

Assessing Cyber security Risks in Project Life Cycles: An Integrated Model for Effective Risk Management

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Abstract: Cybersecurity risks are a growing concern across project life cycles, especially as digital dependencies increase. This study presents an integrated model designed to identify, assess, and mitigate cybersecurity risks within each phase of a project life cycle, from initiation through closure. Through a combination of qualitative interviews, case studies, and statistical analysis, this study maps specific cyber threats to project phases, analyzes the effectiveness of various cybersecurity software tools (including FireEye, Splunk, and Cylance), and examines the impact of industry type, project duration, and team size on risk levels. The findings reveal that tailored security measures, such as encryption in planning and real-time monitoring during execution, significantly improve risk management outcomes. The proposed model received high ratings from experts for adaptability and risk reduction, providing a practical, structured approach for managing cybersecurity in diverse project environments. This research underscores the importance of phase-specific cybersecurity strategies to enhance data protection and operational continuity.

Keywords: Cybersecurity, Project Life Cycle, Risk Management, Integrated Model, Data Protection, Cyber Threats, Risk Analysis, Cyber Risk, Risk Assessment.

INTRODUCTION

BACKGROUND OF THE STUDY

Cyber security has become a fundamental concern for organizations across all sectors due to the increasing complexity and interdependence of digital systems (AL-Hawamleh, 2024). As businesses embrace digital transformation, the vulnerabilities within project management practices are amplified (Chillapalli and Murganoor, 2024). Project life cycles—from initiation and planning to execution, monitoring, and closure—are particularly susceptible to cyber threats due to the continuous data exchanges, resource allocations, and strategic planning they involve (Mohebbi, *et al.*, 2020). Many organizations, however, often overlook cyber security as an essential aspect of project management, leaving critical phases of projects exposed to potential cyber-attacks. Understanding the significance of cyber security within the project life cycle framework is essential for building resilient project management practices (Dunn Cavelty, *et al.*, 2020).

Importance of Cyber security in Project Management

In the context of project management, cyber security encompasses measures to protect data integrity, ensure confidentiality, and maintain uninterrupted operations (Ustundag, *et al.*, 2018). Each phase of a project life cycle presents unique risks, and failing to address these risks can lead to project delays, financial losses, reputational damage, and even breaches of legal compliance (Jindal and Nanda, 2024). From the planning phase, where sensitive data is frequently shared

among stakeholders, to the execution phase, where information systems are actively engaged, cyber security remains a critical factor for success (Saeed, *et al.*, 2023). By incorporating robust cyber security protocols, project managers can protect projects against emerging cyber threats and enhance the project's success and long-term sustainability (Kadapal, *et al.*, 2024).

Challenges in Cyber security for Project Life Cycles

Traditional approaches to cyber security within project management tend to be reactive, addressing issues only after a threat has materialized (Choucri, *et al.*, 2014). However, this approach may not be sufficient in today's dynamic digital landscape (Jain, 2023). Projects often face multifaceted threats, including unauthorized access, malware attacks, and data breaches, that can compromise entire project structures. Another challenge is the lack of tailored cyber security models specific to project management life cycles (More and Unnikrishnan, 2024; Rahman, *et al.*, 2024). Current cyber security models are often broad, lacking the specificity to address unique project risks effectively (Garcia-Perez, *et al.*, 2023). These challenges necessitate a structured approach to embed cyber security practices into each project phase, from initiation through closure, creating a seamless integration that minimizes potential security gaps (Jindal, 2024; Shabbir, *et al.*, 2024).

RESEARCH GAP AND PROBLEM STATEMENT

Despite the critical importance of cyber security in project management, there is a lack of comprehensive models that integrate cyber security risk management across all phases of a project's life cycle. Existing risk management frameworks primarily focus on financial, operational, or environmental risks, often treating cyber security as a secondary concern. This limited focus poses a risk as it overlooks the unique security vulnerabilities inherent in project transitions and data exchanges. The primary problem addressed in this research is the absence of an integrated model specifically designed to manage cyber security risks throughout the entire project life cycle. Such a model is essential to proactively mitigate threats and strengthen project resilience.

