# Lab\_10: PATIENTS PHYSICAL ACTIVITIES PREDICTION USING BOOSTING

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```
In [1]:
```

```
import pandas as pd
from sklearn.model_selection import train_test_split
import warnings
warnings.filterwarnings('ignore')
from sklearn.metrics import precision_score, recall_score,accuracy_score,roc_auc_score,c
from sklearn.ensemble import GradientBoostingClassifier,AdaBoostClassifier
from sklearn.model_selection import GridSearchCV
from sklearn.linear_model import LogisticRegressionCV
from sklearn.ensemble import RandomForestClassifier, VotingClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import cross_val_score
```

## Step-1: [Understand Data]

In [2]:

HA=pd.read\_csv("Human\_Activity\_Data.csv")
HA.head()

#### Out[2]:

	tBodyAcc- mean()-X	tBodyAcc- mean()-Y	tBodyAcc- mean()-Z	tBodyAcc- std()-X	tBodyAcc- std()-Y	tBodyAcc- std()-Z	tBodyAcc- mad()-X	tBodyA mad(
0	0.288585	-0.020294	-0.132905	-0.995279	-0.983111	-0.913526	-0.995112	-0.9831
1	0.278419	-0.016411	-0.123520	-0.998245	-0.975300	-0.960322	-0.998807	-0.9749
2	0.279653	-0.019467	-0.113462	-0.995380	-0.967187	-0.978944	-0.996520	-0.963€
3	0.279174	-0.026201	-0.123283	-0.996091	-0.983403	-0.990675	-0.997099	-0.9827
4	0.276629	-0.016570	-0.115362	-0.998139	-0.980817	-0.990482	-0.998321	-0.9796

5 rows × 562 columns

```
In [3]:
                                                                                M
HA.shape
Out[3]:
(10299, 562)
In [4]:
                                                                                M
HA.size
Out[4]:
5788038
                                                                                M
In [5]:
HA.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10299 entries, 0 to 10298
Columns: 562 entries, tBodyAcc-mean()-X to Activity
dtypes: float64(561), object(1)
memory usage: 44.2+ MB
In [6]:
                                                                                M
HA.columns
Out[6]:
'tBodyAcc-max()-X',
       'fBodyBodyGyroJerkMag-skewness()', 'fBodyBodyGyroJerkMag-kurtosis
()',
      'angle(tBodyAccMean,gravity)', 'angle(tBodyAccJerkMean),gravityMea
n)',
      'angle(tBodyGyroMean,gravityMean)',
      'angle(tBodyGyroJerkMean,gravityMean)', 'angle(X,gravityMean)',
       'angle(Y,gravityMean)', 'angle(Z,gravityMean)', 'Activity'],
     dtype='object', length=562)
```

