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“AI in Recruitment – Exploring Opportunities and Challenges in Modern Hiring Practices”

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Abstract

The integration of Artificial Intelligence (AI) in recruitment has revolutionized modern hiring practices, offering numerous opportunities and challenges. This study explores the impact of AI on recruitment, examining its potential to enhance efficiency, reduce bias, and improve candidate experiences. The research investigates the applications of AI in various stages of the recruitment process, including resume screening, candidate sourcing, and interview assessments.

The study also delves into the challenges associated with AI in recruitment, such as algorithmic bias, data quality issues, and the need for transparency and accountability. Through a comprehensive review of existing literature and empirical analysis, this research aims to provide insights into the benefits and drawbacks of AI in recruitment.

The findings of this study will contribute to a deeper understanding of the role of AI in shaping modern hiring practices, highlighting best practices for implementation and identifying areas for future research. The study's results will be valuable for organizations, recruiters, and job seekers, enabling them to navigate the complexities of AI-driven recruitment and harness its potential to create more efficient, fair, and effective hiring processes.

Keywords: Artificial Intelligence, Recruitment, Hiring Practices, Opportunities, Challenges, Algorithmic Bias, Efficiency, Candidate Experience.

1. Introduction

The recruitment landscape has undergone a significant transformation in recent years, driven by the increasing adoption of technology and digital tools. One of the most notable developments in this space is the integration of Artificial Intelligence (AI) in recruitment processes. AI-powered solutions are being used to automate routine tasks, enhance candidate sourcing, and improve the overall efficiency of hiring processes.

Despite these challenges, the use of AI in recruitment is becoming increasingly prevalent, with many organizations investing heavily in AI-powered recruitment solutions. As the use of AI in recruitment continues to grow, it is essential to explore the opportunities and challenges associated with this technology.

This study aims to contribute to the growing body of research on AI in recruitment, exploring the benefits and drawbacks of AI-powered recruitment solutions. By examining the current state of AI in recruitment, this study will provide insights into the opportunities and challenges associated with this technology, highlighting best practices for implementation and identifying areas for future research.

Research Background:

The recruitment process is a critical function in any organization, responsible for identifying, attracting, and selecting the best talent. Traditional recruitment methods, such as manual resume screening and interviews, can be time-consuming and prone to bias. The use of AI in recruitment has the potential to address these challenges, enhancing the efficiency and effectiveness of hiring processes.

Research Significance:

This study is significant because it will provide insights into the opportunities and challenges associated with AI in recruitment, contributing to a deeper understanding of the role of technology in shaping modern hiring practices. The findings of this study will be valuable for organizations, recruiters, and job seekers, enabling them to navigate the complexities of AI-driven recruitment and harness its potential to create more efficient, fair, and effective hiring processes.

2. Review of literature

Review of literature is the most essential part of the research work as it helps in knowing the gaps thereby getting a clear idea of one's own research. The present chapter is an effort to briefly describe the existing studies on electronic media for news and the patterns of the audience for consumption of news. An effort has been made to access the available literature in the area of electronic news media which explores the different topics like development of electronic media like television and online media for news consumption, preference of electronic media, impact of electronic media/online media on various groups. At last, the chapter visualized the research gaps in the existing explored from various studies literature. The analysis of literature on the above mentioned areas are discussed below.

Laorden, et al (2001) found the recruitment through internet as a very power full and effectiveness in term of opportunity. Through the survey method a sample of company's CEO was taken from the Europe. This study has been conducted from the business point of view rather than delving into the technical aspect and development tool. Further the paper described the actual and emergent model in the e recruitment market, its access brings the change in recruitment and consider as a better recruitment tool which is bringing in companies practices and strategies. They are also analysing the risk and the opportunity for the company and the job candidate. This study helps to understand the e-recruitment as a complementary and efficient tool in HR management strategy and a source of added value to their organisation.

Adsley, David (2002) focused on the recruitment sector as well as advertising on the Internet. With the novelty approach, coupled with the concept of high speed, apparently low cost against traditional sector-specific media, a very high reader audience and expected high response levels can seem superficially attractive to a client's in-house recruitment function. However, author points out that relying on electronic mail communications that are very impersonal does not give much guidance in the way of identifying a person's soft skills. There is no doubt that sector-specific recruitment can benefit from the modern communication technology advances. The progressive specialist consultancy seeks to utilize constructively the potential of the electronic mail and web site facilities within its portfolio of services and procedures to supplement, not replace, its sector expertise and without losing sight of the personal touch need when dealing with human beings.

Van et al (2002) explored Internet as an effective recruitment source and medium in this modern era. The Internet has quickly become a widely used method for recruitment by individuals and organizations. This study examines differences between individuals sourced through the internet versus individuals sourced through non-Internet methods on the basis of post-hire outcomes, individual characteristics, and information realism. Additionally, this study investigates the special case of information technology professionals' use of the internet as a recruitment source. The insights provided by this study enable a better understanding of the consequences of using the Internet as a recruitment source as compared to non-Internet methods. The insights provided by this study help determine the characteristics of an effective Internet-based recruiting system. This study produces results that improve understanding of the Internet as a recruitment source and as a recruitment medium. The importance of recruitment and the Internet's role in recruitment are bound to increase.

3. Methodology

This study uses a descriptive and exploratory research design, based on both primary and secondary data. Primary data is collected through a structured questionnaire using Google Forms, targeting HR professionals and job seekers. Purposive sampling is used to select respondents. The data is analyzed using descriptive statistics and charts to identify trends, opportunities, and challenges in AI-based recruitment.

3.1. Statement of the Problem

As organizations increasingly adopt Artificial Intelligence (AI) in recruitment processes, there is a growing need to understand its real-world impact. While AI offers benefits like speed, efficiency, and data-driven hiring, it also raises concerns about bias, transparency, and ethical issues. Many companies and job seekers are still uncertain about the effectiveness and fairness of AI-driven tools. This study aims to explore these opportunities and challenges to provide deeper insights into the role of AI in modern hiring practices.

3.2. Need for the Study

The integration of Artificial Intelligence (AI) in recruitment is reshaping how organizations attract, evaluate, and hire talent. As AI-powered tools such as resume screening bots, chatbots, and predictive analytics become more widespread, it is crucial to understand their actual impact on the

hiring process. While AI promises increased speed, reduced human error, and data-driven decision-making, it also introduces concerns related to algorithmic bias, lack of transparency, and ethical dilemmas.

There is a clear need to study how effectively these AI tools are being used by companies, and whether they truly enhance the quality of hiring or pose risks to fairness and inclusivity. Both employers and job seekers must be aware of the capabilities and limitations of AI in recruitment. This study aims to bridge the knowledge gap and support informed decision-making by exploring the opportunities and challenges involved in adopting AI for hiring.

3.3. Scope of the Study

This study focuses on the application of Artificial Intelligence (AI) in recruitment processes across different sectors. It aims to examine how AI tools such as resume screening software, chatbots, video interview analyzers, and predictive analytics are transforming the traditional hiring methods. The research evaluates both the positive aspects (such as time efficiency, reduced human bias, and improved candidate matching) and challenges (such as ethical concerns, data privacy, and lack of human touch) associated with AI-driven recruitment.

The study covers insights from HR professionals, recruiters, and job seekers, primarily collected through a structured questionnaire. It also includes a review of secondary data from industry reports and academic literature. While the research includes a broad view of AI in recruitment, it is limited to the current awareness, usage, and perception of AI tools in hiring. Geographically, the study may be limited to a specific region or country depending on the respondent base.

3.4. Research Questions

1. How are AI technologies currently being used in recruitment processes by organizations?
2. What are the key benefits and opportunities AI offers in improving hiring efficiency and decision-making?
3. What challenges or limitations do organizations and candidates face when using AI in recruitment?
4. How do HR professionals and job seekers perceive the fairness and transparency of AI-based hiring tools?

5. What ethical, legal, and privacy concerns are associated with AI-driven recruitment practices?

3.5. Objectives of the Study

1. To explore how AI technologies are transforming modern hiring practices.
2. To identify the key opportunities that AI offers in enhancing recruitment efficiency and effectiveness.
3. To assess ethical, legal, and regulatory concerns in AI-based hiring practices.

3.6. A) Type of Research

This research is a combination of:

Exploratory Research – To explore how AI is being integrated into recruitment and identify emerging trends and unknown challenges.

Descriptive Research – To describe the current practices, perceptions, and outcomes of using AI in recruitment.

Mixed Method Approach – Using both qualitative (opinions, open-ended questions) and quantitative (structured questionnaires, statistics) data collection.

B) Research Design

This study adopts a descriptive and exploratory research design to analyze how Artificial Intelligence is transforming modern recruitment practices. The descriptive component helps in outlining the current use of AI tools like resume screening software, chatbots, and predictive hiring systems, while the exploratory aspect investigates new opportunities, ethical concerns, and challenges that arise with their implementation.

The research design supports both qualitative and quantitative approaches to gather data. Quantitative data is collected through a structured questionnaire (Google Form), which includes multiple-choice and Likert scale questions targeting HR professionals, job seekers, and hiring managers. Qualitative insights are gathered through open-ended questions to capture opinions and personal experiences.

The study uses purposive sampling to select respondents who have exposure to AI in hiring. Data analysis is conducted using descriptive statistics such as percentages, charts, and graphs, which allow for easy interpretation and comparison of responses.

3.7. Data Collection

The study uses both primary and secondary data collection methods.

Primary data is collected through a structured questionnaire (via Google Forms) designed to gather information from HR professionals, recruiters, and job seekers. The questionnaire includes closed-ended, Likert scale, and open-ended questions.

Secondary data is gathered from academic journals, industry reports, websites, and previous research studies related to AI in recruitment.

3.8. Sampling design

a) Sampling Plan

The research uses a non-probability sampling plan, specifically purposive sampling, to select participants who have experience or knowledge of AI-based recruitment processes. This approach ensures relevant and meaningful data is collected from a targeted group.

b) Sampling Method

Purposive Sampling Method is used. Participants are intentionally selected based on their involvement in recruitment processes or their exposure to AI-based hiring tools (e.g., HR professionals, job seekers who've experienced AI screening, and hiring managers).

c) Sampling Frame

The sampling frame includes: HR departments of various companies Job seekers registered on online job portals (e.g., LinkedIn, Naukri)

Recruitment agencies Professionals in hiring-related roles who are accessible online.

Note: Since it's non-probability sampling, the frame is not exhaustive but focused and relevant.

d) Sampling Units

The sampling units are individual respondents who have either Experience in using AI tools for hiring (HR/recruiters), or have been evaluated through AI systems during their job search (candidates)

Plan of Analysis

The data collected from the structured questionnaire will be analyzed using both quantitative and qualitative methods to derive meaningful insights.

1. Quantitative Data Analysis

Data Tabulation: All responses will be compiled and organized in Microsoft Excel or Google Sheets for clarity.

Descriptive Statistics

Frequency counts and percentages will be calculated to understand how many participants responded in a certain way.

Mean, mode, and standard deviation may be used (if needed) to assess central tendencies for scaled questions.

Graphical Representation

Visual tools such as bar graphs, pie charts, and column charts will be used to present trends in a visually appealing way. This analysis will help to identify: The extent to which AI tools are used in recruitment The most common benefits and challenges experienced. The perception of fairness, accuracy, and transparency in AI-driven hiring

2. Qualitative Data Analysis

Open-ended responses (from job seekers and HR professionals) will be grouped into themes such as Ethical concerns Transparency and bias Personal experience with AI in hiring. A thematic analysis approach will be used to extract major opinions, common phrases, or issues raised by respondents. These insights will be summarized and interpreted to complement the quantitative findings.

3. Cross-tabulation (Optional, Advanced)

If sample size allows, cross-tabulation may be done to analyze the relationship between: Role (HR vs Job Seeker) and their level of trust in AI, Industry type vs frequency of AI tool usage Age group vs openness to AI-based hiring.

4. Final Interpretation

The results will be compared with findings from secondary sources (such as articles and research reports) to validate and enrich the conclusions.

4. Analysis and Interpretation

Analysis

AI in Hiring – Analysis Table

Number	Content	Overall Result
1	Recruiters value AI for data-driven hiring insights.	AI improves decision-making more than engagement/sourcing tools.
2	Fairness in AI hiring.	Requires holistic approach: audits, diverse data, human oversight.
3	AI adoption impact.	Most organizations see moderate to high transformation.
4	AI usage in hiring.	Mainly for filtering; advanced tools like predictive analytics less used.
5	AI vs. human judgment.	AI supports decisions, but human control prioritized.
6	Primary goal of AI.	Used to improve recruitment speed and efficiency.
7	AI reduces time-to-hire.	60% significant reduction; small share saw no impact or increase.
8	AI in talent identification.	Over half agree it helps; some uncertainty remains.
9	Ethical concerns.	Top issues: transparency, privacy, bias; fewer fear job loss.
10	Bias mitigation measures.	Audits most common, but 27.3% use no measures.
11	Legal/regulatory concerns.	Mixed views; some concerned, others unconcerned/unsure.

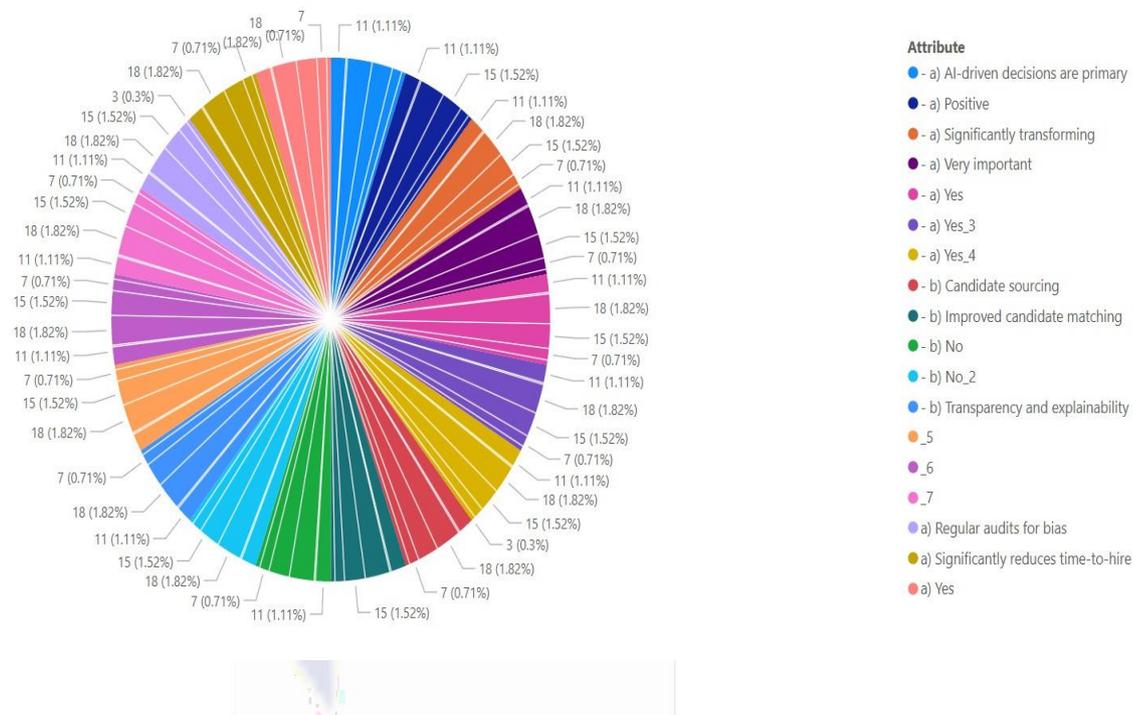
12	Compliance with AI regulations.	About half comply; rest lack compliance or awareness.
13	Importance of AI adoption.	Majority find it important; few unconvinced.
14	AI replacing human judgment.	Opinions evenly split; human intuition still valued.
15	Need for AI training.	Nearly 90% demand training to use AI effectively.
16	Perceptions of AI hiring.	Mostly positive or neutral; few negative views.
17	Future of AI in hiring.	Majority believe AI will evolve to address limits.
18	AI's role in future recruitment.	Strong consensus that AI will be essential.
19	Usefulness in hiring decisions.	Nearly 90% find AI helpful for data-driven hiring.
20	Expectations for AI tools.	73% want more features; varied needs expressed.

1. Recruiters value AI's ability to use data-driven insights for better hiring decisions more than engagement tools or sourcing automation.
2. Organizations recognize fairness in AI hiring as a multi-dimensional issue, requiring a holistic approach rather than relying on a single method to eliminate bias.
3. AI adoption is driving noticeable changes in recruitment, with most organizations experiencing at least moderate transformation, while very few report no impact.
4. Organizations primarily rely on AI for initial filtering tasks, while advanced applications such as predictive analytics (16.4%) remain less commonly adopted.
5. AI is preferred as a supportive tool to enhance decision-making, while human judgment continues to be prioritized to maintain control and accountability in hiring.
6. The primary motivation for AI adoption is to make recruitment faster and more efficient, while improvements in candidate matching, fairness, and experience receive less emphasis.
7. 60% of respondents stated that AI significantly reduces time-to-hire, 25.5% felt it somewhat reduces it, while only a small portion saw no impact (9.1%) or an increase (5.5%).

8. Over half of respondents (54.5%) believe AI effectively helps identify top talent. About 27.3% are unsure, while 18.2% disagree with this view.
9. Among 55 respondents, the top ethical concern with AI in hiring is transparency and explainability (43.6%), followed by data privacy (38.2%) and bias and fairness (32.7%). Only 12.7% expressed concern about job displacement.
10. The most common measure to address bias in AI hiring is regular audits (36.4%), followed by human oversight of AI decisions (25.5%) and training data review (10.9%). However, 27.3% of organizations reported no specific measures at all.
11. 40% of respondents expressed concern about legal or regulatory issues in AI hiring, 45.5% reported no concern, and 14.5% were unsure about such implications.
12. 50.9% of respondents confirmed compliance with local regulations on AI in hiring, 25.5% admitted non-compliance, and 23.6% were unsure.
13. 49.1% of respondents stated that AI adoption in hiring is very important, 30.9% said it is somewhat important, 18.2% considered it not very important, and only 1.8% felt it is not at all important.
14. 43.6% of respondents believe AI will replace human judgment in hiring, 43.6% feel it will not, and 12.7% are unsure.
15. 89.1% of respondents believe additional training on AI in hiring would be beneficial, while only 10.9% do not see its need.
16. 40% of respondents perceive AI-driven hiring positively, 47.3% remain neutral, 9.1% view it negatively, and a very small portion are unsure.
17. A majority (58.2%) believe AI in hiring will evolve to address current limitations, suggesting optimism about its potential to improve, though responses also show a mixed outlook on its future.
18. 76.7% of respondents agree AI will be essential in recruitment, 30.2% are neutral, and 2.3% strongly disagree. The minimal negative response indicates openness to AI integration.
19. 88.4% of respondents find AI helpful in data-driven hiring decisions, with an equal distribution between “Extremely helpful” and “Helpful.” Only 4.7% find AI not helpful.

