Pharos University in
Alexandria Faculty of
Computer Science &
Artificial Intelligence Level1 – Spring 2024/2025



Intro to data science Course Code: DS 102 Lecturer : Dr. Shereen

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Data science final project description

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Dataset Details:

Dataset Name: [iris.csv]

Source: [Kaggle]

Description:

This dataset was retrieved from Kaggle. It contains 150 records about flowers and

SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species

Data Exploration

Number of rows: 150Number of columns: 6

- No missing values or duplicates.

The dataset was loaded using the **Pandas** library to explore its structure and contents. The following steps were performed during the exploration phase:

- The data was read from a CSV file named "iris.csv".
- We used the head() function to preview the first few rows of the dataset and understand the format of each column.
- We used shape and info() to inspect the number of rows and columns, as well as the data types and presence of any null values.
- We discovered that some columns like "species", "SepalLengthCm"

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- We dropped the "ID" column, as it doesn't provide useful information for analysis.
- After processing, the dataset was ready for modeling with all features in numeric format.

Data Processing:

During data processing, we focused on handling outliers and encoding categorical data:

- To identify species outliers, we calculated the Interquartile Range (IQR) as follows:
- Calculated Q1 (25th percentile) and Q3 (75th percentile).
- \circ Computed IQR = Q3 Q1.
- Defined lower and upper bounds to detect potential outliers.
- The detected outliers were inspected to decide whether to retain or remove them.

Data Modeling – Linear Regression or SVC

To predict **species**, we chose **Linear Regression** as our primary model due to its effectiveness in handling numerical data prediction tasks.

- The model was trained using the training portion of the dataset with the following steps:
 - The data was split into training and testing sets (80% for training, 20% for testing).
 - A **Linear Regression** model or super vector classiver was fitted using the training data.
 - We used the trained model to predict car prices from the test set and stored the results in a variable called y_pred.
 - We calculated the error by comparing the **actual vs. predicted** price values using metrics like **Mean Squared Error** (**MSE**) and **R**² **score** to evaluate model performance.

Data Visualization:

We used various visualization techniques to explore the relationships
import matplotlib.pyplot as plt
import plotly.express as px
import seaborn as sns
between variables:
Scatter Plot
Pie plot
Box plot
Histogram
Line
Bar