

Impact of Data Analysis on Financial Performance in Small and Medium Manufacturing Companies

Dinesh Junghare¹ and Sachin Lad¹

¹MIT Art, Design & Technology University, Pune, India

Abstract. The upward-trending digital environment notwithstanding, small- and medium-sized enterprises (SMEs) operating in the manufacturing field are increasingly considering the strategic importance of data analytics to boosting their financial performance. The review provides a conceptual reconsideration of the interconnection between data-analytic adoption and the financial performance of firms. It questions the ability of predictive analytics, business intelligence, and live data processing to enable a greater quality in decision-making, increased cost-effectiveness, improved financial planning, and increased profitability. Based on the recent research conducted in the period between 2014 and 2024, the paper outlines the key financial metrics that are affected by the analytics, especially the return on investment (ROI), profit margins, and operational productivity. According to the empirically proven benefits, however, SMEs still face implementation limitations like the lack of technical infrastructure, poor human resources, challenging data integration, and cybersecurity threats. As a result, some promising future routes are noted in the review, such as the introduction of low-cost analytics or a focus on developing digital skills and the development of policy frameworks that would encompass the needs of SMEs. It concludes that data analytics, when strategically aligned with business goals, offers a powerful pathway for SMEs to achieve financial resilience, competitiveness, and sustainable growth in an increasingly data-driven economy.

1 Introduction

Small and Medium Enterprises (SMEs) are widely recognized as engines of economic growth, employment, and innovation. Within the manufacturing sector, SMEs contribute significantly to industrial productivity, regional development, and export performance. For example, in India, SMEs engaged in manufacturing are responsible for approximately 45% of total industrial output and 40% of the nation's exports [1]. Their role in job creation and localized economic advancement makes them vital to both emerging and developed economies.

The production of small- and medium-sized businesses (SMEs) represents a key industry in the modern economy across the world; however, these companies are faced with barriers

constantly, particularly in profits. Factors, such as the rising input prices, ineffective supply chain models, a high market volatility level, limited access to funding, and the overall competition standards of the bigger organizations, potentially halt its development and sustainability in the long term.

In the recent past, however, a trend that can be considered as very crucial in improving the financial performance is the integration of data analytics into the fundamental organizational processes. Data analytics allows firms to draw practical data-driven inferences on structured and unstructured data and therefore enables them to perform better at forecasting, regulating cost expenditures, and distributing resources more efficiently [2]. The most important factor relating to the methodology is that data is converted into a strategic value used as the foundation of evidence-based decision-making, which makes it less reliant on managerial insight.

With the emergence of Industry 4.0, the diffusion of such novel technologies as big data analytics, artificial intelligence (AI), cloud computing, and predictive modeling has been further expanded. These tools have become increasingly affordable to SMEs because their costs have reduced, and increased cloud-based platforms have become available. Sony recommend that the adoption of these technologies allows the manufacturing companies to digitalize and optimize their value chain to increase the level of productivity and improve financial performance [3]. Also note that data-driven business decision-making improves the strategic alignment, operational reactivity, and profitability.

Even though it can be noted that data analytics is steadily gaining ground in various industries, its proliferation is when applied to small- and medium-sized enterprises (SMEs). The lack of digital literacy of numerous firms, high implementation expenses, the threat of increased cybersecurity issues, the absence of relevant in-house expertise, and not always positive management support are among the challenges firms face [7]. Therefore, some SMEs have shifted to data-oriented enterprises, and others are lagging as their financial competitiveness suffers due to the old ways of doing things. This has given rise to a strong digital divide in the industry, where organizations that deploy analytics gain a distinct competitive advantage in comparison to those that do not.

Considering such dynamics, an intensive synthesis of the current literature is required to clarify the effect of data analytics on the financial performance of manufacturing SMEs. A systematic review can not only map the pros and cons regarding the implementation but also deduce trends regarding implementation, discern the effective instruments and practices, and suggest practice- and policy-focused interventions.

In this paper, five main sections have been established. The former gives a description of the introduction and the background of the study and is then followed by a thematic review of the literature that covers adoption, financial prospects, predictive forecasts, competitive advantage, and impediments to implementation. The third section includes future directions, as well as the formulation of practical recommendations to stakeholders. The fourth synthesizes the main result and contributions of the review, whereas the fifth provides a comprehensive list of works that were used, making assurances of a high academic standard and making ease of research achievable.

