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Final Year Project

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# Glossary

API – Application Programming Interface

AR – Augmented Reality

BDA – Big Data Analytics

BIM – Building Information Modelling

CAD – Computer-Aided Design

CM – Cloud Manufacturing

CPS – Cyber-Physical Systems

DIAMND – Diagnostics and Monitoring

ERP – Enterprise Resource Planning

IoT – Internet of Things

OLE - Object Linking and Embedding

OPC-UA – OLE for Process Control Unified Architecture

PLC – Programmable Logic Controller

PLM – Product Lifecycle Management

SQL – Structured Query Language

VR – Virtual Reality

# Abstract

# Introduction



## Background and Significance of the Study

## Research Objectives

# Literature Review



## Digital Twins



### Origin of Digital Twins.

The concept of the "Digital Twin," which has emerged as a pivotal framework in the realm of engineering and industrial applications, finds its origins in the early 2000s. Dr. Michael Grieves, a scholar at the University of Michigan, is credited with applying and pioneering the foundational ideas behind it (Sjarov et al., 2020). Initially referred to as the "Mirrored Spaces Model," later renamed by NASA’s John Vickers as “digital twin”, the Digital Twin comprises three fundamental components that collectively constitute its essence. These components, seen in Figure 1, consist of the "Real Space," representing the tangible, physical counterpart; the "Virtual Space," serving as the digital replica or simulation of the real-world entity; and the intricate web of connections that interlinks data and information, bridging the gap between the virtual and real products (D’Amico et al., 2019). This innovative framework has since evolved into a versatile and indispensable tool, offering profound insights into various domains, including crane fleet monitoring, where it enables the creation of highly accurate virtual representations of physical assets and facilitates the real-time tracking and analysis of their performance. The developmental trajectory of Digital Twins unfolds across three discernible phases. In its initial instantiation, the digital model lacks the mechanism for automated data exchange between physical and digital entities. Progressing to the second stage, identified as the digital shadow, a paradigm shift is observed with the introduction of automated unidirectional data flow from physical to digital objects. The third and most advanced stage, epitomized by the digital twin, witnesses the establishment of a bidirectional data flow facilitating seamless integration between physical and digital entities (Wang et al., 2020).

A diagram of a space shuttle

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*Figure 1 – Components of a Digital Twin (D’Amico et al., 2019).*

### How do Digital Twins Work?

1. On the physical side, we now collect more
2. and more information about the
3. characteristics of the physical product. We
4. can collect all types of physical
5. measurements from automated quality
6. control stations, such as Coordinate
7. Measuring Machines (CMMs). We can
8. collect the data from the machines that
9. perform operations on the physical part to
10. understand exactly what operations, at
11. what speeds and forces, were applied. For
12. For example, we can collect the torque
13. readings of every bolt that attaches a fuel
14. pump to an engine to ensure that
15. each engine/fuel pump attachment is
16. successfully performed.

Real-world machines are equipped with an assortment of sensors that record critical performance data. These sensors capture information on various aspects of the crane's operations, including parameters such as load capacity, movement, environmental conditions, and more (IBM, n.d.). In the realm of digital twinning for fleet monitoring, the convergence of physical and virtual elements assumes paramount significance. This integration is prominently illustrated through the acquisition of multifaceted physical measurements, derived from the Programmable Logic Controller (PLC) of cranes, which encompass variables such as spatial position and speed of the crane's spreader. These tangible data inputs form the foundation for the construction of a comprehensive digital twin. Furthermore, on the virtual side, the research underpins a substantial augmentation in the depth and breadth of available information. This augmentation is primarily achieved through the incorporation of an extensive array of behavioural characteristics. These attributes, inclusive of various performance parameters, not only facilitate the visual representation of the crane but also empower rigorous testing of its capabilities, ensuring a holistic understanding of its operational dynamics. Although the present investigation emphasises the capacity for virtual testing, it is pertinent to note that for certain applications, the focus may primarily be on generating lightweight virtual models to mirror physical counterparts, with the foremost aim being real-time visualisation of intricate systems, even in cases where comprehensive performance testing may not be feasible or necessary.