```
In [7]:
                                                                                           M
HA['angle(Z,gravityMean)'].value_counts
Out[7]:
<bound method IndexOpsMixin.value_counts of 0</pre>
                                                       -0.058627
1
        -0.054317
2
        -0.049118
3
        -0.047663
4
        -0.043892
10294
         0.184784
10295
         0.182412
10296
         0.181184
10297
         0.187563
10298
         0.188103
Name: angle(Z,gravityMean), Length: 10299, dtype: float64>
Step-2: [Build a small Dataset
In [8]:
                                                                                           M
import numpy as np
a=HA[HA['Activity']=='LAYING'].head(500)
b=HA[HA['Activity']=='SITTING'].head(500)
c=HA[HA['Activity']=='WALKING'].head(500)
In [9]:
                                                                                           M
newdf=pd.concat([a,b,c])
In [10]:
                                                                                           H
newdf.shape
Out[10]:
(1500, 562)
In [11]:
                                                                                           M
newdf.to_csv("Human_Activity_new.csv")
Step-3: [Build GradientBoostingClassifier]
In [12]:
                                                                                           M
df = pd.read_csv("Human_Activity_new.csv")
```

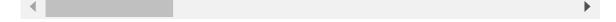
In [13]:

df.head()

## Out[13]:

	Unnamed: 0	tBodyAcc- mean()-X	tBodyAcc- mean()-Y	tBodyAcc- mean()-Z	tBodyAcc- std()-X	tBodyAcc- std()-Y	tBodyAcc- std()-Z	tBodyAc mad()
0	51	0.403474	-0.015074	-0.118167	-0.914811	-0.895231	-0.891748	-0.9176
1	52	0.278373	-0.020561	-0.096825	-0.984883	-0.991118	-0.982112	-0.9879
2	53	0.276555	-0.017869	-0.107621	-0.994195	-0.996372	-0.995615	-0.9949
3	54	0.279575	-0.017276	-0.109481	-0.996135	-0.995812	-0.998689	-0.9963
4	55	0.276527	-0.016819	-0.107983	-0.996775	-0.997256	-0.995422	-0.9971

#### 5 rows × 563 columns



In [14]:

df.shape

## Out[14]:

(1500, 563)

In [15]:

df.size

Out[15]:

844500

In [16]: ▶

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1500 entries, 0 to 1499

Columns: 563 entries, Unnamed: 0 to Activity dtypes: float64(561), int64(1), object(1)

memory usage: 6.4+ MB

In [17]: ▶

df.columns

#### Out[17]:

In [18]: ▶

df.describe()

# Out[18]:

	Unnamed: 0	tBodyAcc- mean()-X	tBodyAcc- mean()-Y	tBodyAcc- mean()-Z	tBodyAcc- std()-X	tBodyAcc- std()-Y	tBo
count	1500.000000	1500.000000	1500.000000	1500.000000	1500.000000	1500.000000	1500.
mean	1430.972000	0.270425	-0.015542	-0.108074	-0.751373	-0.597033	-0.
std	845.331241	0.084685	0.036471	0.055224	0.317106	0.490449	0.
min	27.000000	-1.000000	-0.684097	-1.000000	-0.999300	-0.998524	-0.
25%	726.750000	0.264859	-0.021433	-0.118534	-0.993145	-0.983467	-0.
50%	1407.500000	0.276946	-0.016817	-0.108755	-0.966535	-0.937492	-0.
75%	2133.250000	0.285803	-0.011554	-0.100423	-0.392574	-0.057412	-0.
max	3102.000000	0.559135	0.324130	0.543939	0.057201	0.671192	0.

8 rows × 562 columns

```
3/12/23, 10:48 PM
                                                 PML-lab10--225229144 - Jupyter Notebook
  In [19]:
                                                                                                        M
  X=df.drop('Activity',axis=1)
  X.head()
  Out[19]:
                                                                                   tBodyAc
                 tBodyAcc-
                            tBodyAcc-
                                       tBodyAcc-
                                                  tBodyAcc-
                                                             tBodyAcc-
                                                                        tBodyAcc-
      Unnamed:
                  mean()-X
                             mean()-Y
                                        mean()-Z
                                                      std()-X
                                                                 std()-Y
                                                                            std()-Z
                                                                                      mad(
   0
             51
                  0.403474
                             -0.015074
                                        -0.118167
                                                   -0.914811
                                                              -0.895231
                                                                         -0.891748
                                                                                    -0.9176
   1
             52
                  0.278373
                             -0.020561
                                        -0.096825
                                                   -0.984883
                                                              -0.991118
                                                                          -0.982112
                                                                                    -0.9879
   2
             53
                  0.276555
                             -0.017869
                                        -0.107621
                                                   -0.994195
                                                              -0.996372
                                                                         -0.995615
                                                                                    -0.9949
   3
             54
                  0.279575
                             -0.017276
                                        -0.109481
                                                   -0.996135
                                                              -0.995812
                                                                         -0.998689
                                                                                    -0.9963
             55
                  0.276527
                             -0.016819
                                        -0.107983
                                                   -0.996775
                                                              -0.997256
                                                                         -0.995422
                                                                                    -0.9971
  5 rows × 562 columns
  In [20]:
                                                                                                        M
  y=df['Activity']
  y.head()
  Out[20]:
  0
        LAYING
  1
        LAYING
  2
        LAYING
  3
        LAYING
  4
        LAYING
  Name: Activity, dtype: object
  In [21]:
                                                                                                        M
  X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.3,random_state=42)
  In [22]:
                                                                                                        H
  model = GradientBoostingClassifier(subsample=0.5,n_estimators=100,learning_rate=1.0,max_
  model.fit(X_train,y_train)
  y_pred = model.predict(X_test)
```