1.1 Objectives of the Study

This study aims to review the existing literature to understand how data analytics impacts financial performance in manufacturing SMEs. The specific objectives are:

- To review the role of data analytics in improving financial performance of SMEs

- To explore how SMEs are adopting data-driven practices for cost control and forecasting
- To identify barriers and enablers for successful analytics adoption
- To assess the strategic advantage gained through digital innovation and analytics

1.2 Research Methodology

In strengthening the rigor of this review, a systematic approach was applied for the identified databases. A total of 51 studies were taken, of which 33 met the final inclusion criteria. Studies were included if they: (a) were published between 2014 and 2024; (b) appeared in peer reviewed journals, conference proceedings, or credible industry reports; (c) were written in English; and (d) focused on small and medium manufacturing enterprises or provided transferable insights relevant to SMME analytics adoption. Articles were excluded if they focused solely on large enterprises, discussed non-manufacturing contexts without applicable to SMMEs, lacked empirical or conceptual depth, or were duplicated across databases. This systematic procedure ensured adherence with the thematic area —barriers, enablers, technological readiness, and environmental factors—and aligned with the paper’s qualitative, analytical approach. The current study applies content analysis and identifies and explains the repetitive trends of the scholarly literature with a particular emphasis on adoption patterns, affective dimensions, and hindrances to implementation. The referenced data were accessed in Google Scholar, ScienceDirect, Scopus, and Web of Science. The keywords included data analytics, financial performance, SMEs, manufacturing, Industry 4.0, predictive analytics, big data, and digital transformation. The findings were staged in the form of thematic synthesis, including four interrelated areas: (a) data analytics adoption, (b) financial performance outcomes, (c) predictive forecasting tools, (d) implementation challenges. This framework makes the analysis systematic and comprehensive, which would provide information to both researchers and practitioners. The general aim of the study is of an academic nature and of ensuring that it can direct small and medium enterprises in the way of utilizing data analytics to produce greater financial results.

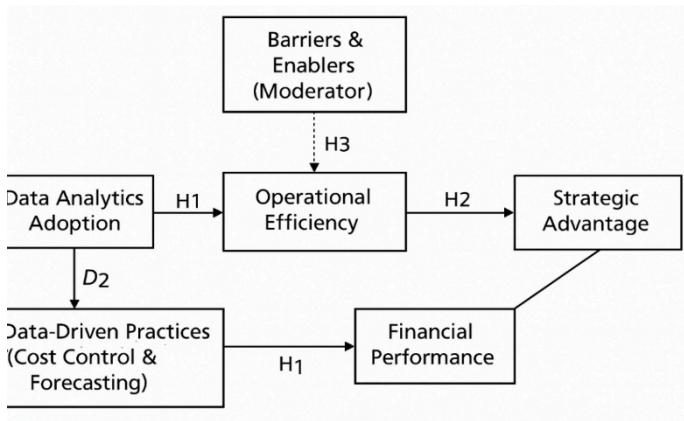


Fig. 1: Conceptual Model, Source: Self-Compiled, Primary Resource

2 Thematic Review: Data Analytics and Financial Performance in SMEs

2.1 Understanding Data Analytics and Financial Metrics in SMEs

In the modern-day digital age, data analytics emerge as one of the pillars in the company strategy, especially regarding small and medium-sized companies (SMEs) working in the manufacturing industry. The results of empirical studies prove that data analytics enables companies to draw inferences out of past and current information, which enhances financial planning and optimization of resources [2, 3, 4]. Generally speaking, there are three major types of data analytics: descriptive analytics helps process the data about the past performance and trends, predictive analytics uses statistical models and machine learning to predict how the financial or operational result may look in the future, and prescriptive analytics advises the reader on the data-driven decision, taking the flow of various financial situations into consideration [5]. Such analysis methods assist SMEs to track financial performance, forecast cash balances, and model cost controls.

