PROTECTING SENSITIVE INFORMATION

AGENDA



- Best Practices on how to protect company and personal data
- Protection techniques throughout the data pipeline
- Deep Dive into preprocessing & anonymization
- Machine Learning models gone rogue



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Panel Discussion with Marie Jadhwani and Pete Stiglich



Monica Kay Royal

Founder & Chief Data Enthusiast





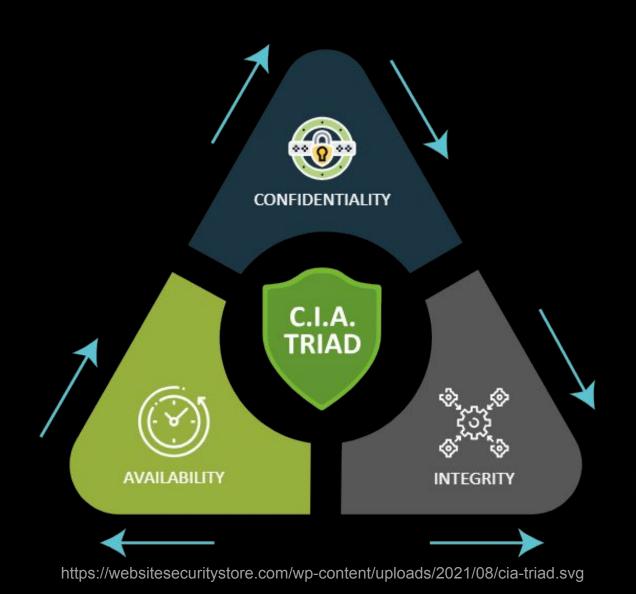
INTRODUCTION

- 16+ years experience
- Accounting & Management Information Systems
- Security/Risk/Compliance
- CPA, CISA, CISM, AAIA
- **Data Career Coaching**
- Instructor at LinkedIn Learning
- Podcast Host: Data Podcast for Nerds!
- Board Member with ISACA: Academic Relations

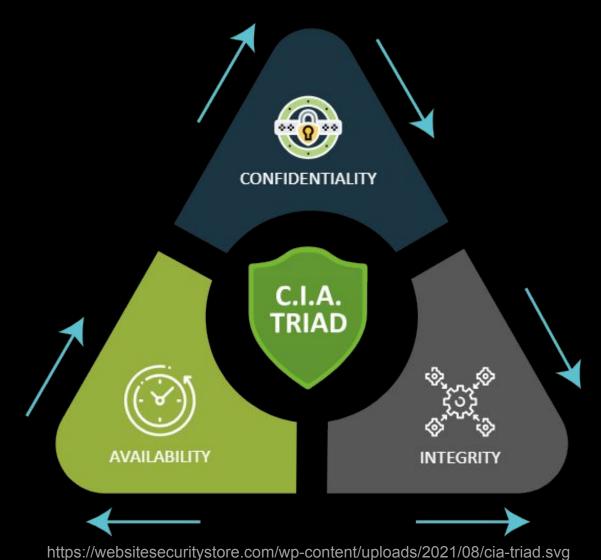




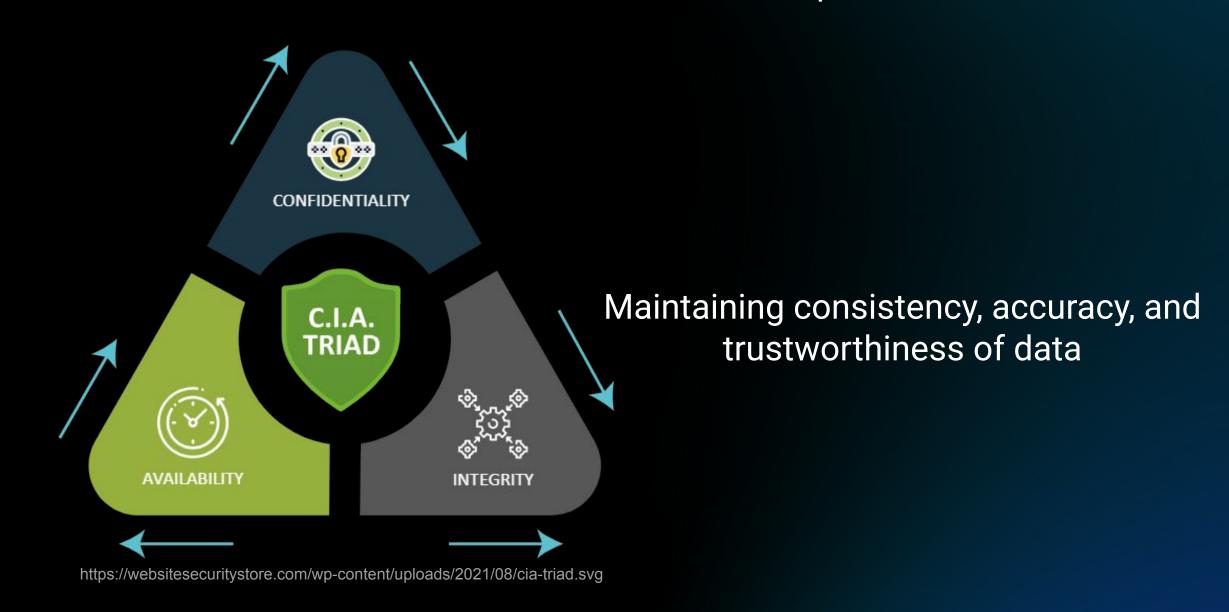
Process of maintaining the confidentiality, integrity, and availability of an organization's data



