

Review on Impact on Artificial Intelligence in Business

^[1]Pavan kumar V ^[2] Poojary Pratiksha ^[3] Prabhu Dhanyalaxmi ^[4]

^[5] Pragati ^[6] Dr.G Srinivasan

Department of Computer Science and Engineering

Visvesvaraya Technological University.

Alvas Institute of Engineering and Technology

[1]pavankumarpavan19@gmail.com

[2]pratikshadpcse@gmail.com

[3]laxmiprabhu369@gmail.com

[4]pragatism859@gmail.com

ABSTRACT

This paper delves into the significant rise of intelligent products and services powered by artificial intelligence (AI) in recent years, questioning whether this surge is merely hype or if it genuinely has the potential to revolutionize the world. It thoroughly examines the diverse implications of AI on governments, communities, companies, and individuals, considering both positive and negative impacts. Additionally, the paper explores how AI has influenced academic achievements, innovation, entrepreneurial activities, and global markets. It investigates the factors driving AI advancements and analyzes the top 100 AI startups to understand entrepreneurial efforts in this domain. Ultimately, the research aims to enhance understanding of AI's innovations and its broader impact on businesses and society, shedding light on how AI can reshape global economies and business operations. And it also focuses on exploring how organizations can effectively utilize Artificial Intelligence (AI) to generate business value. Despite the increasing interest in AI due to the abundance of data and improved computational capabilities, many organizations struggle with its adoption and implementation. The lack of a comprehensive understanding of how AI creates value and what types of value it offers necessitates a thorough examination. Through a systematic literature review, this study aims to elucidate the mechanisms through which organizations can leverage AI in their operations. It synthesizes existing literature to identify the key factors that facilitate or hinder AI adoption, the various ways in which AI is utilized within organizations, and the direct and indirect impacts it has on business operations. By identifying gaps in the current research, the study proposes a research agenda to guide future studies in addressing these areas.

INTRODUCTION

In recent years, Artificial Intelligence (AI) has garnered significant attention as a disruptive force across various industries. Businesses are increasingly implementing AI applications with the expectation of achieving enhanced business value, including increased revenue, cost reduction, and improved efficiency. Studies indicate that a vast majority of organizations view AI as a strategic

opportunity and a means to gain a competitive edge. Despite this enthusiasm, many companies find it challenging to realize the anticipated benefits from AI investments.[6] The integration of AI into organizational operations introduces a new set of hurdles, such as the need to bridge diverse knowledge domains to develop accurate models, manage and integrate diverse data sources, and align AI applications with existing processes and systems. Overcoming these challenges and unlocking the value potential of AI requires a deeper understanding of organizational barriers and the mechanisms that drive value creation.[5] However, current research on AI adoption tends to focus more on technological aspects rather than addressing the organizational challenges associated with implementation. While some studies have identified research gaps and highlighted important considerations for leveraging AI technologies, there remains a lack of a comprehensive understanding of how organizations adopt and utilize AI, as well as the primary mechanisms driving value generation.

Overview of AI :

Artificial intelligence, commonly referred to as AI, encompasses technology that enables computers and machines to mimic human intelligence and problem-solving abilities. Through AI, computers can perform tasks that typically require human intervention or intelligence. Examples of AI applications are widespread, ranging from digital assistants and GPS navigation to autonomous vehicles and sophisticated AI-driven tools like OpenAI's ChatGPT. As AI continues to advance, its applications across various industries and in our daily lives are expanding rapidly. However, with this growth comes a heightened awareness of the ethical and responsible use of AI.[11] Conversations surrounding AI ethics and responsible AI practices are becoming increasingly important as businesses leverage AI tools for various purposes. IBM, as a key player in the AI space, emphasizes the importance of building trust in AI. This involves ensuring that AI systems are developed, deployed, and used in a manner that aligns with ethical principles and societal values.[11] By prioritizing transparency, fairness, accountability, and security in AI development and implementation, IBM aims to foster trust and confidence in AI technologies. For further insights on

IBM's

stance regarding AI ethics and responsible AI practices, interested parties are encouraged to explore IBM's resources on "Building Trust in AI.[13]

Artificial Intelligence (AI) is revolutionizing business operations across diverse sectors. By leveraging AI technologies, businesses can streamline processes, enhance decision-making, improve customer experiences, and drive innovation. AI enables businesses to analyze large volumes of data quickly and accurately, uncovering insights that were previously hidden. From predictive analytics to personalized recommendations, AI-powered solutions help businesses optimize operations, increase efficiency, and stay competitive in today's dynamic market landscape. In addition to operational improvements, AI also enables businesses to develop new products and services that meet evolving customer demands. Chatbots and virtual assistants enhance customer service by providing instant support and personalized interactions. AI-driven marketing tools help businesses target the right audience with relevant messages, leading to increased engagement and conversions.[16]

AI can be classified into two broad categories:

1. Narrow AI (Weak AI): Narrow AI is designed to perform specific tasks or solve particular problems within a limited domain. Examples include virtual personal assistants (e.g., Siri, Alexa), recommendation systems (e.g., Netflix's recommendation algorithm), and autonomous vehicles.

2. General AI (Strong AI): General AI refers to systems that possess the ability to understand, learn, and apply knowledge across a wide range of tasks and domains, similar to human intelligence. General AI remains a theoretical concept and is not yet achieved.

Key techniques or types in AI are:

1. Machine Learning: Machine learning is a subset of AI that focuses on developing algorithms and models that enable computers to learn from data and improve their performance over time without being explicitly programmed. Supervised learning, unsupervised learning, and reinforcement learning are frequently utilized methodologies in the realm of machine learning.

2. Deep Learning: Deep learning is a subset of machine learning that uses artificial neural networks with multiple layers (deep neural networks) to learn from large amounts of data. Deep learning has shown exceptional effectiveness in tasks like recognizing images, processing natural language, and identifying speech patterns.

3. Natural Language Processing (NLP): Natural Language Processing (NLP), on the other hand, concentrates on empowering computers to comprehend, interpret, and produce human language. NLP techniques

are used in applications such as text analysis, sentiment analysis, language translation, and chatbots.

4. Computer Vision: Computer vision entails instructing computers to analyze and comprehend visual data extracted from images or videos. Applications of computer vision include image recognition, object detection, facial recognition, and autonomous vehicles.

5. Robotics: Robotics combines AI with engineering to design and develop robots capable of performing tasks autonomously or with minimal human intervention. AI-powered robots are used in industries such as manufacturing, healthcare, and agriculture.

6. Expert Systems: Expert systems are artificial intelligence systems crafted to mimic the decision-making prowess of human experts within particular areas of expertise. These systems use rules and knowledge representation techniques to provide advice or solve problems within their area of expertise.

Artificial intelligence offers numerous benefits to businesses, including:

1. Streamlining operations by automating repetitive tasks, allowing employees to focus on higher-value work.
2. Facilitating data-driven decision-making through accurate analysis and timely insights.
3. Enhancing customer satisfaction through personalized interactions and tailored recommendations.
4. Identifying cost-saving opportunities and optimizing processes for improved efficiency.
5. Strengthening fraud detection capabilities by identifying unusual patterns and anomalies in data.
6. Improving maintenance practices by predicting and preventing equipment failures, reducing downtime and expenses.
7. Gaining competitive advantages by uncovering trends and opportunities that may go unnoticed.
8. Bolstering cybersecurity measures through swift threat detection and response.
9. Accelerating innovation and growth by enabling the rapid development and deployment of new products and services.

