

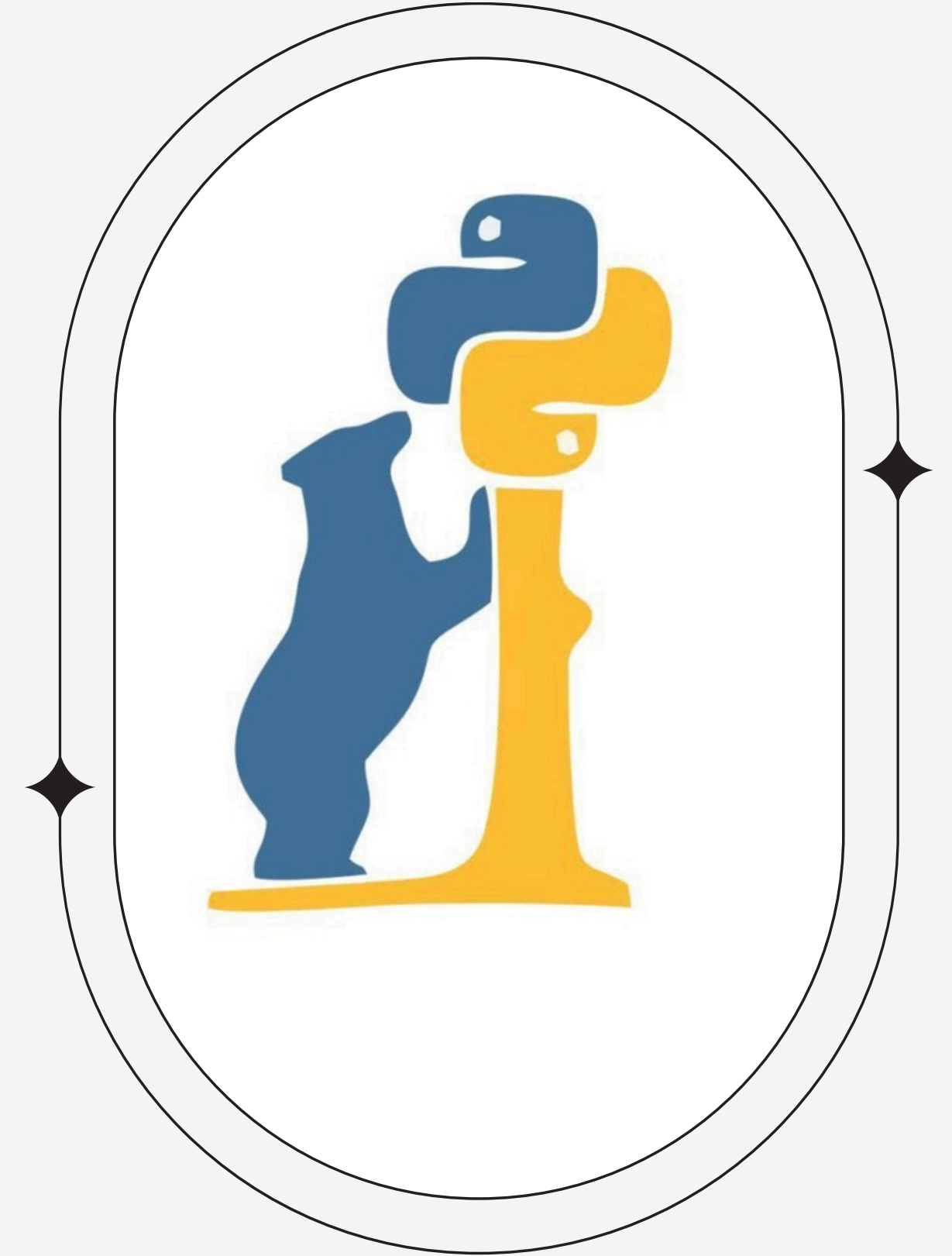
Introduction to PANDAS



Yokshita :)

What is Pandas?

- *Open-source data analysis & manipulation tool*
- *Built on top of NumPy*
- *Created by Wes McKinney in 2008*
- *Highly preferred for data cleaning, transformation, and exploration*



Key Features

- 🎓 Fast and efficient DataFrame object
- 🎓 Tools for reading and writing data
- 🎓 Handling of missing data
- 🎓 Merge and join datasets



Core Data Structures

Series

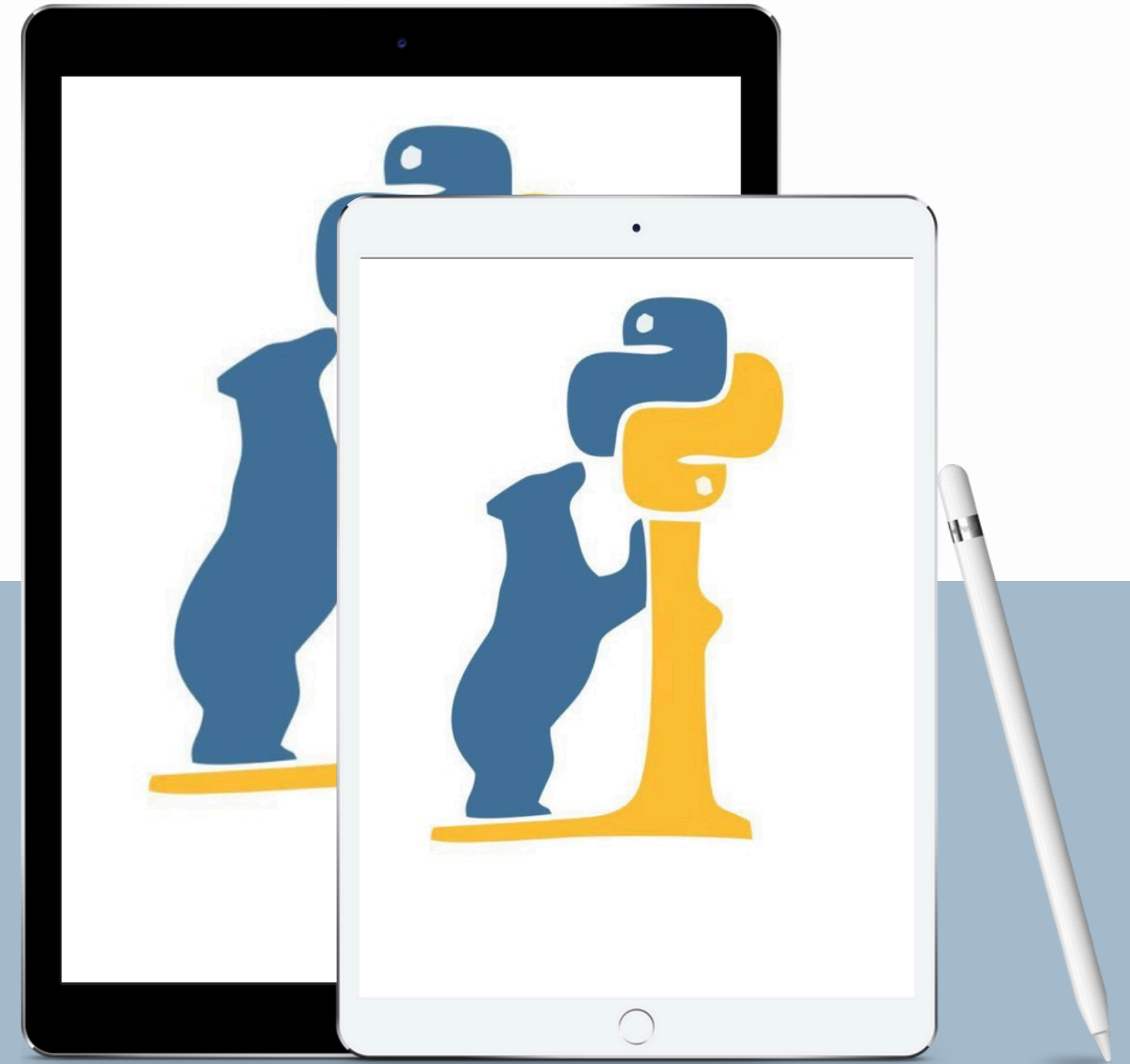
- One-dimensional labeled array
- Example:

```
pd.Series([1, 2, 3])
```

DataFrame

- Two-dimensional labeled data structure
- Example:

```
pd.DataFrame({'Name': ['Alice', 'Bob'], 'Age': [25, 30]})
```



Reading Data

From CSV

- `df = pd.read_csv("data.csv")`

From Excel

- `df = pd.read_excel("data.xlsx")`



Data Exploration

● Viewing Data

df.head(n) *View first n rows (default: 5)*

df.tail(n) *View last n rows (default: 5)*

● Data Summary

df.info() *Overview of data types, non-null values, memory usage*

df.describe() *Statistical summary (mean, std, min, etc.)*

● Data Structure

df.shape *Returns (rows, columns)*

df.columns *List of column names*

df.index *Row index range*



Data Exploration

- **Value Counts**

Frequency of unique values in a column

EXAMPLE `df['column_name'].value_counts()`

- **Checking for Missing Data**

`df.isnull().sum()` *Total missing values per column*

`df.isnull().any()` *Checks if any column has missing values*



Data Cleaning

Handling Missing Values

```
df.dropna(inplace=True)
```

Fill missing values

```
df.fillna(0, inplace=True)  
df.fillna(method='ffill') # Forward fill  
df.fillna(method='bfill') # Backward fill
```

Renaming Columns

```
df.rename(columns={'oldName': 'newName'}, inplace=True)
```

Removing Duplicates

```
df.drop_duplicates(inplace=True)
```

Replacing Values

```
df['Column'].replace('old', 'new', inplace=True)
```



Data Manipulation

Adding Columns

```
df['New_Col'] = [value1, value2, ...]  
df['Double_Age'] = df['Age'] * 2
```

Deleting Columns / Rows

```
df.drop('ColumnName', axis=1, inplace=True) # Drop column  
df.drop(index_number, axis=0, inplace=True) # Drop row
```

Filtering Data

```
df[df['Age'] > 25]
```

Sorting Data

```
df.sort_values(by='Age', ascending=False)
```



Grouping & Aggregating

Basic Syntax

```
df.groupby('Department')
```

Aggregation Functions

```
df.groupby('Department')  
['Salary'].mean()
```

```
df.groupby('Department')  
['Salary'].sum()
```

```
df.groupby('Department')  
['Salary'].count()
```

Multiple Aggregations

```
df.groupby('Department')  
['Salary'].agg(['mean', 'max',  
                'min'])
```

Grouping by Multiple Columns

```
df.groupby(['Department',  
           'Gender'])['Salary'].mean()
```

Merging and Joining

merge()

Combines DataFrames based on column values

```
pd.merge(df1, df2, on='ID', how='inner')
```

join()

Simpler syntax than merge(), joins on index by default

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

concat()

```
pd.concat([df1, df2]) # Row-wise
```

```
pd.concat([df1, df2], axis=1) # Column-wise
```

Real-World Use Cases