An analogous term to data analytics is business intelligence (BI), which refers to the application of systems and tools that convert raw information into useful information that can be used to make strategic decisions. Although, as explained earlier, the traditional reporting system in SMEs makes use of fixed financial reports and spreadsheets, contemporary BI systems enable the use of live dashboards and instant KPIs, letting decision-makers react quickly to the changing rates of profitability, costs, and operational results [6, 7]. BI systems also increase organizational agility through integration of the functions that are mutually dependent in manufacturing enterprises, such as finance, sales, production, and logistics.

Financial performance is considered to be the primary scale of organizational assessment, and the available literature highlights the value of a variety of major indicators—profit margin, return on investment (ROI), cost efficiency, working capital ratio, and cash flow stability—with the assistance of which it is possible to evaluate operational success and investment effectiveness. In terms of small- and medium-sized enterprises (SMEs) manufacturing demands, these variables are proxies to outcome variables, as well as managerial input instruments of the decision-making process regarding price policy, the maintenance of trade stocks, investments, and the development of expansion strategies. Transferring the advanced analytics to financial datasets also adds polish to these indicators as it generates empirically based information, which agrees on strategic alignment and control measures. Using empirical evidence, it is possible to conclude that a reasonable use of data-analytic techniques in financial reporting is that it promotes transparency, reduces the rates of error, and enables regulatory compliance—a fact that is of special interest among SMEs that have to operate in a competitive environment where resources are limited. Accordingly, the professional control of the traditional and advanced reporting systems and the more purposeful synthesis of descriptive, predictive, and prescriptive analytics by means of the business intelligence (BI) tools are a must, to name but a few, considering actors, such as SMEs, that aim at the modernization of their financial management.

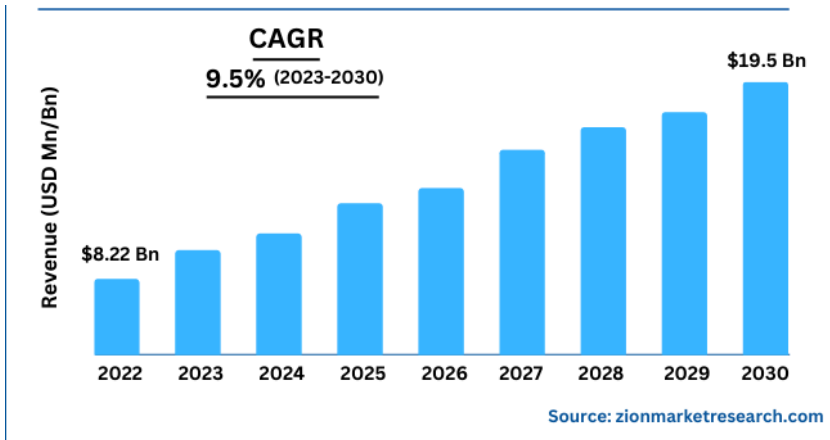


Fig. 2: Global Financial Analytics Market Size (USD Billion, 2022–2030), Source: zionmarketresearch.com

According to the report published by Zion Market Research, the global financial analytics market is forecasted to rise significantly—from USD 8.22 billion in 2022 to USD 19.5 billion in 2030, with a steady CAGR of 9.5% (2023–2030). This global growth trajectory underscores the rising importance of data-driven financial tools in modern business contexts, including SMEs.

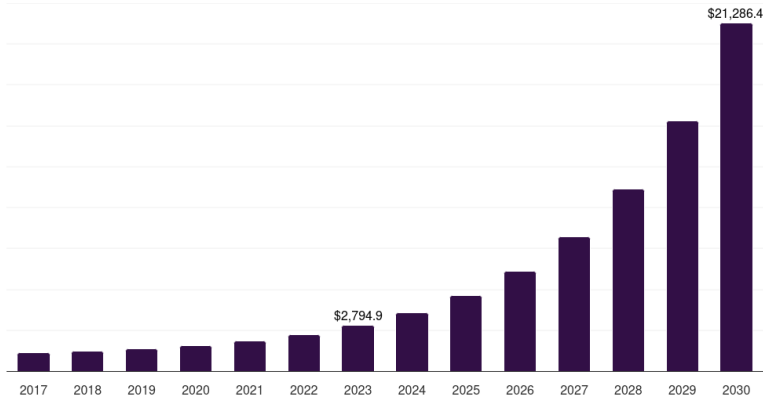


Fig. 3: India Data Analytics Market Size (USD Million, 2017–2030), Source: India Data Analytics Market Data Book

In contrast, the India data analytics market presents a much steeper growth curve. Valued at USD 3,551.8 million in 2024, it is expected to surge to USD 21,286.4 million by 2030, reflecting a CAGR of 35.8% (2025–2030). Within this context, predictive analytics was the leading segment by revenue in 2024, while prescriptive analytics is forecast to be the fastest-growing segment in the near future.

