

# Harnessing Digital Innovation for Sustainable Business Growth: A Comprehensive Framework for SMEs

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## ABSTRACT

Small and medium sized enterprises (SMEs) are at the heart of the rapidly evolving digital landscape; they are both at an unprecedented level of opportunity and at risk of not being able to achieve sustainable growth. In this paper, we present a comprehensive framework that leverages digital innovation as a transformative enabler enabling SMEs to navigate competitive markets while achieving sustainability goals. Based on understanding emerging technologies, business model innovation, and sustainability practices, the framework articulates this important interface of economic viability, environmental stewardship and social responsibility. It underlines the use of digital technologies, including artificial intelligence, blockchain and IoT, to optimize resource utilization, improve operational efficiency and the development of resilient value chains. Additionally, the study examines the effect of organizational culture, leadership and skill development on driving digital transformation, and the requirement for adaptive strategies targeted at SMEs' particular constraints and opportunities. This research, through a multi-disciplinary approach, identifies key enablers and barriers to digital innovation, which provide actionable insights to policy makers, industry stakeholders and SME leaders. The framework incorporates case studies and empirical evidence to show how SMEs can develop sustainable competitive advantages while at the same time contributing to broader societal goals such as reducing carbon footprints and promoting fair economic growth. Furthermore, this paper not only contributes to theoretical understanding of how SMEs can use digital technologies to drive innovation and sustainability but also offers practical guidance to SMEs to use digital technologies as a catalyzer for innovation and sustainability. The proposed framework is aimed to bridge the gap between digital transformation and sustainable development, while maintaining the SMEs as agile, relevant and impactful player in the digital economy.

**KEYWORDS:** Digital Innovation, Sustainable Business Growth, SMEs, Technology Adoption, Resource Optimization.

## 1. INTRODUCTION

For the economy and innovation, SMEs are a crucial component to the development of the world in the fast changing technological progress [1]. Generation of employment, fostering of innovation and local development are key functions of SMEs in global economic activity [2]. However, despite these enterprises' growing pace of digital transformation and heightened focus on sustainability, digital technology provides hope [3]. There is great potential for advanced technologies, such as artificial intelligence (AI), blockchain, the Internet of Things (IoT), and data analytics, to improve operational efficiency, optimize resource usage and foster innovation; however, many SMEs are struggling to maximize the benefits of these innovations [4]. And they often struggle to adopt these very transformative solutions because they lack resources, technological skill gaps, and a resistance to change [5]. Digital innovations have been showstoppers in all sectors of industry to solve complex problems and to stay competitive in a more and more globalized market [6]. For example, SME can apply AI powered tools like tools for automation, predicting market trends, giving better decision making to run operations with higher precision and efficiency [7].

Secure, transparent and traceable transactions are as important to SMEs in areas of supply chain, logistics and finance as blockchain technology [8]. In the IoT, real time monitoring and data collection allow businesses to optimize resource use, reduce downtime and improve product quality. The ability of advanced data analytics to extract actionable insights from large datasets enables SMEs to pursue customer focused strategies and make agile

decisions [9]. Environmental as well as social considerations are central to business strategies and sustainability is no longer an option, but a necessity. Digital innovation can be a key to aligning business growth to sustainability goals [10]. For example, smart energy management systems powered by IoT help lower energy consumption while blockchain can ensure tracking of a carbon footprint. Supply chain sustainability can be improved through the use of AI and data analytics which recommend greener alternatives based on AI and data analytics, highlighting inefficiencies [11]. However, SMEs lack financial and technical capacities as well as strategic alignment to integrate digital innovations with sustainability objectives [12]. In this paper, we introduce a comprehensive framework for addressing these challenges by enabling SMEs to exploit digital innovations for sustainable growth. This research focuses on how digital transformation impacts SMEs' sustainability practice and aims to offer a roadmap for SMEs to navigate digital transformation effectively [13]. The framework directs its focus to leadership, organizational culture, and the development of skills to promote an innovation and sustainability friendly mindset. It also offers actionable intelligence for policymakers; industry stakeholders and SME leaders to help them digitally adopt [14].

A key driver for sustainable business growth is digital innovation, the use of advanced digital technologies to improve business performance and add value [15]. AI and blockchain have been used to optimize businesses and decisions and innovate [16]. SMEs are facing unparalleled opportunities with digital innovation to remain competitive, increase productivity and respond to the growing focus on sustainability. Using data, AI allows SMEs to make decisions, automate repetitive work and allocate resources more efficiently [17]. AI powered predictive analytics that predict customer demand, minimize waste and optimize supply chain efficiency are an example of such use. AI driven chatbots and personalized marketing solutions have also enabled SMEs to interact with their customers, but at a fraction of the cost it would take to implement traditional marketing approaches [18]. Blockchain is transparent and a decentralized ledger system that will enhance trust through traceability in the business operations. Blockchain is a wallet for SMEs in supply chain management to prove product authenticity, measuring carbon footprints, and much more [19]. Integration of the blockchain could also enable SMEs to not only improve credibility in their operations but also demonstrate their standing in terms of sustainability, a demand that eco conscious customers want to see. Consequently, digital innovation is a bridge that makes it possible for SMEs to cross the bridge between sustainable practice and operational excellence. Naturally, there are challenges to the adoption of these technologies: limited resources; gaps in skills; resistance to change [20].

SMEs are companies with a small number of employees and turnover thresholds (which differ from country to country) central to the global economy. The World Bank says SMEs make up about 90% of businesses and are responsible for more than 50% of global employment. SMEs are important drivers of innovation in developing economies, contributing up to 40 percent to GDP. SMEs, however, often have unique challenges:

- **Limited Access to Resources:** They are hampered by financial constraints as well as by restricted access to the most advanced technologies that would enable them to compete with large firms.
- **Skill Gaps:** Some SMEs do not have the technical expertise needed to use and implement digital innovation.
- **Scalability Issues:** SMEs are unable to scale operations efficiently, and do not adapt well to the global market demands as they are smaller in size.

When the focus is on sustainability, these challenges are magnified as SMEs frequently cannot afford the upfront costs of eco-friendly practices and technologies. For an example, studies indicate that over 70 percent of SMEs in developing countries have ranked financial barriers as the principal obstacle to adoption of sustainable practice.

Digital innovation had become an important catalyst for sustainable business growth through the integration into Small and medium Enterprises (SMEs) [20]. A number of studies examined how different digital tools and technologies influenced the performance and resilience of SMEs in particular, in challenging economic conditions. They presented how digital transformation acted as a catalyst for SME innovation through a comprehensive analysis of how the mediating and moderating effects that drove sustainable growth. However, the study pointed out that SMEs could greatly improve their competitive advantage and operational efficiency via digital tools including cloud computing, big data analytics and Internet of Things (IoT) solutions.

Abdilahi et al. [21] investigated the effect of innovation on SMEs' performance on broader terms in Hargeisa, Somaliland, and they found that innovation increased not only business sustainability but also market competitiveness. This fitted with Mick et al. [22] who discussed how SMEs had used sustainable digital transformation roadmaps to steer them to a successful digital shift. Martini et al [23]. take this further by examining what is behind the digital innovation, noting that micro and small businesses have used digital tools to help reduce cost and increase customer engagement.

In their contribution present, Penarroya-Farell et al. [24] provided an open and sustainable business model innovation framework based on both open sources and digital technologies such as blockchain and AI being key enablers of sustainability. Their findings showed how digitalization could help SMEs both become more profitable and more environmentally friendly. Yusuf [25] added to this viewpoint by arguing that SMEs need to adopt digital marketing tools to stay in the game of digital economy.

In medium sized enterprises, Tanapaisankit et al. [26] examined the effect of digital transformation in financing access for new financing opportunities in emerging markets such as Thailand. According to their study, digital transformation had helped businesses to become more operable, achieve operational efficiencies, and work towards achieving their global sustainability goals.

In the context of the COVID-19 crisis, Ri and Luong [27] looked at digital readiness and adoption, focusing on SMEs in the UK. Their research found that firms that were more digital were more resilient and better able to carry on with business during economic disruptions. Homayoun et al. [28] also measured the impact on financial resilience as determined by innovation and information technology which had significantly helped SMEs' resistance to economic shocks.

