

PREDICTIVE MAINTENANCE IN PHARMACEUTICAL MANUFACTURING: INTEGRATING GENAI WITH SAP PLANT MAINTENANCE FOR FDA COMPLIANCE AND QUALITY ASSURANCE

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Abstract - This research to discover the collaboration between GenAI and SAP Plant Maintenance in applying PM initiatives, targeting to improve efficiency, specifying quality assurance, and referring to compliance threats in pharmaceutical manufacturing. Pharmaceutical manufacturing is an extremely regulated sector where severe compliance to quality assurance and compliance models such as, FDA policy is significant. Thus, SAP-PM acts as a cornerstone, asset information, balancing workflow maintenance and obedience documentation effectively. The researcher has applied both secondary qualitative and quantitative methods with case study examples of companies such as Pfizer, Johnson & Johnson, and Sanofi. Additionally, further recommendations were provided including, training, digital infrastructure and others to increase the industry adaptation rate.

Index Terms- Predictive Maintenance, Pharmaceutical manufacturing, AI, ML, IoT, GenAI, SAP Plant Maintenance

I. INTRODUCTION

A. Background to the Study

PdM or Predictive maintenance in pharmaceutical manufacturing is a data-driven context that uses machine learning and sensors to decrease equipment failure. Contemporary maintenance solutions can decrease downtime of the equipment by 50% [1]. Optimized performance has led the pharmaceutical industry to approach

technologies and mitigations that can help prevent delays and maintain efficient operations [2]. PM as a transforming mitigation, leveraging real-time data and contemporary analytics to anticipate breakdowns of tools before they happen. Predictive maintenance fastens the obedience process when applied with SAP PM or “SAP Plant Maintenance,” an effective “enterprise asset management” mitigation. The application of GenAI, on the other hand, improves this ability by computerising predictive analytics, data processing, and anomaly detection.

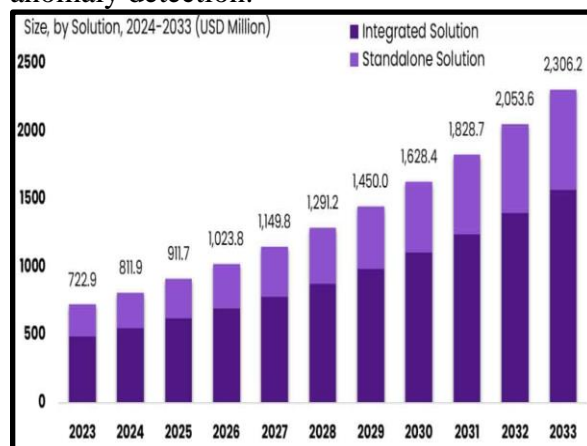


Figure 1: Global AI in Predictive Maintenance Market

[3]

Figure 1 has highlighted the global AI market in PM from 2023 to 2033. This market will expand at the CAGR of 12.3% [3]. Integrated Solutions held a suppressing market position by the mitigation or solution segment of AI in PM, holding more than 68% share [3].

Additionally, GenAI identifies patterns, interprets difficult datasets, and creates actionable measurements, decreasing the flexibility of manual applications and developing informed decision-making. In this regard, GenAI specifies conformity of GMP or “Good Manufacturing Practices” by automating abnormality observation, documentation, and remedial action planning.

B. Overview

PM in pharmaceutical manufacturing grips contemporary analytics, AI, and IoT to anticipate flaws and failures of equipment, specifying continual operations and authoritative obedience. ERP SAP is based on sensor networks [3]. Applying GenAI with “SAP Plant Maintenance” improves data-driven strategies by automating documentation, corrective measures, and detection of anomalies. This supports various industrial processes [4]. This context decreases further delays, controls the credibility of the asset, and specifies adherence to FDA norms and GMP. Thus, by linking insights from PM with obedience assurance, this initiative validates quality manufacturing while decreasing operational threats and costs in pharmaceutical productions.