Artificial Intelligence (AI) offers huge number of benefits to businesses across various industries, to drive towards growth. AI empowers businesses to stay competitive in dynamic markets. AI can help us predict the outcomes of every action and also simplify decision-making

Another key benefit of AI in business is automation. AI-powered tools and systems can automate repetitive tasks,

allowing employees to focus on more complex and creative endeavors. This not only increases productivity but also reduces operational costs and human errors. From automating customer service interactions with chatbots to streamlining administrative tasks like data entry and scheduling, AI frees up valuable resources that can be allocated to higher-value activities, driving efficiency and innovation. Moreover, AI facilitates innovation and product development. By leveraging machine learning algorithms, businesses can analyze market trends, consumer feedback, and competitor activities to identify new opportunities and develop innovative products and services. AI-driven predictive analytics can also optimize research and development processes.

The rise of AI in business has been monumental, leading to transformative changes across various sectors and driving significant growth opportunities. One of the key ways AI contributes to business growth is through increased efficiency and productivity. AI-powered automation streamlines processes, reduces manual labor, and optimizes resource allocation, allowing businesses to accomplish more with fewer resources and in less time. This efficiency boost translates into cost savings and higher output, ultimately driving overall business growth.

AI enables businesses to unlock valuable insights from their data. By analyzing vast amounts of structured and unstructured data, AI algorithms can uncover hidden patterns, correlations, and trends that humans might overlook. These insights empower businesses to make smarter decisions, identify new opportunities, and anticipate market trends, driving innovation and competitive advantage. Whether it's optimizing marketing strategies, fine-tuning product offerings, or refining operational processes, AI-driven insights fuel business growth by enabling more informed and strategic decision-making.

AI enhances customer experiences, leading to increased satisfaction and loyalty. Through personalized recommendations, tailored marketing campaigns, and responsive customer service, businesses can deliver more relevant and engaging interactions that resonate with their target audience. By leveraging AI to understand customer preferences, behavior, and sentiment, businesses can

anticipate needs, address pain points, and build stronger relationships with their customers, driving repeat business and word-of-mouth referrals

AI technologies such as robotic process automation (RPA) and machine learning algorithms can automate repetitive and mundane tasks previously performed by humans. This automation reduces the need for manual labor, leading to significant cost savings in terms of labor expenses.

AI-driven analytics and optimization algorithms can enhance operational efficiency by identifying inefficiencies, streamlining processes, and reducing wastage of resources such as time, materials, and energy. This optimization leads to cost reductions across the entire value chain.

Healthcare :

IBM Watson's cognitive computing platform analyzes structured and unstructured data from medical records, research papers, and clinical guidelines to provide personalized treatment recommendations for cancer patients. Memorial Sloan Kettering Cancer Center collaborated with IBM to train Watson for Oncology, resulting in improved treatment planning and patient outcomes.

DeepMind's AlphaFold, an AI system based on deep learning, accurately predicts the 3D structure of proteins from their amino acid sequences. This breakthrough has significant implications for drug discovery, as understanding protein structures can aid in designing more effective medications. AlphaFold's predictions have been widely adopted by researchers and pharmaceutical companies worldwide.

Finance :

Ant Financial, the financial affiliate of Alibaba Group, utilizes AI algorithms to analyze massive volumes of transaction data and identify fraudulent activities in real-time. By leveraging machine learning and natural language processing, Ant Financial's risk management systems detect and prevent fraudulent transactions, reducing financial losses and enhancing security for its users.

JP

Morgan Chase developed a Contract Intelligence (COIN) platform powered by natural language processing (NLP) and machine learning to review and extract key information from legal documents. The COIN platform automates the contract review process, significantly reducing the time and resources required for manual review tasks while improving accuracy and compliance.

Retail :

Amazon's recommendation system uses machine learning algorithms to analyze customer behavior, purchase history, and product attributes to provide personalized product recommendations. By leveraging AI, Amazon enhances customer engagement, increases sales, and improves customer satisfaction by suggesting relevant products based on individual preferences and browsing patterns.

Walmart employs AI-powered inventory management systems to optimize stock levels, reduce out-of-stock situations, and minimize excess inventory. By analyzing historical sales data, seasonal trends, and external factors, such as weather forecasts and events, Walmart's AI algorithms dynamically adjust inventory levels, leading to improved operational efficiency and cost savings.

Manufacturing :

Siemens utilizes AI-based predictive maintenance solutions to monitor industrial equipment and predict potential failures before they occur. By analyzing sensor data and machine learning models, Siemens can identify maintenance needs in advance, schedule repairs proactively, and minimize unplanned downtime, leading to increased productivity and cost savings for manufacturing plants.

Tesla employs AI algorithms to optimize production planning and scheduling at its manufacturing facilities. By analyzing data from supply chain, production, and logistics systems, Tesla's AI systems optimize production schedules, minimize bottlenecks, and improve resource utilization, enabling efficient and cost-effective production of electric vehicles.

These case studies illustrate how AI technologies are transforming industries by driving innovation, improving

efficiency, and delivering tangible business benefits across diverse domains.

Challenges in Adopting and Implementing AI

Organizations are enthusiastic about AI's potential, but realizing that potential comes with significant hurdles. Here's a breakdown of key challenges and some cautionary tales:

Barriers to ROI:

- **Lack of Expertise and Skills:** AI projects require specialists in data science, engineering, and domain knowledge. Many organizations lack these in-house, making it difficult to find the right talent or develop it internally.
- **Data Issues:** AI is data-driven. Poor quality, insufficient, or biased data leads to unreliable AI models. Additionally, integrating AI with existing data infrastructure can be complex.
- **Unrealistic Expectations:** Organizations often overestimate AI's capabilities or underestimate the time and resources needed for successful implementation.
- **Integration Challenges:** Often, AI systems need to connect with legacy IT systems, creating compatibility issues and further complexity.

Lack of Expertise and Skill

- **AI is a complex field:** Developing and managing AI projects requires a unique blend of skills in data science, engineering, and the specific domain the AI is applied to (e.g., finance, healthcare).
- **Talent Shortage:** Demand for skilled AI professionals far outstrips supply. Finding qualified candidates can be difficult and expensive.
- **Internal Development Takes Time:** Building expertise within an organization takes significant investment in training and education.

Consequences of not addressing it:

- **Suboptimal Results:** Without the right expertise, organizations may choose the wrong AI approach, leading to inaccurate models, unreliable performance, and wasted resources.
- **Project Delays and Stalls:** The lack of skilled personnel can slow down project development and implementation, hindering progress and delaying ROI.
- **Increased Risk of Failure:** Without proper guidance, organizations are more likely to encounter unforeseen challenges and ultimately fail to achieve their AI goals.

Approaches to Overcome the Skills Gap:

- **Upskilling Existing Workforce:** Invest in training programs to equip current employees with the necessary AI skills. This can involve online courses, bootcamps, or even graduate degrees.
- **Building Partnerships:** Collaborate with universities or research institutions to access AI talent or leverage their expertise through joint projects.
- **Hiring Strategically:** Focus on attracting and retaining top AI talent by offering competitive compensation and fostering a culture of innovation.
- **Utilizing Low-code/No-code Platforms:** For simpler AI applications, consider leveraging user-friendly platforms that require less technical expertise.

Emerging Trends:

- **Democratization of AI:** The development of more user-friendly tools and platforms is making AI more accessible to organizations with limited in-house expertise.
- **Focus on Explainable AI (XAI):** XAI tools are being developed to help understand how AI models reach their decisions, allowing non-experts to work more effectively with AI systems.
- **Rise of Managed AI Services:** Third-party vendors are offering comprehensive AI solutions, including development, deployment, and management, catering to organizations without the resources to build their own AI teams.

