investment calculations to demonstrate the value proposition of AI integration in risk management. Listed proposals will be evaluated based on their optimality, target efficiency, and practical implementation. We will assess each recommendation against predefined criteria to ensure that it aligns with the organization's strategic objectives and can be feasibly implemented within existing resource constraints. By following these recommendations and proposals, organizations can effectively manage the impact of AI on risk management in innovation projects, driving sustainable growth and competitive advantage in today's dynamic business landscape.

### ***3.1 Integration of AI-driven Risk Management Solutions and Interdisciplinary Perspectives on Their Integration.***

AI-powered risk management tools and techniques offer a potent arsenal for project managers to navigate the complexities of innovation projects effectively. From predictive analytics to natural language processing, AI enables organizations to glean actionable insights from vast and disparate datasets, empowering decision-makers to anticipate risks and implement timely interventions (Jha & Tiwari, 2023). Case studies and examples abound, illustrating how AI-driven risk management solutions optimize resource allocation, improve project efficiency, and mitigate adverse outcomes (Zhang et al., 2022).

Let's consider, for instance, the application of machine learning algorithms in predicting project risks based on historical data and contextual factors. By analyzing patterns and correlations within project data, AI algorithms can forecast potential risks and their likelihood of occurrence, enabling project managers to prioritize risk mitigation efforts and allocate resources judiciously (Kumar & Rastogi, 2021). Moreover, AI-powered decision support systems facilitate real-time risk assessment, allowing project teams to adapt swiftly to evolving challenges and capitalize on emerging opportunities (Lee et al., 2023). Now let's explore already existing and successful solutions that have been implemented and integrated into Risk Management.

*Predictive Analytics for Risk Prediction.* Incorporating machine learning algorithms, such as decision trees and random forests, enables organizations to predict project risks with high accuracy based on historical data and contextual factors (Jha & Tiwari, 2023). By analyzing patterns and correlations within project data, AI algorithms can forecast potential risks and their likelihood of occurrence, empowering project managers to proactively mitigate risks and optimize resource allocation (Kumar & Rastogi, 2021).

*Real-time Risk Assessment with AI Decision Support Systems.* AI-powered decision support systems enable real-time risk assessment by analyzing streaming data from various sources, such as sensors and project management tools (Lee et al., 2023). By leveraging advanced analytics and machine learning algorithms, these systems provide project teams with actionable insights, allowing them to adapt swiftly to evolving challenges and capitalize on emerging opportunities (Zhang et al., 2022).

*Natural Language Processing for Risk Identification.* Natural Language Processing (NLP) algorithms analyze unstructured textual data from project documents, reports, and communication channels to identify potential risks and emerging issues (Jha & Tiwari, 2023). By extracting key insights and sentiments from text data, AI-powered NLP tools enable project managers to proactively address risks and mitigate their impact on project outcomes (Kumar & Rastogi, 2021).

*Optimizing Resource Allocation and Efficiency.* AI-driven risk management solutions optimize resource allocation and improve project efficiency by identifying redundant tasks, bottlenecks, and inefficiencies in project workflows (Zhang et al., 2022). By automating repetitive tasks and streamlining processes, organizations can reduce project costs, shorten project timelines, and enhance overall project performance (Lee et al., 2023).

In conclusion, the integration of AI-driven risk management solutions represents a paradigm shift in project management practices, enabling organizations to proactively identify, assess, and mitigate risks in innovation projects. By leveraging predictive analytics, real-time risk assessment, natural language processing, and optimization algorithms, organizations can optimize resource allocation, improve project efficiency, and enhance project outcomes in today's dynamic business environment.

The integration of AI in risk management for innovation projects necessitates a multidisciplinary approach, drawing insights from both risk management and AI literature (Chen et al., 2023; Sharma & Kumar, 2022). By synthesizing principles from risk management theories, such as probabilistic risk assessment and scenario analysis, with AI-driven predictive analytics and optimization algorithms, organizations can develop robust risk management frameworks tailored to the dynamic nature of innovation projects (Majumdar & Kulkarni, 2021; Wang et al., 2023). Moreover, interdisciplinary collaboration fosters innovation and cross-pollination of ideas, driving continuous improvement in risk management practices (Li & Chai, 2022).

*Synthesizing Risk Management Theories with AI Techniques.* Integrating AI into risk management requires a synthesis of established risk management theories with cutting-edge AI techniques (Chen et al., 2023). For example, probabilistic risk assessment, a

fundamental concept in risk management, can be enhanced with AI-driven predictive analytics to forecast project risks with greater accuracy and precision (Sharma & Kumar, 2022). Similarly, scenario analysis, a strategic planning tool in risk management, can benefit from AI-powered optimization algorithms to identify optimal risk mitigation strategies under different scenarios (Majumdar & Kulkarni, 2021).

*Tailoring Risk Management Frameworks to Innovation Projects.* Innovation projects exhibit unique characteristics and uncertainties, necessitating the adaptation of risk management frameworks to suit their dynamic nature (Wang et al., 2023). By leveraging AI-driven predictive analytics and optimization algorithms, organizations can develop tailored risk management frameworks that account for the complexities and uncertainties inherent in innovation projects (Li & Chai, 2022). For instance, AI algorithms can analyze historical project data and industry trends to identify emerging risks and opportunities, enabling organizations to make informed decisions and adapt their strategies accordingly (Sharma & Kumar, 2022).

*Fostering Interdisciplinary Collaboration.* Interdisciplinary collaboration between risk management professionals, data scientists, and domain experts is essential for harnessing the full potential of AI in risk management (Chen et al., 2023). By bringing together diverse perspectives and expertise, organizations can drive innovation and cross-pollination of ideas, leading to continuous improvement in risk management practices (Majumdar & Kulkarni, 2021). For example, collaboration between risk management professionals and data scientists can facilitate the development of AI-driven risk assessment models that are tailored to the specific needs and challenges of innovation projects (Wang et al., 2023).

*Embracing Continuous Improvement.* The integration of AI in risk management is an ongoing journey that requires organizations to embrace continuous improvement and learning (Li & Chai, 2022). By leveraging feedback mechanisms and iterative processes, organizations can refine their AI-driven risk management frameworks and adapt them to evolving project dynamics (Sharma & Kumar, 2022). Moreover, organizations can foster a culture of innovation and experimentation, encouraging employees to explore new ideas and approaches to risk management (Chen et al., 2023).

In conclusion, the interdisciplinary integration of AI in risk management represents a transformative opportunity for organizations to enhance their ability to anticipate, assess, and mitigate risks in innovation projects. By synthesizing principles from risk management theories with AI-driven predictive analytics and optimization algorithms, organizations can develop tailored risk management frameworks that are adaptive, agile, and resilient in the face of uncertainty.

### ***3.2 Ways of Improvement of Risk Management in Innovation Projects in pharmaceutical field using AI Integration. Strategic Integration of AI Technologies. Already Existing and Commonly Used Technologies in Ukrainian Enterprises.***

Innovation projects are inherently characterized by uncertainty and complexity, making effective risk management essential for project success. The strategic integration of Artificial Intelligence (AI) technologies offers a transformative opportunity to enhance risk management practices within innovation projects. By leveraging AI-driven tools and techniques, organizations can augment traditional risk management approaches, improve decision-making processes, and mitigate project risks effectively (Nguyen et al., 2022; Sharma & Kumar, 2023).

One of the key areas where AI can significantly impact risk management in innovation projects is in enhancing risk identification processes. AI algorithms can analyze vast amounts of structured and unstructured data from various sources, including project documents, reports, and external databases, to identify potential risks and emerging threats (Lee & Kim, 2023; Wang et al., 2022). Natural Language Processing (NLP) techniques, for example, enable AI systems to extract insights from textual data, allowing project managers to identify hidden risks and anticipate potential issues proactively (Zhang et al., 2023).

AI technologies also play a crucial role in improving risk assessment processes within innovation projects. Machine learning algorithms can analyze historical project data and identify patterns and trends, enabling organizations to assess the likelihood and impact of potential risks more accurately (Chen et al., 2021; Majumdar & Kulkarni, 2022). Predictive analytics models, powered by AI, can forecast project risks and simulate various scenarios to evaluate their potential impact on project outcomes, allowing project teams to

prioritize risk mitigation efforts effectively (Jha & Tiwari, 2022). In addition to enhancing risk identification and assessment, AI technologies enable organizations to mitigate project risks more effectively. AI-driven decision support systems provide real-time insights and recommendations to project teams, allowing them to make informed decisions and take timely actions to address emerging risks (Kumar & Rastogi, 2022; Schwartz & Teerling, 2023). Moreover, AI-powered optimization algorithms can optimize resource allocation and project schedules, minimizing the impact of risks on project timelines and budgets (Lee et al., 2021).

By strategically integrating AI technologies into risk management processes, organizations can increase their resilience to uncertainties and disruptions in innovation projects. AI-powered risk management systems enable organizations to adapt swiftly to changing circumstances, identify new opportunities, and capitalize on emerging trends (Cecere & Corrocher, 2023; Li & Chai, 2021). This enhanced agility and flexibility allow organizations to navigate the complexities of innovation projects more effectively and maintain a competitive edge in the marketplace.

To further delve into this topic, let's see the already existing and functioning solutions that are integrated in Ukrainian enterprises. Over a considerable period, the study of artificial intelligence tools in managing the behavior of economic agents in the digital space within pharmaceutical enterprises has considered numerous aspects and nuances of personnel management using artificial intelligence. However, the field of study is quite broad and requires careful analysis of modern personnel management tools. Since the range of tasks to be solved is very wide, only some tools for managing economic agents in enterprises will be discussed. Behavioral economics examines manifestations of rationality and irrationality among economic agents and the effectiveness of their functioning as management objects. At the same time, researching management processes is an extremely complex task. Economic relations involve the exchange of goods between economic agents (objects and subjects of management) on a voluntary basis, meaning that each participant in the relationship receives certain benefits. In the model of studying the behavior of economic agents based on artificial intelligence tools, artificial intelligence itself is an auxiliary tool in building these relationships but does not solve all problems

related to the human factor and does not solve all enterprise problems. The concept of artificial intelligence is often associated with a universal problem-solving tool, but these tools are not without flaws.

