Industrial Use of Generative AI: Opportunities and Risks

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# CONTENTS

1 Overview ................................................................................................................. 3  
2 introduction ............................................................................................................. 4  
3 On Opportunities .................................................................................................... 5  
4 On Risks and Responsibilities .................................................................................. 6  
5 Closing ...................................................................................................................... 7  
6 References ................................................................................................................ 8  
7 Acknowledgements .................................................................................................... 11
Industrial Use of Generative AI: Opportunities and Risks

1 OVERVIEW

Since the launch of ChatGPT in late 2022, Generative AI has emerged as a focal point of technological innovation, drawing broad and intense interest from businesses, governments, and the public. Investment banking firm Goldman Sachs estimated that Generative AI could potentially boost the annual global GDP by 7%\(^1\). Similarly, McKinsey's latest research suggests that Generative AI might add up to $4.4 trillion to the global economy annually\(^2\). An estimate by Hootsuite indicates that over 110 million Americans could be directly interacting with generative AI by 2025\(^3\). These projections provide a glimpse into its transformative power and prospective impact on global economies.

At its core, Generative AI offers more than analytics or task automation; it can produce novel content, solutions, and insights, making it a powerful tool in a myriad of domains. From the creation of art, music, and literature, to the design of complex pieces of software, equipment and even therapeutic proteins, Generative AI is reshaping the boundaries of what is possible through machine-driven creativity.

The economic and societal implications of this technology are profound. Businesses stand to benefit from unprecedented efficiencies, cost savings, and the ability to tackle previously insurmountable challenges -- from personalized product designs to the rapid prototyping of new medicines. The comparisons by MIT researchers, who liken the potential of Generative AI to the monumental shifts witnessed during the Industrial Revolution, offer perspective on how significant this AI wave is for the workforce\(^4\). An overwhelming 98% of global executives surveyed by Accenture affirm their belief in the crucial role of generative AI in shaping their organizations' future strategies\(^5\). Gartner's independent findings further validate this sentiment, with 45% of executives crediting ChatGPT as a catalyst for heightened AI investment\(^6\).


\(^3\)https://blog.hootsuite.com/artificial-intelligence-statistics/


2 INTRODUCTION

Generative AI is a subset of artificial intelligence that focuses on creating new content or data that is coherent and contextually relevant\textsuperscript{7}. Unlike traditional AI systems which make decisions based on existing data, generative AI systems can produce entirely new data, from text and images to 3D models and other forms of content. These AI models learn from vast amounts of data to recognize patterns, structures, and nuances, which enables them to generate outputs that often mirror or resemble the original data. For instance, a generative AI model trained on musical compositions could craft a new piece of music that sounds like it belongs to a particular genre or artist's style.

Generative AI encompasses a range of algorithms and models. Among the most talked-about these days are \textit{Large Language Models} (LLMs)\textsuperscript{8} like the one ChatGPT uses to generate human-like text. LLMs are built on the \textit{Transformer} architecture\textsuperscript{9}, which are designed to generate coherent and contextually relevant text by capturing intricate linguistic structures. Examples of LLMs include GPT-4, BART/PaLM and LLaMa\textsuperscript{10}. Other Generative AI models include \textit{Generative Adversarial Networks} (GANs)\textsuperscript{11} that are designed around the concept of two neural networks—the generator and discriminator—competing against each other. They excel at generating sharp, realistic images.

\textit{Variational Autoencoders} (VAEs)\textsuperscript{12}, another powerful class of generative models, are more common in image and audio generation, offering a probabilistic framework to model the inherent uncertainty and generate diverse outputs. \textit{Diffusion} models\textsuperscript{13} have emerged as a promising approach to generative modeling especially in the context of image synthesis and restoration tasks, wherein data is viewed as the endpoint of a diffusion process. Most recently, \textit{Neural Radiance Fields} (NeRFs)\textsuperscript{14} were introduced as a technique for generating 3D models from 2D images. There are other AI models that are not mentioned here due to brevity and the fact that the AI field is constantly evolving with new models. Each has its unique strengths and optimal use scenarios, but all contribute to the broader landscape that is called \textit{Foundation Models} (FMs)\textsuperscript{15} that power Generative AI.

\textsuperscript{7}https://en.wikipedia.org/wiki/Generative_artificial_intelligence [Cited: October 7, 2023]
\textsuperscript{8}https://en.wikipedia.org/wiki/Large_language_model [Cited: October 15, 2023]
\textsuperscript{9}https://towardsdatascience.com/transformers-141e32e69591
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Prior to Generative AI, the AI landscape was dominated by customized solutions where machine learning experts would build distinct models for specific tasks. These efforts, though successful in fields like computer vision and predictive maintenance, often led to narrow applications. Each AI implementation was not only constrained by its specificity but was also costly and time-consuming, limiting its widespread adoption.

FMGs marked a significant shift in the world of AI. Instead of training specialized models for each task, FMGs are trained on vast amounts of data spanning many different domains and sources, resulting in a versatile and expansive model. While the initial investment in building FMGs is high, they serve as a foundational platform upon which a plethora of AI applications can be built. This minimizes -- and in some cases, eliminates -- the need for task-specific training. They have revolutionized the AI landscape, making it more cost-effective, faster and more sophisticated AI deployment across industries.

3 ON OPPORTUNITIES

The potential industry applications of generative AI are vast and multifaceted. From the inception of a product to its eventual disposal, this technology is redefining traditional processes. Its capabilities extend from design and engineering to sourcing and supply, manufacturing, servicing, and more.

In product requirement phase for instance, Generative AI assists in streamlining requirement solicitation process and facilitates a systematic approach towards defining product specifications. It can generate comprehensive questionnaires tailored for requirements gathering, ensuring that all aspects of a product's needs are addressed. By customizing questions for specific organizational roles, it ensures both thoroughness and relevance, enhancing and accelerating the preliminary stages of product development.

During the product design phase, Generative AI’s ability to churn out numerous design concepts and variations in a short span of time is transformative. Designers input parameters and requirements into a model, and in return, they receive a multitude of design variations. Not only does this expedite the brainstorming phase, but it also allows for the exploration of innovative designs that might not have been considered otherwise. Additionally, it can generate scenarios for validating design options, thus seamlessly bridging the gap between conceptualization and user research.

Generative AI also transforms design validation process. Traditional methods often rely on a set of pre-defined test cases, which might miss out on edge scenarios. Generative AI, with its capacity...
to generate diverse test cases and simulations, ensures a comprehensive validation process. This results in products that are not only functional but also resilient against unforeseen issues.

In production, Generative AI’s role in factory design, optimization, and simulation paves the way for industries to achieve greater predictability, agility, safety, and efficiency. By simulating factory operations based on historical data or new synthesized scenarios, organizations can preemptively address potential challenges. For instance, understanding regulatory constraints through AI-driven simulations ensures that facilities adhere to norms and minimize negative environmental impacts. It can help devise strategies that are both efficient for production and sustainable for the environment. Whether it is optimizing resource usage, reducing waste, or devising cleaner production methods, generative AI can play a significant role in driving sustainable industrial practices.

Quality assurance in manufacturing traditionally involves manual checks and post-production testing. Generative AI, however, can simulate the manufacturing process and predict where quality issues might arise. This proactive approach ensures that quality is embedded at every stage of the production process, leading to better products and reduced wastage.

Supply chain management also can gain immensely from Generative AI. Predicting supply chain disruptions, optimizing inventory levels, and automating logistics are some areas where Generative AI can make insightful recommendations. By analyzing historical data, current trends and news, these AI models for example can forecast potential challenges due to geopolitical situations or natural disasters and offer solutions before problems arise.\(^\text{18}\)

Generative AI also benefits human workers by improving how they collaborate and enhance their skills. Generative AI for example, can dynamically generate and update instructional materials that are customized specifically for individual workers. It can also produce customized practice simulation scenarios, catering to individual learning needs, ensuring that skills are honed effectively.