20. For the final question, 43 responses were received, with 73% wanting more features. Responses varied, with “Nil” being the most common, reflecting diverse expectations for AI hiring tools.

Interpretation



1. Most respondents (41.9%) identify predictive analytics for candidate selection as the biggest opportunity for AI to enhance recruitment efficiency.
2. A majority (37.2%) believe ensuring fairness in AI-powered hiring tools requires a combination of measures, including audits, diverse training data, and human oversight.
3. AI is widely seen as positively shaping recruitment, with 38.2% stating it is significantly transforming hiring practices and 36.4% noting it is somewhat transforming.
4. Resume screening (49.1%) is the most widely used AI-powered tool in hiring, followed by candidate sourcing (38.2%) and interview scheduling (36.4%).

5. Over half of respondents (50.9%) report that AI mainly informs decisions while humans make the final call, whereas only 20% rely on AI as the primary decision-maker.
6. Increased efficiency is the most valued outcome of AI in recruitment, identified by 43.6% of respondents as the key benefit.
7. AI largely enhances recruitment speed, though its effectiveness varies across organizations depending on implementation.
8. AI is widely regarded as beneficial in talent identification, although doubts persist. Increasing awareness and proper implementation could reduce uncertainty.
9. Transparency and explainability of AI decisions are the leading ethical concerns, followed by privacy and fairness. Fears of job loss are less pressing but still present.
10. Many organizations actively attempt to reduce bias through audits and oversight, but a significant portion still lack structured measures, highlighting the need for stronger frameworks to ensure fairness.
11. Perceptions about legal and regulatory risks are split—while many acknowledge potential risks, a slightly larger group remains unconcerned. “Unsure” responses highlight a lack of awareness, suggesting the need for better clarity and guidance on AI compliance in recruitment.
12. While a majority follow regulatory norms, a significant proportion either lack compliance or awareness. This highlights the need for clearer guidelines and stronger regulatory awareness among organizations using AI in recruitment.
13. AI adoption is widely viewed as crucial in recruitment. However, a section of respondents remains cautious or unconvinced, indicating varying levels of readiness and confidence in AI’s role in hiring.
14. Responses reveal a clear divide, with equal proportions supporting and rejecting the idea of AI fully replacing human judgment. This reflects AI’s power but also shows that many still value human intuition and decision-making.
15. Strong demand for AI-related training in recruitment reflects respondents’ recognition of the importance of skill development to effectively use AI tools in hiring.

16. While many see AI-based hiring as beneficial, a large share remain neutral, possibly due to limited awareness or mixed experiences. Negative perceptions are minimal, indicating overall acceptance with scope for building stronger trust.
17. Respondents expect AI to play a significant role in shaping the future of recruitment. Optimism about its evolution suggests it may become a crucial tool for recruiters, with ongoing research and development needed to address its limitations.
18. A strong consensus emerges on AI's future importance in recruitment. Most respondents are positive about its potential, indicating a significant shift towards AI-driven recruitment practices.
19. AI is perceived as enhancing hiring practices and delivering significant benefits. These insights can inform the development and refinement of AI recruitment tools.
20. Users express a clear demand for improvements and new features in AI hiring tools. Diverse responses indicate varied needs and expectations, providing valuable feedback for developers.

Findings

The study reveals that the adoption of Artificial Intelligence (AI) in recruitment is becoming increasingly prevalent, with organizations using primarily for operational tasks such as resume screening, candidate sourcing, and interview scheduling. However, in line with the Technology Adoption Model (TAM), the findings show that while organizations acknowledge the perceived usefulness of advanced AI applications such as predictive analytics, actual implementation of these tools remains relatively low.

The data further confirms that efficiency and time reduction are the most valued benefits of AI, consistent with Resource-Based View (RBV) theory, where technology is leveraged to optimize organizational processes. Despite this, human judgment continues to be prioritized over full automation, aligning with the Socio-Technical Systems Theory which emphasizes the balance between human and technological inputs.

Ethical considerations emerged as a central concern in this research. Transparency and explainability of AI decisions were the most pressing issues identified, followed by data privacy and fairness. This reflects the principles of Responsible AI frameworks, which highlight the need for explainability and accountability in automated systems. Moreover, the findings show that while organizations are taking measures to mitigate bias, such as audits and human oversight, a considerable portion lacks structured frameworks. This gap highlights a divergence between ethical theory and practice.

The study also uncovers mixed perceptions regarding legal and regulatory risks, with many respondents unsure about compliance. This indicates a lack of institutional knowledge, supporting Institutional Theory, which posits that organizations often adopt practices without fully understanding regulatory implications.

Furthermore, there is strong evidence of a growing demand for AI-related training among HR professionals, which reflects the Knowledge-Based View (KBV) that organizational capability building is essential to extract value from new technologies. While most respondents view AI positively, neutral perceptions suggest limited awareness or mixed experiences, demonstrating the Diffusion of Innovations Theory's point that not all adopters transition from early to late adoption at the same pace.

Finally, the research shows optimism regarding AI's future role in recruitment. Respondents anticipate that AI will evolve to address current limitations, suggesting that the technology is at an inflection point where continuous innovation can shift recruitment practices from operational support to strategic decision-making.

Suggestions

Based on these findings, several theoretical implications emerge. First, organizations should expand the application of AI beyond operational tasks toward advanced predictive analytics and candidate assessment systems. This aligns with the TAM framework, which suggests that perceived usefulness drives adoption; increasing awareness of these benefits could accelerate uptake.

Second, fairness and bias mitigation must be approached holistically by combining audits, diverse training data, human oversight, and transparency—reflecting principles of Responsible AI. This supports the view of Stakeholder Theory, which argues that organizational practices must address the concerns of all stakeholders, including candidates.

Third, there is a clear need for enhanced regulatory compliance and awareness. Drawing from Institutional Theory, organizations should not only adopt AI but also institutionalize compliance through clear guidelines, policy frameworks, and regular training sessions.

Fourth, in line with the Knowledge-Based View, organizations should invest in structured training programs for HR professionals to build the necessary skills for ethical and effective use of AI in recruitment. This training would also address the neutral perceptions revealed in the study by improving awareness and understanding of AI capabilities.

Fifth, developers and organizations should work collaboratively to improve transparency and explainability of AI decisions, as emphasized by Ethical AI frameworks. Increased transparency will strengthen trust, a key factor in successful technology adoption according to the Diffusion of Innovations Theory.

Finally, the research underscores the need for continuous innovation and customization of AI tools to meet diverse organizational needs. This reflects the Dynamic Capabilities Theory, which stresses that organizations must adapt and reconfigure technological assets to maintain a competitive advantage in rapidly changing environments.

Conclusion

This research highlights a defining moment in the evolution of recruitment. Artificial Intelligence has shifted from being a futuristic concept to becoming an indispensable partner in modern hiring practices. The findings reveal that while AI excels at streamlining processes, reducing time-to-hire, and enabling data-driven decision-making, it is not yet a substitute for human judgment. Instead, AI is emerging as a powerful co-pilot—enhancing, rather than replacing, the recruiter’s role.

The study also underscores that technology alone cannot guarantee fairness or transparency. True progress lies in how organizations design, govern, and deploy AI systems. The need for robust frameworks, clear regulatory guidance, and continuous skill development is evident. Organizations that combine technological innovation with ethical responsibility will be better positioned to build trust, attract top talent, and maintain accountability.

Looking ahead, the trajectory of AI in recruitment points toward a future where predictive analytics, explainable algorithms, and tailored candidate experiences converge to create a more efficient, inclusive, and human-centric hiring ecosystem. If embraced thoughtfully, AI will not merely transform recruitment processes—it will redefine what talent acquisition means by shifting the focus from transactional hiring to strategic, evidence-based workforce building.

Ultimately, this research shows that the question is no longer whether AI will shape recruitment, but how organizations will shape AI to serve the dual goals of efficiency and equity. By treating AI not just as a tool but as an evolving partner, organizations can unlock its full potential while preserving the distinctly human qualities—judgment, empathy, and ethical reasoning—that make great hiring decisions possible.

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Title: "An Analytical study on Effectiveness of public sector schemes supporting small enterprises "

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Abstract:

This comprehensive study evaluates the efficacy of public sector initiatives designed to support small enterprises, with a focus on their role in promoting entrepreneurship, stimulating job creation, and driving economic growth. Through a thorough review of existing literature and empirical analysis of select schemes, the research reveals that while these initiatives hold significant potential for fostering small enterprise growth, their impact is frequently compromised by challenges in implementation, insufficient funding, and limited outreach to target beneficiaries. The findings underscore the critical need for more targeted, sustainable, and adaptive support mechanisms that can effectively address the unique needs and challenges faced by small enterprises. By identifying key areas for improvement and proposing strategic recommendations, this study aims to contribute to the development of more effective public sector schemes that can enhance the growth and sustainability of small enterprises, ultimately contributing to the broader economic development and job creation objectives.

Keywords: Public sector initiatives, small enterprises, entrepreneurship development, policy, sustainable support mechanisms.

Introduction :

Small enterprises play a vital role in the economic development of a country, contributing significantly to job creation, innovation, and GDP growth. However, these enterprises often face numerous challenges, including limited access to finance, inadequate infrastructure, and lack of technical expertise. To address these challenges, governments around the world have introduced various public sector schemes to support small enterprises. These schemes aim to provide financial assistance, training, and other forms of support to small enterprises, enabling them to overcome their challenges and achieve their full potential.

Despite the importance of small enterprises and the efforts of governments to support them, there is a need for a comprehensive evaluation of the effectiveness of public sector schemes in achieving their objectives. This study aims to fill this gap by conducting an analytical study on the effectiveness of public sector schemes supporting small enterprises. The study will examine the impact of these schemes on the growth and development of small enterprises, identify the challenges and limitations faced by these enterprises in accessing and benefiting from these schemes, and propose recommendations for improvement.

The significance of this study lies in its potential to inform policymakers and stakeholders about the effectiveness of public sector schemes in supporting small enterprises. By identifying the strengths and weaknesses of these schemes, the study can provide valuable insights for improving their design and implementation, ultimately contributing to the growth and development of small enterprises and the broader economy.

1. Review of literature :

Review of literature presented in Part II of this Chapter presents a wide range of studies in the small and the medium enterprise (SME) sector highlighting different dimensions of implementation of various programmes in many parts of the country. The studies reviewed present some of the problems, along with performance region-wise, role of organisations, and directions for the future.

The reviews bring out the need for a deeper study on major managerial problems faced by SMEs from the angle of production, technology, labour, personnel, marketing, transportation, finance, taxation, and other managerial aspects. There is also need to get a feedback on the entrepreneurs' perception of the impact of liberalisation on the region, and specific product groups.

Dhan and Lydall (1961) discussed various economic aspects such as the relationship between capital, employment and output, location and setting up of the estates and entrepreneurs attitude to the estate etc. at the national level. The study evaluated the performance of industrial estate and covered the period of the first eight years after the program being implemented in the country.

NISIET North-Eastern Regional Centre, Gowhati (1974) conducted market survey in Assam and Meghalaya by selecting 358 SSI units. This study was conducted at the request of the NSIC. Actually NSIC proposed to take up the marketing assistance programs in Assam and Meghalaya for carrying meaningful marketing programs. It needs to study the existing marketing problems. The primary objective of the survey is to identify the products along with adequate production base and identification of consortia that may be covered under NSIC marketing assistance program. For this study it was found that around 38 per cent of the units are facing different types of marketing problems.

Brij Bhushan (1997) explained that development of the small-scale industries sector assumes great significance in strengthening economic reforms to solve the chronic problems of unemployment and, poverty. Though industrial development over a period of about five decades of planning era has exhibited noticeable achievements in many areas, it had not been able to absorb a considerable proportion of the urban unemployment youth, nor could it minimize the pressure of less and unproductive manpower on agricultural land. The miseries and social tension of a huge army of employed rural and urban youth remain a cause of great

concern in the context of the new era of economic liberalisation. In this situation only small-scale industries with low capital and high labor intensive technology of production are of great importance from five major angles namely

- (i) employment generation
- (ii) reduction of disparity in the income distribution
- (iii) prevention of abuses of the monopoly of large industrial houses
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- (iv) prevention of rural migration.

Shambhu Ghatak (2007) in his article, “Overview of SMEs in India”, reviews the change in the definitions of micro, small and medium enterprises (MSMEs) over years, and highlights the contribution of micro and small enterprises to the national economy in terms of number of units, investment in fixed assets, gross output, and employment. He also presents the definitions of MSMEs in a few south Asian countries, discusses the role of industrial clusters, and pinpoints the changing role of the sector in the global economic scenario. After the Micro, Small and Medium Enterprises Development (MSMED) Act was passed in June 2006, and became operational from October 2006, the current definitions of the sector have come into vogue.

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2. Methodology :

This study uses a mixed-methods approach to evaluate public sector schemes for small enterprises. A survey of small enterprise owners/managers will collect quantitative data. Interviews with select participants will gather qualitative insights. Stratified random sampling will ensure representative participation. Quantitative data will be analyzed using descriptive statistics and regression analysis. Qualitative data will be analyzed thematically. SPSS software will be used for data analysis. The study aims to assess scheme effectiveness, identify challenges, and inform policy decisions. Data will be collected from primary and secondary sources, including government reports and existing literature.

2.1 Statement of the Problem :

Despite the plethora of public sector schemes aimed at supporting small enterprises, many of these businesses continue to face significant challenges, including limited access to finance, inadequate infrastructure, and lack of technical expertise. There is a need to assess the effectiveness of these schemes in achieving their intended objectives, identifying the gaps and challenges, and exploring opportunities for improvement. This study aims to investigate the impact and effectiveness of public sector schemes on the growth and development of small enterprises, with a focus on identifying best practices and areas for improvement.

2.2 Need for the Study :

The research study on "An Analytical Study on the Effectiveness of Public Sector Schemes Supporting Small Enterprises" is needed for several reasons:

- There is a need to evaluate the effectiveness of public sector schemes in supporting small enterprises and identify areas for improvement.
- The study's findings can inform policy decisions and help policymakers design more effective schemes to support small enterprises.
- Small enterprises play a crucial role in economic development, and this study can help identify ways to support their growth and sustainability.
- The study can help identify the challenges faced by small enterprises in accessing and benefiting from public sector schemes and propose solutions to address these challenges.

- By evaluating the effectiveness of public sector schemes, the study can help promote entrepreneurship and job creation in small enterprises.
- The study's findings can help enhance the design and implementation of public sector schemes, making them more effective in supporting small enterprises.

2.3 Scope of the study :

- Analyzing small enterprises in a specific country or region.
- Examining small enterprises in specific industries or sectors.
- Investigating public sector schemes providing financial, technical, or infrastructure support to small enterprises.
- Analyzing the impact and effectiveness of these schemes over a specific period.
- Using primary and secondary data sources, including surveys, interviews, government reports, and existing literature.

2.4 Research Questions :

1. What are the key public sector schemes available for small enterprises, and what are their objectives?
2. How effective are these schemes in achieving their intended objectives, such as increasing access to finance, improving infrastructure, and enhancing technical expertise?
3. What are the challenges and limitations faced by small enterprises in accessing and benefiting from these schemes?
4. What are the best practices and success stories of public sector schemes for small enterprises, and how can they be replicated or scaled up?

5. What policy and implementation changes can be made to improve the effectiveness of public sector schemes for small enterprises?

3.5 Objectives of the Study :

1. *To evaluate the effectiveness of public sector schemes in supporting small enterprises*.
2. *To identify the most effective types of public sector schemes for small enterprises*.
3. *To examine the challenges and limitations of public sector schemes in supporting small enterprises*.

3.6 (A) Type of Research :

Descriptive-Analytical Research: The study describes the characteristics of public sector schemes and analyzes their impact on small enterprises.

Evaluative Research: The study evaluates the impact and effectiveness of public sector schemes.

(B) Research Design:

"An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises" is a descriptive and analytical design. This design aims to systematically describe the features, accessibility, and implementation of various public sector schemes

targeted at small enterprises while also analyzing their effectiveness in enhancing business performance. The study will involve the collection of both quantitative data (through surveys) and qualitative insights (through interviews or open-ended responses) to understand the impact of schemes such as PMEGP, MUDRA, and Stand-Up India on factors like revenue growth, employment generation, and business sustainability. This approach enables a comprehensive assessment of how well these schemes meet their objectives and support small enterprise development.

3.7 Data Collection :

For the research title "An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises," data collection will involve a combination of primary and secondary sources. Primary data will be collected through surveys and interviews with small enterprise owners/managers who have benefited from public sector schemes. Secondary data will be gathered from government reports, existing literature, policy documents, and databases related to public sector schemes and small enterprise development. This mixed-methods approach will provide a comprehensive understanding of the effectiveness of public sector schemes in supporting small enterprises.

3.8 Sampling design :

"An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises," a stratified random sampling technique will be used to select a representative sample of small enterprises that have benefited from public sector schemes. The sample size will be determined based on research objectives and population size, with inclusion criteria focusing on small enterprises registered in a

specific region/industry and operational for at least 2 years. This approach will ensure a comprehensive understanding of scheme effectiveness.

A) Sampling Plan :

The sampling plan for "An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises" targets small enterprise owners/managers who have leveraged public sector schemes, utilizing stratified random sampling to select a representative sample, ensuring insightful data collection through surveys and interviews.

B) Sampling Method :

Stratified Random Sampling is the most suitable method, as it enables the division of small enterprises into distinct subgroups based on characteristics like industry, location, or scheme type, and then randomly selects samples from each stratum, ensuring representation, reducing bias, and increasing accuracy, ultimately providing a comprehensive understanding of scheme effectiveness across diverse subgroups.

C) Sampling Frame :

The sampling frame for "An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises" comprises a comprehensive list of small enterprises that have benefited from public sector schemes, sourced from government databases, industry associations, and enterprise directories, ensuring a robust and representative pool of participants for the study.

D) Sampling Units :

The sampling units for "An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises" are individual small enterprises that have benefited from public

sector schemes, specifically the owners or managers of these enterprises who will provide insights on the effectiveness of these schemes through surveys or interviews.

3.9 Plan of Analysis :

The analysis will begin with data preparation, where collected data will be cleaned, coded, and analyzed using statistical software. Descriptive statistics will summarize sample characteristics, while inferential statistics (regression analysis, t-tests) will examine the relationship between scheme benefits and enterprise performance (growth, profitability, sustainability).

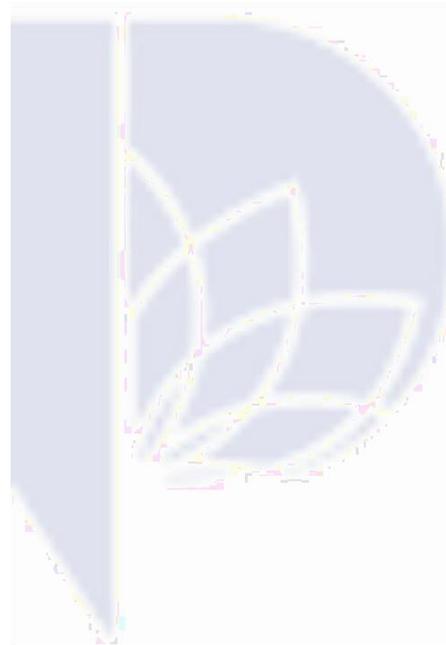
Thematic analysis will be applied to qualitative data from interviews and open-ended questions to gain deeper insights. The results will be interpreted in the context of research objectives, highlighting scheme effectiveness, areas for improvement, and providing actionable recommendations for policymakers and stakeholders.