```
In [23]:
```

H

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

1.0

```
In [24]:
                                                                                           M
print(classification_report(y_test,y_pred))
               precision
                            recall f1-score
                                                 support
      LAYING
                    1.00
                               1.00
                                         1.00
                                                     148
     SITTING
                    1.00
                               1.00
                                         1.00
                                                     141
     WALKING
                    1.00
                               1.00
                                         1.00
                                                     161
                                         1.00
                                                     450
    accuracy
                                         1.00
                                                     450
   macro avg
                    1.00
                               1.00
weighted avg
                               1.00
                                         1.00
                                                     450
                    1.00
Step-4: [Find Best no. of trees and Best Learning Rate using Grid Search and Cross Validation]
In [25]:
                                                                                           M
all_scores = cross_val_score(estimator=model, X=X_train, y=y_train, cv=5)
In [26]:
                                                                                           M
print(all_scores)
[1.
           1.
                      1.
                                           0.9952381]
                                 1.
In [27]:
                                                                                           M
all_scores.mean()
Out[27]:
0.9990476190476191
In [28]:
                                                                                           M
parameter = {'n_estimators': [50, 100, 200, 400], 'learning_rate': [0.1, 0.01]}
In [29]:
model1 = GridSearchCV(estimator=model,
```

param\_grid=parameter,cv=5,n\_jobs=-1)

```
In [30]:
                                                                                          M
model1.fit(X_train,y_train)
Out[30]:
GridSearchCV(cv=5,
             estimator=GradientBoostingClassifier(learning_rate=1.0,
                                                    max depth=10, subsample
=0.5),
             n_jobs=-1,
             param_grid={'learning_rate': [0.1, 0.01],
                          'n_estimators': [50, 100, 200, 400]})
In [31]:
                                                                                          H
y_pred2=model1.predict(X_test)
In [32]:
                                                                                          M
print(accuracy_score(y_test,y_pred2))
1.0
In [33]:
                                                                                          H
print(classification_report(y_test,y_pred2))
              precision
                            recall f1-score
                                                support
                    1.00
                              1.00
                                         1.00
      LAYING
                                                    148
     SITTING
                    1.00
                              1.00
                                         1.00
                                                    141
     WALKING
                    1.00
                              1.00
                                         1.00
                                                    161
                                         1.00
                                                    450
    accuracy
                                         1.00
                                                    450
                    1.00
                              1.00
   macro avg
weighted avg
                    1.00
                              1.00
                                         1.00
                                                    450
In [34]:
                                                                                          M
print(model1.best_estimator_)
```

 $\label{lem:contingClassifier} Gradient Boosting Classifier (\verb|max_depth=10|, n_estimators=50|, subsample=0.5|)$ 

## step 5: [Best AdaBoostClassifier]

```
In [35]:
                                                                                        M
base = DecisionTreeClassifier(max features=4)
model2 = AdaBoostClassifier(base_estimator=base,random_state=0)
param_grid = {'n_estimators': [100, 150, 200], 'learning_rate': [0.01, 0.001]}
model3 = GridSearchCV(model2,param grid,cv=5,n jobs=-1)
model3.fit(X_train,y_train)
Out[35]:
GridSearchCV(cv=5,
             estimator=AdaBoostClassifier(base_estimator=DecisionTreeClas
sifier(max_features=4),
                                           random state=0),
             n jobs=-1,
             param_grid={'learning_rate': [0.01, 0.001],
                          'n_estimators': [100, 150, 200]})
In [36]:
                                                                                        M
y_pred3=model3.predict(X_test)
y_pred3
Out[36]:
array(['WALKING', 'WALKING', 'LAYING', 'SITTING', 'SITTING',
       'WALKING', 'SITTING', 'WALKING', 'LAYING', 'LAYING', 'WALKING',
       'SITTING', 'SITTING', 'LAYING', 'WALKING', 'WALKING', 'SITTIN
G',
       'LAYING', 'LAYING', 'WALKING', 'WALKING', 'WALKING', 'WALKING',
       'WALKING', 'LAYING', 'WALKING', 'WALKING', 'WALKING', 'WALKIN
G',
       'WALKING', 'LAYING', 'WALKING', 'LAYING', 'SITTING',
       'LAYING', 'LAYING', 'WALKING', 'SITTING', 'WALKING', 'SITTING',
       'WALKING', 'SITTING', 'WALKING', 'LAYING', 'LAYING',
       'LAYING', 'SITTING', 'LAYING', 'SITTING', 'WALKING', 'WALKING', 'SITTING', 'LAYING', 'SITTING', 'SITTING', 'SITTING',
       'SITTING', 'SITTING', 'LAYING', 'LAYING', 'SITTING',
       'LAYING', 'WALKING', 'WALKING', 'LAYING', 'LAYING', 'SITTING',
       'SITTING', 'LAYING', 'SITTING', 'WALKING', 'SITTING', 'SITTIN
G',
       'SITTING', 'LAYING', 'WALKING', 'LAYING', 'LAYING', 'SITTING',
In [37]:
                                                                                        M
accuracy_score(y_test,y_pred3)
```

#### Out[37]:

0.9

In [40]:

```
In [38]: ▶
```

```
print(classification_report(y_test,y_pred3))
```

	precision	recall	f1-score	support
LAYING	0.84	0.86	0.85	148
SITTING	0.85	0.84	0.84	141
WALKING	1.00	0.99	0.99	161
accuracy			0.90	450
macro avg	0.90	0.90	0.90	450
weighted avg	0.90	0.90	0.90	450

```
In [39]: ▶
```

```
print(model3.best_estimator_)
```

AdaBoostClassifier(base\_estimator=DecisionTreeClassifier(max\_features=4), learning\_rate=0.01, n\_estimators=100, random\_state=0)

#### Step-6: [Build a LogisticRegressionCV classifier]

```
model4 = LogisticRegressionCV(cv=4,Cs=5,penalty='12')
model4.fit(X_train,y_train)
y_pred2=model4.predict(X_test)
y_pred2
Out[40]:
array(['WALKING', 'WALKING', 'LAYING', 'LAYING', 'LAYING', 'SITTING',
                             'WALKING', 'SITTING', 'WALKING', 'LAYING', 'SITTING', 'WALKIN
G',
                            'SITTING', 'WALKING', 'LAYING', 'WALKING', 'WALKING', 'SITTIN
G',
                             'LAYING', 'LAYING', 'WALKING', 'WALKING', 'WALKING',
                            'WALKING', 'LAYING', 'WALKING', '
G',
                            'WALKING', 'SITTING', 'WALKING', 'WALKING', 'LAYING', 'LAYING',
                            'LAYING', 'LAYING', 'WALKING', 'LAYING', 'WALKING', 'LAYING',
                            'WALKING', 'SITTING', 'WALKING', 'LAYING', 'LAYING', 'LAYING',
                            'SITTING', 'SITTING', 'SITTING', 'WALKING', 'WALKIN
G',
```

'SITTING', 'SITTING', 'LAYING', 'LAYING', 'LAYING', 'SITTING', 'SITTING', 'LAYING', 'LAYING', 'SITTING', 'LAYING', 'LAYING', 'LAYING', 'SITTING', 'SITTING

H

In [41]: ▶

```
accuracy_score(y_test,y_pred2)
```

# Out[41]:

0.99777777777778

In [42]: ▶

```
print(classification_report(y_test,y_pred2))
```

	precision	recall	f1-score	support
LAYING	1.00	1.00	1.00	148
SITTING	1.00	0.99	1.00	141
WALKING	0.99	1.00	1.00	161
accuracy			1.00	450
macro avg	1.00	1.00	1.00	450
weighted avg	1.00	1.00	1.00	450

## Step-7: [Build VotingClassifier]

```
In [43]: ▶
```

```
model4=VotingClassifier(estimators=[('lr',model4),('gbc',model1)], voting='hard')
model4.fit(X_train,y_train)
```

```
Out[43]:
VotingClassifier(estimators=[('lr', LogisticRegressionCV(Cs=5, cv=4)),
                              ('gbc',
                               GridSearchCV(cv=5,
                                             estimator=GradientBoostingClas
sifier(learning_rate=1.0,
max_depth=10,
subsample=0.5),
                                             n jobs=-1,
                                             param_grid={'learning_rate':
[0.1,
0.01],
                                                          'n estimators': [5
0, 100,
                                                                           2
00,
                                                                           4
00]}))])
```