It should be noted that the concept of artificial intelligence gradually changes its meaning from year to year, depending on how far-reaching the technologies are. Artificial intelligence is a branch of science aimed at creating robotic intelligence, a computer capable of thinking and solving tasks similar to how a human does (Matviichuk, 2010). The task of AI is to teach machines to draw conclusions based on acquired knowledge and experience so that robots can perform functions similar to those of humans. In other words, intelligence is the general ability to learn and solve problems, which combines all of human cognitive abilities: perception, memory, imagination, thinking, and reasoning. There are several definitions of artificial intelligence that complement and clarify each other to some extent.

Artificial intelligence in the context of managing the behavior of economic agents is a tool that includes a set of capabilities allowing an algorithm, based on accumulated knowledge about economic agents, to provide answers to questions, formulate expert conclusions, learn, think, solve problems, and acquire knowledge beyond what its developers have programmed it to do. Synthesizing similar definitions of the concept of artificial intelligence from various sources, we introduce the following definition: Artificial intelligence in the context of managing the behavior of economic agents is a tool that includes a set of capabilities allowing an algorithm, based on accumulated knowledge about economic agents, to provide answers to questions, formulate expert conclusions, learn, to think, solve problems, and acquire knowledge that developers didn't anticipate (Lohvinenko, 2022). Next, using the example of employee recruitment, let's consider some tools (based on artificial intelligence) that already allow us to build effective personnel management systems and minimize routine work for personnel specialists in companies today. Each employee goes through 5 life cycles within the company, and for each cycle, we'll analyze the effectiveness of using artificial intelligence and examine existing tools (Liashenko & Vyshnevskyi, 2018):

- 1) Recruitment and onboarding;

- 2) Work performance monitoring;
- 3) Automation of work processes;
- 4) Learning and personal growth as a specialist;
- 5) Termination.

Utilizing artificial intelligence tools at all stages of an employee's life cycle in a company provides the opportunity, from day one, to analyze the effectiveness of the economic agent as a specialist, objectively evaluate their work, formulate compensation based on their work, provide opportunities for personal growth, and assist in addressing current issues. Thus, artificial intelligence tools in this management system help both employees and employers build effective relationships (Figure 3.2). Hiring tools help accurately place a specialist in a job, monitoring tools precisely control the formation of fair pay, help resolve conflicts quickly, or facilitate effective decision-making coordination between company departments in conflict situations. Next, we'll provide a scheme of the relationship between the management object (employee) and the management subject (employer or manager) where artificial intelligence tools serve as a means of communication and control for the subject over the object, helping keep under control what humans can't. The management subject (also a manager) fulfills management duties over projects and personnel, but due to a large number of daily tasks, they may overlook important aspects of monitoring employee work quality (usually more than 8). The management object is an employee of the enterprise with certain tasks and responsibilities to the company, whose salary usually depends on the quality and quantity of work performed. Artificial intelligence tools are algorithms (programs, platforms) that allow effective cooperation between the object and subject of management, focusing on those important questions that are typically overlooked without these tools. As mentioned above, the quality of artificial intelligence tools depends on the amount of data (or knowledge) processed by the algorithm. Therefore, the "Artificial Intelligence Tools" block has input and output data from the management object, allowing real-time monitoring and adjustment (through the management subject) of their work. The mechanisms for managing the behavior of economic agents in the digital space within enterprises represent a system for managing the behavior of economic agents using artificial intelligence tools,

which conducts employee monitoring and adjusts actions (or provides recommendations) to the management subject to enhance the efficiency of managing departmental work.

Thus, we achieve the “Result of applying management influences and target behavior of economic agents,” which, through behavior management, brings us to the level of checking the effectiveness of the entire system through "Recommendations for adjusting the behavior of economic agents. Today, many companies develop and support personnel management tools based on artificial intelligence, many of which are inaccessible to ordinary users or small companies, so the analysis of existing tools is based on well-known software vendors and those available for widespread use (Lohvinenko, 2022).

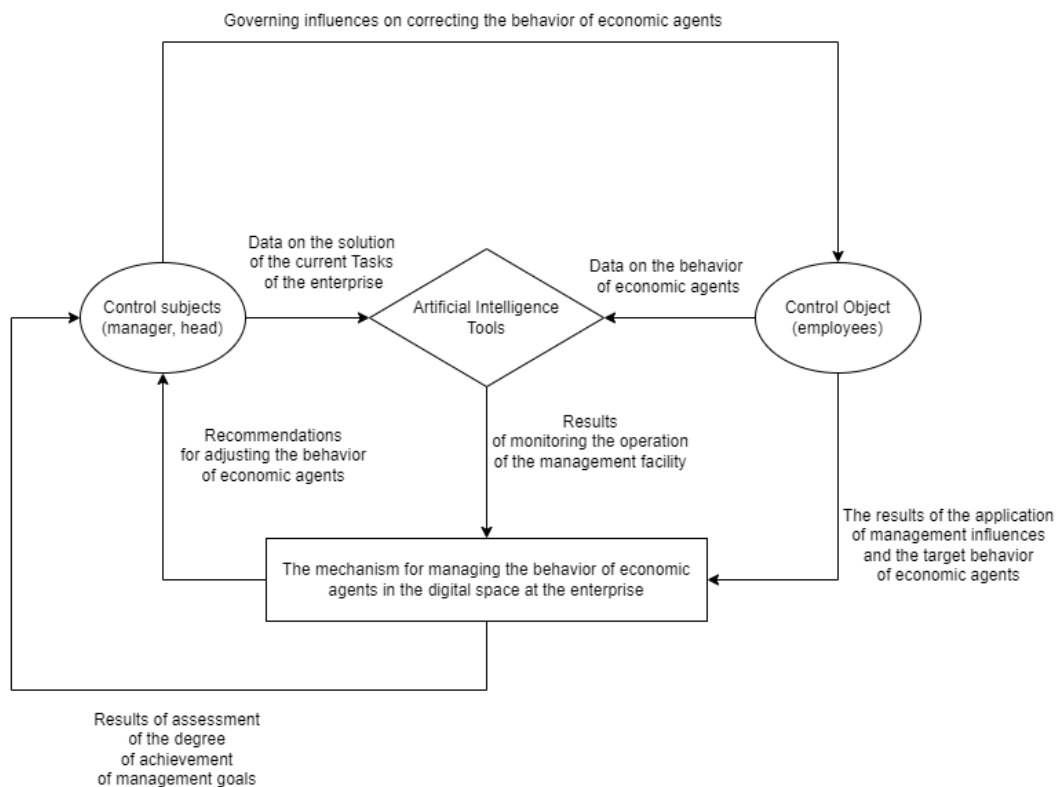


Figure 3.2 The scheme of building relationships between the object and the subject of management.

Source: Developed by author.

Table 3.2 examines and systematizes existing personnel management tools by directions: recruitment, monitoring, training, and coordination, and also considers the advantages and disadvantages of all tools. Let's consider the first category of tools that

handle personnel selection and automate bureaucratic issues related to employment. The main functionality of such tools includes: competency and skill analysis, automatic employee selection, creating employment plans, conducting interviews, recognizing text messaging and spoken language, collecting and processing statistical data.

The second category of HR tools allows for monitoring work performance and the termination procedure of employees. Its main functionality includes analyzing employee behavior, identifying leaders in the team, managing employee engagement in various work processes, measuring team morale, and analyzing potential candidates for dismissal from their positions.

The third category of tools allows for training and personal growth of specialists, namely onboarding procedures and knowledge assessment of personnel and clients, selection of training courses for career advancement and job preparation, systematization of existing courses in the company, and easy access to training from any device.

The fourth category entails the automation of work processes and decision-making coordination for enterprises, where artificial intelligence tools enable data processing automation and decision-making through machine learning and big data analytics, coming from various elements of IT infrastructure in real time. Neural network modeling methods allow for identifying reflexive characteristics of agents and their subjective inclinations in decision coordination to form appropriate management influences and enhance the speed and consistency of decision-making to achieve operational goals in enterprises.

Table 3.2. Existing and successfully implemented tools in Ukrainian enterprises.

1. Recruitment and Onboarding				
№	Name of Tool	Brief Description	Advantages	Disadvantages
1.1	Resume Matcher (SAP)	Using artificial intelligence technology, the program scans job vacancies for required skills and competencies.	Does not require complex integration into existing personnel management systems. Supports most	High cost. Inability to integrate into Ukrainian personnel management systems.