Findings will be presented in a clear and concise manner, using tables, figures, and narratives to facilitate understanding and inform decision-making.

3.10 Limitations of the Study :

- The sample may not be fully representative of all small enterprises benefiting from public sector schemes, potentially limiting generalizability.
- Data collected through surveys and interviews may be subject to self-reporting bias, where respondents may not accurately report their experiences.
- The effectiveness of public sector schemes may vary significantly across different industries, locations, or enterprise sizes, which may not be fully captured in the study.
- The study may only capture short-term effects of public sector schemes, potentially missing long-term impacts.
- The availability and quality of data on small enterprises and public sector schemes may be limited, affecting the accuracy of the analysis.

- The study's methodology may not be able to establish causality between public sector schemes and small enterprise performance.
- The study may focus on specific regions or sectors, limiting its applicability to other contexts.



Analysis :

Public Sector Schemes Analysis & Interpretation

S. No.	Category	Key Findings & Overall Result
1	Awareness of Schemes	Key Findings: 46.2% somewhat familiar, 40.4% very familiar, 13.4% low awareness Overall Result: High awareness, but small group uninformed
2	Utilization of Schemes	Key Findings: 40.4% occasionally availed, 26.9% frequently, 26.9% aware but not

		<p>availed</p> <p>Overall Result: Engagement exists but inconsistent</p>
3	Effectiveness of Schemes	<p>Key Findings: 42.3% moderately effective, 30.8% very effective, 26.9% slightly effective, 3.8% not effective</p> <p>Overall Result: Generally positive but varies</p>
4	Support Received	<p>Key Findings: 34.6% training, 23.1% financial assistance, 23.1% no support, 19.2% market linkages</p> <p>Overall Result: Training dominates, outreach uneven</p>
5	Impact on Productivity	<p>Key Findings: 50% some improvement, 32.7% great extent, 15.4% little improvement, 3.8% no impact</p> <p>Overall Result: Improves productivity at varying levels</p>
6	Timeliness of Support	<p>Key Findings: 26.9% always timely, 34.6% usually, 36.5% delays, small fraction extreme delays</p> <p>Overall Result: Mostly timely but delays common</p>
7	Employment Generation	<p>Key Findings: 59.6% agree, 21.2% strongly agree, 19% disagree/strongly disagree</p> <p>Overall Result: Strong job creation support</p>
8	Most Beneficial Programs	<p>Key Findings: Skill development 55.8% most beneficial</p> <p>Overall Result: Skill development valued most</p>

9	Financial Schemes Usefulness	Key Findings: Stand-Up India & Credit Guarantee Fund each 30.8% useful Overall Result: Preferred financial schemes for risk/startup support
10	Training Program Effectiveness	Key Findings: 46.2% moderately effective Overall Result: Needs better outcomes
11	Infrastructure Schemes	Key Findings: 50% somewhat helpful Overall Result: Helpful but limited impact
12	Agriculture Schemes	Key Findings: 44.2% most effective in agriculture sector Overall Result: Highly impactful in agriculture
13	Mode of Delivery	Key Findings: 50% prefer online portals Overall Result: Digital access preferred
14	Promotional Initiatives	Key Findings: 30.8% MSME Expo, 30.8% digital marketing support Overall Result: Expos & digital marketing aid market access
15	Challenges in Availing	Key Findings: 53.8% lack awareness, 23.1% complex procedures, 17.3% delays, 5.8% non-cooperation Overall Result: Awareness biggest barrier
16	Accessibility	Key Findings: 50% somewhat accessible, 30.8% very easy, 21.2% difficult, 3.8% not accessible Overall Result: Moderately accessible
17	Transparency	Key Findings: 53.8% moderately transparent, 17.3% highly transparent, 15.4% slightly,

		13.5% not transparent Overall Result: Transparency moderate, some doubts
18	Perception of Awareness	Key Findings: 59.6% agree lack of awareness, 21.2% strongly agree Overall Result: Lack of awareness widespread
19	Limitations in Applying	Key Findings: 42.3% lack documentation, 26.9% eligibility, 21.2% digital access, 9.6% unawareness Overall Result: Documentation & eligibility key barriers
20	Meeting Enterprise Needs	Key Findings: 40.4% partially, 30.8% rarely, 21.2% fully, 7.7% not at all Overall Result: Partially meets needs
21	Follow-up Support	Key Findings: 38.5% occasional, 32.7% regular, 23.1% rare, very few none Overall Result: Follow-up irregular.

□ A majority of respondents are aware of public sector schemes, with 46.2% being somewhat familiar, 40.4% very familiar, and only 13.4% having low awareness.

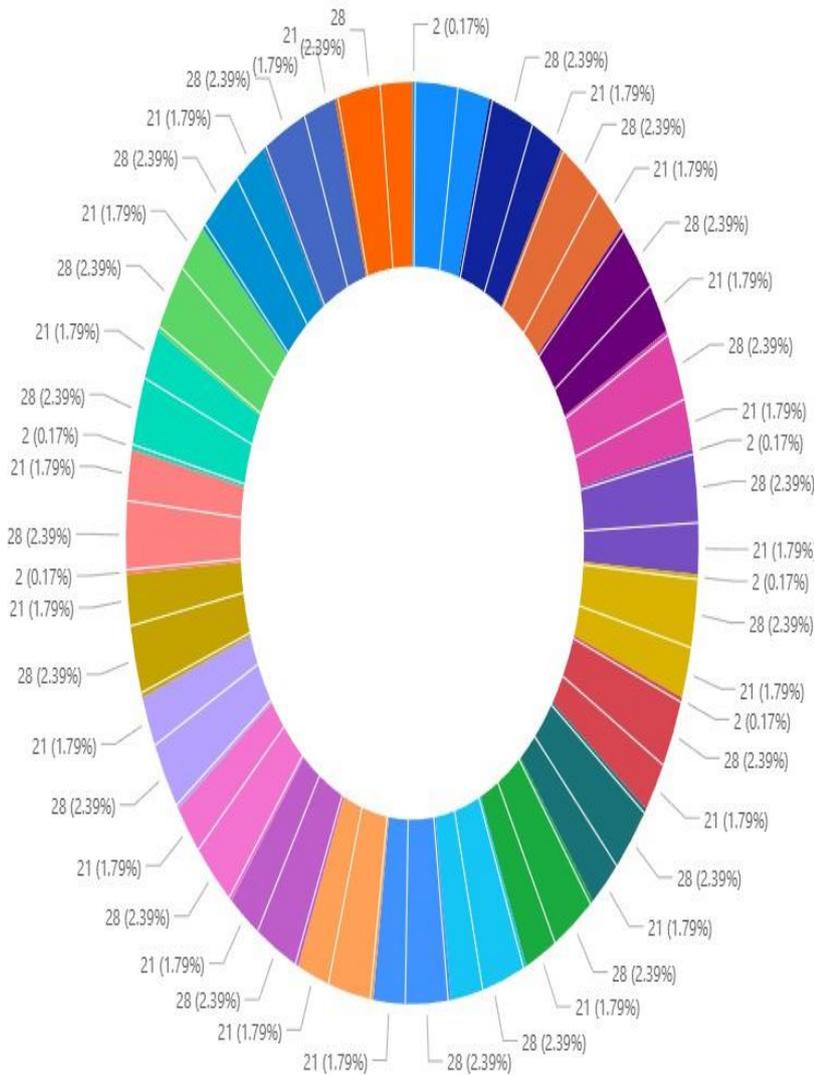
□ Around 40.4% of respondents have availed schemes occasionally, 26.9% frequently, while another 26.9% are aware but have not availed them; only a few remain completely unaware.

□ In terms of effectiveness, 42.3% rated the schemes as moderately effective, 30.8% as very effective, 26.9% as slightly effective, and just 3.8% as not effective.

- When it comes to support received, 34.6% benefited from training and development, 23.1% each from financial assistance or no support, and 19.2% from market linkages.
- Half of the respondents (50%) felt that schemes improved productivity to some extent, 32.7% to a great extent, while 15.4% saw little improvement and 3.8% noticed no impact.
- Regarding timeliness of support, 26.9% reported it was always timely, 34.6% usually timely, 36.5% experienced delays, and a small fraction faced extreme delays.
- Most respondents (59.6%) agree and 21.2% strongly agree that the schemes contribute to employment generation, whereas 19% disagree or strongly disagree.
- Skill development and training programs were the most beneficial (55.8%), while Stand-Up India and the Credit Guarantee Fund (30.8% each) emerged as the most useful financial schemes.
- Stand-Up India and the Credit Guarantee Fund (30.8% each) were rated as highly beneficial, while training programs were mostly considered moderately effective (46.2%).
- Training programs were largely seen as moderately effective (46.2%), while infrastructure-related schemes were found somewhat helpful by half (50%) of the respondents.
- About 50% of respondents considered infrastructure schemes somewhat helpful, while agriculture and allied schemes were rated the most effective (40.4%).
- Agriculture and allied sector schemes were viewed as the most effective (44.2%), with online portals (50%) being the most preferred mode of scheme delivery.
- Half of the respondents (50%) preferred online portals for scheme delivery, while MSME Expo support (30.8%) and digital marketing support (30.8%) were the leading promotional initiatives aiding market access.
- The most common challenge in availing schemes was lack of awareness (53.8%), followed by complex procedures (23.1%) and delays (17.3%).
- Similarly, 53.8% identified lack of awareness as the biggest challenge, while 23.1% pointed to complex procedures, 17.3% to delays in processing, and 5.8% to non-cooperation by officials.

- In terms of accessibility, 50% found the schemes somewhat accessible, 30.8% found them very easy, 21.2% difficult, and 3.8% not accessible at all.
- Regarding transparency, 53.8% believed implementation was moderately transparent, 17.3% highly transparent, 15.4% slightly transparent, and 13.5% not transparent.
- A majority of respondents (59.6%) agreed and 21.2% strongly agreed that there is a lack of awareness about available schemes, while 15.4% disagreed and only 3.8% strongly disagreed.
- The main limitation in applying for schemes was lack of documentation (42.3%), followed by complex eligibility criteria (26.9%), lack of digital access (21.2%), and unawareness (9.6%).
- When asked whether schemes meet enterprise needs, 40.4% said partially, 30.8% rarely, 21.2% fully, and 7.7% not at all.
- Follow-up support was reported as occasional by 38.5%, regular by 32.7%, rare by 23.1%, and non-existent by a small percentage.
- Similarly, most respondents (38.5%) said follow-up occurs occasionally, 32.7% regularly, 23.1% rarely, and very few reported no follow-up at all.

Interpretation :



Attribute

- 1. How familiar are you with public sector schemes for smal...
- 10. How effective are training and development programs ...
- 11. How helpful are infrastructure-based schemes (like indu...
- 12. Which sector-specific schemes are more effective in yo...
- 13. What mode of scheme delivery do you prefer?
- 14. Which promotional schemes have enhanced your mark...
- 15. What are the major challenges in availing public sector ...
- 16. How would you describe the accessibility of public sect...
- 17. How transparent are the processes involved in scheme i...
- 18. Do you think there is a lack of awareness about the avai...
- 19. What limits your ability to apply for government schem...
- 2. Have you ever availed any public sector scheme for your ...
- 20. Do government schemes meet the real needs of your e...
- 21. How often do you receive follow-up or monitoring fro...
- 22. What improvement would you suggest for public sector...
- 3. How would you rate the overall impact of public sector s...
- 4. What kind of support have you received from governme...
- 5. To what extent do public sector schemes help in improvi...
- 6. How timely is the support or funding received through p...
- 7. Are public sector schemes contributing to employment g...
- 8. Which type of scheme has benefited your business the ...
- 9. Which financial support schemes do you consider most ...
- Age:-

- Overall awareness is fairly high, but targeted efforts are needed to reach those with limited familiarity.
- While many businesses engage with schemes, consistent and widespread utilization remains insufficient.
- Respondents generally view public sector schemes positively, though perceptions of effectiveness vary.
- Training is the most widely accessed benefit, yet the absence of support for some highlights uneven program outreach.
- Schemes are seen to enhance productivity, though the level of impact differs across businesses.
- Support is often timely, but frequent delays reduce the efficiency and effectiveness of schemes.
- Public sector schemes are recognized as important for job creation, though a small group perceives limited benefits.
- Entrepreneurs value skill development most, while in financial support, they prefer risk-reducing and startup-friendly schemes.
- Schemes that provide risk coverage and startup support are appreciated, though training programs require improvements for greater impact.
- Training programs need to deliver stronger outcomes, while infrastructure schemes, though useful, could be made more effective.
- Infrastructure support is beneficial but underutilized, whereas agriculture-focused schemes are regarded as highly impactful.
- Strong reliance is observed on agriculture-based schemes, with digital platforms emerging as a preferred mode of delivery.
- Online platforms are considered the most convenient for scheme access, while expos and digital marketing initiatives play a vital role in enhancing market exposure.

- Although half the respondents find schemes somewhat accessible, lack of awareness remains the most critical barrier to effective utilization.
- To improve participation, government schemes must focus on raising awareness and simplifying procedures.
- Even though many find schemes somewhat accessible, significant barriers still restrict full utilization for some beneficiaries.
- While moderate transparency is acknowledged, doubts remain regarding the overall openness of scheme implementation.
- A widespread perception of low awareness emphasizes the need for stronger outreach and effective information dissemination.
- Simplifying documentation requirements and easing eligibility criteria would significantly enhance access to schemes.
- Schemes are beneficial but often fall short of fully meeting enterprise needs.
- Monitoring and follow-up from scheme authorities are inconsistent, with most beneficiaries receiving support only occasionally.
- The follow-up system lacks regularity, resulting in uneven monitoring and support for beneficiaries.

Findings:

The analysis reveals that overall awareness of public sector schemes is fairly high, with most respondents being either very or somewhat familiar, though a small fraction remains less aware. Utilization is not uniform, as many respondents avail schemes only occasionally or not at all despite being aware. Effectiveness is perceived positively by most, but views range from very effective to slightly effective, indicating variation in experiences. Training and development emerge as the most common support received, though some beneficiaries report receiving no assistance, showing gaps in outreach. Respondents agree that schemes contribute to productivity and job creation, yet the level of impact differs, and delays in support delivery reduce overall effectiveness. Skill development and agriculture-based schemes are rated as the most impactful, while financial schemes such as Stand-Up India and Credit Guarantee Fund are valued for reducing risks and supporting startups. Training programs, however, are often seen as only moderately effective, and infrastructure schemes are considered somewhat helpful but underutilized.

Online platforms are the most preferred mode of scheme delivery, supported by MSME expos and digital marketing, highlighting the increasing importance of digital access in outreach. Despite these advantages, barriers persist—chiefly lack of awareness, complex procedures, documentation requirements, and limited digital access. Accessibility is rated moderate by most, but transparency is seen as only partial, with doubts about fairness and openness in implementation. Many respondents feel that schemes meet enterprise needs only partially or rarely, and follow-up mechanisms are inconsistent, with most beneficiaries receiving occasional or irregular support. These findings underline the fact that while schemes have a positive impact, gaps in accessibility, effectiveness, and sustained support hinder their full potential.

Suggestions:

To improve effectiveness, greater emphasis should be placed on targeted awareness campaigns to reach less-informed groups, especially rural and small enterprises. Simplifying processes, reducing documentation burdens, and easing eligibility criteria will encourage wider participation. Training and development programs should be redesigned to focus on practical skill-building and measurable outcomes, while infrastructure support should be scaled up and made more accessible to smaller enterprises. Agriculture-focused schemes should be expanded further, given their proven high impact, while digital platforms should be made more inclusive by addressing the digital divide through offline support and helplines.

Financial schemes like Stand-Up India and the Credit Guarantee Fund can be further strengthened with flexible repayment options and mentorship for startups. Transparency can be improved through public dashboards, progress reports, and grievance redressal mechanisms, ensuring trust and accountability. To better meet enterprise needs, schemes must be customized by sector and business size, offering tailored benefits rather than a one-size-fits-all approach. Additionally, monitoring and follow-up should be standardized and continuous, ensuring beneficiaries receive not only initial support but also sustained guidance. Promoting digital marketing and expos alongside training can help enterprises expand their markets and visibility. Finally, government agencies should focus on building long-term partnerships with entrepreneurs, creating a system where schemes are not only accessible but also drive consistent growth and innovation.

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CONCLUSION:

The analysis highlights that public sector schemes play a significant role in supporting enterprises by enhancing productivity, generating employment, and fostering skill development. While overall awareness is encouraging, irregular utilization, limited transparency, and procedural complexities hinder their full impact. Training and agriculture-based initiatives stand out as the most effective, whereas infrastructure and financial schemes remain moderately beneficial. Challenges such as lack of awareness, documentation issues, and inconsistent follow-up reduce accessibility and long-term sustainability. To maximize their effectiveness, schemes must focus on wider outreach, simplified processes, impactful training, stronger digital delivery mechanisms, and consistent monitoring. Strengthening these aspects will ensure that public sector schemes become more inclusive, accessible, and capable of meeting the diverse needs of beneficiaries.

Title: "An Analytical study on Effectiveness of public sector schemes supporting small enterprises "

Yashika K S (1 MBA), Kusuma C (1 MBA)

Under the guidance of

Dr. Padmashree V

Assistant Professor

Abstract:

This comprehensive study evaluates the efficacy of public sector initiatives designed to support small enterprises, with a focus on their role in promoting entrepreneurship, stimulating job creation, and driving economic growth. Through a thorough review of existing literature and empirical analysis of select schemes, the research reveals that while these initiatives hold significant potential for fostering small enterprise growth, their impact is frequently compromised by challenges in implementation, insufficient funding, and limited outreach to target beneficiaries. The findings underscore the critical need for more targeted, sustainable, and adaptive support mechanisms that can effectively address the unique needs and challenges faced by small enterprises. By identifying key areas for improvement and proposing strategic recommendations, this study aims to contribute to the development of more effective public sector schemes that can enhance the growth and sustainability of small enterprises, ultimately contributing to the broader economic development and job creation objectives.

Keywords: Public sector initiatives, small enterprises, entrepreneurship development, policy, sustainable support mechanisms.

Introduction :

Small enterprises play a vital role in the economic development of a country, contributing significantly to job creation, innovation, and GDP growth. However, these enterprises often face numerous challenges, including limited access to finance, inadequate infrastructure, and lack of technical expertise. To address these challenges, governments around the world have introduced various public sector schemes to support small enterprises. These schemes aim to provide financial assistance, training, and other forms of support to small enterprises, enabling them to overcome their challenges and achieve their full potential.