These trends affirm the crucial role of analytics in SMEs' financial strategy—particularly in emerging markets like India. The adoption of real-time, intelligent systems not only strengthens internal financial management but also prepares SMEs for compliance, agility, and competitive advantage in data-intensive economies.

2.2 Adoption of Data Analytics in Manufacturing SMEs

In modern research, there is a greater focus on data-analytic integration in small and medium manufacturing enterprises (SMEs) as the higher level of digitalization and the pressure to increase the financial and operational performance stimulate this trend. In empirical studies, it is observed that big business has historically attained dominance in regard to adopting analytics, but SMEs are increasingly recognizing the strategic importance of analyzing data in promoting competitiveness [8]. In the case of manufacturing in particular, analytic tooling is implemented to track the effectiveness of production, streamline supply-chain processes, and enhance quality control, which has a direct impact on cutting costs and making a profit [9].

Nonetheless, the rate of adoption of analytics is quite volatile. Factors like digital preparedness, management perception, and external facilitation all play a role in uptake. The notion of Industry 4.0 has contributed to increased adoption rates, as it brings with it such technologies as AI, IoT, machine learning, and real-time data processing, which are used to support decision-making at all levels of a production corporation. Cloud-based analytics solutions delivered also minimized the barriers to SMEs since even resource-constrained companies could use their highly sophisticated features without investing significant capital. A limited use of data-analytic adoption with SMEs is, however, still a reality since there are a variety of issues and limitations, both internally motivated and externally motivated.

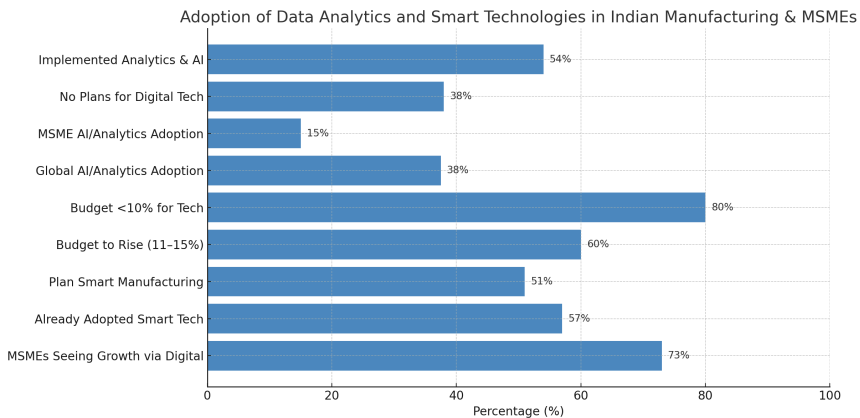


Fig. 4: Adoption of Data Analytics in Indian Manufacturing & MSMEs, Source: Data compiled from PwC India, Analytics India Magazine, ETCFO, IPF Online, The Economic Times

The chart highlights a mixed trend in data analytics adoption among Indian manufacturing and MSMEs. While 54% of manufacturers have adopted AI/analytics, 38% still lack digital plans. Only 15% of MSMEs have adopted such technologies, far below global levels. Budget constraints remain a challenge, though planned investments are rising. Encouragingly, 51% aim to adopt smart manufacturing, and 73% of rural MSMEs report growth through digital tools. Despite progress, financial and technical barriers continue to limit widespread adoption.

Availability of finance always appears as one of the major barriers. The majority of SMEs are characterized by a limited budget and consequently do not want to invest their resources in data infrastructure or analytical software when it is not clear how to measure their short-term payback [9]. Also, the inadequacy of qualified people that can analyse the outcomes of data and convert them into working strategies is also being developed as

another critical problem [10]. As a result of this technical expertise shortage, it is the case that current tools are not used at full capacity or that older decision-making methods are maintained.

