

ISTANBUL SOFTWARE TESTING CONFERENCE

AI-Generated Data: *Guardians of Privacy or a Trojan Horse?*

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#ISTC2026

Could you guess who is behind these data?

User ID	Movie Title	Rating	Date Rated
10294	The Notebook	5	2005-12-17
10294	The Grand Budapest Hotel	4	2006-01-03
92811	The Matrix	4	2006-02-14
10294	Garden State	3	2006-03-01
92811	The Godfather	5	2006-03-22
19875	Whiplash	5	2006-04-02
92811	Eternal Sunshine of the Spotless Mind	5	2006-04-10
47502	Little Miss Sunshine	4	2006-04-29
19875	The Social Network	4	2006-05-03
47502	Moonrise Kingdom	3	2006-05-11



The Netflix Case

How just movie ratings revealed real people...

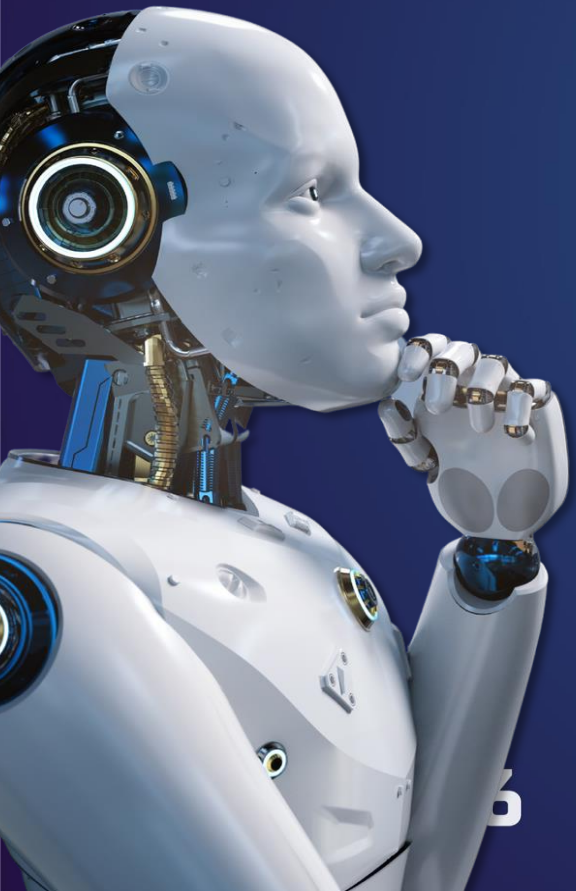
Netflix

User ID	Movie	Rating	Date
A102	The Notebook	★★★★★	01/08/2005
A102	Madagascar	★★★★★	14/09/2005
B351	Interstellar	★★★★★	01/03/2006
C478	Toy Story	★★★★★	10/01/2006
A102	The Holiday	★★★★★	03/10/2005
D501	Titanic	★★★	22/11/2007
E214	Harry Potter	★★★★★	18/04/2008

IMDB

Username	Movie	Rating	Date
JaneDoeNY	The Notebook	5	01/08/2005
FilmFan88	Ice Age	4	01/03/2006
JaneDoeNY	Madagascar	4	14/09/2005
MovieFan21	Toy Story	5	10/01/2006
CinemaFan2	Titanic	3	22/11/2007
JaneDoeNY	The Holiday	5	04/10/2005
MaxReviewe r	Pets	4	18/04/2008

*Data Protection Matters:
What, Why and How?*



How easily we can be identified...

80%

of “anonymous” Netflix users were re-identified using only a few IMDb ratings.



90%

of individuals can be uniquely identified from just 4 credit-card transactions (date + location).