a.k.a Privacy
Preventing sensitive information from unauthorized access attempts

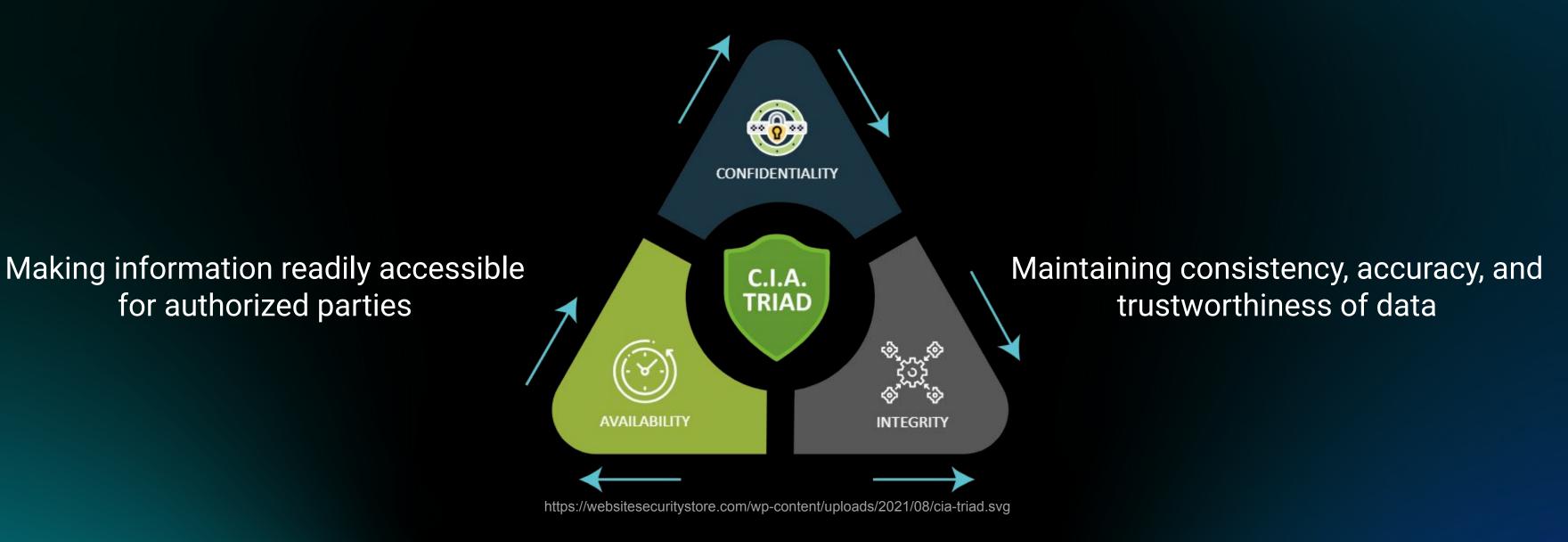


a.k.a Privacy
Preventing sensitive information from unauthorized access attempts



for authorized parties

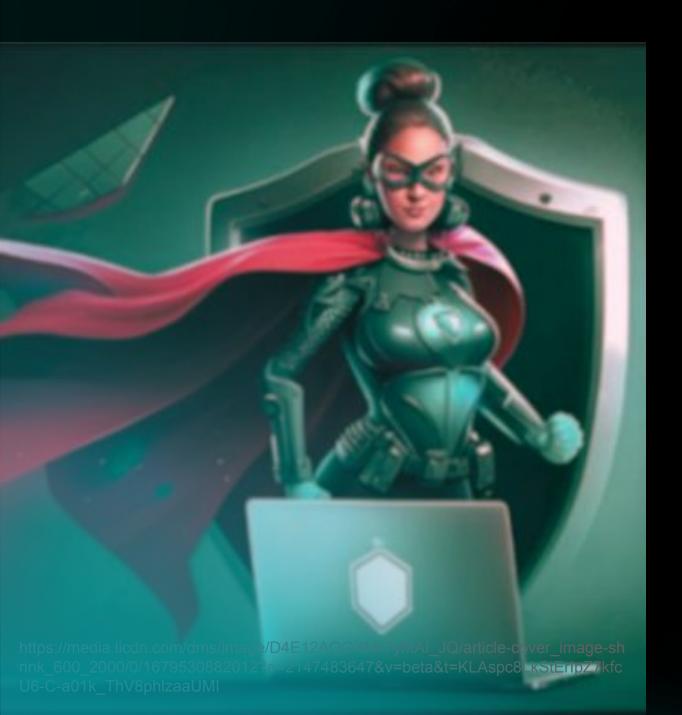
a.k.a Privacy Preventing sensitive information from unauthorized access attempts



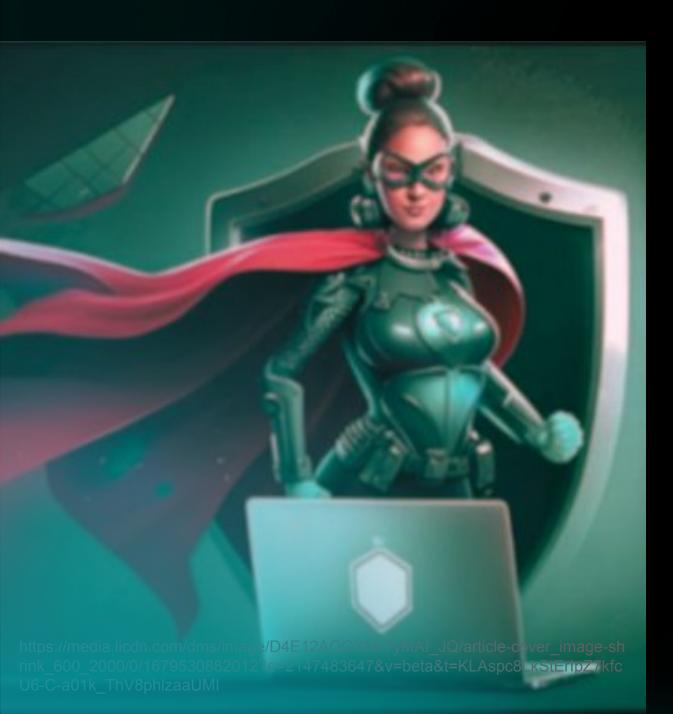
Information security is the practice of protecting information, with a goal to ensure the safety and privacy of critical and sensitive data.



Great Data Great Responsibilities!



Great Data Great Responsibilities!



