

Data visualization (Basics)

- Representation of the data in a pictorial or graphical format
- Allow us to get the intuitive understanding of the data
- Helps to visualize the patterns in the data
- Python offers multiple great graphing libraries that come packed with lots of different features.
- Matplotlib: low level, provides lots of freedom
- Pandas Visualization: easy to use interface, built on Matplotlib
- Seaborn: high-level interface, great default styles
- Plotly: can create interactive plots
- Bokeh: used creating interactive visualizations for modern web browsers"

Primary Objects of matplotlib



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- The [figure] is the overall figure space that can contain one or more plots
- The [axes] is the individual plots that are rendered within the figure

Anatomy of a figure



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Using matplotlib.pyplot library to plot a chart

Installing library -

```
!pip install matplotlib
```

Importing library -

```
In [ ]: import matplotlib.pyplot as plt
```

Use `plt.show()` as the last line of the code. It suppresses the memory address information

```
In [ ]: import numpy as np
        dates = np.arange('2019-01', '2022-01', dtype='datetime64[M]')
        sales = np.array([42390, 77560, 77385, 76039, 42968, 53833, 47205, 68936, 51175, 48
        profits = np.array([ 7206.3 ,  8531.6 , 13155.45,  9885.07,  7304.56,  9689.94, 566
```

```
In [ ]:
```

Resize the figure

```
In [ ]:
```

Customise plots

- color, labels, title, legends, grid

```
In [ ]:
```

Subplots

Visualise sales and profits as subplots

```
In [ ]:
```

Additional chart types

```
In [ ]: import numpy as np
        products = np.array(['Caffe Latte', 'Cappuccino', 'Colombian', 'Darjeeling', 'Decaf
        sales = np.array([52248.0, 14068.0, 71060.0, 60014.0, 69925.0, 27711.0, 19231.0, 24
        profits = np.array([17444.0, 5041.0, 28390.0, 20459.0, 23432.0, 7691.0, -2954.0, 71
        target_profits = np.array([15934.0, 4924.0, 31814.0, 19649.0, 24934.0, 8461.0, 7090
        target_sales = np.array([48909.0, 13070.0, 80916.0, 57368.0, 66906.0, 30402.0, 1821
```

Bar chart

```
In [ ]:
```

Horizontal bar chart

```
In [ ]:
```

Stacked bar chart

```
In [ ]:
```

Side by Side Bar Chart

In []:

Pie chart

In []:

Bullet chart

In []:

Scatter Plot

In []:

Introduction to Pandas Library

- Pandas is an open source library in python which is know for its rich applications and utilities for all kinds of mathematical, financial and statistical functions
- It is useful in data manipulation and analysis
- It provides fast, flexible, and expressive data structures designed to make working with structured (tabular, multidimensional, potentially heterogeneous) and time series data

Installing pandas

In []: `!pip install pandas`

Importing pandas

In []: `import pandas as pd
import numpy as np
import matplotlib.pyplot as plt`

Series

Series -

- are one-dimensional ndarray with axis labels (homogenous data)
- labels need not be unique but must be of immutable type

Creating Series

Ex. Create series using the given list of names

In []:

Extracting elements from series

Indexing based on index number

In []:

Indexing based on index name

In []:

Assigning names as index to marks

In []:

In []:

Filtering Series / Conditional Indexing

In []:

Operations on Series

In []:

Ranking and Sorting

```
series.sort_values( ascending=True , inplace=False , na_option =  
{"first","last"} )
```

```
series.sort_index( ascending=True , inplace=False )
```

```
series.rank( ascending=False , method={"average","min","dense"} , na_option =  
{"top","bottom"} )
```

In []:

Working with NULLs

In []:

Dataframe

A DataFrame is two dimensional data structure where the data is arranged in the tabular format in rows and columns

DataFrame features:

- Columns can be of different data types
- Size of dataframe can be changes
- Axes(rows and columns) are labeled
- Arithmetic operations can be performed on rows and columns