Despite the importance of small enterprises and the efforts of governments to support them, there is a need for a comprehensive evaluation of the effectiveness of public sector schemes in achieving their objectives. This study aims to fill this gap by conducting an analytical study on the effectiveness of public sector schemes supporting small enterprises. The study will examine the impact of these schemes on the growth and development of small enterprises, identify the challenges and limitations faced by these enterprises in accessing and benefiting from these schemes, and propose recommendations for improvement.

The significance of this study lies in its potential to inform policymakers and stakeholders about the effectiveness of public sector schemes in supporting small enterprises. By identifying the strengths and weaknesses of these schemes, the study can provide valuable insights for improving their design and implementation, ultimately contributing to the growth and development of small enterprises and the broader economy.

1. Review of literature :

Review of literature presented in Part II of this Chapter presents a wide range of studies in the small and the medium enterprise (SME) sector highlighting different dimensions of implementation of various programs in many parts of the country. The studies reviewed present some of the problems, along with performance region-wise, role of organizations, and directions for the future.

The reviews bring out the need for a deeper study on major managerial problems faced by SMEs from the angle of production, technology, labor, personnel, marketing, transportation, finance, taxation, and other managerial aspects. There is also need to get a feedback on the entrepreneurs' perception of the impact of liberalization on the region, and specific product groups.

Dhan and Lydall (1961) discussed various economic aspects such as the relationship between capital, employment and output, location and setting up of the estates and entrepreneurs attitude to the estate etc. at the national level. The study evaluated the performance of industrial estate and covered the period of the first eight years after the program being implemented in the country.

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This study uses a mixed-methods approach to evaluate public sector schemes for small enterprises. A survey of small enterprise owners/managers will collect quantitative data. Interviews with select participants will gather qualitative insights. Stratified random sampling will ensure representative participation. Quantitative data will be analyzed using descriptive statistics and regression analysis. Qualitative data will be analyzed thematically. SPSS software will be used for data analysis. The study aims to assess scheme effectiveness, identify challenges, and inform policy decisions. Data will be collected from primary and secondary sources, including government reports and existing literature.

2.1 Statement of the Problem :

Despite the plethora of public sector schemes aimed at supporting small enterprises, many of these businesses continue to face significant challenges, including limited access to finance, inadequate infrastructure, and lack of technical expertise. There is a need to assess the effectiveness of these schemes in achieving their intended objectives, identifying the gaps and challenges, and exploring opportunities for improvement. This study aims to investigate the impact and effectiveness of public sector schemes on the growth and development of small enterprises, with a focus on identifying best practices and areas for improvement.

2.2 Need for the Study :

The research study on "An Analytical Study on the Effectiveness of Public Sector Schemes Supporting Small Enterprises" is needed for several reasons:

- There is a need to evaluate the effectiveness of public sector schemes in supporting small enterprises and identify areas for improvement.
- The study's findings can inform policy decisions and help policymakers design more effective schemes to support small enterprises.
- Small enterprises play a crucial role in economic development, and this study can help identify ways to support their growth and sustainability.

- The study can help identify the challenges faced by small enterprises in accessing and benefiting from public sector schemes and propose solutions to address these challenges.
- By evaluating the effectiveness of public sector schemes, the study can help promote entrepreneurship and job creation in small enterprises.
- The study's findings can help enhance the design and implementation of public sector schemes, making them more effective in supporting small enterprises.

2.3 Scope of the study :

- Analyzing small enterprises in a specific country or region.
- Examining small enterprises in specific industries or sectors.
- Investigating public sector schemes providing financial, technical, or infrastructure support to small enterprises.
- Analyzing the impact and effectiveness of these schemes over a specific period.
- Using primary and secondary data sources, including surveys, interviews, government reports, and existing literature.

2.4 Research Questions :

1. What are the key public sector schemes available for small enterprises, and what are their objectives?
2. How effective are these schemes in achieving their intended objectives, such as increasing access to finance, improving infrastructure, and enhancing technical expertise?
3. What are the challenges and limitations faced by small enterprises in accessing and benefiting from these schemes?
4. What are the best practices and success stories of public sector schemes for small enterprises, and how can they be replicated or scaled up?
5. What policy and implementation changes can be made to improve the effectiveness of public sector schemes for small enterprises?

3.5 Objectives of the Study :

1. *To evaluate the effectiveness of public sector schemes in supporting small enterprises*.
2. *To identify the most effective types of public sector schemes for small enterprises*.
3. *To examine the challenges and limitations of public sector schemes in supporting small enterprises*.

3.6 (A) Type of Research :

Descriptive-Analytical Research: The study describes the characteristics of public sector schemes and analyzes their impact on small enterprises.

Evaluative Research: The study evaluates the impact and effectiveness of public sector schemes.

(B) Research design :

"An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises" is a descriptive and analytical design. This design aims to systematically describe the features, accessibility, and implementation of various public sector schemes targeted at small enterprises while also analyzing their effectiveness in enhancing business performance. The study will involve the collection of both quantitative data (through surveys) and qualitative insights (through interviews or open-ended responses) to understand the impact of schemes such as PMEGP, MUDRA, and Stand-Up India on factors like revenue growth, employment generation, and business sustainability. This approach enables a comprehensive assessment of how well these schemes meet their objectives and support small enterprise development.

3.7 Data Collection :

For the research title "An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises," data collection will involve a combination of primary and secondary sources. Primary data will be collected through surveys and interviews with small enterprise owners/managers who have benefited from public sector schemes. Secondary data will be gathered from government reports, existing literature, policy documents, and databases related to public sector schemes and small enterprise development. This mixed-methods approach will provide a comprehensive understanding of the effectiveness of public sector schemes in supporting small enterprises.

3.8 Sampling design :

"An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises," a stratified random sampling technique will be used to select a representative sample of small enterprises that have benefited from public sector schemes. The sample size will be determined based on research objectives and population size, with inclusion criteria focusing on small enterprises registered in a specific region/industry and operational for at least 2 years. This approach will ensure a comprehensive understanding of scheme effectiveness.

A) Sampling Plan :

The sampling plan for "An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises" targets small enterprise owners/managers who have leveraged public sector schemes, utilizing stratified random sampling to select a representative sample, ensuring insightful data collection through surveys and interviews.

B) Sampling Method :

Stratified Random Sampling is the most suitable method, as it enables the division of small enterprises into distinct subgroups based on characteristics like industry, location, or scheme type, and then randomly selects samples from each stratum, ensuring representation, reducing bias, and increasing accuracy, ultimately providing a comprehensive understanding of scheme effectiveness across diverse subgroups.

C) Sampling Frame :

The sampling frame for "An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises" comprises a comprehensive list of small enterprises that have benefited from public sector schemes, sourced from government databases, industry associations, and enterprise directories, ensuring a robust and representative pool of participants for the study.

D) Sampling Units :

The sampling units for "An Analytical Study on Effectiveness of Public Sector Schemes Supporting Small Enterprises" are individual small enterprises that have benefited from public sector schemes, specifically the owners or managers of these enterprises who will provide insights on the effectiveness of these schemes through surveys or interviews.

3.9 Plan of Analysis :

The analysis will begin with data preparation, where collected data will be cleaned, coded, and analyzed using statistical software. Descriptive statistics will summarize sample characteristics, while inferential statistics (regression analysis, t-tests) will examine the relationship between scheme benefits and enterprise performance (growth, profitability, sustainability).

Thematic analysis will be applied to qualitative data from interviews and open-ended questions to gain deeper insights. The results will be interpreted in the context of research objectives, highlighting scheme effectiveness, areas for improvement, and providing actionable recommendations for policymakers and stakeholders.

Findings will be presented in a clear and concise manner, using tables, figures, and narratives to facilitate understanding and inform decision-making.

3.10 Limitations of the Study :

- The sample may not be fully representative of all small enterprises benefiting from public sector schemes, potentially limiting generalizability.
- Data collected through surveys and interviews may be subject to self-reporting bias, where respondents may not accurately report their experiences.
- The effectiveness of public sector schemes may vary significantly across different industries, locations, or enterprise sizes, which may not be fully captured in the study.
- The study may only capture short-term effects of public sector schemes, potentially missing long-term impacts.
- The availability and quality of data on small enterprises and public sector schemes may be limited, affecting the accuracy of the analysis.
- The study's methodology may not be able to establish causality between public sector schemes and small enterprise performance.
- The study may focus on specific regions or sectors, limiting its applicability to other contexts.

Analysis :

Public Sector Schemes Analysis & Interpretation

S. No.	Category	Key Findings & Overall Result
1	Awareness of Schemes	Key Findings: 46.2% somewhat familiar, 40.4% very familiar, 13.4% low awareness Overall Result: High awareness, but small group uninformed
2	Utilization of Schemes	Key Findings: 40.4% occasionally availed, 26.9% frequently, 26.9% aware but not availed Overall Result: Engagement exists but inconsistent
3	Effectiveness of Schemes	Key Findings: 42.3% moderately effective, 30.8% very effective, 26.9% slightly effective, 3.8% not effective Overall Result: Generally positive but varies
4	Support Received	Key Findings: 34.6% training, 23.1% financial assistance, 23.1% no support, 19.2% market linkages Overall Result: Training dominates, outreach uneven
5	Impact on Productivity	Key Findings: 50% some improvement, 32.7% great extent, 15.4% little improvement, 3.8% no impact Overall Result: Improves productivity at varying levels
6	Timeliness of Support	Key Findings: 26.9% always timely, 34.6% usually, 36.5% delays, small fraction extreme

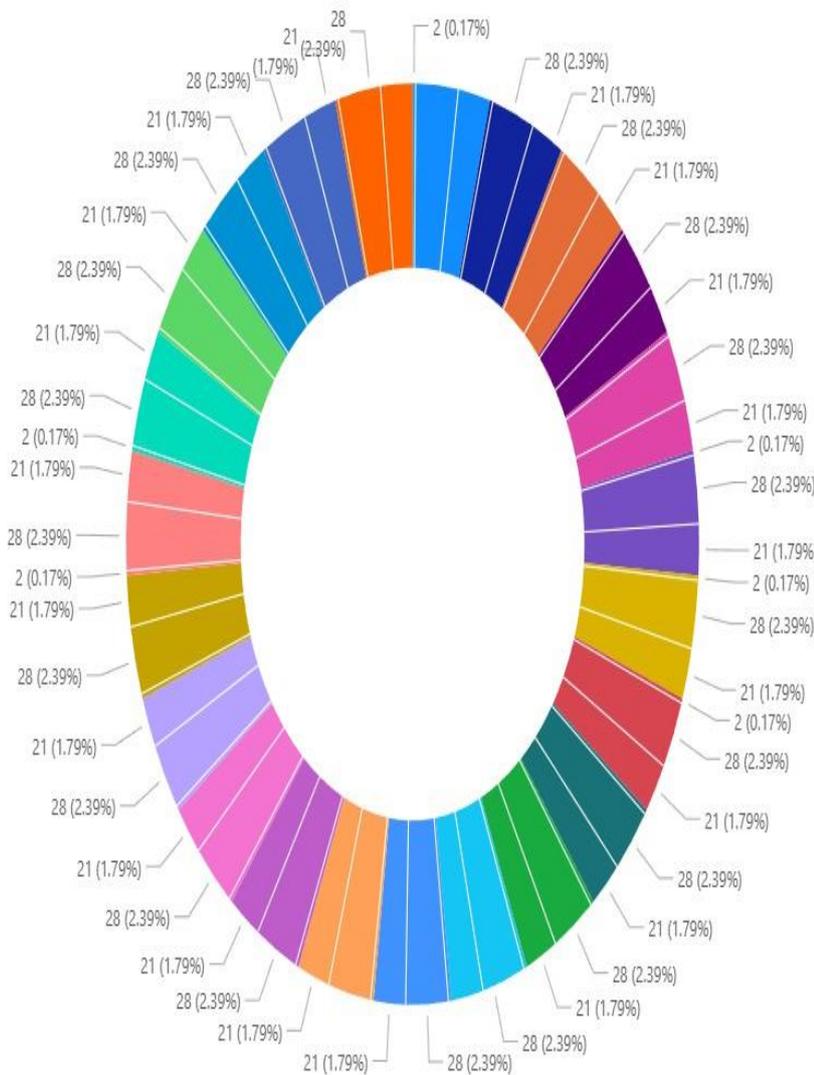
		delays Overall Result: Mostly timely but delays common
7	Employment Generation	Key Findings: 59.6% agree, 21.2% strongly agree, 19% disagree/strongly disagree Overall Result: Strong job creation support
8	Most Beneficial Programs	Key Findings: Skill development 55.8% most beneficial Overall Result: Skill development valued most
9	Financial Schemes Usefulness	Key Findings: Stand-Up India & Credit Guarantee Fund each 30.8% useful Overall Result: Preferred financial schemes for risk/startup support
10	Training Program Effectiveness	Key Findings: 46.2% moderately effective Overall Result: Needs better outcomes
11	Infrastructure Schemes	Key Findings: 50% somewhat helpful Overall Result: Helpful but limited impact
12	Agriculture Schemes	Key Findings: 44.2% most effective in agriculture sector Overall Result: Highly impactful in agriculture
13	Mode of Delivery	Key Findings: 50% prefer online portals Overall Result: Digital access preferred
14	Promotional Initiatives	Key Findings: 30.8% MSME Expo, 30.8% digital marketing support Overall Result: Expos & digital marketing aid market access

15	Challenges in Availing	Key Findings: 53.8% lack awareness, 23.1% complex procedures, 17.3% delays, 5.8% non-cooperation Overall Result: Awareness biggest barrier
16	Accessibility	Key Findings: 50% somewhat accessible, 30.8% very easy, 21.2% difficult, 3.8% not accessible Overall Result: Moderately accessible
17	Transparency	Key Findings: 53.8% moderately transparent, 17.3% highly transparent, 15.4% slightly, 13.5% not transparent Overall Result: Transparency moderate, some doubts
18	Perception of Awareness	Key Findings: 59.6% agree lack of awareness, 21.2% strongly agree Overall Result: Lack of awareness widespread
19	Limitations in Applying	Key Findings: 42.3% lack documentation, 26.9% eligibility, 21.2% digital access, 9.6% unawareness Overall Result: Documentation & eligibility key barriers
20	Meeting Enterprise Needs	Key Findings: 40.4% partially, 30.8% rarely, 21.2% fully, 7.7% not at all Overall Result: Partially meets needs
21	Follow-up Support	Key Findings: 38.5% occasional, 32.7% regular, 23.1% rare, very few none Overall Result: Follow-up irregular.

- A majority of respondents are aware of public sector schemes, with 46.2% being somewhat familiar, 40.4% very familiar, and only 13.4% having low awareness.
- Around 40.4% of respondents have availed schemes occasionally, 26.9% frequently, while another 26.9% are aware but have not availed them; only a few remain completely unaware.
- In terms of effectiveness, 42.3% rated the schemes as moderately effective, 30.8% as very effective, 26.9% as slightly effective, and just 3.8% as not effective.
- When it comes to support received, 34.6% benefited from training and development, 23.1% each from financial assistance or no support, and 19.2% from market linkages.
- Half of the respondents (50%) felt that schemes improved productivity to some extent, 32.7% to a great extent, while 15.4% saw little improvement and 3.8% noticed no impact.
- Regarding timeliness of support, 26.9% reported it was always timely, 34.6% usually timely, 36.5% experienced delays, and a small fraction faced extreme delays.
- Most respondents (59.6%) agree and 21.2% strongly agree that the schemes contribute to employment generation, whereas 19% disagree or strongly disagree.
- Skill development and training programs were the most beneficial (55.8%), while Stand-Up India and the Credit Guarantee Fund (30.8% each) emerged as the most useful financial schemes.
- Stand-Up India and the Credit Guarantee Fund (30.8% each) were rated as highly beneficial, while training programs were mostly considered moderately effective (46.2%).
- Training programs were largely seen as moderately effective (46.2%), while infrastructure-related schemes were found somewhat helpful by half (50%) of the respondents.
- About 50% of respondents considered infrastructure schemes somewhat helpful, while agriculture and allied schemes were rated the most effective (40.4%).
- Agriculture and allied sector schemes were viewed as the most effective (44.2%), with online portals (50%) being the most preferred mode of scheme delivery.

- Half of the respondents (50%) preferred online portals for scheme delivery, while MSME Expo support (30.8%) and digital marketing support (30.8%) were the leading promotional initiatives aiding market access.
- The most common challenge in availing schemes was lack of awareness (53.8%), followed by complex procedures (23.1%) and delays (17.3%).
- Similarly, 53.8% identified lack of awareness as the biggest challenge, while 23.1% pointed to complex procedures, 17.3% to delays in processing, and 5.8% to non-cooperation by officials.
- In terms of accessibility, 50% found the schemes somewhat accessible, 30.8% found them very easy, 21.2% difficult, and 3.8% not accessible at all.
- Regarding transparency, 53.8% believed implementation was moderately transparent, 17.3% highly transparent, 15.4% slightly transparent, and 13.5% not transparent.
- A majority of respondents (59.6%) agreed and 21.2% strongly agreed that there is a lack of awareness about available schemes, while 15.4% disagreed and only 3.8% strongly disagreed.
- The main limitation in applying for schemes was lack of documentation (42.3%), followed by complex eligibility criteria (26.9%), lack of digital access (21.2%), and unawareness (9.6%).
- When asked whether schemes meet enterprise needs, 40.4% said partially, 30.8% rarely, 21.2% fully, and 7.7% not at all.
- Follow-up support was reported as occasional by 38.5%, regular by 32.7%, rare by 23.1%, and non-existent by a small percentage.
- Similarly, most respondents (38.5%) said follow-up occurs occasionally, 32.7% regularly, 23.1% rarely, and very few reported no follow-up at all.

Interpretation :



Attribute

- 1. How familiar are you with public sector schemes for small...
- 2. Have you ever availed any public sector scheme for your ...
- 3. How would you rate the overall impact of public sector s...
- 4. What kind of support have you received from governme...
- 5. To what extent do public sector schemes help in improvi...
- 6. How timely is the support or funding received through p...
- 7. Are public sector schemes contributing to employment g...
- 8. Which type of scheme has benefited your business the ...
- 9. Which financial support schemes do you consider most ...
- Age:-
- 10. How effective are training and development programs ...
- 11. How helpful are infrastructure-based schemes (like indu...
- 12. Which sector-specific schemes are more effective in yo...
- 13. What mode of scheme delivery do you prefer?
- 14. Which promotional schemes have enhanced your mark...
- 15. What are the major challenges in availing public sector ...
- 16. How would you describe the accessibility of public sect...
- 17. How transparent are the processes involved in scheme i...
- 18. Do you think there is a lack of awareness about the avai...
- 19. What limits your ability to apply for government schem...
- 20. Do government schemes meet the real needs of your e...
- 21. How often do you receive follow-up or monitoring fro...
- 22. What improvement would you suggest for public sector...

