

Generative AI (Gen AI)

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	low	medium	high
Impact			
Complexity			

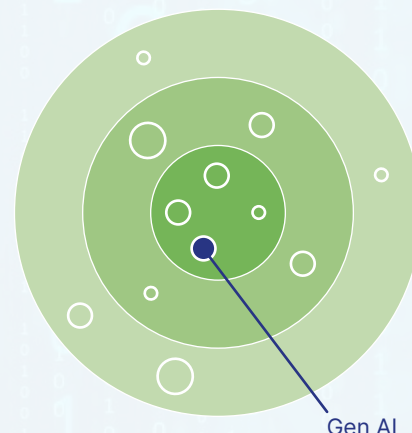
1 INTRODUCTION

Purpose

Generative AI is a subfield of artificial intelligence focused on creating models that can generate new data, such as images, text, music, or even code, that is similar to a given dataset. This involves the use of [neural networks](#), particularly [deep learning models](#) like [Generative Adversarial Networks \(GANs\)](#) and [transformers](#).

Key benefits

Gen AI offers powerful capabilities such as the ability to create realistic synthetic data, enhance creativity through automation, generate personalised content at scale, and improve customer interactions through AI-driven chatbots and digital assistants, leading to innovations in product design, content creation, workflow optimisation, customer service and beyond.



2 KEY CONCEPTS

Fundamental principles

The fundamental principles of Gen AI include unsupervised and semi-supervised learning for training models without labelled data, generative modelling to predict data distributions, and the use of neural network architectures like [GANs](#) and [Variational Autoencoders \(VAEs\)](#) to create new content. In the context of chatbots, transformers are particularly important as they can generate coherent and contextually relevant responses.

Terminology

In Gen AI, [latent space](#) refers to the compressed representation of data in neural networks, [GANs](#) involve two [neural networks](#) (a generator and a discriminator) that compete to improve the generation process, and [transformers](#) are models particularly effective in generating sequential data. [Natural Language Processing \(NLP\)](#) is another critical area, focusing on the interaction between computers and human language.

The Three Components of Generative AI

- Model Training:** leveraging large datasets to train models that can learn underlying patterns.
- Data Generation:** producing new data that closely resembles the original dataset.
- Adaptation:** continuously refining models to improve the quality and relevance of generated content.

3 POPULAR TOOLS AND FRAMEWORKS

Primary tools

- [TensorFlow/Keras](#): widely used for building and training generative models like GANs and VAEs.
- [PyTorch](#): a flexible, widely adopted deep learning framework, favoured for research in generative models.
- [Hugging Face Transformers](#): a library for working with transformer models, ideal for text generation and chatbots.

Comparison

- [GANs vs VAEs](#): GANs are often used for generating high-quality images, while VAEs are more stable and easier to train but may produce blurrier outputs.
- [Transformers \(e.g. GPT-4\) vs Recurrent Neural Networks \(RNNs\)](#): transformers outperform RNNs in text generation tasks due to their ability to capture long-range dependencies more effectively.

4 APPLICATIONS

Industry use cases

- **Creative arts:** AI-generated artwork, music and literature.
> **Best Practice:** [OpenArt](#) and [DALL-E](#).
- **Healthcare:** AI-generated synthetic medical data for research and training.
> **Best Practice:** [GANs for Medical Imaging](#).
- **Marketing:** personalised content generation for ads and customer engagement.
> **Best Practice:** [Copy.ai](#).
- **Customer service:** automated customer support and engaging users in natural conversation.
> **Best Practice:** [Replika](#) for personal AI companions, [Google Assistant](#) for virtual assistants.

Practical examples

- [GPT-4](#): a state-of-the-art language model used for generating human-like text across various domains, including chatbots and digital assistants.
- [DeepAI](#): a GAN variant capable of producing highly realistic images, widely used in entertainment.
- [MuseNet](#): a deep neural network that generates complex musical compositions.