### Digital Twin: Use Case Models

The application of Digital Twin technology in crane monitoring and fleet management unveils a realm of profound utility, effectively harnessing the capabilities of conceptualisation, comparison, and collaboration as outlined by Grieves (2014). Conceptualisation, in the context of crane operations, enables a transformative approach to understanding the status and performance of these heavy machinery assets. Unlike conventional data processing, Digital Twins offer the unique advantage of real-time, visual representation, eliminating the need for manual translation of visual information into symbolic data. Through the Digital Twin, operators can simultaneously visualise the physical crane's condition and its virtual counterpart, allowing for a seamless comprehension of crucial data.

Moreover, the concept of comparison becomes an indispensable analytical tool in the context of crane and fleet monitoring. The Digital Twin allows for the immediate evaluation of desired operational outcomes against actual results, eliminating the inefficiencies associated with manual data cross-referencing. By overlaying the ideal characteristics and tolerance corridors, the Digital Twin empowers users to swiftly assess whether the cranes and fleet are performing within acceptable parameters, with deviations colour-coded for instant recognition. These comparisons extend to various measurements, including tensile strength, torque readings, and other critical performance metrics, enhancing real-time decision-making.

Collaboration in crane and fleet management takes on a new dimension with the Digital Twin. Traditionally, operational assessments and troubleshooting were confined to a local context. However, the Digital Twin enables a shared conceptualisation that can be accessed and visualised by teams worldwide, transcending geographical boundaries. This global perspective allows stakeholders from various locations to not only monitor their fleet but also compare their performance with fleets across the globe. In the event of an issue in one fleet, the solution can be promptly identified and shared with other fleets, fostering collaborative innovation on a global scale.

In summary, the application of Digital Twins in crane monitoring and fleet management aligns seamlessly with the conceptualisation, comparison, and collaboration framework proposed by Michael Grieves. This technological advancement not only streamlines crane operations but also empowers global teams to collaborate in real-time, driving innovation and efficiency across the fleet management landscape.

### Choosing Unity 3D for Visualisation of Digital Twins

Unity3D serves as the linchpin in the landscape of digital twin development, offering an array of potent features and capabilities meticulously tuned to cater to the specific demands of digital twin applications. At its core, Unity3D excels in data ingestion and optimisation. This powerful technology seamlessly imports data from diverse formats, including BIM (Building Information Modelling) and CAD (Computer-Aided Design). It integrates data from various systems such as PLM (Product Lifecycle Management), ERP (Enterprise Resource Planning), and IoT (Internet of Things). Unity's data preparation tools are nothing short of impressive, facilitating the import and optimisation of over 70 formats. This results in the creation of a unified, real-time representation of physical assets that forms the bedrock of digital twins.

When it comes to flexible and efficient creation tools for digital twins, Unity3D stands out as a global leader. Renowned as the foremost real-time 3D platform worldwide, Unity is further enhanced by a suite of complementary products that expedite the creation, editing, and real-time iteration of interactive 3D content. This accelerates the development process, enabling rapid deployment of digital twins.

Unity3D also shines in the domain of dynamic visualisation, supporting an extensive range of devices and platforms. With compatibility for over 20 platforms, including HoloLens, Quest, Windows, Mac, iOS, Android, and more, Unity3D emerges as a versatile choice for digital twin applications. It's not just versatility; Unity is a leading platform for crafting content for AR and VR applications, underpinning a substantial portion of head-worn AR experiences.

To streamline digital twin development, Unity3D provides advanced simulation services. These services encompass sensor and robotics emulation, performance-optimised simulation testing, and training, among others. Collectively, these features expedite decision-making processes. Unity3D's hallmark features, including versatility, real-time capabilities, and extensive support for diverse devices and platforms, establish it as an indispensable platform for the visualisation and deployment of digital twins (Unity, n.d.).

The decision to adopt Unity as the foundational platform for the digital twin application is grounded in a solid foundation of reasons. Spatial rendering, especially for spatial-oriented data, presents a complex challenge that has long been the focus of the game industry. This challenge has led to the development of specialised software, often called game engines, which offer comprehensive toolsets and reusable components finely tuned for 3D rendering. In this landscape of options, Unity emerged as the optimal choice for the project, bolstered by familiarity with the platform, rooted in a background as a game development student (Leskovsky et al., 2020).