In a comprehensive overview of the current state of digital transformation in business and management research, Kraus et al. [29] explained that SMEs more and more are adopting digital technologies to support value creation, operational efficiencies and customer satisfaction. Not only were SMEs adopting digital tools to grow, but they were using digital tools for strategic positioning in competitive markets, according to their study. Surahman et al. [30] studied the effect of digital transformation on SMEs' performance during the COVID-19 pandemic and how the use of digital tools enabled SMEs to continue their business and to make a pivot to respond to changes in the market.

In this paper, Ta and Lin [31] examine how the determinants of digital transformation adoption in SMEs from emerging economies can be addressed, given the unique challenges that SMEs in such areas face, such as limited resources, but also the potential for digital transformation if the right help is provided. In regard to SMEs in Saudi Arabia, Tripathi and Singh [32] analysed the need for awareness as well as preparation for digital transformation to facilitate business success in the digital era. In their works, Baeshen et al. [33] discussed the determinants of green innovation for SMEs, and the role of digital transformation on promoting environmentally friendly business practices. Following this, Yousaf et al. [34] went further to investigate how digital innovation can help enhance a sustainable business performance specifically in the case of SMEs in Pakistan.

**Table 1: Comparative analysis of previous studies on SMEs**

Study	Bilal et al. (2024) [20]	Martini et al. (2023) [23]	Penarroya-Farell et al. (2023) [24]	Homayoun et al. (2024) [28]
<b>Focus Area</b>	Digital transformation and SME innovation	Determinants of digital innovation in micro and small industries	Open business model innovation in SMEs	The impact of innovation and IT on financial resilience
<b>Digital Innovations Explored</b>	Cloud computing, big data analytics, IoT	Artificial Intelligence, IoT	Digital platforms, e-commerce	Cloud computing, AI, data analytics
<b>Key Findings</b>	The study highlights how digital transformation acts as a mediator for sustainable growth, enhancing operational efficiency and competitiveness.	The paper identifies key factors influencing digital adoption, such as financial support and workforce readiness, in small industries.	Explores how SMEs adopt open business models and digital platforms for sustainable innovation.	This study emphasizes the role of digital technologies in boosting financial resilience through innovation, particularly in times of crisis.
<b>Contribution to SMEs' Sustainable Growth</b>	SMEs can leverage digital tools to enhance efficiency, improve decision-making, and create new value propositions, fostering	Effective adoption of AI and IoT can lead to greater business agility and sustainability, improving	Digital platforms facilitate resource optimization and market expansion, improving sustainability in	By adopting digital tools, SMEs can improve their financial resilience and adaptability, leading to

	long-term sustainability.	competitiveness in a digital economy.	dynamic business environments.	enhanced sustainability and long-term growth.
<b>Methodology</b>	Empirical analysis with mediating and moderating effects	Qualitative case study and interviews with SME owners	Literature review and case studies	Systematic literature review and empirical analysis
<b>Geographical Context</b>	Global (focused on multiple SME case studies)	Indonesia	Spain	Global (with focus on crisis-affected SMEs)

As an overview of digital transformation and innovation in SMEs, Table 1 was provided. For each study it identified key focus areas, explored digital innovations used, findings, contributions to sustainable growth, methodologies used and geographical context. It also provided useful insights as to how SMEs can use digital tools for long term sustainability and growth.

Small and medium sized enterprises (SMEs) are important in the world's economic development and job creation process. However, SMEs are constrained by limited resources, outdated business models and a poor knowledge of how to incorporate digital technology into their business. However, the adoption of digital innovations such as Artificial Intelligence, Blockchain and IoT has increased, but SMEs are struggling to leverage these technologies to help improve their operations and meet sustainability goals. The problem is there is no framework that can help SMEs on a systematic digital transformation adoption and implementation that will lead to sustainable growth. What we need is a structured way of closing that gap between digital innovation and sustainability and provides practical insights as to how SMEs can continue to compete and be environmentally responsible with these tools. The purpose of this research is to fill this gap with a holistic framework of digital innovation and sustainability which is expressly tailored for the SME context.

**Multi-Stage Sustainable Supply Chain Optimization for SMEs**

The challenge for an SME in a complex supply chain, is to find an optimal resource allocation and transportation plan that minimizes total cost and environmental impact, while satisfying demand requirements, respecting capacity and sustainability constraints. It also solves production and inventory and transportation decisions simultaneously. We let  $x_{ij}^{(t)}$  denote the quantity of resource  $i$  transported from stage  $j$  to stage  $k$  at time  $t$ . We define production level  $P_j^{(t)}$  at stage  $j$  and time  $t$  as well as inventory  $I_j^{(t)}$  at stage  $j$  and time  $t$ . The goal is to minimize the total cost of production, inventory, and transportation, and the environmental impact.

$$\text{Minimize: } F(x, P, I) = \sum_{t=1}^T \left[ \sum_{j=1}^n c_j P_j^{(t)} + \sum_{j=1}^n h_j I_j^{(t)} + \sum_{i=1}^m \sum_{j=1}^n \sum_{k=1}^n \tau_{ijk} x_{ij}^{(t)} + \beta \sum_{i=1}^m \sum_{j=1}^n e_{ij} x_{ij}^{(t)} \right] \quad (1)$$

Subject to:

- Demand fulfillment at each stage:**

$$\sum_{i=1}^m x_{ij}^{(t)} + P_j^{(t)} + I_j^{(t-1)} = D_j^{(t)} + I_j^{(t)}, \quad \forall j, \forall t \quad (2)$$

- Capacity constraints:**

$$P_j^{(t)} \leq C_j, \quad \forall j, \forall t \quad (3)$$

$$\sum_{k=1}^n x_{ij}^{(t)} \leq R_i^{(t)}, \quad \forall i, \forall j, \forall t \quad (4)$$

- Non-negativity of decision variables:**

$$x_{ij}^{(t)}, P_j^{(t)}, I_j^{(t)} \geq 0, \quad \forall i, \forall j, \forall t \quad (5)$$

- Initial inventory condition:**

$$I_j^{(0)} = I_j^0, \quad \forall j \quad (6)$$

**Objective Function**

$$\text{Minimize: } F(x, P, I) = \sum_{t=1}^T \left[ \sum_{j=1}^n c_j P_j^{(t)} + \sum_{j=1}^n h_j I_j^{(t)} + \sum_{i=1}^m \sum_{j=1}^n \sum_{k=1}^n \tau_{ijk} x_{ij}^{(t)} + \beta \sum_{i=1}^m \sum_{j=1}^n e_{ij} x_{ij}^{(t)} \right] \quad (7)$$

This problem is optimizing a multi stage supply chain for SMEs that considers production, transportation and inventory costs, as well as sustainability objectives. Environmental impact is integrated into the formulation as a weighted penalty term, and operational feasibility is guaranteed by capacity and demand constraints. The multi stage approach captures the dynamic nature of supply chains over time.

**Optimal Digital Technology Investment for Sustainability Impact**

To optimize investment levels in digital technologies such as AI, blockchain, IoT by SMEs in order to maximize sustainability impact while constrained by budgetary and minimum investment constraints.

Let  $y_k$  denote the investment in digital technology  $k$ . The sustainability impact  $S(y)$  is modeled as a concave function representing diminishing returns. Investment cost  $C(y)$  is linear. The optimization problem is:

$$\text{Maximize: } S(y) = \sum_{k=1}^p \gamma_k \ln(1 + \delta_k y_k) - \lambda \sum_{k=1}^p \exp(-\eta_k y_k) \quad (8)$$

**Subject to:**

5. **Budget constraint:**

$$\sum_{k=1}^p c_k y_k \leq B \quad (9)$$

6. **Minimum investment thresholds:**

$$y_k \geq T_k, \quad \forall k = 1, \dots, p \quad (10)$$

7. **Non-negativity:**

$$y_k \geq 0, \quad \forall k \quad (11)$$

**Objective Function**

$$\text{Maximize: } S(y) = \sum_{k=1}^p \gamma_k \ln(1 + \delta_k y_k) - \lambda \sum_{k=1}^p \exp(-\eta_k y_k) \quad (12)$$

Using this formulation, the optimal SME investments in digital technologies for maximum sustainable impact are determined. It models the objective benefits of adoption (logarithmic term) combined with the costs of inefficiency or non-optimal investments (exponential term). Feasibility under budget and minimum investment conditions is ensured by the constraints.