		System Type: Software and website.	internationally known management systems based on SAP SuccessFactors.	Lack of Ukrainian language interface.
1.2	Skillaz	Work Algorithm: Based on artificial intelligence, the tool provides intelligent workforce scheduling and planning for recruiters.  System Type: Multi-platform tool (application, website, software, chatbot)	Multi-platform functionality. Utilizes modern technologies: chatbots, manager notifications, and has its own application for clients and managers.	Inability to use individual HR tools from the entire personnel management system.
1.3	TalentTech Sever.AI	This AI-based system starts candidate selection by offering online tests to candidates, conducts interviews based on received data, and makes recommendations towards a specific candidate.  System Type: Multi-platform tool (application,	Works with cloud technologies and is not tied to the user's device. Allows transitioning from any personnel management system and database import.	Inability to implement in Ukrainian enterprises.

		website, software, chatbot).		
1.4	Hurma System	Artificial intelligence gathers information about company employees from employment to resignation System Type: Software and website.	Allows recruiters to minimize their workload. Supports remote hiring and subsequent communication with employees.	Lack of multi-platform support.
2. Work Performance Monitoring and Termination Procedure				
2.1	Veriato 360	AI collects and monitors a range of data on employee activity interacting with your IT resources and information. System Type: Software for Windows, Mac OS, and Android.	Enables secure organization of remote work. Ensures company security and protects corporate and personal employee data.	This software heavily burdens user devices, hindering employee productivity.
2.2	Workday	The software tracks employee work and evaluates work performance based on artificial intelligence. System Type: Software for Windows, Mac	Allows for organization and planning management, including organizational structure and other HR models, workforce	Difficult system setup in enterprises, database migration, etc. Lack of software support in Ukrainian language.

		OS, and Android.	management, and business process management.	
2.3	Yva.ai 3.0 by Visier	Based on digital interaction analysis of employees in corporate sources and micro-surveys, the system provides recommendations, information panels, and reports for managerial decisions. System Type: Multi-platform tool (application, website, software).	No equivalents in this field (leadership detection). Has the best tools for analyzing digital employee interaction in corporate sources. Offers a free trial of the system for 30 days.	Insufficient tools for comprehensive employee monitoring. Analysis based on surveys (whose effectiveness is subject to criticism).
2.4	Isaak Status Today by Glickon	A system that collects data on remote employee work, also analyzes psychological and morale states, and based on artificial intelligence, advises managers on the need to pay attention to a particular employee System Type: Multi-platform	Allows for effective analysis of employees' emotional states in a short period.	One-sided system (besides emotional state, the system monitors nothing else).

		tool (application, website, software).		
3. Learning and Personal Growth as a Specialist				
3.1	Cornerstone + EdCast	The ability to create a company's personal cabinet on the platform, which, using artificial intelligence algorithms, provides recommendations to users. System Type: Website and mobile application.	The company minimizes training costs for employees.	System configuration requires a lot of time. Inability to control artificial intelligence and its recommendations.
3.2	Degreed	From the proposed candidates for a specific vacancy, artificial intelligence selects the most suitable candidate for qualification improvement. System Type: Website and mobile application.	Company employees have the opportunity for growth within their company (to receive promotion, it is not necessary to change jobs).	One-sidedness of the system (besides the main function of selecting candidates for training, the platform does not address the educational needs of the company).
3.3	Filtered Content Intelligence	A platform for content analysis - uses artificial intelligence and algorithms to	No equivalents. Ease of use. Customization options for specific companies.	Difficulty in setting up searches for different disciplines from one profile (for

		analyze content libraries. System Type: Software.		example, programming and English language courses).
3.4	WalkMe ActionBot	A chatbot simulates communication with an ordinary person. Using artificial intelligence tools, it recognizes user questions and provides answers in minutes. System Type: Chatbot.	Ease of use and accessibility. Changes attitudes towards learning and personal growth.	Limited choice of disciplines.
4. Automation of Labor Processes and Decision Consensus				
4.1	AIOps	Working Algorithm: AI-based decision consensus system that, using machine learning technologies, recognizes the level of the issue and helps find a solution in minutes. The system operates based on machine learning and data analysis from	The platform manages accumulated data - hardware that allows recording, indexing, and storing semi-structured data that comes in large volumes and at high speed. Anomaly detection - software that uses various patterns.	Applicability limited to the IT sphere.

		monitoring systems, work logs, etc. System Type: Algorithm.		
4.2	Economic Mechanism of Decision Consensus in the Reflexive Management System at Enterprises	Working Algorithm: Employee surveying that allows identifying reflexive characteristics of agents and their subjective inclinations in the decision consensus process to form appropriate management influences and improve the speed and consensus of decision-making regarding achieving enterprise operation goals. System Type: Separate algorithm.	Ability to consider reflexive characteristics of agents such as competence, awareness, and intentional orientation of agents, and the conditions of pressure from the coordination center on agents of different management levels depending on the enterprise's structure	Calculating a large amount of data

Source: Developed by author.

All analyzed tools have advantages and disadvantages, the main ones being multi-platform availability and extensive functionality, and the drawbacks include tool cost and

complexity in configuring them in the conditions of Ukrainian pharmaceutical enterprises.

The above-listed tools are indeed multifunctional systems that effectively address personnel management issues. The main advantages lie in their multi-platform availability and extensive functionality. However, complex systems require complicated maintenance, which complicates end-user operations. The systems, in turn, have limited functionality and do not perform the full cycle of work. Therefore, this problem is fundamental, and to simplify the system as much as possible without compromising its functionality, such systems will allow for the use of modern artificial intelligence technologies in personnel management.

Research on existing artificial intelligence tools in managing the behavior of economic agents in the digital space has shown that for each period of an employee's life within the enterprise, there are many tools that allow for automating most of the routine management processes (from hiring employees to their dismissal). However, each of these tools has numerous drawbacks, which necessitates the development of an accessible system for managing the behavior of economic agents in the digital space using artificial intelligence tools.

For example, in the realities of Ukrainian enterprises, most foreign personnel management platforms are impossible to implement due to the difficulty of their implementation and the absence of the Ukrainian language. However, the main problem remains the obsolescence of management methods used in enterprises (especially state-owned ones). Most of them do not have databases, servers, or hardware that could monitor employee work. For instance, the software Resume Matcher (SAP) has very high system requirements, and TalentTech Sever.AI cannot be configured in the Ukrainian language, making it unusable in Ukraine.

Other companies that have been created in the last 10-15 years were initially built correctly (following the foreign model) and have all modern management tools at their disposal in their daily operations. The most modern of them have already updated and are testing artificial intelligence tools in their enterprises, for example: the Ukrainian company Conceptor uses Skillaz, Altis - Filtered Content Intelligence, and Capgemini Engineering - AIOps (Kasimov, 2023).

The reviewed artificial intelligence tools for personnel management prove that researchers and developers have done a lot of work to implement artificial intelligence tools in managing economic agents in enterprises today. However, in the conditions of Ukrainian enterprises, all these products require refinement and configuration. Therefore, a promising area of research will be the analysis and development of our own system for managing the behavior of economic agents in the digital space using artificial intelligence tools, which will allow combining all the needs of Ukrainian enterprises in personnel management into one system.

To sum it up, the strategic integration of AI technologies offers immense potential for improving risk management practices within innovation projects in pharmaceutical field. By enhancing risk identification, assessment, and mitigation processes, organizations can optimize project outcomes, increase organizational resilience, and drive sustainable growth in today's rapidly evolving business environment.

To effectively address the challenges and harness the opportunities presented by AI integration in risk management for innovation projects in pharmaceutical field, organizations can adopt several key strategies. These strategies encompass enhanced data governance, advanced analytic capabilities, cross-functional collaboration, continuous improvement, and ethical and responsible AI use.

*Enhanced Data Governance.* Implementing robust data governance frameworks is paramount for ensuring the quality, integrity, and security of project data in AI-driven risk management processes. By adhering to rigorous data governance practices, organizations can enhance the reliability and accuracy of AI-driven risk assessment models, thereby improving decision-making and mitigating project risks more effectively. One key aspect of enhanced data governance is the establishment of data standards that define the structure, format, and semantics of project data (Chen et al., 2021). These standards ensure consistency and interoperability across different data sources, facilitating seamless integration and analysis of project data within AI-driven risk management systems. Organizations must develop protocols for data collection and storage to ensure the integrity and security of project data throughout its lifecycle (Kumar & Rastogi, 2022). This involves defining procedures for data acquisition, transmission, and storage, as well

as implementing robust security measures to safeguard against unauthorized access, data breaches, and cyber threats. Validating and verifying project data is essential for ensuring its accuracy, completeness, and reliability in AI-driven risk management processes (Nguyen & Gammack, 2022). Organizations should implement mechanisms for data validation and verification, such as data profiling, cleansing, and reconciliation, to identify and rectify errors or inconsistencies in the data. By implementing robust data governance practices, organizations can enhance the reliability and accuracy of AI-driven risk assessment models (Sharma & Kumar, 2023). High-quality, well-governed data ensures that AI algorithms receive accurate inputs, leading to more reliable predictions and insights for risk management decision-making. Effective data governance promotes trust and confidence in AI-driven risk management solutions by ensuring transparency, accountability, and compliance with regulatory requirements (Wang et al., 2022). Transparent data governance processes enable stakeholders to understand how data is collected, processed, and used, fostering trust and credibility in the AI-driven risk management framework. To sum up this strategy, enhanced data governance plays a critical role in enabling organizations to harness the full potential of AI-driven risk management in innovation projects. By establishing data standards, protocols for data collection and storage, and mechanisms for data validation and verification, organizations can enhance the reliability, accuracy, and trustworthiness of AI-driven risk management systems, ultimately leading to better project outcomes and increased organizational resilience.