Creating Dataframes

```
In [ ]: employees = {"Name" : ["Jack", "Bill", "Lizie", "Jane", "George"],  
                    "Designation" : ["HR", "Manager", "Developer", "Intern", "Manager"],  
                    "Salary": [40000, 60000, 25000, 12000, 70000]}
```

```
df = pd.DataFrame(employees)  
df
```

Accessing Dataframes

```
In [ ]:
```

Operations on dataframes

Ex. Average Salary

```
In [ ]:
```

Ex. Average Salary of managers

```
In [ ]:
```

Concataneting and Merging Dataframes

```
In [ ]: df_jan = pd.DataFrame({"Order ID" : range(101, 111), "Sales" : np.random.randint(10  
df_feb = pd.DataFrame({"Order ID" : range(111, 121), "Sales" : np.random.randint(10  
df_mar = pd.DataFrame({"Order ID" : range(121, 131), "Sales" : np.random.randint(10
```

```
In [ ]:
```

```
In [ ]:
```

Concatenate

```
pd.concat(tuple of dfs, ignore_index = False, axis=0)
```

In []:

Merging Dataframes

```
df1.merge(df2, how="", left_on="", right_on="", left_index= "", ,  
right_index="")
```

In []:

```
df_emp = pd.DataFrame({"Name" : ["Jack", "Bill", "Lizie", "Jane", "George"],  
                        "Designation" : ["HR", "Manager", "Developer", "Intern", "Manager"]})  
df_emp
```

In []:

```
base_salaries = pd.DataFrame({"Designation" : ["HR", "Developer", "Manager", "Senior"],  
                              "Salary": [40000, 25000, 70000, 100000]})  
base_salaries
```

Inner Merge

In []:

Left Merge

In []:

Right Merge

In []:

Outer Merge

In []:

Reading data from Data Sources

In []:

```
import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt
```

Reading data from MYSQL or SQLITE3

In []:

```
!pip install sqlalchemy
```

Examples using Coffee Shop Dataset

```
In [ ]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

Ex. Read data from `coffee_sales.csv`

```
In [ ]:
```

Drop a column or row from dataframe

```
In [ ]:
```

Drop null rows

```
df.dropna( axis = 0 , how = "any" , inplace = False )
```

- axis 0 for row or 1 for column
- how - {any or all}

```
In [ ]:
```

Renaming a Columns

Rename Columns (column 5 - 8 are not accessible)

```
In [ ]: headers = ["ShopID", "Year/Month", "Product", "Product Type", "State", "Target Prof
```

Rename Single Column

```
In [ ]:
```

Understanding Data in Dataframe

- `df.shape` - gives the size of the dataframe in the format (row_count x column_count)
- `df.dtypes` - returns a Series with the data type of each column
- `df.info()` - prints information about a DataFrame including the index dtype and columns, non-null values and memory usage
- `df.head()` - prints the first 5 rows of you dataset including column header and the content of each row
- `df.tail()` - prints the last 5 rows of you dataset including column header and the content of each row

```
In [ ]: df_coffee.shape
```

```
In [ ]: df_coffee.dtypes
```

```
In [ ]: df_coffee.info()
```

```
In [ ]: df_coffee.head()
```

```
In [ ]: df_coffee.head(3)
```

```
In [ ]: df_coffee.tail()
```

```
In [ ]: df_coffee.tail(3)
```

Cleaning data

```
df.apply()
```

Convertinf Franchise into str

```
In [ ]:
```

Ex. Converting Sales and Profits columns to float types

```
In [ ]:
```

Working with null values

`df.isna()` - Detect missing values. Return a boolean same-sized object indicating if the values are NA.

`df.fillna(value=None, inplace=False)` - Fill NA/NaN values using the specified method.