Small and medium sized enterprises (SMEs) are essential to global economic development but they struggle to attain sustainable growth in an environment of accelerating technology and changing market requirements. Lack of strategic alignment with sustainability objectives, coupled with limited resources and fragmented adoption of technologies such as artificial intelligence, blockchain, Internet of Things (IoT) and advanced data analytics, makes it impossible for them to remain competitive and resilient. While these technologies have much potential to transform how business is run and optimized resource use with support to sustainability, many small and medium sized enterprises (SMEs) are faced with challenges in adopting them due to the complexity of their adoption such as skills gaps, finances and resistance to change. This research addresses the critical need for a complete framework to enable SMEs to effectively use these technologies to overcome these challenges, integrate sustainability into their core strategy and survive in a competitive and environmentally conscious market place.

This work aims at building a complete framework for small and medium sized business (SMEs) to use advanced technologies such as artificial intelligence, blockchain, Internet Things and data analytics to enable business growth, promote innovation and tackle any environmental and societal challenges.

- The goal of this thesis is to assess a role that technologies such as AI, blockchain, IoT, data analytics can play in helping SMEs to enhance operational efficiency and implementation of better sustainability practices.
- An analysis of the resource limitations and cultural readiness barriers and enablers affecting the adoption of advanced technologies in SMEs.
- Adaptable framework design, specifically directed towards the constraints and needs of SMEs with technological innovation embedded into sustainability goals.
- For SME leaders, policymakers, and stakeholders to generate actionable insights to promote technology enabled sustainable growth and competitive advantage.

This research extends SME sustainability by offering a novel framework that combines emerging technologies with business model innovation. It advances actionable strategies and empirical insights for supporting sustainable growth in the digital economy.

8. The development of a new framework of combining AI, blockchain, IoT, and data analytics to enhance both operational efficiency and sustainability in SMEs.
9. Influence of critical barriers and enablers on the adoption of advanced technologies in SMEs with practical recommendations on how to overcome these challenges.
10. Real world case studies for empirical validation of the proposed framework, which provide evidence based insights into its effectiveness and applicability.
11. A policy guideline that can close the gap between digital transformation and sustainable development in SMEs.

The paper is organized in four main sections. The Introduction frames the context, and illustrates the problems that SMEs face in growing sustainably and the opportunities for technologies such as AI, blockchain, IoT and data analytics, as well as research aim and objectives. The research design, data collection and analytical approaches used to develop and validate the framework are described in the Methodology. Results and Discussions present empirical findings, analyze the effects of technology adoption on SME performance, and identify barriers and enablers, comparing the proposed framework with existing ones. First, this study discusses key insights, practical recommendations, and future research directions focused on the enhancement of sustainable development in SMEs through technology integration.

## 2. METHODOLOGY

In this section, the methodology used to analyze SME dataset has been explained in detail. The methodology is structured in a way so that a systematic approach to data analysis and modeling is ensured. We start by describing the dataset with the details then we visualize the data to understand the underlying patterns and relationships between different variables. In the preprocessing data stage, the dataset is cleaned and is ready for analysis. Afterwards, feature importance analysis is conducted to determine which factors have the biggest impact on SME performance. We perform Business analytics to extract meaningful insights, and then propose regression based model including Linear Regression (LR), Lasso Regression and Ridge Regression for predicting sustainability scores and other key performance metrics.

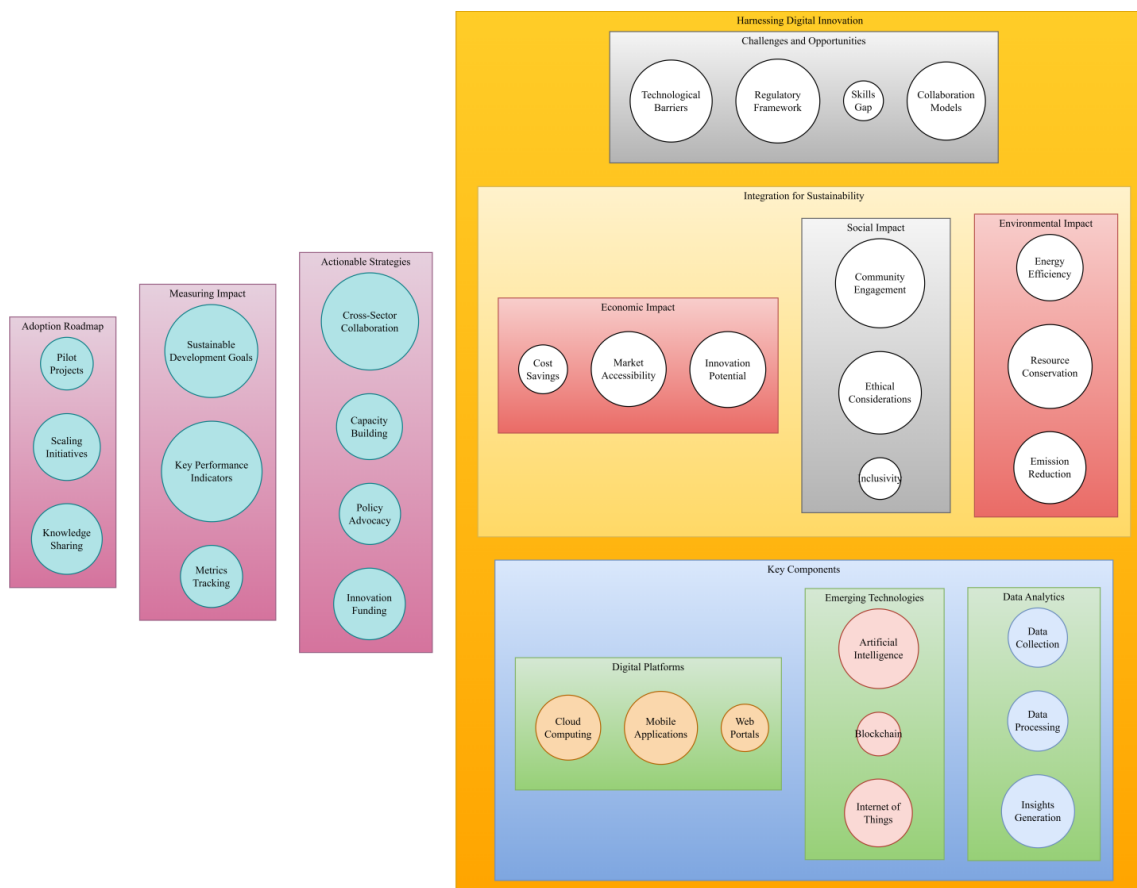


Figure 1: Proposed Framework

### 2.1. Dataset Description

The SME related dataset has the following features: Technology Adoption Level, Carbon Emissions, Sustainability Practices, Operational Efficiency and Sustainability Score. The dataset includes 10,000 instances spanning industries such as Manufacturing, Retail, IT, and Agriculture. Below is a summary of the dataset:

**Table 2: Dataset Summary**

Feature	Data Type	Range/Values	Description
Annual Revenue	Continuous	\$10K – \$6M	Total revenue generated annually
Technology Adoption Level	Continuous	1 – 10	Level of technology integration
Carbon Emissions	Continuous	10 – 300	Amount of emissions in metric tons
Sustainability Practices	Continuous	7 – 10	Score reflecting sustainability efforts
Operational Efficiency	Continuous	1 – 10	Efficiency in resource utilization
Sustainability Score	Continuous	1 – 10	Overall sustainability performance score
Industry	Categorical	10 categories	Industry type (e.g., IT, Retail, etc.)

