

# Exploratory Data Analysis: Iris dataset

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# Iris Dataset

**iris setosa**



petal

sepal

**iris versicolor**



petal

sepal

**iris virginica**



petal

sepal

# Data Summary:

- Number of Samples: 150
- Number of Features: 4
- Target Classes: Setosa, Versicolor, Virginica

# Organization of Dataset

The **data matrix** refers to the array of numbers

$$\mathbf{X} = \begin{pmatrix} x_{11} & x_{12} & \cdots & x_{1p} \\ x_{21} & x_{22} & \cdots & x_{2p} \\ x_{31} & x_{32} & \cdots & x_{3p} \\ \vdots & \vdots & \ddots & \vdots \\ x_{n1} & x_{n2} & \cdots & x_{np} \end{pmatrix}$$

where  $x_{ij}$  is the  $j$ -th variable collected from the  $i$ -th item (e.g., subject).

- items/subjects are rows
- variables are columns

$\mathbf{X}$  is a data matrix of order  $n \times p$  (# items by # variables).

# Collection of Column Vectors

We can view a data matrix as a collection of **column vectors**:

$$\mathbf{X} = \begin{pmatrix} | & | & \cdots & | \\ \mathbf{x}_1 & \mathbf{x}_2 & \cdots & \mathbf{x}_p \\ | & | & & | \end{pmatrix}$$

where  $\mathbf{x}_j$  is the  $j$ -th column of  $\mathbf{X}$  for  $j \in \{1, \dots, p\}$ .

The  $n \times 1$  vector  $\mathbf{x}_j$  gives the  $j$ -th variable's scores for the  $n$  items.

# Collection of row vectors

We can view a data matrix as a collection of **row vectors**:

$$\mathbf{X} = \begin{pmatrix} \text{---} & \mathbf{x}'_1 & \text{---} \\ \text{---} & \mathbf{x}'_2 & \text{---} \\ & \vdots & \\ \text{---} & \mathbf{x}'_n & \text{---} \end{pmatrix}$$

where  $\mathbf{x}'_i$  is the  $i$ -th row of  $\mathbf{X}$  for  $i \in \{1, \dots, n\}$ .

The  $1 \times p$  vector  $\mathbf{x}'_i$  gives the  $i$ -th item's scores for the  $p$  variables.

# Exploratory Data Analysis: Iris dataset

- To read Iris dataset (iris.csv) the easiest way is to use data frame of pandas library in python.
- Import following libraries of python:
  - `import os`
  - `import pandas as pd`
  - `import numpy as np`
  - `from matplotlib import pyplot as plt`
  - `Import seaborn as sns`
- Keep your python program and dataset (iris.csv) in a same folder.
- Load the Iris dataset from a local drive.
- `data_path = os.getcwd()`
- `data = pd.read_csv(os.path.join(data_path, 'iris.csv'))`

- #Add header in the dataset
- `data.columns = ['Sepal_Length', 'Sepal_Width', 'Petal_Length', 'Petal_Width', 'Species']`
- #Gaining information from data
- `iris_df.info()`
- #We need to know the overall statistical information of the dataset
- `iris_df.describe()`
- Checking the distribution of each species in the dataset
- `iris_df['Species'].value_counts()`
- `sns.countplot(iris_df['Species'])`
- `plt.title('Species Count')`



# Covariance of Dataset

The **covariance matrix** refers to the symmetric array of numbers

$$\mathbf{S} = \begin{pmatrix} s_1^2 & s_{12} & s_{13} & \cdots & s_{1p} \\ s_{21} & s_2^2 & s_{23} & \cdots & s_{2p} \\ s_{31} & s_{32} & s_3^2 & \cdots & s_{3p} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ s_{p1} & s_{p2} & s_{p3} & \cdots & s_p^2 \end{pmatrix}$$

where

- $s_j^2 = (1/n) \sum_{i=1}^n (x_{ij} - \bar{x}_j)^2$  is the **variance** of the  $j$ -th variable
- $s_{jk} = (1/n) \sum_{i=1}^n (x_{ij} - \bar{x}_j)(x_{ik} - \bar{x}_k)$  is the **covariance** between the  $j$ -th and  $k$ -th variables
- $\bar{x}_j = (1/n) \sum_{i=1}^n x_{ij}$  is the mean of the  $j$ -th variable

# Correlation of Dataset

The **correlation matrix** refers to the symmetric array of numbers

$$\mathbf{R} = \begin{pmatrix} 1 & r_{12} & r_{13} & \cdots & r_{1p} \\ r_{21} & 1 & r_{23} & \cdots & r_{2p} \\ r_{31} & r_{32} & 1 & \cdots & r_{3p} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ r_{p1} & r_{p2} & r_{p3} & \cdots & 1 \end{pmatrix}$$

where

$$r_{jk} = \frac{s_{jk}}{s_j s_k} = \frac{\sum_{i=1}^n (x_{ij} - \bar{x}_j)(x_{ik} - \bar{x}_k)}{\sqrt{\sum_{i=1}^n (x_{ij} - \bar{x}_j)^2} \sqrt{\sum_{i=1}^n (x_{ik} - \bar{x}_k)^2}}$$

is the Pearson correlation coefficient between variables  $\mathbf{x}_j$  and  $\mathbf{x}_k$ .