

Original Article

The Convergence of Artificial Intelligence and Product Strategy: A Data Science Perspective on Market Disruption

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Abstract: Across many industries, AI is changing how businesses operate, develop their products, examine the market, and reach consumers. Using both AI and product strategy makes it possible for companies to gain useful knowledge from huge datasets and use that to bring new ideas to the market. The paper investigates how AI and strategic product management interact when viewed through the lens of data science. We examine how integrating machine learning, natural language processing, and predictive analytics into product lifecycle management can increase innovation and improve customer experience. The paper also examines how AI helps predict upcoming market movements, adjust experiences for each person, and check product performance as it occurs. We use cases, illustrate data, and look at how AI is used in today's industry to discuss the effects of this convergence and recommend strategies for using AI. The study discusses ethical matters, rules for handling data, and the lack of skills companies encounter when using AI for products. In general, the research explains that AI could strengthen both product strategy and leadership by driving the direction of products.

Keywords: Artificial Intelligence, Product Strategy, Data Science, Machine Learning, Predictive Analytics, Strategic Management, Innovation.

I. INTRODUCTION

A) Importance of Product Strategy in Business Success

Thanks to digital transformation, AI has become a key source of new ideas and competitiveness, moving from the background of future ideas into center stage for today's businesses. [1-4] AI-based tools support many top organizations in making better decisions, predicting what will happen, distributing resources more efficiently, and adapting to market upsets. As a result, companies can see in detail how consumers act and predict what is coming up, guiding smarter and more effective decisions. Tesla, Amazon, and Netflix are good examples of how combining AI with products can disturb established industries. Amazon, Netflix, and Tesla all rely on AI: Amazon recommends products and streamlines supplies, Netflix guides its streaming offerings, and Tesla makes their vehicles smart and enhances its products. They demonstrate how AI increases customer satisfaction, improves operations, and supports the company's market position. AI has become essential to business strategy today, transforming many industries by defining new paths for achievement.



Fig. 1 Importance of Product Strategy in Business Success



- **Defining Product Strategy:** A product strategy directs the making, placing in the market, and management of a product's entire life. Marketing brings the vision for the product in line with the company's goals, the market's needs, and what customers expect. Ensuring your product strategy is clear allows for unity in decisions so all actions help the business and bring value to buyers.
- **Driving Competitive Advantage:** A good product strategy helps businesses stand out. Finding what makes a company different, focusing on the right parts of the market, and looking ahead at market trends help create unique products. With separate methods, brands ensure customer loyalty, draw new clients, and retain profits for the future.
- **Optimizing Resource Allocation:** Using a product strategy shows managers which investments are most important and how to use their resources wisely. It leads organizations to concentrate on ideas that matter most and saves time and money from investing in ideas with little or no potential. This strategic approach shortens product development and ensures every dollar spent is well used.
- **Enhancing Customer Satisfaction:** Effective solutions for their issues are possible when centering product planning on customer demands and what they tell you. When customers are pleased with what they buy, they tend to keep coming back, suggesting the product to others, and sharing useful information to make it better.
- **Enabling Adaptability and Growth:** To succeed, a product strategy needs ways to track shifts in the market and new technologies. Being agile means businesses can respond to changes, improve frequently, and take advantage of new projects. So, organizations are more capable of growing and adapting to changes in unpredictable surroundings.
- **Aligning Cross-Functional Teams:** Product strategy enables teams from marketing, sales, development, and executive levels to work toward the same objectives and important things. This way of working helps different teams coordinate, reduce conflicts, and speed up the delivery of what customers and businesses want. In brief, having a solid product strategy is key to making a business successful. It influences the design of products, ensures the company's market importance, helps it become more competitive, efficiently uses resources, makes customers happy, and builds a healthy workplace. Attempting to grow without a clear product strategy increases the risk of departments working against each other and important chances being missed in the marketplace.