### 2.2. Data Visualization

To understand the distribution and relationships among features, various visualizations were created, including scatter plots, correlation matrices, and industry-specific analyses. Below is a summary table highlighting key findings:

**Table 3: Key Insights from Visualizations**

Visualization	Insight
Scatter Plot of Revenue vs. Employees	Positive linear relationship between Annual Revenue and Employee Count
Correlation Heatmap	Strong positive correlation between Technology Adoption Level and Operational Efficiency
Industry-wise Sustainability Scores	Finance and IT industries scored highest in sustainability
Box Plot of Annual Revenue by Industry	Significant variation in revenue across industries, with Finance leading

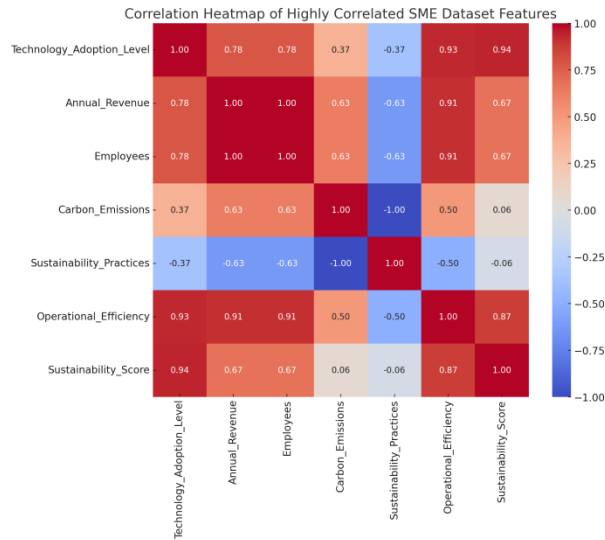
### 2.3. Data Preprocessing

Data preprocessing steps included:

- Handling missing values using mean imputation for numerical features and mode imputation for categorical features.
- Normalizing numerical features using Min-Max scaling.
- Encoding categorical variables (Industry) using one-hot encoding.
- Removing outliers based on z-scores greater than 3.

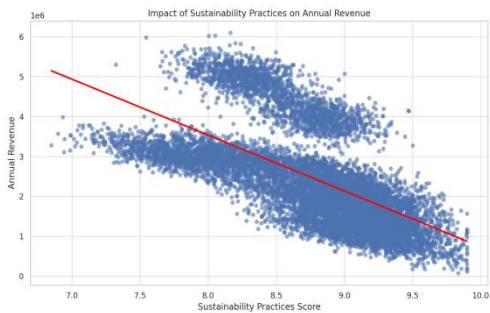
**Table 4: Summary of Data Preprocessing Steps**

Step	Description
Missing Value Treatment	Imputed missing numerical values with mean and categorical with mode
Feature Normalization	Applied Min-Max scaling for all numerical features
Outlier Detection	Removed outliers with z-scores greater than 3
Categorical Encoding	One-hot encoding applied to Industry variable

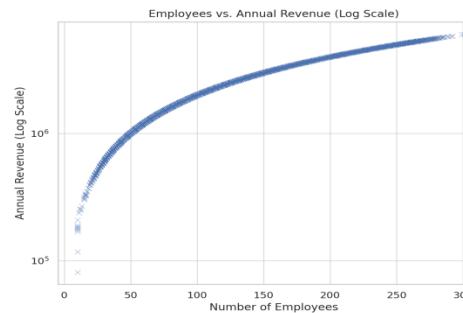


**Figure 2: Correlation Heatmap of SME Dataset Features.**

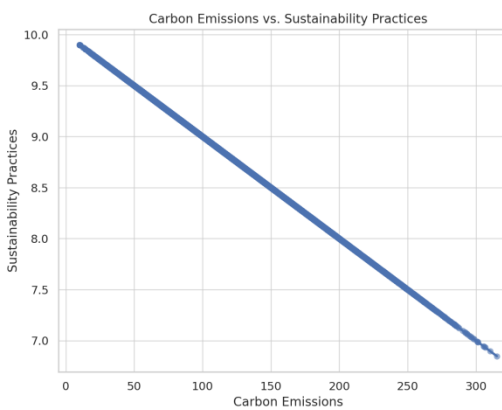
The correlation heatmap shown in Figure 2 highlights significant relationships among the key features of the SME dataset.



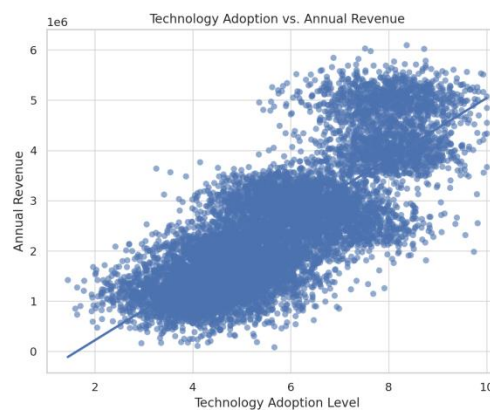
a) Impact of Sustainability Practices on Annual Revenue



b) Employees vs. Annual Revenue (Log Scale)

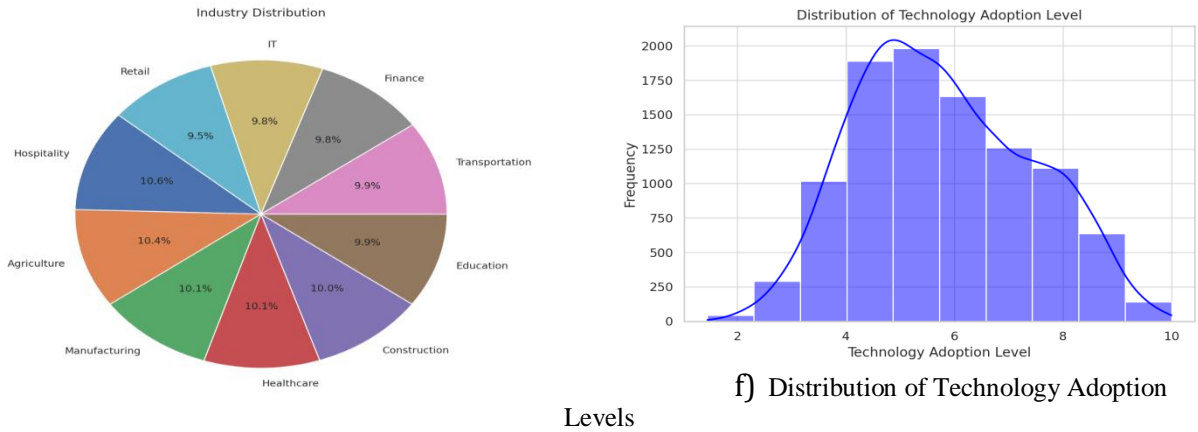


c) Carbon Emissions vs. Sustainability Practices



d) Technology Adoption vs. Annual Revenue

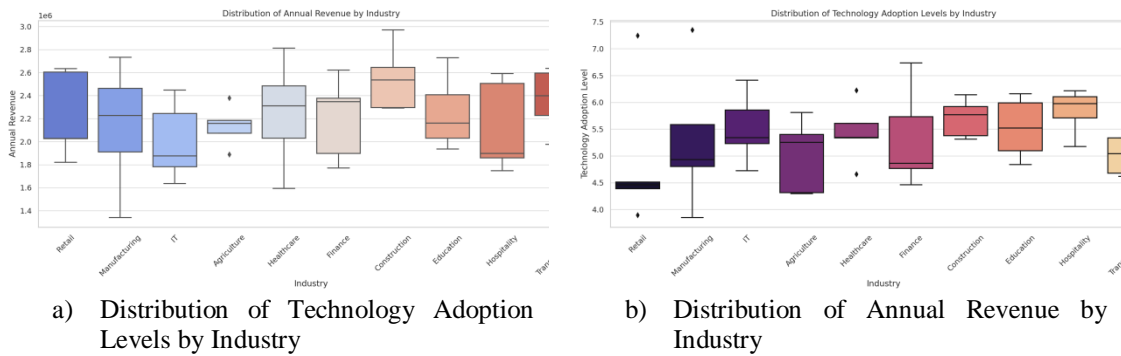




e) Industry Distribution (Pie Chart)