C. Problem Statement

Manufacturers in the pharmaceutical sector observe major threats in maintaining the usefulness and workability of equipment, specifying regulatory compliance, and decreasing downtime. The continuance of production machines needs to be audited by the regulator in this sector [5]. Manual maintenance practices commonly lead to reactive reactions, creating major regulatory uncertainties, deviations in quality, and certain downtime. This study refers to this issue by applying GenAI along with SAP-PM to apply PM initiatives. This study cultivates how AI-based valuations and real-time data analytics smooth compliance with the FDA

norms, improve the detection of faults and specify GMP conformance. By showing developments in operational efficacy and compliance validation, this study shows a metamorphic context to mitigating these major industry-based issues.

D. Objectives

The primary purposes of this research are: 1. To specify the correlation between GenAI and SAP PM in applying predictive maintenance initiatives for fault detection and mitigation. 2. To explore the application of analytics with SAP PM to streamline the maintenance system for FDA Compliance and Quality Assurance. 3. To identify issues while integrating GenAI with SAP Plant Maintenance for FDA Compliance and Quality Assurance in Pharmaceutical Manufacturing. 4. To propose strategies that help to increase GenAI application with SAP Plant Maintenance integration in Pharmaceutical Manufacturing. These research objectives aim to discover the collaboration between GenAI and SAP Plant Maintenance in applying PM initiatives, targeting to improve efficiency, specifying quality assurance, and referring to compliance threats in pharmaceutical manufacturing.

E. Scope and Significance

This study highlights the application of AI along with SAP-PM to apply strategies of predictive maintenance in pharmaceutical manufacturing. Predictive maintenance became one of the active fields for Industry 4.0 [6]. This investigates the contributions of AI-based evaluations and real-time data analytics to improve the workability of instruments, specify FDA regulations, and maintain operational efficiencies. Additionally, study outcomes will show developmental mitigations and strategies to minimise industry threats, maintain the quality of products, and decreasing downtime. This study validates innovation, sustainability, and regulatory compliance in

the pharmaceutical market by highlighting the contribution of tools and technologies in PM.

II. LITERATURE REVIEW

A. GenAI with SAP Plant Maintenance

GenAI incorporates automation to interpret wide datasets from IoT sensors, observing lineups and anticipating causes of equipment failures before they happen. A GenAI tool was used through prompt engineering in production activities [7]. This decreases accidental downtime, regulates the credibility of assets, and specifies conformity to the norms of FDA and GMP.

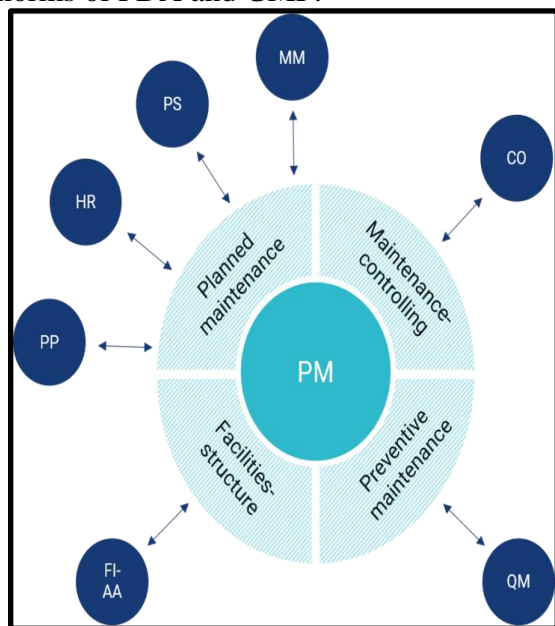


Figure 2: SAP Plant Maintenance
[8]

Figure 2 shows the attributes of PM including, planned maintenance, predictive maintenance, facilities structure, and maintenance control. In this regard, Pfizer which is an American multinational pharmaceutical corporation, integrates “Intelligent Asset Management mitigations” of SAP applied with AI-based analytics to improve its control, tracking, and specific compliance of GMP [9]. By initiating predictive maintenance, this company has decreased its maintenance charges, decreased disruptions in its production, and

developed the performance of the instruments. This context validates quality production while referring to rigid administrative demands. The coordination between SAP and GenAI shows a paradigm transformation to achieve operational particularity, sustainability, and compliance in pharmaceutical manufacturing.

