Random Forest

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In this project, random forest algorithm is used to train the model. Feature importance is plotted for the analysis of the dataset.

Dataset: Breast Cancer Wisconsin

Source: https://www.kaggle.com/uciml/breast-cancer-wisconsin-data

Features: ID Number, Clump Thickness, Uniformity of Cell Size, Uniformity of Cell Shape, Marginal Adhesion, Single Epithelial Cell Size, Bare Nuclei, Bland Chromatin, Normal Nucleoli, Mitoses

Label: Class

0.1 Preliminary works

1. Import libraries

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

from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestRegressor
```

2. Get a dataset

```
[]: from google.colab import files data_to_load = files.upload()
```

<IPython.core.display.HTML object>

Saving breast-cancer-wisconsin.csv to breast-cancer-wisconsin.csv

3. Create train, test and validation sets

```
[]: data = pd.read_csv("breast-cancer-wisconsin.csv").dropna()
data.columns = ['ID Number', 'Clump Thickness', 'Uniformity of Cell Size',

→'Uniformity of Cell Shape', 'Marginal Adhesion', 'Single Epithelial Cell Size'

→, 'Bare Nuclei', 'Bland Chromatin', 'Normal Nucleoli', 'Mitoses', 'Class']
```

0.2 Random Forest algorithm

```
[]: rf = RandomForestRegressor(n_estimators=100)
    rf.fit(X_train, Y_train)
    Y_pred = rf.predict(X_test)
    print("%.4f" %rf.score(X_train, Y_train))
    print("%.4f" %rf.score(X_test, Y_test))
```

0.9832 0.8521

Feature importance is plotted for the analysis.

```
[]: def plot_feature_importance(importance,names,model_type):
       #Create arrays from feature importance and feature names
       feature_importance = np.array(importance)
       feature_names = np.array(names)
       #Create a DataFrame using a Dictionary
       data={'feature_names':feature_names,'feature_importance':feature_importance}
       df = pd.DataFrame(data)
       #Sort the DataFrame in order decreasing feature importance
       df.sort_values(by=['feature_importance'], ascending=False,inplace=True)
       #Define size of bar plot
      plt.figure(figsize=(10,8))
       #Plot Searborn bar chart
       sns.barplot(x=df['feature_importance'], y=df['feature_names'])
       #Add chart labels
      plt.title(model_type + 'FEATURE IMPORTANCE')
      plt.xlabel('FEATURE IMPORTANCE')
      plt.ylabel('FEATURE NAMES')
```

```
[]: plot_feature_importance(rf.feature_importances_,X_train.columns,'RANDOM FOREST')
```