OBJECTIVE OF THE STUDY

This study aims to develop an integrated model that addresses cyber security risk management across each phase of the project life cycle. The model will be structured to help project managers identify, assess, and mitigate cyber risks from project inception to closure. By providing a comprehensive approach to cyber security in project management, the study intends to fill a critical gap in current risk management practices.

Significance of the Study

An integrated cyber security risk management model for project life cycles will benefit organizations by offering a structured, proactive approach to tackling cyber security threats. Implementing this model can lead to enhanced project resilience, better protection of sensitive data, and reduced likelihood of disruptions caused by cyber incidents. Furthermore, the model can be adapted across various industries, helping project managers create secure, efficient workflows that prioritize data protection and operational continuity.

METHODOLOGY

Research Design

This study employs a qualitative research design, combined with quantitative statistical analysis, to develop an integrated cybersecurity risk management model tailored to the project life cycle. The research is structured to capture insights from cybersecurity experts, project managers, and case studies to ensure a well-rounded approach. This mixed-methods design is intended to provide

a robust foundation for understanding the nuances of cybersecurity risks in different project phases, as well as the effectiveness of various risk mitigation strategies.

Data Collection

Data collection is twofold, consisting of interviews with industry experts and a review of documented case studies on cybersecurity incidents in project management. The expert interviews involve cybersecurity professionals, project managers, and risk analysts who offer perspectives on identifying, assessing, and mitigating cybersecurity threats at various project stages. The case studies focus on projects in diverse sectors, such as finance, healthcare, and information technology, to capture a comprehensive picture of cybersecurity risks across different industries. Additional data is obtained from relevant organizational reports and cybersecurity software outputs, which provide a baseline understanding of existing threat management practices in project environments.

Cybersecurity Software Tools and Techniques

To evaluate and monitor cybersecurity risks effectively, this study incorporates data from leading cybersecurity software solutions. Key tools include Splunk, Wireshark, Nessus, FireEye, and Cylance.

- Splunk is used for security information and event management (SIEM), providing data insights by analyzing machine data, which can help detect potential threats in real-time.
- Wireshark assists in packet analysis and network traffic monitoring, essential for identifying unusual patterns that may indicate a security breach during the project's execution phase.
- Nessus is employed for vulnerability assessment, scanning systems and networks for potential weak points that could be exploited by attackers.
- FireEye provides advanced threat detection capabilities, including real-time monitoring and malware protection, crucial for protecting critical project data.
- Cylance utilizes artificial intelligence for predictive threat protection, aiding in preemptively identifying and addressing potential risks.

Each software solution is chosen for its unique capabilities, aligning with specific phases of the project life cycle where risks are most prevalent. For example, Splunk's SIEM capabilities are

leveraged during the monitoring phase to track activity continuously, while Nessus is particularly useful in the planning phase for identifying initial vulnerabilities.

DATA ANALYSIS

The collected qualitative data from interviews and case studies are analyzed thematically to extract recurring themes and patterns related to cybersecurity risks and risk management strategies. These themes are then categorized according to the project life cycle phases, helping to map specific types of cyber threats to each phase. Quantitative data derived from software tools and case study analyses are statistically evaluated to determine the frequency and impact of different cyber threats.

For the quantitative analysis, descriptive statistics such as frequency counts and percentages are used to categorize the types of cybersecurity incidents by project phase and industry. To understand the correlation between specific project phases and types of cyber threats, correlation analysis is performed. Additionally, regression analysis helps determine the factors most predictive of cybersecurity incidents, considering variables like project duration, team size, and industry sector.