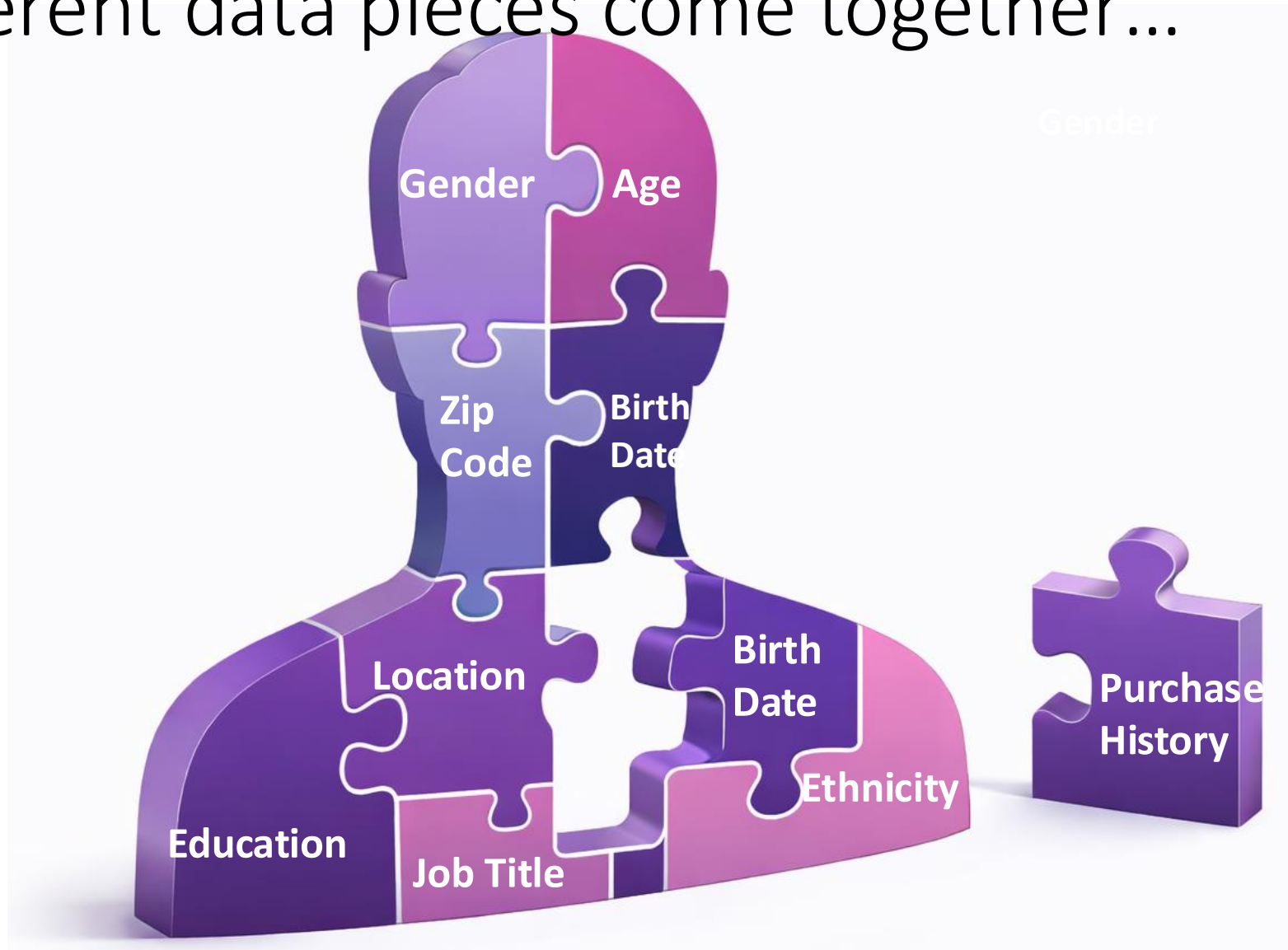


99.98%

99.98% of Americans can be re-identified using only 15 demographic attributes.



When different data pieces come together...