- Overall awareness is fairly high, but targeted efforts are needed to reach those with limited familiarity.
- While many businesses engage with schemes, consistent and widespread utilization remains insufficient.
- Respondents generally view public sector schemes positively, though perceptions of effectiveness vary.
- Training is the most widely accessed benefit, yet the absence of support for some highlights uneven program outreach.
- Schemes are seen to enhance productivity, though the level of impact differs across businesses.
- Support is often timely, but frequent delays reduce the efficiency and effectiveness of schemes.
- Public sector schemes are recognized as important for job creation, though a small group perceives limited benefits.
- Entrepreneurs value skill development most, while in financial support, they prefer risk-reducing and startup-friendly schemes.
- Schemes that provide risk coverage and startup support are appreciated, though training programs require improvements for greater impact.
- Training programs need to deliver stronger outcomes, while infrastructure schemes, though useful, could be made more effective.
- Infrastructure support is beneficial but underutilized, whereas agriculture-focused schemes are regarded as highly impactful.
- Strong reliance is observed on agriculture-based schemes, with digital platforms emerging as a preferred mode of delivery.
- Online platforms are considered the most convenient for scheme access, while expos and digital marketing initiatives play a vital role in enhancing market exposure.

- Although half the respondents find schemes somewhat accessible, lack of awareness remains the most critical barrier to effective utilization.
- To improve participation, government schemes must focus on raising awareness and simplifying procedures.
- Even though many find schemes somewhat accessible, significant barriers still restrict full utilization for some beneficiaries.
- While moderate transparency is acknowledged, doubts remain regarding the overall openness of scheme implementation.
- A widespread perception of low awareness emphasizes the need for stronger outreach and effective information dissemination.
- Simplifying documentation requirements and easing eligibility criteria would significantly enhance access to schemes.
- Schemes are beneficial but often fall short of fully meeting enterprise needs.
- Monitoring and follow-up from scheme authorities are inconsistent, with most beneficiaries receiving support only occasionally.
- The follow-up system lacks regularity, resulting in uneven monitoring and support for beneficiaries.

Findings:

The analysis reveals that overall awareness of public sector schemes is fairly high, with most respondents being either very or somewhat familiar, though a small fraction remains less aware. Utilization is not uniform, as many respondents avail schemes only occasionally or not at all despite being aware. Effectiveness is perceived positively by most, but views range from very effective to slightly effective, indicating variation in experiences. Training and development emerge as the most common support received, though some beneficiaries report receiving no assistance, showing gaps in outreach. Respondents agree that schemes contribute to productivity and job creation, yet the level of impact differs, and delays in support delivery reduce overall effectiveness. Skill development and agriculture-based schemes are rated as the most impactful, while financial schemes such as Stand-Up India and Credit Guarantee Fund are valued for reducing risks and supporting startups. Training programs, however, are often seen as only moderately effective, and infrastructure schemes are considered somewhat helpful but underutilized.

Online platforms are the most preferred mode of scheme delivery, supported by MSME expos and digital marketing, highlighting the increasing importance of digital access in outreach. Despite these advantages, barriers persist—chiefly lack of awareness, complex procedures, documentation requirements, and limited digital access. Accessibility is rated moderate by most, but transparency is seen as only partial, with doubts about fairness and openness in implementation. Many respondents feel that schemes meet enterprise needs only partially or rarely, and follow-up mechanisms are inconsistent, with most beneficiaries receiving occasional or irregular support. These findings underline the fact that while schemes have a positive impact, gaps in accessibility, effectiveness, and sustained support hinder their full potential.

Suggestions:

To improve effectiveness, greater emphasis should be placed on targeted awareness campaigns to reach less-informed groups, especially rural and small enterprises. Simplifying processes, reducing documentation burdens, and easing eligibility criteria will encourage wider participation. Training and development programs should be redesigned to focus on practical skill-building and measurable outcomes, while infrastructure support should be scaled up and made more accessible to smaller enterprises. Agriculture-focused schemes should be expanded further, given their proven high impact, while digital platforms should be made more inclusive by addressing the digital divide through offline support and helplines.

Financial schemes like Stand-Up India and the Credit Guarantee Fund can be further strengthened with flexible repayment options and mentorship for startups. Transparency can be improved through public dashboards, progress reports, and grievance redressal mechanisms, ensuring trust and accountability. To better meet enterprise needs, schemes must be customized by sector and business size, offering tailored benefits rather than a one-size-fits-all approach. Additionally, monitoring and follow-up should be standardized and continuous, ensuring beneficiaries receive not only initial support but also sustained guidance. Promoting digital marketing and expos alongside training can help enterprises expand their markets and visibility. Finally, government agencies should focus on building long-term partnerships with entrepreneurs, creating a system where schemes are not only accessible but also drive consistent growth and innovation.

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Nagayya, D. and B. Sobha Rani (2008), “Financing Small and Medium Enterprises in the Emerging Scenario,” *GITAM Journal of Management*, Vol.6 No.3, July - September, pp. 127-152.

CONCLUSION:

The analysis highlights that public sector schemes play a significant role in supporting enterprises by enhancing productivity, generating employment, and fostering skill development. While overall awareness is encouraging, irregular utilization, limited transparency, and procedural complexities hinder their full impact. Training and agriculture-based initiatives stand out as the most effective, whereas infrastructure and financial schemes remain moderately beneficial. Challenges such as lack of awareness, documentation issues, and inconsistent follow-up reduce accessibility and long-term sustainability. To maximize their effectiveness, schemes must focus on wider outreach, simplified processes, impactful training, stronger digital delivery mechanisms, and consistent monitoring. Strengthening these aspects will ensure that public sector schemes become more inclusive, accessible, and capable of meeting the diverse needs of beneficiaries.

“Artificial Intelligence for Workplace Well-Being: A Comprehensive Review”

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Abstract:

The role of Artificial Intelligence (AI) in enhancing employee well-being has garnered significant attention in recent years, as organizations seek innovative solutions to improve workplace mental health. This review explores how Artificial Intelligence (AI) can improve employee well-being by supporting mental health in the workplace. AI tools like sentiment analysis, stress detection, and mental health chatbots are being used to monitor and assist employees in real-time. By analysing data, AI can identify risks to well-being and provide personalized support, creating a healthier work environment. The paper highlights the benefits, challenges, and ethical issues of using AI for employee well-being and its impact on mental health in organizations.

Keywords: Artificial Intelligence, Employee Well-being, Mental Health, Workplace Wellness, Sentiment Analysis, Stress Detection, Personalized Wellness, Employee Support, Ethics

Introduction:

Employee well-being is essential for improving organizational performance, as it directly impacts productivity, engagement, and overall success. When employees are healthy, both physically and mentally, they are more focused, efficient, and creative in their work.

Promoting well-being can reduce stress, lower absenteeism, and help maintain a smooth workflow. It also lowers turnover rates because employees who feel valued are more likely to stay with the company, reducing recruitment and training costs. High job satisfaction from

well-being efforts boosts engagement, loyalty, and teamwork, which drives innovation and better performance.

Review of literature

- 1. Mishchuk et al., (2023).** The impact on labour markets is expected to be considerable, in terms of job destruction (robotics, automation and artificial intelligence (AI) making several types of jobs redundant and several types of job tasks irrelevant), highly routinized jobs, and job creation (newly emerging professions requiring new skills' acquisitions)
- 2. Kang, H., & Yu, E. (2020).** The impact of artificial intelligence on employee health and well-being: Evidence from wearable health monitoring devices and how wearable devices powered by AI are used to monitor and promote the health of employees, providing insights into stress management, physical health, and overall well-being
- 3. Peffers, C., & O'Reilly, D. (2019).** Wearable technology and its implications for employee well-being in the workplace. the adoption of AI-powered wearable technology that tracks health metrics such as heart rate variability, and its effects on employees' psychological and physical well-being.
- 4. Fitzpatrick, K. K., Darcy, A., & Vierhile, M. (2017).** Delivering Cognitive Behavioral Therapy to treat depression and anxiety in youth via mobile and digital technologies AI-driven systems for delivering cognitive behavioural therapy (CBT) to employees dealing with stress or anxiety. These systems, such as chatbots and virtual therapists, have shown potential in workplace settings for mental health support.
- 5. Mirkovic, J., & Furlan, N. (2018).** Employee monitoring and its effects on well-being: Ethical concerns with AI surveillance They explores the ethical challenges related to AI surveillance in the workplace and how constant monitoring can create stress and negatively impact employee well-being.
- 6. Miner, A. S., Milstein, A., & Schueller, S. M. (2016).** Talking therapy: A large-scale study of the effectiveness of an AI-powered chatbot for mental health support the effectiveness of AI-powered chatbots in providing low-cost, accessible mental health support to employees, reducing feelings of burnout and anxiety

Research gap

Despite the There is a gap in understanding the long-term impact of AI tools on employee mental health. Most research focuses on the development and immediate benefits of AI-driven interventions, but little attention is given to how these tools affect employees in different work environments and cultures. Additionally, challenges around integrating AI with existing mental health systems, data privacy, trust, and algorithmic bias are not fully explored. Ethical issues, such as potential stigmatization or misdiagnosis, also need more attention. More research is needed to assess both the effectiveness and ethical implications of AI in employee well-being.

Research Aim and Objectives

Aim: To explore how AI technologies can improve mental health and well-being in the workplace.

Objectives:

Review AI tools used for mental health monitoring.

Assess the impact of AI-driven interventions on employee well-being.

Identify challenges and ethical considerations in using AI for mental health.

Research methodology

This research will focus on a case study that highlights the impact of AI on employee well-being. One relevant case is **Unilever's Implementation of AI Forecasting Tools**, which showcases how AI was integrated into the company's operations to enhance both efficiency and employee satisfaction

CASE STUDY: Implementation of AI Forecasting Tools at Unilever.

In recent years, Unilever, one of the world's leading consumer goods companies, has focused on leveraging technology to enhance operational efficiency and employee well-being. In particular, the integration of artificial intelligence (AI) into their forecasting processes has proven transformative. This case study outlines how AI forecasting tools were implemented, the challenges faced, and the resulting benefits for both the organization and its employees, particularly in enhancing work-life balance.

Challenges:

Overloading Employees: Managers struggled to predict workloads, leading to employee burnout during peak times.

Data Silos: Data was scattered, making it hard for managers to get a complete view of workloads and resources.

Inefficient Resource Allocation: Manual forecasting created imbalances, with some teams overworked and others underutilized.

Implementation of AI Tools:

AI Integration: Unilever adopted advanced AI forecasting tools that used historical data, market trends, and other factors to predict workloads accurately.

Training: Employees learned to use the AI tools, interpret data, and apply insights to decision-making.

Real-Time Analytics: AI provided managers with instant visibility into workload predictions and resource allocation.

Feedback Loops: Employees reported workloads and stress levels, refining AI predictions for greater accuracy.

Results:

Improved Workload Management: Managers could anticipate peak times, redistribute tasks more effectively, and reduce employee overload.

Enhanced Work-Life Balance: With better workload management, employees experienced less stress, leading to improved well-being and job satisfaction.

Increased Productivity: A more balanced distribution of tasks allowed employees to focus without feeling overwhelmed, boosting overall efficiency.

Data-Driven Decisions: Real-time data allowed for more agile decision-making and strategic planning.

Cultural Shift: AI tools fostered collaboration and transparency, with employees feeling their feedback was valued.

Higher Employee Retention: A balanced work environment led to increased employee satisfaction, reducing turnover rates.

Conclusion: By implementing AI forecasting tools, Unilever successfully addressed workload unpredictability, improved resource allocation, and enhanced both operational efficiency and employee well-being. This initiative not only boosted productivity but also fostered a positive workplace culture, positioning Unilever as a leader in employee engagement. The case study highlights how AI can be leveraged not just for operational gain but also to improve the employee experience, setting a benchmark for future workforce management practices.

Future Directions

Future research should focus on longitudinal studies to evaluate the long-term effects of AI initiatives on employee well-being. Additionally, it would be beneficial to explore the perceptions of employees regarding AI's role in their work lives and well-being.

Conclusion

AI has the potential to significantly enhance employee well-being by providing resources that support mental health, improve work-life balance, increase productivity, and offer personalized development opportunities. However, organizations must navigate the associated challenges to harness the full potential of these technologies. A balanced approach—integrating human empathy with AI efficiency—may serve as the most effective strategy for enhancing employee well-being in the future.

AI plays a key role in supporting employee well-being by offering tools that can improve health, reduce stress, and create a positive work environment. AI-driven solutions, such as personalized wellness programs, mental health support, and workload management, can help employees stay healthy and productive. These AI applications contribute to lower absenteeism, improved job satisfaction, and a supportive work culture. By investing in AI for employee well-being, companies not only improve their employees' health and happiness but

also benefit from better overall performance and reduced costs. AI, therefore, is an important tool in enhancing employee well-being, leading to sustained organizational success.

Mental Health Support

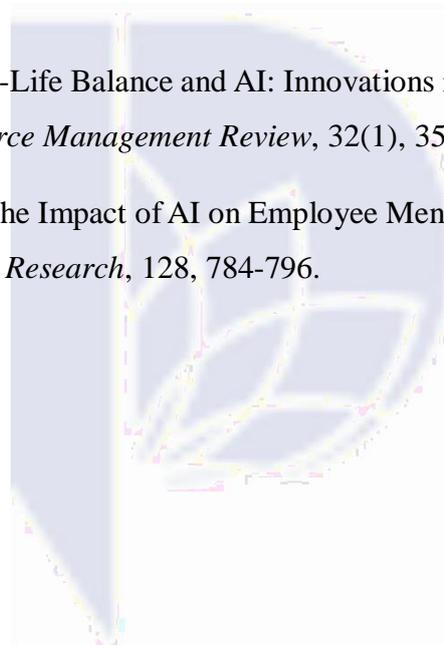
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**“Bridging Humans and Machines: The Significance of Communication
in AI Prompting”**

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Abstract

Artificial Intelligence (AI) has emerged as a transformative force across diverse domains, reshaping how humans interact with information, knowledge creation, and communication. Central to this interaction is the use of AI prompts—the structured or unstructured instructions through which human intentions are translated into machine-understandable language. The process of crafting prompts is not merely a technical task but an act of communication that requires precision, contextual awareness, and linguistic clarity. This research article investigates the significance of communication in the design, refinement, and optimization of AI prompts, emphasizing how the principles of communication theory, language use, and discourse analysis can be applied to enhance human–AI interaction. Drawing from both theoretical frameworks and practical case studies, the article underscores that effective communication in prompt formulation directly influences the accuracy, relevance, and creativity of AI-generated outputs. Furthermore, the study highlights how prompt-based communication expands the scope of AI applications across fields such as education, literature, research, and professional practice. In education, AI-supported communication fosters innovative teaching and learning strategies; in literature and creative writing, it provides new avenues for expression and interpretation; in research, it aids critical inquiry and data-driven analysis; and in professional settings, it improves productivity and problem-solving.

By bridging the gap between linguistic practice and technological application, this study provides a comprehensive understanding of how communication functions as both a tool and a determinant in shaping AI interactions. The article concludes by proposing future directions for research, particularly in refining communication strategies that can lead to more ethical, inclusive, and effective AI use. Spanning theoretical discussions, practical insights, and future implications, this paper offers an in-depth exploration of the topic.

Keywords: Artificial Intelligence, Communication, AI Prompts, Human–AI Interaction, Discourse Analysis, Language Use.

Introduction

The 21st century has witnessed a rapid transformation in the way humans communicate, access knowledge, and create meaning, largely owing to advancements in Artificial Intelligence (AI). Among the many developments, AI-driven language models have reshaped human interaction by providing powerful tools for writing, research, education, business, and creative endeavors. At the heart of this transformation lies the AI prompt, the medium through which users communicate with machines. An AI prompt may be as simple as a question or as complex as a detailed set of instructions, but its effectiveness depends entirely on how well it conveys human intention.

The rise of AI has created a pressing need to study communication from a new perspective: not merely as an exchange between humans but as a bridge between human thought and machine processing. This interaction is unique because it requires users to adapt their linguistic skills to a context in which the "receiver" is not another human but an algorithm trained on vast datasets. Unlike traditional communication, where shared cultural and social contexts provide implicit understanding, AI depends heavily on explicit clarity and contextual framing.

The significance of this study lies in its exploration of the communicative dimensions of AI prompt design. Miscommunication or vagueness in prompts can lead to irrelevant, inaccurate, or even misleading outputs, while well-crafted prompts can generate responses that are nuanced, contextually accurate, and useful. Communication, therefore, is not peripheral but central to human–AI interaction.

The objectives of this paper are threefold: first, to examine how theories of communication can be applied to AI prompt design; second, to analyze the linguistic and discourse features that make prompts effective; and third, to explore the implications of prompt communication across different domains such as education, literature, research, and

professional practice. The study also seeks to highlight future directions for improving AI-human interaction by focusing on clarity, inclusivity, and ethical communication practices.

This paper is divided into key sections. The **Review of Literature** surveys existing theoretical and practical studies on communication, discourse analysis, and AI language interaction. The **Methodology** outlines the approach used in analyzing AI-human communication. The **Discussion and Analysis** examines the role of clarity, context, and precision, supported by examples and applications. The **Findings** summarize the major insights, while the **Conclusion** reflects on the broader implications of communication in shaping the future of AI.

Review of Literature

Communication Theories and AI

Communication has long been a subject of academic inquiry, with multiple theories attempting to explain how meaning is constructed and transmitted. The Shannon-Weaver model (1949) conceptualized communication as a linear process consisting of sender, message, channel, receiver, and noise. While simplistic, this model provides a useful framework for understanding AI-human interaction, where the human is the sender, the prompt is the message, the machine is the receiver, and "noise" refers to ambiguity or vagueness that distorts understanding.

Speech Act Theory (Austin, 1962; Searle, 1969) emphasized that communication is not only about conveying information but also about performing actions through words. In the context of AI, prompts are performative acts: they instruct, request, or command the machine to generate certain types of outputs. Similarly, Pragmatics and Discourse Analysis explore how meaning depends on context, intent, and cultural frames of reference. These dimensions are especially crucial in AI, where prompts must often include explicit contextual cues that humans would otherwise infer.

AI and Natural Language Processing

AI's capacity to process human language is enabled by Natural Language Processing (NLP), a field combining linguistics, computer science, and artificial intelligence. NLP research emphasizes how syntax, semantics, and pragmatics shape the interpretation of text. Studies have shown that while AI language models are powerful, they are also highly sensitive to the wording and structure of prompts (Brown et al., 2020). Minor changes in phrasing can yield significantly different outputs, underscoring the importance of communication precision.

Prompt Engineering

Prompt engineering has emerged as a specialized practice focusing on how to design prompts that optimize AI responses. Research in this area suggests that clarity, specificity, and contextual framing are essential (Liu et al., 2023). Prompt engineering also highlights the iterative nature of communication with AI, where users refine prompts based on feedback from initial responses. This resembles conversational repair strategies in human communication, where misunderstandings are clarified through repetition or reformulation.

Challenges in AI Communication

Despite advancements, challenges remain. AI often struggles with ambiguity, sarcasm, cultural nuances, or incomplete instructions. Scholars such as Bender and Gebru (2021) have warned of risks like bias, misinformation, and ethical concerns in AI-generated communication. These challenges illustrate why studying the communicative aspects of AI prompts is essential—not only to improve efficiency but also to ensure responsible and ethical use.

