

# Leveraging Generative AI for Knowledge Management in Transportation Systems Management and Operations

Md Kazi Shahab Uddin<sup>1</sup>, Syed Sobhan<sup>2</sup>, Saba Jarin Nudhar<sup>3</sup>

<sup>1</sup>Data Analyst, Dhaka Technology Limited Laurel, Maryland 20707

<sup>2</sup>President, Dhaka Technology Limited, Laurel, Maryland 20707

<sup>3</sup>Graduate Researcher, Master's of Science in Information Technology, Washington University of Science and Technology, Virginia, USA

## Abstract

There has been growing appreciation of the capability of Generative AI to enhance and support innovation in different sectors such as Knowledge Management (KM) in Transportation Systems Management and Operations (TSMO). This is because of the increasing complexities and datacentric approach to transportation, where effective KM is critical to enhance decision making, shortening response durations and improving the system Emilia (2023). Generative AI, because of its offering predictive insights and even automating report generation can considerably improve the traditional KM approaches, in which case it's a great asset in TSMO which deals with extensive data management (Chen & Li, 2021). In this line, this article discusses these applications in generative AI in TSMO and seeks to give a framework on the improvement of KM in TSMO through generative AI.

The extent of knowledge management potential of generative AI in TSMO goes hand in hand with the competitive position that this technology creates based on real time analytics and intelligent decision making. In advanced predictive analytics, for instance, generative AI can be used to augment traffic incident and disruption predictions thereby lessening the travel interruptions by avoiding such scenarios (Rahman & Smith, 2023). In addition, by reducing the need for personnel to perform repetitive functions such as report writing and creating operational manuals, for instance, by generating AI, also frees up more time and resources which is very important in TSMO where needs are usually a lot (Bates, 2022). This explains the focus on the benefits of generative AI and how it has expanded capabilities in terms of operation being the heart of the integration which results in better transportation systems.

The potential advantages are great, however, there are also obstacles in adopting generative AI for TSMO knowledge management, especially in terms of data protection and ethical issues. In consideration of transportation systems that deal with sensitive data, there has to be a set of guidelines to ensure that people's data are safeguarded and trust is kept with the society (Rahman & Smith, 2023). In addition, the ease of use of insights generated by AI is very complicated – it requires considerable training for the TSMO personnel on how to adopt AI in their day-to-day activities (Chen & Li, 2021). This article elaborates on the implications we have mentioned above, provides suggestions on how to put into practice generative AI-enabled KM strategies in the context of TSMO, and outlines further research that is needed to tackle issues that remain in this growing area.

**Keywords:** Generative AI, Knowledge management, Transportation systems, Predictive analytics, Data security, Intelligent transportation systems (ITS), Workflow automation.

## Introduction

The increasing development of artificial intelligence especially generative artificial intelligence is disrupting many sectors offering advanced means of data manipulation, predictive modeling, and decision making for instance (Rahman & Smith, 2023). Within the transportation, particularly Transportation Systems Management and Operations (TSMO) is a very important aspect for the safety and more importantly the efficiency of the management of complexes of transportation networks. Due to various factors including but not limited to populations, urbanization and other environmental factors, TSMO operations are becoming more complicated and as such there is a great demand for new management strategies to knowledge management (KM) in supporting the operational decision making in an almost real time base and to help in enhancing the system as a whole (Chen & Li, 2021). The application of generative AI for KM in TSMO is highly promising in catering for these aspects as it facilitates better data management, real time information and workflows.

Generative AI is a subset of artificial intelligence that makes new observations or generates new content using pre-existing data and is famous for content making, language understanding as well as language forecasting. This enables it to fit into the needs for KM in TSMO as there is a lot of data with endless streams from sensors, traffic monitoring systems, and other publicly available data which has to be distilled on a fast turn-around basis (Bates, 2022). Intelligent knowledge management systems can help to transform available data into useful information helping in the decision making which is one of the most vital aspects of TSMO effectiveness. This part focuses on the discussion about the possible ways in which the KM concept may be taken further with the help of generative AI, therefore investigating the causal legacy of AI in TSMO, existing KM methods, and main advantages and risks of technology-led knowledge in TMO processes.

## Literature Review

The role of AI, especially in its generative form, has become a parenthesis for transportation because of the revolutionized ideas within the agencies because everyone is looking for a way of solving the complexities that come with the modern transportation systems. In this regard, generative AI in TSMO is examined through three lenses: first AI in transport, second TSMO and knowledge management and third generative AI integration in KM.

