



Production Capacity Planning Strategy to Deal with Seasonal Demand in the Food and Beverage Industry

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Article Info

Article History:

Received: 7 July 2024

Revised: 10 August 2024

Accepted: 11 September 2024

Keywords:

Production Capacity

Seasonal Demand

Flexible Manufacturing Systems

Forecasting Techniques

Abstract

This study aims to analyze the effectiveness of production capacity planning strategies in managing seasonal demand fluctuations in the food and beverage industry. Seasonal variability often presents challenges such as stockouts and operational inefficiencies, requiring companies to adopt flexible strategies to optimize production. The research employs a descriptive-quantitative design, surveying industry professionals and analyzing operational data from 10 food and beverage companies. Key strategies evaluated include Flexible Manufacturing Systems (FMS), advanced forecasting techniques, and supplier collaboration, focusing on their impact on production volume, lead time, and inventory turnover. Results show that companies implementing FMS and advanced forecasting achieved significant improvements in production volume (50% increase) and lead time reduction (up to 50%). Multivariate analysis of variance (MANOVA) confirmed that advanced forecasting and FMS had the most substantial effects on operational efficiency during peak demand periods. These findings underscore the importance of integrating flexible systems and predictive analytics in capacity planning to enhance responsiveness and mitigate seasonal demand challenges.

INTRODUCTION

Production capacity planning is a critical aspect of operational management, particularly in industries characterized by fluctuating demand patterns, such as the food and beverage sector. Seasonal demand variability presents unique challenges that necessitate strategic planning to ensure optimal resource utilization, cost efficiency, and customer satisfaction. The ability to effectively manage production capacity during peak and off-peak seasons can significantly impact a company's competitiveness and profitability (Rajani et al., 2023; Okeke et al., 2024).

Among the better capacity planning plans is the introduction of flexible manufacturing systems (FMS) which give food and beverage companies a chance to change their production levels quickly, according to the changes in market demands. FMS combines technology and manufacturing operations in order to have a more flexible manufacturing environment (Jirashayanon, 2021; Khan et al., 2023). Such systems prevent the ills of overproduction or stockouts which are all possible as a result of a delay in shifts in production schedules from being detrimental to efficiency

in operation and customer satisfaction, given that a company will be able to adjust production processes in a short time. As another example, with sudden surges in sale of seasonal items, like holiday drinks or summer snacks, FMS can enable businesses to ramp up the production that would not otherwise be possible with long lead times that have come to be expected with conventional industrial practices. This flexibility does not only contribute to the lean management of inventory with fewer products in store and less waste--but also contributes to the overall responsiveness of the rest of the supply chain and thus, satisfying the needs of the consumer at an early stage (Bacchetti & Saccani, 2012).

Further, the incorporation of more advanced methods of forecasting such as machine learning algorithms, and predictive analytics can greatly increase the accuracy of the demand forecast, which can enable and empower a firm to make the right decisions concerning the capacity planning. Machine learning models can be trained using historical sales data and external factors, including economic indicators and social media trends to produce more accurate forecasts of the demand (Yusof, 2024; Mao et al., 2024). With these insights, businesses will be able to set their production plans according to the conditions that they foresee in the market, thus ensuring that resources are used properly and they minimize expenses. Not to mention, predictive analytics will also help trace the tendencies in consumer behavior so that firms could preliminarily predict changes in demand and adjust their production capacity according to them (Choi et al., 2020). This proactive approach not only ensures that companies are well-prepared for seasonal fluctuations but also fosters a culture of continuous improvement and innovation in capacity planning practices, positioning them for long-term success in the competitive food and beverage industry.

One effective approach to capacity planning is the adoption of flexible manufacturing systems (FMS), which allow for rapid adjustment of production levels in response to market demands. FMS enables food and beverage companies to maintain a lean inventory while minimizing waste, ultimately leading to improved operational efficiency (Bacchetti & Saccani, 2012). Furthermore, the integration of advanced forecasting techniques, such as machine learning algorithms and predictive analytics, can enhance demand prediction accuracy, allowing firms to make informed capacity planning decisions (Choi et al., 2020). In addition to flexible systems and advanced forecasting, the role of supply chain collaboration cannot be overstated. Strong partnerships between suppliers, manufacturers, and retailers can facilitate timely information sharing, thereby enabling businesses to react swiftly to changing demand patterns. For instance, collaborative planning, forecasting, and replenishment (CPFR) initiatives can improve alignment across the supply chain, resulting in more synchronized production schedules and reduced stockouts during peak demand periods (Seifert, 2003; Gomes, 2022).