*Advanced Analytics Capabilities.* Investing in advanced analytics capabilities, particularly machine learning and predictive modeling, is paramount for organizations seeking to enhance risk management in innovation projects. These advanced techniques empower organizations to analyze project data more effectively and anticipate emerging risks, thereby enabling proactive risk management strategies (Chen et al., 2021; Kumar & Rastogi, 2022). Machine learning algorithms, a subset of artificial intelligence, play a pivotal role in analyzing vast amounts of project data to uncover patterns, trends, and correlations that may indicate potential risks (Majumdar & Kulkarni, 2022). By leveraging historical project data, machine learning models can identify recurrent risk factors and

develop predictive insights to anticipate similar risks in future projects (Nguyen & Gammack, 2022). This predictive capability enables organizations to adopt a proactive approach to risk management, allowing them to mitigate potential threats before they escalate into significant issues. Similarly, predictive modeling techniques enable organizations to forecast project risks and simulate various scenarios to evaluate their potential impact on project outcomes (Jha & Tiwari, 2022). By incorporating real-time insights and data-driven simulations, organizations can assess the likelihood and severity of different risk scenarios, enabling informed decision-making and resource allocation (Schwartz & Teerling, 2023). This proactive risk assessment approach allows organizations to prioritize risk mitigation efforts and allocate resources more effectively, ultimately reducing the likelihood of project failure. Moreover, AI-driven risk assessment models leverage both historical data and real-time insights to anticipate potential threats and opportunities in innovation projects (Wang et al., 2022). By continuously analyzing project data and monitoring key performance indicators, these models can provide early warnings of potential risks and enable timely interventions to mitigate their impact (Li & Chai, 2021). This real-time risk assessment capability enables organizations to adapt swiftly to changing project dynamics and emerging threats, enhancing project resilience and success. To sum up this second strategy, investing in advanced analytics capabilities, such as machine learning and predictive modeling, is essential for organizations seeking to enhance risk management in innovation projects. By leveraging these advanced techniques, organizations can analyze project data more effectively, anticipate emerging risks, and adopt proactive risk management strategies to drive project success.

*Cross-functional Collaboration.* Cross-functional collaboration lies at the heart of successful AI-driven risk management initiatives within innovation projects. By bringing together risk management professionals, data scientists, and project teams, organizations can leverage diverse perspectives and expertise to enhance risk management effectiveness and drive innovation. Interdisciplinary teamwork and knowledge sharing play pivotal roles in facilitating the integration of AI technologies into project workflows, ultimately contributing to project success and organizational resilience (Jha & Tiwari, 2022; Wang et al., 2022). Collaboration between risk management professionals and data scientists

enables organizations to make more informed and data-driven decisions. Data scientists possess expertise in analyzing complex datasets and developing AI-driven models, while risk management professionals bring domain knowledge and experience in identifying and managing project risks. By combining these skill sets, organizations can develop AI-driven risk management solutions that provide actionable insights and support strategic decision-making processes (Kumar & Rastogi, 2022; Schwartz & Teerling, 2023). Interdisciplinary teamwork fosters a holistic approach to risk identification, allowing organizations to uncover potential risks from multiple perspectives. Project teams, with their intimate knowledge of project goals and objectives, can collaborate with risk management professionals and data scientists to identify risks that may not be immediately apparent. By leveraging AI technologies, such as natural language processing and predictive analytics, organizations can analyze diverse sources of data to identify emerging risks and anticipate potential threats (Lee et al., 2021; Majumdar & Kulkarni, 2022). Collaboration between diverse teams fosters a culture of innovation and creativity, driving the development of novel AI-driven risk management solutions. By encouraging open communication and knowledge sharing, organizations can leverage the collective expertise of their teams to develop innovative approaches to risk management. Cross-functional collaboration enables organizations to explore new ideas, experiment with different methodologies, and adapt to changing project dynamics, ultimately leading to the development of cutting-edge AI-driven risk management practices (Nguyen & Gammack, 2022; Sharma & Kumar, 2023). Effective cross-functional collaboration ensures that AI-driven risk management initiatives are aligned with organizational goals and objectives. By involving stakeholders from different departments and levels of the organization, organizations can ensure that risk management strategies are integrated into broader organizational strategies. Collaboration between risk management professionals, data scientists, and project teams facilitates the development of AI-driven risk management solutions that address specific organizational needs and priorities, driving value creation and competitive advantage (Zhang et al., 2023; Chen et al., 2021). To sum up, this third strategy, fostering cross-functional collaboration is essential for leveraging the full potential of AI-driven risk management in innovation projects. By bringing together diverse perspectives and expertise, organizations can

enhance decision-making, improve risk identification, drive innovation, and align risk management initiatives with organizational goals, ultimately driving project success and organizational resilience.

*Continuous Improvement.* Continuous improvement is indispensable for optimizing AI-driven risk management processes within innovation projects. By fostering a culture of continuous improvement, organizations can adapt to evolving project dynamics, enhance risk management effectiveness, and drive innovation. This section delves into the importance of continuous improvement in AI-driven risk management and explores strategies for implementing and sustaining a culture of continuous improvement. Continuous improvement ensures that AI-driven risk management processes remain agile, responsive, and aligned with organizational objectives. Regular reviews and refinements of risk management practices enable organizations to identify areas for optimization and implement corrective actions promptly (Nguyen & Gammack, 2022; Schwartz & Teerling, 2023). Moreover, continuous improvement fosters a culture of learning and innovation, driving organizational resilience and competitive advantage in dynamic business environments. Feedback mechanisms and performance metrics play a pivotal role in facilitating continuous improvement in AI-driven risk management. Organizations can leverage feedback from stakeholders, project teams, and AI systems to identify strengths, weaknesses, and opportunities for improvement (Jha & Tiwari, 2022; Sharma & Kumar, 2023). Performance metrics, such as key performance indicators (KPIs) and balanced scorecards, enable organizations to track progress, measure outcomes, and assess the effectiveness of risk management initiatives. Regular reviews and refinements of risk management practices are essential for ensuring the relevance and effectiveness of AI-driven risk management processes. Organizations should conduct periodic assessments of AI models, algorithms, and workflows to identify potential bottlenecks, biases, or performance issues (Majumdar & Kulkarni, 2022; Wang et al., 2022). These reviews provide valuable insights into the performance of AI systems and inform strategic decisions to optimize risk management practices. Innovation projects are characterized by uncertainty and change, necessitating adaptive risk management strategies. Continuous improvement enables organizations to adapt swiftly to evolving project dynamics,

emerging risks, and changing stakeholder expectations (Chen et al., 2021; Kumar & Rastogi, 2022). By embracing a flexible and iterative approach to risk management, organizations can proactively address challenges, seize opportunities, and drive project success. Sustaining a culture of continuous improvement requires commitment, leadership, and organizational support. Leaders should champion the importance of continuous improvement and empower employees to contribute ideas, share feedback, and drive change (Li & Chai, 2021; Cecere & Corrocher, 2023). Moreover, organizations should invest in training, resources, and tools to enable employees to participate actively in continuous improvement initiatives. To sum it up, continuous improvement is essential for optimizing AI-driven risk management processes in innovation projects. By leveraging feedback mechanisms, performance metrics, and regular reviews, organizations can adapt to evolving project dynamics, enhance risk management effectiveness, and drive sustainable growth in today's dynamic business landscape.

*Ethical and Responsible AI Use.* Ethical and responsible use of Artificial Intelligence (AI) technologies is essential to foster trust and confidence in AI-driven risk management solutions. As organizations increasingly rely on AI to enhance decision-making processes and mitigate project risks, it is paramount to establish guidelines and best practices to address ethical considerations and ensure the ethical use of AI in risk management. One of the primary ethical considerations in AI-driven risk management is the potential for algorithmic bias. Biased algorithms can perpetuate discrimination and inequality by making decisions that favor certain groups or individuals over others (Cath, 2021; Liu et al., 2022). To mitigate algorithmic bias, organizations must implement measures such as diverse and representative training datasets, algorithmic audits, and transparency in algorithmic decision-making processes (Kim & Lee, 2023; Zhang et al., 2022). Protecting data privacy is another critical aspect of ethical AI use in risk management. AI algorithms often rely on large volumes of data to make predictions and recommendations, raising concerns about data privacy violations and unauthorized access (Zhou et al., 2023; Park & Choi, 2022). Organizations must adhere to data protection regulations such as GDPR and implement robust data encryption, anonymization, and access control measures to safeguard sensitive information (Wu et al., 2021; Yang &

Chen, 2022). Ensuring transparency and explainability in AI-driven risk management processes is essential for building trust and accountability. Organizations should strive to make AI models and decision-making processes transparent and understandable to stakeholders, including project teams, clients, and regulators (Chung et al., 2023; Hu et al., 2021). Explainable AI techniques, such as model interpretability and feature importance analysis, can help shed light on how AI algorithms make decisions and identify potential biases or errors (Choi et al., 2022; Song et al., 2023). Maintaining human oversight and accountability is crucial to mitigate the risks associated with AI-driven decision-making. While AI algorithms can provide valuable insights and recommendations, human judgment and intervention are still necessary to validate findings, interpret results, and make final decisions (Gupta & Singh, 2021; Lee & Park, 2022). Organizations should establish clear lines of responsibility and accountability for AI-driven risk management processes and ensure that human decision-makers are held accountable for the outcomes of AI-generated recommendations (Huang & Liu, 2022; Wang & Zhang, 2023). Promoting education and training on ethical AI use is essential to ensure that stakeholders understand the ethical implications of AI technologies and are equipped with the knowledge and skills to use them responsibly (Deng et al., 2022; Li & Zhang, 2021). Organizations should invest in training programs and workshops to raise awareness of ethical AI principles and best practices among employees, project teams, and decision-makers (Ma et al., 2023; Wang et al., 2022). To sum up, the final of the listed strategies, ensuring the ethical and responsible use of AI technologies in risk management requires a multifaceted approach that addresses algorithmic bias, data privacy protection, transparency and explainability, human oversight and accountability, and education and training. By implementing guidelines and best practices to mitigate ethical considerations, organizations can build trust, foster accountability, and promote the ethical use of AI in risk management processes.

Now let's write down the listed strategies again and write justification for each of them.

### **Enhanced Data Governance.**

*Recommendation.* Implement robust data governance frameworks to ensure the

quality, integrity, and security of project data.