The key to the small- and medium-sized enterprises (SMEs) adoption of data on a large scale is the cultural and organizational determinants, in addition to technological and economical determinants, that pose decisive obstacles to large-scale adoption of data within the SMEs. The level of resistance to managerial innovation is often present despite the ease of accommodating the perceived benefits of data utilization [11]. Adding to this, the issues of data security, integration complexities, and regulatory compliance, especially in the fields that handle proprietary or sensitive data, remain a major roadblock to the widespread adoption [6].

At the same time, these barriers are slowly being alleviated by external drivers like government-driven digital programs, training provided by the vendors, and collaboration with industry-wide players. As an example, policy structures in countries such as Germany and Singapore have granted monetary support and knowledge subsidies to enable digital transformation within SMEs [11]. In addition, the use of pilot-based implementation and modular analytics tools supporting the specific needs of SMEs has shown success in making the process not only more confident but also more skilful when it comes to data analytics utilization [8]. On the whole, it can be stated that adoption is a continuous process, manufacturing SMEs increasingly turn to data analytics as a tool that helps to advance financially and on operations.

2.3 Link Between Data Analytics and Financial Performance

The relationship between data analytics and financial performance of the small and medium manufacturing enterprises (SMEs) has attracted more academic and industry attention. Empirical investigations have shown that integration of analytics into central business processes is a relevant step that significantly increases the ability of a firm to improve profitability, optimize resource utilization, and maximize return on an investment. In the manufacturing industry, in the case of SMEs, analytics can help them have more precise control of their inventory, predicting demand and production, which is generating fewer operation expenses and more financial stability.

There is empirical evidence that when financial reporting and budgeting are performed in real-time data integration, rapid decision-making is realized and there are no drawbacks to using out-of-date or inaccurate information. Through the employment of predictive models, companies may anticipate the changes in customer demand, adjust production plans, and avoid the accumulation of unnecessary stores, which are the main trackers of cost-saving [12]. Machine learning and time-series analysis enable SMEs to forecast finances and calculate the cash flow and investment requirements to reduce liquidity risks and the cost of financing [13].

The positive returns yielded by the implementation of an analytics-based strategy by SMEs have been supported by case-study evidence. Analytics are used by manufacturing enterprises to implement capital-allocation decisions that yield high-quality project selection and funding prioritization, which translates into better financial performance. Moreover, the implementation of analytics in strategic planning allows SMEs to monitor their profit margins, overhead ratios, and unit costs more efficiently and offer the management the responses to act on.

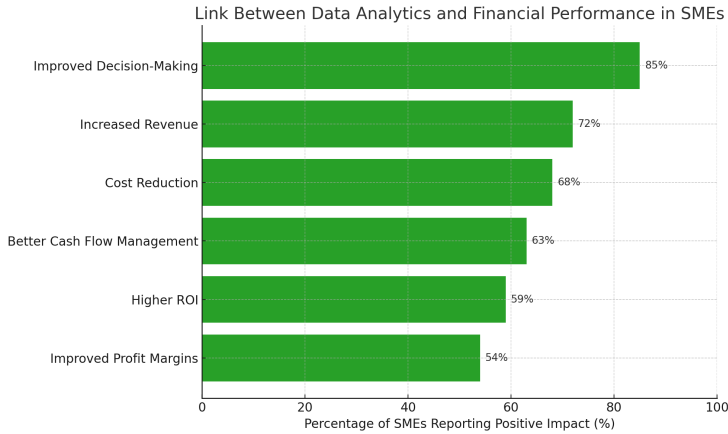


Fig. 5: Link Between Data Analytics and Financial Performance in SMEs., Source: Adapted from insights reported by PwC India (2024), Analytics India Magazine, and Deloitte SME Analytics Report (2023).

The chart demonstrates that data analytics significantly contributes to financial performance in SMEs. A large majority (85%) report improved decision-making, followed by increased revenue (72%) and cost reduction (68%). More than half of the firms also saw gains in cash flow management, ROI, and profit margins, highlighting the strong positive impact of analytics on financial outcomes.

Empirical studies show that the application of data analytics in the manufacturing of small and medium enterprises (SMEs) can improve financial performance regarding the alignment of operations. Companies that coordinate their financial objectives with operational measures—including uptime of machinery, defects, and delays in the supply chain—have been able to record significant enhancements in productivity and waste minimization, thus boosting bottom-line performance. Besides, the companies that use analytics to track the customer buying behaviour and sales rates will be able to optimize the pricing, as well as the marketing investment, which will have a direct impact on increasing revenues [14].