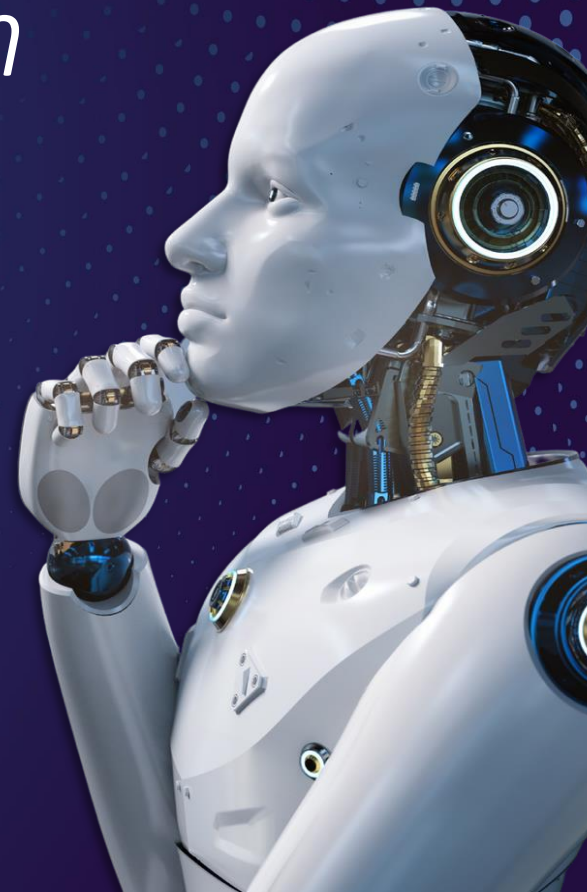
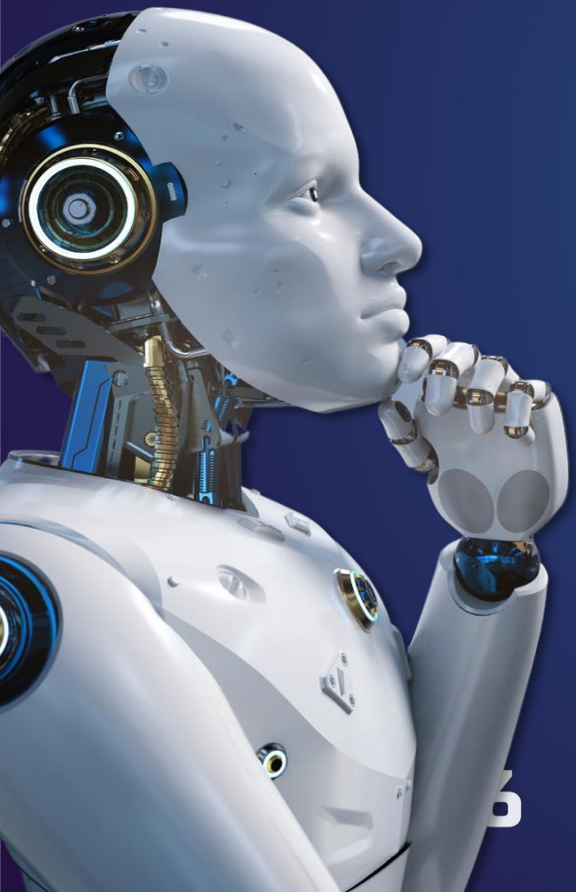
Why Data Protection Matters

Trust takes years to build....