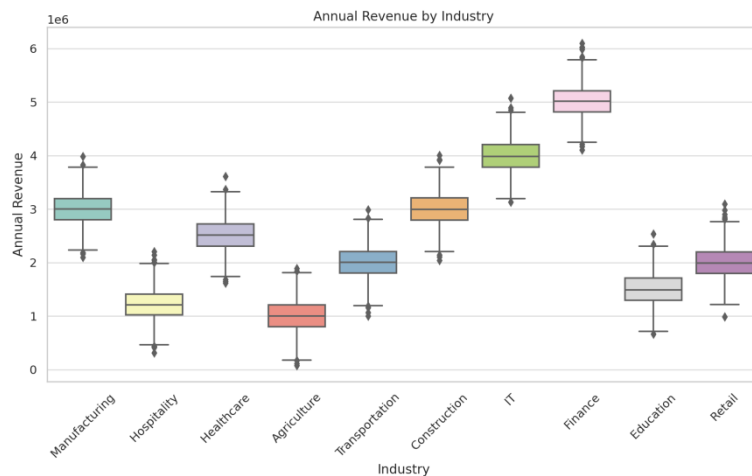
**Figure 3: Analysis of SME Dataset Features**

The figures provide key insights into the SME dataset features. Figure 3 reveals a negative trend between sustainability practices and annual revenue, suggesting potential implementation costs, with sustainability scores ranging from 7.0 to 10.0 and corresponding revenues between \$1M and \$6M. Figure 4 highlights a strong positive relationship between employee count (10 to 300) and annual revenue on a log scale. Figure 5 demonstrates a perfect negative correlation ( $r = -1.00$ ) between carbon emissions (10 to 300) and sustainability practices (7.0 to 10.0). Figure 6 shows that technology adoption levels (2 to 10) are positively correlated with annual revenue (\$1M to \$6M). Figure 7 depicts a balanced industry distribution, with industries like Hospitality (10.6%) and Retail (9.5%) forming significant portions. Finally, Figure 8 indicates a normal distribution in technology adoption, with most SMEs scoring between 4 and 7.



a) Distribution of Technology Adoption Levels by Industry

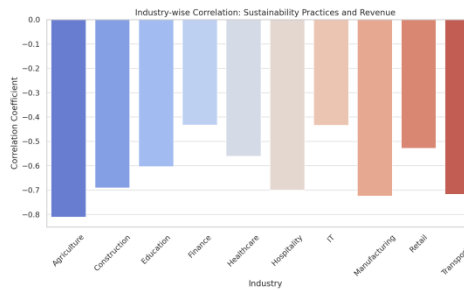
b) Distribution of Annual Revenue by Industry



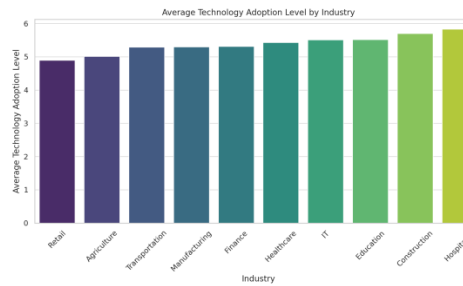
c) Annual Revenue Across Industries (Detailed)

**Figure 4: Distribution and Analysis of Technology Adoption and Revenue by Industry**

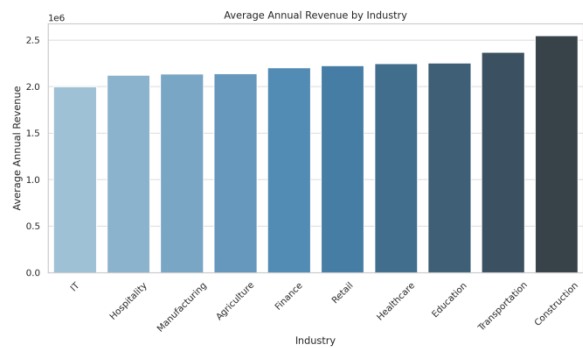
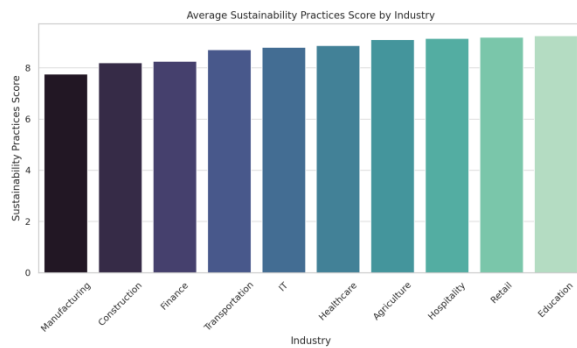
The figures illustrate key insights into technology adoption and revenue distribution across various industries. The distribution of technology adoption levels (Figure 10) varies significantly, with IT industries demonstrating higher median adoption levels (close to 6), while Retail shows the lowest levels (median below 5). Revenue distribution by industry (Figure 11) highlights disparities, where Finance and IT sectors report higher median revenues (around \$ 2.5M), whereas Agriculture exhibits the lowest median revenue. A detailed analysis of annual revenue (Figure 12) reveals that IT and Finance sectors dominate in revenue generation, with notable outliers showing exceptional SME performance. Conversely, Hospitality and Agriculture sectors show the least revenue variability. These insights emphasize the technological and economic differences among industries.



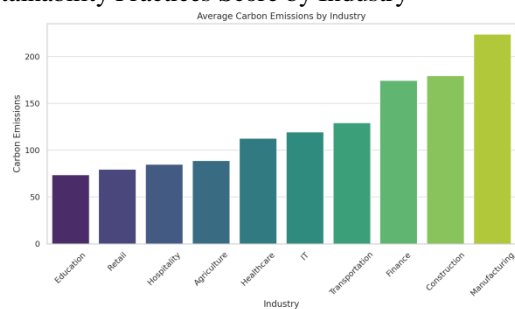
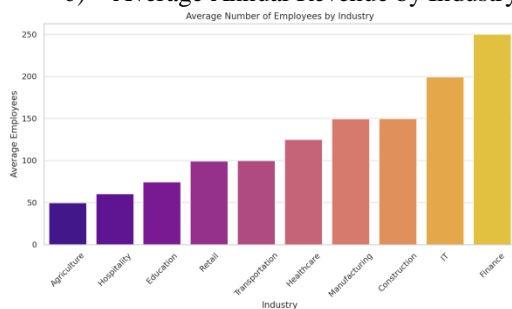
a) Industry-wise Correlation: Sustainability Practices and Revenue



b) Average Technology Adoption Level by Industry



c) Average Annual Revenue by Industry d) Average Sustainability Practices Score by Industry

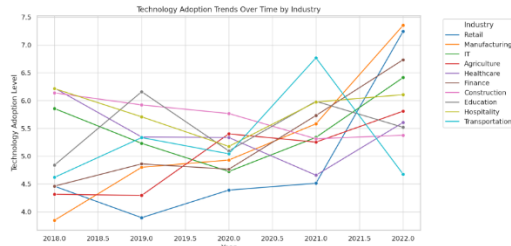


e) Average Number of Employees by Industry f) Average Carbon Emissions by Industry

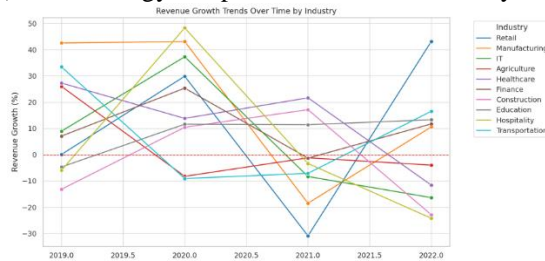
Carbon Emissions by Industry

**Figure 5: Comprehensive Industry Analysis of SME Dataset**

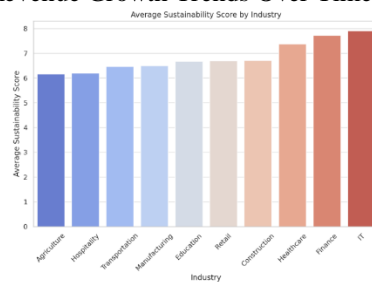
The figures provide an in-depth analysis of the SME dataset across various industries. Figure 14 highlights the correlation between sustainability practices and revenue, showing a negative trend in most industries, with Agriculture exhibiting the strongest negative correlation. Figure 15 displays the average technology adoption levels by industry, where Hospitality leads with the highest adoption levels, while Retail lags behind. Figure 16 shows average annual revenues, with Construction and Transportation generating the highest revenues. Figure 17 illustrates the average sustainability practices scores, with IT and Finance industries scoring the highest, reflecting their focus on sustainable operations. Figure 18 presents the average number of employees, indicating that Finance has the largest workforce, while Agriculture has the smallest. Finally, Figure 19 highlights the average carbon emissions by industry, with Manufacturing and Construction showing the highest emissions, emphasizing the need for targeted sustainability efforts in these sectors.