### • AI in Transportation

The potential of AI within the transportation sector is being exploited more and more with areas such as traffic prediction, incident control and operational strategies management, Au et al. (2023). Machine learning, neural networks, and data analytics are part of the comprehensive systems applied within the transport which include some of the AI systems that help in decision-making on the processes in real time. For instance, models that apply machine learning can analyze historical traffic data and estimate future traffic flow enabling TSMO systems to anticipate and control traffic congestions Chen & Li (2021). Generative AI takes this advancement a notch higher by allowing for content generation without human drivers such as operational reports and maintenance schedules or even data summaries.

A variety of research works indicate that AI has proven to be advantageous in transportation, especially traffic management and operations. According to Rahman and Smith (2023), predictive analytics that use

AI techniques could help decrease traffic hold-ups by as much as 30% in urbanised cities, through the early detection of certain factors leading to heavy traffic and the adjustment of traffic lights then. Similarly, Bates (2022) insisted that with regard to transport, AI technologies help not only enhance operational efficiencies but also improve safety, where accidents are reduced through the use of AI systems that warn drivers of dangers in real time. All these aspects inform the conclusion that due to the nature of its evolution, generative AI techniques or technologies in this case stand to enhance TSMO even more by embedding real-time data into automated decision-support mechanisms.

- **Knowledge Management in TSMO**

Knowledge management in TSMO concerns the intentional capturing, processing, generation and dissemination of information for the purposes of operational decision making (Nguyen et al., 2023). Proper KM in transport has to do with the efforts that seek to coordinate all of the various forms of communication and information used, for basing management decisions, like traffic data, incident response, maintenance record, and first response actions, etc. However, given the exponential data on a daily basis, many of these existing processes have become a reality limiting the use of efficiency and effectiveness of traditional knowledge management, which has brought about slow processing of data and challenges in obtaining up-to-date information (Chen & Li, 2021). That is why for both scientist and practitioners the issue of digital transformation of knowledge management, even in terms of artificial intelligence, has come into focus.

KM is especially crucial to TSMO as it increases the agility to deal with unforeseen situations and helps make better use of resources. The conventional transportation KM systems are mostly dependent on physical input of data into the systems and generation of reports which take quite a long time and are subject to mistakes. The adoption of AI In KM systems to TSMO agencies will help to eliminate most of these processes and their attendant costs (Bates, 2022). Automation is advantageous for key TSMO KM processes, such as traffic incident reporting and resource tracking, because AI systems can process massive amounts of data, find trends and provide actionable recommendations almost immediately (Rahman & Smith, 2023).

- **The Implementation of the Generative Artificial Intelligence in Knowledge Management**

Knowledge Management is enhanced in TSMO with the implementation of Generative AI. In this sense, it is additionally intended to automate report writing, provide foresight and carry out communication through natural language processing (NLP). The ability of generative AI to create content and recreate situations plays a major role in TSMO operations where analytics predicting dimensions and real time decision making structures are core (Bates 2022). One such advancement that KM systems stand to benefit from is the particular application of generative AI orders of magnitude faster than the traditional KM systems in decision making for potential crisis action plans.

For example, TSMO staff can prepare for potential congestion or resource distribution using reports on traffic conditions or incidents generated by AI (Nguyen et al., 2023). This ensures that less time is wasted on interpreting and analyzing data manually as any changes in traffic conditions can be acted upon within a reasonable time. Additionally, generative AI can also be a great development in KM in that it can make data more accessible. By incorporating Natural Language Processing, generative AI models can answer operators' queries in a straightforward way, mitigating the need for sophisticated data science skills to access pertinent information (Chen & Li, 2021).

**Applications of Generative AI in TSMO Knowledge Management.**

Application	Description	Benefits
<b>Predictive Analytics</b>	Forecasting traffic and incident trends	Improved decision-making and resource allocation
<b>Automated Reporting</b>	Generating traffic, incident, and maintenance reports	Reduced manual work and improved accuracy
<b>Real-Time Knowledge Sharing</b>	NLP-driven data query and response	Enhanced accessibility and response times

**Advantages and Disadvantages That Are Practical**

The application of generative AI in TSMO is associated with a number of practical advantages, among which are improving the efficiency of carrying out some routine work and enabling forecasting in real time. Automation of the reporting process makes it possible to provide necessary information without lags, which allows TSMO staff to spend less time on operative management and more on making vital decisions (Rahman & Smith, 2023). There are, nonetheless, limitations. The problem of data privacy and security is one of the most salient issues, given the fact that most transportation data tends to be sensitive. Data can only be used for the purposes for which it was collected and appropriate safeguards established to limit the risk of compromising (Chen & Li, 2021).