Nevertheless, existing literature points out that the financial drama of data analytics will depend on the success with which the organization will integrate it into the decision-making culture. Without proper alignments of the strategy or effective data governance, their mediocre performance can continue, notwithstanding the presence of technological tools. As a result, enduring financial returns are possibly higher when analytics is not only viewed as a technical tool but rather as an ongoing organizational capability built within the existing systems of performance management.

Summing up, there is strong evidence that data analytics encourages a positive financial performance of manufacturing SMEs by enabling precise forecasting, cost cuts, better capital choices, and maximization of revenue, although the aforementioned effect is dependent on the degree of strategic fit, the integration of operational silos, and organizational data literacy.

2.4 Predictive Analytics and Financial Forecasting

Predictive analytics has been regarded as a determinative tool of strategic and financial planning among small and medium manufacturing enterprises (SMEs) whose ultimate goal is to increase forecasting accuracy and reduce financial risk. The method combines the tools

of statistical modelling, machine-learning algorithms, and data-mining methods to explain any future trend, behaviour, and outcome by using both historic and real-time data. Such an application of predictive analytics to SMEs promotes a proactive attitude to decision-making by enabling the firms to anticipate changes in their cash flows, to optimize investment strategies, and to manage their resources in a more prudent way.

The most common methods that are implemented in predictive analytics include regression analysis, time-series forecasting, and classification models, which are becoming a part of cloud-based financial-planning platforms and enterprise resource planning systems, also known as ERPs. Time-series forecasting has especially been widely applied to revenue forecasting and demand estimation, as well as inventory control, which directly impacts working capital and profitability. In the case of manufacturing SMEs, the predictive analytics will be able to maintain an ongoing record on sales patterns, materials consumption, and supply chain variability, which will help in better budgeting and buying decisions.

Another area that predictive analytics can produce tangible value is managing cash flow. Due to the stringent liquidity conditions of SMEs, the propensity to predict contingencies of receivables, payables, and short-term liabilities in terms of ability may significantly reduce the risk of insolvency or credit default. With the help of machine learning, companies will be in a position to identify seasonal fluctuations, identify late payments, and predict financial needs, which can make it easier to coordinate activities with the lenders and financial firms.

The predictive model is another critical process that can be used in making investment decisions and establishing risk reduction efforts. To illustrate, risk-scoring algorithms can be used by managers to assess the financial solvency of potential clients or suppliers, whereas credit-scoring frameworks allow assessing borrowers, credit lines, and terms thereof [15]. Predictive instruments are also essential in capital-budgeting scenarios and simulations of returns on investment that reduce the degree of speculation and chart the extent of data-driven expansion. Such abilities are especially relevant to small- and medium-sized enterprises (SMEs) that do not enjoy sufficient financial reserves to absorb high-risk ventures.

Similarly, predictive analytics reinforces strategic agility as it allows the companies to simulate the effects of internal adjustments (e.g., cost-structure fluctuations or workforce optimization) and the external environment (e.g., inflation, demand fluctuations, supplier delays) on subsequent financial performance. As a result, the manufacturing SMEs will be in a position to enhance their resilience to facing the volatility and discontinuous shocks in the market, such as the economic downturn and the supply-chain bottlenecks.

Despite such merits, the adoption of predictive analytics among the SMEs is at an embryonic stage. Issues like poor data quality, lack of training, and limited compute infrastructure remain limiting factors. However, these divides are being decreased by advancements in customer-friendly analytics tools and the multiple features of the technology associated with them that link to online financial applications, making them more available and less expensive for SMEs to use.

2.5 Challenges in Implementation

The literature also observed contradict views regarding the use of modern technologies. Several scholars, , claim that cloud-based platforms, AI tools, and open-source analytics have significantly reduced adoption barriers, making advanced analytics less costly for smaller firms. However, empirical work [9] challenges this, showing that many

SMEs—especially in rural or resource constrained regions—still lack the digital infrastructure, high-speed connectivity, or technical support needed to capitalize on these technologies. This highlights a gap between theoretical technological accessibility and practical technological usability.