Unity earns favour for several compelling reasons. It provides extensive support for all essential aspects of the planned development, both directly and indirectly. Unity's user-friendliness ensures ease of learning, and its cost-effective pricing conditions are noteworthy. Moreover, Unity boasts comprehensive documentation and is distinguished for its rapid growth, continuously introducing new functionalities.

By choosing Unity, the potential of this versatile 3D engine is unlocked. It empowers the crafting of three-dimensional objects within a virtual space, offering dynamic manipulation, movement, and rotation. It also allows for the seamless integration of data from IoT devices. In the case of the crane, equipped with a multitude of IoT sensors, Unity's prowess in gathering and processing data from these sensors is invaluable. In the context of digital twin development, reliance on Figure 2, a schematic diagram illustrating the integration of Digital Twins within Unity3D serves as a valuable reference for the project (Gao et al., 2023). These capabilities lay the foundation for the immersive environment that the digital twin requires.

A diagram of a cloud server

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*Figure 2 – A schematic diagram of using Digital Twins in Unity3D (Gao et al., 2023).*

The camera, a pivotal component in 3D applications, plays a central role in shaping the user's viewpoint and impacting application control and display. Our application offers a spectrum of camera view modes, catering to diverse user needs, from PC desktop viewing to immersive VR experiences with headsets like Oculus. Unity's cross-platform compatibility is a standout advantage, allowing us to develop a unified application seamlessly running across platforms, spanning PCs, mobile phones, and the web. Unity further equips us with robust VR and AR tools that intuitively adapt the camera and interface to accommodate users and their equipment, whether involving a joystick, headsets, or other devices.

This combined section emphasises Unity3D's pivotal role in digital twin development and offers a comprehensive perspective on the reasons for choosing Unity as the foundational platform for our digital twin application.

## Case Study



### Importance of Case Studies

In the realm of technological advancements and systems improvement, case studies play a pivotal role in showcasing the significance of innovation. The case of DIAMND (Diagnostics and Monitoring), a Crane Management System, serves as a compelling example of how such studies shed light on the transformation of existing systems. It highlights the importance of critically examining and addressing the challenges posed by legacy technologies, especially when it comes to aesthetics and functionality. The importance of this case study lies in its potential to inspire others to explore new, more efficient solutions and improve the user experience, as well as to create visually appealing interfaces for data management systems.

### DIAMND: An Overview

In this case study, the goal is to address the limitations of the DIAMND system and propose a more effective solution. The existing DIAMND system, used for crane management, presents several challenges, especially in terms of its appearance and functionality. It relies on data collection from various sources, including direct connections to a crane's PLCs through SignalR and OLE for Process Control Unified Architecture (OPC-UA), hourly trace files containing approximately 35,000 signals, and feedback arrays within the PLCs to populate job and load statistics tables in Structured Query Language (SQL).

Throughout this project, active engagement with members of the sales and engineering teams at Liebherr has been crucial in gathering insights and requirements for the improved system. These inputs have played a significant role in shaping the approach. This case study highlights the potential of modern technology and data-driven solutions in not only overcoming the limitations of legacy systems like DIAMND but also in improving the overall user experience and aesthetics of crane management operations.



#### Addressing The Challenges

In this section, the existing DIAMND system is examined, highlighting the imperative need for its transformation. DIAMND serves as the primary approach to crane management, but it presents a series of challenges, particularly in terms of aesthetics and functionality. These challenges stem from its reliance on data acquisition from various sources, including direct connections to a crane's PLC through SignalR and OPC-UA, hourly trace files containing approximately 35,000 signals, and feedback arrays within the PLCs, which are used to populate job and load statistics tables in SQL.

One of the significant challenges is the complexity of data management. The DIAMND system grapples with the intricacies of data acquisition, storage, and presentation. Diverse data sources not only make data management convoluted but also introduce noise and irrelevant information into the system. This noise can obscure critical data, contributing to inefficiencies and suboptimal data aesthetics.