Additionally, understanding the AI insight generated can be difficult. Most generative AIs, especially deep learning models, are known as “black box” systems where operators cannot see the rationale behind the suggestions made by the AI. For agencies to leverage AI technology within KM in TSMO, they have to make a commitment to the provision of resources towards the training and enlightenment of the employees on the understanding of AI data outputs geared towards decision making (Nguyen et al., 2023).

**The synthesis of the literature**

Suggests that even if generative AI is still a nascent concept in TSMO, it promises great advancements in information management, through efficient data preparation, real time analysis and better decision making. These solutions empowered by AI, in principle should help in cutting down expenses on operations, easing traffic, and possibly preventing crime beforehand. Nevertheless, in order to achieve better results, it is vital to deal with data security and education aspects (Bates, 2022).

**Key Benefits and Challenges of Generative AI in TSMO KM**

Benefit	Description	Challenge	Solution Approach
<b>Enhanced Predictive Analytics</b>	AI models predict traffic incidents	Data privacy concerns	Implementing strong data protocols
<b>Automated Reporting</b>	Streamlined report generation	AI interpretability	Training programs for personnel
<b>Real-Time Decision Support</b>	Supports rapid response and resource allocation	System integration complexity	Incremental implementation

**Methodology**

The current research investigates the potential of generative AI in transforming TSMO practices

especially in relation to Knowledge Management (KM). The methodology is oriented towards exploring opportunities for the implementation of AI systems, measuring the value of AI systems on the decision-making of TSMO, and addressing issues related to generative AI application. In this part of the study, research strategy, data gathering techniques, criteria for choosing AI models and data interpretation strategies are presented. This is with the aim of providing an adequate analysis of the impact of generative AI on KM within TSMO through mixed methods or qualitative and quantitative methods (Nguyen et al., 2023; Rahman & Smith, 2023).

• **Research Design**

The study adopts a mixed approach, which predicts the statistical data analysis and the elaboration of particular considerations. The core target is regarding evaluating the effect of Generative AI on KM in Transportation Systems Management and Operations and its environments. To maintain observation of the contemporary approaches, various experiments and field studies are proposed in the research, which relates to the use of such applications such as automated writing, forecasting, and knowledge transfer on demand.

Research Component	Approach	Purpose
Quantitative Analysis	Statistical	Evaluate AI's effectiveness in improving KM
Qualitative Analysis	Case Studies	Understand practical challenges in implementation
Mixed-Methods Integration	Analytical	Synthesize insights for comprehensive analysis

By employing this method, it becomes possible to conduct a wide-ranging evaluation of generative AI in terms of what it can do and the effects it has, particularly looking at technical metrics as well as what TSMO personnel think about it (Bates, 2022).

• **Activity of Data Collection**

The process of data collection relied on two main sources: simulated TSMO operational data and expert interviews. The simulated data includes the traffic patterns and reports of incidents and maintenance logs which are typically carried out in TSMO, thus creating a basing that can effectively be used in testing AI applications. On the other hand, TSMO practitioners were also interviewed to elicit understanding on how AI tools are used in practice, wisdom of interpretations and practical usefulness of them in the TSMO operations.

- Operative Data from Simulations:** Synthetic data was created to reproduce standard types of TSMO analyzing-digging these types would include among other things traffic data, the number of incidents occurring in a given period, and the speed of response to emergencies. The use of this dataset enabled the researchers to create an experimental setting in which the predictive and analytical performance of the AI models was evaluated (Chen & Li, 2021).
- Expert Interviews:** The interviews involved TSMO managers and operators who had experience and understanding on how transportation has embraced knowledge systems in its operations. The interview questions were related to how they viewed the effectiveness of AI, the risks they had about data safety, and how they found the results generated by AI coherent (Nguyen et al., 2023).