B) Role of Data Science in AI Integration

Bringing AI into the product strategy relies on data science, which connects raw data and insights businesses can use. Data science is gathering, cleansing, working with, and studying large data sets to reveal patterns, trends, and relationships influencing decisions. AI models need accurate and relevant data to work properly, so data science checks that the data is suitable for use. Regarding AI in product strategy, data science supports the strong development of machine learning and prediction models. It converts customer actions, market information, and business figures that are difficult to use into a format that machines can process. Feature engineering is part of the process where important factors affecting outcomes are found, improving the model's reliability and performance. Also, data science helps machines improve by providing ongoing learning. Statistical analysis and visualization of data help data scientists watch the results produced by AI models, spot issues and confirm that they line up with company objectives. Continuous evaluation of the algorithms allows them to meet new challenges and customer preferences. In addition, data science promotes understanding and openness of AI models, meaning stakeholders can recognize the basis for AI-based solutions and conclusions. Clarity in recommendations is important for leading trusting relationships between departments and managing ethical topics such as bias and fairness. Data science is the main structure for bringing AI into product strategy. It changes raw data into effective and clear inputs that aid AI in creating useful and smart product choices. If data science processes are not conducted correctly, AI tools could be inadequate and provide wrong information. As a result, the importance of data science for AI is obvious.

II. LITERATURE SURVEY

A) Historical Overview of AI in Business

Since the 1980s, when expert systems first appeared, AI has played a role in business. These systems, e.g., XCON developed by Digital Equipment Corporation, mimicked the decisions that experienced people could make. While their initial use was limited, they quickly revealed that business automation was possible. [5-8] As advanced technology became available in the 1990s and 2000s, it became clear that simple rule-based systems were limited, leading to the move toward more flexible approaches. The recent growth in big data and access to cloud computing has made AI much more powerful and faster. AI researchers use different tools, including deep learning, reinforcement learning, and unsupervised learning. These technologies can go through large collections of unstructured data and point out patterns and trends that people could not possibly find on such a scale. Now, AI allows businesses to do more than automate. They also make decisions automatically, offer personal experiences, and forecast upcoming trends, which changes how they handle processes.

B) Review of Product Strategy Frameworks

The usual frameworks for product strategy have led to well-made decisions in businesses for a long time. For example, Ansoff's Matrix guides companies in judging new growth opportunities using market penetration, product innovation, market

development, and diversification. Likewise, using Porter's Five Forces helps judge how competitive an industry is and influences future strategies. Even though these frameworks give detailed strategic planning guidance, they are considered too rigid and fit today's fast business environment. Adding AI to strategic planning has moved these old tools into modern systems that continually process real-time data, model several market situations, and forecast customer actions accurately. Machine learning helps restaurants more accurately than traditional ways, so products and marketing can be tailored to each segment. For example, AI prototypes help with sentiment analysis and forecasting, which conventional models cannot handle well. Because of this, companies are beginning to use the practical lessons learned from standard frameworks and the advanced tools of AI to shape strategies that react more quickly.

C) AI Applications in Strategic Management

AI is applied to many areas in strategic management, such as anticipating future trends, supporting decisions, and streamlining operations. More and more, AI systems are helping to predict changes in marketing trends, problems in the supply chain, and expected earnings with greater certainty. Today, using NLP and machine learning, decision support systems can summarize and present useful insights from a large amount of data to executives. IBM Watson is one example frequently mentioned since it has performed well in healthcare, finance, and legal services. Watson supports doctors by evaluating patient data against many medical records and suggests evidence-backed treatment options. In investment decisions, AI monitors financial markets and news worldwide. AI is used in strategic operations to improve how much inventory businesses need, as well as the work of employees with the help of predictive maintenance and automation. They show that AI is valuable for supporting and steering important business actions and offering an edge to companies when markets change.

D) Gaps in Existing Research

Despite extensive research on using AI in business, a real need exists for full integration of AI, data science, and product strategy. A lot of what is written about these topics emphasizes only a few specific fields, like marketing automation or improvements to daily operations. Not many studies connect AI-based insights to product development and the company's position in the market. Because AI is scattered, organizations cannot use it best for all their strategy work. Not enough research reports demonstrate how AI efforts continue to influence business strategies in numerous industries and organizations. Strategic management research hasn't fully explored the topic of ethical and governance problems in AI related to algorithmic bias and transparency. As a way to close these gaps, this study presents an integrated framework for bringing AI, goals, resources, and market conditions into harmony. The idea is to guide organizations in adopting AI by moving from theory to actionable ways of using AI in their strategy-making.

