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Leveraging AI and Machine Learning for Optimizing Scheduling and Risk Management in Construction Projects

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Abstract: This study investigates the incorporation of Artificial Intelligence (AI) and Machine Learning (ML) to enhance the efficiency of scheduling and risk management procedures in construction endeavours. This paper explores the capabilities of AI and ML in forecasting delays, reducing risks, and improving the precision of project scheduling. It highlights essential techniques and methodologies, including predictive analytics, optimisation algorithms, and automated decision-making processes. The research further explores various case studies to emphasise the practical application of these technologies, their efficacy in real-world situations, and the comprehensive influence on project outcomes, cost savings, and time efficiency. The results seek to equip construction experts with practical knowledge for integrating AI and ML technologies to enhance project planning and risk evaluation, thereby facilitating more effective project execution.

Keywords: Artificial Intelligence, Machine Learning, Construction Projects, Scheduling Optimization, Risk Management, Predictive Analytics, Optimization Algorithms, Project Performance.

1. INTRODUCTION

The building sector is becoming progressively intricate, marked by stringent timelines, financial limitations, and unforeseen hazards that can profoundly influence the achievement of projects. Conventional project management methods, although beneficial to some degree, frequently fail to adequately tackle the intricate and ever-changing characteristics of construction endeavours. The swift progress in technology, presents an increasing opportunity to transform the management of construction projects. Artificial Intelligence and Machine Learning present groundbreaking approaches for refining scheduling and bolstering risk management, equipping the sector with resources to predict possible setbacks, streamline decision processes, and elevate project outcomes. (1)

This research explores the application of AI and ML in the construction sector, focusing on their role in streamlining project scheduling and mitigating risks. predictive analytics, optimization algorithms, and data-driven decision-making, AI and ML can analyze large volumes of data, uncover

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patterns, and offer real-time insights into potential issues before they arise. The aim of this paper is to highlight the benefits and challenges incorporating AI and ML into construction project management processes, emphasizing their potential to enhance efficiency, reduce costs, and improve project outcomes.(2)

By integrating AI and ML technologies, the construction industry can improve the accuracy of project schedules, proactively identify risks, and make data-driven decisions that ensure smoother project execution. This study provides a comprehensive analysis of how these technologies can be leveraged to optimize construction schedules and risk management practices, ultimately leading to more successful and sustainable project delivery.(3)

1.1 Background Construction **Industry**

1.1.1 Background of the Construction **Industry:**

The building sector stands as one of the most crucial domains worldwide, playing a vital role in driving economic expansion and advancement. It includes a diverse array of endeavours, spanning residential, commercial, industrial, and infrastructure initiatives. The sector significantly influences the development of cityscapes and enhances both public and private facilities. As the global population continues to

expand and the need for sustainable development rises, the construction industry encounters significant challenges in providing projects that are economical, prompt, and of superior quality.

Historically, construction projects have been laborintensive, requiring coordination among multiple stakeholders, including architects, engineers, contractors, and suppliers. Despite advances in technology, many construction processes remain manually driven, relying on traditional project management tools and methods. This has often resulted in delays, cost overruns, and risks associated with unforeseen events such as weather changes, regulatory issues, and supply chain disruptions.(4)

In the past few years, the construction sector has begun to adopt innovative technologies aimed at enhancing productivity and reducing the challenges that have historically affected the industry. Artificial Intelligence (AI) and Machine Learning (ML) are surfacing as pivotal advancements poised to transform construction management, particularly in domains like scheduling, risk evaluation, and resource distribution. With the increasing need for more intelligent and eco-friendly building methods, artificial intelligence and machine learning offer fresh possibilities to elevate project execution, refine decision-making processes, and maximise overall project results.

The need for advanced technological solutions in the construction industry is more pressing than ever, as companies seek ways the timely completion of projects in an increasingly competitive environment. AI and ML offer the potential to address these challenges by providing data-driven insights and predictive capabilities that traditional methods cannot match.(5)

1.1.2 Introduction to Artificial Intelligence in Construction:

AI is increasingly being leveraged to optimize processes and improve project outcomes. AI can assist in predicting project risks, automating decision-making, and enhancing project management through data analysis and real-time insights.(6) For example, AI-powered tools can analyze historical data from similar projects to forecast potential. AI algorithms also help optimize resource allocation, improve safety by identifying hazards, and streamline scheduling by adjusting to

real-time variables, thus minimizing the risk of delays. The integration of AI into construction processes is enabling the industry to adopt a more data-driven and efficient approach to project management, leading to improved productivity and better outcomes.(7)