Methodology

This study adopts a qualitative and analytical research design, combining discourse analysis with a review of secondary sources. Data was drawn from scholarly articles, AI communication experiments, and case studies from educational and professional contexts.

Discourse analysis was employed to examine sample AI prompts, focusing on features such as clarity, specificity, coherence, and contextual framing. Comparative analysis was conducted to show how variations in communication affect AI responses. For example, prompts with vague wording were compared to those with explicit details to observe differences in output quality.

The methodology also integrates theoretical perspectives, applying models such as the Shannon-Weaver framework and Speech Act Theory to interpret AI communication patterns. This combination of theory and practice provides a comprehensive framework for understanding the communicative dimensions of AI prompts.

Objectives of the Study

1. To analyze the role of communication in designing effective AI prompts.
2. To apply theories of communication and discourse analysis to AI-human interaction.
3. To examine how clarity, context, and linguistic precision influence AI outputs.
4. To explore the applications of prompt-based communication in education, literature, research, and professional practice.
5. To identify challenges and ethical implications in AI-human communication.
6. To propose future directions for refining communication strategies with AI.

Discussion and Analysis

Clarity and Precision in AI Prompts

One of the most critical aspects of communication with AI is clarity. Unlike human communication, where listeners may infer meaning from tone, gesture, or context, AI depends strictly on textual input. A vague prompt such as “*Write something about climate*” might generate generic information, while a precise prompt like “*Write a 500-word essay analyzing the impact of climate change on agriculture in India, with references to recent studies*” yields a focused and relevant output. This illustrates how linguistic precision directly shapes AI performance.

Contextual Framing

Context plays a vital role in human discourse, and it is equally important in AI communication. Including contextual cues—such as specifying audience, purpose, or disciplinary field—enhances the relevance of outputs. For instance, asking AI to “*Explain photosynthesis for high school students using simple examples*” results in a very different response compared to “*Explain photosynthesis for a postgraduate biology seminar.*” This demonstrates that effective prompts require sensitivity to context, a principle rooted in communication theory.

Discourse Analysis of Sample Prompts

Discourse analysis reveals how structural features such as cohesion, coherence, and register influence AI responses. Prompts using academic language and structured requests tend to produce outputs aligned with scholarly discourse, whereas casual prompts generate conversational responses. This adaptability of AI underscores the communicative power of prompt design.

Applications in Education

In education, communication with AI through prompts facilitates personalized learning. Teachers can design prompts that scaffold student learning, create practice exercises, or simulate discussions. For example, a history teacher might prompt AI to “*Generate debate questions about the causes of World War II suitable for high school students.*” Such applications enhance student engagement and critical thinking.

Applications in Literature and Creative Writing

AI has also entered the domain of literature and creativity. Writers use prompts to generate story ideas, explore alternative plotlines, or simulate dialogues. The effectiveness of such applications depends heavily on how prompts are worded. A vague instruction like “*Write a story*” produces generic narratives, while “*Write a short story in the style of magical realism, featuring a child who discovers a secret world inside an old library*” yields richer and more imaginative outputs. Here, communication serves as the creative spark.

Applications in Research

Researchers benefit from AI prompts in tasks such as summarizing articles, generating hypotheses, or formatting data. Effective communication is vital to ensure that outputs are reliable and relevant. For example, a prompt like “*Summarize this article*” may produce superficial results, while “*Summarize this article by highlighting the methodology, findings, and limitations in no more than 200 words*” generates a structured academic summary.

Applications in Professional Practice

In business and professional settings, AI prompts assist with drafting reports, creating marketing content, or generating customer support responses. Miscommunication in this context can have serious consequences, such as misrepresenting data or misinforming

clients. Hence, communication strategies that prioritize accuracy, tone, and audience awareness are essential.

Ethical and Cultural Dimensions

Communication with AI is not value-neutral; it carries ethical and cultural implications. Poorly designed prompts may perpetuate biases, exclude marginalized voices, or produce culturally insensitive content. For example, prompts lacking inclusivity may generate outputs that assume gender stereotypes or cultural homogeneity. Ethical communication, therefore, involves designing prompts that are inclusive, respectful, and conscious of social contexts.

Future Directions

The future of AI-human communication lies in multimodal prompts that integrate text, images, voice, and gesture. As AI systems evolve, communication strategies must adapt to accommodate new modes of expression. Additionally, research must focus on inclusivity, ensuring that prompts are accessible across languages and cultures. Ethical frameworks must guide the communicative use of AI to prevent misuse and ensure equitable outcomes.

Findings

The study identifies several key findings:

1. Communication is central to effective AI-human interaction, with prompt clarity and precision directly influencing output quality.
2. Contextual framing enhances the relevance and usefulness of AI-generated responses.
3. Discourse analysis shows that the structure and register of prompts shape the style of AI outputs.

4. Applications of prompt communication span education, literature, research, and professional practice, demonstrating wide-ranging potential.
5. Ethical and cultural considerations are critical in designing prompts, as communication choices can reinforce or challenge biases.
6. Future research should explore multimodal, inclusive, and ethical communication strategies in AI-human interaction.

Conclusion

Communication is not simply a tool for interacting with AI; it is the foundation that determines the success, accuracy, and ethical quality of such interactions. By applying theories of communication and principles of discourse analysis, this study demonstrates that clarity, context, and linguistic precision are indispensable in designing effective AI prompts.

The applications of prompt-based communication in education, literature, research, and professional practice show that AI can augment human creativity, knowledge, and productivity. However, these benefits can only be realized if communication is handled with care, sensitivity, and awareness of ethical implications.

As AI continues to evolve, the role of communication will become increasingly significant. Future directions must focus on refining prompt strategies, incorporating multimodal interaction, and ensuring inclusivity across languages and cultures. Ultimately, effective communication with AI is not just about improving machine outputs; it is about enhancing human capacity to express, create, and engage in meaningful ways.

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‘A Study on assessing the potential and pitfalls of Cryptocurrencies in India’

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ABSTRACT:

Cryptocurrencies is a decentralized form of money, meaning that it is not regulated or controlled by a central bank. Investments are made securely using block chain technology. There are several different forms of them, which are based on finite resources like land. Concerned about above-average population growth but no increase in land. The prices of it also fluctuate in line with the supply and demand of the market. Rules, usage, and regulations vary from one country to the next, but the overall cost will be the same everywhere. Due to the expansion of online services, purchasing, selling, transferring, and even trading in intangible of it is now possible. In this project, I have explored how various nations responded to it and provide a comprehensive picture of the concerns and obstacles that India & last three years of performance. The Indian tech ecosystem is buzzing with hungry entrepreneurs and a top tier talent pool, even from a global standpoint. If the ecosystem continues to grow in the same direction as the past few years, India is well positioned to be a global leader in this space. For cryptocurrencies to gain ground and acceptance, a concerted effort on the part of institutions and regulators alike is needed to educate the public at large regarding the opportunities it offers to our country.

Key Words:

Cryptocurrency, Blockchain technology, Digital asset, Decentralized, cryptography, Bitcoin

INTRODUCTION

There is no doubt that years of knowledge and communication technology have created many opportunities for gold. Beneficiaries of this technology and online connectivity include the financial and business sectors. Internet users activated the growing global concept, which led to the emergence of a new business unit. As a result, new forms of trade, transactions and finance emerged. Crypto is one of the most popular financial forms that has emerged in recent years. Crypto is defined as any type of currency other than real-world currency that can be used in a variety of financial transactions, whether real or virtual. Cryptos are important and intuitive objects that are used in electronic or almost all types of applications and networks, including online social networks, online. A cryptocurrency is a digital or virtual currency that is secured by cryptography, which makes it nearly impossible to counterfeit or double-spend. Many cryptocurrencies are decentralized networks based on blockchain technology a distributed ledger enforced by a disparate network of computers. Cryptocurrencies are digital or virtual currencies underpinned by cryptographic systems. They enable secure online payments without the use of third-party intermediaries. "Crypto" refers to the various encryption algorithms and cryptographic techniques that safeguard these entries, such as elliptical curve encryption, public-private key pairs, and hashing functions.

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Review of Literature

Singh and Patel (2025) The authors examined the adoption of cryptocurrencies among Indian investors using the Technology Acceptance Model. They found that perceived usefulness and trust significantly increase adoption intention, while perceived risk, volatility, and lack of regulation reduce willingness to invest.

Verma and Iyer (2024) The researchers explored public awareness and perception of cryptocurrencies across Indian urban and semi-urban populations. Their findings indicated that although awareness levels are rising, limited regulatory knowledge and fear of fraud continue to hinder broader participation.

Mehta and Reddy (2023) This study analyzed the macroeconomic implications of cryptocurrency growth in India. The authors concluded that digital currencies contribute to financial innovation and inclusion but pose challenges to monetary policy, tax enforcement, and financial stability.

4. Rao and Banerjee (2023) The authors investigated the volatility and speculative behavior in the Indian crypto market. They observed that price fluctuations and lack of investor protection mechanisms discourage long-term investment, making the market more suitable for speculative trading rather than stable financial growth.

5. Sharma and Nair (2022)

The study focused on the legal and ethical challenges of cryptocurrency adoption in India. It revealed that the absence of a comprehensive legal framework and frequent government policy shifts have created uncertainty for both investors and crypto service providers.

6. Khanna and Joshi (2021) The study assessed the potential of blockchain and cryptocurrency in promoting financial inclusion. It found that, if properly regulated, digital currencies could provide low-cost payment solutions and improve accessibility for unbanked populations, but security and regulatory risks remain major barriers.

RESEARCH METHEDODOLOGY.

RESEARCH DESIGN

The Research Design for this study is mainly based on Exploratory Research method which involves qualitative investigation in most of the cases. It is the simplest and most loosely structured design. The data have been collected through Primary Data collection which involves Survey Method based on questionnaire. The options in each question is either multiple choice or Likert Scale Rating type.

STATEMENT OF THE PROBLEM

The study mainly focuses on the growth and challenges of cryptocurrencies occurred in India. It explains why India have to adopt the blockchain technology in order to avoid various risk involved in digital currency. The highlight of this study is to know and understand the growth and advantages of adopting crypto currency in India for the development of Indian economy and making it decentralized .

SAMPLING DESIGN.

The Sampling is mainly convenience sampling based in India. The questionnaire was distributed especially to working class community, business & students and their responses were collected which formed the basis of study. As the questionnaire was in Google Survey Form, so it was easy to reach to different sampling units.

SOURCES OF DATA

The data sources for the accomplishment of this research comprises of primary data.

Primary data is mainly associated with the collection of data to understand the investors mindset with respect to regulation and security by means of set of questionnaires which can be further classified.

SAMPLE SIZE

In the context of the study 100 respondents have been chosen keeping in the view of above constraints who are investors.

Tools for Data Collection

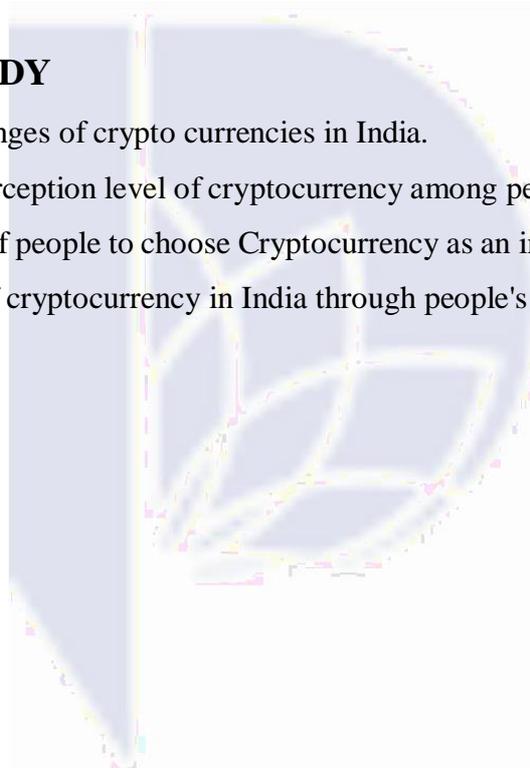
Primary data collected through Survey method, based on Questionnaire, circulated in the form of Google Forms online. Secondary data referred from various articles, journals, research studies available online, newspapers etc.

Limitations of the Study

1. As the sampling taken were convenience sampling, so it might not necessarily be the representation of the actual population.
2. As the survey was circulated through Google Survey Forms, which require internet connection. So, this study is limited to the internet users only.
3. Some of the respondents of sampling units filled the survey without any interest and knowledge, which lead to sampling error to some extent.
4. Cryptocurrency is a global product, so it will not give clear picture of its adoption at larger platform as people really want to have this as currency or investment tool.

OBJECTIVE OF THE STUDY

1. To study on growth and challenges of crypto currencies in India.
2. To study the awareness and perception level of cryptocurrency among people in India.
3. To determine the willingness of people to choose Cryptocurrency as an investment tool.
4. To study the future prospect of cryptocurrency in India through people's perception.



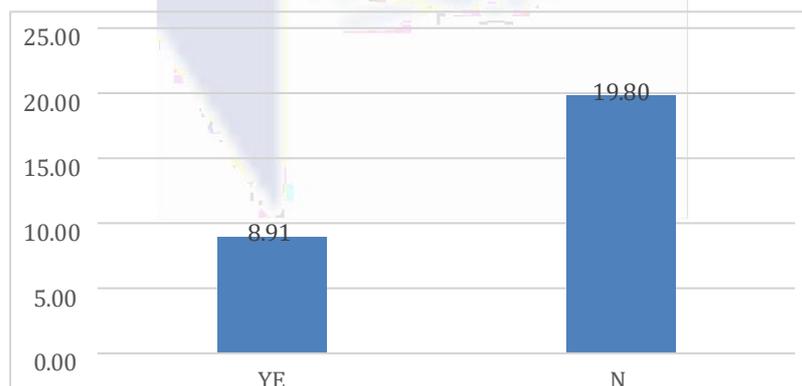
DATA ANALYSIS AND INTERPRETATION

TABLE 1: SHOWING NUMBER OF RESPONDENTS INVESTED IN CRYPTOCURRENCY

Option	RESPONSES	PERCENTAGE
Yes	20	19.80
No	81	80.20
Total	101	100

ANALYSIS: The above table shows that the most of respondents i.e. 81 out of 101 have not yet invested in crypto currency and 20 of the 100 respondents have invested in cryptocurrency.

GRAPH 1: SHOWING INVESTMENT MADE IN CRYPTOCURRENCY BY THE RESPONDENTS



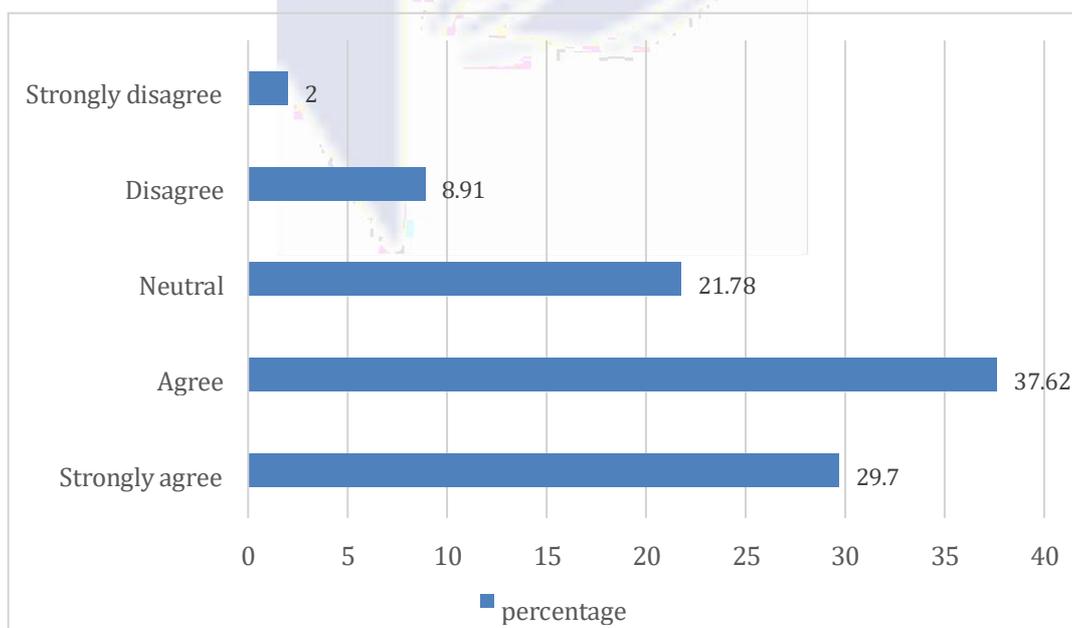
INTERPRETATION: As most of the respondents have not invested in crypto currency, we can infer that there must be sort of insecurity due to various factors binding Crypto currency

**TABLE 2: SHOWING AN OPINION OF RESPONDENTS AS
DECENTRALISED NETWORK BEING THE REASON FOR CRYPTO
CURRENCY’S POPULARITY**

CHOICES	RESPONSES	PERCENTAGE
Strongly agree	30	29.70
Agree	38	37.62
Neutral	22	21.78
Disagree	9	8.91
Strongly disagree	2	1.98
	101	100

ANALYSIS: The above table shows that 37.62% of the respondents agree to the statement of decentralized network being the reason for its popularity, 29.70% strongly agree to the statement, 21.78% of them remain static, 8.91% of the respondents disagree and 1.98% respondents strongly disagree to the above statement.