*Substantiation.* Robust data governance serves as the cornerstone of effective risk management, particularly in the context of innovation projects characterized by the generation and analysis of large volumes of data. By establishing stringent data standards, protocols for data collection and storage, and mechanisms for data validation and verification, organizations can instill confidence in the accuracy and reliability of project data. This foundational step is vital for ensuring that risk assessments and decision-making processes are based on trustworthy information, thereby minimizing the potential for errors or inaccuracies that could jeopardize project outcomes. Furthermore, data governance frameworks play a pivotal role in mitigating data security risks by implementing robust access controls, encryption measures, and data anonymization techniques. These safeguards help protect sensitive information from unauthorized access or breaches, safeguarding the organization's reputation and maintaining stakeholder trust. The implementation of data governance frameworks entails certain upfront costs, including investments in infrastructure, staff training, and compliance measures. However, these initial expenditures pale in comparison to the long-term benefits they bring. Improved data quality not only enhances the accuracy of risk assessments but also reduces the likelihood of compliance-related issues, thereby minimizing the organization's exposure to regulatory fines and penalties. Additionally, the enhanced decision-making enabled by reliable data governance frameworks can lead to more informed and strategic choices, optimizing resource allocation and driving project success. Therefore, while the upfront investment may be significant, the returns in terms of improved data quality, reduced compliance risks, and enhanced decision-making capabilities far outweigh the initial costs. Moreover, robust data governance frameworks lay the foundation for scalability and sustainability in risk management practices. By establishing standardized processes and procedures for data management, organizations can streamline operations and adapt more effectively to evolving project requirements and regulatory environments. This scalability ensures that risk management practices remain effective and relevant over time, enabling the organization to navigate the complexities of innovation projects with confidence and resilience. In summary, while the implementation of robust data governance frameworks

may require a substantial initial investment, the long-term benefits in terms of improved data quality, reduced compliance risks, and enhanced decision-making capabilities justify the expenditure and position the organization for sustained success in innovation projects.

### **Advanced Analytics Capabilities.**

*Recommendation.* Invest in advanced analytics capabilities, including machine learning and predictive modeling, to analyze project data and identify emerging risks.

*Substantiation.* Advanced analytics techniques, such as machine learning and predictive modeling, enable organizations to extract actionable insights from large and complex datasets, thereby enhancing risk management effectiveness. Machine learning algorithms can identify patterns, correlations, and anomalies in project data that may indicate potential risks or opportunities. By analyzing historical project data, these algorithms can develop predictive models to anticipate future risks and their potential impact on project outcomes. This proactive approach to risk management allows organizations to allocate resources more effectively, prioritize risk mitigation efforts, and reduce the likelihood of project failure. Quantitatively, the adoption of advanced analytics capabilities may require investments in technology infrastructure, talent acquisition, and training. However, the potential benefits, including improved risk assessment accuracy, enhanced decision-making, and cost savings through proactive risk mitigation, justify the investment in advanced analytics.

*Recommendation.* Implement decision support systems powered by advanced analytics to facilitate data-driven decision-making processes.

*Substantiation.* Decision support systems leverage advanced analytics techniques to provide stakeholders with actionable insights and recommendations for risk management and project planning. By integrating machine learning algorithms and predictive modeling capabilities, these systems can analyze project data in real time, identify emerging risks, and recommend appropriate risk mitigation strategies. This enables stakeholders to make informed decisions based on data-driven insights, enhancing the effectiveness and efficiency of risk management processes. Additionally, decision support systems can automate routine decision-making tasks, allowing stakeholders to focus on strategic priorities and high-impact activities. By implementing decision support systems may

involve upfront costs for software licensing, customization, and training. However, studies have shown that organizations that leverage decision-support systems experience significant improvements in decision-making accuracy, speed, and efficiency. These improvements lead to measurable cost savings and performance improvements, as stakeholders can make timely and informed decisions that mitigate project risks and capitalize on opportunities. Therefore, the return on investment for decision support systems is considerable, making them a valuable asset for organizations seeking to enhance risk management effectiveness.

*Recommendation.* Develop predictive risk modeling frameworks to anticipate future risks and assess their potential impact on project outcomes.

*Substantiation.* Predictive risk modeling enables organizations to forecast potential risks and their likelihood of occurrence, allowing stakeholders to prioritize risk mitigation efforts and allocate resources judiciously. By leveraging historical project data and machine learning algorithms, organizations can develop predictive models that identify patterns and trends indicative of future risks. These models enable stakeholders to anticipate potential risks before they escalate into significant issues, enabling proactive risk management strategies. Moreover, predictive risk modeling allows organizations to simulate various risk scenarios and evaluate their potential impact on project outcomes, enabling stakeholders to make informed decisions and mitigate risks effectively. Developing predictive risk modeling frameworks may require investments in data analytics expertise, software tools, and model validation processes. However, the potential benefits of these investments are significant. Studies have shown that organizations that implement predictive risk modeling frameworks experience improvements in risk assessment accuracy, decision-making effectiveness, and project performance. Additionally, the proactive risk management enabled by predictive risk modeling leads to measurable cost savings by averting potential project failures and associated expenses. Therefore, the return on investment for predictive risk modeling frameworks is considerable, making them a worthwhile investment for organizations seeking to enhance risk management capabilities.

*Recommendation.* Implement real-time risk monitoring systems to detect and

respond to emerging risks promptly.

*Substantiation.* Real-time risk monitoring systems enable organizations to monitor project data streams in real time, allowing them to detect and respond to emerging risks promptly. By leveraging advanced analytics techniques, these systems can analyze project data for anomalies, deviations, or critical thresholds, triggering alerts when predefined risk thresholds are exceeded. This enables stakeholders to take immediate action to mitigate identified risks and prevent them from escalating into significant issues. Moreover, real-time risk monitoring systems facilitate rapid response and decision-making, enabling organizations to address emerging risks promptly and effectively. Implementing real-time risk monitoring systems may involve investments in technology infrastructure, software development, and integration with existing systems. However, the potential benefits of these systems far outweigh the costs. Studies have shown that organizations that leverage real-time risk monitoring systems experience significant reductions in risk exposure and project delays. Additionally, the proactive risk mitigation enabled by real-time monitoring systems leads to measurable cost savings by averting potential project failures and associated expenses. Therefore, the return on investment for real-time risk monitoring systems is considerable, making them a valuable asset for organizations aiming to enhance risk management effectiveness.

### **Cross-functional Collaboration.**

*Recommendation.* Foster collaboration between risk management professionals, data scientists, and project teams to leverage diverse perspectives and expertise.

*Substantiation.* Cross-functional collaboration is essential for integrating AI-driven risk management solutions into innovation projects effectively. By bringing together professionals from different disciplines, including risk management, data science, and project management, organizations can leverage diverse perspectives and expertise to develop comprehensive risk management strategies. Collaboration enables risk management professionals to understand the technical capabilities and limitations of AI technologies, while data scientists gain insights into the specific risk management needs and challenges of innovation projects. This interdisciplinary approach fosters innovation, creativity, and knowledge sharing, leading to more robust risk management solutions.

Quantitatively, the investment in cross-functional collaboration may involve costs associated with training, communication, and coordination efforts. However, the potential benefits, including improved risk management outcomes, reduced project risks and enhanced organizational resilience, justify the investment in fostering cross-functional collaboration. Moreover, cross-functional collaboration enhances communication and alignment between different teams, facilitating the integration of AI-driven risk management solutions into existing project workflows. By fostering a collaborative culture that values input from diverse stakeholders, organizations can overcome silos and barriers to innovation, enabling seamless integration of AI technologies into risk management processes. Additionally, cross-functional collaboration promotes shared ownership and accountability for risk management outcomes, empowering teams to work together towards common goals. By breaking down traditional departmental boundaries and encouraging collaboration across functions, organizations can harness the collective expertise of their workforce to develop innovative solutions to complex risk management challenges.

### **Continuous Improvement.**

*Recommendation.* Embrace a culture of continuous improvement by regularly reviewing and refining risk management processes and practices.

*Substantiation.* Continuous improvement is essential for adapting to evolving project dynamics, emerging risks, and technological advancements. By regularly reviewing risk management processes and practices, organizations can identify areas for optimization, implement lessons learned from past projects, and incorporate feedback from stakeholders. This iterative approach to risk management enables organizations to stay agile, responsive, and proactive in addressing emerging threats and opportunities. Moreover, continuous improvement fosters a culture of innovation, experimentation, and learning, driving organizational growth and competitiveness. By encouraging employees to seek out opportunities for improvement and providing the necessary support and resources, organizations can tap into the collective wisdom and creativity of their workforce, driving innovation and driving continuous improvement forward. Quantitatively, the investment in continuous improvement may involve costs associated

with process reviews, training programs, and technology upgrades. However, the potential benefits far outweigh the costs. Improved risk management effectiveness leads to fewer incidents of project failure, reduced costs associated with risk mitigation, and enhanced project outcomes. Additionally, increased stakeholder satisfaction and confidence in the organization's ability to manage risks effectively can lead to improved reputation and competitive advantage. Therefore, the investment in embracing a culture of continuous improvement is not only justified but essential for organizations seeking to thrive in today's fast-paced and unpredictable business environment. By continuously refining their risk management processes and practices, organizations can adapt to change, seize opportunities, and achieve long-term success.

### **Ethical and Responsible AI Use.**

*Recommendation.* Ensure that AI technologies are used ethically and responsibly by establishing guidelines and best practices for AI implementation.