```
In [ ]:
```

Ex. Identify Sales made by 'Caffe Latte'

```
In [ ]:
```

Removing Duplicate Data

```
In [ ]:
```

Replacing values

`df.replace(old_value, new_value, inplace=True)`

In []:

Adding a new Column by calculation

Ex. Create columns showing `Sales` and `Profit` targets achieved

In []:

Ex. Count the number times Targets are achieved

In []:

Creating a bar chart to view Target Status

using matplotlib

In []:

using pandas

In []:

using seaborn

In []:

Setting and Resetting Index

Setting Index

`df.set_index(keys, drop=True, inplace=False,)` - Set the DataFrame index (row labels) using one or more existing columns or arrays (of the correct length). The index can replace the existing index or expand on it.

In []:

Resetting Index

`df.reset_index(level=None, drop=False, inplace=False,)` - Reset the index of the DataFrame, and use the default one instead. If the DataFrame has a MultiIndex, this method can remove one or more levels.

In []:

Indexing and Slicing using loc and iloc

Using loc to retrieve data

- loc is label-based
- specify the name of the rows and columns that we need to filter out

Ex. Extract data for franchise 203

In []:

Ex. Extract City column

In []:

Ex. Extract Sales column for Franchise - 203

In []:

Ex. Extract Sales and Profit column for Franchise - 203, 504

In []:

Using iloc to retrieve data

- iloc is integer index-based
- specify rows and columns by their integer index.

Ex. Extract row at index 2

In []:

Ex. Extract rows at index position 2,3,4

In []:

Ex. Extract column at index 0

In []:

Ex. Extract column from index 0 to 2

In []:

Ex. Extra rows 0 to 2 and columns 0 to 2

In []:

Working with dates

In []:

Insert a column in between

```
df.insert( index , column_name , default_value )
```

Create columns Year and Month - extract data using pd.DatetimeIndex

In []:

Extract data for 2018

In []:

Extract data for Jan - 2018

In []:

Extract data for Jan - 2018 and 2019

In []:

Extract data starting from April - 2019

In []:

Extract data from Jan-2019 to Apr-2019

In []:

Ranking and Sorting Dataframes

Ex. Rank the products in descending order of Sales

In []:

Ex. Sort the data in ascending order of Rank

In []:

Grouping Dataframes

```
df.groupby(by=None, as_index=True, sort=True, dropna=True)
```

Ex. Find product wise total Sales - bar chart

In []:

use of `agg()`

Ex. Extract Monthly Sales and Profit

In []:

Analysing Dataframes

- univariate analysis - boxplot, histogram, `value_counts()`, `countplot`, `describe()`
- bivariate analysis
 - categorical X numerical - `barchart`, `piechart`
 - 2 numerical - scatter plot
 - 2 categorical - `crosstab`
- multivariate - pivot table

Univariate Analysis

Summary Statistics

`df.describe()` - Generates descriptive statistics. Descriptive statistics include those that summarize the central tendency, dispersion and shape of a dataset's distribution, excluding NaN values. Analyzes both numeric and object series, as well as DataFrame column sets of mixed data types. The output will vary depending on what is provided.

In []:

`df.value_counts(normalize = False)` - returns a Series containing counts of unique rows in the DataFrame

In []:

Histogram

In []:

Box and Whisker Plot



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Bivariate Analysis

`pd.crosstab(index, columns, values=None, aggfunc=None, normalize=False)` - Computes a simple cross tabulation of two (or more) factors. By default computes a

frequency table of the factors unless an array of values and an aggregation function are passed.

Ex. Number of franchise where a product is sold across each state

In []:

Ex. Product and the number of time Sales Target achieved

In []:

`df.pivot_table(values=None, index=None, columns=None, aggfunc='mean')` - creates a spreadsheet-style pivot table as a **DataFrame**. The levels in the pivot table will be stored in **MultilIndex** objects (hierarchical indexes) on the index and columns of the result **DataFrame**.

In []:

Barchart

Display Sales across products

In []:

Piechart

In []:

Scatter Plot

In []:

Line Chart

Ex. Plot line chart showing monthly sales

using pandas

In []:

using seaborn

In []: