

DATA PREPERATION

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- 1.Data Cleaning
- 2. Categorical Data
- 3. Normalisation/Standarisation
- 4. Feature Selection

PLAN

DATA CLEANING • MISSING VALUES

```
Entrée [1]: dt.isnull().sum
```

DATA CLEANING

• DELETING THE MISSING VALUES

```
Entrée [6]: dt.dropna(axis=1)#cols

dt.dropna(axis=0)#rows
```

DATA CLEANING

• REMPLACING THE MISSING VALUES

```
Entrée [7]: #remplacing the missing values with 0
    dt.fillna(0)
    #remplacing the missing values with the mean
    dt.fillna(dt.mean())
    #remplacing the missing values with the mode
    dt.fillna(dt.mode())
```

DATA CLEANING • DUPLICATED ROWS

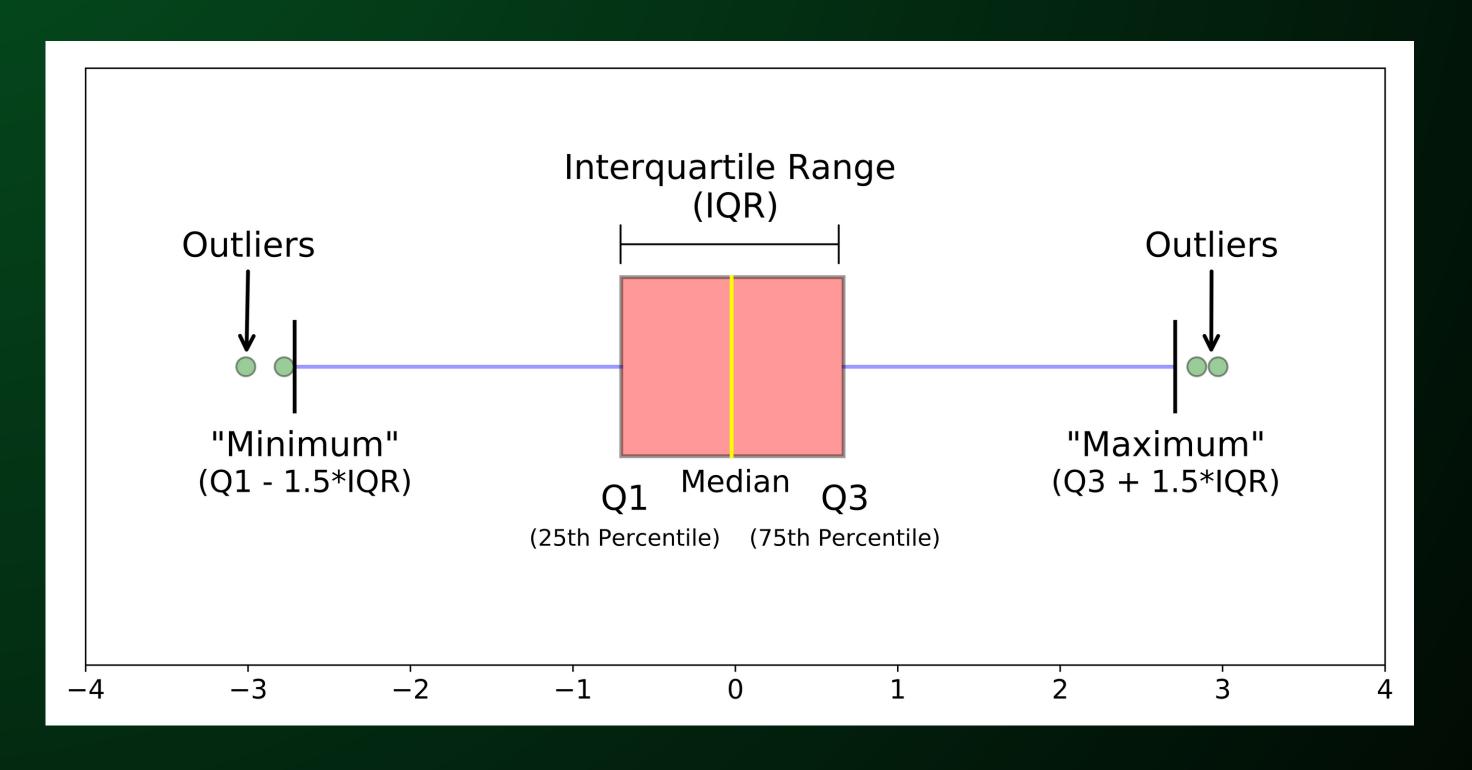
```
Entrée [13]: True in list(df.duplicated())
Out[13]: True
Entrée [16]: df = df.drop_duplicates()
```

DATA CLEANING

• OUTLIERS



DATA CLEANING • DELETING THE OUTLIERS



```
Entrée [8]: import pandas as pd

def remove_outliers(dt, column_name):
    data = dt[column_name]

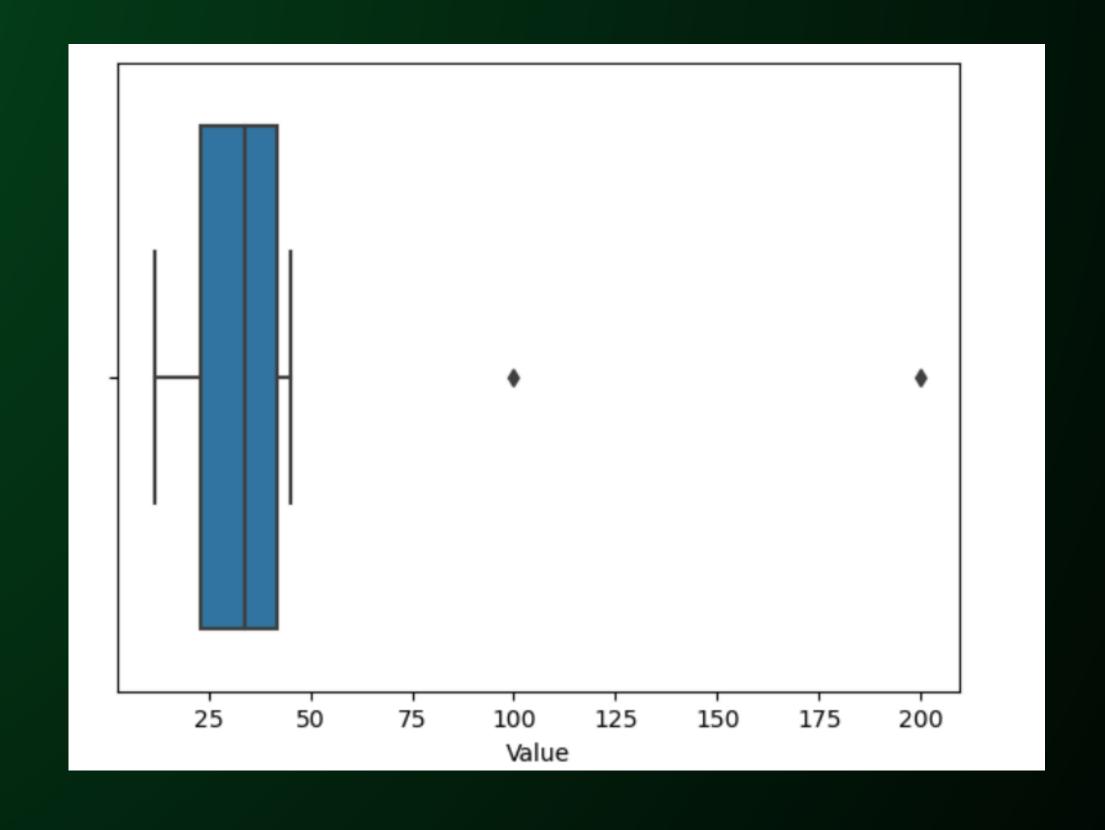
    q1 = data.quantile(0.25)
    q3 = data.quantile(0.75)
