

Generalized Overview of Go-to-Market Concept for Smart Manufacturing

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Abstract - The integration of artificial intelligence (AI) and machine learning (ML) algorithms through computer vision has emerged as a pivotal force in revolutionizing the manufacturing industry. This paper presents an exploration of the multifaceted impact of AI and ML on manufacturing processes, with a focus on cost-saving measures, increased productivity, reduced production time, improved quality checks via object detection, optimized capital utilization, and the automation of existing processes. Additionally, we introduce the concept of Go-to-Market (GTM) strategies for business model generation in the context of a web-based service offering AI and ML solutions to the manufacturing sector. This review aims to provide insights into how the amalgamation of AI, ML, and GTM strategies reshapes the manufacturing landscape, offering a holistic approach to the development of manufacturing business mode.

Key Words: Go-to-Market, Artificial Intelligence, ML Algorithms, Computer vision, Business Model Generation, Smart Manufacturing, Manufacturing industry.

1.INTRODUCTION:

The landscape of industrial processes is undergoing a transformative revolution with the advent of smart manufacturing, marked by the integration of cutting-edge technologies, particularly Artificial Intelligence (AI) and Machine Learning (ML) algorithms [4][1]. This research embarks on a comprehensive exploration, emphasizing the development of a generalized go-to-market concept within the realm of AI implementation in smart manufacturing [5][2]. The significance of this concept lies in its pivotal role in augmenting Improved quality, reduced decision-making time, increased productivity, optimized resource utilization and maximizing automation to minimize manual error.

In navigating the transition from traditional methodologies, such as manual quality checks, paper-based documentation, manual data analysis, and decision-making, to intelligent manufacturing with automated quality checks, digitalized workflows, adaptive manufacturing, connected systems, and real-time analytics, our study identifies key value propositions specific to the manufacturing sector [3]. These

include heightened manufacturing flexibility, increased efficiency, and the ability to respond rapidly to evolving market demands [1]. To facilitate this paradigm shift, we propose an AI-assisted approach supported by trained ML algorithms, tailored for enterprises of various scales, presenting a lucrative business opportunity by providing scalable solutions [2].

Aligned with sustainability objectives, the implementation of AI projects not only enhances operational processes in manufacturing, assembly, and production lines but also demonstrates the business viability of the proposed architecture [6]. Our investigation, particularly addressing challenges in the wind energy sector, yields valuable insights into revenue streams and market positioning for businesses operating in intelligent manufacturing for smart factories [7].

In conclusion, this research contributes not only to academic knowledge but strategically outlines a business model for automating the industrial sector. By leveraging emerging technologies, businesses can position themselves as leaders, delivering innovative, sustainable solutions and consulting services [8][1]. This approach is crucial for manufacturers navigating the dynamic landscape within the frameworks of Industry 4.0 and Industry 5.0, ushering in a new era of industrial excellence and innovation.

1.1 Artificial Intelligence in manufacturing Industry

Manufacturing is undergoing a revolution thanks to artificial intelligence (AI), which is generating previously unheard-of increases in production and efficiency [1][2]. Manufacturers can anticipate equipment breakdowns and cut downtime and maintenance costs by using machine learning-powered predictive maintenance [1]. AI-powered automation streamlines processes, improves accuracy, and quickens production cycles [2]. AI-driven computer vision improves quality control by eliminating waste and guaranteeing high standards [1]. The industry's transition is further aided by increased demand forecasting, collaborative robots (co-bots), and resource optimization [2]. Even while artificial

intelligence (AI) has many advantages, its full potential cannot be achieved unless issues like data security and worker preparedness are resolved [1].

The impact of AI extends beyond the factory floor, influencing demand forecasting, inventory management, and customer relationship management [1]. Despite the transformative benefits, challenges like data security, workforce adaptation, and ethical considerations necessitate careful navigation [9]. The integration of AI in manufacturing signifies a fundamental shift, propelling the industry towards a future characterized by smart, efficient, and sustainable practices [2]. As manufacturers continue to embrace AI, they unlock new possibilities for innovation, competitiveness, and overall business success in the ever-evolving global market [1].

1.2 Object detection via computer vision and ML algorithms.

Object detection through computer vision and machine learning algorithms has become a cornerstone in modern manufacturing, specifically revolutionizing quality checks and real-time supervision. This technology facilitates automated inspections, utilizing high-resolution cameras and advanced ML algorithms to swiftly identify defects or deviations from quality standards [6][7]. This not only enhances the accuracy of quality control processes but also expedites inspections, reducing reliance on time-consuming manual checks.