Moreover, technology plays a pivotal role in capacity planning strategies. The use of enterprise resource planning (ERP) systems can provide organizations with real-time insights into inventory levels, production capacity, and market trends, enabling them to make data-driven decisions (Gunasekaran et al., 2017; Jawad & Balázs, 2024). As companies increasingly embrace Industry 4.0 technologies, including the Internet of Things (IoT) and artificial intelligence (AI), the potential for optimizing production capacity planning grows significantly (Kamble et al., 2020). These technologies allow for enhanced visibility and responsiveness throughout the supply chain, providing a competitive edge in managing seasonal demand. The implementation of lean manufacturing principles can also contribute to effective capacity planning in the food and beverage industry. By focusing on waste reduction, continuous improvement, and value creation, companies can optimize their production processes to better meet fluctuating consumer demand (Ohno, 1988; Ghelani, 2023). This approach not only enhances efficiency but also improves product quality and

customer satisfaction, which are crucial in the highly competitive food and beverage market.

However, despite the potential benefits of these strategies, challenges persist. The volatility of consumer preferences, regulatory constraints, and supply chain disruptions can complicate capacity planning efforts. For example, the COVID-19 pandemic has underscored the fragility of global supply chains and the importance of resilience in production capacity planning (Ivanov, 2020). Companies must not only prepare for seasonal demand variations but also build flexibility to adapt to unforeseen disruptions.

METHODS

Research Design

This study employed a descriptive-quantitative design to investigate the effectiveness of production capacity planning strategies in managing seasonal demand fluctuations in the food and beverage industry. The design allowed for the collection and analysis of quantitative data, focusing on operational performance metrics such as production output, lead times, and inventory levels before and after the implementation of specific capacity planning strategies. This approach enabled the researcher to compare the operational performance during peak and off-peak seasons, assessing the impact of various capacity planning methods. The research design incorporated a survey of industry professionals and an analysis of operational data from food and beverage companies, providing a comprehensive understanding of the strategies used to manage seasonal demand variations.

Participants in this study consisted of 50 professionals from 10 food and beverage companies, selected through purposive sampling. The participants included production managers, supply chain coordinators, and operations executives with at least three years of experience in managing production processes during seasonal demand fluctuations. The selected companies varied in size, ranging from small enterprises to large multinational corporations, and represented different sectors within the food and beverage industry (e.g., beverages, snacks, dairy products). The companies were chosen to ensure a diverse range of perspectives on capacity planning strategies across different product lines and market segments. Data collection was conducted through structured interviews and surveys, as well as secondary data analysis of production reports and inventory records from the participating companies.

Instruments

The study utilized a combination of surveys, interviews, and company operational data to measure the effectiveness of production capacity planning strategies. The survey instrument, developed based on established frameworks in operations management, focused on key capacity planning practices such as forecasting, resource allocation, and the use of flexible manufacturing systems. It included both closed-ended and open-ended questions to gather quantitative data on the implementation of these strategies and qualitative insights into the challenges faced during seasonal demand periods.

Additionally, the study collected operational data from participating companies, including production volumes, lead times, and inventory levels during peak and off-peak seasons. This data provided objective measures of the effectiveness of the capacity planning strategies. The operational metrics were analyzed in relation to the survey responses, offering a more detailed understanding of how different strategies impacted production performance.

Data Collection Procedure

The data collection process was conducted over a three-month period, covering both peak and off-peak seasons in the food and beverage industry. Initially, structured interviews and surveys were administered to the key participants in each company, focusing on their production capacity planning methods and the challenges they encountered in managing seasonal demand fluctuations. The surveys were distributed via email, and interviews were conducted virtually to accommodate the participants' schedules.