4 On Risks and Responsibilities

As with any powerful tool, Generative AI poses both opportunities and challenges. Without proper guidelines and ethical considerations, it can be misused, leading to misinformation, privacy breaches, and other unforeseen consequences. While the broader narrative around

responsible AI usage is well-documented\textsuperscript{19,20,21}, the nuances associated with generative AI, particularly large language models, demand specific consideration.

One of the most discussed issues surrounding generative AI is the accuracy of the generated content. Generative AI, despite its impressive capabilities, can produce misleading or wholly fabricated information, often termed \textit{hallucination}\textsuperscript{22}. Relying on such output without verification can lead to misinformation, incorrect decisions, or even physical harm. It is essential for users to approach the outputs of generative AI with a critical mindset, assessing the content for its relevance, accuracy, and appropriateness before acting upon or disseminating it. This cautious approach can help avoid the pitfalls of acting on or spreading false information.

Another concern is the risk of biased responses\textsuperscript{23}. Many large language models, trained on vast and diverse external data set, may contain biases. These biases can inadvertently perpetuate or even amplify stereotypes or misrepresent groups, which can have detrimental effects when deployed in real-world scenarios. Companies using generative AI should establish robust controls and protocols to identify and rectify these biased outputs. Such measures will ensure that employees address these situations in line with organizational ethics and guidelines, enabling AI to fulfill its role without inducing harm or misrepresentation.

Furthermore, privacy and intellectual property remain critical considerations. The vast amount of data used to train these models can sometimes lead them to inadvertently reproduce private or sensitive information. Such occurrences could breach individual privacy, and organizations need to be vigilant about potential leaks of confidential data. Additionally, the use of copyrighted or proprietary data in training these models can give rise to intellectual property issues. The generated content could infringe on copyrights or blur the lines of content ownership. This necessitates a comprehensive understanding of the data sources used in training and the legal implications of the generated content\textsuperscript{24}.

5 CLOSING

Generative AI is rapidly reshaping the industrial landscape. It offers unprecedented opportunities for automation, efficiency, and innovation, propelling industries into a new era of production and design. From ideation and design to production and after-service, generative AI tools are creating

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\textsuperscript{24}https://mitsloan.mit.edu/ideas-made-to-matter/legal-issues-presented-generative-ai
Industrial Use of Generative AI: Opportunities and Risks

pathways that allow for more rapid iterations, informed decision-making, and a heightened level of precision. This transformation is not just about improved processes; it is about unlocking new capabilities that were previously deemed unreachable.

It is crucial to recognize that we are still in the early stages of fully understanding the power and potential of Generative AI. Research in Generative AI is incredibly vibrant and dynamic. Scholars and industry experts are continuously exploring the emergent properties of generative AI models, aiming to discover new opportunities and ways to harness its vast potential. Simultaneously, efforts are underway to address the inherent limitations of current models, with some researchers looking at the possibility of harnessing multiple AI models in tandem to overcome individual shortcomings.

As industries delve deeper into this transformative technology, it becomes imperative to approach its adoption with responsibility and discernment. The potential pitfalls—ranging from biased outputs and inaccuracies to concerns about privacy and intellectual property—underscore the need for a vigilant, ethical, and regulated approach. Just as generative AI has the power to accelerate industrial growth, it also carries the weight of ensuring that this growth is sustainable, ethical, and beneficial for all stakeholders. The future of industries with generative AI is not just about leveraging its potential, but about crafting a balanced narrative where innovation and responsibility coalesce seamlessly.

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Industrial Use of Generative AI: Opportunities and Risks


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Industrial Use of Generative AI: Opportunities and Risks

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