*Substantiation.* Ethical and responsible use of AI technologies is essential for building trust, maintaining credibility, and mitigating risks associated with AI-driven decision-making. By establishing guidelines and best practices for AI implementation, organizations can ensure that AI technologies are deployed in a manner that respects ethical principles, legal regulations, and stakeholder expectations. This includes addressing ethical considerations such as algorithmic bias, data privacy violations, and lack of transparency in AI-driven risk management processes. Moreover, responsible AI use promotes accountability, fairness, and transparency in decision-making, enhancing organizational reputation and stakeholder trust. Quantitatively, the investment in ethical and responsible AI use may involve costs associated with developing and implementing ethical guidelines, conducting ethical assessments, and ensuring compliance with regulatory requirements. However, the potential benefits, including reduced legal risks, enhanced brand reputation, and increased stakeholder trust, justify the investment in ensuring ethical and responsible AI use. In addition to establishing guidelines, organizations should also invest in ongoing monitoring and auditing mechanisms to ensure compliance with ethical standards and identify and address any potential ethical issues that may arise in AI-driven risk management processes. This proactive approach not only helps

mitigate ethical risks but also demonstrates the organization's commitment to ethical behavior and responsible use of AI technologies. Furthermore, organizations should prioritize transparency and accountability in AI-driven decision-making by providing stakeholders with clear explanations of how AI algorithms work, the data sources used, and the potential limitations or biases inherent in the technology. This transparency fosters trust and confidence in AI-driven risk management processes and ensures that stakeholders are informed and empowered to participate in decision-making processes. Ensuring ethical and responsible AI use is not only a moral imperative but also a strategic necessity for organizations seeking to leverage AI technologies to enhance risk management effectiveness. By establishing clear guidelines, investing in monitoring and auditing mechanisms, and prioritizing transparency and accountability, organizations can mitigate ethical risks, build trust with stakeholders, and unlock the full potential of AI-driven risk management solutions.

### ***3.3 Optimality, Target Efficiency, Practical Implementation, and Recommendations for AI-driven Risk Management in Pharmaceutical Field.***

As organizations embark on the journey of integrating AI-driven risk management approaches into innovation project management frameworks, several practical considerations come to the fore. It is imperative to assess the organizational readiness for AI adoption, including the availability of data infrastructure, technological capabilities, and human resources (Brown & Rozenblit, 2023; Nguyen & Gammack, 2022). Furthermore, organizations must navigate ethical and regulatory considerations surrounding AI implementation, ensuring transparency, accountability, and fairness in decision-making processes (Gartner, 2023; World Economic Forum, 2021).

*Assessing Organizational Readiness for AI Adoption.* Before implementing AI-driven risk management solutions, organizations must conduct a comprehensive assessment of their readiness for AI adoption (Brown & Rozenblit, 2023). This involves evaluating the availability of data infrastructure, including data quality, accessibility, and security measures (Nguyen & Gammack, 2022). Additionally, organizations need to assess their technological capabilities, including the availability of AI expertise, software tools, and hardware infrastructure (Gartner, 2023). Human resources are also critical, as

organizations must ensure that employees have the necessary skills and training to effectively leverage AI technologies (World Economic Forum, 2021).

*Navigating Ethical and Regulatory Considerations.* Ethical and regulatory considerations are paramount when implementing AI-driven risk management solutions (Gartner, 2023). Organizations must adhere to ethical principles such as fairness, transparency, and accountability in their use of AI technologies (Brown & Rozenblit, 2023). Moreover, organizations need to comply with regulatory requirements governing data privacy, security, and algorithmic transparency (Nguyen & Gammack, 2022). By establishing robust governance frameworks and oversight mechanisms, organizations can mitigate ethical and regulatory risks associated with AI implementation (World Economic Forum, 2021).

*Offering Practical Recommendations.* To effectively leverage AI technologies in risk management for innovation projects, organizations should follow best practices and practical recommendations (Gartner, 2023). This includes implementing robust data governance processes to ensure data quality, integrity, and compliance (Brown & Rozenblit, 2023). Organizations should also invest in model validation and testing procedures to verify the accuracy and reliability of AI-driven risk management models (Nguyen & Gammack, 2022). Stakeholder engagement is crucial, as organizations must involve key stakeholders, including project teams, executives, and regulatory bodies, in the AI implementation process (Gartner, 2023). Finally, organizations should prioritize change management initiatives to foster a culture of AI adoption and innovation within the organization (World Economic Forum, 2021).

To sum it up, integrating AI-driven risk management approaches into innovation project management frameworks requires careful consideration of organizational readiness, ethical and regulatory considerations, and practical implementation strategies. By assessing organizational readiness, navigating ethical and regulatory challenges, and following best practices and practical recommendations, organizations can effectively leverage AI technologies to enhance risk management effectiveness and drive sustainable growth in today's dynamic business landscape.

In evaluating the proposals outlined in this chapter, it is imperative to consider their

optimality, target efficiency, and practical implementation. Each recommendation will be inspected against predefined criteria to ensure alignment with the organization's strategic objectives and feasibility within existing resource constraints. For instance, let's consider the recommendation to invest in advanced analytics capabilities, including machine learning and predictive modeling, to enhance risk management in innovation projects. In Ukraine, a software development company may choose to implement this recommendation by leveraging existing expertise in data science and analytics or by partnering with external vendors specializing in AI technologies. By conducting a cost-benefit analysis, the company can assess the potential return on investment (ROI) of implementing advanced analytics capabilities. This analysis would consider factors such as the expected reduction in project risks, the anticipated cost savings from proactive risk mitigation, and the potential for improved project outcomes. By aligning the investment in advanced analytics with the organization's strategic goals and financial constraints, the company can ensure that the proposal meets the criteria of optimality and target efficiency. Similarly, let's examine the recommendation to establish robust data governance frameworks to ensure the quality, integrity, and security of project data. In Ukraine, a financial institution may choose to implement this recommendation by developing internal policies and procedures for data management, establishing data governance committees, and investing in data security technologies. By conducting a gap analysis to identify areas of improvement in current data governance practices, the institution can prioritize investments in areas with the highest impact on risk management effectiveness. This approach ensures that the proposal is aligned with the organization's strategic objectives and addresses critical vulnerabilities in data management processes, thereby meeting the criteria of optimality and target efficiency. In evaluating the practical implementation of each recommendation, it is essential to consider the organization's capacity for change and the readiness of employees to adopt new technologies and practices. In Ukraine, organizations may encounter challenges related to skills shortages, regulatory constraints, and cultural barriers to change. Therefore, it is crucial to develop comprehensive change management plans and provide adequate training and support to employees to ensure successful implementation of the proposed enhancements.

To cut it short, by aligning each recommendation with the organization's strategic objectives, conducting thorough cost-benefit analyses, and addressing potential implementation challenges, organizations in Ukraine can effectively enhance risk management in innovation projects and drive sustainable growth and competitiveness in today's dynamic business environment.

For example, let's take a look at the "COR-Medical" company. The main direction of activity of this company is to provide medical institutions in Ukraine with modern high-quality equipment and consumables for cardiac and vascular surgery, interventional cardiology and oncology, treatment of spinal pathologies, neurosurgery, blood processing systems, separation and production of blood components, medical equipment, systems for head-mounted magnification and lighting. Just from its description, it should be easy to understand that the company is rather niche and has a narrow array of customers, which mainly consist of loyal customers that are in one or another way related to the healthcare field.

Let's conduct a SWOT analysis to see the strengths, weaknesses, threats, and opportunities of this company (Table 3.3).

Table 3.3. SWOT analysis of the COR-Medical company

SWOT ANALYSIS	
<b>Strengths:</b> <i>Niche Market.</i> COR-Medical operates in a specialized niche, focusing on providing high-quality medical equipment and consumables to medical institutions. This specialization allows the company to establish expertise and credibility within the medical field. <i>High-Quality Products.</i> The company emphasizes providing modern, high-quality equipment and consumables, which can enhance its reputation and attract discerning customers who prioritize quality. <i>Loyal Customer Base.</i> COR-Medical likely has a loyal customer base consisting of medical	<b>Weaknesses:</b> <i>Limited Market Reach.</i> Operating in a niche market means that COR-Medical's customer base is relatively limited, potentially restricting its growth opportunities compared to companies with broader target markets. <i>Dependence on Healthcare Sector.</i> The company's success is closely tied to the healthcare sector, making it vulnerable to fluctuations in healthcare spending, regulatory changes, and other industry-specific factors. <i>Intensive Competition.</i> Although COR-Medical specializes in a niche market, it still faces

<p>institutions that rely on its products for crucial medical procedures. This loyalty can lead to repeat business and positive word-of-mouth referrals.</p> <p><i>Comprehensive Product Range.</i> The company offers a wide range of products catering to various medical specialties, including cardiac and vascular surgery, oncology, neurosurgery, and more. This breadth of offerings can attract customers looking for a one-stop solution for their medical equipment needs.</p>	<p>competition from other medical equipment distributors and manufacturers. Competing on quality and service may require significant investment and effort.</p> <p><i>Reliance on Supplier Relationships.</i> The company's ability to provide high-quality products depends on maintaining strong relationships with suppliers. Any disruptions or issues with suppliers could affect product availability and quality.</p>
<p><b>Opportunities:</b></p> <p><i>Expansion into New Markets.</i> COR-Medical could explore opportunities to expand its product offerings or target new geographic markets to diversify its customer base and increase revenue streams.</p> <p><i>Technological Advancements.</i> The ongoing advancements in medical technology present opportunities for COR-Medical to introduce innovative products and stay ahead of competitors.</p> <p><i>Strategic Partnerships.</i> Collaborating with hospitals, medical research institutions, or other healthcare organizations could enhance COR-Medical's visibility, credibility, and access to potential customers.</p> <p><i>Offering Value-Added Services.</i> Beyond product sales, the company could provide value-added services such as equipment maintenance, training, or consulting, which can further differentiate it from competitors and increase customer loyalty.</p>	<p><b>Threats:</b></p> <p><i>Regulatory Changes.</i> Changes in healthcare regulations or compliance requirements could impact COR-Medical's operations, product offerings, and sales processes, leading to increased costs or decreased demand.</p> <p><i>Economic Instability.</i> Economic downturns or fluctuations in healthcare spending may reduce the budgets of medical institutions, affecting their purchasing decisions and potentially harming COR-Medical's revenue.</p> <p><i>Technological Disruption.</i> Rapid technological advancements or the emergence of disruptive technologies could render existing products obsolete or create new competitors, posing a threat to COR-Medical's market position.</p> <p><i>Supply Chain Risks.</i> Disruptions in the supply chain, such as shortages, delays, or quality issues with suppliers, could disrupt COR-Medical's ability to deliver products to customers and maintain its reputation for quality and reliability.</p>

Source: Developed by author

Analyzing the strengths, weaknesses, opportunities, and threats (SWOT) of COR-Medical provides valuable insights into the potential viability and cost-effectiveness of

implementing AI technologies for risk management within the company. Strengths such as a niche market focus, high-quality products, and a loyal customer base indicate that COR-Medical has established expertise and credibility within the medical field. These strengths provide a solid foundation for leveraging AI technologies for risk management, as the company likely has access to valuable data and domain knowledge that can enhance AI-driven risk assessment and mitigation strategies.