B. FDA Compliance and Quality Assurance in Pharmaceutical Manufacturing

Developers in the pharmaceutical sector need to serve rigid norms including, GMP, and track their documentation and maintenance accuracy for their inspection and further audits. Pharmaceutical companies need to specify that their computerized processes are outlined, developed, and maintained under the applicable norms and industry standards, such as GAMP 5 [10]. Failure to do this leads to legal penalties, reputational damage, and increased costs for production.



Figure 3: Principal Standards of Quality Assurance
[11]

QA has a major role in specifying the efficacy, safety, and compliance of medicines. The above figure highlights the Principal Standards of QA including, process approach, relationship management, improvement, and others [11]. Thus, applying GenAI along with SAP-PM creates

a strong mitigation to achieve these demands. GenAI improves predictive analytics and real-time observation of data with wide data processes to observe major processing faults initially. This action decreased accidental disruptions that hamper and degrade product quality. SAP-PM commends this by simplifying the workflow automation in the documentation and specifying the attributability of corrective measures. Both GenAI and SAP PM increase the credibility and trustworthiness of the assets, specify conformity to FDA norms, and create immediate readiness for the audits. Thus, organisations need to create and execute validation plans, test protocols, risk assessments, and specific comprehensive documentation of compliance activities [10]. Automation in threat and anomaly detection through GenAI decreases human delusions, whereas SAP PM specifies that conservation actions are linked with authoritative standards, strengthening the quality assurance of the products.

C. GenAI threats

Applying GenAI along with SAP-PM in pharmaceutical manufacturing for FDA compliance and quality assurance imposes major threats such as **regulatory alignment**. Specifying algorithms of AI and their data outputs merged with strict FDA norms, including, “**21 CFR Part 11**” for digital records and signs. Additionally, **data integrity** as another issue refers to the management of delicate product production and acquiescence data needs strong cybersecurity actions to decrease data breaches and specify FDA-regulated integrity of data [12].

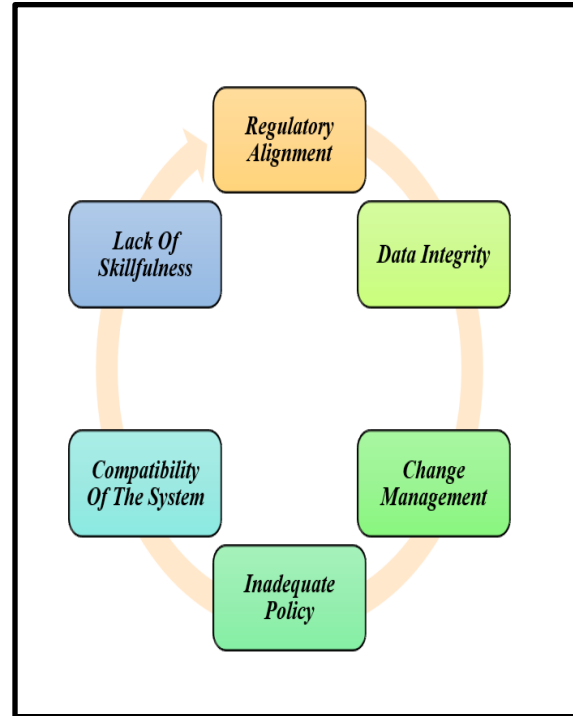


Figure 4: adaptation issue of GenAI with SAP-PM

Change management has been identified as another issue from the above image. Hesitations from the workforce and employees who are not comfortable with AI-based processes hamper its usage and further utilisation in production. **Inadequate Policy** as another threat refers to most of the company's lack of specific IT security policies and mechanisms for AI tool adoption [12]. Additionally, **compatibility of the system** as another issue refers to applying GenAI instruments along with present SAP-PM processes became problematic because of varying formats of data. Lastly, **lack of skillfulness** is another issue that indicates that companies can observe a lack of skilled workforce to handle AI and analyse insights from predictive maintenance proactively.