Another issue is the outdated user interface. As highlighted in Figure 3 and Figure 4 below, the user interface of the DIAMND system is visually unappealing and does not align with modern design principles. This not only impacts the user experience but also underscores the pressing need for a modern and visually pleasing solution. It's worth noting that the current interface appears thrown together, lacking proper labels, and missing the company's distinctive touch, including its logo.

*A screenshot of a computer

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*Figure 3 – A view of the main spreader information displayed in DIAMND (Liebherr, 2023).*

A screenshot of a computer

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*Figure 4 – A view of some spreader information displayed in DIAMND (Liebherr, 2023).*

#### Proposed Solutions

To address these formidable challenges, a comprehensive transformation of the DIAMND system is proposed to streamline data management and enhance the user experience. Firstly, the utilisation of an Application Programming Interface (API) is recommended to seamlessly query data from an OPC-UA Server, connected to PLC, for certain variables and send it to an Azure database. This streamlined approach simplifies data acquisition, ensuring that relevant information is obtained swiftly and accurately. Secondly, data will be securely stored in an Azure database, offering enhanced data management capabilities. The Azure platform provides scalability, reliability, and accessibility, facilitating efficient data storage and retrieval. Lastly, to improve data aesthetics and user-friendliness, the implementation of Power BI for data visualisation is proposed. This powerful tool enables the creation of clear and visually appealing data presentations, making it easier for users to derive insights from the information.

The proposed solutions promise to mitigate the challenges faced by the existing DIAMND system, offering a path toward more efficient, user-friendly, and visually appealing crane management, in alignment with contemporary standards and user expectations.

### Lessons from Previous Implementations

Drawing on lessons from previous implementations, the DIAMND case study provides valuable insights for future projects. By examining the challenges and successes of this transformation, lessons can be extracted that extend beyond crane management. The key takeaway is the importance of aligning technology with user expectations and needs. Learning from this case study can guide future implementations, ensuring they are more efficient, user-friendly, and aesthetically pleasing.

## Internet of Things (IoT)



### Fundamentals of IoT

The IoT represents a transformative concept introduced by Kelvin Ashton in 1999, facilitating the connection of physical objects through the Internet to establish a platform for various activities (Gamil et al., 2020). The IoT framework encompasses a network of physical objects embedded with sensors, software, and other technologies, enabling data exchange with other devices and systems over the Internet. The current IoT landscape boasts around 14.76 billion connected devices (Howarth, 2023), with Oracle (n.d.) projecting a surge to over 22 billion by 2025.

The IoT framework is theoretically organised into four distinct layers, seen in Figure 5, that collectively contribute to its functionality. The application layer serves as a hub for various applications and services, ranging from smart cities and homes to transportation, utilities, and healthcare. In this layer, IoT manifests its diverse applications, becoming an integral part of modern living. The perception layer introduces sensory technologies like temperature, vibration, pressure sensors, and RFID sensors, allowing devices to gain awareness of their surroundings. This layer is pivotal in facilitating the acquisition of real-world data by IoT devices. The network layer is the communication backbone, encompassing both software and physical components that enable data transmission between devices and receivers. Its role is fundamental in ensuring seamless connectivity and interaction within the IoT ecosystem. Finally, the physical layer constitutes the basic hardware elements, including physical components, smart appliances, and power supplies, forming the infrastructure that supports the networking of smart objects. Each layer plays a crucial role in shaping the intricate fabric of the IoT (Kumar et al., 2016).