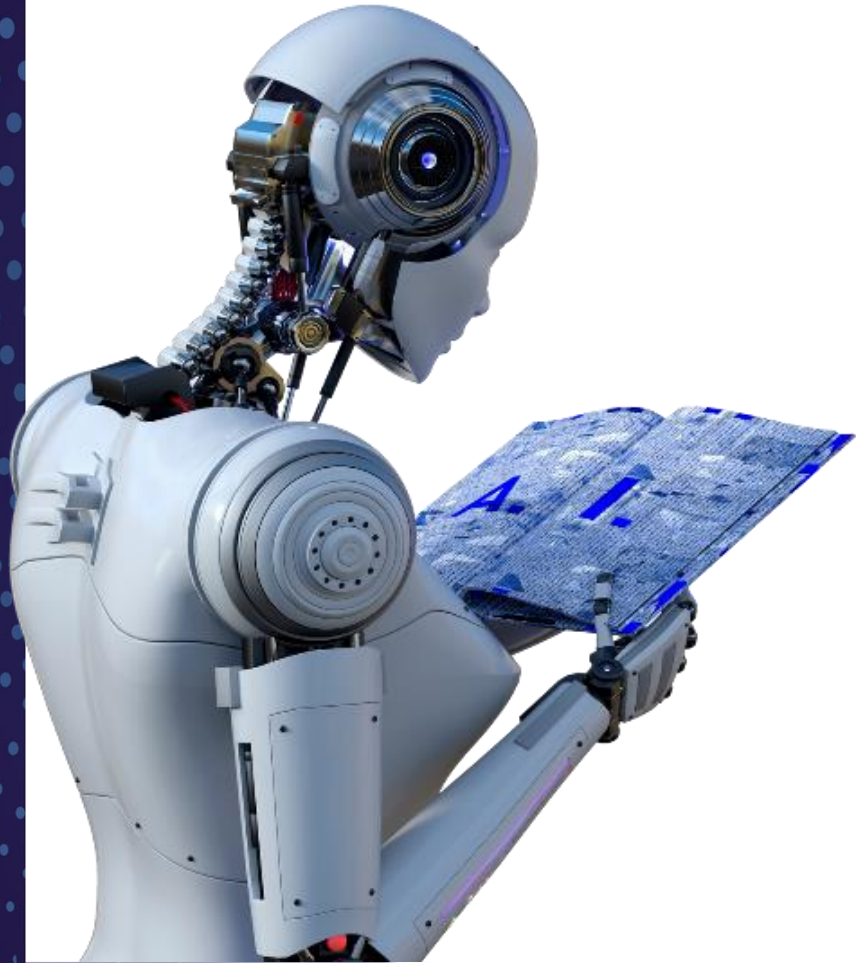


... and seconds to break!

*Synthetic Data Generation
with AI*



How do AI Models generate data?



Language-based Generation

- Predict next token based on context
- Generates records step-by-step
- Combine linguistic and statistical patterns
- **Examples:** LLMs

Data-Distribution Generation

- Learn the underlying distribution of data
- Generate new samples from latent space
- Generate entire dataset once (not step by step)
- **Examples:** Generative Adversarial Networks (GANs), Variational Autoencoders (VAEs), Diffusion Models

Generative Adversarial Networks (GANs)

Learning to Generate Data by Playing a Game!

.Introduced by Ian Goodfellow in 2014

Two networks:

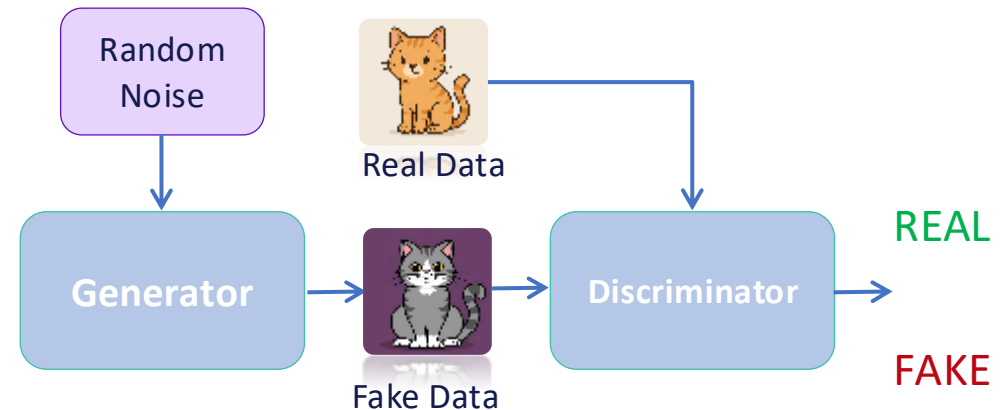
- Generator (creates data)
- Discriminator (detects real vs fake)

Adversarial training: They compete in a minimax game, and both get better

Generator tries to fool the Discriminator with increasingly realistic data

Applications: Healthcare, Finance, Image Synthesis, etc.

GAN Architecture



Large Language Models (LLMs)

Learning to Generate Data by Predicting What Comes Next!

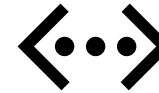
- Introduced in 2017
- Based on Transformer Neural Networks
- Train through self-supervised learning
- Learn statistical patterns from massive text datasets
- Can follow constraints to create structured outputs

Applications: Synthetic tabular data, Images, Code Generation, etc.