In the realm of real-time supervision, object detection provides continuous monitoring of production processes. By swiftly analysing the flow of items on the assembly line, these algorithms can promptly identify issues or inefficiencies [7]. This immediate feedback empowers manufacturers to make on-the-fly adjustments, ensuring that production stays aligned with quality benchmarks and operates at optimal efficiency. The seamless integration of object detection in quality assurance and real-time supervision signifies a transformative shift, where manufacturing processes become not only more precise but also adaptive and responsive to evolving demands. This amalgamation of computer vision and ML technologies is propelling the industry toward a future where quality control is intelligent, efficient, and an integral part of the production landscape [7].

1.2 Go-to-Market concept for smart manufacturing business model generation.

In the context of smart manufacturing driven by artificial intelligence (AI) and machine learning (ML) algorithms, the synergy of Go-to-Market (GTM) strategy and business model generation becomes paramount for success. A Go-to-Market strategy tailored for the smart manufacturing industry involves a comprehensive plan to introduce and promote AI

and ML-based solutions that optimize manufacturing processes.

The synergy between the GTM strategy and the business model is crucial in the dynamic field of smart manufacturing. The GTM strategy dictates how AI and ML solutions are introduced to manufacturing stakeholders, aligning with the core principles of the business model. In return, the business model provides the foundational structure for executing the GTM strategy effectively. This holistic approach not only optimizes manufacturing operations through the integration of AI and ML but also establishes a resilient and competitive presence in the rapidly evolving landscape of smart manufacturing.

2. Methodology.

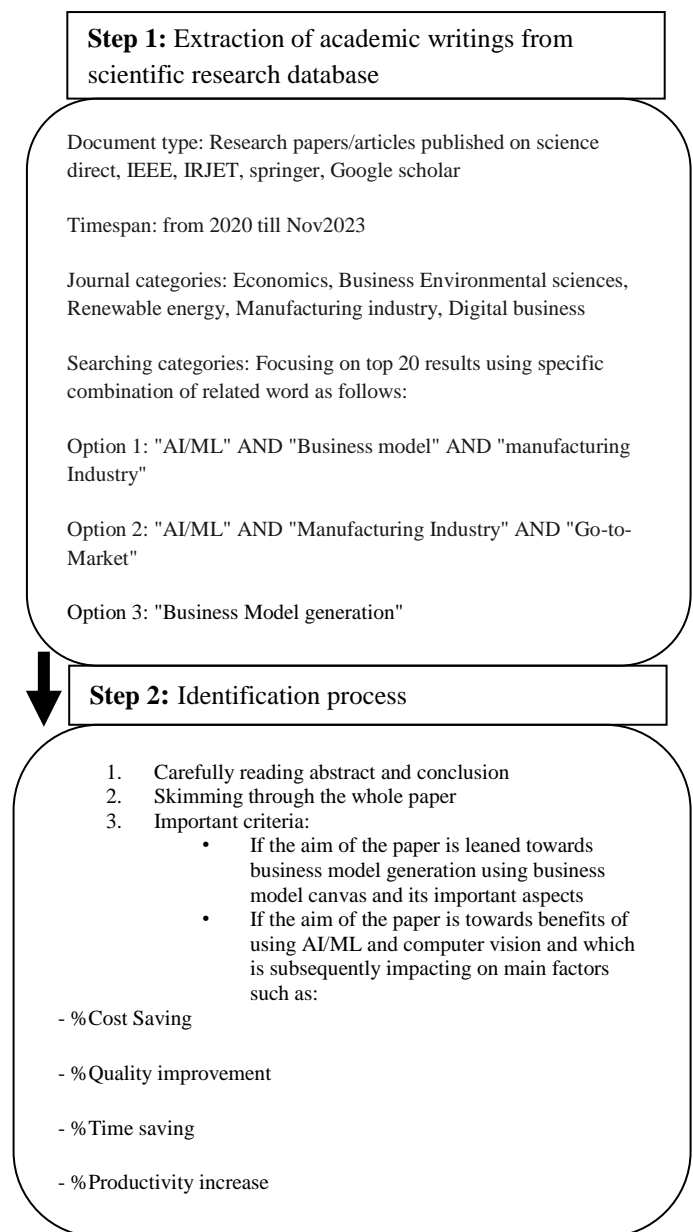


Fig -1: Review Methodology

To comprehensively explore the multifaceted impact of artificial intelligence (AI) and machine learning (ML) on the manufacturing industry, a research methodology combining a scoping review and qualitative content analysis was employed. The scoping review aimed to provide an initial understanding of the implications of AI and ML in manufacturing. Given the novelty and rapid development of these technologies, a systematic scoping review method was chosen, encompassing both scientific and non-scientific sources. This inclusive approach ensured coverage of developments at the forefront of both scientific research and public discourse. For this review paper, due to no access to paid publications like Science Direct, Elsevier, or Springer databases, Google Scholar was used, and only articles with free access and institutional access were reviewed. Our methodology comprises two research steps: the first entails

identifying, reading, and comprehending significant documents, while the second involves a bibliometric examination of the identified articles. The initial process consists of Step 1: extracting academic writings from a scientific research database using relevant keywords related to the reviewed topic, and Step 2: identifying pertinent articles by thoroughly reading each paper. If the literature review contains information related to the keywords mentioned at the beginning, the article is selected for inclusion in the bibliographic data collection and classification

Table -1: Bibliometrix data collection and classification

Year	Author	Journal	Article type	Subtopic	Aims & findings	Methodology
2022	Pratheek M Goutham, Rohit Y R, T S Nanjundeswaras wamy	International Research Journal of Engineering and Technology	Article	smart manufacturing; manufacturing; technologies; challenges; automated systems; intelligent manufacturing systems	The article highlights that automated manufacturing, once envisioned as 100% automated, has shifted away from this ideal due to practical business considerations. The focus of smart manufacturing is not solely on automation levels but on a firm's independence, development, simulation, and optimization. The article emphasizes the importance of mirroring the physical sector in virtual worlds to enhance an organization's smartness. It presents a vision of smart manufacturing with six distinctive components, supported by ten hypotheses describing the principles of smart manufacturing.	Qualitative study
2021	Fotis Kitsios and Maria Kamariotou	MDPI Sustainability	Article	artificial intelligence; business strategy; digital transformation; information technology. Industry 4.0; sustainable competitive advantage	The literature review examines strategic applications of AI in corporate strategy, emphasizing predictive analytics, image recognition, and decision support systems. Despite knowledge gaps, it highlights theoretical and practical implications, managerial considerations, and suggests future research avenues, including exploring AI-human collaboration in decision-making and addressing language and source limitations. The article proposes a starting point for future studies, encouraging further exploration and potential bibliometric analyses.	Qualitative study
2023	Valentina De Simonea, Valentina Di Pasqualea, Salvatore Mirandaa	4th International Conference on Industry 4.0 and Smart Manufacturing	Article	Artificial Intelligence; Machine Learning; Micro, Small, and Medium Enterprises; Data Analytics	This research, based on a scoping literature review, outlines the growing interest in AI and ML applications in manufacturing MSMEs. It identifies limited interest from micro-enterprises, key application fields, and challenges such as data issues and a need for simplification. Future research should explore specific applications, technologies' integration, and regional variations for a more comprehensive understanding.	Qualitative study

Year	Author	Journal	Article type	Subtopic	Aims & findings	Methodology
2023	Navitha Singh Sewpersadh	Journal of Innovation and Entrepreneurship	Article	Business model, Customer relationship, Digital transformation, Value drivers	The study emphasizes the need for multidisciplinary research in the dynamic business landscape. It explores discontinuous shifts in business models by analyzing interdisciplinary literature	Qualitative study
2020	Assunta Di Vaio, Rosa Palladino, Rohail Hassanb, Octavio Escobar	Journal of Business Research	Article	artificial intelligence; business strategy; digital transformation; information technology. Industry 4.0; sustainable competitive advantage	The paper reviews AI in Sustainable Business Models, highlighting gaps in current research, such as the adoption of sustainable practices and the integration of AI into production processes. It emphasizes the need for ethical AI implementation, global collaboration, and public awareness to promote sustainability.	Qualitative study
2023	Min-Fan Ricky Lee	MDPI Processes	Review Article	intelligent control; process optimization; smart manufacturing; soft computing	Recent advancements in intelligent control theory enhance process optimization and smart manufacturing. Challenges persist, urging further research in deep learning integration, real-time adaptability, interdisciplinary approaches, scalability, security, human-AI collaboration, benchmarks, economic implications, and environmental impact. A structured procedure is proposed for needs assessment, technique selection, integration, continuous monitoring, periodic review, and update to maximize efficiency and effectiveness.	Qualitative study
2020	Wiebke Reim, Josef Åström and Oliver Eriksson	Entrepreneurship and Innovation	Review Article	business models; business model innovation; artificial intelligence; implementation. road map	The report underscores the growing interest in AI across industries but notes challenges hindering its effective implementation. Four key takeaways include understanding organizational capabilities, refining AI implementation skills, achieving internal acceptance, and fostering business model innovation. The generic framework, applicable to diverse businesses, suggests a need for further research in AI readiness, digital transformation complexities, industrial ecosystem roles, and AI business models with a focus on digital servitization.	Qualitative study