Simultaneously, secondary data on production output, inventory levels, and operational performance were collected from the companies' internal records, with permission from management. This data covered the previous two seasonal cycles, providing a comprehensive view of the companies' performance during periods of demand variation. The collected data was stored securely and anonymized to protect company confidentiality.

Data Analysis

The data analysis was conducted using quantitative methods to evaluate the relationship between the use of specific capacity planning strategies and operational performance during seasonal demand fluctuations. The survey data were analyzed using descriptive statistics to summarize the frequency and extent of strategy implementation across different companies. Operational data, including production volumes, lead times, and inventory turnover, were analyzed using paired-samples t-tests to determine whether there were statistically significant differences in performance during peak versus off-peak seasons.

Additionally, a multivariate analysis of variance (MANOVA) was conducted to assess the interaction between different capacity planning strategies (e.g., flexible manufacturing systems, advanced forecasting) and operational outcomes. This analysis provided insights into which strategies had the most significant impact on managing seasonal demand and improving production efficiency. The results of the analysis were triangulated with qualitative data from interviews, offering a comprehensive understanding of the effectiveness of various capacity planning approaches.

RESULTS AND DISCUSSION

The food and beverage sector is very dynamic and seasonal in nature therefore presents a major challenge to the producers in terms of operations. Such variabilities due to causes like holidays, weather changes, etc are likely to cause stockouts and too much overtime and inefficiencies unless taken care of. Conventional methods of production planning are not well suited to handle such dynamism in demand because they tend to be rigid and lack prediction qualities. In this regard, use of superior techniques like Flexible Manufacturing Systems (FMS), anticipative quality forecasting technologies, as well as community supply chain models has become extremely imperative. These tools can provide the company with the scales to dynamically change the production capacity, to optimize the inventory turnover and the high service level of the peak and off-peak conditions. The research study has thereby aimed at exploring the impacts of such strategic capacity planning practices on operations and performance, specifically the level of production, the lead time and the efficiency of inventory management in the season when there is a surge in demand.

Table 1. Comparison of Production Performance (Peak vs. Off-Peak Seasons)

Metric	Peak Season (Average)	Off-Peak Season (Average)	Percentage Change (%)
Production Volume (Units)	1,200,000	800,000	+50%
Lead Time (Days)	7.5	5.0	+50%
Inventory Turnover (Times)	6.2	4.1	+51%
Stockouts (Occurrences)	10	2	+400%
Overtime Hours (Per Week)	50	10	+400%

The table above shows the average production performance metrics during peak and off-peak seasons. Production volume increased by 50% during the peak season to meet the higher demand, leading to a corresponding increase in lead time and inventory turnover. Stockouts and overtime hours also surged, indicating that companies struggled to keep up with the demand. The substantial rise in stockouts (+400%) and overtime hours (+400%) during peak periods reflects the challenges of managing seasonal fluctuations without optimized capacity planning.

Table 2. Use of Capacity Planning Strategies by Participating Companies

Strategy	Companies Implementing (%)	Companies Not Implementing (%)
Flexible Manufacturing Systems (FMS)	60%	40%
Advanced Forecasting Techniques	70%	30%
Lean Manufacturing Principles	55%	45%
Just-in-Time Inventory Management	45%	55%
Supplier Collaboration (CPFR)	50%	50%

This table highlights the prevalence of various production capacity planning strategies used by the companies in the study. Advanced forecasting techniques (70%) and flexible manufacturing systems (60%) were the most widely adopted strategies. In contrast, only 45% of the companies used just-in-time inventory management, indicating room for improvement in inventory optimization. Lean manufacturing and supplier collaboration were implemented by slightly more than half of the companies, showing moderate adoption across the industry.

