

import pandas as pd

import warnings

warnings.filterwarnings('ignore')

from sklearn.metrics import accuracy\_score, classification\_report

from sklearn.ensemble import GradientBoostingClassifier, AdaBoostClassifier,

from sklearn.linear\_model import LogisticRegression CV.

from sklearn.model\_selection import train\_test\_split, GridSearchCV.

from sklearn.tree import DecisionTreeClassifier

from sklearn.preprocessing import LabelEncoder.

Step: 1

Original = pd.read\_csv("Human-Activity-Data.csv")

Original.head()

Original.columns

Original.shape

Original.dtypes

Original.info()

Original.value\_counts()

label\_encoder = LabelEncoder()

Original["label-Activity"] = label\_encoder.fit\_transform(Original["Activity"])

Step: 2

Original.Activity.value\_counts()

Original.label-Activity.value\_counts()

## Lab10. Patients Physical Activities Prediction using Boosting

### Objectives

In this lab, you will recognize physical activities such as 'laying', 'sitting' or 'walking' using Gradient Boosting, AdaBoost and VotingClassifiers.

### Learning Outcomes

After completing this lab, you will be able to

- Create a small dataset with selected rows based on fewer target labels
- Build GradientBoostingClassifier, fit and predict on test data
- Print accuracy and classification report
- Find the best no. of decision trees and learning rate using GridSearch and Cross Validation
- Build AdaBoost classifier model with GridSearchCV, fit and predict
- Select best parameter values for n\_estimators and learning\_rate
- Build LogisticRegressionCV model, fit, predict and print scores
- Build VotingClassifier using other models, fit, predict and print scores
- Interpret results and parameter values
- Change parameter values and play around with models

### Business Use Case

As you a data scientist, one popular hospital in your city has asked you to design a model based on the mobile phone data of their patients. Your model will help the hospital to understand the physical activity level of their patients. The physical activities are classified into 3 levels - Laying, Sitting and Walking. Based on the prediction, Doctors will recommend health diets, exercise and physical activities to the new patients.

### Dataset

You will use a dataset named, *Human\_Activity\_Data.csv*.

#### Step1. [Understand Data]

- Using Pandas, import "*Human\_Activity\_Data.csv*" file and print properties such as head, shape, columns, dtype, info and value\_counts.

#### Step2. [Build a small dataset]

- As it is a big dataset, execution time will be long for training and testing. So build a small dataset with only 3 classes, laying, sitting and walking, where each class with 500 samples. So, shape of this new dataframe will be (1500, 563). That is 1500 rows and 563 features.
- Store this new dataframe as a CSV file.

#### Step3. [ Build GradientBoostingClassifier]

- Import your new reduced CSV file
- Print basic properties of the new CSV file
- Split it into training set and test set (30% samples for testing)
- Create GradientBoostingClassifier, fit and predict
- Print accuracy and classification report

#### Step4. [Find Best no. of trees and Best Learning Rate using Grid Search and Cross Validation]

- Create GridSearchCV model with GradientBoostingClassifier
- Parameters: `param_grid = {'n_estimators': [50, 100, 200, 400], 'learning_rate': [0.1, 0.01]}`
- Perform fit and predict
- Print accuracy, classification report
- Print best parameters such as best no. of trees and learning rate. Use the attribute `best_estimator_`

Take first 3000 samples:

tem1 = Original [Original ['Activity'] == 'LAYING'] [:500]

tem2 = Original [Original [Original ['Activity'] == 'SITTING'] [:500]]

tem3 = Original [Original ['Activity'] == 'WALKING'] [:500]

tem4 = Original [Original ['Activity'] == 'STANDING'] [:500]

tem5 = Original [Original ['Activity'] == 'WALKING-UPSTAIRS'] [:500]

tem6 = Original [Original ['Activity'] == 'WALKING-DOWNSTAIRS'] [:500]

new\_df = pd.concat([tem1, tem2, tem3, tem4, tem5, tem6])

new\_df.to\_csv('human-activity-clipped 3000.csv')

new\_df.head()

new\_df.shape

new\_df.columns

new\_df.dtypes

new\_df.value\_counts()

X = new\_df.drop(['Activity', 'label-Activity'], axis=1)

y = new\_df['label-Activity']

X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=0.2)

Build Gradient Boosting:

model = GradientBoostingClassifier(subsample=0.5, n\_estimators=100)

model.fit(X\_train, y\_train)

y\_pred = model.predict(X\_test)

print(accuracy\_score(y\_test, y\_pred))

print(classification\_report(y\_test, y\_pred))

**Step5. [Build AdaBoostClassifier]**

- Create AdaBoostClassifier with DecisionTreeClassifier
- Create GridSearchCV with AdaBoostClassifier model that you created as before
- Parameters: `param_grid = {'n_estimators': [100, 150, 200], 'learning_rate': [0.01, 0.001]}`
- Perform fit, predict
- Print accuracy, classification report
- Print best parameters such as best no. of trees and learning rate. Use the attribute `best_estimator_`

**Step6. [Build LogisticRegressionCV classifier]**

- Create a LogisticRegressionCV model with the parameters `Cs=5, cv=4, penalty='l2'`.
- Perform fit and predict
- Print classification report

**Step7. [ Build VotingClassifier]**

- Build VotingClassifier model with GradientBoostingClassifier and LogisticRegressionCV that you created in the previous steps
- Perform fit and predict operations
- Print classification report

