#### 1. Introduction to Pandas

- Pandas is a Python library for data manipulation and analysis.
- It provides two main data structures:
  - Series → 1D labeled array (vertical)
  - DataFrame → 2D table made up of multiple Series

# 2. Importing Pandas

```
import pandas as pd
import numpy as np
```

#### 3. Pandas Series

### **Creating a Series**

```
11 = [10, 20, 30]
12 = ['a', 'b', 'c']
print(pd.Series(11, index=12))
```

### Key Points:

- A **Series** can hold any data type (int, float, string, etc.)
- It is 1-dimensional and labeled vertically
- Custom index can be assigned

### 4. Pandas DataFrame

#### **Creating a DataFrame**

```
data = {
     'Name': ['Ali', 'Qasim'],
     'Age': [11, 12]
}
df = pd.DataFrame(data)
print(df)
```

A **DataFrame** is made up of multiple Series combined together.

# 5. Selecting and Indexing Columns

#### **Selecting Columns**

```
df['Name']  # Single column
df[['Name', 'Age']]  # Multiple columns
```

#### **Adding a New Column**

```
df['Hi'] = [1, 2]
```

### Removing a Column

```
df.drop('Name', axis=1)  # Removes temporarily
df.drop('Name', axis=1, inplace=True)  # Removes permanently
```

### Axis Meaning:

- axis = 0 → rows (horizontal)
- axis = 1 → columns (vertical)

# 6. Selecting Rows

#### **Using loc (label-based)**

```
df.loc[0]  # Select first row
df.loc[[0, 1]]  # Select multiple rows
df.loc[[0, 1], ['Name', 'Age']] # Select specific rows & columns
```

## **Using iloc (index-based)**

```
df.iloc[0:2, 0:2] # Rows 0-1, Columns 0-1
```

## 7. Conditional Selections

```
df[df['Age'] > 19]
df[(df['Age'] > 19) & (df['City'] == 'Paris')]
```

#### Tip:

Always use parentheses () when combining multiple conditions with & or |.

# 8. Handling Missing Data

### **Identifying Missing Values**

```
df.isna()  # Shows True for missing values
df.isna().sum()  # Count of missing values per column
df.isna().any()  # True if column has any NaN
```

### **Dropping Missing Values**

```
df.dropna(thresh=2) # Keeps rows with at least 2 non-null values
```

# Filling Missing Values

```
df.fillna(0)  # Fill NaN with 0
df.fillna(df.mean())  # Fill NaN with column mean
```

# 9. Merging and Joining DataFrames

#### Merging (based on common column key)

```
data1 = pd.DataFrame({
    'A': [1, 2, np.nan, 4, 5],
    'B': [np.nan, 2, 3, 4, 5]
})
data2 = pd.DataFrame({
    'A': [1, 2, np.nan, 4, 5],
    'D': [np.nan, 22, 43, 4, 5]
})
pd.merge(data1, data2, on='A')
                                       # Merge on common column A
pd.merge(data1, data2, how='inner')
                                       # Only common values
pd.merge(data1, data2, how='outer')
                                       # All values from both
pd.merge(data1, data2, how='left')
                                       # All from left
pd.merge(data1, data2, how='right') # All from right
```

#### Joining (based on index)

```
df1.join(df2, how='outer')
```

# 10. Grouping and Aggregation

### **Example**

```
data = {
    'Category': ['A', 'B', 'C'],
    'Store': ['51', '55', '51'],
    'Sales': [100, 200, 300],
    'Quantity': [10, 15, 5],
    'Date': pd.date_range('2023-01-01', periods=3)
}

df = pd.DataFrame(data)
```

```
# Group by single column
df.groupby('Category')['Sales'].sum()
# Group by multiple columns
df.groupby(['Store', 'Category'])['Sales'].sum()
```

#### **Aggregation**

```
df['Sales'].max()
df['Sales'].mean()
df['Sales'].median()
df['Sales'].agg(['mean', 'max', 'min'])
```

#### 11. Pivot Tables and Cross Tabs

#### **Pivot Table**

```
pd.pivot_table(df, values='Sales', index='Region', columns='Product')
pd.pivot_table(df, values='Sales', index='Region', columns='Product',
aggfunc='median')
```

#### **Cross Tabulation**

```
pd.crosstab(df['Region'], df['Product']) # Counts occurrences
```

# 12. Common DataFrame Operations

```
print(df.shape) # (rows, columns)
print(df.columns) # Column names
print(df.info()) # Data info and types
print(df.describe()) # Summary statistics
```

# 13. Applying Functions

```
def sqr(x):
    return x ** 2

df['B'] = df['B'].apply(sqr)
```

# 14. Extracting Features from Text Columns

#### **Example 1: Extract Numbers from Text**

```
def extract_episode(data):
    check = False
    num = ""
    for i in data:
        if i == '(':
            check = True
           continue
        if i == ')':
           break
        if check:
           num += i
    return num
def just_numbers(data):
   value = ""
    for i in data:
        if i >= '0' and i <= '9':
           value += i
    return value
df = pd.read_csv(r'C:\Users\Rapid\Downloads\anime.csv')
df['Episodes'] = df['Title'].apply(extract_episode)
df['Episodes'] = df['Episodes'].apply(just_numbers)
df['Episodes'] = df['Episodes'].astype(int)
```

# **Example 2: Extract Text After a Symbol**

```
def extract_time(data):
```

```
txt = ""
for i in range(len(data)):
    if data[i] == ')':
        for j in range(i+1, i+20):
            txt += data[j]
        return txt

df['Total Time'] = df['Title'].apply(extract_time)
```

# **15. Summary of Common Functions**

Function	Description
pd.read_csv()	Read CSV file
df.to_csv()	Save DataFrame to CSV
df.head()	Display first 5 rows
df.tail()	Display last 5 rows
<pre>df.sort_values(by='col ')</pre>	Sort by a column
<pre>df.rename(columns={'A' :'B'})</pre>	Rename columns
<pre>df.drop_duplicates()</pre>	Remove duplicate rows
<pre>df.reset_index(drop=Tr ue)</pre>	Reset index