5 IMPLEMENTATION INSIGHTS

Best practices and tips

- **Model Pre-training:** utilise pre-trained models to save time and resources, fine-tuning them for specific applications like customer service chatbots.
- **Data Augmentation:** increase dataset diversity through techniques like data augmentation to improve model robustness.
- **Bias Mitigation:** actively work to identify and mitigate biases in generated content.

Common challenges

- **Model Overfitting:** avoid overfitting by using regularisation techniques and ensuring a diverse training dataset.
- **Ethical Considerations:** address potential misuse, particularly in chatbots where issues like bias or misinformation can arise.
- **Reputational and Legal Risks:** from e.g., wrong answers, offensive language, digital trust issues, and discrimination.
- **Quality Control:** ensure the quality and reliability of generated content.

6 KEY TRENDS AND PREDICTIONS

Top milestones in Generative AI

- **2014:** introduction of [GANs by Ian Goodfellow](#), revolutionising generative modelling.
- **2018:** GPT-2 Release by OpenAI, showcasing the potential of transformer models in text generation, paving the way for more advanced chatbots.
- **2021:** DALL-E Release by OpenAI, demonstrating powerful AI image generation from textual descriptions.

Current trends

- **AI for Creative Collaboration:** increasing use of AI as a tool for artists and designers to enhance creativity.
- **Realistic Content Creation:** advancements in GANs and transformers are leading to more realistic and convincing AI-generated content, including conversational agents.
- **Cross-Domain Applications:** expansion of generative AI applications across various industries.

Future predictions

- **Generative AI in Virtual Worlds:** expanding use in creating realistic virtual environments for gaming, simulation and metaverse applications, with chatbots as guides or NPCs (non-playable characters).
- **AI-Generated Code:** AI tools that can generate and optimise software code, aiding developers and reducing time to market. These tools are increasingly integrated into programming chatbots.
- **AI Ethics and Regulation:** increased focus on ethical considerations and regulatory frameworks for AI-generated content.

→ See our separate fact sheet on **Artificial Intelligence (AI)**.

7 KEY RESOURCES AND MOST HELPFUL LINKS

Websites and blogs

- [OpenAI Blog](#): insights and updates on the latest advancements in generative AI.
- [Distill.pub](#): research articles explaining complex AI topics, including generative models.
- [AI Products Catalogue](#): a collection of over 5,000 AI tools.

Online courses

- [Generative Adversarial Networks \(GANs\) Specialization](#): by Andrew Ng.
- [Building Generative Adversarial Networks \(GANs\)](#): by DataCamp.
- [Generative AI with TensorFlow](#): by Pluralsight.

Communities and forums

- [Kaggle](#): a platform to learn and practise generative AI with datasets and competitions.
- [MongoDB](#): for discussions on the latest in AI research and applications, incl. chatbots.
- [The Secrets of Agents](#): explore different generative AI projects and guides.

8 GLOSSARY

Common terms and definitions

- **GAN** (Generative Adversarial Network): a model that generates new data by pitting two networks against each other.
- **Transformer**: a model architecture designed to handle sequential data, excelling in text generation, essential for chatbots.
- **Latent Space**: a representation of compressed data within neural networks used for generating new data.
- **Overfitting**: a model's excessive adaptation to training data, leading to poor generalisation to new data.
- **Fine-tuning**: the process of adapting a pre-trained model to a specific task or dataset, often used in chatbot development.

Authors



Dr Torsten Wingenter

Torsten established Digital Innovations at Lufthansa, founded the FlyingLab, and was responsible for the digital strategies of Austrian, Lufthansa and Swiss airlines. Today, as the "Inno Doc", he is digital advisor, coach and catalyst, interim manager and fire fighter for many organisations in their pursuit for digital innovations.

www.inno-doc.com



Prof Marc K Peter

Marc was an executive at eBay, E*TRADE (ANZ) and LexisNexis. Today, he is the "Digital Prof" at Rochester-Bern Executive Programs, the University of Rochester, at FHNW and at CSU in Australia. His research and teaching covers digital transformation, digital technology, digital leadership, cybersecurity and digital marketing.

www.digitalprof.com