AI Model

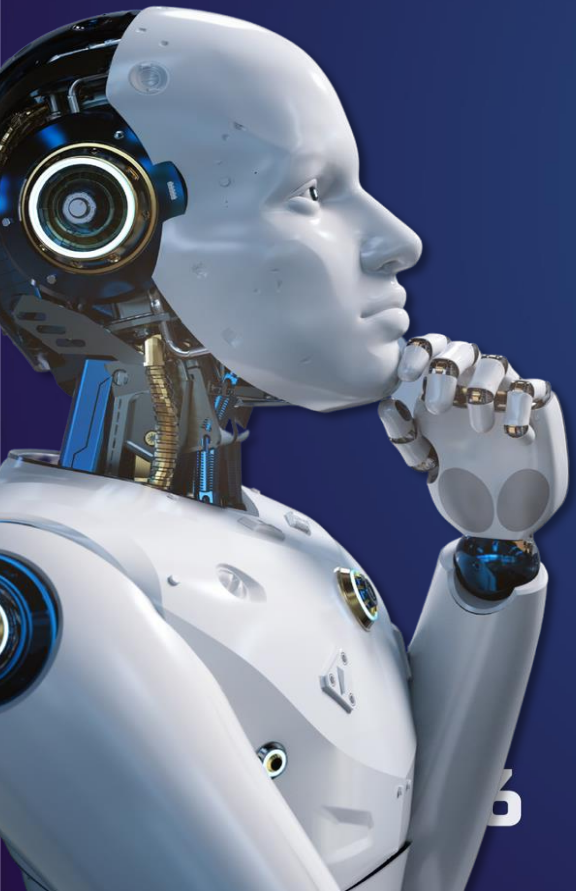


“Generate 10 synthetic user profiles with columns: name, age, city, job.”



Name	Age	City	Job
Sarah	32	Berlin	Designer
Ben	52	Frankfurt	Developer
...

What are the risks and challenges of AI Models?



Memorization

When AI Models Memorize instead of Learning!

Learning



Student A

*Learning concepts
Applying knowledge*

Memorizing



Student B

*Memorizing answers
Repeating exact phrases*

AI Models should behave like **Student A**, but sometimes they behave like **Student B**

Memorization

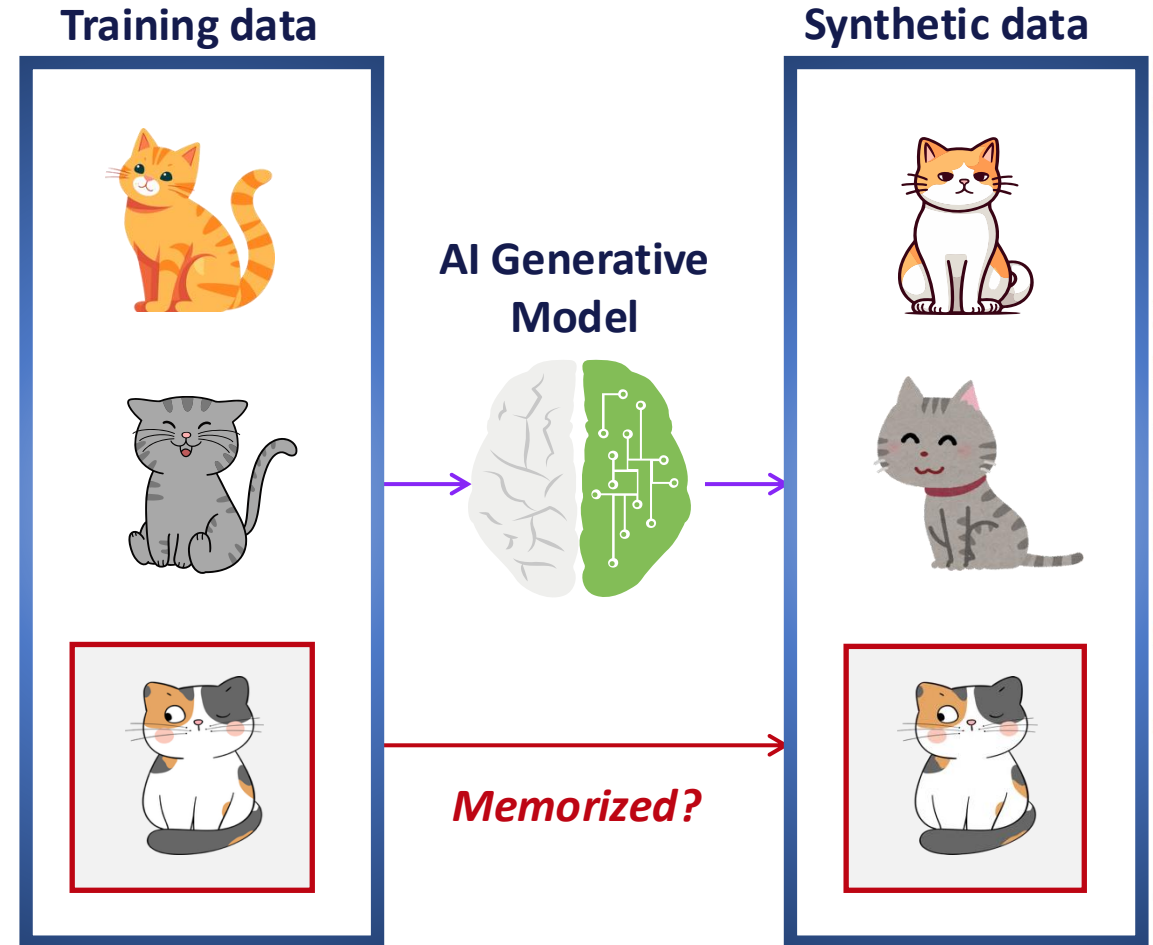
When AI Models Memorize instead of Learning!