Protecting Sensitive Information

Data Access Data Collection

Data Preprocessing

Data Sharing Data Storage



Data Access Data Collection

Data Preprocessing

Data Sharing

Data Storage



need to protect

Data Access Data Collection

Data Preprocessing

Data Sharing Data Storage





need to protect

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need to protect

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need to protect

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Data Preprocessing

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need to protect

Data Access Data Collection

Data Preprocessing

Data Sharing

Data Storage













need to protect

Data Access Data Collection

Data Preprocessing

Data Sharing Data Storage







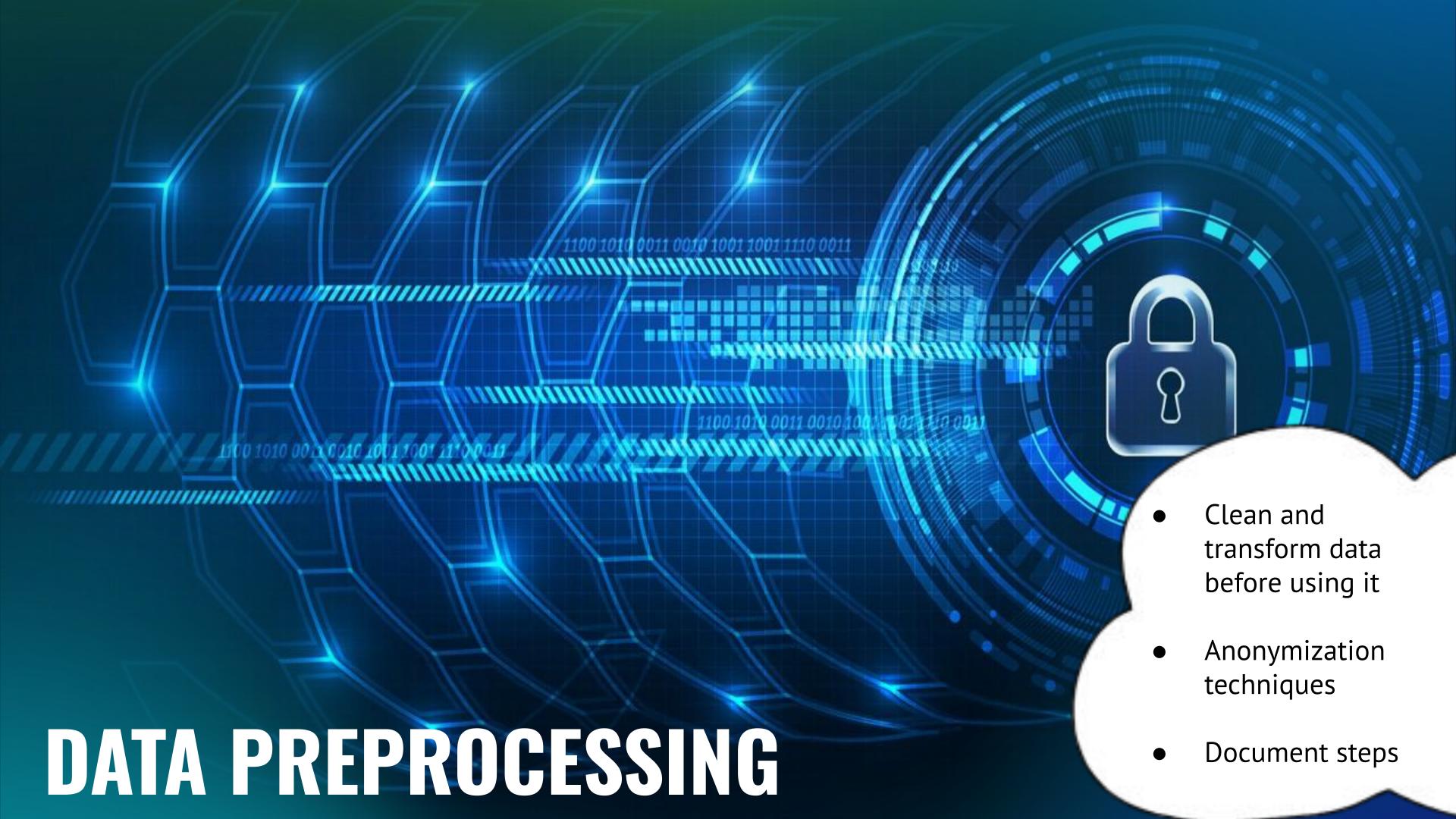


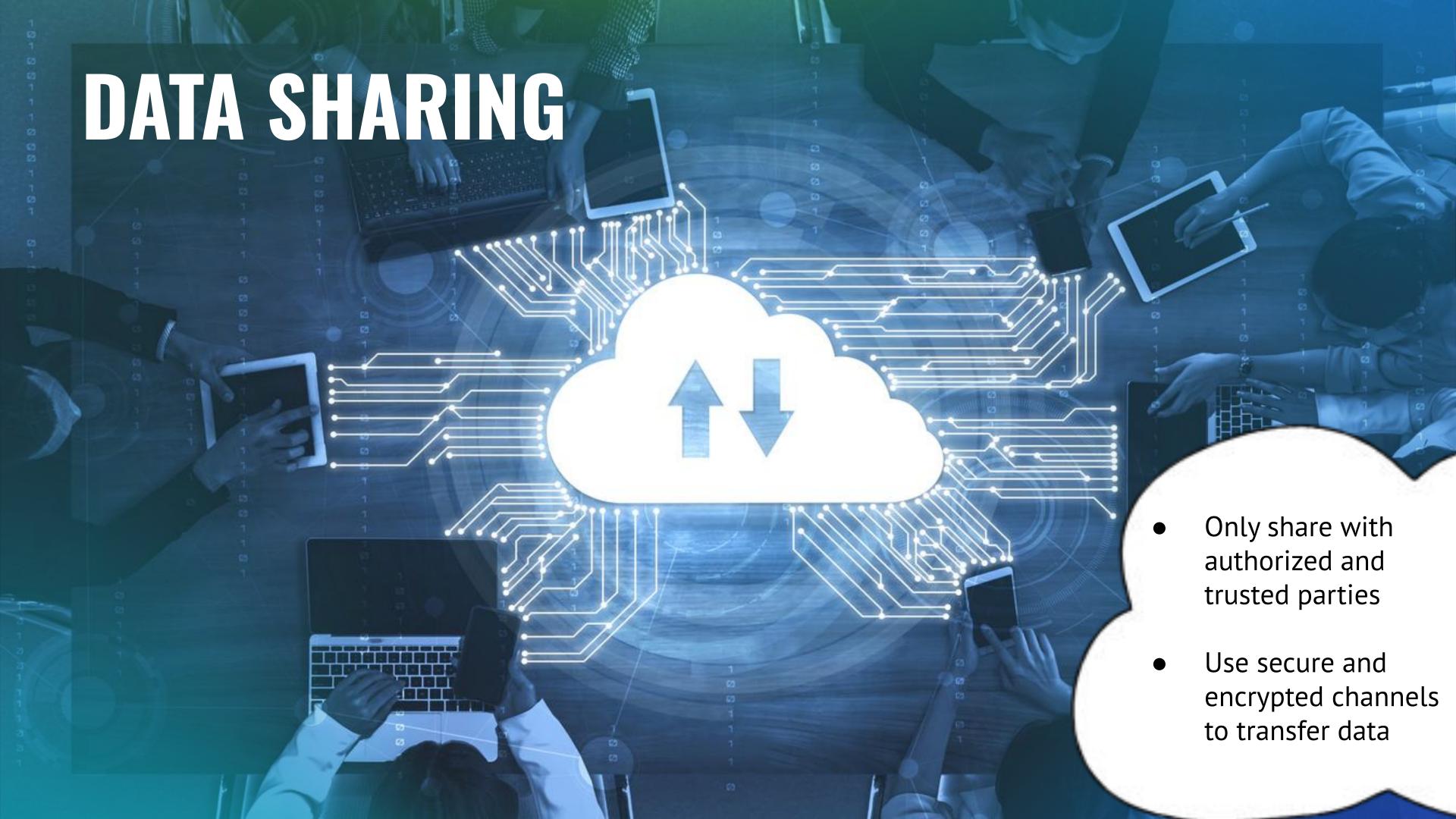


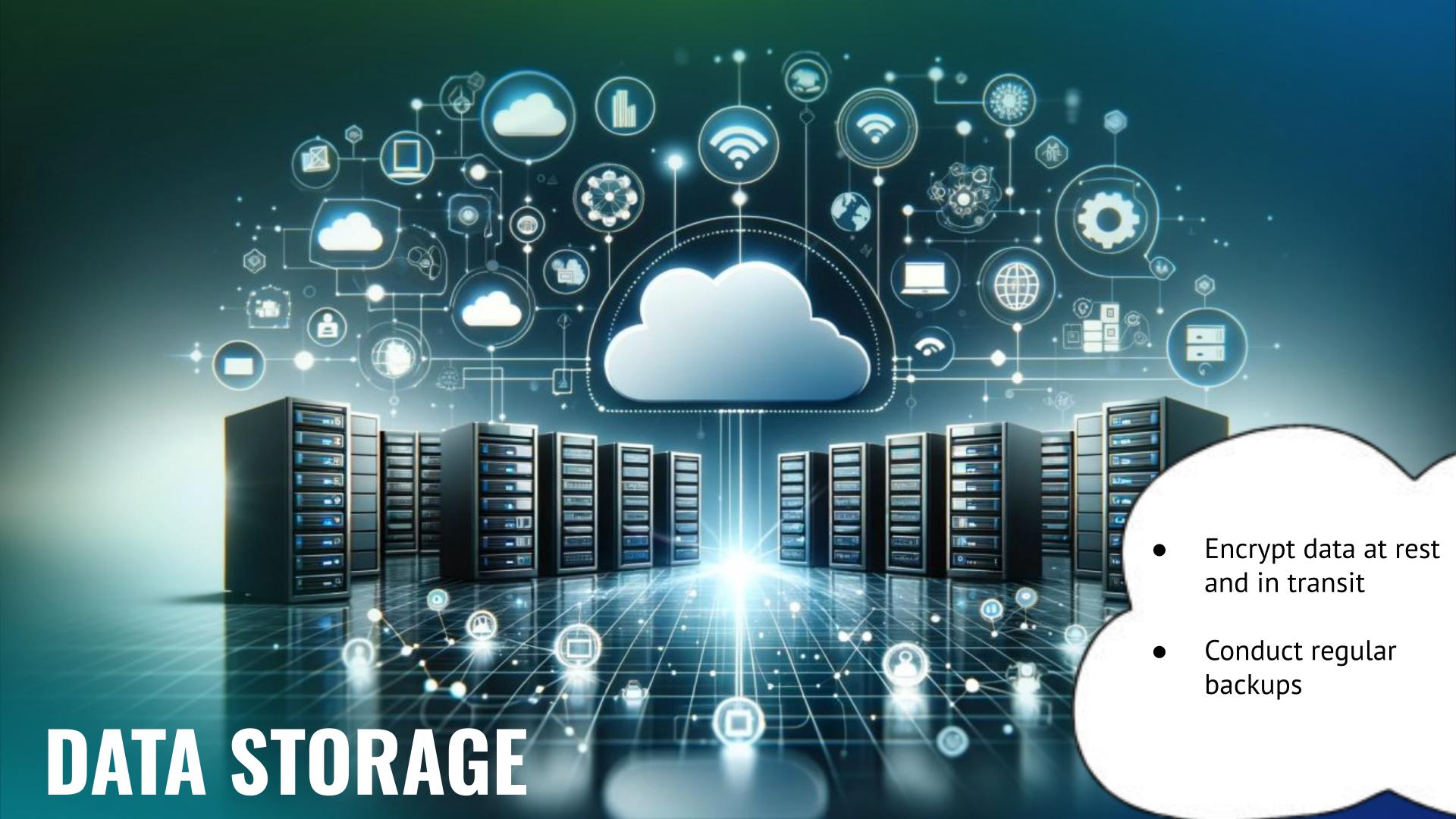














Why This Matters



Why This Matters

Data is EVERYWHERE!