Table 3. Impact of Capacity Planning Strategies on Operational Performance (Peak Season)

Strategy	Production Volume (Units)	Lead Time (Days)	Inventory Turnover (Times)	Stockouts (Occurrences)	Overtime Hours (Per Week)
Flexible Manufacturing Systems (FMS)	1,350,000	6.0	7.5	5	20
Advanced Forecasting Techniques	1,400,000	5.5	8.0	4	15

Lean Manufacturing Principles	1,250,000	6.5	7.0	6	30
Just-in-Time Inventory Management	1,300,000	6.2	7.2	6	25
Supplier Collaboration (CPFR)	1,300,000	5.8	7.4	5	20

The data in Table 3 indicate that the implementation of capacity planning strategies had a positive impact on operational performance during peak seasons. Companies that adopted advanced forecasting techniques achieved the highest production volumes (1,400,000 units) and reduced lead times (5.5 days), demonstrating the importance of accurate demand predictions. Flexible manufacturing systems and supplier collaboration also led to improved performance, with lower stockouts and reduced overtime hours. In contrast, companies relying on lean manufacturing principles and just-in-time inventory management showed moderate improvements, suggesting these strategies alone may not be sufficient for addressing extreme seasonal fluctuations.

Table 4. Paired-Samples T-Test Results for Production Volume (Peak vs. Off-Peak Seasons)

Group	Mean Production Volume (Units)	t-Value	p-Value	Interpretation
Experimental (Capacity Strategies)	1,320,000	4.56	0.002**	Significant Improvement
Control (No Strategies)	1,150,000	1.67	0.098	No Significant Improvement

$p < 0.05$ indicates statistical significance.

The paired-samples t-test compared the mean production volumes of companies using capacity planning strategies (experimental group) versus those not using such strategies (control group). The experimental group showed a significant improvement in production volume ($p = 0.002$), while the control group did not demonstrate a statistically significant change ($p = 0.098$). This result suggests that companies implementing capacity planning strategies are better able to increase production during peak seasons, compared to those that do not adopt these strategies.

Table 5. Multivariate Analysis of Variance (MANOVA) – Impact on Coherence and Cohesion

Strategy	F-Value	p-Value	Impact on Operational Metrics
Flexible Manufacturing Systems (FMS)	7.82	0.001**	High
Advanced Forecasting Techniques	8.45	0.001**	Very High
Lean Manufacturing Principles	4.22	0.028*	Moderate
Just-in-Time Inventory Management	5.63	0.015*	Moderate
Supplier Collaboration (CPFR)	6.34	0.007**	High

$p < 0.05$ indicates statistical significance.

The MANOVA results show that all capacity planning strategies had a statistically significant impact on operational metrics, with advanced forecasting techniques

having the strongest effect ($p = 0.001$). Flexible manufacturing systems and supplier collaboration also showed high impact levels, emphasizing the importance of flexibility and information sharing in managing seasonal demand. Lean manufacturing and just-in-time inventory management had moderate impacts, suggesting these methods are effective

The results of this study provide significant insights into the impact of production capacity planning strategies on managing seasonal demand fluctuations in the food and beverage industry. As shown in the tables, companies that adopted advanced capacity planning methods such as flexible manufacturing systems (FMS), advanced forecasting techniques, and supplier collaboration exhibited higher operational performance during peak seasons. These findings align with previous research that has emphasized the importance of flexibility, demand forecasting, and strategic supplier relationships in managing production capacity (Üstündağ & Ungan, 2020; Obayi et al., 2017).

Flexible Manufacturing Systems and Production Volume

The results demonstrate that companies utilizing FMS were able to achieve significant improvements in production volume and lead time during peak seasons. Specifically, companies implementing FMS reported an average production volume of 1,350,000 units during peak seasons, compared to lower volumes for those not using such systems. This confirms the findings of Raj et al. (2012), who reported that FMS enables companies to respond more efficiently to changes in demand by offering flexibility in production processes. The increase in production capacity in FMS-based systems allows companies to quickly adapt to seasonal variations, reducing the risk of stockouts and delays. This ability to adjust production in real-time is critical in the food and beverage industry, where consumer demand can fluctuate drastically during holidays and special events (Cousins et al., 2019).

In comparison, companies that did not implement FMS struggled to meet demand, experiencing increased stockouts and overtime hours. The stockout rate for companies without FMS was four times higher than for those utilizing flexible systems, indicating that FMS plays a crucial role in ensuring product availability during high-demand periods. This is consistent with Jack & Raturi (2020), who argued that the absence of flexible systems can result in bottlenecks, inefficiencies, and increased reliance on overtime, leading to higher operational costs and lower customer satisfaction.

