

Human Activity Recognition using Smartphone Data with Machine Learning

Importing necessary libraries for the project

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
```

Reading the data-set

```
train = pd.read_csv("train-1.csv")
test = pd.read_csv("test.csv")
```

```
train['Data'] = 'Train'
test['Data'] = 'Test'
both = pd.concat([train, test], axis=0).reset_index(drop=True)
both['subject'] = '#' + both['subject'].astype(str)
```

```
train.shape, test.shape
```

```
((7352, 564), (2947, 564))
```

```
both.head()
```

| | tBodyAcc- mean()-X | tBodyAcc- mean()-Y | tBodyAcc- mean()-Z | tBodyAcc- std()-X | tBodyAcc- std()-Y | tBodyAcc- std()-Z | tBodyAcc- mad()-X | tBodyAcc- mad()-Y |
|---|-----------------------|-----------------------|-----------------------|----------------------|----------------------|----------------------|----------------------|----------------------|
| 0 | 0.288585 | -0.020294 | -0.132905 | -0.995279 | -0.983111 | -0.913526 | -0.995112 | -0.995112 |
| 1 | 0.278419 | -0.016411 | -0.123520 | -0.998245 | -0.975300 | -0.960322 | -0.998807 | -0.998807 |
| 2 | 0.279653 | -0.019467 | -0.113462 | -0.995380 | -0.967187 | -0.978944 | -0.996520 | -0.996520 |
| 3 | 0.279174 | -0.026201 | -0.123283 | -0.996091 | -0.983403 | -0.990675 | -0.997099 | -0.997099 |
| 4 | 0.276629 | -0.016570 | -0.115362 | -0.998139 | -0.980817 | -0.990482 | -0.998321 | -0.998321 |

```
5 rows × 564 columns
```

```
both.dtypes.value_counts()
```

```
float64    561
object      3
```

dtype: int64

```
def basic_details(df):
    b = pd.DataFrame()
    b['Missing value'] = df.isnull().sum()
    b['N unique value'] = df.nunique()
    b['dtype'] = df.dtypes
    return b
```

```
basic_details(both)
```

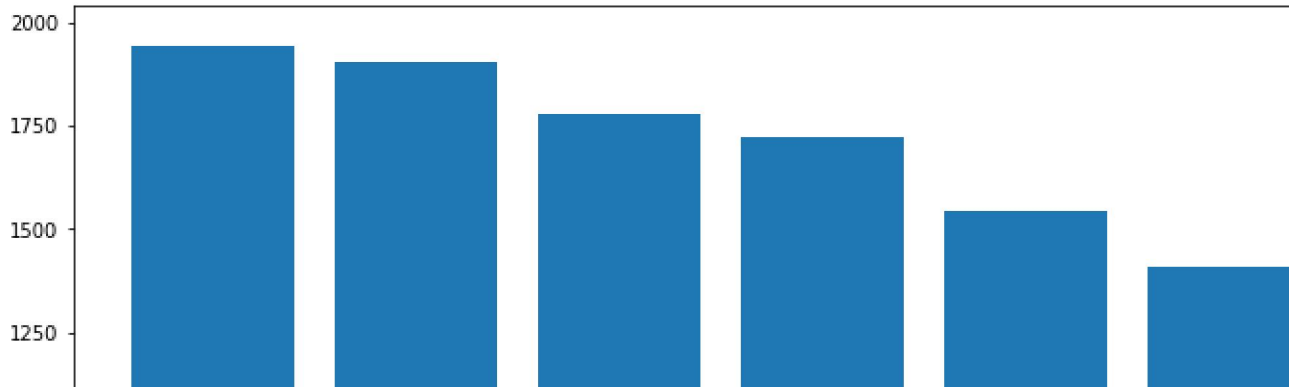
| | Missing value | N unique value | dtype |
|-----------------------------|---------------|----------------|---------|
| tBodyAcc-mean()-X | 0 | 10292 | float64 |
| tBodyAcc-mean()-Y | 0 | 10299 | float64 |
| tBodyAcc-mean()-Z | 0 | 10293 | float64 |
| tBodyAcc-std()-X | 0 | 10295 | float64 |
| tBodyAcc-std()-Y | 0 | 10297 | float64 |
| ... | ... | ... | ... |
| angle(Y,gravityMean) | 0 | 10299 | float64 |
| angle(Z,gravityMean) | 0 | 10299 | float64 |
| subject | 0 | 30 | object |
| Activity | 0 | 6 | object |
| Data | 0 | 2 | object |

564 rows × 3 columns