III. METHODOLOGY

A) Research Design

This project uses mixed-method research to analyze how Artificial Intelligence (AI), data science, and product strategy come together. The combination of qualitative and quantitative approaches enables the researchers to see important details and broad trends. This section looks closely at several firms in the tech, finance, and healthcare fields that have included AI in their planning. We create these case studies by conducting semi-structured interviews with product managers, data scientists, and strategy executives. [9-12] This process enables a detailed study of an organization's routines, how it makes decisions, and how ready its workers are for AI. The emergence of success strategies and the discovery missing points are enabled by including qualitative data, usually missed by quantitative studies alone. At the same time, we examine numerical information from industry research, questionnaires on AI usage, financial success measures, and publicly accessible records about AI use. The aim is to uncover whether AI is related to success in the market, customer loyalty, and making innovations. Various statistical instruments, such as regression analysis and tools for visualizing data, are applied to analyze and confirm what has been spotted in qualitative research. Combining these two methods increases the honesty and dependability of the results. Interviewing healthcare experts provides real insights at the time, and secondary reading from journals, white papers, and benchmarking reports supplies information about the whole field. Thanks to this model, the research supplies useful recommendations to businesses looking to implement AI effectively. A mixed-method approach creates a strong framework for running an AI product strategy.

B) AI Techniques Used

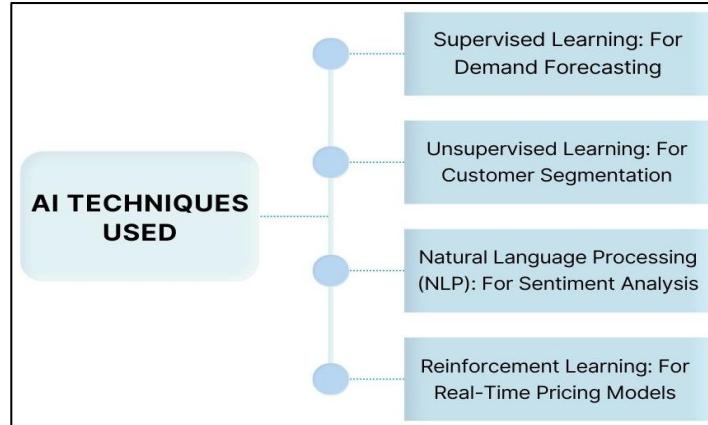


Fig. 2 AI Techniques Used

- **Supervised Learning: For Demand Forecasting:** Training predictive models in supervised learning depends on data that has already been labeled. This study uses linear regression, decision trees, and neural networks to forecast product demand in different market segments. They base their decisions on historical sales results, seasonal trends, what marketing channels do, and economic changes. Thanks to spotting trends and links in this data, supervised learning allows companies to improve how they handle inventory, arrange their supply chain, and set up promotions.
- **Unsupervised Learning: For Customer Segmentation,** unsupervised techniques, such as K-means and hierarchical clustering, are applied to separate customers for marketing segmentation. They examine customer data that lack labels, focusing on items like purchase choices, browsing, and how involved a customer is in revealing hidden commonalities. Traditional demographic segmentation methods differ from unsupervised learning, which can discover behavior-based groups. Based on these insights, companies can make their products, price their services, and run advertisements that fit certain customer groups, making customers happier.
- **Natural Language Processing (NLP): For Sentiment Analysis:** For this study, Natural Language Processing (NLP) is applied to look at customer feedback, reviews, and social media mentions. This technology, sentiment analysis, is important in NLP and lets organizations understand how people feel about their products and experiences. Using a large amount of text, NLP tools can detect if a text is positive, negative, or neutral and pinpoint important themes discussed within the text. Because of these insights, product managers can improve what they sell, deal with customer issues, and successfully manage the brand's reputation in a competitive setting.
- **Reinforcement Learning: For Real-Time Pricing Models:** Real-time dynamic pricing models are built using reinforcement learning. An AI agent develops the best pricing strategies by engaging with its surroundings, checking demand, seeing other competitors, and analyzing what customers do. Eventually, the model gains the most returns, such as extra revenue or market share, by responding to shifting trends in the market. The high flexibility in setting prices in e-commerce and online businesses helps to control customer purchases and helps e-commerce businesses make more profits.