We work with data every single day













~ Monica Kay Royal

CRIME-AS-A-SERVICE

According to Help Net Security, Crime-as-a-Service (CaaS) is the practice of experienced cybercriminals selling access to the tools and knowledge needed to execute cybercrime. CaaS enables cybercriminals to outsource various aspects of their operations, such as malware development, ransomware distribution, botnet rental, or data theft.

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The cost of global cybercrime has been estimated to reach

US \$10.5tn by 2025

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According to a paper from Cornell, ML models can learn too much from the training set which can be leaked and used to recreate part of the dataset if breached.



ML MODEL LEAKAGE



Trained the model using conventional training algorithms for image classification (row 1)

Experiments

- White-Box Attacks: adversary can inspect the parameters of the model after it is available
 - Correlated Value Encoding Attack (row 2)
 - Sign Encoding Attack (row 3)
- Black-Box Attacks: adversary can only send queries to a prediction API
 - Capacity Abuse Attack (row 4)

COUNTERMEASURES



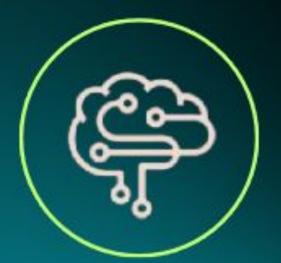
Data Minimization

Only collect necessary data



Perturbation

Creating random noise



Synthetic Data

Create artificially generated data







DATA PREPROCESSING

ML MODELS



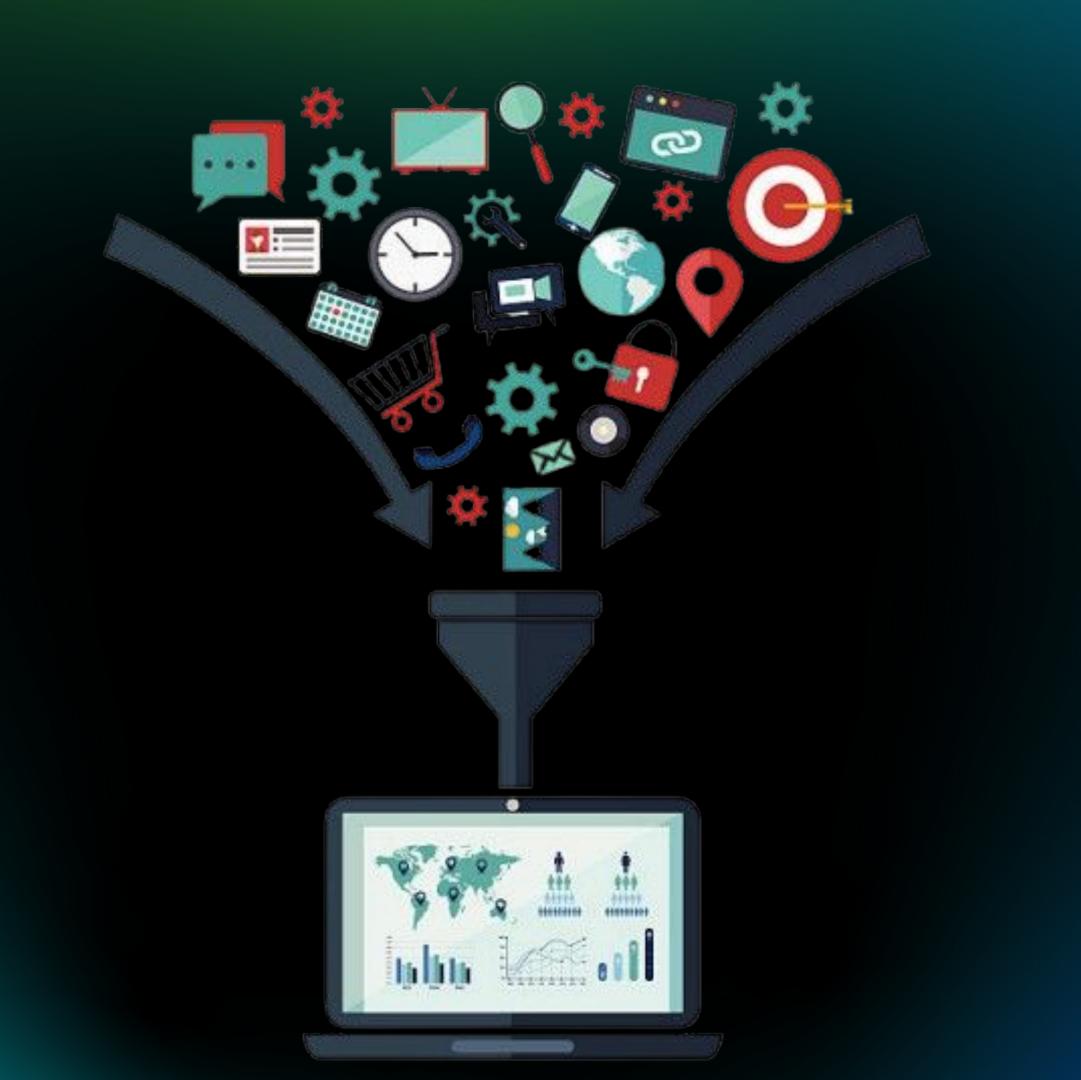
What is a model?

A program that can find patterns or make decisions from a previously unseen dataset.

Types of Decisions...

