

The Pivotal Role of Data Visualization

- Transforms complex datasets into clear visual stories
- Acts like a detective's tool: organizing scattered clues into solutions
- Real-world impact:
 - ❖ Healthcare → heat maps reveal disease spread
 - **❖ Business** → line graphs show sales trends

Benefits:

- Quickly absorb information
- Spot patterns, correlations, and trends
- Enable data-driven decisions
- Communicate findings effectively

Python's Data Visualization Libraries

- ❖ Matplotlib → Flexible, powerful 2D/3D plots; steep learning curve
- ❖ Seaborn → Simplifies plotting; excellent for statistical visualization
- ❖ Plotly → Interactive plots; higher complexity, but engaging visuals
- ❖ Pandas Visualization → Simple plots; integrates with data workflows
- ❖ Plotnine → Inspired by R's ggplot2; effective for layered graphics
- ❖ Altair → Declarative, clean, and user-friendly statistical graphics

Beyond Python

❖ Tableau, Power BI, D3.js, R's ggplot2 →
Other effective visualization tools

Getting Started with Google Colab

Google Colab – A free, browser-based coding environment requiring no setup and offering free computing power. To get started:

- ❖ Visit the Google Colab website.
- Click on 'File' > 'New notebook' to create a new notebook.
- You are now in a Python environment. You can write code in the cells and run them by clicking the play button on the left or by pressing Shift+Enter.

Getting Started with Google Colab

As a test run, paste the following simple Python code into a cell:

1 print("Hello, Data Visualization!")

- Running the cell prints: "Hello, Data Visualization!"
- Google Colab will be the primary tool for this course
 - Used to interact with Python
 - Employ data visualization libraries

Getting Acquainted with Google Colab

- Cloud-based Python environment
- Runs in the browser, no setup needed
- Works like a Python notebook powered by Google

Setting Up Your First Notebook

- Go to Google Colab website
- ❖ Click File → New notebook

Writing & Executing Code

- Use a code cell (click + Code)
- Type Python code
- Run with Shift+Enter → output shown below cell

Saving & Sharing Notebooks

- Saved directly to Google Drive
- ❖ File → Save to store work
- Share button → invite via email or link
- Option to download for offline sharing

Harnessing Pandas for Data Manipulation

- Core Python library for data analysis
- Provides DataFrames (spreadsheet-like structures)

Uses in Workflow:

- Load datasets (CSV, Excel, SQL, etc.)
- Filter data with simple conditions (e.g., Age > 30)

Utilizing Pandas in Your Data Science Workflow

Performing basic data analysis: Pandas allows statistical analysis, e.g., df.describe() gives descriptive statistics of a DataFrame.

Go to this link:

https://github.com/MikkoDT/MexEE402_Al/blob/main/Python_Visualization/Pandas_for_Data_ _Manipulation.ipynb

In this code:

- Import Pandas library
- Load dataset from URL into a DataFrame
- Print column names of the DataFrame

- Filter data based on conditions
- Use describe() for descriptive statistics
- Prepares and structures data for effective visualization

Matplotlib—Your Tool for Effective Data Visualization

Matplotlib Overview

- Versatile Python plotting library
- Supports static, animated, and interactive plots
- Works seamlessly with Pandas DataFrames

Basic Plots with Matplotlib

- Line Chart (plot) → Track changes over time
- ❖ Bar Chart (bar) → Compare categorical data
- ❖ Histogram (hist) → Show data distribution

Go to this link:

https://github.com/MikkoDT/MexEE402_AI/blob/main/Python_Visualization/Creating_Basic_Plots_with_Matplotlib.ipynb

Each plot reveals unique insights into beer servings, highlighting the importance of data visualization for understanding data.

Understanding Histograms

- Graphical representation of data grouped into bins
- Bins = value ranges; height = frequency of data points
- Provides a clear view of data distribution

Key Insights from Histograms:

- Frequency of values in specific ranges
- Detection of outliers
- Identification of skewness (how the data distribution leans or tilts on a histogram)
 - Symmetrical (no skew): Data is evenly distributed around the center (bell-shaped).
 - ❖ Positive skew (right-skewed): Long tail extends to the right → more values concentrated on the lower end.
 - ❖ Negative skew (left-skewed): Long tail extends to the left → more values concentrated on the higher end.

Real-World Example

Teacher analyzing exam scores

- ❖ Create bins (0-10, 11-20, etc.)
- Visualize class performance distribution

Creating Histograms in Google Colab

- Use Matplotlib to generate histograms
- Steps:
 - Open Google Colab
 - Start a new notebook

Go to this link:

https://github.com/MikkoDT/MexEE402_AI/tree/main/Python_Visualization/Scores

Importing Necessary Libraries

import pandas as pd import matplotlib.pyplot as plt

Real-World Example

Creating and Loading Your Data

For our illustration, we'll use a CSV file containing the final exam scores of a class. You'll need to create this file. Follow these steps:

- 1. Open a text editor on your computer, such as Notepad on Windows or TextEdit on Mac. Or in Microsoft excel.
- 2. Copy and paste the following lines into your text editor:

Save the CSV file "scores.csv"

```
scores
85
90
78
92
88
76
95
89
```

Real-World Example

Upload the CSV file to Google Colab:

- In Google Colab, click the folder icon (left sidebar)
- Click Upload to session storage (upward arrow icon)
- Select scores.csv from your local files
- File is uploaded to your Colab session

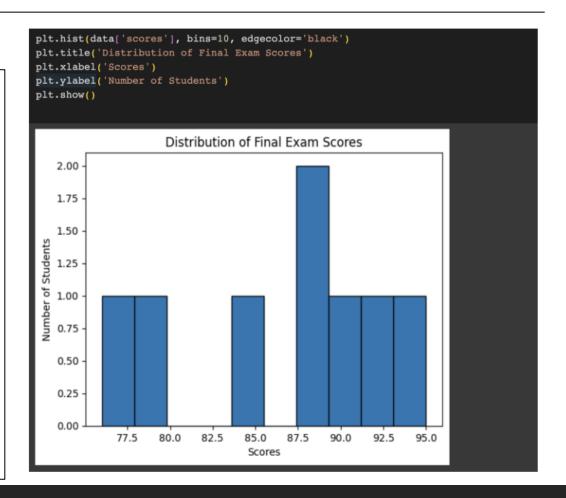
Load the data into a pandas DataFrame:

Finally, load the data into a pandas DataFrame using the following code:

- Generates a histogram of final exam scores
- Bins = score ranges
- Height of bins = number of students in each range

Real-World Example

- Generates a histogram of final exam scores
- Bins = score ranges
- Height of bins = number of students in each range



Understanding Time Series Data

- Collection of data points over time (chronological order matters)
- Example: Hourly temperature readings form a time series
- Applications:
 - **❖ Finance** → stock trends
 - ❖ Weather → forecasting patterns
 - **❖ Healthcare** → tracking patient vitals
 - ❖ E-commerce → analyzing website traffic

Key Value:

Explains the past and provides insights into the future

Visualizing Time Series Data

- Line charts → simple & effective for time series
- Reveal patterns, trends, and outliers (data points that deviate significantly from the overall pattern)
 - Appear as unusually high or low values in a time series
 - May indicate errors, rare events, or important anomalies

Importing Libraries

We will need to import two Python libraries, pandas and matplotlib.pyplot, to get started with data manipulation and visualization respectively. You can import these libraries by running the following code in a new cell:

import pandas as pd

import matplotlib.pyplot as plt

Go to this link:

https://github.com/MikkoDT/MexEE402_Al/tree/main/Python_Visualization/Line_Charts

Creating and Loading Data

For the purpose of this tutorial, we'll consider a dataset representing the daily temperature of a city for a month. Let's walk through the steps to create this data in a CSV file and subsequently load it into our notebook.

Begin by creating a CSV file with the following content, which includes the date and corresponding temperature:

Save the CSV file "temperature.csv"

```
Date, Temperature
2023-01-01,15
2023-01-02.18
2023-01-03,20
2023-01-04,17
2023-01-05,16
2023-01-06.19
2023-01-07,21
2023-01-08.16
2023-01-09.17
2023-01-10.18
```

Saving the CSV File

- Save as temperature_data.csv
- Ensure extension is .csv (not .txt)
- **Uploading to Google Colab**
- Click folder icon → Upload (up arrow)
- Select temperature_data.csv to upload

Using Pandas

```
data =
pd.read_csv("temperature_data.csv")
```

Plotting the Data

- Use plt.plot() to create a line chart
- Customize with:
 - ❖ Title → "Daily Temperature Over a Month"
 - **❖ X-axis** → Date (rotated 45° for clarity)
 - **❖ Y-axis** → Temperature
 - ❖ plt.tight_layout() → ensures labels fit properly
- Output: Line chart showing daily temperature trends
- Helps identify patterns and anomalies for deeper analysis/forecasting

```
plt.plot(df['Date'], df['Temperature'])
plt.title('Daily Temperature Over a Month')
plt.xlabel('Date')
plt.ylabel('Temperature')
plt.xticks(rotation=45) # Rotates the x-axis
labels by 45 degrees
plt.tight_layout() # Adjusts the layout so
everything fits in the figure
plt.show()
```

```
plt.plot(df['Date'], df['Temperature'])
plt.title('Daily Temperature Over a Month')
plt.xlabel('Date')
plt.ylabel('Temperature')
plt.xticks(rotation=45) # This will rotate the x-axis labels by 45 degrees
plt.tight layout() # This will adjust the layout so everything fits in the figure
plt.show()
                       Daily Temperature Over a Month
   21
   20
   18
   17
   16
```

5. Using Scatter Plots in Google Colab

Understanding Scatter Plots

- Display relationship between two variables using Cartesian coordinates
- X-axis = one variable, Y-axis = another
- Each point = data observation
- Example:
 - City planner analyzing population density vs. number of parks
 - Scatter plot reveals possible positive correlation

Why Scatter Plots Matter

- Identify correlations
 (positive, negative, or none)
- Spot trends (increases, decreases, fluctuations)
- Detect outliers or anomalies in data

5. Using Scatter Plots in Google Colab

Go to this link:

https://github.com/MikkoDT/MexEE402_AI/tree/main/Python_Visualization/Scatter_Plot