C) Data Collection Tools



Fig. 3 Data Collection Tools

- **Web Scraping Tools for Market Trend Analysis:** This data is collected through web scraping tools on websites, news websites, internet shops, and competitor pages. [13-17] Through these tools, companies can automatically get useful information about pricing, new products, customers' thoughts, and rising trends in different countries. What happens is that aggregating web data makes it possible to uncover general trends in industries and what interests consumers. With

this data, businesses can plan their future steps in response to what happens in the market and with competing companies.

- **APIs for Social Media Sentiment:** User-generated sentiment analysis content is obtained using Twitter, Facebook, and Reddi APIst. Using these APIs makes it possible to automatically retrieve posts, comments, and hashtags about specific products, brands, or topics on the market. Taking advantage of mass unstructured data, sentiment analysis systems assist in gauging views from the public, spotting changes in consumer attitudes, and detecting what's trending online. Using this insight helps form how a brand talks to customers, answers customer concerns and makes product modifications based on real-time social discussions.
- **SQL Databases for Internal Product Performance Metrics:** Businesses use Structured Query Language (SQL) databases to store and control their data on products, covering sales results, returns, customer loyalty, and stock amounts. Thanks to these databases, historical and current data can be processed and examined easily, giving a good base to support supervised learning techniques and comparison of system performances. With SQL-based data systems, everything businesses need is accessible, scalable, and accurate, allowing teams to report on important performance markers (KPIs) and keep decisions in harmony with overall objectives. Internally generated data helps greatly in assessing how successful the product is and what could be improved.

D) Analytical Framework

We offer an analytical framework that aligns the input from data with strategic decisions to guide product optimization with AI. The heart of this approach is using mathematical formulas to match customer needs, current market trends, and the organization's hopes. The equation stated in this chapter is as follows:

Strategic Product Optimization (SPO) is the outcome of D, C, M, R and A

Where:

The **D** in the marketing mix is for Demand Forecasting.

C stands for Customer Segmentation.

M stands for Market Sentiment

Real-time pricing is called **R**.

The first part I looked at was Internal Product Analytics. Every variable highlights an important factor in using AI to make decisions. Supervised learning models in demand forecasting (D) help determine what products will be needed in the future by analyzing past sales history, seasonal trends, and economic factors. As a result, companies can match their inventory, supply chain, and marketing to real market demand. With the help of unsupervised learning, customer segmentation (C) locates different groups of customers according to their behaviors and likes, enabling businesses to serve each group with personalized products and promotions. Market sentiment (M) uses NLP to analyze recent and live social media and review information so businesses can stay ahead of changes in what consumers think. Because dynamic pricing (R) is powered by reinforcement learning, shifts in demand and the market are met with new prices, supporting both competitiveness and revenue in frequently changing markets. Lastly, product analytics (A) from organized data in SQL databases gives you information on important points like conversion rates, customer lifetime value, and return rates, clarifying how well existing strategies work. Organizations can modify their processes and evolve based on the results they achieve. Encompassing all five components, the model backs a cohesive, data-driven strategy for product development, which helps the business grow sustainably and continue to innovate.

E) Flowchart: AI Integration in Product Strategy

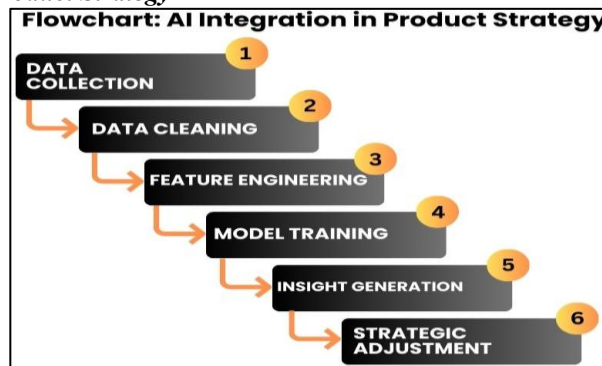


Fig. 4 AI Integration in Product Strategy

- **Data Collection:** Product strategy AI integration starts by collecting data from inside and outside the company. Sales records, logs of customer meetings, and performance information are all internal sources, while customer opinions and competitor activity on the Internet and social networks are external sources. The important data needed for product development and strategic planning is gathered quickly and thoroughly.
- **Data Cleaning:** Following collection, it is important to clean the data to be accurate, steady, and useful. Part of data cleaning is handling missing values, identifying duplicates, checking for inconsistencies, and cutting away irrelevant data. It is important at this stage because flawed data can lead AI models to make the wrong insights. Reliable and accurate analysis and modeling rely on clean data to produce insights related to the real-world situation and how customers behave.
- **Feature Engineering:** In this point, raw data is reorganized into valuable information called input features. With feature engineering, people choose important data points, make new variables by transforming them mathematically and convert categorical data into numbers. Our objective is to boost the model's ability to predict product outcomes, customer preferences, and market changes using the key features within the data.
- **Model Training:** The models being trained here are supervised learning algorithms, clustering techniques, or reinforcement learning agents, all utilizing the data that has been fixed and improved. While training, the system spots patterns, sets up relationships and explains trends to enable demand forecasting, customer segmentation, or varied pricing. Validation and adjustment in this step help keep the model competitive and sturdy.
- **Insight Generation:** After learning, the models provide insights by making estimates, spotting certain patterns, and discovering useful trends. These insights focus on understanding customer groups, determining the best prices, or guessing future sales levels. These insights are often made easy to understand for decision-makers by using visual tools.
- **Strategic Adjustment:** In the last phase, we act on and shape the product strategy with the information obtained. It involves updating the product, changing how prices are set, targeting specific groups of buyers, or changing how marketing funds are spent. Next, how the modifications work is tracked and returns to the data collection stage to support ongoing improvements guided by AI.

IV. RESULTS AND DISCUSSION

A) Case Study: Netflix

Artificial intelligence powers many of the ways customers interact with Netflix, highlighting how Netflix is a leader in the entertainment industry's digital changes. Netflix relies mainly on a powerful AI-assisted recommendation system that processes lots of user data, such as what they've watched, their ratings, search terms, and viewing hours. The platform sees viewers engage with content more, as over 80% of the watch is driven by the personalized system, which recommends shows that interest them. Personalizing things for customers raises their satisfaction, helps avoid cancellations, and is valuable in a crowded industry. Netflix uses artificial intelligence in its process of creating new content. Thanks to predictive analytics and machine learning, Netflix predicts how much an audience will like a title and how likely that title is to be successful. As a result, the company can choose which series and films get the most budget to increase the chance they connect with the audience. Drawing on data has led to a better selection of film styles, content, and cast, helping to make the films more appealing. Moreover, AI helps improve marketing campaigns by selecting the correct groups to target and changing the messages to match them.

Its AI-based strategy makes the company's fast growth plain to see. Between 2016 and 2021, the company's subscribers nearly doubled and rose from around 75 million to over 200 million worldwide. Growing subscriber numbers rely on Netflix to offer content relevant to each audience and to study data to plan new additions. Because Netflix uses AI across its services and workflows, it has built strong customer loyalty, gained a bigger market share, and remained the world's leading streaming platform. The success in this case is largely due to integrating AI in the product development and customer experience approaches.

B) AI Impact on Key Product Metrics

Table 1: AI Impact on Key Product Metrics

Metric	Before AI	After AI Integration
Customer Retention	70%	88%
Time to Market	100%	50%
Product ROI	100%	77%