Model Development

Based on the findings from the data analysis, a structured model for cybersecurity risk management across project life cycles is developed. The model incorporates risk identification, assessment, mitigation, and monitoring strategies tailored to each project phase. Key metrics from the statistical analysis, such as the most frequent types of threats in each phase and the effectiveness of various mitigation strategies, guide the model's framework. By integrating insights from both qualitative themes and quantitative results, the model provides a

comprehensive, data-driven approach to managing cybersecurity risks in projects.

Validation of the Model

The proposed model is validated by applying it to selected projects from various industries to assess its practical effectiveness. Simulated cybersecurity scenarios are also used to test the model's responsiveness to real-time cyber threats. Feedback from cybersecurity professionals and project managers is collected post-implementation to identify any areas for improvement, ensuring the model's applicability and effectiveness in real-world project environments.

ETHICAL CONSIDERATIONS

Given the sensitive nature of cybersecurity data, strict ethical standards are upheld throughout the research process. Confidentiality agreements are in place to protect organizational and individual data. All participants are informed about the research purpose, ensuring transparency and voluntary participation in interviews.

This methodology provides a thorough framework for examining cybersecurity risks in project life cycles and developing an integrated, adaptable model for effective risk management. The use of cybersecurity software, combined with both qualitative and quantitative analyses, ensures a comprehensive, practical, and actionable approach to addressing cybersecurity risks in project management.

Cybersecurity Risks Across Project Life Cycle Phases

Based on thematic analysis, various cyber threats were identified across project life cycle phases. The statistical analysis also quantified the frequency and impact of these risks, leading to a comprehensive understanding of their distribution and criticality across different stages.

RESULTS

Table 1: Frequency and Type of Cyber Threats Across Project Phases

Project Phase	Threat Type	Frequency (%)	Severity (1-5)
Initiation	Phishing	15%	3
Planning	Data Breach	25%	5
Execution	Malware Infiltration	30%	4
Monitoring & Control	Unauthorized Access	20%	4
Closure	Inadequate Data Disposal	10%	3

Table 1 shows that execution and planning phases exhibited the highest occurrences of cyber threats, with malware and data breaches being predominant. This aligns with observed practices

where these stages involve extensive data handling and system utilization, making them highly vulnerable.

Correlation Analysis of Project Phases and Cyber Threats

To examine potential relationships between project phases and cyber threats, correlation analysis was

conducted. Results indicated a moderate correlation between data breaches and the planning phase, while execution strongly correlated with malware risks.

Table 2: Correlation Coefficients Between Project Phases and Cyber Threat Types

Cyber Threat	Initiation	Planning	Execution	Monitoring & Control	Closure
Phishing	0.48	0.22	0.34	0.29	0.12
Data Breach	0.31	0.67	0.42	0.33	0.18
Malware Infiltration	0.23	0.41	0.79	0.26	0.09
Unauthorized Access	0.18	0.35	0.26	0.65	0.20
Inadequate Data Disposal	0.07	0.15	0.12	0.24	0.73

In Table 2, the correlation coefficients demonstrate significant correlations between data breaches in the planning phase (0.67) and malware in the execution phase (0.79). This suggests a need for heightened security protocols in these phases.

Regression Analysis on Predictive Factors of Cybersecurity Incidents

A regression analysis was conducted to determine predictors of cyber incidents, focusing on variables like project duration, team size, and industry type.

Table 3: Regression Analysis Results

Predictor Variable	Coefficient	Standard Error	p-value
Project Duration	0.31	0.08	0.002
Team Size	0.18	0.07	0.015
Industry Type	0.42	0.09	0.001

Table 3 shows that industry type (coefficient = 0.42) is the most significant predictor of cybersecurity incidents, with p-values under 0.05 across all predictors. This highlights that specific industries may require tailored cybersecurity approaches based on inherent risk levels.

Cybersecurity Software Performance in Mitigating Threats

The selected cybersecurity software was assessed based on threat detection, response time, and user-friendliness. Performance data from software trials indicate each tool's effectiveness.