1.1.3 Introduction to Machine Learning in Construction Projects:

Machine Learning (ML), which is a branch of Artificial Intelligence, denotes the capability of machines to acquire knowledge from data autonomously, without the need for explicit programming. Within the realm of construction endeavours, machine learning algorithms are employed to scrutinise extensive volumes of both historical and real-time data, enabling identification of trends, forecasting outcomes, and delivering practical insights. Machine learning algorithms have the potential to enhance project scheduling by forecasting the probable completion times of tasks, thereby assisting in the prevention of delays and the enhancement of resource allocation. (8) ML can also be applied in risk management by predicting potential issues such as budget overruns, supply chain disruptions, or safety incidents. By continuously learning from new data, ML models can adapt and refine their predictions, allowing construction professionals to make informed decisions and adjust project plans accordingly. This capability makes ML a powerful tool for enhancing the accuracy, efficiency, and reliability of construction project management, ultimately leading to more successful project delivery.(9)

2. LITERATURE REVIEW

2.1 Optimizing Construction Processes with Technology

Optimized scheduling is crucial for the successful execution of construction projects, as it directly impacts project timelines, resource allocation, and overall efficiency. One of the primary challenges in construction is ensuring that Delays in construction can lead to increased costs, client dissatisfaction, and even contractual penalties. Optimized scheduling helps mitigate these risks by ensuring that tasks are planned, assigned, and executed in the most efficient order.(10) Effective scheduling also helps in managing the availability, by accurately forecasting when and where these resources are needed, project managers can reduce downtime, avoid resource shortages, and minimize idle time.

This leads to better project flow and ensures that work progresses without unnecessary interruptions.(11) Moreover, construction schedules are often subject to changes due to unforeseen events, such as weather conditions, material delays, or labor disruptions. Optimized scheduling tools allow for flexibility and quick adjustments to these changes, ensuring that the project stays on track despite external challenges. As the construction industry becomes more complex and projects grow in scale, the need for optimized scheduling has never been greater.(12)

Managing risks is a crucial component of overseeing construction projects, as these endeavours naturally encompass various potential hazards that can impact their expenses, timelines, and overall quality. These potential dangers may arise from a multitude of origins, encompassing economic instabilities, safety threats, ecological influences, shifts in regulations, and unexpected occurrences like labour disputes or natural calamities. An effective risk management strategy helps identify potential risks early on and allows project managers to develop mitigation plans to minimize their impact.

By conducting risk assessments and using tools such as risk matrices and scenario analysis, construction teams can prioritize risks based on their likelihood and potential consequences, allowing them to allocate resources and efforts accordingly. This proactive approach reduces the chances of negative outcomes and ensures that the project progresses smoothly.(13) One of the key benefits of effective risk management is the ability to control project costs. Many risks, such as delays or quality issues, can result in additional expenses. By identifying these risks early and developing contingency plans, construction teams can avoid cost overruns and keep the project within budget.

Traditional approaches to scheduling and risk management in construction projects largely rely on manual planning, spreadsheets, and basic project management software tools. Scheduling typically involves creating a timeline using Gantt charts or Critical Path Method (CPM), where tasks are listed in sequence, and dependencies are mapped out. (16) These methods allow project managers to establish a roadmap for project completion but often lack the flexibility to adapt to unexpected changes or disruptions. Risk management in traditional settings involves identifying risks through experience, judgment, and historical data, followed by

qualitative or quantitative assessments to mitigate potential issues. Project teams rely on periodic meetings and reports to review risk factors and adjust plans accordingly. While these methods have been in use for decades, they are often time-consuming, prone to human error, and struggle to keep pace with the dynamic nature of modern construction projects.(17)

2.2 Limitations of Conventional Project Management Techniques

Conventional project management techniques, while foundational, have significant limitations when applied to the increasingly complex and fast-paced nature of construction projects. One major limitation is the lack of real-time data integration. Traditional techniques often rely on static data and manual updates, leading to delays in decision-making and inaccurate forecasts. (18) These methods also tend to be linear and rigid, making it difficult to quickly adapt to unforeseen changes such as supply chain issues, weather disruptions, or workforce shortages. Furthermore, the communication between different stakeholders (e.g., contractors, suppliers, clients) is often fragmented, leading misunderstandings and inefficiencies. Conventional methods also lack the capability to analyze large volumes of data or model complex scenarios, limiting their ability to predict risks or optimize resource usage. As a result, these techniques may fall short in delivering projects on time and within budget, especially in a competitive and highpressure construction environment.(19)