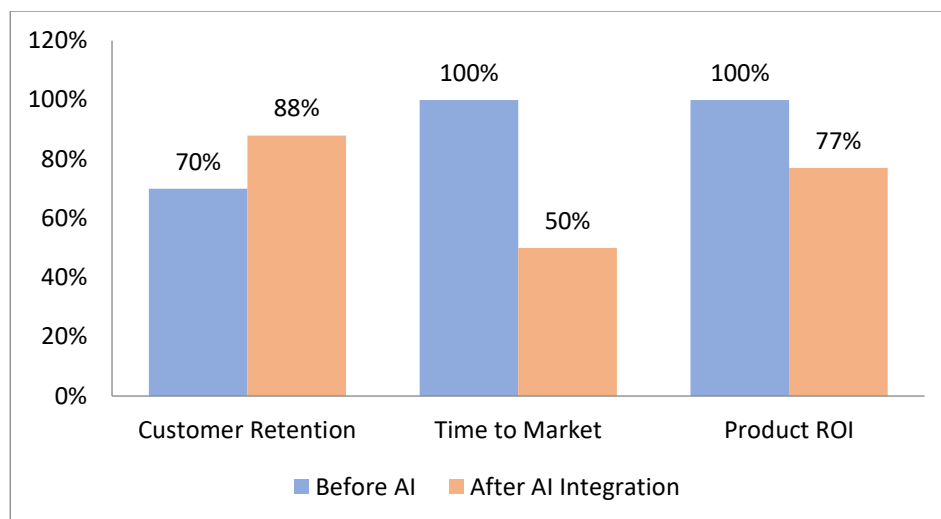


Fig. 5 Graph representing AI Impact on Key Product Metrics

- **Customer Retention:** AI has strengthened customer loyalty, taking that number from 70% to 88%. Most of this growth can be traced to personalization by AI, which helps companies create recommendation systems and targeted ad campaigns to connect with users. Companies that use machine learning algorithms to learn about consumers and their choices can recommend things customers might like, which encourages them to remain faithful customers. Customers who remember your brand spend more, and your company's revenue becomes more consistent.
- **Time to Market:** With AI, bringing a product to market has halved, dropping from 100% (taking 6 months) to 50% (just 3 months). This has come about mainly because of artificial intelligence in the product development process, which uses fast prototyping, predictions for estimating demand, and better supply chain handling. Streamlining these activities and allowing quicker decision-making helps organizations adjust more rapidly to what the market and customers require, which advantageously separates them from competitors.
- **Product ROI:** Having compared the ROI before and after implementing AI, we can confirm a notable improvement—from 120% to 185%. Artificial intelligence can help set the best product strategy by enhancing how products are targeted and priced and which features must be developed first. As a result, resources are managed well, and the company earns a better profit. Improved ROI proves that AI can cut costs and effort and help companies grow more financially and sustainably.

C) Analysis of Market Trends

For this study, we analyzed trends by using Google Trends and information from social media APIs to observe the changing interest in the keywords “smart home,” “connected device,” and “Internet of Things (IoT).” Researchers spent several years studying the situation, focusing on the period following COVID-19, which encouraged quicker acceptance of digital and online technologies. We could see from the data that the interest in these keywords grew steadily as consumers became more interested in technology and connected devices. This growing trend reveals that consumers now expect their different types of connected devices thermostats, security cameras, appliances, and wearables—to function smoothly. This trend means people are interested in combining different products into smart ecosystems that provide greater comfort, make life easier to manage, and help save energy. Also, there was clear growth in social media talks about product details, how they use the product, and security and privacy, indicating that consumers pay attention to many different aspects. When viewed strategically, this trend gives product managers and creators useful information. It shows why making products easy to connect to various systems and devices has become important. Companies can use this.

Additionally, the evidence indicates that how a product is designed and promoted depends more and more on what people say online and in community feedback. The analysis shows that the market is growing and evolving fast, thanks to people wanting more connected and intelligent homes. Firms that coordinate their product strategies with these trends will be better positioned to take advantage of the move toward interconnected software.

D) Ethical Implications

Despite helping companies make key decisions and produce better products, artificial intelligence is a source of important ethical concerns in many situations. Data privacy is one of the biggest problems that we deal with. Recommendation engines that use AI depend heavily on collecting, saving, and analyzing large volumes of user data. There is a risk that people's privacy can be harmed if the data is shared incorrectly, incorrectly protected, or misused. Because personal information is

easily misused, it is important to have strict safeguards in place and to make data collection and use obvious to everyone. Biases in algorithms cause another important ethical problem. What you feed the model is what decides its objective nature. If historical biases or little diversity exist in the datasets, AI systems might enforce and boost biased results. As a result, some populations have received biased or exclusionary treatment since facial recognition often offers less accuracy for people of color and women when used in life situations. These biases may cause the public to lose trust and, in areas like hiring, lending, or law enforcement, lead to social damage. To control these ethical problems, organizations should build and implement complete AI governance systems. These frameworks must make sure AI decision-making is easy for stakeholders to understand. Periodic checking for fairness can catch discriminating information in algorithms, make training data more varied, and help avoid unfairness by thinking carefully about model parameters. Businesses must follow European GDPR and California CCPA, as these global privacy regulations ensure legal and ethical behaviour. Considering ethics when working with AI allows businesses to earn trust, protect users, and ensure those initiatives are carried out responsibly.