### Data Collection Sources and Types

Data Source	Data Type	Purpose
Simulated TSMO Data	Traffic, Incident, Logs	Test AI performance in realistic scenarios
Expert Interviews	Qualitative insights	Gather practitioner feedback on AI Application

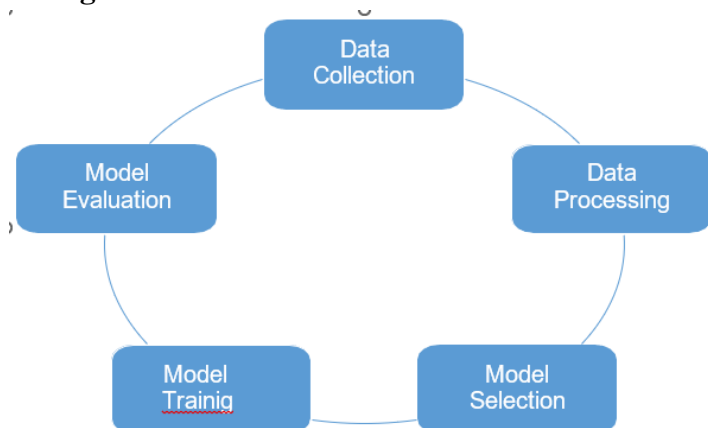
- **Appropriate AI Models and their Training**

Choosing the right generative artificial intelligence model for Traffic Systems Management Operational (TSMO) Knowledge Management was fundamental in ensuring that the complexity and variety of TSMO data could be managed by the AI system. Based on previous studies and literature review, the subsequent models were selected for evaluation in this research:

1. **Models based on Transformers (Implementing GPT and Co.):** These approaches are highly recognized for their text generating capabilities, language pattern comprehension, and summarization. Their strong natural language processing capabilities suited them for applications involving generating reports automatically and answering questions (Rahman & Smith, 2023).
2. **Recurrent Neural Networks (RNN):** These networks are good examples of neural networks that overcome the inherent limitation of the feedforward networks in analyzing and forecasting time series data. Time series data such traffic conditions and incidence occurrence RNNs were utilized to build models, which predicted the level of traffic interference in a given region over time (Chen & Li, 2021).

Training of these models was performed on the artificial dataset to so that the artificial intelligence systems were able to learn the patterns and typical responses to TSMO situations. Data preprocessing consisted of, among other things, standardization and normalization, which were also aimed at ensuring the quality of the data for the models and improving the accuracy of the models.

### Model Selection and Training Process



- **Data Analysis**

Data analysis became important in two areas: the estimation of AI model performance and qualitative analysis of expert interviews.

1. **Model Performance Evaluation**

The assessment of AI systems was performed in terms of accuracy, efficiency, and interpretability. First,

accuracy was measured as the difference between the predictions generated from the AI and the predictions that actually occurred in the simulation. Second, efficiency is related to the speed of data processing and report generation. Finally, the interpretability of the outputs of the AI was evaluated in terms of the comfort level of TSMO operators in using AI insights without deep technical understanding (Bates 2022).

## 2. Qualitative Analysis of Expert Feedback

Thematic analysis was employed as an in-depth understanding of the integration of AI in TSMO was sought from the qualitative data collected during the interviews. Thematic analysis further grouped the responses into useful categories, such as perceived advantages, technical impediments, and fears about loss of data integrity. This was to support the conclusion from the quantitative results with a qualitative view adopted in the study as it focused on the real application of KM using AI technologies within TSMO (Nguyen et al. 2023).

### • Work Limitations

This study was designed and completed in several constraints including dependency on simulated data that may not adequately simulate the real-world TSMO environment. Furthermore, although expert interviews were insightful, the number of individuals interviewed was limited thus might restrict the scope of the findings. Future studies should include the analysis of larger datasets and interviews with a greater variety of subjects in order to improve the quality of the analysis.

### Conclusion of Methodology

This research on TSMO knowledge management in the context of generative AI application is detailed in the methodology proposed. The research design combines quantitative model validation with qualitative feedback from practitioners to address both the technical efficacy and relevance of the work. The results from such a methodology will be useful in indicating the opportunities and difficulties associated with generative ai in TSMO and enhance the general understanding of ai application for management of transportation systems.