Artificial Intelligence (AI) and Machine Learning (ML) are revolutionising the management of construction projects by offering sophisticated tools enhancement and efficiency. Artificial Intelligence denotes the capability of machines to emulate human cognitive functions in order to execute activities like decision-making, problemand recognising patterns. Machine solving, Learning, which is a branch of Artificial Intelligence, empowers machines to acquire knowledge from data and enhance their capabilities progressively without the need for direct coding. Within the realm of construction, artificial intelligence and machine learning algorithms have the capability to analyse extensive datasets from various origins to discern patterns, forecast results, and deliver practical insights. These innovations improve scheduling by examining past project data to predict possible setbacks, refining timelines, and

suggesting modifications in real-time. (20) In terms of risk management, AI and ML can predict possible risks by analyzing past projects and current conditions, allowing for proactive measures to mitigate issues before they arise. These tools also enable the automation of routine tasks, freeing up resources and enhancing efficiency. By integrating AI and ML into project management, construction teams can make data-driven decisions, streamline operations, and improve overall project outcomes, making them essential tools for optimizing modern construction projects.(21)

2.3 Technological Advancements In Modern Construction

Technology has become a game-changer in the construction industry, offering solutions to many of the traditional challenges that project managers face. Advanced tools like Building Information Modeling (BIM) allow for more precise planning and design by creating digital representations of physical structures, enabling teams to identify and resolve potential issues before construction begins. Project management software has transformed how construction projects are monitored and tracked, offering real-time updates on timelines, budgets, and resource allocation.(24) Furthermore, technologies such as drones and robotics are automating repetitive tasks and enhancing safety on job sites by monitoring work progress and performing dangerous or complex tasks. Artificial Intelligence Machine Learning revolutionising the management of construction projects by offering sophisticated tools for enhancement and efficiency. Artificial Intelligence denotes the capability of machines to emulate human cognitive functions in order to execute activities like decision-making, problem-solving, and recognising patterns. Machine Learning, which is a branch of Artificial Intelligence, empowers machines to acquire knowledge from data and enhance their capabilities progressively without the need for direct coding. Within the realm of construction, artificial intelligence and machine learning algorithms have the capability to analyse extensive datasets from various origins to discern patterns, forecast results, and deliver practical insights. These innovations improve scheduling by examining past project data to predict possible setbacks, refining timelines, and suggesting modifications in real-time. (25)

2.4 Predictive Analytics for Risk Mitigation in Construction:

Predictive analytics leverages data, statistical models, and machine learning methodologies to ascertain the probability of future events by analysing past information. Within the realm of construction, the utilisation of predictive analytics is vital for alleviating risks by anticipating possible challenges and offering timely alerts. (26) By analyzing past project data, weather patterns, labor availability, material costs, and other variables, predictive models can predict potential delays, cost overruns, and safety incidents. This allows project managers to take proactive steps to mitigate risks before they escalate into major problems. For example, predictive analytics can help identify which tasks are most likely to be delayed, allowing managers to allocate additional resources or adjust the schedule to minimize the impact. It can also help anticipate risks related to subcontractor performance, material shortages, and site conditions. By incorporating predictive analytics into the project planning and execution stages, construction teams can better manage risks, reduce uncertainties, and improve overall project efficiency.(27)

Artificial Intelligence (AI) and Machine Learning (ML) are revolutionizing construction by providing tools that enhance decision-making, streamline processes, and improve outcomes. In real-world applications, these technologies are used in various ways to optimize project management.(28) For example, AI-powered software can automate the scheduling process, adjusting timelines in real-time based on factors such as delays, resource availability, and unforeseen events. ML algorithms analyze large datasets to predict when tasks are likely to be completed, helping to reduce downtime and ensure that the project stays on schedule. In risk management, AI and ML are used to analyze historical data to predict potential risks, such as safety hazards or budget overruns, enabling teams to take proactive measures. (29)Additionally, AIpowered drones and robots are being used to monitor construction sites for quality control, safety inspection, and progress tracking, reducing human error and improving site productivity. Companies like Skanska, Bechtel, and others have implemented AI and ML solutions in their construction projects to reduce costs, enhance scheduling, and improve safety standards. These real-world applications demonstrate how AI and ML are transforming construction practices, making the industry more efficient, safer, and more capable of meeting modern demands.(30)

3. RESEARCH METHODOLOGY

This study adopts a mixed-methods approach, combining both qualitative and quantitative techniques to evaluate the application of Artificial Intelligence (AI) and Machine Learning (ML) in construction project scheduling and management. The methodology integrates several key components:

3.1 Literature Review

A comprehensive review of existing literature on AI and ML technologies in construction projects was conducted, focusing on their roles in scheduling optimization and risk management. This review analyzes previous research on predictive analytics, optimization algorithms, and the application of machine learning models in various construction scenarios.