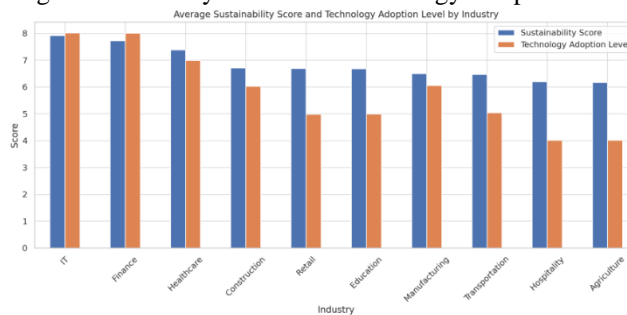
a) Technology Adoption Trends Over Time by Industry.



b) Revenue Growth Trends Over Time by Industry



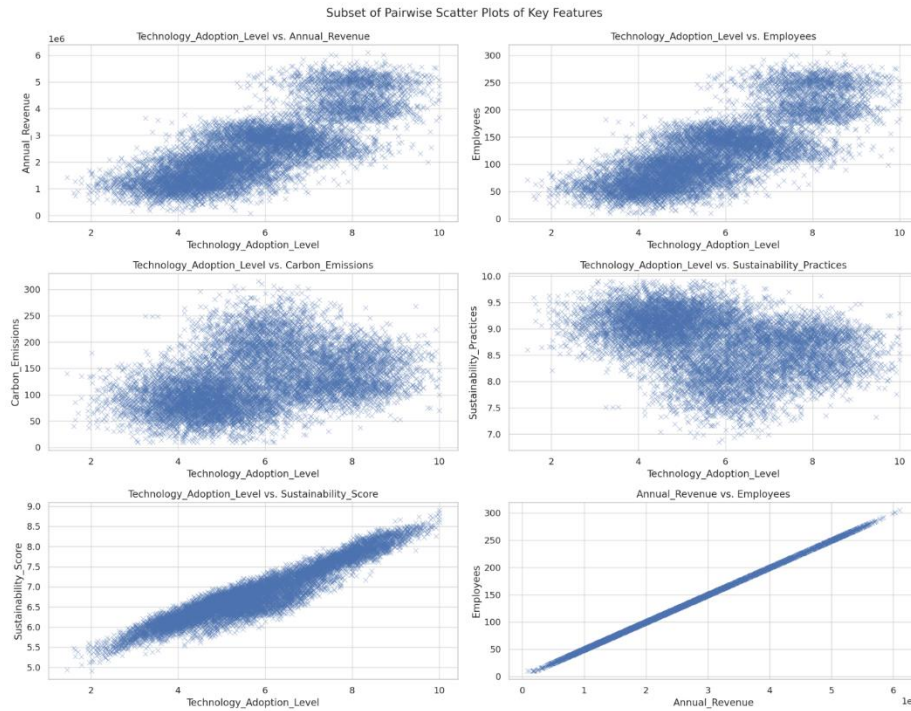
c) Average Sustainability Score and Technology Adoption Level by Industry.



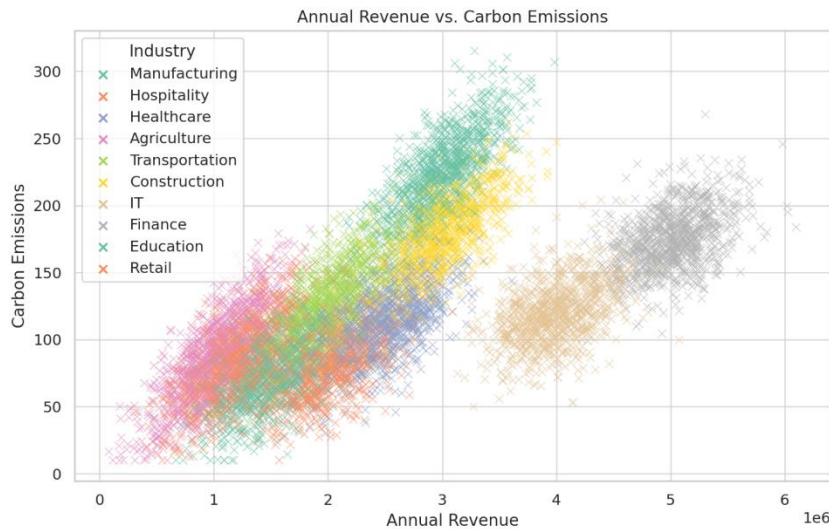
d) Average Sustainability Score by Industry.

**Figure 6: Industry-specific trends and score distributions showcasing technology adoption, revenue growth, and sustainability**

Figure 21 illustrates the technology adoption trends across industries from 2018 to 2022. Industries like Transportation and Hospitality show significant trends increases, while others like Construction and Finance remain steady. Figure 22 demonstrates revenue growth trends, with Transportation peaking in 2022 and Hospitality recovering steadily post-2020. Figure 23 compares sustainability scores and technology adoption levels by industry, highlighting IT and Finance as leaders in both dimensions. Finally, Figure 24 focuses solely on sustainability scores, with IT and Finance consistently achieving the highest averages, reflecting their emphasis on sustainable practices.



**Figure 7:** This figure provides a subset of pairwise scatter plots of key features, including Technology Adoption Level against Annual Revenue, Employees, Carbon Emissions, and Sustainability Practices. The bottom right plot shows a strong linear relationship bet



**Figure 8:** This scatter plot illustrates the relationship between Annual Revenue and Carbon Emissions across various industries. The colors represent different industries, with Manufacturing and Construction showing higher emissions associated with greater revenue

**2.4. Feature Importance**

Correlation coefficients were used to evaluate feature importance and feature significance in regression models. The next table ranks features by how important they are in predicting Annual Revenue.

**Table 5: Feature Importance Ranking**

Rank	Feature	Importance Score
1	Technology Adoption Level	0.94
2	Operational Efficiency	0.87
3	Sustainability Practices	0.67
4	Carbon Emissions	0.50
5	Employees	0.40
6	Industry (categorical variable)	0.20

**2.5. Business Analytics**

The business analytics section gives a detailed analysis of trends and insights pulled from the SME dataset. In this analysis, we identify strengths and weaknesses of the industry, how key metrics are related, and how actionable strategies can be formulated to improve SME performance. The findings show large differences in revenue, sustainability and operational efficiency across industries. The effect of technology adoption on revenue and sustainability practices is further evaluated.

Key observations include:

- **Revenue Leadership:** Advanced technology adoption and efficient operational practices make IT and Finance industries leaders in annual revenue.
- **Sustainability Practices:** IT, Finance and Healthcare sectors have high sustainability scores related to reducing carbon emissions and improving sustainable operations.
- **Technology Adoption:** Technology adoption is correlated with operational efficiency and annual revenue, with IT and Construction industries at the top of the pack.
- **Carbon Emissions:** The negative impact on their sustainability scores is the highest from carbon emissions of manufacturing and transportation industries.