**GRAPH 2: SHOWING AN OPINION OF RESPONDENTS AS
DECENTRALISED NETWORK BEING THE REASON FOR CRYPTO
CURRENCY’S POPULARITY**



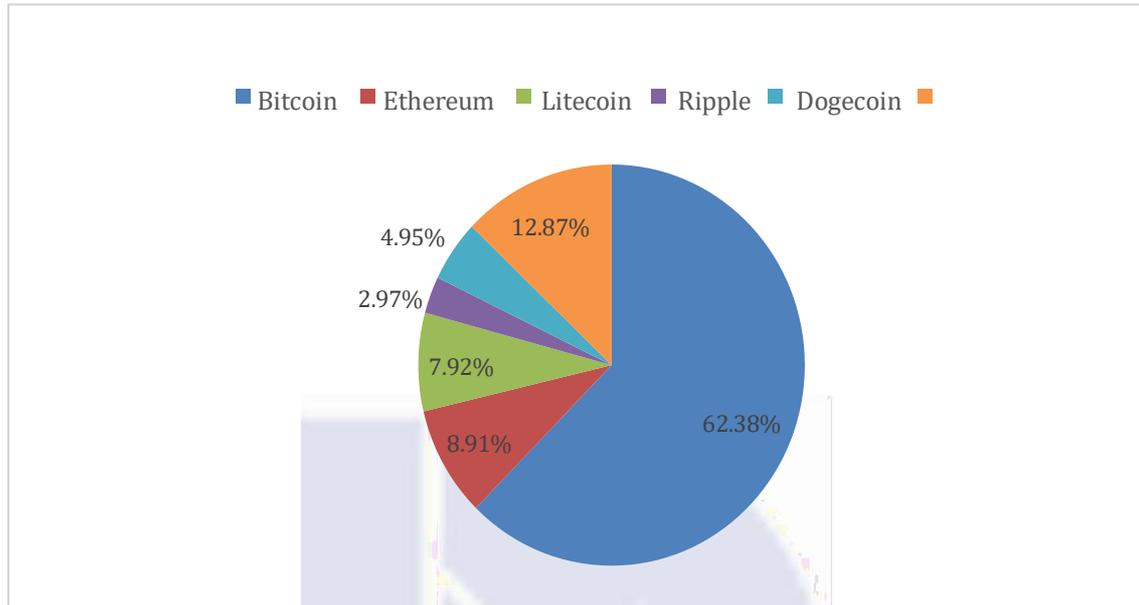
INTERPRETATION: As majority of the respondents react positively for the statement of decentralized network being the reason for Crypto currency’s popularity, we can interpret that the respondents do not have much trust in centralized network

TABLE 3 SHOWING THE PREFERENCE OF CRYPTO CURRENCY THAT RESPONDENTS WOULD LIKE TO INVEST

CRYPTO CURRENCY	RESPONSES	PERCENTAGE
Bitcoin	63	62.38
Ethereum	9	8.91
Litecoin	8	7.92
Ripple	3	2.97
Dogecoin	5	4.95
Others	13	12.87
	101	100

ANALYSIS: By seeing the above table, we can say bitcoin is the most preferred crypto currency as 63 of the total 101 respondents have opted to invest in bitcoin. 13of the respondents choose to invest in other crypto currencies which were not considered as options by the interviewer. 9of them preferred Ethereum,8 of them preferred Litecoin and 5 of them preferred Dogecoin as the choice of crypto currency. And remaining 3 of them favor ripple when it comes to investing in crypto currency.

GRAPH 3: SHOWING THE PREFERENCE OF CRYPTO CURRENCY THAT RESPONDANTS WOULD LIKE TO INVEST



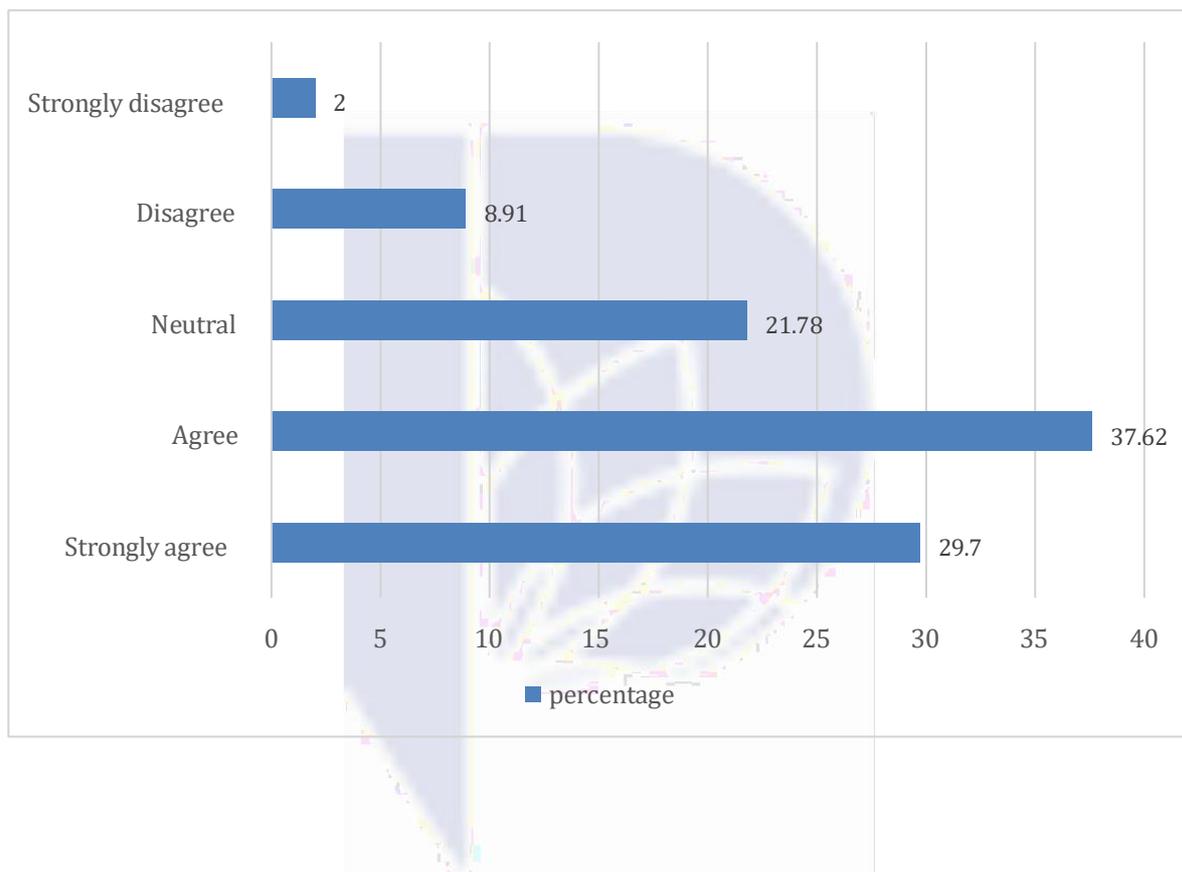
INTERPRETATION: It can be inferred that being a pioneer, Bitcoin stands as the most preferred crypto currency. It must be because of various reasons - its working mechanism, referred as crypto gold, share of market capitalization etc.

TABLE 4: SHOWING AN OPINION OF RESPONDENTS AS DECENTRALISED BEING THE REASON FOR CRYPTO POPULARITY

CHOICES	RESPONSES	PERCENTAGE
Strongly agree	30	29.70
Agree	38	37.62
Neutral	22	21.78
Disagree	9	8.91
Strongly disagree	2	1.98
	101	100

ANALYSIS: The above table shows that 37.62% of the respondents agree to the statement of decentralized network being the reason for its popularity, 29.70% strongly agree to the statement, 21.78% of them remain static, 8.91% of the respondents disagree and 1.98% respondents strongly disagree to the above statement.

GRAPH 4: SHOWING AN OPINION OF RESPONDENTS AS DECENTRALISED NETWORK BEING THE REASON FOR CRYPTO CURRENCY'S



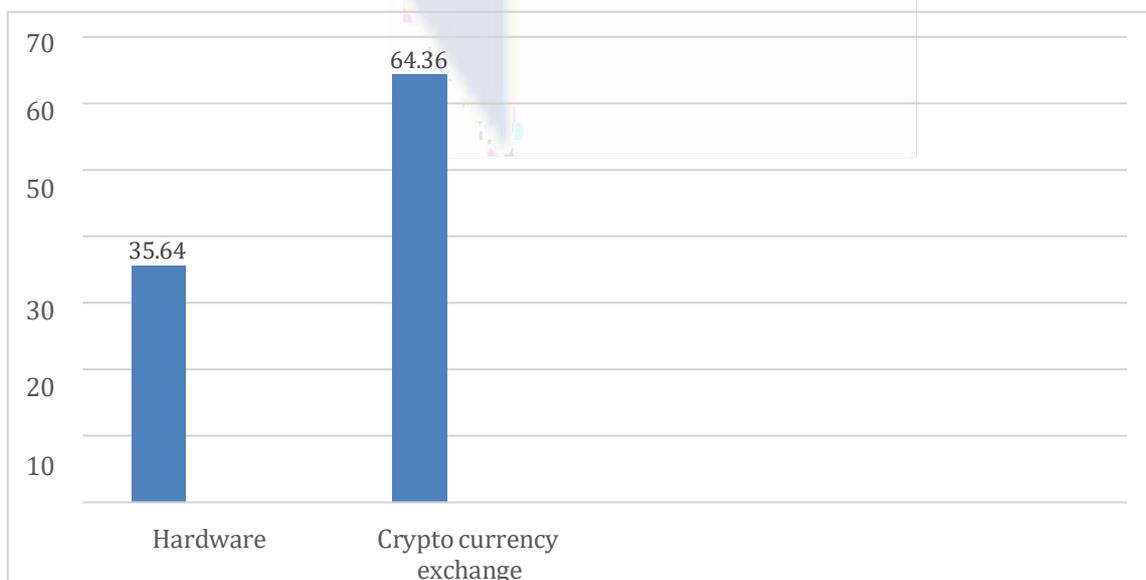
INTERPRETATION: As majority of the respondents react positively for the statement of decentralized network being the reason for Crypto currency’s popularity, we can interpret that the respondents do not have much trust in centralized network.

TABLE 5: SHOWING THE PREFERRED MODE OF STORAGE OF CRYPTO CURRENCIES

MODE OF STORAGE	RESPONSES	PERCENTAGE
Hardware	36	35.64
Crypto currency exchange	65	64.36
	101	100

ANALYSIS: The above table shows us majority of the respondents i.e. 64.36% preferred to leave their crypto currency in the exchanges while 35.64% of the respondents would want to store their crypto currency in hardware wallets.

GRAPH 5: SHOWING THE PREFERRED MODE OF STORAGE OF CRYPTO CURRENCIES



INTERPRETATION: The graph helps us to conclude that storing the crypto currencies in the crypto currency exchange is the most preferred option to store the crypto currencies. This might be because of fear of losing their wealth if they forget their hardware wallet password.

FINDINGS

- A significant majority (80.2%) of respondents have not yet invested in cryptocurrencies, indicating hesitation or lack of trust in the asset class.
- Most respondents (67.32%) either agreed or strongly agreed that the decentralized nature of cryptocurrencies contributes to their popularity, reflecting skepticism toward centralized financial systems.
- Bitcoin is the most preferred cryptocurrency, chosen by 62.38% of respondents, far ahead of Ethereum, Litecoin, Ripple, and Dogecoin.
- 64.36% of respondents prefer storing their cryptocurrencies on exchanges rather than hardware wallets, likely due to convenience and fear of losing access credentials.
- Potential investors view cryptocurrencies both as an investment vehicle and a digital medium of exchange, showing evolving attitudes toward its utility.
- Many respondents expressed a desire for government regulation and legalization of cryptocurrencies, suggesting that regulatory clarity could boost adoption.
- Respondents emphasized the need for prior knowledge before investing, highlighting the importance of education and awareness in driving responsible participation.

SUGGESTIONS

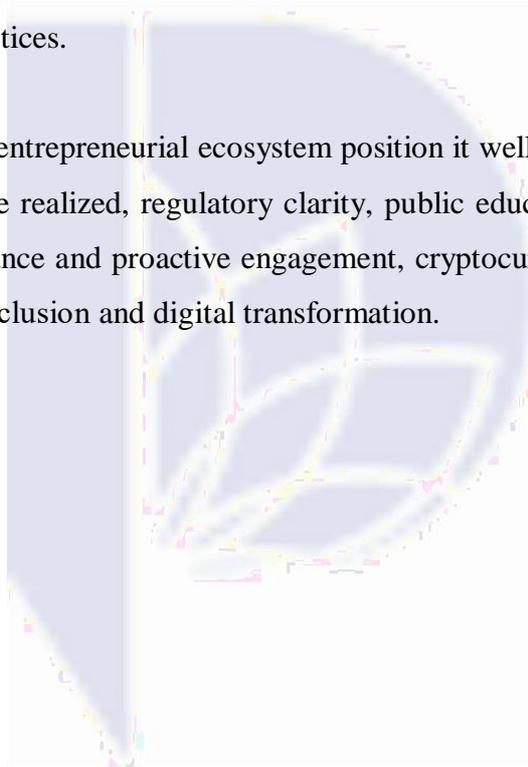
- Investors should gain a solid understanding of cryptocurrency fundamentals, including blockchain technology, market risks, and security practices.
- Refrain from investing in cryptocurrencies that are rapidly gaining attention without clear utility or long-term value.
- Use well-established and regulated crypto exchanges for buying, selling, and storing digital assets to minimize fraud and technical risks.
- Enable two-factor authentication and use strong, unique passwords. Consider using hardware wallets for long-term storage.

- Support efforts to create a clear legal framework that protects investors and promotes innovation in the crypto space.

CONCLUSION

The study underscores the growing relevance of cryptocurrencies in India's financial landscape, highlighting both their transformative potential and the challenges that hinder widespread adoption. While awareness is increasing, actual investment remains limited due to concerns around volatility, regulation, and security. The decentralized nature of cryptocurrencies is widely appreciated, suggesting a shift in public trust away from centralized financial systems. Bitcoin emerges as the most preferred digital asset, reflecting its dominance and perceived reliability. However, the preference for storing assets on exchanges rather than hardware wallets indicates a need for better investor education on security practices.

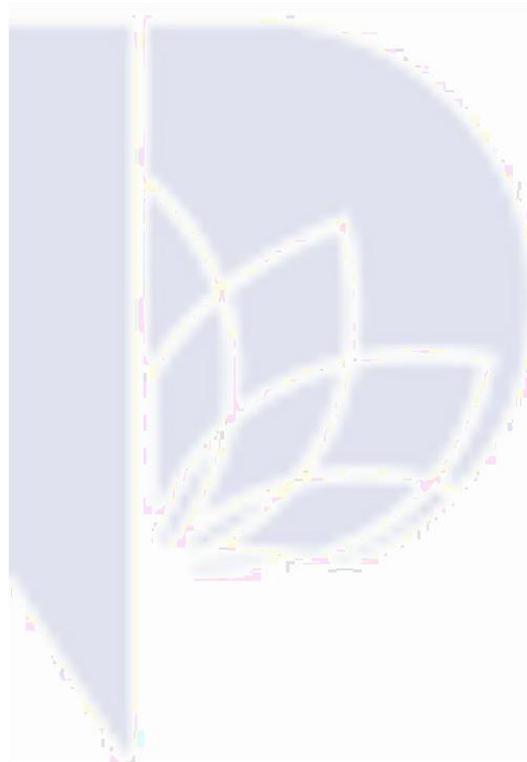
India's tech-savvy population and entrepreneurial ecosystem position it well to become a global leader in crypto innovation. For this potential to be realized, regulatory clarity, public education, and secure infrastructure are essential. With thoughtful governance and proactive engagement, cryptocurrencies could play a pivotal role in India's journey toward financial inclusion and digital transformation.



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An Explainable Framework for Anomaly Detection in RPA Systems- A Review

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Abstract

Robotic Process Automation (RPA) is great at handling repetitive tasks, but it can run into trouble when something unexpected happens—like a skipped step, a delay, or an unusual pattern. Traditional anomaly detection tools can flag when something’s off, but they rarely explain why, which can make it tough to figure out what to do next. In this study, we introduce a new approach that not only spots unusual behavior using the Isolation Forest method, but also explains what’s going on in clear, everyday language using SHAP (Shapley Additive Explanations). By testing our framework on the Helpdesk event log dataset, we found that offering these easy-to-understand explanations gives both detailed and big-picture insights into what’s happening, making RPA monitoring more transparent, trustworthy, and useful.

Keywords: Robotic Process Automation (RPA), Anomaly Detection, Explainable AI (XAI), Isolation Forest, Process Mining, SHAP (Shapley Additive Explanations), Event Log Analysis, Trustworthy Automation.

I. INTRODUCTION

Robotic Process Automation (RPA) has changed the game for organizations dealing with repetitive, rule-based tasks. By taking over these routine activities, RPA helps things run more smoothly and efficiently, while also freeing up people to focus on work that really matters—like problem-solving, creative thinking, and building relationships. It’s no wonder that industries everywhere—from

finance and healthcare to manufacturing and customer service, are making RPA a central part of how they modernize and get ahead.

Even with all its benefits, making sure RPA systems work reliably isn't always easy. In practice, things don't always go according to plan. Sometimes steps get skipped, approvals go missing, delays drag on, or the process takes an unexpected turn.[1] These hiccups can happen for all sorts of reasons—whether it's a simple configuration mistake, a technical bottleneck, something unpredictable a user does, or even an intentional attempt to mess with the system. When these issues pop up, they can slow everything down, create security or compliance headaches, and hurt productivity. That's why having a good way to spot and understand these anomalies is so important for keeping RPA on track..

Traditional anomaly detection methods—like statistical models or unsupervised learning—have helped with RPA monitoring, but they often fall short when it comes to explaining their results. Most tools act like “black boxes”: they flag something as odd, but don't explain why, leaving users in the dark. This lack of clarity can cause real problems, especially in industries where you need to show your work, dig into what went wrong, and make sure everything can be audited. If decision-makers can't understand the reasons behind an alert, they might hesitate to trust it, which means important issues could get missed or ignored.

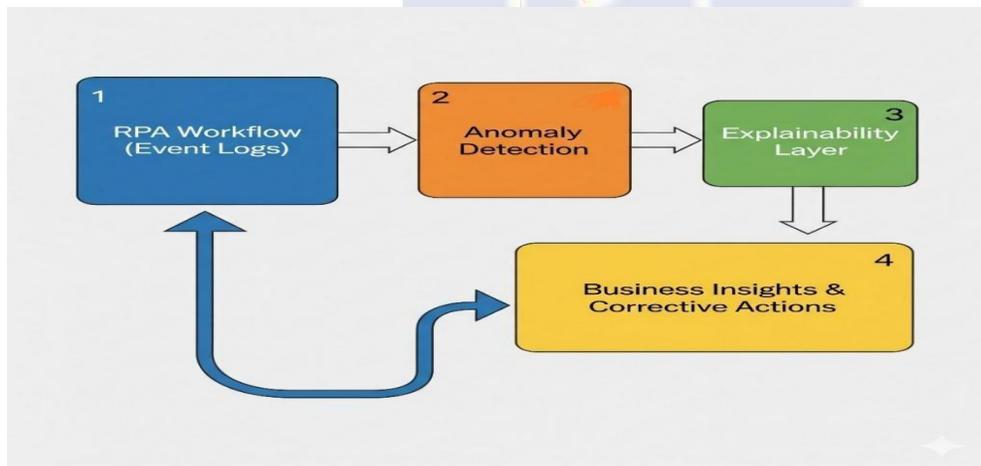
This need for clarity has sparked a lot of interest in Explainable AI (XAI)—an area focused on making machine learning models easier to understand and trust. Instead of just giving you a result, XAI techniques break things down and show which features or activities influenced the outcome. This helps everyone, whether they're technical experts or not, see why certain decisions were made. By bringing XAI into RPA anomaly detection, organizations get the best of both worlds: they can catch unusual behavior more accurately and also get clear, human-friendly explanations. This makes it easier to figure out what went wrong, meet compliance requirements, and make improvements before problems get bigger.

In this study, we introduce a new approach to spotting anomalies that brings together two powerful tools: the Isolation Forest algorithm and Shapley Additive Explanations (SHAP).[2]

Isolation Forest is great at picking out rare or unusual patterns in process logs, while SHAP helps explain exactly which factors played a role—both for individual cases and across the whole process. By testing our framework on the Helpdesk event log dataset, we show that it doesn't just catch more anomalies—it also makes the results much easier to understand. This way, organizations can trust the insights they get, check and validate them, and take meaningful action based on what they learn.