A diagram of a network layer

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*Figure 5 – The four layers that make up the IoT (Kumar et al., 2016).*

### IoT and Industry 4.0

Industry 4.0 represents an important paradigm shift in the manufacturing sector and involves the integration of information and communication technologies into industrial processes. Formed in Germany, Industry 4.0 represents the fourth industrial revolution, after the era of mechanical energy (Industry 1.0), mass production (Industry 2.0), and the digital revolution (Industry 3.0). At its core, Industry 4.0 relies on the fusion of cyber-physical systems (CPS), cloud manufacturing (CM), and the IoT. CPSs comprise machines, storage systems, and production facilities capable of autonomously exchanging information, triggering actions, and monitoring one another. These systems combine virtual and physical elements of production by integrating analogue and digital devices. The Internet of Things is the key technology of Industry 4.0, providing the platform to connect CPSs through a network of sensors, actuators, and devices. CM, a concept born from Industry 4.0, harnesses the capabilities of cloud computing in external data centres to optimise production processes. This harmonious integration of technologies underpins the Industry 4.0 revolution, fostering a new era of smart manufacturing characterised by efficiency, connectivity, and data-driven decision-making (Ben-Daya et al., 2017).

### IoT Integration and Digital Twinning Integration

After elucidating the fundamentals of IoT and its crucial role in Industry 4.0, attention now turns to the transformative amalgamation with Digital Twinning, unveiling its practical application in refining construction processes. The convergence of IoT with Digital Twinning heralds a groundbreaking advancement in technological capabilities. The seamless connectivity facilitated by IoT aligns seamlessly with Digital Twinning's virtual replication of physical entities (Gamil et al., 2020). Within this integrated framework, IoT sensors and devices continually gather real-time data from the physical environment, ensuring a constant update of corresponding digital twins. This dynamic interconnection significantly enhances comprehension of the physical system's behaviour, performance, and potential issues.

In the context of the overarching Digital Twinning in Cranes project, the synergy between IoT and Digital Twinning goes beyond mere connectivity. It enables the creation of virtual replicas of cranes that are continuously updated in real-time. This functionality not only facilitates advanced monitoring but also empowers predictive maintenance capabilities. This harmonious integration becomes a catalyst for optimising crane operations, resulting in reduced downtime and an overall enhancement of efficiency in construction processes.

Moreover, the significance of real-time monitoring is underscored by its application in construction project management. Observations derived from websites and sensor-based information prove pivotal in advancing critical stages of construction projects (BIM Engineering, 2018). These insights, by minimising delays and fostering efficient operational strategies, play an invaluable role. IoT solutions complement this by providing real-time alerts to supervisors concerning resource shortages or operational issues, highlighting real-time monitoring as a top-tier application of IoT. This proactive approach effectively mitigates downtime caused by stockouts or employee performance issues. The amalgamation of IoT-driven real-time observations and Digital Twinning's virtual replication establishes a comprehensive framework for elevating construction project management and operational efficiency.

## Big Data and Visualisation



### An Overview

In the contemporary business landscape, success is intricately tied to the effective utilisation of data. The evolution of technology and the internet has led to an unprecedented proliferation of information, making data a cornerstone of every successful enterprise.

Big Data refers to extensive datasets characterised by complex structures that present challenges in storage, analysis, and visualisation. With the continuous growth in data generation from various sources such as online transactions, social interactions, and IoT devices, businesses and organisations are compelled to explore innovative approaches for managing and extracting value from these vast datasets (Allaymoun et al., 2022).

Big data encompasses vast and exponentially growing quantities of information. Traditional data analytics tools face challenges in analysing such massive datasets, with examples including the daily generation of over 1 TB of data by the New York Stock Exchange and 400 TB+ daily data by social media platforms like Facebook (Rana et al., 2023).

### Classifying Big Data

Delving into the taxonomy of big data, Rana et al. (2023) provide valuable insights into its various types, namely structured, unstructured, and semi-structured data. Each category brings its own set of characteristics, highlighting the multifaceted nature of data in contemporary analytics.

Structured data is represented in a well-defined manner, often in the form of rows and columns. It is easily amenable to data models, facilitating relationships, updates, deletions, and modifications. The security features of structured big data are also relatively straightforward.

In contrast, unstructured data lacks a definite structure and cannot be easily fit into data models. This type of data is often portable and scalable, presenting challenges in storage due to the absence of a proper schema.

Semi-structured data possesses some structure but does not conform to a rigid data model. It includes metadata for grouping and describing data, offering flexibility and portability. While queries on structured big data are more efficient, semi-structured data accommodates diverse properties and sizes within the same group.