**Step8. [ Interpret your results]**

- Analyze your results
- Change parameters and play with your code

```

base = DecisionTreeClassifier(max_features=14)
base2 = AdaBoostClassifier(base_estimator=base, random_state=0)
param_grid = {'n_estimators': [100, 150, 200], 'learning_rate': [0.01, 0.001]}
model1 = GridSearchCV(base2, param_grid, cv=10, n_jobs=-1)
model2 = model1.fit(X_train, y_train)
y_pred2 = model2.predict(X_test)
print(accuracy_score(y_test, y_pred2))
print(classification_report(y_test, y_pred2))
model2.best_estimator_

Build Logistic Regression:
model3 = LogisticRegressionCV(cv=5, Cs=10, penalty='l2')
model3.fit(X_train, y_train)
y_pred3 = model3.predict(X_test)

```

Print (accuracy\_score(y\_test, y\_pred3))

print (classification\_report(y\_test, y\_pred3))

Find Best No. of trees:

param\_grid = {'n\_estimators': [50, 100, 200, 400],  
'learning\_rate': [0.1, 0.01]}

all\_scores = Cross\_val\_score(estimator= model\_3, X=X\_train,  
y=y\_train, cv=5)

print(all\_scores)

model\_4 = GridSearchCV(estimator= model\_3, param\_grid=param  
\_grid, cv=5, n\_jobs=-1).

model\_4 = fit(X\_train, y\_train)

y\_pred\_4 = model\_4.predict(X\_test)

print (accuracy\_score(y\_test, y\_pred\_4))

print (Classification\_report(y\_test, y\_pred\_4))

model\_4 = best\_estimator\_

Build Voting classifier for 3000 samples:

model\_5 = VotingClassifier(estimators = [('lr', model\_3),

('gbc', gbc)], voting='soft',

model\_5 = fit(X\_train, y\_train)

y\_pred\_5 = model\_5.predict(X\_test)

Print (accuracy\_score(y\_test, y\_pred\_5))

Print (classification\_report(y\_test, y\_pred\_5))

From this 3000 samples you should:

temp1 = new\_df[new\_df['Activity'] == 'LAYING'][:500]

temp2 = new\_df[new\_df['Activity'] == 'SLEEPING'][:500]

**NOTE**

```
temp3: new_df['Activity'] == ['WALKING'][:500]
```

```
df = pd.concat([temp1, temp2, temp3])
```

```
df.to_csv("human_activity - clipped 1500.csv")
```

```
df = pd.read_csv("human_activity - clipped 1500.csv")
```

```
df.head()
```

```
df.shape
```

```
df.columns
```

```
df.dtypes
```

```
df.info()
```

```
df.value_counts()
```

Step 3 - [Build Gradient Boosting Classifier]

```
X = df.drop(['Activity', 'label_Activity'], axis=1)
```

```
y = df['label_Activity']
```

```
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2)
```

```
model = GradientBoostingClassifier(subsample=0.5, n_estimators=100,
    learning_rate=0.1)
```

```
model.fit(X_train, y_train)
```

```
y_pred = model.predict(X_test)
```

```
y_pred
```

accuracy = score(y\_test, y\_pred)

~~print~~ print(classification\_report(y\_test, y\_pred))

Step: 4

param\_grid = {'n\_estimators': [50, 100, 200, 400], 'learning\_rate': [0.1, 0.01]}

all\_scores = cross\_val\_score(estimator=model, X=X\_train, y=y\_train, cv=5)

print(all\_scores)

model\_2 = GridSearchCV(estimator=model, param\_grid=param\_grid, cv=5, n\_jobs=-1)

print(all\_scores)

model\_2 = GridSearchCV(estimator=model, param\_grid=param\_grid, cv=5, n\_jobs=-1)

model\_2.fit(X\_train, y\_train)

y\_pred\_2 = model\_2.predict(X\_test)

y\_pred\_2

accuracy\_score(y\_test, y\_pred\_2)

model\_2.best\_estimator\_

Step: 5

base = DecisionTreeClassifier(max\_features=4)

base\_2 = AdaBoostClassifier(base\_estimator=base, random\_state=0)

param\_grid = {'n\_estimators': [100, 150, 200], 'learning\_rate': [0.01, 0.01]}

model\_3 = GridSearchCV(base\_2, param\_grid, cv=10, n\_jobs=-1)

model\_3.fit(X\_train, y\_train)

y\_pred\_3 = model\_3.predict(X\_test)

y\_pred\_3

accuracy\_score(y\_test, y\_pred\_3)

print(classification\_report(y\_test, y\_pred\_3))

NOTE

model 3- best\_estimator.

Step: 6

model 4 = LogisticRegressionCV (Cv=4, C3=5, penalty='l2')

model 4.fit (X\_train, y\_train)

y\_pred4 = model 4.predict (X\_test)

y\_pred4.

accuracy = score (y\_test, y\_pred4)

print (classification\_report (y\_test, y\_pred4))

Step: 7

model 5 = VotingClassifier (estimators = [( 'lr', model 4),  
( 'gbc', gbc)], voting = 'soft')

model 5.fit (X\_train, y\_train)

y\_pred 5 = model 5.predict (X\_test)

y\_pred 5

accuracy = score (y\_test, y\_pred 5)

print (classification\_report (y\_test, y\_pred 5))

Step: 8

Gradient Boosting Classifier

model 6 = GradientBoostingClassifier (n\_estimators = 50,  
learning\_rate = 1.0,

model 6.fit (X\_train, y\_train)

y\_pred 6 = model 6.predict (X\_test)

y\_pred 6



accuracy\_score(y\_test, y\_pred)

print(ClassificationReport(y\_test, y\_pred))

Ada Boost Classifier:

base 3 = AdaBoostClassifier(base\_estimator=base,

model 7 = GridSearchCV(base 3, param\_grid, cv=5, n\_jobs=-1)

model 7.fit(X\_train, y\_train)

y\_pred 7 = model 7.predict(X\_test)

y\_pred 7.

accuracy\_score(y\_test, y\_pred 7)

print(ClassificationReport(y\_test, y\_pred 7))