- Underwriting Process
- Targeted Advertising
- Hiring Process





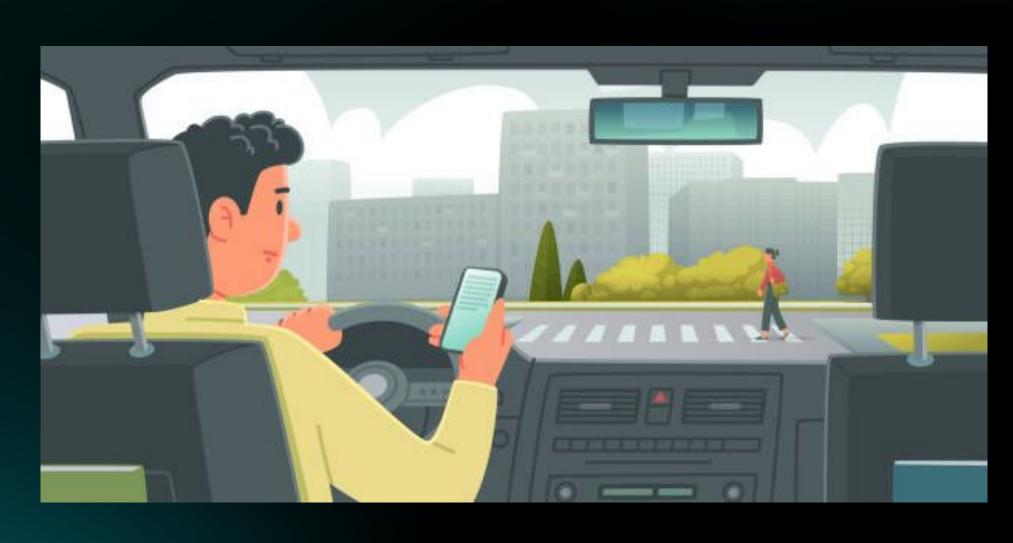
UNDERWRITING





UNDERWRITING



















EMPLOYMENT APPLICATION

nal Information
Email Address
on & Availability
Enter the date you can start work





SPECIAL OFFER

Earn a \$200 Statement Credit and 20,000 Bonus Miles*

Plus enjoy your first checked bag free on Delta Flights

\$0 introductory annual fee for the first year, then \$99

*Statement credit issued approximately 8-12 weeks after you make a Delta purch miles will be issued after you make \$1,000 in purchases on your new Card in you

Rates & Fees

Offer Terms

Apply for the Delta SkyMiles® Gold American Express Card

Southwest's

esto

Select card below or compare cards







Southwest Rapid Rewards® Plus Credit Card

Earn 40,000 points

after you spend \$1,000 on purchases in the first 3 months from account opening.*

Apply now >

†Pricing & Terms | *Offer Details

\$69 annual fee applied to your first billing statement.†

Southwest
Rapid Rewards

GARY KELLY

VISA

Already a Cardmember? >



HIRING PROCESS







HIRING PROCESS







Employment History

Purchasing Patterns

Personality Tests

Personal Income

Demographics

Credit Scores

Locations

Behaviors

GPA

ANONYMIZATION TECHNIQUES



Mask

Concealing specific parts of data



Slightly modifying dates





Generalization

Aggregating data at a higher level

Data Swapping

Shuffling specific attributes



Pseudonymization

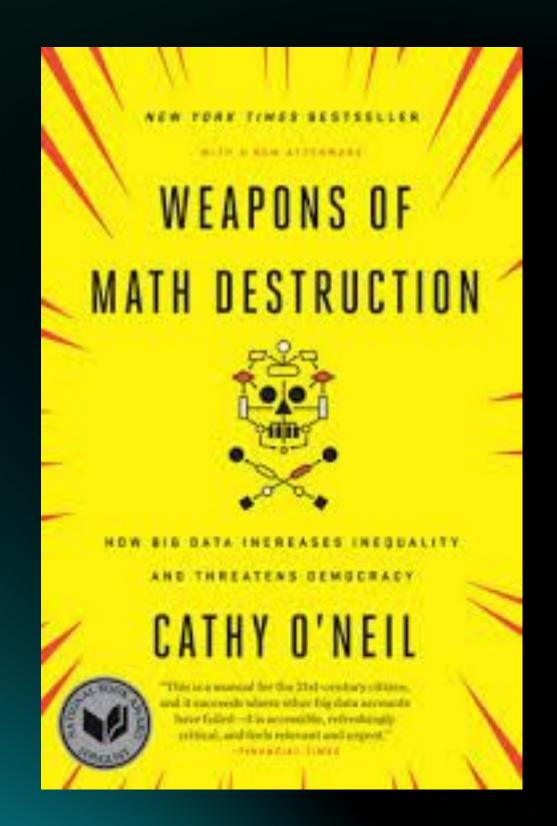
Creating pseudo names to protect identities