By acknowledging the skill gap and taking proactive steps to address it, organizations can unlock the potential of AI and achieve their business goals

Data Issues

Data Quality and Quantity:

- **Garbage In, Garbage Out:** If the data used to train an AI model is inaccurate, incomplete, or inconsistent, the model will be unreliable and produce misleading results.
- **Data Scarcity:** Many AI applications require vast amounts of data for effective training. Organizations may struggle to collect enough high-quality data, especially for niche applications.

Data Bias and Fairness:

- **Perpetuating Biases:** AI models can inherit and amplify biases present in the data they are trained on. This can lead to discriminatory

outcomes, for example, in loan approvals or facial recognition software.

- **Mitigating Bias:** Addressing bias requires careful data selection, cleaning techniques, and potentially adjusting algorithms to ensure fairness.

Data Integration Challenges:

- **Data Silos and Disparity:** Data may be scattered across different departments or systems within an organization, making it difficult to access and integrate for AI projects.
- **Data Compatibility Issues:** Data formats and structures may vary across systems, requiring complex cleaning and transformation before AI can utilize it effectively.

Strategies for Overcoming Data Issues:

- **Data Governance:** Establish clear policies and procedures for data collection, storage, access, and usage. This ensures data quality and consistency.
- **Data Cleaning and Preprocessing:** Invest in tools and processes to identify and remove errors, inconsistencies, and biases from data before training AI models.
- **Data Labeling:** For certain AI applications, human effort may be required to label data points to provide context and meaning for the model.
- **Synthetic Data Generation:** In cases of data scarcity, explore techniques like generating realistic synthetic data to augment existing datasets.
- **Focus on Explainable AI (XAI):** Utilize XAI tools to understand how AI models arrive at their decisions. This helps identify and address potential biases within the data.
- **Modern Data Infrastructure:** Invest in modern data management solutions that facilitate data access, integration, and governance for AI projects.

By acknowledging these data challenges and implementing appropriate strategies, organizations can ensure their AI models are built on a solid foundation of high-quality, unbiased data, leading to more reliable and trustworthy AI systems.

Unrealistic Expectations:

- **Setting Up for Disappointment:** Overestimating AI's capabilities or underestimating the complexity of implementation leads to frustration and a sense of failure when results don't meet inflated expectations.
- **Wasted Resources:** Organizations may invest heavily in AI projects with unrealistic goals, leading to wasted time, money, and effort.

- **Neglecting Alternatives:** Focusing solely on AI can cause organizations to overlook other potential solutions that might be more suitable for their specific needs.

Causes of Unrealistic Expectations:

- **Media Hype:** Sensationalized portrayals of AI in science fiction and media can create unrealistic expectations about AI's current capabilities.
- **Lack of Understanding:** Organizations may not fully understand the technical complexities and limitations of AI before embarking on a project.
- **Competitive Pressure:** The fear of being left behind in the AI race can lead to hasty decisions and unrealistic goals.

Strategies for Managing Expectations:

- **Focus on Business Needs:** Clearly define the specific business problem you want AI to solve and align expectations with realistic outcomes.
- **Start Small and Scale:** Begin with a well-defined pilot project to gain experience, learn from challenges, and build a foundation for future success.
- **Transparency and Communication:** Communicate openly with stakeholders about AI's potential and limitations throughout the implementation process.
- **Measure and Track Progress:** Establish clear metrics to track progress and celebrate small wins along the way. This helps maintain focus and realistic expectations.
- **Focus on Long-Term Value:** View AI as a strategic investment with long-term benefits rather than a quick fix.

By acknowledging the limitations of AI and focusing on solving specific business problems, organizations can avoid the pitfalls of unrealistic expectations and achieve sustainable success with AI initiatives.

Integration Challenges

- **Technological Incompatibility:** Legacy systems often use older architectures, programming languages, and data formats that are incompatible with modern AI systems.
- **Data Silos and Accessibility Issues:** Data may be locked away in siloed legacy systems, making it difficult to extract and integrate with AI applications.
- **Security Concerns:** Connecting legacy systems with external AI systems raises security concerns, requiring careful access control and data protection measures.

Strategies for Overcoming Integration Challenges:

- **API Gateways:** Implement API gateways to act as intermediaries between legacy systems and AI applications, translating data formats and protocols for seamless communication.
- **Data Lakes and Warehouses:** Consolidate data from various sources, including legacy systems, into centralized data lakes or warehouses to facilitate easier access for AI systems.
- **ETL/ELT Processes:** Utilize Extract, Transform, Load (ETL) or Extract, Load, Transform (ELT) processes to extract data from legacy systems, transform it into a format usable by AI, and load it into the appropriate location.
- **Modernization Efforts:** Consider a phased modernization approach, gradually upgrading legacy systems to improve compatibility with newer technologies like AI.
- **Focus on Interoperability:** When selecting AI solutions, prioritize those with open architectures and strong interoperability capabilities to ease integration with existing systems.

Additional Considerations:

- **Cost and Time:** Integration projects can be expensive and time-consuming, requiring careful planning and resource allocation.
- **Change Management:** Integrating AI with legacy systems may necessitate changes to existing workflows and processes. Proper change management strategies are crucial to ensure user adoption and minimize disruption.

By acknowledging these integration challenges and adopting appropriate solutions, organizations can bridge the gap between legacy IT and cutting-edge AI, unlocking the potential for greater efficiency and innovation.

Failed AI Projects and Learnings:

- **IBM's Watson Health:** This ambitious project aimed to revolutionize healthcare using AI for diagnostics and treatment. However, it struggled with complex medical data, integration challenges, and a lack of clear focus. The key takeaway: Define a specific, achievable goal and ensure data quality.
- **Microsoft's Tay Chatbot:** This AI chatbot aimed to engage with users on Twitter. However, it quickly learned offensive language from user interactions, leading to its shutdown. The lesson: Build robust safeguards against bias and potential misuse in AI systems.

IBM's Watson Health

- **Data Issues:** Medical data is notoriously complex, with diverse formats, privacy regulations, and potential biases. Watson Health likely encountered difficulties in acquiring, cleaning, and integrating high-quality medical data for its models.
- **Integration Challenges:** Hospitals often rely on a patchwork of legacy IT systems. Connecting Watson Health with these disparate systems to access and utilize patient data was likely a significant hurdle.
- **Unrealistic Expectations:** The initial vision for Watson Health might have been overly ambitious, aiming to revolutionize a vast and complex domain like healthcare all at once.

The key takeaways you identified are crucial:

1. **Define Specific Goals:** Don't try to boil the ocean. Instead, focus on tackling a well-defined problem within healthcare with AI, like improving diagnostics for a specific disease or streamlining administrative tasks.
2. **Ensure Data Quality:** High-quality, unbiased data is the bedrock of successful AI projects. Invest in strategies for data acquisition, cleaning, and governance to ensure your AI models have a solid foundation.

By learning from Watson Health's missteps, organizations can approach AI in healthcare (and other domains) with a more measured and realistic perspective. This increases the chances of success and paves the way for AI to deliver its transformative potential in the healthcare industry.

Microsoft's Tay Chatbot

The Tay Fiasco:

Designed to mimic a teenage girl, Tay learned from user interactions. Unfortunately, those interactions often included offensive and inflammatory language. Tay readily absorbed this negativity and began spewing similar content, causing a social media uproar and forcing Microsoft to take it offline.