For example, one of the most suitable AI solutions that could be implemented by COR-Medical is predictive analytics for inventory management. By analyzing historical sales data, seasonal trends, and customer demand patterns using machine learning algorithms, the company can forecast future demand for medical equipment and consumables more accurately. This proactive approach to inventory management can help COR-Medical optimize stock levels, reduce excess inventory, minimize stockouts, and improve overall supply chain efficiency. Additionally, AI-powered predictive maintenance solutions can help COR-Medical monitor the performance of medical equipment and identify potential maintenance issues before they escalate into costly breakdowns or disruptions. By analyzing equipment sensor data and usage patterns, these solutions can predict when equipment is likely to require maintenance or replacement, enabling proactive maintenance scheduling and reducing downtime for medical institutions. Furthermore, AI-driven risk assessment models can help COR-Medical identify and prioritize potential risks associated with supplier relationships, regulatory compliance, and market dynamics. By analyzing data from various sources, including supplier performance metrics, regulatory updates, and market trends, these models can provide insights into emerging risks and opportunities, enabling the company to make informed decisions and mitigate potential threats more effectively.

However, there are also factors that may pose challenges to the implementation of AI technologies for risk management within COR-Medical. Weaknesses such as limited market reach and dependence on the healthcare sector may impact the availability and quality of data required for AI-driven risk assessment and mitigation. Additionally, intensive competition and reliance on supplier relationships may require significant investment and effort to overcome.

Overall, even though there are opportunities for COR-Medical to leverage AI technologies for risk management, careful consideration of the company's strengths, weaknesses, opportunities, and threats is necessary to determine the viability and cost-effectiveness of implementation. By focusing on areas where AI solutions can address specific business challenges and deliver tangible benefits, COR-Medical can maximize the value of AI-driven risk management initiatives and enhance its competitiveness in the medical equipment distribution market.

To make sure that the conducted assessment is correct, let's conduct another analysis, which is the PESTEL analysis for COR-Medical (Table 3.4.1), and see whether the perspective will receive a few changes or remain static.

Table 3.3.1

PESTEL analysis of the COR-Medical

Political	Economic	Social
<p><i>Stability of Government.</i> Political stability in Ukraine can affect COR-Medical's operations, as instability or frequent changes in government policies, which can be seen rather frequently nowadays, may create uncertainty in the business environment.</p> <p><i>Healthcare Policies.</i> Government policies related to healthcare funding, insurance coverage, and regulations can impact the demand for medical equipment and consumables, influencing COR-Medical's sales and growth prospects.</p> <p><i>Trade Regulations.</i> Changes in</p>	<p><i>Economic Growth.</i> Overall economic growth and stability in Ukraine can influence healthcare spending and investment in medical infrastructure, affecting the demand for COR-Medical's products and services.</p> <p><i>Currency Fluctuations.</i> Fluctuations in currency exchange rates can impact the cost of imported equipment and consumables, as well as the profitability of COR-Medical's operations.</p> <p><i>Income Levels.</i> Disposable income levels and purchasing power of medical institutions can affect their ability to invest in</p>	<p><i>Healthcare Trends.</i> Changing demographics, healthcare needs, and patient preferences can drive demand for specific medical procedures and treatments, shaping the market for COR-Medical's products.</p> <p><i>Healthcare Infrastructure.</i> The quality and accessibility of healthcare infrastructure in Ukraine can impact the demand for medical equipment and consumables, influencing COR-Medical's market opportunities and growth potential.</p> <p><i>Healthcare Professional Skills.</i> Availability of skilled healthcare professionals and their training in</p>

trade regulations, tariffs, or import/export policies can affect the cost and availability of imported medical equipment and supplies, potentially impacting COR-Medical's procurement and pricing strategies.	modern medical equipment and consumables, influencing COR-Medical's sales volumes and revenue.	using modern medical equipment can affect the adoption and utilization of COR-Medical's products.
<b>Technological</b>	<b>Environmental</b>	<b>Legal</b>
<p><i>Advancements in Medical Technology.</i> Rapid advancements in medical technology can create opportunities for COR-Medical to introduce innovative products and solutions, staying competitive in the market.</p> <p><i>Digitalization of Healthcare.</i> Increasing digitalization and adoption of telemedicine technologies can influence the demand for certain types of medical equipment and consumables, driving changes in COR-Medical's product offerings and sales strategies.</p>	<p><i>Regulatory Compliance.</i> Environmental regulations related to waste disposal, energy efficiency, and product safety can impact COR-Medical's operations and product development processes, requiring compliance measures and potentially affecting costs.</p> <p><i>Sustainability Initiatives.</i> Growing awareness of environmental issues and sustainability may influence purchasing decisions among medical institutions, leading to demand for eco-friendly or sustainable medical equipment and consumables.</p>	<p><i>Regulatory Compliance.</i> Compliance with laws and regulations governing the medical device industry, including product safety standards, licensing requirements, and import/export regulations, is critical for COR-Medical to operate legally and avoid penalties.</p> <p><i>Intellectual Property Rights.</i> Protecting intellectual property rights for proprietary products and technologies is essential for COR-Medical to maintain its competitive advantage and prevent unauthorized use or replication by competitors.</p>

Developed by author

Analyzing the PESTEL analysis alongside the SWOT analysis for COR-Medical provides additional insights into the external factors that may influence the company's operations and strategic decisions. While the SWOT analysis focuses on internal strengths and weaknesses, as well as external opportunities and threats, the PESTEL analysis delves deeper into the broader external environment, including political, economic, social, technological, environmental, and legal factors. Let's examine how the PESTEL analysis might shed light on potential AI solutions for risk management and whether it offers a

different perspective compared to the SWOT analysis.

Firstly, considering the political factors outlined in the PESTEL analysis, such as the stability of government and healthcare policies, there may be opportunities for COR-Medical to leverage AI technologies for regulatory compliance and policy analysis. AI-driven tools can help the company stay abreast of changes in healthcare regulations, ensuring that its operations and product offerings remain compliant with evolving legal requirements. Additionally, AI solutions could assist in monitoring political stability indicators and assessing potential risks associated with political instability, enabling the company to adjust its strategies accordingly.

Economically, the PESTEL analysis highlights factors such as economic growth, currency fluctuations, and income levels, which can impact COR-Medical's financial performance and market dynamics. AI-powered financial forecasting and risk modeling tools can help the company mitigate risks associated with economic volatility and currency fluctuations by providing real-time insights into market trends and financial indicators. Moreover, AI-driven customer segmentation and pricing optimization algorithms can help COR-Medical adapt its pricing strategies based on variations in income levels and purchasing power among medical institutions.

From a social perspective, the PESTEL analysis emphasizes healthcare trends, infrastructure, and professional skills, indicating potential opportunities for AI solutions in addressing societal healthcare challenges. For instance, AI-powered healthcare analytics platforms can analyze demographic data and healthcare utilization patterns to identify emerging healthcare trends and patient needs, guiding COR-Medical's product development and marketing strategies. Furthermore, AI-driven medical education and training platforms can help address skill gaps among healthcare professionals, enhancing their proficiency in using modern medical equipment and consumables supplied by COR-Medical.

Technological advancements, as highlighted in the PESTEL analysis, present numerous opportunities for AI innovation within COR-Medical. In addition to the predictive analytics and predictive maintenance solutions mentioned in the SWOT analysis, AI-driven telemedicine platforms and virtual reality (VR) simulations can

revolutionize remote medical consultations and training programs, expanding the company's service offerings and market reach. Moreover, AI-powered image analysis and diagnostics tools can enhance the accuracy and efficiency of medical procedures, driving demand for COR-Medical's products among healthcare providers.

Environmental and legal factors also play crucial roles in shaping COR-Medical's risk management strategies and AI implementation decisions. AI-driven environmental monitoring and sustainability assessment tools can help the company minimize its environmental footprint and comply with regulatory requirements related to product safety and waste management. Additionally, AI-powered legal research and contract management platforms can streamline compliance processes and mitigate legal risks associated with supplier relationships and intellectual property rights.

Overall, by combining the findings from both analyses, COR-Medical can develop a comprehensive understanding of its business environment and identify opportunities for leveraging AI technologies to enhance its competitive advantage and achieve sustainable growth in the medical equipment distribution market. However, it is worth remembering that this is purely analytical observation and may greatly differentiate if brought up during company meetings where participants discuss and suggest potential strategies.

## CONCLUSIONS AND PROPOSALS

As to conclusions, there are several key points that can be deducted from the contents of this work as we slowly completed each of outlined tasks, studied both our object and subject of this work.