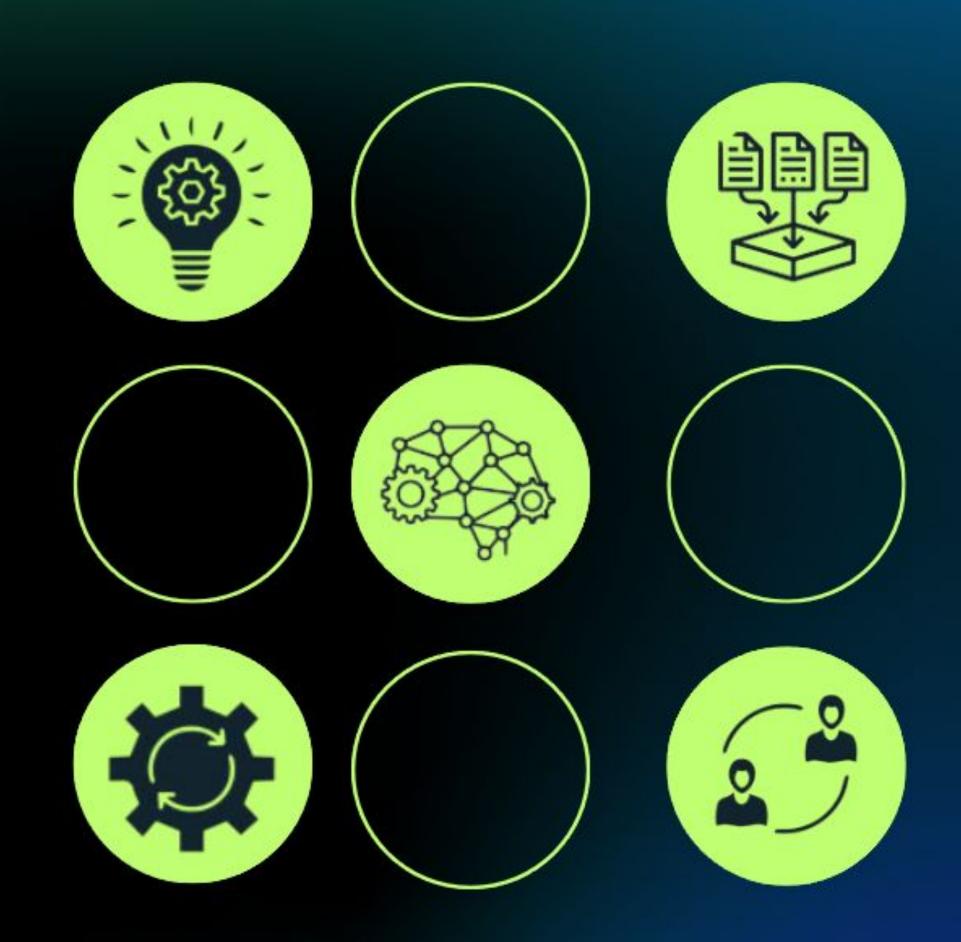


Tokenization

Replacing sensitive data with unique identifiers







Reevaluate Success



- Reevaluate Success
- Feedback Loops



- Reevaluate Success
- Feedback Loops
- Transparency



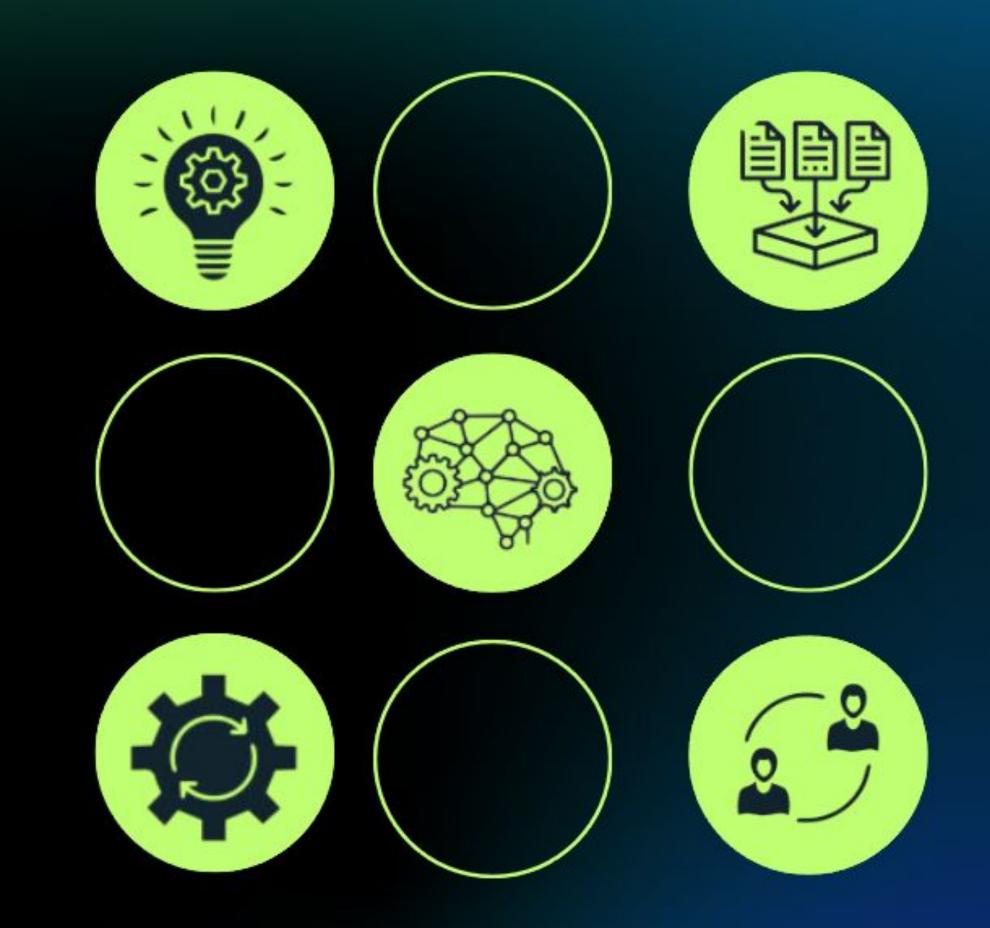
- Reevaluate Success
- Feedback Loops
- Transparency
- Regulation Updates

Fair Credit Reporting Act (FCRA)

Equal Credit Opportunity Act (ECOA)

Americans with Disabilities Act (ADA)

- Reevaluate Success
- Feedback Loops
- Transparency
- Regulation Updates
- Algorithmic Audits





FUTURETRENDS

- Privacy-Preserving AI / ML Frameworks
- Advanced Anonymization Techniques
- Cross-disciplinary Approaches
- Educational Initiatives







<u>Faker</u> is a Python package that generates fake data for you. Installation: <u>Help Link</u> Open Anaconda prompt command to install:

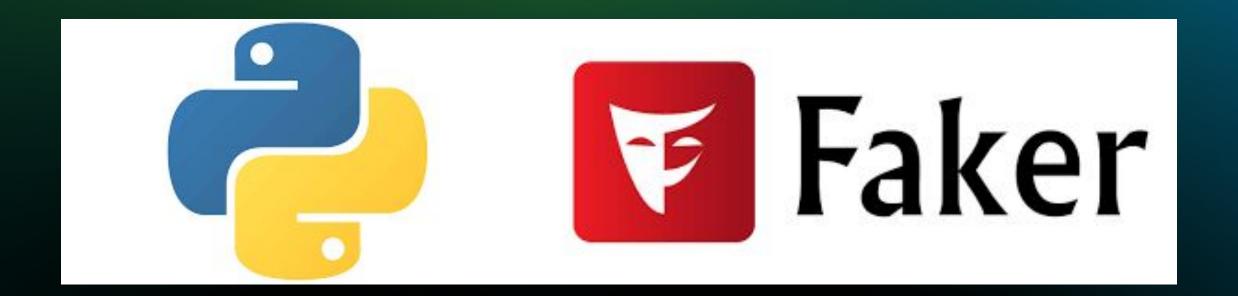
```
conda install -c conda-forge faker
```