```
activity = both['Activity']
label_counts = activity.value_counts()

plt.figure(figsize= (12, 8))
plt.bar(label_counts.index, label_counts)
```

<BarContainer object of 6 artists>



```
Data = both['Data']
Subject = both['subject']
train = both.copy()
train = train.drop(['Data','subject','Activity'], axis =1)
```

Scaling the data

```
# Standard Scaler
from sklearn.preprocessing import StandardScaler
slc = StandardScaler()
train = slc.fit_transform(train)
```

```
# dimensionality reduction
from sklearn.decomposition import PCA
pca = PCA(n_components=0.9, random_state=0)
train = pca.fit_transform(train)
```

Splitting data for training and testing

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(train, activity, test_size = 0.2, rand
```

```
num_folds = 10
seed = 0
scoring = 'accuracy'
results = {}
accuracy = {}
```

Algorithm to recognize human activity

```
# Finalizing the model and comparing the test, predict results
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import confusion_matrix, accuracy_score, classification_report
from sklearn.model_selection import KFold, cross_val_score
model = KNeighborsClassifier(algorithm= 'auto', n_neighbors= 8, p= 1, weights= 'distance')
```

```

_ = cross_val_score(model, X_train, y_train, cv=10, scoring=scoring)
results["GScv"] = (_.mean(), _.std())

model.fit(X_train, y_train)
y_predict = model.predict(X_test)

accuracy["GScv"] = accuracy_score(y_test, y_predict)

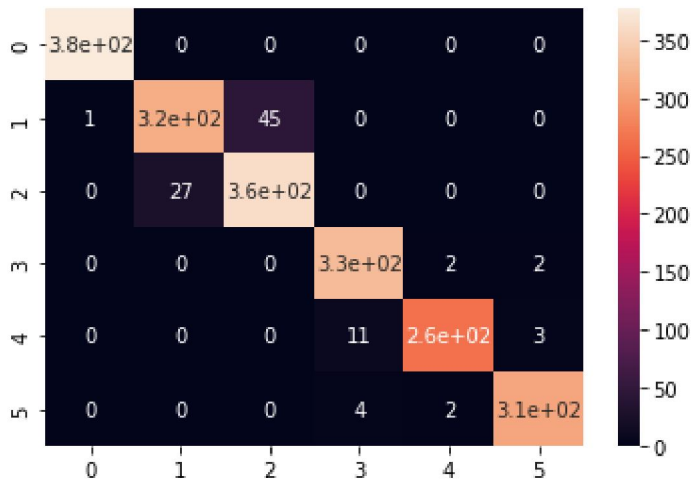
print(classification_report(y_test, y_predict))

cm= confusion_matrix(y_test, y_predict)
sns.heatmap(cm, annot=True)

```

| | precision | recall | f1-score | support |
|--------------------|-----------|--------|----------|---------|
| LAYING | 1.00 | 1.00 | 1.00 | 377 |
| SITTING | 0.92 | 0.87 | 0.90 | 364 |
| STANDING | 0.89 | 0.93 | 0.91 | 390 |
| WALKING | 0.96 | 0.99 | 0.97 | 335 |
| WALKING_DOWNSTAIRS | 0.99 | 0.95 | 0.97 | 278 |
| WALKING_UPSTAIRS | 0.98 | 0.98 | 0.98 | 316 |
| accuracy | | | 0.95 | 2060 |
| macro avg | 0.96 | 0.95 | 0.95 | 2060 |
| weighted avg | 0.95 | 0.95 | 0.95 | 2060 |

<matplotlib.axes._subplots.AxesSubplot at 0x7f22739fcfd0>



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