D. Increase Adaptation of GenAI with SAP Plant Maintenance

Adaption of GenAI with SAP-PM in pharmaceutical sectors can be increased in many ways, such as **collaborative partnerships**. Effective collaboration and knowledge-sharing are core for public health

to avoid stumbling blocks [14]. Collaborations with regulatory experts and technology providers to apply strong and submissive mitigations. A **pilot program** as another strategy refers to the starting of production with pilot projects to increase facilities, reform the development process, and reflect ROI. After these conducting **employee training** can signify the integration of GenAI with SAP-PM. **Conduct daily and weekly basis training** sessions to create knowledge and professionalism in AI-based instruments and impose compliance among the workforce. Furthermore, **Modernization in the Infrastructure** as another strategy refers to the investment in IoT-based sessions and cloud computing domains to validate quick integration of data.

III. METHODOLOGY

A. Research Design

Research design is a plan to answer research questions and objectives by gathering, analysing, and interpreting data. Thus, to identify the role of “Predictive Maintenance in Pharmaceutical Manufacturing with GenAI and SAP Plant Maintenance for FDA Compliance” the researcher has chosen an “**explanatory research design.**” Therefore, it is an approach to finding details about why something occurs.

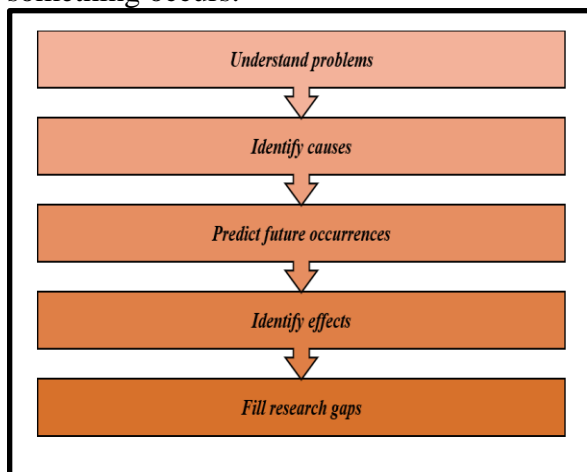


Figure 5: Benefits of explanatory research design

The above figure has highlighted several benefits of “explanatory research design” in

any research such as, Explanatory design navigates the researcher to observe the reason for the phenomenon. This is building a structure or plan for the research [15]. This also guides researchers to anticipate future occurrences of the phenomena. Moreover, this research design is approached here as it can create evaluations into how predictive maintenance affects compliance with the FDA, operational efficiency, and ensures product quality. Observing underlying components and testing theoretical and statistical measurements, crates in-depth thinking and assurance of the initiative to fulfil research objectives.

B. Data Collection

This research applied a mono research method based on **secondary quantitative and qualitative** data sources and analysis techniques. Data sources used for the **secondary qualitative method** are **journals-articles, case study examples**, and others. Case study examples include company data operating in the pharmaceutical sector. Additionally, **statistical values, charts, graphs, industry reports**, and **metrics** are collected and further analysed in a **secondary quantitative method**. Both research strategies help create a strong base of GenAI with SAP Plant Maintenance for FDA Compliance and Quality Assurance in the Pharmaceutical sector.

C. Case Studies/Examples

Case Study 1: SAP Data Intelligence for deliver life-saving vaccines

Pfizer integrates SAP Data Intelligence to create real-time information, and trusted data required to create and deliver life-saving vaccines at the time of COVID-19 [16]. This has included an incorporation of a new ML tool such as, “Smart Data Query (SDQ).”

Case Study 2: Johnson & Johnson’s Collaborative Integration to AI & Predictive Analytics

Johnson & Johnson collaboratively applied AI and PA, creating faster abnormality

detection during crisis times. This company has created its global recognition through the COVID-19 surveillance dashboard with the help of data visualization and predictive analytics systems [17].

Case Study 3: Sanofi “all in” approach

Sanofi is “all in” on AI and data science data science to fasten breakthroughs for its patients. This company has created several AI programs to decrease research times with the help of “automated time-sink” and “predictive modelling” tasks [18].

D. Evaluation Metrics

Evaluation metrics for this study are **Accuracy, OEE “Overall Equipment Effectiveness,” Compliance Rate, and Downtime Reduction.** For example, Prediction Accuracy is a measure that measures the potentiality of artificial intelligence to anticipate further failures [2]. Additionally, downtime reduction measures the reduction in accidental downtime of the tools or equipment in the sector.

IV. RESULTS

A. Data Presentation

The integration rate of GenAI sector-wise is increasing as organisations value authoritative compliance and overall operational efficiency. Market adaptation of GenAI is regulated by developments in IoT, AI, and increasing recruitment for data-based mitigations [7]. Pharmaceutical companies’ organisations such as, Pfizer show its successful integration, cultivating the ability of GenAI to reform the sector.

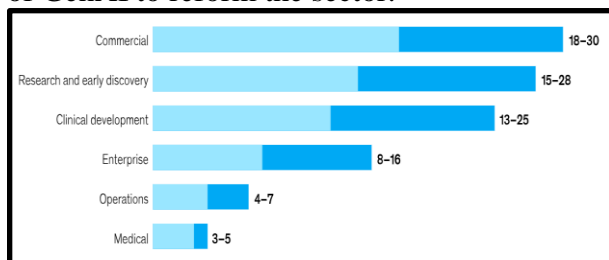


Figure 6: GenAI in the Pharmaceutical Sector
[19]

Figure 6 has highlighted that GenAI is anticipated to create \$60 billion to \$110 billion in 1-year value around the value chain of the pharmaceutical sector [19]. GenAI cultivates quality assurance as well as FDA compliance through automation in data interpretation, reporting, and abnormality detection of data. GenAI validates real-time tracking of tools and regulates predictive maintenance that reduces variations of quality and specifies homogeneous adherence to GMP.

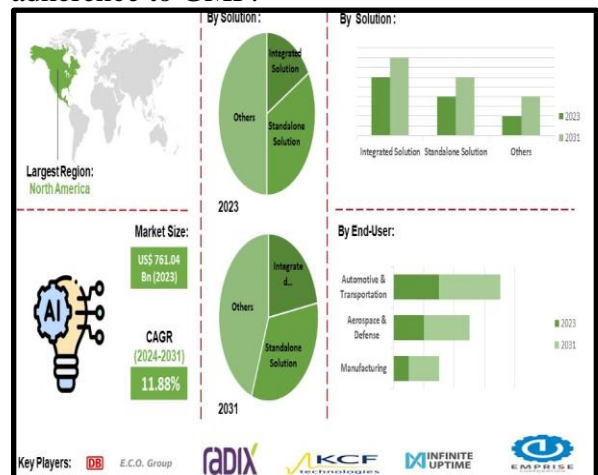


Figure 7: AI-driven Predictive Maintenance Market Overview
[20]

Figure 7 has created an overview of the AI-driven Predictive Maintenance Market which was valued at around USD 761.04 million and it is anticipated to touch 1.868.51 million at 11% CARG at the time of the forecast timeline for 2024 to 2031 [20]. Hence, the role of PA or predictive maintenance is pivotal in the pharmaceutical sector and its production by specifying the credibility and particularity of tools applied in the extremely operated manufacturing methods. PA leverages AI and IoT sensors to track the longevity of tools [10]. This process decreased certain downtime problems, decreased repair charges, and further disruptions that hampered the compliance and credibility of the products with FDA and GMP standards. Predictive maintenance also guides the preparedness of audits by

generating detailed, and detectable maintenance documentation.