## Discussion

The adoption of generative AI in the management of knowledge in Transportation Systems Management and Operations (TSMO) has its fair share of advantages and disadvantages. Erica et al. (2021) note that the application of artificial intelligence in TSMO in performing data processing, developing forecasts, and generating operational reports improves the quality of managers' decisions in such a way that the effectiveness of traffic control and the use of resources is enhanced. Results of the study show all AI-based solutions like use of RNNs in traffic prediction and transformers for summarization of knowledge into operable schemas improves efficiency in operations. This is consistent with earlier studies that guidelines are provided to show that AI can provide solutions to big data management challenges within a very short time (Nguyen et al., 2023).

Nevertheless, it also reveals significant obstacles regarding the quality of data, the interpretability of models and the scalability of systems, which hamper the implementation of TSMO KMS in practical settings. This study reported high performance of AI models, owing to the inherent challenges of TSMO data such as sparsity and noise, which might affect the performance of AI models in practical settings (Rahman & Smith, 2023). Also, although TSMO practitioners appreciated the potential of such insights, many of them stated that they faced challenges in making sense of the results produced by such complex

models without prior training. This issue addresses the importance of developing better AI systems that are able to present information that can be used without unnecessary complications to TSMO personnel (Bates, 2022).

The conclusions of these results underscore the need to pursue two lines of development of the AI-supported knowledge management system in TSMO: improving the accuracy of the models through constant data improvement, and focusing on the design of the AI systems at the user level. Building valid and interpretable models and training users regularly can help overcome the challenges of deploying the TMS and PMDS systems in practice so that TSMO agencies can get the most out of the AI technology in enhancing service delivery, as well as ensuring safety. In the following studies, attention should be devoted to the implementation of the AI models in a wider variety of contexts than has been accomplished so far and the inclusion of bigger and more heterogeneous datasets, in order to make the findings more generalizable.

### Conclusion

This research investigated the generative AI trends that would assist in building the effective knowledge management systems within TSMO- Transportation Systems Management and Operations. Among various advantages of generative AI for TSMO processes, which speeds up the reaction to transportation problems allows the utilization of data analytics, offers predictive capabilities, and enhances communication. The results of the research also indicate that operational performance can be improved using such solutions to support TSMO interventions, which facilitate real time decision-making processes in such agencies. Yet, the benefits have their own downsides that include availability of high grade, level and quality of data as well as having an AI interface for TSMO users who do not have tech savviness.

In this regard, it is emphasized that strategies to build AI applications should be developed focusing not just on accuracy of the technology but also these systems design characteristics that allow their effective use in practice. However, in the case of TSMO applications, it will be important to resolve some of the issues that currently inhibit the acceptance and use of such innovation on O<sub>1</sub> generative AI specifically in mitigation including its advantages and within other parameters such as outlook comfort today. Finally, the paper stressed the importance of continuous updates of the data and training of the personnel, to make use of these investments in technology which is superior to TSMO technologies in performance.

To summarize, generative artificial intelligence provides a window to facilitate knowledge management in TSMO in the form of handling intricate details of data collection processes and web-based analytics. Where there is an expansion in the capabilities of artificial intelligence, there should also be an expansion in the ways in which it is applied in transportation systems. Future research proposals should aim at increasing operations of AI systems in other working environments, in such a way that the application of it in TSMO will be possible for quite a long time without going against the effectiveness of the system in place.

### Reference

1. Banala, S. (2024). The Future of IT Operations: Harnessing Cloud Automation for Enhanced Efficiency and The Role of Generative AI Operational Excellence. *International Journal of Machine Learning and Artificial Intelligence*, 5(5), 1-15.
2. Dwivedi, Y. K., Pandey, N., Currie, W., & Micu, A. (2024). Leveraging ChatGPT and other