Table 4: Cybersecurity Software Performance Metrics

Software	Threat Detection (%)	Response Time (ms)	User-Friendliness (1-5)
Splunk	92%	500	4.2
Wireshark	85%	650	3.8
Nessus	88%	700	4.0
FireEye	95%	450	4.5
Cylance	90%	480	4.1

Table 4 indicates that FireEye achieved the highest detection rate (95%) and the fastest response time (450 ms), making it the most effective software tool in the study. Splunk and Cylance also performed well, with strong detection rates and reasonable response times, positioning them as valuable tools for threat monitoring.

Effectiveness of Cybersecurity Measures Across Project Phases

To assess the cybersecurity strategies implemented during each project phase, the effectiveness of various measures (e.g., encryption, multi-factor authentication) was analyzed.

Table 5: Effectiveness of Cybersecurity Measures

Cybersecurity Measure	Initiation	Planning	Execution	Monitoring & Control	Closure
Encryption	High	Very High	High	Moderate	Low
Multi-Factor Authentication	Moderate	High	Very High	Very High	Moderate
Real-Time Monitoring	Low	Moderate	High	Very High	Low
Data Disposal Protocols	Low	Moderate	Moderate	High	Very High

Table 5 reveals that encryption is particularly effective in the planning phase, while real-time monitoring is crucial during execution and monitoring phases. Data disposal protocols show the highest effectiveness in the closure phase, indicating a need for strong closure protocols.

Table 6: Expert Feedback on Model Effectiveness

Metric	Mean Rating (1-5)	Standard Deviation
Risk Reduction	4.6	0.4
Ease of Implementation	4.2	0.5
Adaptability Across Sectors	4.5	0.3

As shown in Table 6, the model received high ratings across all metrics, with an average rating of 4.5 in adaptability. This positive feedback suggests that the integrated model effectively addresses cybersecurity risks and can be implemented across various project types.

DISCUSSION

Interpretation of Cybersecurity Risks Across Project Phases

The analysis revealed a distinct distribution of cybersecurity risks across project life cycle phases. The planning and execution phases demonstrated the highest risk levels, primarily due to the extensive data sharing, access points, and system interactions inherent in these stages (AL-Hawamleh, 2024). The high occurrence of data breaches in the planning phase and malware infiltration in the execution phase suggests that these stages require specialized security measures. This finding emphasizes the need for robust cybersecurity protocols during project initiation and execution, including proactive threat assessments and encryption strategies (Aji, *et al.*, 2013). The thematic identification of phishing and inadequate data disposal in initiation and closure phases, respectively, also indicates that these stages often suffer from oversight in cybersecurity planning (Braa, *et al.*, 2007).

Correlation Between Project Phases and Cyber Threats

The correlation analysis showed moderate to high correlations between certain project phases and specific cyber threats, notably data breaches in the planning phase and malware during execution. These correlations support the notion that each phase of a project has distinct vulnerabilities (Zhang, *et al.*, 2013). For instance, the high correlation between data breaches and the planning phase could be attributed to the frequent sharing of sensitive information with stakeholders, which increases exposure to external threats. Similarly,

Model Validation and Feedback

After implementing the integrated cybersecurity model, feedback from experts was collected to validate its effectiveness. Experts rated the model's impact on risk reduction, ease of implementation, and adaptability.

malware infiltration's strong link to the execution phase highlights the susceptibility of active systems to cyberattacks, pointing to a critical need for real-time monitoring during project implementation (Monostori, *et al.*, 2016). These insights underscore the value of tailoring cybersecurity measures to each project phase.

Predictive Factors of Cybersecurity Incidents

The regression analysis identified industry type as the most significant predictor of cybersecurity incidents, followed by project duration and team size. This result aligns with the notion that certain industries—such as finance, healthcare, and technology—are inherently more exposed to cyber threats due to high-value data and extensive use of digital systems (Colledani, *et al.*, 2014; Fagnant & Kockelman, 2015). Additionally, longer project durations and larger team sizes appear to increase the risk, potentially due to a higher number of access points and longer exposure periods to threats. These findings suggest that project managers in high-risk industries should adopt more stringent cybersecurity measures, particularly for projects with larger teams or extended timelines (Kambatla, *et al.*, 2014).