```
In [44]:
                                                                                                                                                                                                                                       M
y_pred3=model4.predict(X_test)
y_pred3
Out[44]:
array(['WALKING', 'WALKING', 'LAYING', 'LAYING', 'LAYING', 'SITTING',
                    'WALKING', 'SITTING', 'WALKING', 'LAYING', 'SITTING', 'WALKIN
G',
                    'SITTING', 'WALKING', 'LAYING', 'WALKING', 'WALKING', 'SITTIN
G',
                    'LAYING', 'LAYING', 'WALKING', 'WALKING', 'WALKING',
                    'WALKING', 'LAYING', 'WALKING', 'WALKING', 'WALKING', 'WALKIN
G',
                    'WALKING', 'SITTING', 'WALKING', 'WALKING', 'LAYING', 'LAYING',
                    'LAYING', 'LAYING', 'WALKING', 'LAYING', 'WALKING', 'LAYING',
                   'WALKING', 'SITTING', 'WALKING', 'LAYING', 'LAYING', 'LAYING',
                    'SITTING', 'SITTING', 'SITTING', 'WALKING', 'WALKIN
G',
                    'SITTING', 'SITTING', 'LAYING', 'LAYING', 'SITTING',
                   'SITTING', 'SITTING', 'LAYING', 'SITTING', 'LAYING',
                    'LAYING', 'WALKING', 'WALKING', 'LAYING', 'LAYING', 'SITTING', 'SITTING', 'LAYING', 'SITTING', 'WALKING', 'SITTING', 'SIT
In [45]:
                                                                                                                                                                                                                                       M
accuracy_score(y_test,y_pred3)
Out[45]:
1.0
In [46]:
                                                                                                                                                                                                                                       M
print(classification_report(y_test,y_pred3))
                                      precision
                                                                        recall f1-score
                                                                                                                            support
                LAYING
                                                                                                         1.00
                                                                                                                                       148
                                                   1.00
                                                                              1.00
             SITTING
                                                   1.00
                                                                              1.00
                                                                                                         1.00
                                                                                                                                       141
             WALKING
                                                   1.00
                                                                              1.00
                                                                                                         1.00
                                                                                                                                       161
                                                                                                                                       450
                                                                                                         1.00
           accuracy
                                                   1.00
                                                                              1.00
                                                                                                         1.00
                                                                                                                                       450
        macro avg
weighted avg
                                                   1.00
                                                                              1.00
                                                                                                         1.00
                                                                                                                                       450
Step-8: [Interpret your results]
                                                                                                                                                                                                                                       M
In [47]:
print(model1.best estimator )
```

localhost:8888/notebooks/PML-lab10--225229144.ipynb#NAME:-UMESH-C

GradientBoostingClassifier(max\_depth=10, n\_estimators=50, subsample=0.5)

```
In [48]:
print(model3.best estimator )
AdaBoostClassifier(base_estimator=DecisionTreeClassifier(max_features=4),
                                              learning rate=0.01, n estimators=100, random state=0)
GradientBoostingClassifier
                                                                                                                                                                                                              M
In [49]:
classifierF = GradientBoostingClassifier(n_estimators=50,max_features=4)
all_scoresF = cross_val_score(estimator=classifierF, X=X_train, y=y_train, cv=5)
parameter = {'n_estimators': [50, 100, 200, 400], 'learning_rate': [0.1, 0.01]}
modelGB = GridSearchCV(estimator=classifierF,param_grid=parameter,cv=5,n_jobs=-1)
modelGB.fit(X_train,y_train)
Out[49]:
GridSearchCV(cv=5,
                               estimator=GradientBoostingClassifier(max_features=4,
                                                                                                                        n estimators=50),
                               n jobs=-1,
                               param_grid={'learning_rate': [0.1, 0.01],
                                                             'n estimators': [50, 100, 200, 400]})
                                                                                                                                                                                                              M
In [50]:
y_predGB=model3.predict(X_test)
y predGB
Out[50]:
array(['WALKING', 'WALKING', 'LAYING', 'LAYING', 'SITTING', 'SITTING',
                  'WALKING', 'SITTING', 'WALKING', 'LAYING', 'LAYING', 'WALKING',
                 'SITTING', 'SITTING', 'LAYING', 'WALKING', 'WALKING', 'SITTIN
G',
                  'LAYING', 'LAYING', 'WALKING', 'WALKING', 'WALKING',
                  'WALKING', 'LAYING', 'WALKING', '
G',
                  'WALKING', 'LAYING', 'WALKING', 'WALKING', 'LAYING', 'SITTING',
                  'LAYING', 'LAYING', 'WALKING', 'SITTING', 'WALKING', 'SITTING',
                  'WALKING', 'SITTING', 'WALKING', 'LAYING', 'LAYING', 'LAYING',
                  'LAYING', 'SITTING', 'LAYING', 'SITTING', 'WALKING', 'WALKING',
                 'SITTING', 'LAYING', 'SITTING', 'LAYING', 'SITTING', 'SITTING', 'SITTING', 'SITTING', 'LAYING', 'LAYING', 'SITTING', 'SITTING', 'LAYING', 'LAYING', 'LAYING', 'LAYING', 'SITTING', 'LAYING', 'LAYING', 'LAYING', 'SITTING',
                  'SITTING', 'LAYING', 'SITTING', 'WALKING', 'SITTING', 'SITTIN
G',
                 'SITTING', 'LAYING', 'WALKING', 'LAYING', 'LAYING', 'SITTING',
```

H

In [51]: ▶