    iqr = q3 - q1

    lower_bound = q1 - 1.5 * iqr
    upper_bound = q3 + 1.5 * iqr

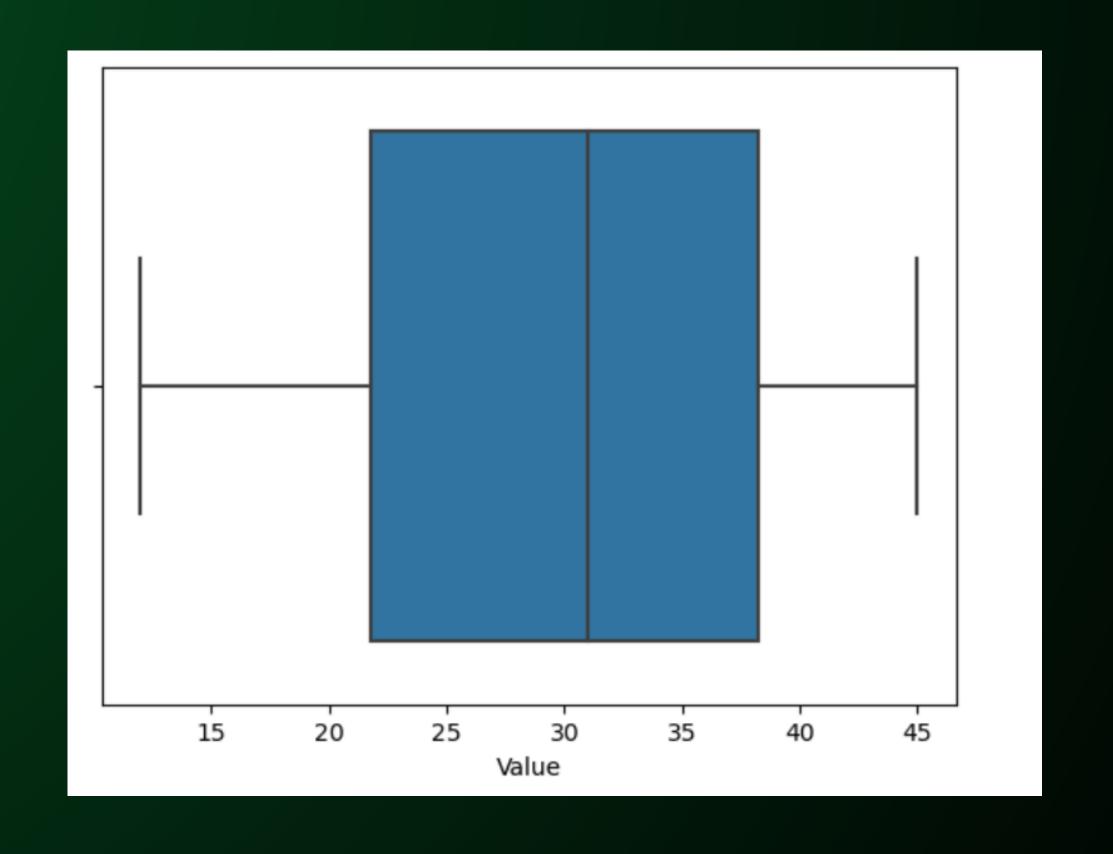
    filtered_dt = dt[(dt[column_name] >= lower_bound) & (dt[column_name] <= upper_bound)]

    return filtered_dt</pre>
```