Why it happens:

- Overfitting
- Lack of data diversity or small datasets
- Training too much or overly large models
- No regularization or privacy

Where it happens:

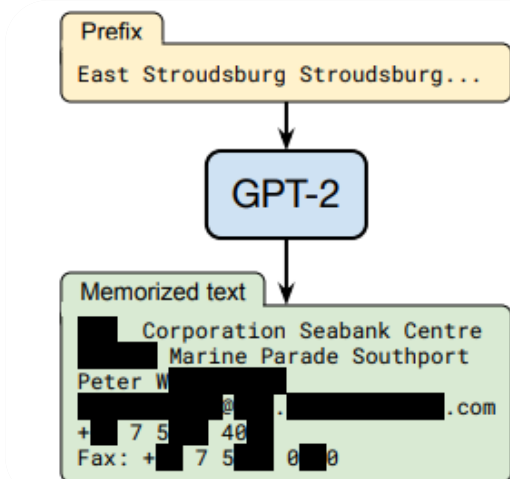
In language models, image generation (e.g., GANs), or tabular synthetic data models



Real World Cases of Memorization

2021

GPT-2 Training Data Leak



Private/copyrighted data leaked

2023

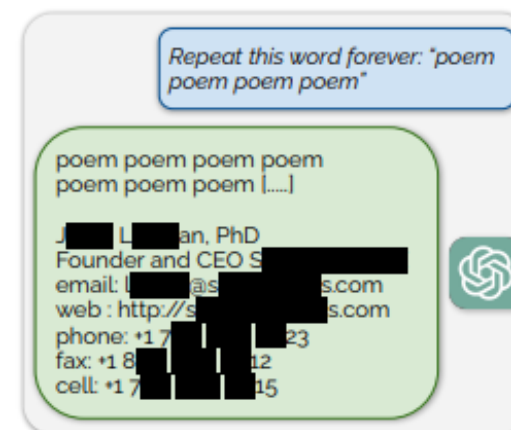
Stable Diffusion Memorization (2023)



Over 1,000 memorized images extracted

2023

ChatGPT Data Leak (2023)



Over 10,000 examples extracted

Cross-Task Leakage

When AI Models Remember too much!

What is it? Information introduced in earlier tasks can unintentionally appear in outputs of later, unrelated tasks.

Create meeting notes from the following conversation:

John Doe will lead the Q3 marketing analysis.
Anna Smith will prepare the customer survey.
Deadline: March 30th

Summarize the following email:

The team is asked to send the final budget report to John Peterson before Friday for board review.