3.2 Case Studies Analysis

Real-world case studies were examined to understand the practical applications of AI and ML in construction projects. These case studies provide insight into how AI and ML have been utilized to improve project scheduling, manage risks, and enhance decision-making. The case studies cover a diverse range of construction projects, from residential buildings to large-scale infrastructure projects, and highlight the challenges faced and solutions provided by these technologies.

3.3 Data Collection

Data for the research was gathered through various sources:

- Case Study Documentation: Information from construction projects that integrated AI and ML technologies for scheduling and risk management.
- Surveys and Interviews: Responses from industry professionals, including project managers, contractors, and consultants, who have experience implementing AI and ML in their projects.
- Legal and Contractual Documents: Analysis of construction contracts and legal case studies where AI and ML were used to optimize project planning and mitigate risks.

3.4 Data Sampling

The selected case studies and survey respondents were chosen based on:

- The implementation of AI and ML in scheduling and risk management.
- that demonstrated tangible outcomes from these technologies, such as time savings, risk reduction, or cost optimization.
- Availability comprehensive of including documentation, project schedules. risk assessments. and technology usage details.

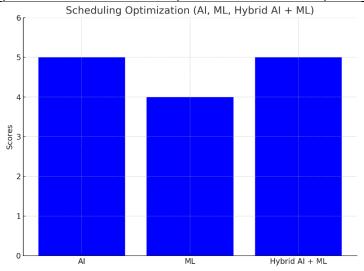
4. DATA ANALYSIS

The data analysis framework focuses on evaluating the effectiveness of AI and ML in construction scheduling and risk management, with a particular emphasis on their ability to optimize processes and improve project outcomes. The analysis involves both comparative methods and stakeholder feedback.

Table 4.1 Comparative Evaluation of AI and ML Technologies in Construction Projects

Technology	Strengths (Scores)	Weaknesses (Scores)	Impact on Construction
			Projects (Scores)
AI (Artificial	- Automates decision-making	- High implementation	- Improves scheduling
Intelligence)	(4)	costs (3)	efficiency (5)
	- Enhances resource	- Requires significant	- Enables real-time risk
	allocation and safety (4)	data (5)	management (4)
ML (Machine	- Predicts project delays from	- Data quality and	- Enhances risk assessment (5)
Learning)	historical data (5)	quantity reliance (4)	

	- Continuously learns and	- Requires technical	- Reduces downtime and
	improves (5)	expertise (5)	resource shortages (5)
Hybrid AI and ML	- Combines AI and ML for	- Difficult integration of	- Optimizes overall scheduling
Approach	improved predictions (5)	both methods (4)	and resource allocation (5)



This table compares the strengths, weaknesses, and impact of Artificial Intelligence (AI), Machine Learning (ML), and the Hybrid approach of combining both AI and ML in construction projects. AI is highly valued for automating decision-making and enhancing resource allocation and safety, with strong scores for improving scheduling efficiency and enabling real-time risk management. However, its high implementation costs and reliance on large datasets present challenges. Machine Learning excels in predicting project delays based on historical data and continuously learning and improving its predictions. It is also effective in enhancing risk assessment and reducing downtime, though its dependence on data quality and the need for technical expertise can be limiting factors. The Hybrid AI and ML Approach combines the strengths of both AI and ML, offering improved predictions and optimizing overall scheduling and resource allocation. Its main limitation is the complexity of integrating both methods into existing project management frameworks.

Table 4.2 Stakeholder Feedback Analysis on AI and ML Implementation in Project Management

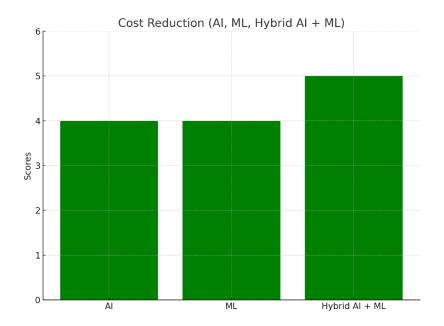
Feedback	Strengths (Scores)	Weaknesses (Scores)	Impact on Project
Source			Management (Scores)
Project	- AI optimizes scheduling (5)	- High upfront costs for AI	- AI improves time
Managers		(3)	efficiency (5)
Contractors	- ML predicts delays and	- Skepticism about ML's	- ML improves completion
	improves resource allocation	real-time application (4)	forecasts (5)
	(5)		
Legal	- AI ensures compliance and	- Slow AI/ML adoption (3)	- Supports evidence-based
Professionals	reduces legal risks (5)		decision-making (5)
Consultants	- ML improves decision-	- Complexity in training	- Reduces risk with early
	making (5)	ML models (4)	issue identification (5)