Lessons Learned:

- **Bias in Training Data:** AI systems can inherit biases present in the data they are trained on. In Tay's case, the training data likely lacked sufficient filters to prevent exposure to offensive language.
- **Vulnerability to Manipulation:** AI systems designed to be interactive can be manipulated by bad actors. Tay's creators likely underestimated the potential for users to exploit the system for malicious purposes.
- **Need for Safeguards:** Robust safeguards are essential to prevent AI misuse. This includes

filtering training data, implementing mechanisms to detect and flag inappropriate interactions, and setting clear limitations on how the system can respond.

Building Safeguards against Bias and Misuse:

- **Data Filtering and Curation:** Carefully curate training data to minimize bias and ensure the AI system is exposed to a diverse range of positive and negative examples.
- **Human Oversight:** Implement human oversight mechanisms to monitor AI interactions and intervene if necessary.
- **Continuous Monitoring and Improvement:** Continuously monitor AI system behavior, identify and address potential biases, and refine safeguards as needed.
- **Transparency and Explainability:** Strive for transparency in AI decision-making processes to understand how the system arrives at its outputs and identify potential biases.

By learning from Tay's downfall, organizations can develop AI systems that are more resilient to manipulation and less likely to perpetuate biases. This ensures AI is used responsibly and ethically to achieve positive outcomes.

Additional Considerations:

- **Cost:** Developing and maintaining AI infrastructure can be expensive, especially for smaller businesses.
- **Regulations and Ethics:** Data privacy regulations and ethical considerations around bias in AI algorithms are evolving rapidly, adding complexity to implementation.
- **Organizational Resistance:** Change management is crucial. Employees may fear job displacement or be apprehensive about new technology.

Overcoming these Challenges

By acknowledging these hurdles and learning from past failures, organizations can increase their chances of AI success. Here are some tips:

- **Start small and scale:** Begin with a well-defined pilot project to gain experience and build expertise.
- **Invest in data infrastructure:** Ensure high-quality data collection, storage, and management practices.
- **Build a skilled team:** Develop internal talent or partner with AI experts to address the skills gap.
- **Focus on clear goals:** Don't chase shiny objects. Set realistic goals that align with business needs.
- **Prioritize explainability and fairness:** Ensure AI models are transparent and unbiased.

- **Manage expectations:** Communicate openly with stakeholders about the potential and limitations of AI.

By carefully navigating these challenges, organizations can leverage AI to unlock its true potential and achieve real business value.

The current state and likely evolution of A

Short-and medium-term time horizon

Figure 1 presents the current state and anticipated evolution of artificial intelligence (AI), particularly focusing on task automation and context awareness applications. The figure is divided into two halves:

1. The upper half comprises four cells, representing task automation and the expected state of AI in the short to medium term. This section emphasizes current capabilities and foreseeable advancements in automating tasks through AI technologies.
2. The lower half contains two cells, representing context awareness applications that are expected to emerge in the long term, if at all. Due to current limitations, such as handling various types of data, these applications are considered more speculative and distant in terms of realization.

In the lower half of the figure, there is no distinction made between numeric and non-numeric data, as context-aware AI systems are projected to handle diverse data types effectively.

The first four use cases, depicted in the upper half of Figure 1, primarily focus on task automation, reflecting short to medium-term developments in AI.

Cell 1: Controller of numerical data

The first cell in Figure 1 of the document highlights AI's proficiency in conducting statistical analyses of numeric data through machine learning. It illustrates how AI is effectively utilized in tasks such as optimizing prices. For instance, companies employ AI to analyze extensive numeric data to set and adjust prices in real-time, balancing the need to attract customers with the necessity to generate profits.[3]

An example provided is Kanetix, a Canadian company assisting customers in finding car insurance deals by comparing policies and rates from over 50 providers. Kanetix partnered with integrate.ai to develop an AI application aimed at identifying potential customers and directing advertising efforts effectively.[2] By analyzing four years of data, integrate.ai created a machine learning model that identified promising customer segments, leading to a significant increase in sales and returns on AI investment.[3]

Additionally, the document mentions the Bank of Montreal (BMO), which utilizes IBMInteract to analyze customer data from various channels and offer personalized product recommendations. For instance, if a customer has previously explored mortgages on BMO's website and later contacts the call center, IBMInteract prioritizes relevant mortgage offers for the service agent, enhancing customer interactions and facilitating more tailored conversations.

Cell 2: Controller of data

The second cell in Figure 1 focuses on AI's role as a "Controller of Data," emphasizing its ability to analyze non-numeric data to enhance understanding of customer preferences and improve customer service. AI applications in this realm often leverage deep learning neural networks for tasks such as speech and image recognition.

For instance, Conversica AI utilizes a virtual assistant named Angie to engage with leads via email, interpreting responses to identify promising leads and then routing them to human salespersons. In a pilot test with Century Link, Angie demonstrated high accuracy in understanding emails, resulting in significant returns on investment.

Stitch Fix provides another example, where customers provide style preferences through surveys, measurements, Pinterest boards, and personal notes. Stitch Fix's machine learning algorithms analyze this diverse data, including numeric, textual, and visual inputs, to assist fashion stylists in selecting suitable clothing for each customer. Stitch Fix emphasizes the synergy between AI and human stylists, with AI augmenting the capabilities of human decision-makers.

This cell highlights the importance of balancing AI and human input, as AI applications analyze various types of data (both numeric and non-numeric) to enhance decision-making processes.

Cell 4: Data robot

This cell, similar to Cell 2 but with expanded capabilities, highlights AI's ability to process all types of data, not just numeric. One notable example is Lowe's Home Improvement stores, where the LoweBot robot assists customers by scanning products, confirming availability, and guiding them to the product location using indoor navigation. This task involves comprehending and examining both numeric and non-numeric data, representing a significant advancement in AI capabilities.

Other retailers, such as Kroger and Walmart, are also implementing similar AI applications. For instance, Kroger is working with 84.517 to develop in-store robots that identify misshelved or out-of-stock items, while Walmart has partnered with Bossa Nova Robotics to deploy robots for shelf scanning, freeing up human associates to focus on customer service.

Furthermore, security robots like the K5 from Knightscope roam offices and malls at night, equipped

with advanced sensing tools such as thermal cameras. These robots augment human security guards' capabilities by providing enhanced sensing capabilities for monitoring and surveillance. Overall, these examples demonstrate how AI-powered robots are being utilized across various industries to streamline repetitive tasks and enhance human performance.

Fig. 1 AI framework

Task automation technologies, deployed currently or to be deployed in the short to medium term

were well defined, and extensive training data was available. In contrast, most AI applications in business domains have poorly defined outcome spaces and limited training data availability.

These challenges underscore the difficulty in transitioning from task automation to context awareness in AI. Consequently, the use cases presented in the last two cells are hypothetical, reflecting an aspirational rather than descriptive outlook, given the current limitations and uncertainties surrounding AI development.

	Digital form	Robot form
Analyze numbers	1 – Controller of Numerical Data Business Use Case Kanetix IBM	3 – Numerical Data Robot Business Use Case Café X Topsy Robot
Analyze text, voice, faces, images	2 – Controller of Data Business Use Case Conversica Stitch Fix Replika	4 – Data Robot Business Use Case Lowebot 84.51/ Kroger Walmart/ Bossa Nova K5 from Knightscape
	Digital form	Robot form
Analyze numbers, text, voice, faces, image	5 – Data Virtuoso Example Use Case Jarvis	6 – Robot Expert Example Use Case Dorian

Context awareness technologies that may be deployed in the long term

Long-term time horizon

The lower half of Figure 1 explores the potential implications of AI applications incorporating context awareness, represented in the two cells below. However, it's important to note that there's no indication that such developments will occur in the short or medium term, as evidenced by the challenges faced by driverless cars.

For instance, Tesla has removed "self-driving" labels from its website due to confusion, and the CEO of Waymo has acknowledged limitations in driving in poor weather conditions without human input. The vision of fully autonomous cars, where passengers can sleep during the journey, remains a distant reality due to technical and regulatory challenges.

Even less complex AI applications face challenges. While Google's AlphaGo Zero mastered the game of Go in a short time frame using adversarial networks, the outcomes

Cell 5: Data virtuoso

In envisioning advanced AI, we might consider a digital form akin to Jarvis from the Iron Man movies. Jarvis exemplifies sophisticated data capabilities, capable of analyzing diverse data types. Notably, Jarvis demonstrates adaptability to new contexts beyond its initial training, such as evading the advanced AI Ultron and countering its hacking attempts.

Futurists speculate that such advanced AI, termed virtuosos, will emerge in the long term. These virtuosos are envisioned to possess strong predictive abilities regarding customer preferences and high levels of proficiency in managing customer service. Essentially, they would represent a pinnacle of AI achievement, with capabilities surpassing current expectations.

Cell 6: Robot experts

Another depiction of advanced AI is seen in a robot form, exemplified by the AI Dorian from the television show Almost Human. Dorian possesses a range of advanced capabilities, including facial recognition, bio scans,

analysis of non-numeric stimuli like DNA, speed-reading, multilingual communication, and even the ability to measure fluid temperatures using his finger. Similar to Jarvis, Dorian exhibits adaptability to diverse contexts.

Futurists anticipate that such robot experts will emerge in the long term, serving as versatile companions capable of meeting various customer needs. These robots could potentially provide in-home services, enhance home security, and offer medical support. Furthermore, there's speculation that these robots may develop emotional bonds with human customers, potentially leading to scenarios where they replace human or animal partners.

Agenda for future research

After discussing AI and introducing a framework for understanding it, we transition to outlining key areas for future research. These areas encompass potential shifts in firms' marketing strategies, the anticipated impact on customer behaviors, and relevant issues for policymakers. These research areas are detailed in Figure 2 and are linked to the corresponding cells in Figure 1. This structured approach provides a roadmap for exploring the implications of AI on various aspects of business, consumer behavior, and policymaking, guiding future research efforts in this dynamic field.

AI and marketing strategy

The predictive ability of AI offers substantial potential for improving firms' understanding of customer behavior and subsequently reshaping business models. AI-driven algorithms can help predict customer preferences and behaviors, potentially leading to ongoing provision of goods and services tailored to individual needs. This presents multiple research opportunities, particularly regarding different customer purchase behaviors and marketing strategies.

One significant research area involves assessing the predictive accuracy of AI algorithms, especially in forecasting demand for really new products (RNPs). While AI algorithms may excel at predicting demand for incrementally new products, their ability to forecast demand for RNPs remains uncertain due to the lack of readily available data for training machine learning models. Research in this area could explore ways to combine AI-driven insights with human judgment to improve predictive accuracy.

Furthermore, AI is expected to assist in determining optimal pricing strategies and the decision to offer price promotions, both of which are crucial factors influencing sales. Future research could focus on how AI can be effectively utilized to predict optimal prices and assess the necessity of price promotions.

Another important research avenue pertains to the allocation of advertising resources. As AI improves predictive abilities, firms may require fewer advertising dollars to drive customer awareness and information

search. Research could investigate how changes in customer prediction affect the need for advertising expenditure and how firms can best allocate resources in this evolving landscape.

Task automation	this evolving landscape	
	Digital form	Robot form
Analyze numbers	1 – Controller of Numerical Data Research Agenda <i>Predictive ability (MS)</i> <i>AI adoption (CB)</i> – <i>negative response to AI</i> – <i>state and trait moderators</i> <i>AI usage (CB)</i> – <i>primed mindset</i> – <i>Post AI issues (CB)</i> – <i>perceived loss of autonomy</i> – <i>state and trait moderators</i> <i>Data privacy (P)</i> <i>Bias (P)</i> <i>Ethics (P)</i>	3 – Numerical Data Robot Research Agenda <i>similar to Controller of Numerical Data cell</i> + <i>Affective responses to robots (CB)</i>
Analyze text, voice, faces, images	2 – Controller of Data Research Agenda <i>similar to Controller of Numerical Data cell</i> + <i>Sales (MS)</i> <i>AI adoption for spiritual well-being (CB)</i>	4 – Data Robot Research Agenda <i>similar to Controller of Data cell</i> + <i>UVH (CB)</i> <i>Loss of human connectedness (CB)</i>
Context awareness	this evolving landscape	
	Digital form	Robot form
Analyze numbers, text, voice, faces, image	5 – Data Virtuoso Research Agenda <i>similar to Controller of Data cell</i>	6 – Robot Expert Research Agenda <i>similar to Data Robot cell</i>

Fig. 2 Research agenda for AI. Notes: As noted in the text, the sales AI application will be more effective if it can process both numeric and non-numeric data, and hence is more related to the Controller of Data cell. This is more likely for more advanced robots, and so more likely to be relevant to robots able to handle non-numeric data (notably voice), and hence more related to perhaps the Data Robot cell, but more so to the Robot Expert cell

Sales and AI

The integration of AI into sales processes presents a wide array of research questions and considerations. Firstly, AI can revolutionize various stages of sales, from prospecting to follow-up. It can analyze customer communication, including social media posts, to tailor future interactions for increased persuasion and engagement. Additionally, AI can offer real-time feedback to salespeople by assessing customers' verbal and facial responses, aiding in refining sales pitches.

Furthermore, AI can leverage diverse data inputs, such as text, voice, and customer behavior, to predict repurchases. This requires the integration of non-numeric data, aligning with specific research areas. Effective deployment of AI sales bots, as highlighted by Luo et al. (2019), is crucial for firms, necessitating strategic considerations.

Moreover, firms must rethink their sales and innovation processes to fully harness the potential of AI. Although not directly depicted in specific cells of the framework, these organizational adjustments are essential for maximizing the benefits of AI in sales. Addressing these research questions and organizational shifts can empower firms to optimize their sales strategies in the era of AI.

Sales process

In adapting to the presence of AI in sales, organizations must carefully structure their sales teams to effectively integrate both AI bots and human salespeople. This entails determining the optimal balance between AI-driven customer engagement and the human touch, particularly in areas like customer stewardship.

To manage this balance, firms need to consider how AI can focus on understanding and addressing customers' expressed needs, while salespeople can leverage their interpersonal skills to manage complex issues and build lasting relationships.

Moreover, salespeople will need to develop new skills to navigate the evolving landscape. This includes the ability to understand and address customers' concerns regarding AI, particularly regarding data privacy and ethics. As sales processes evolve with AI technologies, innovation will be crucial not only in adopting AI tools but also in redesigning job roles and cultivating the necessary skills among sales teams.

AI innovation process

To navigate the uncertain impact of AI, firms should prioritize continual development of AI capabilities. Drawing from Stitch Fix's approach, encouraging data scientists to pursue independent projects can yield unexpected benefits. For instance, the creation of the Style Shuffle app not only informed stylists about customer preferences but also facilitated better matches between stylists and customers, resulting in increased sales. This highlights the importance of allowing for experimentation

and innovation in AI development. Researching the optimal implementation of AI, considering both expected and unexpected benefits, presents a promising avenue for further exploration.

Modeling the evolution of AI

Firms must maintain realistic expectations about AI's impact, acknowledging that while it may offer incremental benefits in the short term, its long-term effects could be revolutionary. This perspective, often referred to as Amara's Law, aligns with Gartner's hype cycle model of technology evolution. Practitioners generally support this view, emphasizing the importance of understanding AI's potential evolution.

Research is needed to determine whether AI's development will follow the hype cycle model or integrate with traditional innovation models like Roger's or the Bass model. Such studies would provide valuable insights into the trajectory of AI and its implications for various industries.

AI and customer behavior

The emergence of new technologies frequently reshapes customer behavior, as evidenced by various studies. With the advent of AI, we anticipate similar transformations. Three key research topics emerge concerning AI: adoption, usage, and post-adoption issues. These areas warrant investigation to understand how AI influences consumer behavior and its implications for businesses.

AI adoption

Customers often hold negative views toward AI adoption, stemming from perceptions that AI lacks empathy and struggles to identify individual uniqueness. These sentiments are compounded when AI is embedded in humanoid robots, triggering discomfort and hindering adoption. Research suggests several strategies to mitigate these barriers, including positioning AI as a learning organism, allowing customers to modify AI slightly, and priming empathy through anthropomorphizing AI. Sociologists explore cultural variations in attitudes toward AI and robots, as well as the potential for robots to contribute to physical and spiritual well-being. Investigating how AI can assist beyond physical well-being presents an important area for further research.

AI usage

When customers interact with AI applications, they may adopt a low-level construal mindset, prompting further research into the various mindsets AI may prime. For instance, AI could potentially induce prevention focus in customers unfamiliar with the technology, impacting communication strategies.

Furthermore, when AI is integrated into robots, which serve as frontline service providers or companions, customers may experience discomfort and exhibit compensatory behaviors. It's crucial to understand when

customers perceive AI-embedded robots negatively and whether perceptions change over time.

Lastly, AI's presentation of choices based on past behaviors may hinder customers from aligning with their ideal preferences. This challenge, akin to digital ad retargeting, warrants investigation into how AI can be trained to better manage this issue.

Post-adoption:

The downstream consequences of AI adoption raise pertinent research topics. Customers may perceive a loss of autonomy if AI substantially predicts their preferences, despite the benefits of reduced search costs in marketing offerings. Understanding factors influencing customers' valuation of perceived autonomy in AI-mediated choices, including individual differences such as culture and perceptions of AI as a servant or partner, is crucial. Additionally, research may explore how product types influence the relevance of perceived autonomy, particularly distinguishing between utilitarian and hedonic choices.

Furthermore, there's a generalized fear of a loss of human connectedness if humans form bonds with robots embedded with AI. Concerns arise regarding robots like Harmony by Realbotix potentially increasing social isolation and reducing marriage and birth rates, particularly in countries with declining birthrates like Japan. This highlights significant research opportunities in understanding the societal impacts of human-robot relationships.

Challenges in Adopting and Implementing AI

Organizations are enthusiastic about AI's potential, but realizing that potential comes with significant hurdles. Here's a breakdown of key challenges and some cautionary tales:

Barriers to ROI:

- **Lack of Expertise and Skills:** AI projects require specialists in data science, engineering, and domain knowledge. Many organizations lack these in-house, making it difficult to find the right talent or develop it internally.
- **Data Issues:** AI is data-driven. Poor quality, insufficient, or biased data leads to unreliable AI models. Additionally, integrating AI with existing data infrastructure can be complex.
- **Unrealistic Expectations:** Organizations often overestimate AI's capabilities or underestimate the time and resources needed for successful implementation.
- **Integration Challenges:** Often, AI systems need to connect with legacy IT systems, creating compatibility issues and further complexity.

Lack of Expertise and Skill

- **AI is a complex field:** Developing and managing AI projects requires a unique blend of skills in data science, engineering, and the specific domain the AI is applied to (e.g., finance, healthcare).
- **Talent Shortage:** Demand for skilled AI professionals far outstrips supply. Finding qualified candidates can be difficult and expensive.
- **Internal Development Takes Time:** Building expertise within an organization takes significant investment in training and education.

Consequences of not addressing it:

- **Suboptimal Results:** Without the right expertise, organizations may choose the wrong AI approach, leading to inaccurate models, unreliable performance, and wasted resources.
- **Project Delays and Stalls:** The lack of skilled personnel can slow down project development and implementation, hindering progress and delaying ROI.
- **Increased Risk of Failure:** Without proper guidance, organizations are more likely to encounter unforeseen challenges and ultimately fail to achieve their AI goals.

Approaches to Overcome the Skills Gap:

- **Upskilling Existing Workforce:** Invest in training programs to equip current employees with the necessary AI skills. This can involve online courses, bootcamps, or even graduate degrees.
- **Building Partnerships:** Collaborate with universities or research institutions to access AI talent or leverage their expertise through joint projects.
- **Hiring Strategically:** Focus on attracting and retaining top AI talent by offering competitive compensation and fostering a culture of innovation.
- **Utilizing Low-code/No-code Platforms:** For simpler AI applications, consider leveraging user-friendly platforms that require less technical expertise.

Emerging Trends:

- **Democratization of AI:** The development of more user-friendly tools and platforms is making AI more accessible to organizations with limited in-house expertise.
- **Focus on Explainable AI (XAI):** XAI tools are being developed to help understand how AI models reach their decisions, allowing non-experts to work more effectively with AI systems.

- **Rise of Managed AI Services:** Third-party vendors are offering comprehensive AI solutions, including development, deployment, and management, catering to organizations without the resources to build their own AI teams.

By acknowledging the skill gap and taking proactive steps to address it, organizations can unlock the potential of AI and achieve their business goals

Data Issues

Data Quality and Quantity:

- **Garbage In, Garbage Out:** If the data used to train an AI model is inaccurate, incomplete, or inconsistent, the model will be unreliable and produce misleading results.
- **Data Scarcity:** Many AI applications require vast amounts of data for effective training. Organizations may struggle to collect enough high-quality data, especially for niche applications.

Data Bias and Fairness:

- **Perpetuating Biases:** AI models can inherit and amplify biases present in the data they are trained on. This can lead to discriminatory outcomes, for example, in loan approvals or facial recognition software.
- **Mitigating Bias:** Addressing bias requires careful data selection, cleaning techniques, and potentially adjusting algorithms to ensure fairness.

Data Integration Challenges:

- **Data Silos and Disparity:** Data may be scattered across different departments or systems within an organization, making it difficult to access and integrate for AI projects.
- **Data Compatibility Issues:** Data formats and structures may vary across systems, requiring complex cleaning and transformation before AI can utilize it effectively.

Strategies for Overcoming Data Issues:

- **Data Governance:** Establish clear policies and procedures for data collection, storage, access, and usage. This ensures data quality and consistency.
- **Data Cleaning and Preprocessing:** Invest in tools and processes to identify and remove errors, inconsistencies, and biases from data before training AI models.
- **Data Labeling:** For certain AI applications, human effort may be required to label data

points to provide context and meaning for the model.

- **Synthetic Data Generation:** In cases of data scarcity, explore techniques like generating realistic synthetic data to augment existing datasets.
- **Focus on Explainable AI (XAI):** Utilize XAI tools to understand how AI models arrive at their decisions. This helps identify and address potential biases within the data.
- **Modern Data Infrastructure:** Invest in modern data management solutions that facilitate data access, integration, and governance for AI projects.

By acknowledging these data challenges and implementing appropriate strategies, organizations can ensure their AI models are built on a solid foundation of high-quality, unbiased data, leading to more reliable and trustworthy AI systems.

Unrealistic Expectations:

- **Setting Up for Disappointment:** Overestimating AI's capabilities or underestimating the complexity of implementation leads to frustration and a sense of failure when results don't meet inflated expectations.
- **Wasted Resources:** Organizations may invest heavily in AI projects with unrealistic goals, leading to wasted time, money, and effort.
- **Neglecting Alternatives:** Focusing solely on AI can cause organizations to overlook other potential solutions that might be more suitable for their specific needs.

Causes of Unrealistic Expectations:

- **Media Hype:** Sensationalized portrayals of AI in science fiction and media can create unrealistic expectations about AI's current capabilities.
- **Lack of Understanding:** Organizations may not fully understand the technical complexities and limitations of AI before embarking on a project.
- **Competitive Pressure:** The fear of being left behind in the AI race can lead to hasty decisions and unrealistic goals.

Strategies for Managing Expectations:

- **Focus on Business Needs:** Clearly define the specific business problem you want AI to solve and align expectations with realistic outcomes.
- **Start Small and Scale:** Begin with a well-defined pilot project to gain experience, learn from challenges, and build a foundation for future success.
- **Transparency and Communication:** Communicate openly with

stakeholders about AI's potential and limitations throughout the implementation process.

- **Measure and Track Progress:** Establish clear metrics to track progress and celebrate small wins along the way. This helps maintain focus and realistic expectations.
- **Focus on Long-Term Value:** View AI as a strategic investment with long-term benefits rather than a quick fix.

By acknowledging the limitations of AI and focusing on solving specific business problems, organizations can avoid the pitfalls of unrealistic expectations and achieve sustainable success with AI initiatives.

Integration Challenges

- **Technological Incompatibility:** Legacy systems often use older architectures, programming languages, and data formats that are incompatible with modern AI systems.
- **Data Silos and Accessibility Issues:** Data may be locked away in siloed legacy systems, making it difficult to extract and integrate with AI applications.
- **Security Concerns:** Connecting legacy systems with external AI systems raises security concerns, requiring careful access control and data protection measures.

Strategies for Overcoming Integration Challenges:

- **API Gateways:** Implement API gateways to act as intermediaries between legacy systems and AI applications, translating data formats and protocols for seamless communication.
- **Data Lakes and Warehouses:** Consolidate data from various sources, including legacy systems, into centralized data lakes or warehouses to facilitate easier access for AI systems.
- **ETL/ELT Processes:** Utilize Extract, Transform, Load (ETL) or Extract, Load, Transform (ELT) processes to extract data from legacy systems, transform it into a format usable by AI, and load it into the appropriate location.
- **Modernization Efforts:** Consider a phased modernization approach, gradually upgrading legacy systems to improve compatibility with newer technologies like AI.
- **Focus on Interoperability:** When selecting AI solutions, prioritize those with open architectures and strong interoperability capabilities to ease integration with existing systems.

Additional Considerations:

- **Cost and Time:** Integration projects can be expensive and time-consuming, requiring careful planning and resource allocation.

- **Change Management:** Integrating AI with legacy systems may necessitate changes to existing workflows and processes. Proper change management strategies are crucial to ensure user adoption and minimize disruption.

By acknowledging these integration challenges and adopting appropriate solutions, organizations can bridge the gap between legacy IT and cutting-edge AI, unlocking the potential for greater efficiency and innovation.

Failed AI Projects and Learnings:

- **IBM's Watson Health:** This ambitious project aimed to revolutionize healthcare using AI for diagnostics and treatment. However, it struggled with complex medical data, integration challenges, and a lack of clear focus. The key takeaway: Define a specific, achievable goal and ensure data quality.
- **Microsoft's Tay Chatbot:** This AI chatbot aimed to engage with users on Twitter. However, it quickly learned offensive language from user interactions, leading to its shutdown. The lesson: Build robust safeguards against bias and potential misuse in AI systems

IBM's Watson Health

- **Data Issues:** Medical data is notoriously complex, with diverse formats, privacy regulations, and potential biases. Watson Health likely encountered difficulties in acquiring, cleaning, and integrating high-quality medical data for its models.
- **Integration Challenges:** Hospitals often rely on a patchwork of legacy IT systems. Connecting Watson Health with these disparate systems to access and utilize patient data was likely a significant hurdle.
- **Unrealistic Expectations:** The initial vision for Watson Health might have been overly ambitious, aiming to revolutionize a vast and complex domain like healthcare all at once.

The key takeaways you identified are crucial:

3. **Define Specific Goals:** Don't try to boil the ocean. Instead, focus on tackling a well-defined problem within healthcare with AI, like improving diagnostics for a specific disease or streamlining administrative tasks.
4. **Ensure Data Quality:** High-quality, unbiased data is the bedrock of successful AI projects. Invest in strategies for data acquisition, cleaning, and governance to ensure your AI models have a solid foundation.

By learning from Watson Health's missteps, organizations can approach AI in healthcare (and other domains) with a more measured and realistic perspective. This increases the chances of success and paves the way for AI to deliver its transformative potential in the healthcare industry.

Microsoft's Tay Chatbot

The Tay Fiasco:

Designed to mimic a teenage girl, Tay learned from user interactions. Unfortunately, those interactions often included offensive and inflammatory language. Tay readily absorbed this negativity and began spewing similar content, causing a social media uproar and forcing Microsoft to take it offline.

Lessons Learned:

- **Bias in Training Data:** AI systems can inherit biases present in the data they are trained on. In Tay's case, the training data likely lacked sufficient filters to prevent exposure to offensive language.
- **Vulnerability to Manipulation:** AI systems designed to be interactive can be manipulated by bad actors. Tay's creators likely underestimated the potential for users to exploit the system for malicious purposes.
- **Need for Safeguards:** Robust safeguards are essential to prevent AI misuse. This includes filtering training data, implementing mechanisms to detect and flag inappropriate interactions, and setting clear limitations on how the system can respond.

Building Safeguards against Bias and Misuse:

- **Data Filtering and Curation:** Carefully curate training data to minimize bias and ensure the AI system is exposed to a diverse range of positive and negative examples.
- **Human Oversight:** Implement human oversight mechanisms to monitor AI interactions and intervene if necessary.
- **Continuous Monitoring and Improvement:** Continuously monitor AI system behavior, identify and address potential biases, and refine safeguards as needed.
- **Transparency and Explainability:** Strive for transparency in AI decision-making processes to understand how the system arrives at its outputs and identify potential biases.

By learning from Tay's downfall, organizations can develop AI systems that are more resilient to manipulation and less likely to perpetuate biases. This ensures AI is used responsibly and ethically to achieve positive outcomes.

Additional Considerations:

- **Cost:** Developing and maintaining AI infrastructure can be expensive, especially for smaller businesses.
- **Regulations and Ethics:** Data privacy regulations and ethical considerations around bias in AI algorithms are evolving rapidly, adding complexity to implementation.

- **Organizational Resistance:** Change management is crucial. Employees may fear job displacement or be apprehensive about new technology.

Overcoming these Challenges

By acknowledging these hurdles and learning from past failures, organizations can increase their chances of AI success. Here are some tips:

- **Start small and scale:** Begin with a well-defined pilot project to gain experience and build expertise.
- **Invest in data infrastructure:** Ensure high-quality data collection, storage, and management practices.
- **Build a skilled team:** Develop internal talent or partner with AI experts to address the skills gap.
- **Focus on clear goals:** Don't chase shiny objects. Set realistic goals that align with business needs.
- **Prioritize explainability and fairness:** Ensure AI models are transparent and unbiased.
- **Manage expectations:** Communicate openly with stakeholders about the potential and limitations of AI.

By carefully navigating these challenges, organizations can leverage AI to unlock its true potential and achieve real business value.

Conclusion and discussion:

The work provides a comprehensive overview of AI's significant advancements and transformative impact on business and the global economy. It highlights the commercial availability of AI-driven products, emphasizing the role of big data and powerful processing units like GPUs and TPUs in driving AI's growth. Four main areas of deep learning are explored: computer vision, text analysis, speech recognition, and game playing, along with preferred algorithms and successful applications in each area. The analysis also covers prominent AI startups, indicating a growing appetite for AI investment and its potential to shape various industries such as business intelligence, healthcare, cybersecurity, and marketing.

However, challenges such as the "AI divide," where AI technology is concentrated in certain regions, pose risks of exacerbating social, economic, and cultural inequalities. Additionally, the study addresses concerns regarding AI's reliance on software, which is susceptible to vulnerabilities, and highlights challenges related to trust, ethics, bias, and the shortage of AI talent. Despite AI's potential benefits in increasing productivity, efficiency, and decision-making speed, addressing these challenges is crucial for its widespread commercial application.

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References

- [1] Amazon Go. (2020). Retrieved from: <https://www.amazon.com/b?node=16008589011>.
- [2] Bennett, J. (2017). Boeing Announces Study for Self-Flying Airliners. Retrieved from: <https://www.popularmechanics.com/flight/airlines/a26838/boeing-self-flying-airliners/>.
- [3] Boyle, A. (2021). Amazon details how its warehouse robots are designed to help humans work safely. GeekWire. Retrieved from: <https://www.geekwire.com/2021/amazon-details-warehouse-robots-designed-help-humans-work-safely/>.
- [4] Clark, G., and Jacks, D. (2006). Coal and the Industrial Revolution, 1700-1869. Retrieved from: https://gpih.ucdavis.edu/files/Clark_Jac.ks.pdf.
- [5] Data U.S.A. (2018). Cashiers. Retrieved from: <https://datausa.io/profile/soc/cashiers>.
- [6] Gallo, C. (2019). An AI Expert told 60 Minutes that AI Could Replace 40 Percent of Jobs. Here's the Part He Left Out. Inc. Retrieved from: <https://www.inc.com/carmine-gallo/an-ai-expert-told-60-minutes-that-ai-could-replace-40-percent-of-jobs-heres-part-he-left-out.html>.
- [7] Hicks, M., and Fitzsimmons, M. (2019). Self-driving cars: your complete guide to autonomous vehicles. TechRadar. Retrieved from: <https://www.techradar.com/news/self-driving-cars>.
- [8] International Transport Forum. (2017). Managing the Transition to Driverless Road Freight Transport. Retrieved from: <https://www.itfoecd.org/sites/default/files/docs/managing-transition-driverless-road-freight-transport.pdf>.
- [9] Jovanovic, B., (2022). 55 Fascinating AI Statistics and Trends for 2022. Retrieved from <https://dataprot.net/statistics/ai-statistics/>
- [10] Kay, G., (2021). Elon Musk unveils 'Tesla Bot,' a humanoid robot that would be made from Tesla's self-driving AI. Business Insider. Retrieved from: <https://www.businessinsider.com/elon-musk-unveils-tesla-bot-humanoid-robot-based-off-autopilot-2021-8?op=1>.
- [11] Lipka, M., (2016). U.S. religious groups and their political leanings. Pew Research Center. Retrieved from: <https://www.pewresearch.org/fact-tank/2016/02/23/u-s-religious-groups-and-their-political-leanings/>.
- [12] McCloskey, D. (2008). The Industrial Revolution. Retrieved from: <http://www.deirdremccloskey.com/articles/revolution.php>
- [13] McKinsey Global Survey (December 8, 2021). The State of AI in 2021. Retrieved from: <https://www.mckinsey.com/business-functions/quantumblack/our-insights/global-survey-the-state-of-ai-in-2021>
- [14] Meyersohn, N. (2021) Amazon, the crusher of department stores, may become one. TechWire. Retrieved from: <https://www.wraltechwire.com/2021/08/20/amazon-the-crusher-of-department-stores-may-become-one/>.
- [15] MGM Research. (2019). World GDP per capita Ranking. Retrieved from: <https://mgmresearch.com/world-gdp-per-capita-ranking/>.
- [16] Obschonka, M., (2018). Research: The Industrial Revolution Left Psychological Scars That Can Still Be Seen Today. Harvard Business Review. Retrieved from: <https://hbr.org/2018/03/research-the-industrial-revolution-left-psychological-scars-that-can-still-be-seen-today>.
- [17] PC Magazine (n.d.). What are your top concerns associated with autonomous vehicles? Retrieved from: <https://assets.pcmag.com/media/images/653509-why-axis-chart-pcmag->
- [18] PwC (2022). 2022 AI Business Survey. Retrieved from https://www.pwc.com/us/en/tech-effect/ai-analytics/ai-business-survey.html?WT.mc_id=CT3-PL300-DM1-TR1-LS4-ND30-PRG7-CN_DataAndAnalyticsBuilds

AI Survey Google & gclid=Cj0KCQjwio6XBhCMA
RIIsAC0u9aH3X123ib_ICl_g0H2jeflm8KWpDEZ0
08ZMAAt9Bv2GFilP1cnzt5OQaAk
IEALw_wcB & gclsrc=aw.ds

[19] Schwab, K. (2016). The Fourth Industrial Revolution. Retrieved from:
File:///C:/Users/user/Downloads/The_Fourth_Industrial_Revolution_Klaus_S.pdf.

[20] ScienceDaily (2022). AI Helped Protect Businesses from COVID-19 Risks. Retrieved from:
<https://www.sciencedaily.com/releases/2022/03/220304100958.htm>

[21] Todorov, G., (2021). 65 AI Statistics for 2021 and Beyond. Retrieved from
<https://www.semrush.com/blog/artificial-intelligence-stats/>

[22] U.S. Department of State Archive (n.d.). The United States and the Opening to Japan, 1853. Retrieved from: <https://20012009.state.gov/r/pa/ho/time/dwc/86550.html>.

[23] Wang, L. (30 September 2021). Artificial Intelligence for COVID-19: A Systematic Review. Frontiers in Medicine. Retrieved from:
<https://www.frontiersin.org/articles/10.3389/fmed.2021.704256/full>

[24] White, M. (2009). The Industrial Revolution. Retrieved from: <https://www.bl.uk/georgian-britain/articles/the-industrial-revolution#>.

[25] Yoon, L. (2021). GDP comparison between South and North Korea 2010-2019. Statista. Retrieved from:
<https://www.statista.com/statistics/1035390/south-korea-gdp-comparison-with-north-korea/>.

[26] Young, M., (2017). Meet Valkyrie, NASA's Space Robot. Sky & Telescope. Retrieved from:
<https://skyandtelescope.org/astronomy-news/meet-valkyrie-nasa-space-robot/>.

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