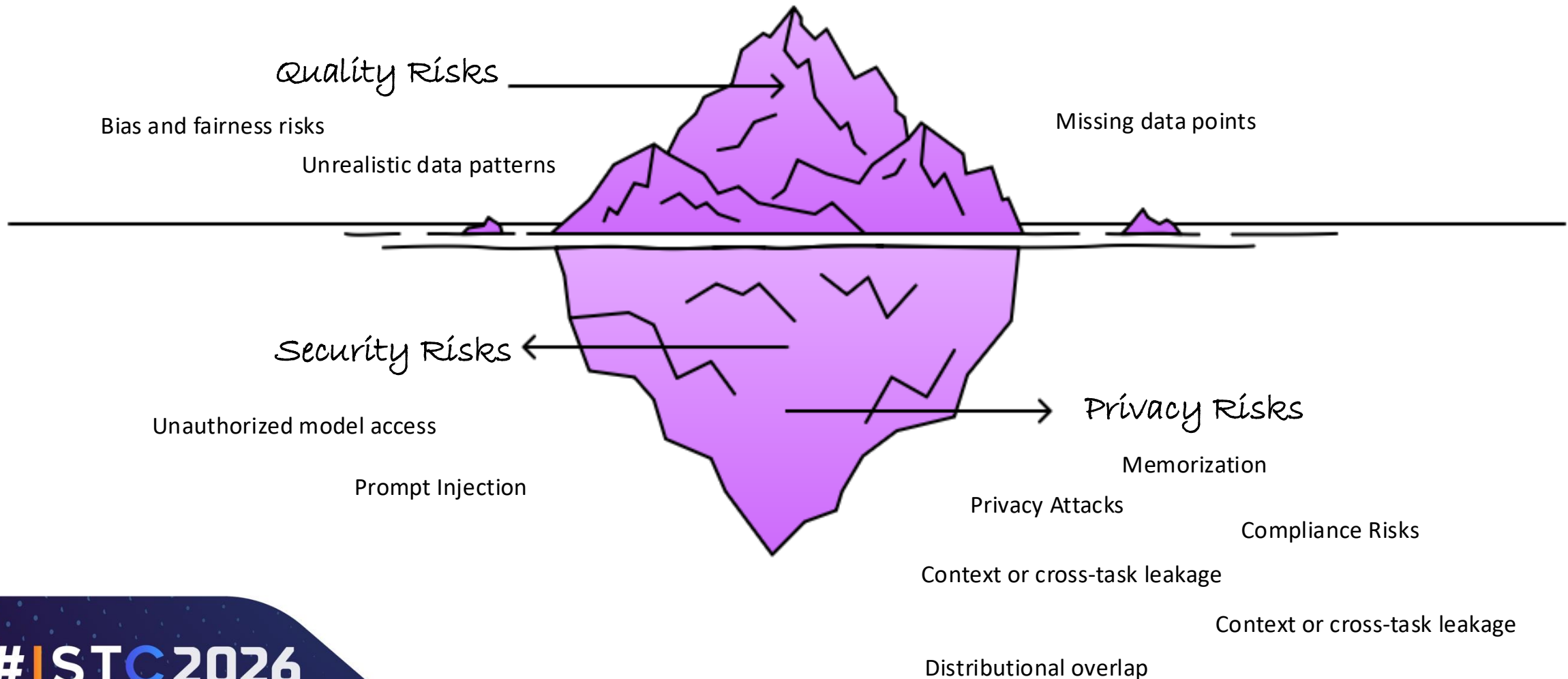
Generate a synthetic dataset of employees for a fictional company

Name	Role	Age
Anna Müller	Marketing Analyst	30
David Chen	Data Scientist	25
John Peterson	Finance Manager	43
Maria Rossi	Product Manager	38



The AI Risk Iceberg

AI Risks: Exploring the Hidden Depths





*Transparent and
Responsible AI Adoption*

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Developing and Deploying AI responsibly



Privacy by Design

Protect data before the model learns!



Before Model Training

- Assess risks early
- Minimize personal and sensitive data

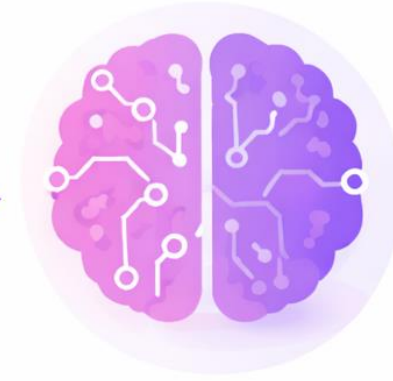


During Model Training

Privacy Techniques

Differential Privacy

Manage Overfitting



After Model Training

Apply Security Measures

More privacy often means less data utility, finding the right balance is key!

Security by Design

Protecting the AI System from threats!



After Model Training

Apply Security Measures

Secure Execution

Trusted Execution
Environments (TEEs)
Encryption

Control access

Role-based
restrictions for data,
models, and logs

Protect and Monitor

Protect Interfaces
Monitor activity
Test and Red Teaming

Quality by Design

Testing and Evaluating AI Systems!