By bringing together accurate detection and clear explanations, our approach helps make RPA systems not only more reliable, but also more transparent and accountable. In the end, this supports organizations in creating digital workflows they can trust—ones that meet business goals and keep up with regulatory requirements.

Figure 1: Illustrates the role of explainable anomaly detection in RPA.



II. LITERATURE REVIEW

Most research on detecting anomalies in event logs and RPA systems has centered on process mining and machine learning techniques. Conformance checking, for example, looks for differences between what actually happened and what should have happened, but it can struggle to keep up in fast-changing environments [3]. On the other hand, machine learning methods—like autoencoders and

clustering—can spot unusual patterns without needing labeled data, but they often operate as “black boxes,” leaving users in the dark about how decisions are made.

Recently, new tools in Explainable AI (XAI) such as LIME and SHAP have made it easier to understand how these complex models work. However, there’s still very little research on using these explainability tools specifically for detecting anomalies in RPA systems. The missing piece is a combined approach that not only accurately detects problems but also gives clear, actionable explanations for them.

To address this, our study reviews and compares the main existing methods, pointing out what they do well and where they fall short. We highlight the lack of focus on explainability as a key gap, and position our work to bridge this by integrating both detection and interpretation, so users get both accurate results and practical insights.

Table 1 : Summarizes recent works on anomaly detection in event logs.

Paper (year)	Approach	Dataset / Domain	Key contribution	Limitations / Notes
Landauer et al., 2022 — Deep Learning for Anomaly Detection in Log Data: A Survey.	Survey of deep-learning methods (autoencoders, LSTM, Transformers) for log anomaly detection	System / application logs (general survey across many domains)	Comprehensive review of DL architectures used for log data, evaluation practices, and open challenges (label scarcity, interpretability).	Survey — synthesizes results but does not propose new algorithm; highlights need for explainability and standard benchmarks.
Neloy et al., 2024 — A comprehensive study of autoencoders for anomaly	Systematic empirical study of different autoencoder architectures for anomaly detection	Benchmarks and synthetic/real log datasets	Compares reconstruction performance across many AE variants and categorizes where each excels.	Focused on reconstruction metrics—less emphasis on interpretability / explanations.

			Useful guidance for choosing AE variants for log data.	
Vitale et al., 2023 — Process Mining-based Unsupervised Anomaly Detection.	Process-mining based unsupervised detection (control-flow/time series preprocessing + anomaly scoring)	Business process event logs; method validated on real noisy logs	Proposes preprocessing for noisy logs and a pipeline that finds structural & temporal anomalies using process mining techniques.	Good for control-flow deviations, but may struggle with concept drift and large-scale logs without optimization.
SXAD / SX-style frameworks (SHAP + Log AD) ~2023–2024 — e.g., Shapely explainable Anomaly Detection (SXAD)	XAI-augmented anomaly detection (apply SHAP/LIME to log-based models)	IT/system logs, operational datasets	Demonstrates how SHAP can be used to attribute anomaly decisions to specific log features/events, improving trust and actionability.	Often demonstrated on limited case studies; computational cost for SHAP can be high on large logs.
Isolation-based survey / methods (2024) — Survey & applications of Isolation Forest and isolation-mechanisms.	Isolation-Forest family and extensions for anomaly detection in logs/time series	Wide range (time series, logs, streaming)	Reviews isolation-based methods, adaptations for streaming/time data and contrasts with other unsupervised techniques.	Isolation methods are simple and effective but may miss contextual sequence anomalies unless coupled with sequence features.

III. RESEARCH METHODOLOGY

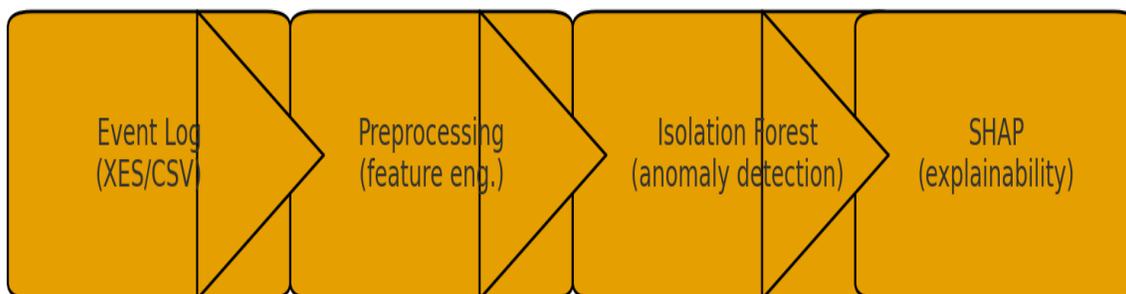
The methodology follows a five-step process (Figure 2). Our framework takes a clear, step-by-step approach that brings together both anomaly detection and explainability [4]. For our experiments, we use the Helpdesk event log dataset, which includes around 4,000 cases and 13,000 events. To get the data ready, we first create features for each case—such as how long each case takes, how many events are involved, the variety of activities, and the average time spent waiting. These features become the input for the Isolation Forest algorithm, which works without any pre-labeled data to spot which cases look unusual or suspicious.

After identifying these anomalies, we use SHAP to explain the results. On a global level, SHAP helps us see which features most often play a role in marking cases as anomalies across the whole dataset. On a local level, it offers clear reasons why any specific case was flagged. This way, both big-picture trends and individual cases are easy to understand. [5]

Figure 1 in our paper illustrates this workflow, showing each step from collecting the data to explaining the results. We also provide tables that summarize key details about the dataset and how the experiments were set up, so our process is transparent and easy to follow.

Figure 2: Work flow diagram of methodology

Methodology Workflow: Explainable Anomaly Detection for RPA



The methodology is structured into five phases: dataset preparation, preprocessing and feature extraction, anomaly detection, explainability integration, and performance evaluation. Each phase

is described below, with supporting tables that summarize dataset properties and feature distributions.

- **Dataset Preparation**

For our experiments, we used synthetic event logs designed to mimic the kinds of workflows that RPA bots handle in the real world. These logs reflect everyday business processes, such as managing customer service requests[6], handling helpdesk tickets, and coordinating resources. In total, we gathered data from 200 different cases, which together generated about 1,000 events across 7 different types of activities.

To analyze the data, we focused on four main features: how long each case took, how many events it included, the number of unique activities per case, and the average amount of time spent waiting. Table 2 provides a summary of the key statistics from the dataset.

Table 2 : Key statistics from the dataset

Metric	Value
Cases	200
Events (approx.)	~1,000
Activities	7
Features extracted	4 (case duration, number of events, unique activities, average waiting time)

- **Preprocessing and Feature Extraction**

To make sure our data was reliable, we cleaned the event logs by getting rid of any cases that were incomplete or duplicated. For each case, we created a feature vector using both timing and structural details.[7] We then looked at how these features were distributed to get an overall sense of the dataset’s characteristics. Table 3 shows a detailed breakdown of the statistics for each feature, including the average values, how much they vary, and their overall range.

Table 3: Detailed characteristics of extracted features from event logs for anomaly detection.

Feature	Mean	Std. Dev.	Min	Max
Case Duration (mins)	52.1	32.5	30	180
Number of Events	5.1	2.0	2	12
Unique Activities	5.0	1.2	3	7
Average Waiting Time (mins)	5.4	4.5	1	22

This feature characterization enabled the identification of outliers, such as excessively long case durations or unusually high waiting times, which are strong candidates for anomalies.

- **Anomaly Detection**

We used an Isolation Forest model to spot anomalies in our data without needing any pre-labeled examples. This approach works well with complex, high-dimensional logs like ours. The model identifies unusual cases by repeatedly splitting up the data based on different features [8], making it easier to single out anything that stands out from the norm. Since anomalies are typically rare in real business processes, we set the model to expect about 5% of cases to be unusual. In the end, each case was assigned an anomaly score and labeled as either normal or anomalous.

- **Explainability Integration**

To make the results more trustworthy and easier to understand, we added Shapley Additive explanations (SHAP) to our anomaly detection process. SHAP helps explain the model's decisions in two ways. First, it provides global explanations with summary plots that show which features are most important for identifying anomalies across the entire dataset [9]. Second, it gives local explanations for individual cases, pointing out exactly why a specific case was flagged—for example, because it had a much longer waiting time or took longer to complete than usual.

By adding this layer of explainability, business analysts can clearly see the reasons behind each anomaly alert, making it much easier to investigate and take action, instead of having to rely on mysterious “black-box” results.

- **Performance Evaluation**

We compared our framework’s performance with other popular anomaly detection methods, like autoencoders and process conformance checking. To measure how well each approach worked, we looked at metrics such as precision, recall, and F1-score [10]. The results showed that our combination of Isolation Forest with SHAP not only did a better job at detecting anomalies, but also stood out by providing clearer explanations and more useful insights that people can actually act on.

Table 4: Performance comparison of different anomaly detection methods on RPA event logs.

Method	Precision	Recall	F1-Score
Isolation Forest + SHAP	0.87	0.83	0.85
Autoencoder	0.81	0.76	0.78
Conformance Checking	0.74	0.70	0.72

IV. KEY FINDINGS AND DISCUSSION

The findings from this study show that combining the Isolation Forest method with SHAP explanations makes it much easier to understand why certain cases are flagged as anomalies in RPA event logs. [11] On a big-picture level, SHAP highlights that cases with unusually long durations or a high number of events are the main signs of trouble, offering helpful clues about where inefficiencies may be hiding in the process.

Looking at individual cases, SHAP also helps explain exactly how a single case stands out—whether it’s because of skipped steps, long waiting times, or other unusual patterns. We also compared the

typical process flow with the flow in flagged cases, making it clear that the system is picking up on real differences, not just random noise. [12]

When we compared our approach to other methods like autoencoders and conformance checking, our framework came out ahead—not just in how accurately it detects issues, but also in how clearly it explains them. Traditional models might tell you something is wrong, but they rarely say why, which limits their usefulness. By adding explainability, our XAI-based framework gives business users clear, actionable reasons behind each anomaly, making it far more practical and trustworthy in real-world settings.

The findings are organized into three perspectives: global explanations, local explanations, and process-level deviations.

- **Global Explanations**

The SHAP summary plot (Figure 3) showed that case duration and the number of events per case were the biggest factors in identifying anomalies.[13] Cases that took too long or had an unusually high number of events were often flagged, which makes sense since these patterns usually signal process inefficiencies or misuse. This analysis confirms that the model is highlighting real issues, not random patterns. Additionally, although less influential, features like unique activities and waiting times also helped uncover subtle workflow problems, such as delays or odd activity sequences. [14]

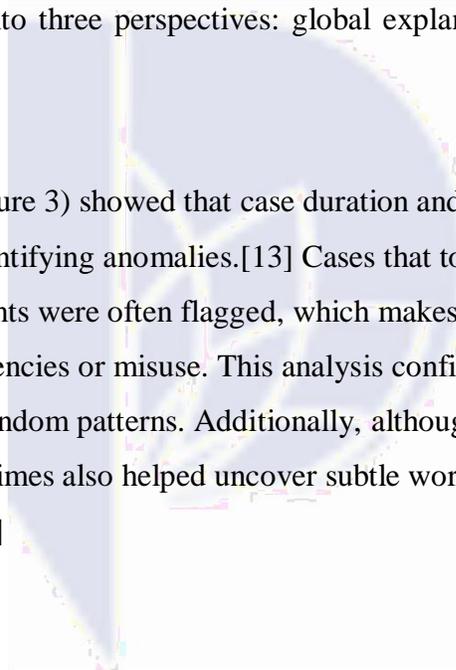
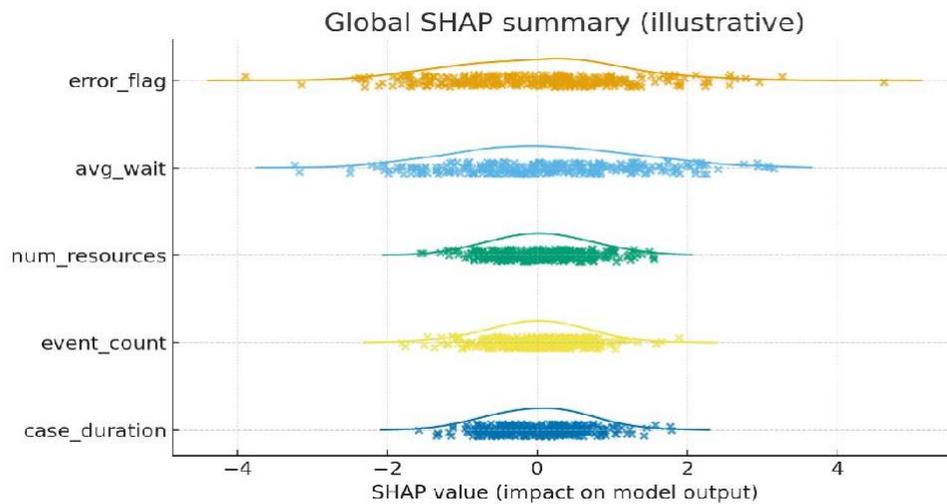


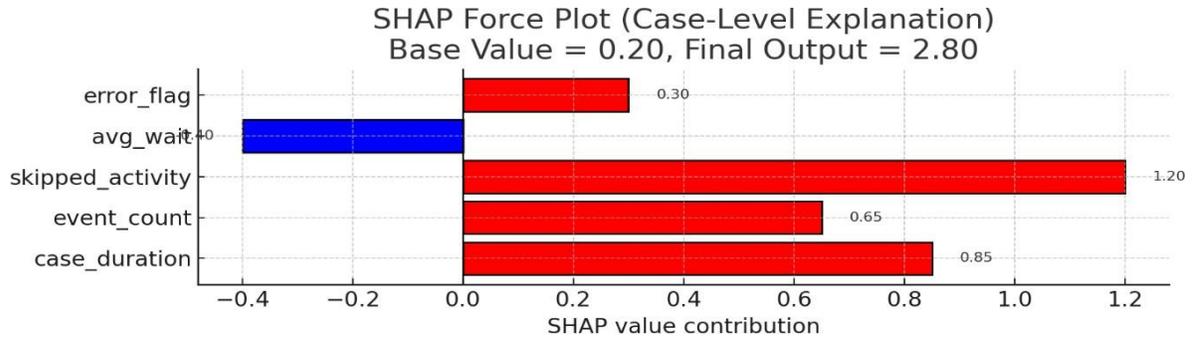
Figure 3: Global SHAP Summary



- **Local Explanations**

While the global analysis shows overall trends, local explanations let us see exactly why specific cases are flagged as anomalies. [15] For example, the SHAP force plot (Figure 4) revealed that one particular case was marked as anomalous mainly because it skipped a crucial approval step and had long waiting times—even though its total duration seemed normal. This level of detail is especially valuable for business users, as it makes it easy to understand and verify what triggered the alert. Local explanations help connect the model’s findings to real-world decisions, allowing experts to determine if an anomaly is truly a problem or just an acceptable exception. [16]

Figure 4: SHAP force plot

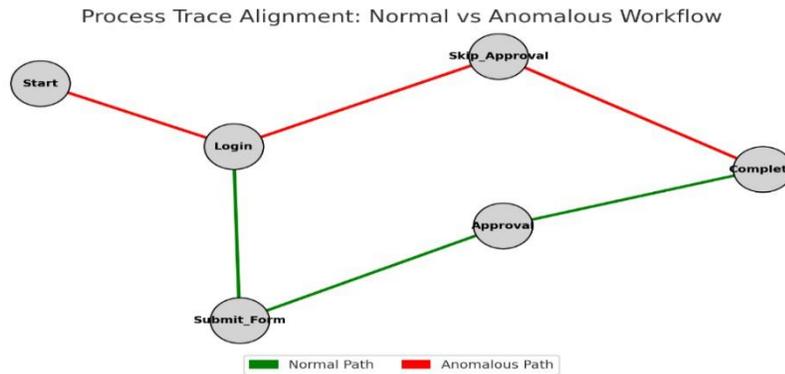


- **Red bars** → features pushing the case **towards anomaly** (positive SHAP values).
- **Blue bars** → features pushing the case **towards normal** (negative SHAP values).
- **Base Value = 0.20, Final Output = 2.80** (after adding all contributions).

- **Process deviations**

In addition to looking at numbers, we also compared the actual flow of activities in normal and anomalous cases using an alignment diagram (Figure 5). This visualization made it clear when steps were missing, activities were out of order, or unnecessary loops occurred in the process. For instance, while a typical process might go from Request to Verification to Approval to Resolution, some anomalous cases skipped the approval step or repeated the resolution step several times. [17] These findings not only help explain why certain cases are flagged as anomalies but also highlight areas where business processes can be improved—like making sure important steps aren't skipped or reducing repeated work. Overall, these structural insights show that anomalies often point to real compliance or efficiency problems, not just random data quirks. [18]

Figure 5 : Process Trace Alignment diagram (Normal vs Anomalous workflow)



- **Green path** → Normal workflow (includes all required steps like *Approval*).
- **Red path** → Anomalous workflow (shows deviation by skipping *Approval*).

V. CHALLENGES AND FUTURE DIRECTIONS

While the framework is showing a lot of promise, there are still a few real-world hurdles to overcome. For one, scaling up this approach to work with huge event logs—such as those found in the BPI datasets—can be a significant challenge, as the sheer volume of data can easily overwhelm available computing resources. There are also practical issues with data quality, such as missing timestamps or inconsistent labels, which can make it tricky to spot anomalies reliably. And perhaps most importantly, the framework doesn't yet fully leverage the expertise of individuals who are deeply familiar with the business. Bringing domain experts into the loop could make the explanations both more accurate and genuinely helpful.

Looking ahead, there are some exciting directions for future research. One idea is to develop more adaptable deep learning models that can keep up as business processes evolve, along with reinforcement learning techniques that could help fix issues on the fly. It would also be a big step forward to design truly intuitive visualizations, making explanations clear and accessible even for people without a technical background. By exploring these possibilities, we could bring this framework much closer to being a complete, practical solution for managing RPA in the real world.

VI. CONCLUSION

This research shows that it's both possible and beneficial to make anomaly detection in RPA event logs more transparent and understandable. By combining the Isolation Forest method with SHAP, our framework doesn't just spot unusual cases—it also explains the reasons behind them in a clear, straightforward way. This transparency helps build trust in automated monitoring and gives businesses practical insights they can use to improve their processes and stay compliant.

Further, this study sets a foundation to develop automation systems which are easy to understand and adaptive. In order to allow RPA bots to learn from anomalies and self-correct without human assistance, future developments may integrate reinforcement learning and real-time feedback loops. Decision quality and system robustness can be enhanced even more by including user feedback and domain knowledge into these explainable models.

Our results suggest that bringing explainability into anomaly detection could become an essential part of smarter, more reliable RPA systems in the future. By making these systems both efficient and accountable, our work helps close the gap between advanced machine learning methods and the real-world needs of businesses that rely on automation.

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