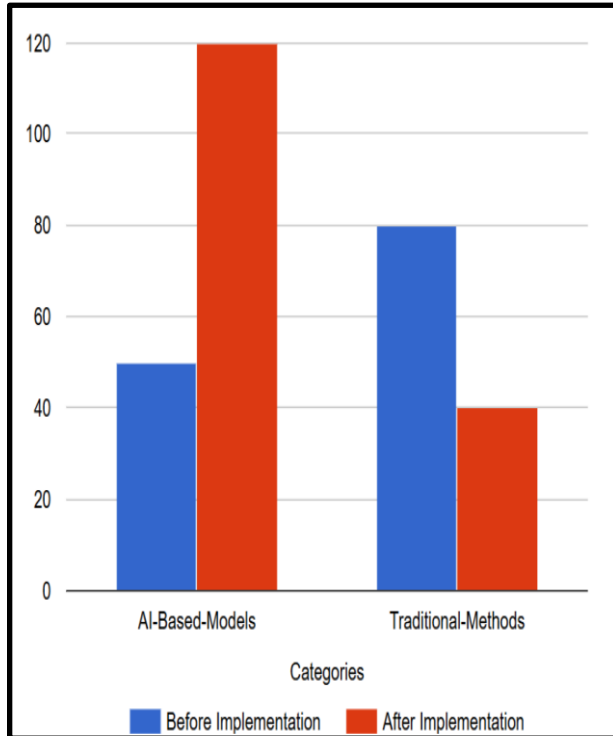


Figure 8: Number of Failures anticipated [21]

The above bar chart has highlighted the effect of Predictive Maintenance Failure Prediction before and after Implementation by comparing measurements between AI-based and traditional models. Thus, compared to the traditional model fault detection rate is higher in AI-based models after applying PM [21]. Future Technology Trends in PM include AI (25%), ML (20%), Edge Computing (15%), and IoT (15%) [21]. With the help of AI, Data-driven PM anticipates challenges, automates decision-making, and optimises all-over performances.

B. Findings

Minor deviations in pharmaceutical manufacturing can significantly impact product quality. PM specifies that the tool is designed to monitor critical parameters, to ensure reliability and the production of high-quality pharmaceutical products. The 1st graph shows the GenAI adaptation in the Pharmaceutical Sector [19]. GenAI can

produce 60 billion to \$110 a year in economic value for medical-product sectors and pharma companies as it can increase productivity levels by accelerating the system of highlighting elements for possible new medication interventions [19]. Additionally, the second graph has shown an all-over market overview of AI-driven Predictive Maintenance. This market is segmented depending on mitigation and sector. Depending on mitigation, this market is categorised into standalone and integrated mitigation. By sector, the PM market is segmented into transportation, automotive, telecommunication, and healthcare [20]. Figure 7 has highlighted the effect of PM on Failure Prediction both in AI-based and traditional frameworks, where early n Failure Prediction has increased in AI-based frameworks after applying PM [21].

D. Case Study Outcomes

Case Study	Company	Key Outcome
SAP Data Intelligence for delivering life-saving vaccines	Pfizer	This has included the incorporation of Smart Data Query (SDQ) and as an outcome vaccine clinical trial data was ready to be rechecked in only 22 hours after achieving the base efficacy case counts [16].
Johnson & Johnson's Collaborative	Johnson & Johnson	As an outcome, this company has

Integration to AI & Predictive Analytics		observed more out-of-the-box and cutting-edge applications of AI from J&J [17].
Sanofi “all in” approach	Sanofi	As an outcome, AI allows R&D departments to accelerate and scale processes and targets therapeutic domains such as immunology, neurology, and others by 20 to 30% [18].

	drugs that can be repurposed by investigating real-world evidence sources, such as AI.”	mentions in clinical trials [7].		the combinations [7].
[10]	“QM S is to prevent quality issues, detect and correct deviations, and drive continuous improvement in	Pivotal roles of comprehensive, strategic, and comprehensiveness in digitalisation [10].	Data integrity challenges, and resistance to change [10].	Digital transformation improves productivity and efficiency [10].

