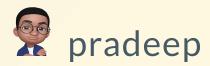
# **Basic Data Mining with Python**



## What is Data Mining?

" Data Mining is a process of extracting insights from data.





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# Technology/Tools

Programming Language

• Python, R

#### Software

Weka, RapidMiner, Excel

#### Cloud

• R Studio Cloud, Power BI, Tableau, Google Collab





## What problem to solve?

Classification - Supervised Learning

Clustering - Unsupervised





# Methodology

Apa tu





### What is Data?





# Python

#### What

- High-Level Programming Language.
- Emphasizes on code readibilty.
- Rank = 1\* for 2021 \*(IEEE Spectrum)
- Consist of fantastic libraries!





## **Python Libraries**

A Python library is a collection of related modules. It makes Python Programming simpler and convenient for the programmer.

# Pandas
import pandas as pd





# **Reading Data**

Usually we can use pandas library. Pandas store the imported data as DataFrame.

```
# Default sep = ','
df = pd.read_csv("iris_dirty.csv")
```

or if you want to use another separator, simply add sep='\t'

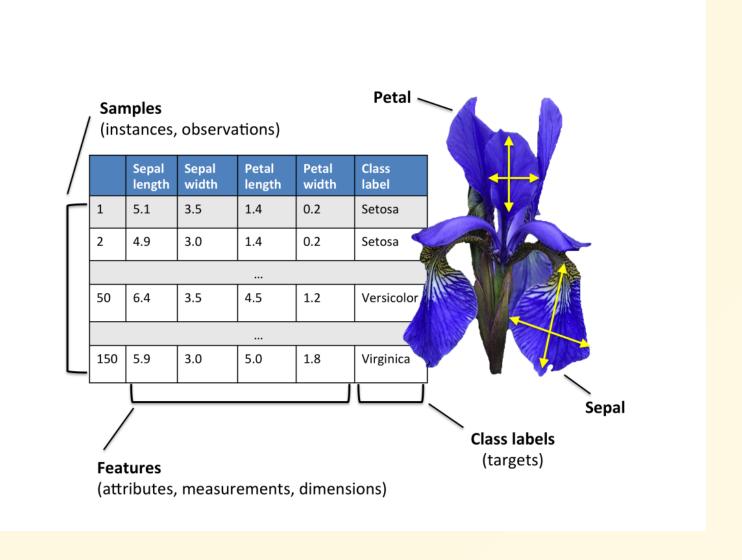
```
df = pd.read_csv("file_name.csv", sep = '\t')
```





#### **Iris Dataset**

Fisher's Iris data set introduced by Ronald Fisher in his 1936 paper.







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#### **View Data**

You can have a look at the first five rows with .head():

```
# by default is 5 rows
df.head()
# you can also customize the #-rows
df.head(10)
```

or the last five rows with .tail():

```
df.tail()
```





#### **Data Info**

The shape property returns a tuple representing the dimensionality of the DataFrame.

df.shape

The info() method prints information about the DataFrame.

df.info





## **Statistical Description**

All standard statistical operations are present in Pandas:

```
# Show the statistical summary on the numerical columns
df.describe()
# or individually
df.mean()
```

```
# Show the statistical summary on the categorical columns
df.describe(include = 'object')
```





#### **Finding Missing Values**

It is common to have not-a-number (NaN) values in your data set.

```
# Will give the total number of NaN in each column
df.isna().sum()
```





#### **Handling Missing Values**

```
# Remove the rows with NaN, not recommended df.dropna()
```

```
# fill NaN with 0, also not recommended
df.fillna(0)

# fill NaN with mean, better
df1 = df.fillna(df.mean(numeric_only=True))
```





#### **Problematic Values**

Typo can be considered problematic.

```
# Count unique categorical values
df1.Species.unique()
df1['Species'].value_counts()

# View problematic values
df1.iloc[[7]]
```





**Handling Problematic Values** 

We can replace with the correct value using replace()

```
df2 = df1.replace(['SETSA'], 'setosa')
```

Cleaning done!

Check out my Kaggle post for more data cleaning example.





#### **Data Visualization**

#### Why?

- Visualizing data prior to analysis is a good practice.
- Statistical description do not fully depict the data set in its entirety.

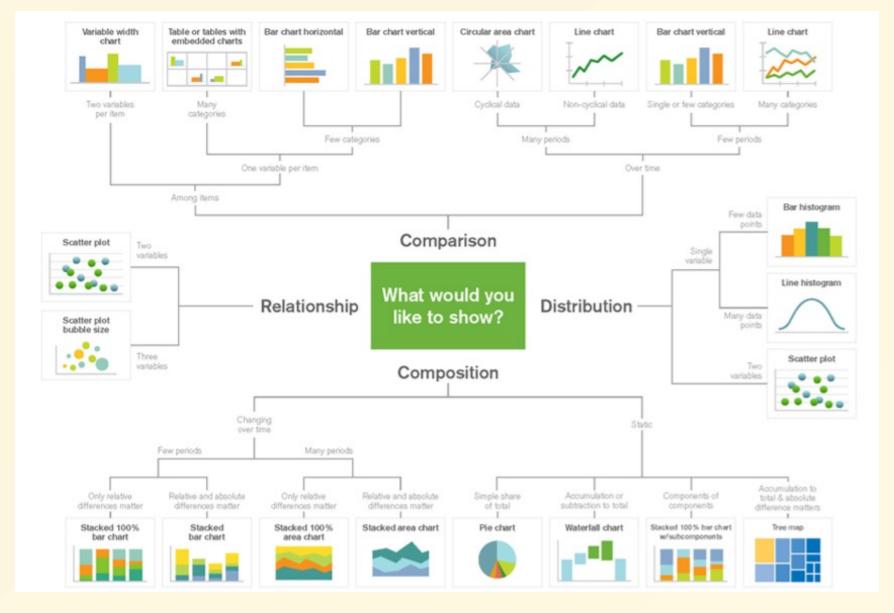
Check out my video \*HERE explaining the importance of visualizing data when analyzing it.

\*promo syok sendiri





# **Cheat Sheet**







#### **Data Visualization**

#### Scatter plot

```
df2.plot.scatter(x = 'Petal.Length', y = 'Petal.Width')
```

#### Using colour as third variable

```
# Dictionary mapping colour with categorical values
colors = {'setosa':'red','virginica':'blue','versicolor':'green'}
df2.plot.scatter(x = 'Petal.Length', y = 'Petal.Width', c = df2['Species'].map(colors))
```





## **Machine Learning**

scikit-learn

```
df2.plot.scatter(x = 'Petal.Length', y = 'Petal.Width')
```