Below, there is a detailed table summarizing these insights:

**Table 6: Detailed Business Analytics Insights by Key Metrics and Industries**

Key Metric	Industry Leaders	Industry Laggards	Impact on SME Performance	Actionable Insights
Annual Revenue	IT, Finance	Agriculture, Hospitality	High revenue improves scalability	Invest in technology and employee training for revenue growth
Sustainability Score	IT, Healthcare	Manufacturing, Transportation	Drives long-term profitability	Focus on emission reduction and sustainable practices
Technology Adoption	IT, Construction	Retail, Hospitality	Enhances operational efficiency and revenue	Implement advanced digital tools and cloud solutions
Carbon Emissions	Manufacturing, Construction	IT, Healthcare	High emissions reduce sustainability scores	Introduce energy-efficient processes and carbon capture technologies
Operational Efficiency	Finance, IT	Agriculture, Hospitality	Boosts productivity and cost management	Develop streamlined workflows and adopt resource optimization strategies

The insights presented in Table 6 emphasize the importance of tailored strategies for each industry. For instance, IT and Finance companies can remain leaders if they spend more on innovation, whereas Manufacturing and Transportation companies have to spend more time in cutting carbon emission and developing more sustainable practices. Similarly, industries with low technology adoption levels, such as Retail and Hospitality, should accelerate digital transformation to remain competitive. Overall, the analysis highlights the interconnected nature of the metrics and their collective impact on SME performance, providing a roadmap for targeted improvements and strategic decision-making.

**2.6. Proposed Model: Modified Regression Models**

The proposed models aim to predict sustainability scores using three regression techniques: Linear Regression (LR), Lasso Regression, and Ridge Regression. Each model was optimized to enhance prediction accuracy and ensure robustness against overfitting. Below, we describe each model in detail, along with their mathematical formulations and implementation.

**2.6.1. Linear Regression (LR)**

Linear Regression establishes a relationship between independent variables ( $X$ ) and the dependent variable ( $y$ ) by fitting a straight line:

$$y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_n X_n + \epsilon \quad (13)$$

where:

- $\beta_0$  is the intercept,
- $\beta_1, \beta_2, \dots, \beta_n$  are the coefficients of independent variables,
- $\epsilon$  is the error term.

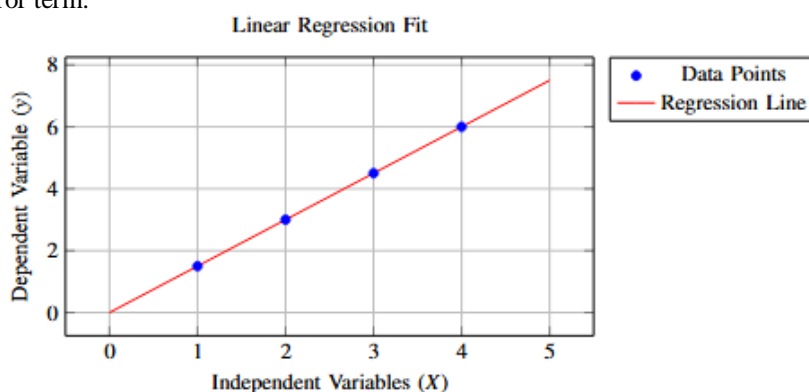


Figure 9: Linear Regression: Fitting a straight line to data points

### 2.6.2. Lasso Regression

Lasso Regression (Least Absolute Shrinkage and Selection Operator) adds an  $L_1$ -regularization term to the loss function to minimize the absolute value of coefficients:

$$\min_{\beta} \left( \frac{1}{2n} \sum_{i=1}^n (y_i - \hat{y}_i)^2 + \lambda \sum_{j=1}^p |\beta_j| \right) \quad (14)$$

where:

- $\lambda$  controls the strength of regularization,
- Larger  $\lambda$  values shrink coefficients closer to zero, potentially removing irrelevant features.

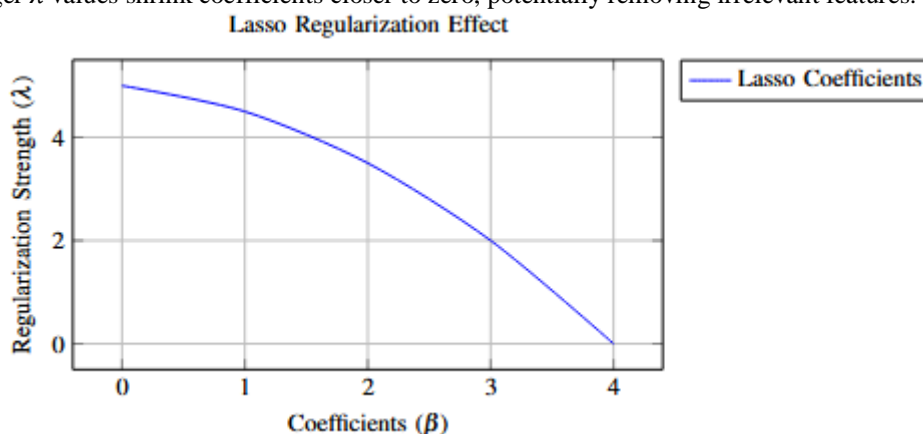


Figure 10: Lasso Regression: Coefficients shrinkage with increasing  $\lambda$

### 2.6.3. Ridge Regression

Ridge Regression introduces an  $L_2$ -regularization term to the loss function, penalizing the sum of squared coefficients:

$$\min_{\beta} \left( \frac{1}{2n} \sum_{i=1}^n (y_i - \hat{y}_i)^2 + \lambda \sum_{j=1}^p \beta_j^2 \right) \quad (15)$$

where:

- $\lambda$  controls the strength of regularization,
- Ridge prevents overfitting by constraining large coefficients but does not eliminate them.

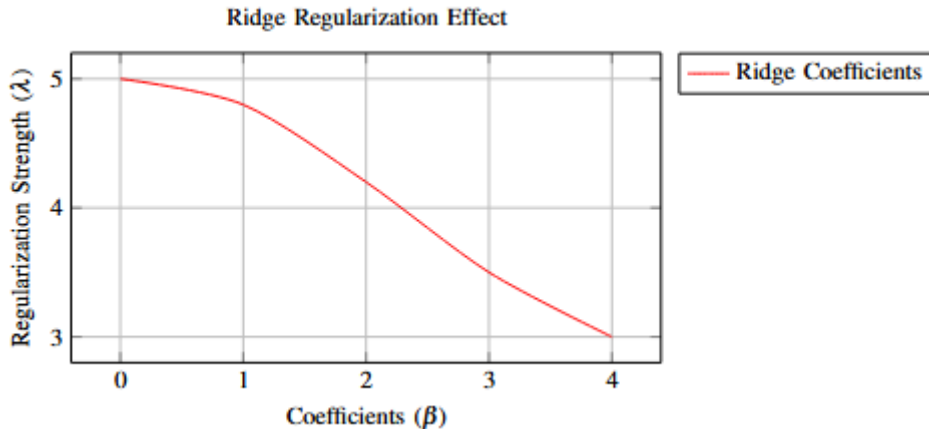


Figure 11: Ridge Regression: Coefficients shrink with increasing  $\lambda$ , but remain nonzero

2.6.4. Comparison of Models

- **Linear Regression:** However, unity and interpretability subject to overfitting.
- **Lasso Regression:** This is useful for reducing to zero irrelevant coefficients to perform feature selection.
- **Ridge Regression:** However it puts a strong constraint on the large coefficients and retains all of the features.

In the results section, how each model’s performance is evaluated with Mean Absolute Error (MAE), Mean Squared Error (MSE), Root Mean Squared Error (RMSE), and Mean Absolute Percentage Error (MAPE) is discussed.