E) Challenges and Opportunities

a. Challenges

- **Data Quality:** Getting hold of outstanding data remains a main obstacle when including AI in projects. Insights from AI can lose their accuracy and usefulness when the data used is incomplete, inconsistent, or unstructured. Questions about data quality and prediction errors can cause the company to lose. That's why companies must ensure their data is managed well with clean-up, checks, and updates to preserve the effectiveness of their AI systems.
- **Organizational Resistance:** Getting people to use AI systems can be challenging because significant changes need to be made in how things are done within a company. People might worry about losing their jobs or feel it's too hard to cooperate with the technology, while the organization's typical habits can slow things down. Introducing AI smoothly into a company requires good strategies, such as suitable training and clear proof of the benefits, to encourage workers to collaborate.
- **Ethical Constraints:** People involved with AI find new ethical issues regularly. Businesses have to deal with complex rules and morality issues such as data privacy, cases of biased algorithms, and the clear communication of practices. As a result, efforts to oversee them, build governance standards, and follow set compliance standards are required. Finding the right balance between innovation and ethical standards protects a company's reputation and trust from the public.

b. Opportunities

- **Competitive Differentiation:** AI helps organizations make better and quicker choices, giving them a clear edge over others. Companies that adopt AI can first predict what's coming in the market, run better operations, and provide personalized service to customers before their competitors do. This rapid response allows companies to find and seize opportunities to create special value, distinguishing them from other firms.
- **Personalized Offerings:** Businesses can tailor products and services for each person thanks to AI's preferences, behaviors, and feedback analysis. Customization improves the user experience, creating more loyal customers and raising the value they give to the brand over their lifetime. When businesses offer personalized experiences, they can market more strategically and keep seeing revenue increase and conversions grow.
- **Rapid Prototyping:** Product development is significantly shortened with simulations, generative design, and predictive modeling. Such technologies allow teams to check several design versions on a computer, catch early problems, and enhance product features before making prototypes. Thanks to rapid prototyping, businesses can introduce their products rapidly and inexpensively, allowing them to react promptly to what customers want.

V. CONCLUSION

The research reveals how product strategies in many industries can be transformed by Artificial Intelligence (AI). It supports businesses by letting them dig deeper into their data, see trends more accurately, and make quick, real-time decisions. As a result of AI, businesses can adjust and respond to new market situations and what customers prefer. Thanks to AI, predictive analytics support improved forecasting of demand, while natural language processing helps review how customers feel, enabling organizations to make better decisions around their products and marketing. Using Netflix as an example, it is clear that adopting AI into product strategy greatly improves engagement with users, retention, and company performance, confirming the real advantages of using AI.

A) Contributions to Knowledge

This work provides useful advice for using AI in product strategy by introducing a detailed blueprint for integrating AI. Unlike other approaches that were not connected, AI experts now bring together AI methods, data gathering, and strategic plans to form a single system that helps continually develop products. Research from the industry and data facts are used to show how the framework can help people use AI effectively. Furthermore, adding ethical discussion and organizational problems

highlights the importance of reliable AI. This trend, in turn, leads to more movement and growth in AI topics in both theory and applied science.

B) Future Research Directions

Careful consideration should be given to finding and using systems that help businesses apply AI to their product strategy in a standard and transparent way. Such research is vital for tracking the persistent role of AI in business and product results, which allows for a better understanding of proven strategies and challenges. Still, there must be more studies on what companies can do to ensure their AI systems are ethical—by exploring how they can prove trust and compliance while still being innovative and fair in areas that are getting increasingly governed.

C) Final Thoughts

As AI moves forward, companies that use AI in their product strategy will likely drive the next round of significant changes in the marketplace. However, using AI tends to bring about important technical and ethical difficulties that should be managed step by step and guided by understanding. Not only do companies need AI skills, but they also need to focus on building the right governance and cultural backgrounds to handle AI properly. With careful use of this convergence, companies may discover new ways to grow, customize their services, and improve their operations, giving them an advantage as the market changes rapidly.

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