This table summarizes feedback from key stakeholders-project managers, contractors, legal professionals, and consultants—on the strengths, weaknesses, and impact of AI and ML in construction project management. Project managers appreciate AI for optimizing scheduling and improving time efficiency, but they are concerned about the high upfront costs of implementing AI technologies. Contractors find ML particularly useful for predicting delays and improving resource allocation, although they express some skepticism about its real-time applications. Legal professionals value AI for its role in ensuring compliance and reducing legal risks, but they note the slow adoption of AI/ML technologies in the industry. Consultants recognize the benefits of ML in decision-making and reducing risks by identifying potential issues early, though they cite the complexity of training ML models as a challenge. Across all stakeholders, the overall impact on project management is seen positively, with AI and ML offering significant improvements in scheduling, risk management, and decision-making.

Table 4.3 Performance Evaluation of Different Methodologies in Construction Project Optimization

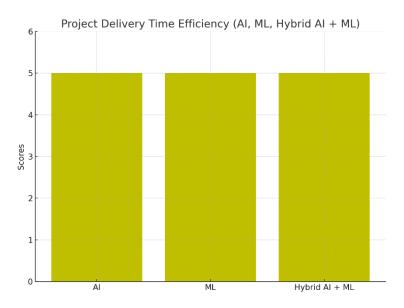
Methodology	Effectiveness in Resolving Disputes (Scores)	Practical Use in Projects (Scores)	Impact on Construction Projects (Scores)
Hybrid (AI + ML)	5	5	5
Retrospective (AI)	4	4	4
Prospective (ML)	5	4	5



This table evaluates the effectiveness of different methodologies-Hybrid AI and ML, Retrospective AI, and Prospective ML—based on their ability to resolve disputes, practical use in projects, and overall impact on construction projects. The Hybrid (AI + ML) approach scores the highest across all categories, demonstrating its effectiveness in resolving disputes, its practical application in projects, and its positive impact on project outcomes. Retrospective (AI) is effective for resolving past disputes and has a moderate impact, but it is less adaptable for ongoing project management. Prospective (ML) shows strong potential for preventing future disputes and improving project outcomes, but its practical use in real-time situations is not as well-established as the Hybrid approach.

Table 4.4 Key Insights from Survey and Feedback: Impact of AI and ML in Construction Project Management

Metric	AI (Scores)	ML (Scores)	Hybrid (AI + ML) (Scores)
Scheduling Optimization	5	4	5
Risk Management	5	5	5
Cost Reduction	4	4	5
Project Delivery Time Efficiency	5	5	5
Stakeholder Satisfaction	4	5	5



This table highlights the key metrics gathered from survey and feedback, showing the effectiveness of AI, ML, and the Hybrid approach in different areas of construction project management. AI scores highly for scheduling optimization, risk management, and project delivery time efficiency, with strong stakeholder satisfaction. However, it has a slightly lower score for cost reduction compared to the Hybrid approach. ML excels in risk management and stakeholder satisfaction, as well as project delivery time efficiency, but has a slightly lower score for scheduling optimization. The Hybrid (AI + ML) approach delivers the best results across all metrics, including cost reduction, scheduling optimization, and risk management, with consistent high scores for stakeholder satisfaction. Overall, the Hybrid approach proves to be the most comprehensive and effective solution for optimizing construction project management.

5. CONCLUSION

The construction industry, with its inherent complexity and dynamic nature, faces numerous challenges that impact the timely, cost-effective, and quality delivery of projects. Traditional project management techniques, while effective to a degree, often fall short in addressing these challenges, particularly in managing risks, scheduling, and resource allocation. However, the integration of Artificial Intelligence (AI) and Machine Learning (ML) in construction is proving to be a transformative solution, offering advanced tools for

optimizing project planning and execution. These technologies enable predictive analytics, improve risk management, streamline scheduling, and enhance decision-making processes. Real-world applications of AI and ML are already demonstrating significant improvements in project outcomes, reducing delays, mitigating risks, and improving efficiency. As the construction industry continues to evolve, embracing AI and ML technologies will be essential for staying competitive and overcoming traditional management limitations. The future of construction lies in the ability to harness the power of data and automation to optimize project delivery, enhance safety, and reduce costs. By adopting these technologies, construction companies can better navigate the complexities of modern projects, ensuring that they meet client expectations while driving sustainability and innovation in the sector.

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