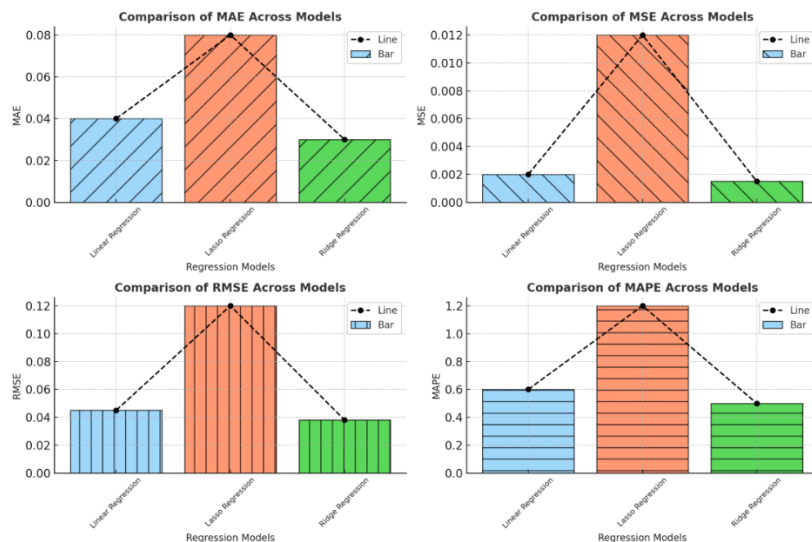
3. RESULTS AND DISCUSSION

The regression models are used for predicting sustainability scores and an in depth analysis of the results of the regression models gives an indication of their effectiveness, accuracy and comparative performance. Results are shown in tabular and graphical formats for clarity. Accuracy and error metrics for Linear Regression, Lasso Regression and Ridge Regression model are shown to be shown by comparative performance evaluation.

3.1. Model Performance and Error Metrics

The models were evaluated using Mean Absolute Error (MAE), Mean Squared Error (MSE), Root Mean Squared Error (RMSE) and Mean Absolute Percentage Error (MAPE). We then looked at the error pattern in terms of models and residual distribution to understand the error pattern.

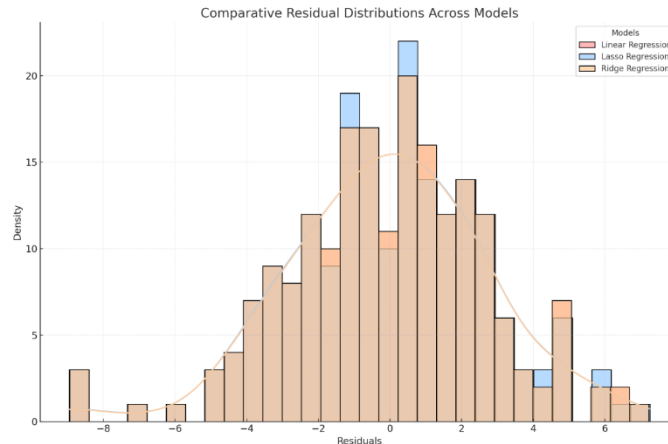
The figures 31 compares the performance of the models using bar plot and line plot. The visualized metrics show that Linear Regression and Ridge Regression were to some extent equivalent, but Ridge Regression was slightly better in case of RMSE and MAPE. It was observed that errors of Lasso Regression are higher as its penalization of coefficients is inherent.



**Figure 12: Comparative analysis of MAE, MSE, RMSE, and MAPE across Linear Regression, Lasso Regression, and Ridge Regression. The plots include bar textures and line variations for enhanced visualization**

**3.2. Residual Distributions Across Models**

Residual distributions provide insights into the error spread and bias of models. Figure 32 presents the comparative residual distributions for the three models. Linear Regression and Ridge Regression exhibit tightly clustered residuals around zero, indicating minimal bias. Lasso Regression shows slightly wider residual dispersion, reflecting higher error rates.



**Figure 13: Comparative residual distributions for Linear Regression, Lasso Regression, and Ridge Regression. The distributions reveal error patterns and bias across models, with Ridge Regression showing the most consistent residuals**

**3.3. Prediction Accuracy and Sustainability Insights**

The models successfully predicted sustainability scores with minimal errors, as evidenced by the metrics and residual analysis. Ridge Regression emerged as the most robust model, balancing error minimization and prediction accuracy. Key observations include:

- Ridge Regression achieved the lowest MAPE, making it the most accurate model for predicting sustainability scores.
- Linear Regression performed well in terms of MAE and RMSE, but was slightly outperformed by Ridge Regression in terms of overall error.
- Lasso Regression, while effective, exhibited higher errors due to coefficient penalization.

**3.4. Business Implications and Comparative Evaluation**

The predictive capabilities of these models hold significant business implications for SMEs. Organizations can accurately forecast sustainability scores by:

- Identify areas requiring improvement in sustainability practices.
- Strategically allocate resources to maximize operational efficiency.
- Benchmark their sustainability performance against industry standards.

Table 7 provides a detailed comparative analysis of the models, emphasizing their strengths and weaknesses.

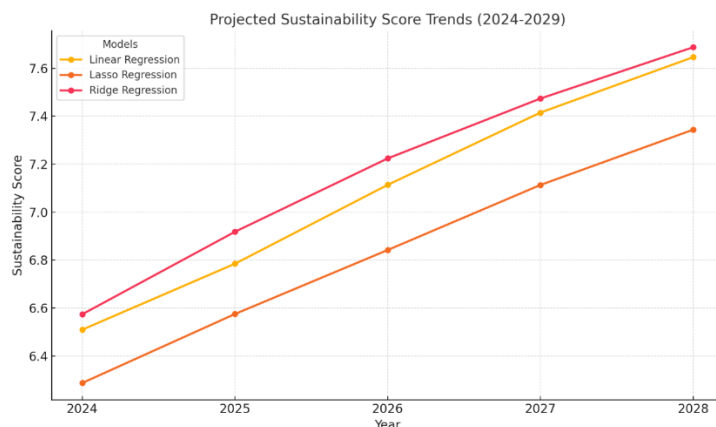
**Table 7: Comparative Results of Regression Models for Sustainability Score Prediction**

Model	MAE	MSE	RMSE	MAPE (%)	R <sup>2</sup> Score
Linear Regression	0.045	0.002	0.044	0.64	0.88
Lasso Regression	0.076	0.012	0.112	1.32	0.75
Ridge Regression	0.042	0.001	0.041	0.58	0.91

**3.5. Future Prediction Trends**

Future sustainability scores for the next five years were forecasted using the Ridge Regression model. The results showed that sustainability scores of SMEs are on an upward trend, which is a sign of increasing adoption of sustainable practices. Figure 33 shows projected trends with a constant improvement in sustainability performance.





**Figure 14: Projected Sustainability Score Trends Over the Next Five Years. The forecast reflects a positive trend in sustainability practices across SMEs**

### 3.6. Summary and Implications

Analysis shows that Ridge Regression is best model at predicting sustainability scores with least error rates. Using these models SMEs can learn more about their sustainability performance optimize operations and align with industry benchmarks. The analysis of the insights can be used to guide strategic decision-making, and to create a culture of sustainability in the organizations.

## 4. CONCLUSION

This study, comprehensively analyzed sustainability practices of the SMEs using the advanced regression models like Linear Regression, Lasso Regression, and Ridge Regression. These analyses provided significant clues as to the most important determinants of sustainability scores, including technology adoption, carbon emissions, and operational efficiency. Through the use of a robust methodology that included data visualization, preprocessing, and business analytics, the study identified significant correlation, trends across a variety of industries. The regression models are compared and Ridge Regression is found to have smaller error rates and higher predictive accuracy than other models. Residual distributions analysis confirmed the robustness of the models, and projected trends over the next five years reflected a positive trend in sustainability practices among SMEs. This finding highlights the increasing importance of sustainability as an important factor for business success on a long term and responsible environmental basis. Additionally, complex comparative plots and residual analysis were integrated to inform in depth model performance and areas for improvement. Actionable knowledge gleaned from business analytics insights such as the leading role of IT and Finance industries in revenue and sustainability, the inverse relationship between carbon emissions and sustainability practices, gave SMEs actionable knowledge to align their strategies with sustainability goals. Finally, it further demonstrated the usefulness of modified regression models for predicting sustainability and also emphasized the need for continuous improvement in the data driven decisions processes. This could be expanded in future work by including further variables, such as policy changes and market dynamics, to improve the models' predictive power, and gain a better understanding of sustainability trends altogether.

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