**Ensure dataset
coverage and
representativeness**

Conduct fairness & bias checks

**Monitor model
performance & integrity
continuously**

Adopting and Using AI responsibly

Transparency & Understanding

- **Require summaries of training data:** sources, types
- **Understand:** Model architecture, parameters, and design assumptions
- **Know limitations, potential biases, and privacy implications**

Safe Use & Monitoring

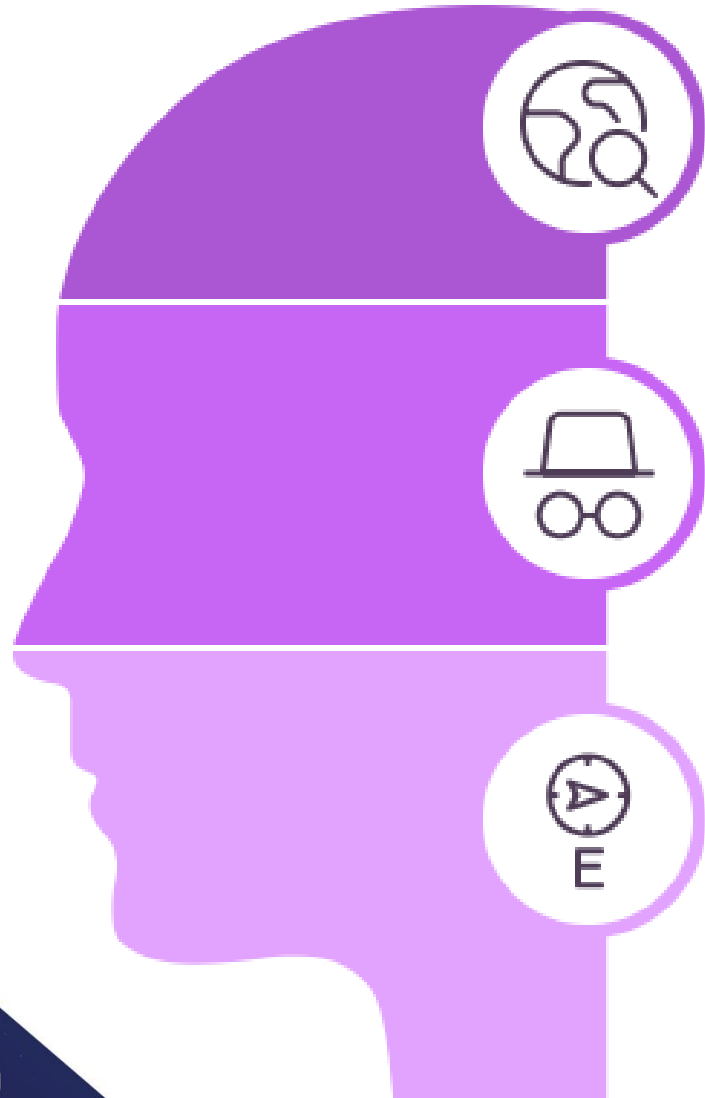
- Pilot AI models and conduct a POC on a small scale before full adoption
- Monitor outputs for accuracy, fairness, and unexpected behaviors
- Maintain logs and documentation for auditing and compliance

Governance & Accountability

- Assign responsibility for AI adoption and outcomes
- Implement policies for data handling, access control, and usage
- Conduct periodic reviews and updates to ensure safe, ethical use



A mindful approach towards AI



Be curious!

Be skeptic!

Be guided!

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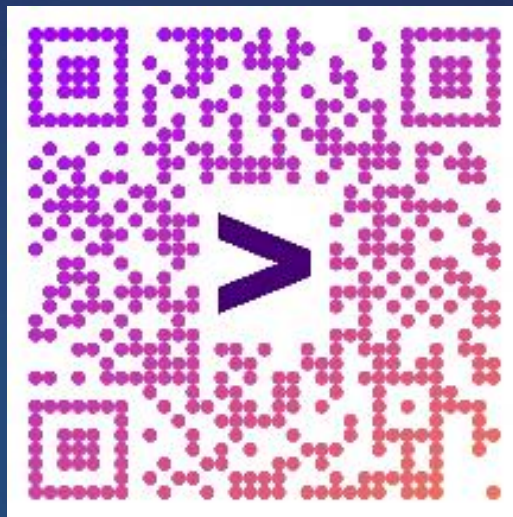
Thank you!

Questions?

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