A diagram of a big data

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*Figure 6 – Big Data Types (Rana et al., 2023).*

### Benefits and Challenges of Big Data Analysis

The analysis of large datasets offers substantial benefits, including the development of efficient techniques for predicting future observations and gaining insights into the relationships between different variables. Big data analytics, with its focus on exploring heterogeneity and commonality across subpopulations, provides a unique opportunity to uncover hidden structures and extract essential common traits.

The landscape of big data analytics is evolving rapidly, driven by technological innovations such as big data and cloud computing. Cloud-based delivery models, exemplified by platforms like Amazon’s Big Data Analytics and SAP Big Data Analytics, offer scalable and accessible solutions for organisations seeking to harness the power of big data (Allaymoun et al., 2022).

Despite the advancements, challenges persist in efficiently pricing and distributing data in big data services. To address this, an auction-based big data market model is proposed, incorporating considerations of data size and analytics performance. The integration of machine learning algorithms and Bayesian profit maximisation auctions aims to provide a rational and computationally efficient mechanism for optimising service pricing and data distribution (Allaymoun et al., 2022).

### Data Visualisation in the Era of Big Data

Data visualisation, the graphical representation of information, has long been a valuable tool for conveying complex concepts quickly and effectively. Traditionally, data visualisation has been instrumental in detecting patterns in data; however, with the exponential growth of data, traditional approaches are becoming obsolete (Allaymoun et al., 2022). Now, more than ever, the importance of data visualisation is huge: it helps people see, interact with, and better understand data. Whether simple or complex, the right visualisation can bring everyone on the same page, regardless of their level of expertise.

In the realm of IoT, Data visualisation emerges as a state-of-the-art technology. The continuous stream of information from IoT devices gains exponential value through meaningful insights derived from visualisation techniques. Visualisation serves as a bridge between raw data streams and actionable insights, enhancing users' understanding of data patterns and trends.

The intersection of Big Data and the IoT is a critical juncture in the technological landscape. IoT focused on assigning IP addresses to every object and enabling their interconnectedness, generates massive volumes of data. Big Data analytics becomes indispensable in extracting meaningful conclusions from the raw data churned out by trillions of interconnected devices.

The characteristics of IoT data align with the defining features of Big Data, encompassing volume, variety, velocity, veracity, and value. The sheer volume of data generated by IoT devices, its diverse forms, real-time streaming, reliability, and the practical value it provides contribute to categorising IoT data as Big Data.

A diagram of a big data flow

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*Figure 7 – Big Data and IoT relationship (Mukherjee et al., 2022).*

The symbiotic relationship between Big Data and IoT is evident in their mutual benefits. While Big Data enables real-time analysis of IoT-generated data, the growth in IoT technologies prompts a demand for greater Big Data capacities. This reciprocal interaction drives technological advancements in both fields.

### Data Visualisation in the IoT Landscape

Data analytics in IoT involves analysing datasets to extract fundamental conclusions and valuable insights. Effective data analytics is crucial for advancing IoT applications and ventures, providing the necessary tools for making informed decisions based on the analysed data.

The transformative potential of insights derived from IoT data hinges on robust reporting and visualisation tools. Key factors influencing effective data visualisation in IoT include identifying pertinent information, selecting an appropriate reporting style, simplifying reports, considering enterprise data integrations, and establishing best practices for streamlined reporting (Il-Agure & Dempere, 2022).

IoT visualisation systems incorporate custom dashboard interfaces to aid users in analysing raw metrics. These dashboards provide real-time updates, interactive elements, and clarity, enhancing operators' confidence in AI models. Various visualisation models, including Tableau, Thingsboard, IBM Watson, Grafana, and Kibana Platform, offer diverse approaches to presenting and simulating IoT metrics.