3. Academic discussion:

Year	Author	Journal	Article type	Subtopic	Aims & findings	Methodology
2021	Sung Wook Kim ¹ , Jun Ho Kong ² , Sang Won Lee ³ , Seungchul Lee	International Journal of Precision Engineering and Manufacturing	Review Article	Artificial intelligence · Deep learning · Fault detection and diagnosis · Condition monitoring · Manufacturing process	The application of AI in manufacturing faces challenges in modeling highly nonlinear phenomena. Despite being in its infancy, recent literature suggests significant potential for AI to revolutionize manufacturing, extending to medical analysis, finance, and more. Overcoming limitations involves exploring sub-branches of deep learning, offering promising convergence with other engineering sectors. The review emphasizes harnessing AI effectively in future-oriented manufacturing.	Qualitative study
2020	McKinsey and company		Insight report	Transforming advanced manufacturing through Industry 4.0	The past decade witnessed disruptions challenging companies, driving the adoption of Industry 4.0. Exemplified by the Global Lighthouse Network, AI-driven transformations in advanced manufacturing improve KPIs, with notable use cases like flexible automation and digital performance management leading to efficiency, quality, and competitiveness gains. Advanced technology, including AI-based inspections and smart maintenance, significantly impacts task accuracy and time-to-market. Synergistic use cases in smart manufacturing, particularly in automotive and white goods companies, demonstrate amplified results, cost reduction, and increased efficiency through Industry 4.0 technologies. Successful Industry 4.0 journeys leverage impactful use cases, ensuring transformative impact and readiness for future disruptions.	Industry report
2022	IBM Industrial product		Insight report	Manufacturing 4.0: From data to decisions	Manufacturing 4.0 integrates AI and ML, enhancing efficiency and decision-making. Despite potential, few use data effectively, hindering process improvement. Data Transformers, advanced in digital maturity, leverage AI, achieve cyber resilience, and excel in manufacturing and workforce capabilities, outperforming peers in revenue growth.	Industry report

It is clear from reading academic journals and pertinent studies about AI applications in the manufacturing industry that companies looking to stay ahead of the competition must employ AI-assisted manufacturing procedures. There are numerous opportunities for the development of a digital product given the rapid growth of artificial intelligence in smart manufacturing, which is driven by attractive growth factors like improved quality, increased productivity, cost savings through optimized resource planning, and increased profitability. This product has the ability to greatly support and enhance a wide range of product manufacturing processes thanks to superior data analysis. No convincing proof of the existence of a well-defined go-to-market concept or a comprehensive business model designed expressly for AI-assisted industrial processes has been discovered in the academic literature or web research.

4. CONCLUSIONS

In summary, the examined research articles all highlight how artificial intelligence (AI) is revolutionizing smart manufacturing. The first article's depiction of automated manufacturing's developing nature highlights a move toward comprehensive smart manufacturing concepts. Understanding AI's strategic applications, corporate ramifications, and function in sustainable business models is aided by the articles that follow. The difficulties and possibilities that are shown underscore how new AI is in the manufacturing sector and call for more research.

Research on AI and machine learning in manufacturing MSMEs identifies growing interest and challenges. Ethical AI implementation, intelligent control theory advancements, and Industry 4.0 success stories underscore the multifaceted impact of AI. The concept of Manufacturing 4.0 integrates AI and ML, demonstrating enhanced efficiency and decision-making. The identification of advanced Data Transformers emphasizes the importance of digital maturity.

In summary, these studies collectively advocate for ongoing research, multidisciplinary approaches, ethical considerations, and strategic implementations to fully realize AI's transformative potential in the manufacturing landscape.

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