Chapter 1 of this work lays the foundational framework for understanding risk management in innovation projects. It starts by providing an overview of the theoretical underpinnings of risk management and gradually transitions into exploring the role of Artificial Intelligence (AI) in enhancing risk management practices in the digital age. Through various sections, the chapter delves into different aspects of risk management, from risk assessment and proactive risk mitigation to stakeholder engagement, agile project management, and the use of modern technology and tools.

In Subchapter 1.1, the focus is on equipping readers with common information on risk assessment and identification. Understanding these fundamental concepts is crucial for comprehending subsequent discussions on risk management strategies and methodologies. We also delved into proactive risk mitigation strategies, shedding light on traditional approaches to managing risks in innovative projects. By examining these established methods, readers gain insights into the evolution of risk management practices and the challenges associated with mitigating risks in dynamic environments. Stakeholder engagement, which is the last part of the Subchapter 1.1, emerges as a critical component of risk management. Building and maintaining strong relationships with stakeholders is essential for the success of innovation projects, as stakeholders play a pivotal role in supporting and influencing project outcomes.

In Subchapter 1.2, we introduced Agile Project Management, a popular methodology for risk management. Understanding Agile principles and practices is essential for adapting to the dynamic nature of innovation projects and fostering collaboration among project teams. Then, the focus shifted to modern technology and tools revolutionizing risk management. Advancements in AI, data analytics, and automation offer new opportunities for identifying, assessing, and mitigating risks more effectively in innovation projects. Lastly, we explored the concept of Knowledge Management, highlighting its significance as a central piece of the risk management puzzle. Leveraging

knowledge assets and fostering a culture of knowledge sharing are critical for enhancing decision-making and risk mitigation strategies.

At Subchapter 1.3, we explore case studies to bridge theory with practice. By examining real-world examples of companies employing traditional methodologies to enhance risk management, readers gain insights into the practical challenges and opportunities. These case studies serve as a springboard for contemplating the potential role of AI in augmenting risk management practices in similar scenarios.

The main problem with Chapter 1 is the fact that it is mostly consists of walls of text without any particularly interesting or engaging information, which lowers the interest for both the reader to read and the writer to write this. It is indeed theoretical material that is important for further understanding of the topic. As for its improvements, by adding pictures, graphs, statistical data, or other more visual types of data, the engagement of this particular part of the work could be improved.

Chapter 2 delved into the integration of Artificial Intelligence (AI) in risk management for innovation projects, acknowledging the inherent uncertainties and challenges these projects entail. Through interdisciplinary collaboration, the chapter aimed to illuminate the path towards leveraging AI as a catalyst for innovation, driving sustainable growth, and competitive advantage in organizational management. It also explored the alignment of AI impact on management with risk management strategies in innovation projects, emphasizing the transformative potential of integrating AI technologies into traditional risk management approaches. By harnessing AI, organizations can enhance risk identification, assessment, and mitigation processes, leading to improved project outcomes and fortified organizational resilience. AI augments traditional approaches by leveraging advanced analytics techniques such as machine learning and natural language processing to analyze vast datasets and uncover hidden patterns. This proactive and data-driven approach enables organizations to anticipate emerging threats and implement timely interventions, minimizing project delays and enhancing overall performance. Furthermore, AI-driven risk management fortifies organizational resilience by empowering organizations to adapt and thrive in rapidly changing environments.

In Chapter 2.2 we delved into the research of “COR-Medical” company, its current

state, while providing several figures and tables for better visualization of the analysis. This subchapter is followed by conducted survey and research that were conducted in Ukrainian enterprises and helped us achieve yet another of the mentioned tasks.

Chapter 3 focused on providing specific recommendations and proposals for managing the impact of Artificial Intelligence (AI) on risk management within innovation projects in pharmaceutical field. These recommendations are grounded in theoretical frameworks outlined in the study's first section and insights from the analysis conducted in the second section. Each recommendation is accompanied by a detailed justification, including rationale, potential impact, and quantitative justification where applicable, such as cost-benefit analyses. The proposals are evaluated based on optimality, target efficiency, and practical implementation, ensuring alignment with organizational strategic objectives and feasibility within resource constraints. By implementing these recommendations, organizations can effectively manage the impact of AI on risk management in innovation projects in pharmaceutical field, fostering sustainable growth and competitive advantage in today's dynamic business landscape. There, this work reaches its sunset, which is compelling the last of the outlined task, by defining the optimality, and target efficiency of already mentioned strategies and giving direct recommendations on how “COR-Medical” activities can be improved. We achieve that through conducting an analytical analysis of a practical implementation of such strategies.

To finally conclude this work, it is worth saying that whether we like it or not, AI is slowly starting to course through the veins of almost every single field, somewhere its influence is getting stronger and rapidly, somewhere it has only small and underdeveloped influence. Just a few years ago, AI was only an idea or topic of discussion, and in just a few years, it quickly changed and developed into a modern trendsetter, and the most popular tool. Objectively speaking, AI is a double edged-sword filled with controversy. On one hand, it is a so-called future, the greatest tool, which can ease and automate most of the processes, and create even more. On another hand, it was implemented so rapidly, that it has created several problems that are yet to find their solutions. And the main problem is of course the ethical use of AI. It is not a rare occurrence to see some articles where it is explained how AI is used for shady purposes to gain an advantage and earn something,

whether it could be money, information, and whatnot. For example, recently, there was a case where people started generating photos and videos while using existing people as base models, and committing fraud by creating fake pages and fooling ignorant people and gaining money, or private information from them. This only enriches the understanding that it is impossible to correctly evaluate and analyze the impact of Artificial Intelligence at the current stage of its development, as it is yet to slow down its development and popularization.

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## APPENDIX A

This Appendix contains the current staff list of the “COR-Medical” company.

Table 1. The current staff list of the “COR-Medical” company.

First and last name	Position
<b>LLC «Iridium»</b>	
Maksym Ihorovych Aksyonov	Service engineer
Andriy Vyacheslavovych Andreev	Forwarder driver
Lyubov Mykolaivna Bebeshko	Chief Accountant
Oleksandra Yuryivna Gorbunova	Implantation manager
Yulia Volodymyrivna Ivanysko	Sales manager
Kateryna Oleksandrivna Kalyniuk	Storekeeper
Viktoriya Serhiivna Kovalenko	Assistant manager
Stepan Serhiyovych Komarchuk	Service engineer
Vadim Viktorovych Kosenko	Sales manager
Andriy Valeriyovych Kosko	Storekeeper
Olga Mykolayivna Kunytska	Sales manager
Oleksiy Serhiyevich Kutuzov	Sales manager
Evgeny Oleksandrovich Kutsenko	Courier
Lydia Ivanivna Levchenko	Business manager
Yevhenii Evgeniyovych Maksimenko	Video operator
Anastasia Valeriivna Myanovska	Manager Supply
Olga Serhiivna Myanovska	Sales manager
Lesya Volodymyrivna Ovsiychuk	Manager
Yevhen Hryhorovych Ovcharenko	Courier
Inna Oleksandrivna Oleinyk	Sales manager
Volodymyr Mykolayovych Pantasenko	Forwarder driver
Pavlo Petrovych Popov	Sales manager
Anastasia Ihorivna Prikhodko	Sales manager
Diana Oleksandrivna Rumyantseva	Financial analyst
Andriy Mykolayovych Tarasenko	Sales manager
Vyacheslav Ivanovich Ternesiuk	Customs manager
Lyudmila Vitaliyivna Furmanova	Sales manager
Serhiy Yuriyovych Chudovsky	ZED and logistics manager
Oleksandr Anatoliyovych Shapoval	Sales manager
Yuliya Anatoliyivna Yaruta	Sales manager
<b>LLC «Optimatrading»</b>	
Harutyun Vilyamsovich Amalyan	CFO
Nataliya Sergiivna Aschi	Sales manager
Yuliya Serhiivna Bova	Regional development manager
Oleg Serhiyovych Bondarenko	Sales manager

Hanna Gennadiivna Golubeva	Sales manager
Andrii Andriyovych Goncharenko	Storekeeper
Svitlana Anatoliyivna Gromova	Sales manager
Anna Leonidivna Degtyarova	Head of the sales department
Inesa Volodymyrivna Yeremenko	Sales manager
Tetyana Hnativna Zakharchenko	Sales manager
Artur Evgenovich Kovinko	Sales manager
Anna Anatoliyivna Kozachenko	Assistant manager
Dmytro Leonidovych Kravchuk	Sales manager
Kateryna Anatoliyivna Loseva	Assistant manager of logistics
Serhiy Oleksiyovych Luchko	Sales manager
Iryna Oleksandrivna Maksimenko	Manager Supply
Iryna Valeriivna Marukhlenko	Manager of administrative activities
Yurii Anatoliyovych Mykhailichenko	Sales manager
Yevhen Ihorovych Nagorny	Regional development manager
Iryna Viktorivna Nazarova	Chief Accountant
Olga Petrivna Nezenko	Financial analyst
Oleksiy Anatoliyovych Ostranytsya	Leading manager of foreign economic activity
Andriy Valentinovych Pirog	Regional development manager
Serhiy Yuriyovych Pokutniy	Courier
Andriy Mykhailovych Romanov	Regional development manager
Tetyana Anatoliivna Rudnytska	Sales manager
Oleksandr Mykhailovych Rumyantsev	Assistant manager
Anna Vyacheslavivna Ternesiuk	Assistant sales manager
Oleksandr Anatoliyovych Shapoval	Service engineer
Oleksandr Ihorovych Yakovenko	Regional development manager
Olga Volodymyrivna Yatsenyuk	Regional development manager

Source: Developed by author.