Import package

from faker import Faker

Faker has the ability to print/get a lot of different fake data, for instance, it can print fake name, address, email, text, etc. Important most commonly used faker commands

```
fake.name()
fake.address()
fake.email()
fake.text()
fake.country()
```



OUTPUT: (Different every time)

vwilson@hotmail.com

Belgium

Shane Hunter

Commodi vel libero placeat quibusdam odio odio consequatur. Ducimus libero quae optio non quidem. Facilis quas impedit quo.

26.5687745 -124.802165

http://www.turner.com/



```
Getting setup
  %pip install openai
  %pip install pandas
  %pip install scikit-learn
  %pip install matplotlib
  from openai import OpenAI
  import os
  import re
  import numpy as np
  import pandas as pd
  from sklearn.cluster import KMeans
  import matplotlib.pyplot as plt
  import json
  import matplotlib
  client = OpenAI(api_key=os.environ.get("OPENAI_API_KEY", "<your OpenAI API key if not set as env var>"))
```



1. CSV with a structure prompt @

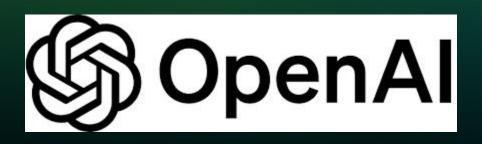
Here we create data in the simplest way. You can quickly generate data by addressing 3 key points: telling it the format of the data (CSV), the schema, and useful information regarding how columns relate (the LLM will be able to deduce this from the column names but a helping hand will improve performance).



```
datagen_model = "gpt-4o-mini"
 question = """
 Create a CSV file with 10 rows of housing data.
 Each row should include the following fields:
  - id (incrementing integer starting at 1)
  - house size (m^2)
  - house price
  - location
  - number of bedrooms
Make sure that the numbers make sense (i.e. more rooms is usually bigger size, more expensive locations
increase price. more size is usually higher price etc. make sure all the numbers make sense). Also only
respond with the CSV.
 response = client.chat.completions.create(
   model=datagen_model,
   messages=[
     {"role": "system", "content": "You are a helpful assistant designed to generate synthetic data."},
     {"role": "user", "content": question}
 res = response.choices[0].message.content
 print(res)
```



```
```csv
id, house_size_m2, house_price, location, number of bedrooms
1,50,150000, Suburban, 2
2,75,250000,City Center,3
3,100,350000,Suburban,4
4,120,450000,Suburban,4
5,80,300000, City Center, 3
6,90,400000, City Center, 3
7,150,600000, Premium Area,5
8,200,750000, Premium Area,5
9,55,180000,Suburban,2
10,300,950000, Premium Area,6
...
```



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```



```
output_string = ""
for i in range(3):
 question = f"""
 I am creating input output training pairs to fine tune my gpt model. The usecase is a retailer generating a description
 The format should be of the form:
 1.
 Input: product_name, category
 Output: description
 2.
 Input: product_name, category
 Output: description
 Do not add any extra characters around that formatting as it will make the output parsing break.
 Create as many training pairs as possible.
 response = client.chat.completions.create(
 model=datagen model,
 messages=
 {"role": "system", "content": "You are a helpful assistant designed to generate synthetic data."},
 {"role": "user", "content": question}
 res = response.choices[0].message.content
 output_string += res + "\n" + "\n"
print(output string[:1000]) #displaying truncated response
```

I am creating input output training pairs to fine tune my gpt model. The usecase is a retailer generating a description for a product from a product catalogue. I want the input to be product name and category (to which the product belongs to) and output to be description.



1.

Input: Wireless Bluetooth Headphones, Electronics

Output: Immerse yourself in high-quality sound with these Wireless Bluetooth Headphones, featuring active noise cancellation and a comfortable over-ear design for extended listening sessions.

2.

Input: Organic Green Tea, Beverages

Output: Enjoy a refreshing cup of Organic Green Tea, sourced from the finest leaves, packed with antioxidants, and perfect for a healthy, invigorating boost anytime.

3.

Input: Stainless Steel Kitchen Knife, Kitchenware

Output: Cut with precision and ease using this Stainless Steel Kitchen Knife, designed with an ergonomic handle and a sharp blade for all your culinary tasks.

4.

Input: Hiking Backpack, Outdoor Gear

Output: Explore the great outdoors with this durable Hiking Backpack, featuring multiple compartments for optimal organization and a breathable design for ultimate comfort on long treks.

5.

Input: Air Fryer, Kitchen Appliances

Output: Cook your favorite meals with less oil using this Air Fryer





### PANEL:

### USE CASES AND SECURITY CONSIDERATIONS FOR GENAI



Marie Jadhwani



**Pete Stiglich** 



# Protecting Data for Analysis and Machine Learning





### RESOURCES

- O'Neil, C. (2016). Weapons of Math Destruction, How Big Data Increases Inequality and Threatens Democracy. Penguin Books
- Song, C. Ristenpart, T. Shmatikov, V. (2017). Machine Learning Models that Remember Too Much. Association for Computing Machinery.
- Python Faker Library: <a href="https://www.geeksforgeeks.org/python/python-faker-library/">https://www.geeksforgeeks.org/python/python-faker-library/</a>
- OpenAl Cookbook: Senthethic data generation (Part 1): <a href="https://cookbook.openai.com/examples/sdg1">https://cookbook.openai.com/examples/sdg1</a>





https://www.linkedin.com/in/monicakayroyal/



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Protecting Data for Analysis and Machine Learning

