

# Fundamentals of Data Security

By Prof.Vishal Chugh



# What is Data Security ?

Data security is the practice of protecting digital information from unauthorized access, corruption or loss throughout its entire lifecycle

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# Process to Secure and Access the Data

## Encryption

Encryption is the process of converting information into a coded format, making it unreadable to anyone without the proper key

## Decryption

It is the process of converting encrypted data back into its original, readable form

# Major types of Encryption used in Industry

## Symmetric

A single secret key is used to encrypt and decrypt data.

## Asymmetric

Uses a pair of keys: a public key that can be shared with anyone for encryption and a private key that must be kept secret for decryption.

# Pros

## Symmetric

Generally faster and more efficient than asymmetric encryption, making it suitable for encrypting large amounts of data.

## Asymmetric

Offers better security for key distribution, as the private key never needs to be shared.

# Cons

## Symmetric

Requires secure key distribution, as the same key must be shared between the sender and recipient.

## Asymmetric

Generally slower and less efficient than symmetric encryption, making it less suitable for large data volumes.

# Encryption in Python

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# Import the Fernet from Cryptography

```
from cryptography.fernet import Fernet
```



# Generate Key and store it in .key file

```
[ ] key = Fernet.generate_key()
```

```
[ ] print("Your key value is:", key)
```

➡ Your key value is: b'V7q2zssFAK45587Rr5QyVomC1W0QwKrpr3ZQN6jlGhw='

```
[ ] with open('filekey.key', 'wb') as f:  
    f.write(key)
```

# Create Encryption

```
# Encryption
with open('filekey.key', 'rb') as f:
    key = f.read()
```

```
# Create a Fernet object using the key
fernet = Fernet(key)
```

```
# Open the file to be encrypted in binary read mode
with open('file_name.csv', 'rb') as f:
    original = f.read()
```

```
# Encrypt the file content
encrypted = fernet.encrypt(original)
```

```
# Overwrite the original file with the encrypted data
with open('file_name.csv', 'wb') as f:
    f.write(encrypted)
```

# Decryption in Python

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# Import the Fernet from Cryptography

```
from cryptography.fernet import Fernet
```

## Load .key file and encrypted data file

```
# Load the key again
with open('filekey.key', 'rb') as f:
    key = f.read()
```

```
# Create a Fernet object
fernet = Fernet(key)
```

```
# Read the encrypted data from the file
with open('file_name.csv', 'rb') as f:
    encrypted = f.read()
```

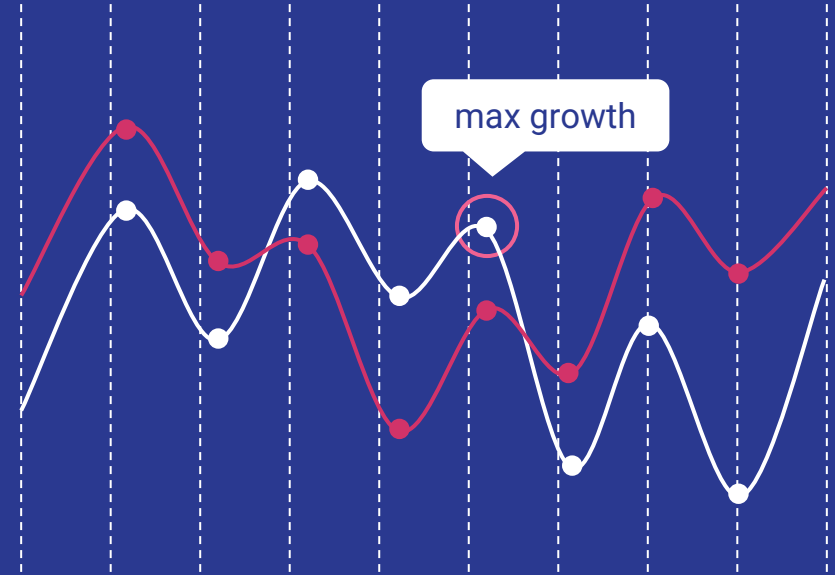
## Create Decryption

```
# Decrypt the encrypted data  
decrypted = fernet.decrypt(encrypted)
```

```
# Write the decrypted data back to the file  
with open('file_name.csv', 'wb') as f:  
    f.write(decrypted)
```

# Common Steps in Python

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# First Step Import Necessary Libraries

```
# Import Libraries
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
import copy
```



# Load data in dataframe using pandas

```
# Loading the dataset  
variable_name = pd.read_csv('file_name.csv')  
variable_name = pd.read_excel('file_name.xlsx')  
variable_name = pd.read_json('file_name.json')
```