### Data Visualisation and Unity

The integration of Unity, a robust game development engine, with Adobe Photoshop, introduces a dynamic synergy that transcends traditional boundaries in data visualisation, especially in light of the limited availability of dedicated visualisation tools compatible with Unity. While Unity provides a powerful platform for creating immersive and interactive environments, the existing tools for intricate data visualisation within Unity are scarce. This scarcity highlights the significance of incorporating external software, such as Adobe Photoshop, to meet the advanced visualisation needs. By combining Unity's game development prowess with Photoshop's sophisticated visualization capabilities, creators can unlock a new dimension in visual storytelling, addressing the challenges posed by the absence of dedicated data visualisation tools tailored for Unity. The result is a harmonious blend of Unity's interactive potential and Photoshop's graphic finesse, offering a unique solution to the limitations of conventional data visualisation tools within the Unity framework.

In addition to leveraging Adobe Photoshop for advanced visualisations in Unity, another viable option is the integration of Power BI, a robust business analytics tool, seamlessly embedded within the Unity environment. This integration not only expands the visualisation capabilities within Unity but also provides a user-friendly interface for dynamic data exploration. By embedding Power BI into Unity, developers and designers can harness its rich features for data analysis, reporting, and interactive dashboards, seamlessly merging the functionalities of both platforms. The combination of Unity, Adobe Photoshop, and Power BI offers a comprehensive solution to the challenges posed by the lack of dedicated visualisation tools, allowing creators to craft visually engaging and data-driven experiences within a unified development environment.

In the context of handling substantial datasets, opting for a business intelligence (BI) tool like Microsoft Power BI, as advocated by Rana et al. (2023), proves advantageous compared to conventional tools like Excel. One notable advantage lies in the superior processing speed of Power BI, outpacing Excel's capabilities when dealing with extensive data volumes. The visualisations crafted within Power BI are not only faster but also more aesthetically appealing, enhancing the overall user experience. The utility of Power Queries further facilitates the manipulation of vast datasets with ease, providing a streamlined approach to data management. In contrast, Excel encounters limitations, capping at 1.4 million rows and 16.38 thousand columns, rendering it inadequate for handling big data scenarios. Relying on Power BI, as recommended by Rana et al. (2023), addresses these shortcomings, offering a robust solution for efficient data processing, visually compelling representations, and seamless data manipulation in the realm of extensive datasets.

### Data Acquisition in Industry 4.0

Data acquisition in Industry 4.0 big data analytics systems involves collecting data from field devices for storage, visualisation, and analytics. Common data communication protocols such as OPC-UA and Modbus enable real-time or batch-oriented data collection. IoT gateways play a crucial role in data gathering, providing services like protocol translation, encryption, data processing, and wireless networking (Kahveci et al., 2022).

The accessibility and affordability of sensors allow industrial devices to generate massive amounts of data. IoT-enabled cloud platforms, exemplified by solutions like GE’s Predix, ABB’s Ability, and Microsoft Azure, offer capabilities for analysing raw production data. However, these platforms introduce dependencies on external connectivity, proprietary technologies, and custom implementation (Kahveci et al., 2022).

## Communication Protocols

### Real Time Monitoring

## Cloud Computing

Cloud computing, a transformative paradigm, provides users with internet-based access to diverse computing services, eliminating the need for on-site infrastructure. This model, featuring on-demand self-service, broad network access, resource pooling, rapid elasticity, and measured service, allows flexible resource management and cost-effectiveness. With three service models—Infrastructure as a Service (IaaS), Platform as a Service (PaaS), and Software as a Service (SaaS)—cloud computing supports various applications, from basic storage to advanced data analytics. In the context of the IoT, cloud technologies are crucial for managing the exponential growth of data. Cloud service models, including private, public, and hybrid options, offer versatile solutions for IoT integration. While financial considerations impact deployment choices, cloud computing remains a vital enabler for scalable and efficient IoT operations, aligning with industry trends (Khan et al., 2020).



# Methodology



## Research Approach

## Data Collection Methods

## Data Analysis Techniques

# Implementation



## Real-World Application of Digital Twinning

## IoT Integration and Case Studies

## Big Data and Visualisation Implementations

# Results



## Key Findings from the Implementation

## Data Analysis Results

## Successes and Challenges Encountered

# Discussion



## Interpretation of Results

## Comparing Findings with the Literature

## Insights Gained from the Study

# Conclusion



## Recap of Key Points

## Implications for Industry and Research

## Future Directions

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