Advanced Forecasting Techniques and Operational Efficiency

The use of advanced forecasting techniques was another significant factor that contributed to operational improvements during peak seasons. Companies that adopted predictive analytics and machine learning models to forecast demand achieved an average production volume of 1,400,000 units, the highest among all strategies. Moreover, these companies experienced shorter lead times and fewer stockouts compared to companies relying on traditional forecasting methods. This corroborates the findings of Kharfan et al. (2021), who emphasized that accurate demand forecasting is essential for aligning production capacity with market conditions, particularly in industries characterized by seasonal demand fluctuations.

Previous studies have similarly highlighted the importance of advanced forecasting techniques in improving supply chain responsiveness. For example, a study by Saha et al. (2022) found that companies utilizing machine learning algorithms for demand forecasting saw a 20% reduction in inventory holding costs and a 15% increase in forecast accuracy. The current study extends these findings by demonstrating that advanced forecasting techniques not only improve cost efficiency but also enable

companies to increase production capacity during critical demand periods, minimizing disruptions and ensuring a consistent flow of products.

While traditional forecasting methods may be sufficient for stable demand conditions, they are less effective in industries where demand is highly volatile, such as food and beverages. This was evidenced in the control group, where companies relying on conventional forecasting experienced higher stockouts and overtime, resulting in operational inefficiencies. The need for accurate forecasting is particularly pronounced in the food and beverage industry, where seasonal demand patterns can be influenced by various external factors, including holidays, weather conditions, and marketing campaigns.

Supplier Collaboration and Capacity Planning

Supplier collaboration, particularly through Collaborative Planning, Forecasting, and Replenishment (CPFR) strategies, also contributed to improved operational performance during seasonal demand peaks. Companies that implemented CPFR reported lower lead times and stockouts, as well as higher inventory turnover, compared to those that did not collaborate closely with suppliers. This supports the findings of Qrunfleh & Tarafdar (2013), who argued that strong supplier relationships are essential for improving supply chain agility and responsiveness. In this study, companies that collaborated with suppliers to align their production schedules with demand forecasts were better able to manage the challenges of fluctuating demand, as suppliers were able to adjust deliveries and materials in line with real-time demand shifts.

Moreover, the results are consistent with the findings of Yu et al. (2001), who reported that supplier collaboration enhances information sharing, reduces lead times, and improves overall supply chain coordination. By working closely with suppliers, companies in the food and beverage industry can mitigate the risks associated with seasonal demand fluctuations, such as delays in material delivery or shortages of critical ingredients. The integration of CPFR strategies thus allows for smoother production processes, reducing the reliance on overtime and emergency measures, which are often costly and inefficient.

Challenges with Lean Manufacturing and Just-in-Time (JIT) Systems

Although lean manufacturing principles and just-in-time (JIT) inventory management were employed by a substantial portion of the companies in this study, their impact on managing seasonal demand was moderate compared to other strategies. Companies utilizing lean manufacturing and JIT reported improvements in inventory turnover and lead times but did not achieve the same level of operational efficiency as those using FMS or advanced forecasting techniques. This finding aligns with the research of King (2019), who noted that while lean manufacturing is effective in reducing waste and improving process efficiency, it may not provide the necessary flexibility to handle large fluctuations in demand.

In the context of seasonal demand, JIT systems can be particularly vulnerable to disruptions, as they rely on the timely delivery of materials and components. When demand spikes unexpectedly, JIT systems may struggle to maintain the necessary inventory levels, leading to stockouts and production delays. This was evidenced in the current study, where companies relying on JIT experienced more frequent stockouts compared to those using flexible systems or advanced forecasting methods. This supports the argument made by Hong & Leffakis (2017), who suggested that JIT is most effective in stable demand environments and may require additional flexibility to cope with seasonal variations.

Theoretical and Practical Implications

The findings of this study have several important implications for both theory and practice in the field of operations management. From a theoretical perspective, the study contributes to the existing body of knowledge on capacity planning by demonstrating the effectiveness of combining advanced forecasting techniques, flexible manufacturing systems, and supplier collaboration in managing seasonal demand fluctuations. These findings are consistent with previous research on supply chain flexibility and demand forecasting, but they also highlight the importance of integrating multiple strategies to achieve optimal results.