Across the literature, several critical research gaps become visible. First, most existing studies focus on urban or digitally mature SMEs, leaving rural and micro manufacturers significantly low presence despite their unique structural and challenges. Second, the majority of studies employ cross-sectional designs, offering static snapshots of adoption barriers rather than tracing how capabilities evolve over time. Longitudinal studies are needed to better understand how learning, maturity, and technology integration progress within SMMEs. Third, “soft factors” such as leadership styles, employee attitudes, resistance to change, and psychological readiness remain underexplored despite strong evidence suggesting that cultural barriers often outweigh technological ones. Finally, most existing frameworks adopt generic SME models rather than sector-specific approaches; manufacturing subsectors such as automotive components, precision engineering, and agro processing require tailored analytics adoption pathways.

In summary, while existing literature strongly agrees on the strategic value of data analytics for SMMEs, it reveals substantial differences in how barriers and enablers are prioritized. Furthermore, contradictions surrounding technology accessibility and organizational readiness point to the need for more nuanced, context sensitive research. The gaps identified—particularly those related to rural inclusion, longitudinal insights, soft factor measurement, and subsector-specific frameworks—outline important avenues for future empirical and theoretical exploration.

The problem of functional integration of data: in many cases companies cannot integrate and standardize data on finance, production, logistics, and sales functions. Historical systems, paper-based records, and diverse data structures create challenges to the creation of a coherent analytics framework. As a result, the quality of data is usually undermined, redundancy runs wild, and timely decision-making suffers.

The lack of skilled people. However, data scientists, business analysts, the ability to analyse statistical data are not present in many SMEs [14]. In comparison to big-sized corporations, such organizations are rarely able to provide the required funds to employ additional analytics specialists or to offer long-lasting training sessions. The generated human capital gap hinders the commercial capacity of SMEs to plan, comprehend, and utilize data-informed information in strategic programs [4]. In such a situation, when suitable analytics tools are available, they are not used by staff adequately because of their low technical skills.

Technical infrastructure in itself is another barrier. Most of the effective data analytics solutions often need contemporary IT environments, cloud computing functionalities, and real-time processing systems. However, numerous SMEs utilize obsolete software and hardware setups, which are not compatible enough with superior analytics systems [2]. In addition, poor internet connectivity and lack of adequate investments in cybersecurity predispose such enterprises to exposures that discourage use of digital tools, especially in the rural or semi-urban areas.

Over the past years, the issue of cybersecurity and data privacy has become more pronounced, as companies store and analyze quite sensitive data of their clients, operations, and finances. Such amassed data increases the chances of data theft and breach of regulatory compliance. Still, a very large share of SMEs is unaware of data governance models and do not have the internal organization on a level to meet the provisions of the instruments, the General Data Protection Regulation, or the Digital Personal Data

Protection Act of India. This shortcoming often discourages these organizations from increasing their analytics abilities.

Besides, complications facing the quantification of the return on investments (ROI) usually wear out support of analytics programs within the interior. In SMEs, decision-makers are usually interested in short-term outcomes; thus, it is difficult to prove intangible or long-term values of analytics to financial assets. There is no direct connection between analytics and bottom-line performance, and therefore, many companies do not want to invest significant resources in implementation.

Therefore, despite the possibilities of revolutionizing manufacturing SMEs through data analytics, its implementation is limited by organizational, technological, financial, and cultural constraints. Such limitations require more than tech enhancement but require management attitude, customized skill acquisition, and proper strategic alignment of IT and business operations.

3 Future Directions and Practical Implications

The modern digital environment gives manufacturing SMEs a chance to cross the traditional data-related barriers and integrate the analytics solutions into their activities more efficiently. One thing that can be specially noted is the rise of cheap, accessible analytics services that are specifically oriented towards small businesses. APIs and solutions like Microsoft Power BI, Google Data Studio, and open-source solutions like KNIME and R provide cost-effective entry points that these companies can use to kick-start data-transformation projects without incurring excessive financial cost.

In line with that, training, as well as building up digital skills, is also an essential area of study focus that should be conducted in the future. Many SMEs are still faced with the limitations of not having adequate analytical skills in their departments. Sprint upskilling opportunities and vocational training, along with strategic alliances with higher educational establishments, can equip personnel with industry basics before getting transfers to another post or leaving it completely [4]. Management of data-driven company culture also helps to make sure that analytics acquisition is not just an artificial adoption but becomes an essential part of daily decision-making.