	Value	
0	12	
1	17	
2	21	
3	24	
4	30	
5	32	
6	36	
7	39	
8	41	
9	45	
10	100	
11	200	



	Value
0	12
1	17
2	21
3	24
4	30
5	32
6	36
7	39
8	41
9	45



CATEGORICAL DATA

• LABEL ENCODER

```
Entrée [20]: data = {'class': ['a', 'b', 'b', 'c', 'a', 'e']}
             dt = pd.DataFrame(data)
Entrée [21]: dt
   Out[21]:
                 class
                    b
                    b
                    a
```

CATEGORICAL DATA

• LABEL ENCODER

```
Entrée [22]: from sklearn.preprocessing import LabelEncoder
             le = LabelEncoder()
             dt['class'] = le.fit_transform(dt['class'])
Entrée [23]: dt
   Out[23]:
                class
```

CATEGORICAL DATA

DUMMIES

```
Entrée [30]: one_hot_encoded = pd.get_dummies(dt['class'])
            df_encoded = pd.concat([dt, one_hot_encoded], axis=1)
            df_encoded
   Out[30]:
                  a 1 0 0 0
                  b 0 1 0 0
                  a 1 0 0 0
                  e 0 0 0 1
```

NORMALISATION / STANDARISATION

NORMALISATION

$$x_{\text{norm}} = \frac{x - \min(x)}{\max(x) - \min(x)}$$

STANDARISATION

$$Z=rac{x-\mu}{\sigma}$$

```
Entrée [1]: from sklearn.preprocessing import MinMaxScaler
          import numpy as np
          data = np.array([[100, 0.001],
                         [8, 0.05],
                         [50, 0.005],
                         [88, 0.07],
                         [4, 0.1]])
          min_max_scaler = MinMaxScaler()
          normalized_data = min_max_scaler.fit_transform(data)
Entrée [2]: normalized_data
   Out[2]: array([[1.
                         , 0.
                [0.04166667, 0.49494949],
                [0.47916667, 0.04040404],
                                             Entrée [3]: from sklearn.preprocessing import StandardScaler
                [0.875 , 0.6969697 ],
                [0. , 1. ]])
                                                           standard scaler = StandardScaler()
                                                           standardized data = standard scaler.fit transform(data)
                                             Entrée [4]: standardized_data
                                                  Out[4]: array([[ 1.26398112, -1.16389967],
                                                                  [-1.06174414, 0.12639634],
                                                                   [ 0. , -1.05856939],
                                                                   [ 0.96062565, 0.65304778],
                                                                   [-1.16286263, 1.44302493]])
```

FEATURE SELECTION

CORRELATION COEFFICIENT

$$\rho(X,Y) = \frac{\operatorname{Cov}(X,Y)}{\sqrt{\operatorname{Var}(X)}\sqrt{\operatorname{Var}(Y)}}$$

Corrélation	Négative	Positive
Faible	de -0,5 à 0,0	de 0,0 à 0,5
Forte	de −1,0 à −0,5	de 0,5 à 1,0

FEATURE SELECTION

CORRELATION MATRIX

SOIT XI (∀ I ∈ {1,...,N}) DES VARIABLES, LA MATRICE DE CORRÉLATION EST LA MATRICE R DÉFINIT PAR:

$$\forall i,j \in \{1,...,n\}$$

TRANSFORMATION DES DONNÉES

SÉLECTION DES CARACTÉRISTIQUES

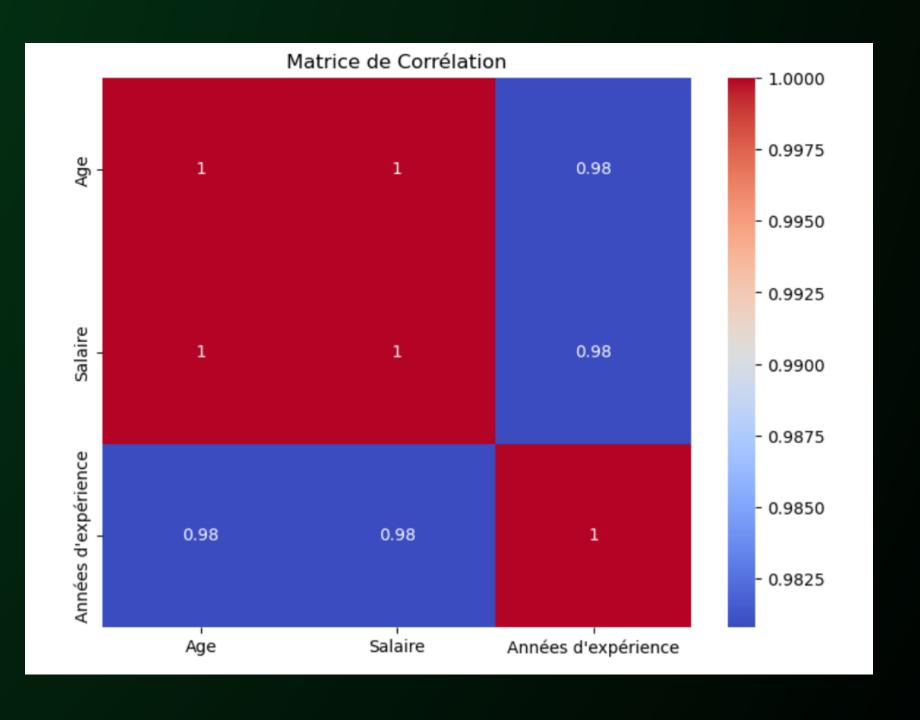
• MATRICE DE CORRÉLATION

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

data = {
    'Age': [25, 30, 35, 40, 45],
    'Salaire': [30000, 40000, 50000, 60000, 70000],
    'Années d\'expérience': [2, 3, 7, 10, 15]
}

df = pd.DataFrame(data)
    correlation_matrix = df.corr()

plt.figure(figsize=(8, 6))
    sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')
    plt.title('Matrice de Corrélation')
    plt.show()
```



YALAH JBDO DB DOK PCS