# Generative Al for Synthetic Data

Generative AI is currently in the spotlight with the release of ChatGPT, but it has already been making significant contributions to data and analytics (D&A) through synthetic data. This solution can help fill gaps in real-world data sources and even improve model outcomes. How are data and analytics professionals currently using synthetic data and what challenges do they face?

One-Minute Insights:

	Organizations adopt AI-generated synthetic data because of challenges with real-world data accessibility, complexity and availability
	Partially synthetic data is the most common approach and text-based is the most-used type of synthetic data
~~	Leaders have seen improvements in model accuracy and efficiency as a result of synthetic data
୶ୡ	Most challenges with synthetic data are inherited from limited, poor quality or biased real-world source data
	To ensure synthetic data quality, most leaders have implemented best practices like using multiple data sources and synthetic dataset validation

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Data collection: Apr 1 - 14, 2023

Respondents: 150 IT and D&A leaders who work with or oversee groups that work with AI-generated synthetic data at their organization

### Challenges with real-world data accessibility, complexity and availability have led organizations to adopt Al-generated synthetic data

Most IT and D&A leaders surveyed say their organization adopted AI-generated synthetic data because of **challenges with real-world data accessibility (60%)**, **complexity (57%)**, or **availability (51%)**.

3% of respondents say their organization did not face any challenges with real-world data.

Which challenges has your organization faced with real-world data that led to adopting AI-generated synthetic data? Select all that apply.



Unbalanced data **31%** | We haven't faced any challenges with real-world data **3%** | Other **0%** 

— 80% n = 150

### Thoughts on using and creating AI-generated synthetic data





## Fully synthetic data is less likely to be used than partially synthetic data; text-based is the most common type

Most respondents say their organization uses **partially synthetic data (63%)** or a combination of partially and fully synthetic data **(20%)**.



50%

of respondents say their organization **generates synthetic data** through a custom-built solution with **open-source tools**, while **31%** turn to **vendor solutions** to generate their synthetic data.

#### How is synthetic data generated at your organization?



### Concerns and challenges with AI-generated synthetic data



"It is in [an] early stage and will be tough to adopt across [the] entire organization and also ROI cannot be [easily] calculated. Regulatory issues are a major concern."

- C-suite, finance industry, 10,000+ employees



"AI generated [techniques have] a high level of myopic bias, selecting the right vendor for data remains a challenge."

- Manager, finance industry, 1,000 - 5,000 employees

Question: Do you have any final thoughts to share on AI-generated synthetic data?

## Synthetic data can improve model accuracy and efficiency, but many have faced challenges with lack of or low quality real-world source data



The most often realized benefits of synthetic data at respondents' organizations are **improved model accuracy (60%)**, **improved model efficiency (56%)** and **mitigated data privacy concerns (45%)**.

How has synthetic data benefited your organization? Select all that apply.

Increased efficiency of data teams **25%** | Rebalanced datasets **23%** | Reduced data breach risks **19%** | Reduced overfitting **14%** | None of these **3%** | Other **0%** 

n = 150

About half (51%) of respondents have dealt with a lack of real-world source data for the synthetic data at their organization. More than one-third have experienced challenges with inherited bias in synthetic data (46%), low quality real-world source data (41%) or inaccuracy caused by statistical noise (34%).

Only 2% of respondents have **not experienced any challenges** with synthetic data at their organization.





— 80% n = 150

Lack of expertise 25% | Insufficient resources 24% | Accuracy degradation 23% | Integration with existing data systems 14% | Cost of computing power 12% | Selecting the right vendor 11% | Determining appropriate utility metrics 11% | Legal/ethical concerns (e.g., re-identification risk) 10% | We haven't experienced any challenges with synthetic data 2% | Other 0%

### Most have implemented best practices to ensure quality of their synthetic data



of respondents use **multiple data sources for generative models** to ensure their synthetic data **quality is high**. Synthetic dataset validation **(59%)** and data quality checks before use in generative models **(50%)** are also common best practices among respondents.

### What best practices have you implemented to ensure the quality of your synthetic data is high? Select all that apply.





n=150

### Risks and considerations for AI-generated synthetic data



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### **Respondent Breakdown**



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