```
accuracy_score(y_test,y_predGB)
```

## Out[51]:

0.9

In [52]:

```
print(classification_report(y_test,y_predGB))
```

	precision	recall	f1-score	support
LAYING	0.84	0.86	0.85	148
SITTING	0.85	0.84	0.84	141
WALKING	1.00	0.99	0.99	161
accuracy			0.90	450
macro avg	0.90	0.90	0.90	450
weighted avg	0.90	0.90	0.90	450

#### AdaBoostClassifier

```
In [53]: ▶
```

```
modelABC = AdaBoostClassifier(base_estimator=DecisionTreeClassifier(),
learning_rate=0.01,
n_estimators=100,
random_state=0)
param_grid = {'n_estimators': [100, 150, 200], 'learning_rate': [0.01, 0.001]}
modelGSCV = GridSearchCV(modelABC,param_grid,cv=5,n_jobs=-1)
modelGSCV.fit(X_train,y_train)
```

#### Out[53]:

```
In [54]:
                                                                                                                                                                                                                                              M
y_predGSCV=model3.predict(X_test)
y predGSCV
Out[54]:
array(['WALKING', 'WALKING', 'LAYING', 'SITTING', 'SITTING',
                     'WALKING', 'SITTING', 'WALKING', 'LAYING', 'LAYING', 'WALKING',
                    'SITTING', 'SITTING', 'LAYING', 'WALKING', 'WALKING', 'SITTIN
G',
                    'LAYING', 'LAYING', 'WALKING', 'WALKING', 'WALKING',
                    'WALKING', 'LAYING', 'WALKING', 'WALKING', 'WALKING', 'WALKIN
G',
                    'WALKING', 'LAYING', 'WALKING', 'WALKING', 'LAYING', 'SITTING',
                    'LAYING', 'LAYING', 'WALKING', 'SITTING', 'WALKING', 'SITTING',
                    'WALKING', 'SITTING', 'WALKING', 'LAYING', 'LAYING', 'LAYING',
                    'LAYING', 'SITTING', 'LAYING', 'SITTING', 'WALKING', 'WALKING',
                    'SITTING', 'LAYING', 'SITTING', 'LAYING', 'SITTING', 'SITTING', 'SITTING', 'SITTING', 'LAYING', 'LAYING', 'SITTING', 'LAYING', 'LAYING', 'LAYING', 'LAYING', 'LAYING', 'SITTING', 'LAYING', 'SITTING', 'SITTING',
                    'SITTING', 'LAYING', 'SITTING', 'WALKING', 'SITTING', 'SITTIN
G',
                    'SITTING', 'LAYING', 'WALKING', 'LAYING', 'LAYING', 'SITTING',
In [55]:
                                                                                                                                                                                                                                              M
accuracy_score(y_test,y_predGSCV)
Out[55]:
0.9
In [56]:
                                                                                                                                                                                                                                              M
print(classification_report(y_test,y_predGSCV))
                                       precision
                                                                           recall f1-score
                                                                                                                                support
                LAYING
                                                     0.84
                                                                                0.86
                                                                                                            0.85
                                                                                                                                           148
              SITTING
                                                     0.85
                                                                                0.84
                                                                                                            0.84
                                                                                                                                           141
                                                                                                            0.99
              WALKING
                                                     1.00
                                                                                0.99
                                                                                                                                           161
                                                                                                            0.90
                                                                                                                                          450
           accuracy
                                                     0.90
                                                                                0.90
                                                                                                            0.90
                                                                                                                                          450
         macro avg
                                                                                0.90
                                                                                                            0.90
weighted avg
                                                     0.90
                                                                                                                                           450
                                                                                                                                                                                                                                              H
In [ ]:
```