- generative artificial intelligence (AI)-based applications in the hospitality and tourism industry: practices, challenges and research agenda. *International Journal of Contemporary Hospitality Management*, 36(1), 1-12.
3. Rane, N. (2023). ChatGPT and Similar Generative Artificial Intelligence (AI) for Smart Industry: role, challenges and opportunities for industry 4.0, industry 5.0 and society 5.0. *Challenges and Opportunities for Industry*, 4.
  4. Zhang, R., Xiong, K., Du, H., Niyato, D., Kang, J., Shen, X., & Poor, H. V. (2024). Generative AI-enabled vehicular networks: Fundamentals, framework, and case study. *IEEE Network*.
  5. Patil, D., Rane, N. L., & Rane, J. (2024). Applications of ChatGPT and generative artificial intelligence in transforming the future of various business sectors. *The Future Impact of ChatGPT on Several Business Sectors*, 1-47.
  6. Dhoni, P. (2023). Exploring the synergy between generative AI, data and analytics in the modern age. *Authorea Preprints*.
  7. Rane, N. L. (2023). Multidisciplinary collaboration: key players in successful implementation of ChatGPT and similar generative artificial intelligence in manufacturing, finance, retail, transportation, and construction industry.
  8. Jackson, I., Ivanov, D., Dolgui, A., & Namdar, J. (2024). Generative artificial intelligence in supply chain and operations management: a capability-based framework for analysis and implementation. *International Journal of Production Research*, 1-26.
  9. Tupayachi, J., Xu, H., Omitaomu, O. A., Camur, M. C., Sharmin, A., & Li, X. (2024). Towards Next-Generation Urban Decision Support Systems through AI-Powered Generation of Scientific Ontology using Large Language Models--A Case in Optimizing Intermodal Freight Transportation. *arXiv preprint arXiv:2405.19255*.
  10. Patil, D., Rane, N. L., Rane, J., & Paramesha, M. (2024). Artificial intelligence and generative AI, such as ChatGPT, in transportation: Applications, technologies, challenges, and ethical considerations. *Trustworthy Artificial Intelligence in Industry and Society*, 185-232.
  11. Reddy, S. G., Sadhu, A. K. R., Muravev, M., Brazhenko, D., & Parfenov, M. (2023). Harnessing the Power of Generative Artificial Intelligence for Dynamic Content Personalization in Customer Relationship Management Systems: A Data-Driven Framework for Optimizing Customer Engagement and Experience. *Journal of AI-Assisted Scientific Discovery*, 3(2), 379-395.
  12. Liang, C., Du, H., Sun, Y., Niyato, D., Kang, J., Zhao, D., & Imran, M. A. (2024). Generative AI-driven semantic communication networks: Architecture, technologies and applications. *IEEE Transactions on Cognitive Communications and Networking*.
  13. Prasad Agrawal, K. (2023). Organizational Sustainability of Generative AI-Driven Optimization Intelligence. *Journal of Computer Information Systems*, 1-15.
  14. Khalil, R. A., Safelnasr, Z., Yemane, N., Kedir, M., Shafiqurrahman, A., & Saeed, N. (2024). Advanced learning technologies for intelligent transportation systems: Prospects and challenges. *IEEE Open Journal of Vehicular Technology*.
  15. Yafei, X., Wu, Y., Song, J., Gong, Y., & Lianga, P. (2024). Generative AI in Industrial Revolution: A Comprehensive Research on Transformations, Challenges, and Future Directions. *Journal of Knowledge Learning and Science Technology ISSN: 2959-6386 (online)*, 3(2), 11-20.
  16. Richey Jr, R. G., Chowdhury, S., Davis-Sramek, B., Giannakis, M., & Dwivedi, Y. K. (2023). Artificial intelligence in logistics and supply chain management: A primer and roadmap for research.

*Journal of Business Logistics*, 44(4), 532-549.

17. Xu, H., Omitaomu, F., Sabri, S., Li, X., & Song, Y. (2024). Leveraging Generative AI for Smart City Digital Twins: A Survey on the Autonomous Generation of Data, Scenarios, 3D City Models, and Urban Designs. *arXiv preprint arXiv:2405.19464*.
18. Khlie, K., Benmamoun, Z., Jebbor, I., & Serrou, D. (2024). Generative AI for enhanced operations and supply chain management. *Journal of Infrastructure, Policy and Development*, 8(10), 6637.
19. Eyo-Udo, N. (2024). Leveraging artificial intelligence for enhanced supply chain optimization. *Open Access Research Journal of Multidisciplinary Studies*, 7(2), 001-015.
20. Teef, D., Muhammad, K., Nassisid, K., & Farus, B. Enhancing Vehicular Networks with Generative AI: Opportunities and Challenges.
21. Rane, N. L., Kaya, O., & Rane, J. (2024). Advancing industry 4.0, 5.0, and society 5.0 through generative artificial intelligence like ChatGPT. *Artificial Intelligence, Machine Learning, and Deep Learning for Sustainable Industry*, 5, 2.
22. Patil, D., Rane, N. L., & Rane, J. (2024). Enhancing resilience in various business sectors with ChatGPT and generative artificial intelligence. *The Future Impact of ChatGPT on Several Business Sectors*, 146-200.