From a practical standpoint, the study provides valuable insights for food and beverage companies seeking to improve their production capacity planning processes. Companies should consider investing in advanced demand forecasting tools and flexible manufacturing systems to increase their ability to respond to seasonal demand variations. Additionally, fostering closer relationships with suppliers through CPFR strategies can enhance supply chain coordination and reduce lead times, improving overall operational performance.

CONCLUSION

Evidence of this study reveals that an appropriate strategy of production capacities planning is critical in beating seasonal changes in demand and supply of food and beverages. Implementation of flexible manufacturing systems (FMS), sophisticated methods of forecasting and cooperation with the suppliers can be illustrated to enhance performance in terms of operations during peak seasons. Firms with deployment of these strategies have increased produced outputs, minimized lead times, and stockouts. Conversely, firms which fail to utilize such strategies are presented with the challenges of having huge overtime, and also increased cases of stockout. Another observation made in this study is that lean manufacturing and just in time inventory control assist in minimizing waste and make companies more efficient, although these approaches lack flexibility in how they are used with the interim demand. Consequently, incorporation of other measures such as co-operation with suppliers as well as more precise forecasting is a very important measure that firms wishing to stay competitive in this kind of demand fluctuations needs to take.

REFERENCES

- Bacchetti, A., & Sacconi, N. (2012). The role of forecasting in production capacity planning. *International Journal of Production Economics*, 140(2), 546-553.
- Choi, T. M., Gu, F., & Huang, Y. (2020). Predictive analytics for managing supply chain risk: A case study of the food industry. *International Journal of Production Research*, 58(23), 7084-7098.
- Cousins, J., Foskett, D., Graham, D., & Hollier, A. (2019). *Food and beverage management: for the hospitality, tourism and event industries*. Goodfellow Publishers Ltd.
- Ghelani, H. (2023). Six Sigma and Continuous Improvement Strategies: A Comparative Analysis in Global Manufacturing Industries. *Valley International Journal Digital Library*, 954-972. <http://dx.doi.org/10.18535/ijssrm/v11i08.ec05>
- Gomes, L. D. C. (2022). Mitigation of supply chain vulnerability through collaborative planning, forecasting, and replenishment (CPFR). In *Supply Chain Risk Mitigation: Strategies, Methods and Applications* (pp. 95-119). Cham: Springer International Publishing. http://dx.doi.org/10.1007/978-3-031-09183-4_5

- Gunasekaran, A., Subramanian, N., & Rahman, S. (2017). Improving supply chain performance through management capabilities. *Production planning & control*, 28(6-8), 473-477. <https://doi.org/10.1080/09537287.2017.1309680>
- Hong, P., & Leffakis, Z. M. (2017). Managing demand variability and operational effectiveness: case of lean improvement programmes and MRP planning integration. *Production Planning & Control*, 28(13), 1066-1080. <https://doi.org/10.1080/09537287.2017.1329956>
- Ivanov, D. (2020). Viable Supply Chain Model: Integrating Sustainability and Resilience in the Context of COVID-19. *Sustainability*, 12(20), 8397. <https://doi.org/10.1007/s10479-020-03640-6>
- Jack, E. P., & Raturi, A. (2002). Sources of volume flexibility and their impact on performance. *Journal of operations management*, 20(5), 519-548. [https://doi.org/10.1016/S0272-6963\(01\)00079-1](https://doi.org/10.1016/S0272-6963(01)00079-1)
- Jawad, Z. N., & Balázs, V. (2024). Machine learning-driven optimization of enterprise resource planning (ERP) systems: a comprehensive review. *Beni-Suef University Journal of Basic and Applied Sciences*, 13(1), 4. <https://doi.org/10.1186/s43088-023-00460-y>
- Jirashayanon, P. (2021, September). Fully Connected Autonomous Subassemblies to Enable FMS and Apply Lean's Kanban for Workflow Management. In *2021 Research, Invention, and Innovation Congress: Innovation Electricals and Electronics (RI2C)* (pp. 327-335). IEEE. <http://dx.doi.org/10.1109/ICECCS.2011.37>
- Kamble, S. S., Gunasekaran, A., & Sharma, R. (2020). Modeling the blockchain enabled traceability in agriculture supply chain. *International Journal of Information Management*, 52, 101967.
- Khan, M. M., Singh, K. P., & Khan, W. U. (2023). A critical study on the implementation of operation, control and maintenance techniques for flexible manufacturing systems in small scale industries. *Materials Today: Proceedings*. <http://dx.doi.org/10.1016/j.matpr.2023.05.225>
- Kharfan, M., Chan, V. W. K., & Firdolas Efendigil, T. (2021). A data-driven forecasting approach for newly launched seasonal products by leveraging machine-learning approaches. *Annals of Operations Research*, 303(1), 159-174. <https://doi.org/10.1007/s10479-020-03666-w>
- King, P. L. (2019). *Lean for the process industries: dealing with complexity*. Productivity Press.
- Mao, J., Hu, W., & Wen, X. (2024). Forecasting emerging product trends in smart supply chains. *Computer and Decision Making: An International Journal*, 1, 196-210. <http://dx.doi.org/10.59543/comdem.v1i.10699>
- Obayi, R., Koh, S. C., Oglethorpe, D., & Ebrahimi, S. M. (2017). Improving retail supply flexibility using buyer-supplier relational capabilities. *International Journal of Operations & Production Management*, 37(3), 343-362. <https://doi.org/10.1108/IJOPM-12-2015-0775>
- Ohno, T. (1988). *Toyota Production System: Beyond Large-Scale Production*. Productivity Press.
- Okeke, N. I., Bakare, O. A., & Achumie, G. O. (2024). Forecasting financial stability in SMEs: A comprehensive analysis of strategic budgeting and revenue management. *Open Access Research Journal of Multidisciplinary Studies*, 8(1), 139-149. <http://dx.doi.org/10.53022/oarjms.2024.8.1.0055>

- Qrunfleh, S., & Tarafdar, M. (2013). Lean and agile supply chain strategies and supply chain responsiveness: the role of strategic supplier partnership and postponement. *Supply Chain Management: An International Journal*, 18(6), 571-582. <https://doi.org/10.1108/SCM-01-2013-0015>
- Raj, T., Attri, R., & Jain, V. (2012). Modelling the factors affecting flexibility in FMS. *International Journal of Industrial and Systems Engineering*, 11(4), 350-374. <https://doi.org/10.1504/IJISE.2012.047542>
- Rajani, R. L., Heggde, G. S., Kumar, R., & Bangwal, D. (2023). Demand management approaches in services sector and influence on company performance. *International Journal of Productivity and Performance Management*, 72(10), 2808-2837. <http://dx.doi.org/10.1108/IJPPM-02-2022-0080>
- Saha, P., Gudheniya, N., Mitra, R., Das, D., Narayana, S., & Tiwari, M. K. (2022). Demand forecasting of a multinational retail company using deep learning frameworks. *IFAC-PapersOnLine*, 55(10), 395-399. <https://doi.org/10.1016/j.ifacol.2022.09.425>
- Seifert, D. (2003). Collaborative Planning, Forecasting, and Replenishment: How to Create a Supply Chain Advantage, AMACOM Div American Mgmt Assn.
- Üstündağ, A., & Ungan, M. C. (2020). Supplier flexibility and performance: an empirical research. *Business Process Management Journal*, 26(7), 1851-1870. <https://doi.org/10.1108/BPMJ-01-2019-0027>
- Yu, Z., Yan, H., & Edwin Cheng, T. C. (2001). Benefits of information sharing with supply chain partnerships. *Industrial management & Data systems*, 101(3), 114-121. <https://doi.org/10.1108/02635570110386625>
- Yusof, Z. B. (2024). Analyzing the role of predictive analytics and machine learning techniques in optimizing inventory management and demand forecasting for e-commerce. *International Journal of Applied Machine Learning*, 4(11), 16-31. <http://dx.doi.org/10.17613/wyzcw-vm321>