# See the first 5 rows and last 5 rows of data

```
# See the first 5 rows and last 4 rows in the loaded dataset  
variable_name.head() # -> Shows first 5 rows  
variable_name.tail() # -> Shows last 5 rows
```

## Inspecting the data types of all columns

```
# Datatypes  
variable_name.dtypes
```

## Inspecting the number of rows and column

```
# How many rows and columns are there  
variable_name.shape
```

## Inspecting the Duplicate Records

```
# Inspect duplicate records  
variable_name.duplicated().sum()
```

# Inspecting the missing values

```
# Inspecting missing values in the dataset  
variable_name.isna().sum()
```

# Creating copy of the original dataset

```
# Creating copy of the dataset  
variable_name_copy = variable_name.copy(deep = True)
```

## Creating Box Plot in missing values column to see outliers

```
# Show the box plot for the column where there are missing values  
sns.boxplot(data= variable_name['Column_1'])
```



For numeric data without outliers we use mean to fill missing values

```
# Replace missing values with mean of column_1  
variable_name['column_1'].fillna(df['column_1'].mean(), inplace= True)
```

For numeric data with outliers we use Median to fill missing values

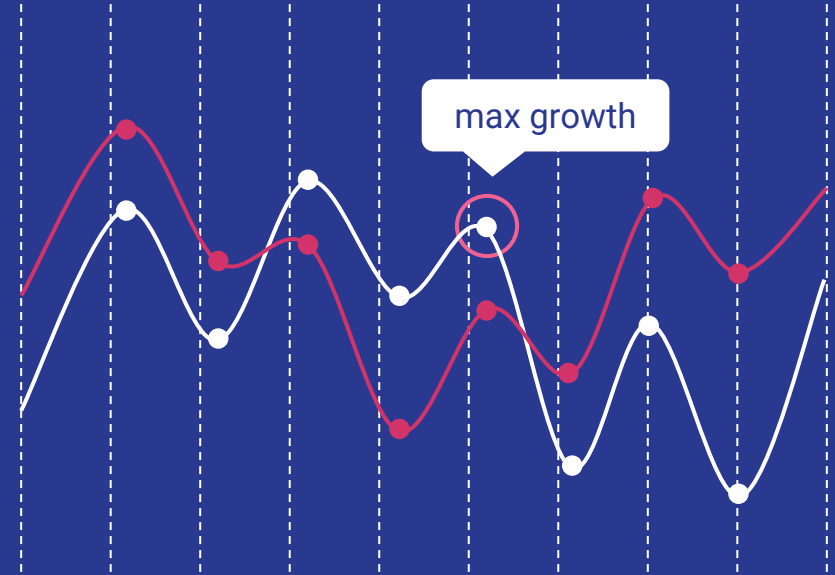
```
# Replace missing values with median of column_1  
variable_name['column_1'].fillna(df['column_1'].median(), inplace= True)
```

For Categorical data we use mode to fill missing values

```
# Replace missing values with mode of column_1  
variable_name['column_1'].fillna(df['column_1'].mode(), inplace= True)
```

# Case Study

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# Case Study: Customer Persona Segmentation for "TrendMart"

## Background

TrendMart is a retail & e-commerce lifestyle brand aiming to personalize marketing strategies. They have collected both **quantitative** (demographics, shopping behavior) and **qualitative** (goals, challenges, lifestyle) customer data.

The marketing team wants to group customers into distinct personas to:

- Target promotions effectively
- Choose the right communication channels
- Improve customer retention

# Questions

## Part 1: Data Preparation

1. Import both datasets in Python and merge them using **Customer\_ID** as the key.
2. Display the first 5 rows of the merged dataset.

## Part 2: Persona Grouping by Demographics

3. Group customers by **Location** (Urban/Rural) and find the **average Annual Income**.
4. Find the **most common Communication Preference** for Urban customers.
5. Group customers by **Technology Usage** and calculate the **average Online Purchase Frequency**.

## Part 3: Persona Grouping by Qualitative Data

6. Find the number of customers whose **Goals and Aspirations** is "**Career growth**".
7. Group customers by **Lifestyle and values** and find the **average Annual Income**.
8. Find the most common **Decision-Making trigger** for customers with **High Technology Usage**.

## Part 4: Combined Segmentation

9. Create a segment of customers who are:
  - Urban
  - Annual Income > ₹1,000,000
  - Technology Usage = "High"Display their **Name, Age, Occupation, and Goals**.
10. Count how many customers have "Discounts and offers" as their **Decision-Making trigger** and shop online more than 5 times a month.