Effectiveness of Cybersecurity Software

The performance analysis of cybersecurity software tools revealed that FireEye, Splunk, and Cylance were the most effective in detecting and responding to threats. FireEye, with the highest detection rate (95%) and quickest response time (450 ms), demonstrated superior performance in handling real-time threats. Splunk and Cylance also showed strong results, particularly in real-time monitoring, which is essential during the execution and monitoring phases. These findings suggest that utilizing a combination of these tools can provide comprehensive protection across the project life cycle (Masood & Java, 2015). Additionally, the software's user-friendliness ratings indicate that accessible, easy-to-use interfaces contribute to

more effective risk management, as team members can quickly implement protective actions when needed (Xia, *et al.*, 2023).

Effectiveness of Cybersecurity Measures Across Project Phases

The assessment of cybersecurity measures across phases showed that encryption was highly effective in the planning phase, while real-time monitoring and multi-factor authentication were crucial during execution and monitoring. These results support the need for targeted security practices at each project stage (Munawar, *et al.*, 2020). The high effectiveness of data disposal protocols in the closure phase suggests that secure data archiving or disposal is essential for preventing residual vulnerabilities after project completion. This targeted approach provides valuable guidance for project managers, indicating that specific cybersecurity practices yield better protection when aligned with phase-specific risks (Zeydan, *et al.*, 2024).

Model Validation and Practical Implications

Feedback from cybersecurity experts indicated that the proposed model was effective in reducing risks, adaptable across sectors, and relatively easy to implement. The high ratings for adaptability and ease of implementation suggest that this model has practical applications across diverse industries (Akinsola & Akinde, 2024). By integrating security protocols that address specific phase vulnerabilities, this model enables project managers to proactively mitigate threats, enhance data protection, and maintain continuity throughout project life cycles (Pelluru, 2021). Additionally, the model's effectiveness in risk reduction highlights its potential to serve as a best-practice framework for organizations seeking to improve cybersecurity in project management (Kitchin, R. & Dodge, 2020; Khan, *et al.*, 2022).

LIMITATIONS AND FUTURE RESEARCH

Although the study provides valuable insights, it has limitations. The sample size for interviews and case studies was relatively small, potentially limiting the generalizability of findings. Additionally, the model's effectiveness was validated through limited case applications; broader testing across various project types and industries would provide more comprehensive validation. Future research should focus on expanding the sample size, testing the model in diverse sectors, and incorporating evolving cybersecurity threats to enhance the model's

robustness. Further development could also involve integrating artificial intelligence-driven tools, which may provide even more predictive and adaptive cybersecurity capabilities.

This study underscores the critical importance of incorporating phase-specific cybersecurity measures across project life cycles. The findings reinforce that each project phase presents unique vulnerabilities, which can be effectively managed through targeted strategies and the use of advanced cybersecurity tools. The proposed integrated model offers a comprehensive approach for identifying, assessing, and mitigating cyber threats, equipping project managers with a proactive framework to safeguard data integrity and project continuity.

CONCLUSION

This study highlights the pressing need for cybersecurity risk management throughout the project life cycle, emphasizing a proactive, integrated approach to mitigate cyber threats effectively. By analyzing cybersecurity risks specific to each project phase and evaluating the performance of leading cybersecurity software, the study demonstrates that tailored security measures can significantly enhance a project's resilience. The proposed model, validated through expert feedback, offers a practical framework for identifying, assessing, and managing cybersecurity risks from initiation to closure, addressing each phase's unique vulnerabilities. The findings indicate that a phase-focused, data-driven strategy, supported by robust software tools, not only strengthens risk management but also promotes operational continuity and data integrity. Moving forward, the model provides a versatile foundation for organizations across industries, supporting adaptive cybersecurity practices in the face of evolving threats and setting a benchmark for cybersecurity standards in project management.

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