Table 1: Case Study Outcomes of three companies

E. Comparative Analysis

Aspects of Literature	Focus	Key findings	Challenges	Solutions
[7]	“Introduce a network specialized in predicting	AI effectively accelerated the method of highlighting drug	Threats in GenAI to Manipulate Drug Development [7].	The threat was mitigated by programmatically constructing

	processes and products”			
[13]	“Focus on the realm of Generative AI (Gen AI).”	No-code AI domains show a transforming opportunity for democratizing AI [13].	GenAI challenges and gaps in the “interdisciplinary concept of Hybrid Intelligence” [13].	Organizational upskilling and collaborative innovation [13].
[14]	“Investigate the types of partnerships that have emerged during the pandemic to develop COVID-19 vaccines”	Collaboration is core for initiatives and potentialities in the race to fight crises [14].	Policy and healthcare are challenges [14].	Policies are required to encourage collaboration and knowledge-sharing for interventions [14].

Table 2: Elaborations of literature review sources

V. DISCUSSION

A. Interpretation of Results

Applying GenAI with SAP-PM for predictive maintenance in the pharmaceutical sector creates major facilities. GenAI improves the interpretation of real-time data, decreasing certain system downtime and anticipating disruptions of the tools initially. This also specifies compliance added with GMP and FDA standards. Additionally, SAP-PM is a cornerstone that balances workflow maintenance and obedience documentation effectively [7]. This proactive context regulates the credibility, and reliability of the assets, decreases further product and tool maintenance charges, and specifies continual quality assurance of the product. On the other hand, change management, Inadequate Policy, compatibility of the system, and others have been identified as GenAI integration issues in the sector [12]. The application consolidates maintenance workflows, validates readiness of audits, and initiates automation in the documentation. This technological-based mitigation develops overall operational efficiencies, specifies administrative adherence, and decreases threats, making it significant for the pharmaceutical sector, specifically in its productions.

B. Practical Implications

Practical usefulness and implication of this study lines in the development of operational adaptability, decreased downtime, and developed obedience to FDA standards. Application of GenAI with SAP-PM imposes higher product maintenance, decreasing major and expensive failures of tools and disruptions in production. This process also faster the documentation process and helps by creating real-time data decision-making. Hence, by exploiting PA, the pharmaceutical

sector and companies specify the quality of products, optimises further costs, and specify authoritative outcomes, increasing overall performance and creating competitive advantages.

C. Challenges and Limitations

However, this research had several boundaries. The qualitative samples included just 2 to 3 case study examples operating in the pharma sector such as Pfizer, Johnson & Johnson, and Sanofi [16]. Other limitations observed included the restriction of data by companies because of sharing constraints between 1 or more agencies. Additionally, too much dependence on secondary sources also hampers the credibility and the value of the research outcomes.

D. Recommendations

In case of industry experts, willing to incorporate GenAI with SAP Plant Maintenance for FDA Compliance and Quality Assurance in Pharmaceutical Manufacturing. That needs to initiate the process with a strong **digital infrastructure**, specifying quick data integration of AI, and IoT sensors [7]. Additionally, by **concentrating on ongoing tracking and optimisation**, experts improve the performance, trustworthiness, and efficiency of the equipment. Furthermore, industry experts can concentrate on **providing employee training** to raise the adaptation and usefulness of GenAI with SAP Plant Maintenance.

VI. CONCLUSION AND FUTURE WORK

In conclusion, maintaining the best possible production systems and optimized configuration conditions were major requirements in the pharmaceutical sector to not only demolish malfunctions that hamper the correct functioning of the production but also specify the highest possible quality of products. Additionally, future work of this study needs to concentrate on reforming algorithms of GenAI for corrective PM and

improving the application of IoT, AI, and blockchain technology. Diversifying scalability around several pharmaceutical companies and referring issues such as lack of skilfulness and others to AI-based processes will be major. Thus, future investigation into authoritative alignment and optimization navigates innovation and incorporation.

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