Policy and government intervention are essential breeders of adoption on a wider scale. The process of strategic investment into the digital infrastructure, including the extension of broadband services, cloud-access subsidies, and the development of solid cybersecurity systems, also allows SMEs in less developed areas to become active users of analytics technology. The shift towards a data-driven form of business could also be accomplished more promptly with the help of public and private initiatives in the form of consultative support or tax credit schemes in the context of digital transformation projects [11].

Besides, the demand is under a continuously expanding pattern in respect to analytics that are sector-specific and scalable. The manufacturing SMEs are very heterogeneous in terms of production methodologies, regulatory requirements, and complexity of operation. Their one-size-fits-all models are thus not very successful. Novel solutions to this shortcoming will need to be addressed in the future by offering flexible and customizable tools equal to the specific requirements of the respective industrial sector, precision engineering, textiles, or food processing [15]. These tailored tools should be scalable, ensuring that firms can gradually expand their analytics capabilities as their operations grow.

In essence, the path forward involves not just greater access to technology but also a deeper integration of analytics into the strategic DNA of manufacturing SMEs. With the right

ecosystem of tools, talent, and support, analytics can become a lasting source of financial and competitive strength for these enterprises.

4 Conclusion

The operations that are enabled by data analytics form a transformative paradigm for small- and medium-sized enterprises (SMEs) in the manufacturing industry. The current review shows that with the use of predictive modelling, real-time tracking, and business-intelligence processes, data analytics has a proven direct, positive impact on core financial figures, including profitability, ROI, cost containment, and efficiency of operation. According to literature findings, evidence has shown that data analytics tools are very useful in facilitating a switch to proactive decision-making, particularly when implemented successfully. Predictive modelling enables companies to anticipate changes in the market, allocate resources in an optimal manner, and plan budgets in the best way possible, thus reducing wastage, avoiding financial hazards, and being flexible enough to react to market fluctuations. With competition in the economy being more intense and more focused on digitization, such capabilities are essential to the survival and expansion of SMEs, especially within a resource-constrained manufacturing environment.

However, there are remaining obstacles that do not allow their complete achievement. Data infrastructures are weak, the early cost of investment is high, there is a lack of skills, there are still data silos, and organizational resistance to change is quite evident. Most SMEs still rely on the traditional systems and do not have the strategic frameworks to map out the analytics with the long-term financial goals. Moreover, the implementation is also difficult due to cybersecurity threats and data-privacy fears, especially in the industries that work with confidential production or client data.

Even so, the outlook on the future direction of analytics in SMEs is becoming more optimistic. Access to affordable tools that work in the cloud, advancement of educational programs, and policy-driven support of analytics are indicative of a gradual democratization of access to the technologies. However, with further integration of the Industry 4.0 applications, such as the Internet of Things (IoT), Artificial Intelligence (AI), and blockchain, into the manufacturing systems, the space to extend its existing performance through analytics would constantly be expanded.

The empirical conclusions of the paper at hand, through the pragmatic perspective, explain the necessity of a multi-stakeholder framework. Policymakers are encouraged to promote the development of digital infrastructures and introduce offers that will speed up the adoption of analytics. At the same time, technology suppliers should come up with ways to design technology in such a manner that it will suit the unique limitations and requirements of small- and medium-sized businesses (SMEs). Business leaders in deploying resources to institutionalize data-driven reasoning through all levels of an organization have an opportunity to develop internal skills that will support the long-run analytics integration.

In the future, research and investigation ought to shift to empirical studies, which aim at separating industry segments and comparing the long-term financial returns on the use of analytics. Comparative analysis of early adopters and laggards in particular manufacturing settings, the explanation of leadership management in digital change, and the creation of scalable models of analytics, applicable to different types of manufacturing SMEs, should be vital future research directions.

To conclude, data analytics will serve as a tremendous strategic driver of improving the financial standing of manufacturing SMEs. With the combined effort of joint effort,

conscious policy involvement, and continual innovation, it will be able to become an unavoidable support posture of future resiliency, competitiveness, and sustainable growth

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