

Model Development Phase Template

Date	21 June 2024
Team ID	739680
Project Title	Estimating Presence or Absence of Smoking through bio signals
Maximum Marks	4 Marks

Initial Model Training Code, Model Validation and Evaluation Report

The initial model training code will be showcased in the future through a screenshot. The model validation and evaluation report will include classification reports, accuracy, and confusion matrices for multiple models, presented through respective screenshots.

Initial Model Training Code:

```

Model Building With Random Forest Classifier

In [53]: from sklearn.ensemble import RandomForestClassifier
In [54]: rf = RandomForestClassifier(criterion = 'entropy', random_state = 0)
In [55]: rf
Out[55]: RandomForestClassifier(criterion='entropy', random_state=0)
In [56]: rf.fit(x_train,y_train)
Out[56]: RandomForestClassifier(criterion='entropy', random_state=0)
In [57]: y_train_pred = rf.predict(x_train)
In [58]: y_test_pred = rf.predict(x_test)
In [59]: #Confusion Matrix For Training Data With Random Forest Classifier
In [60]: confusion_matrix(y_train , y_train_pred)
Out[60]: array([[24548,  0],
               [ 1, 14005]], dtype=int64)
In [61]: #Accuracy For Training Data With Random Forest Classifier
In [62]: accuracy_score(y_train,y_train_pred)*100
Out[62]: 99.99740616813219
In [63]: #Classification Report For Training Data With Random Forest Classifier
In [64]: print(classification_report(y_train,y_train_pred))

              precision    recall  f1-score   support

     0       1.00         1.00         1.00        24548
     1       1.00         1.00         1.00        14005

 accuracy          1.00         1.00         1.00        38553
 macro avg         1.00         1.00         1.00        38553
 weighted avg         1.00         1.00         1.00        38553

```

```

Model Building With Decision Tree

from sklearn.tree import DecisionTreeClassifier

des1 = DecisionTreeClassifier(criterion = 'entropy', random_state = 0)

des1.fit(x_train,y_train)
DecisionTreeClassifier(criterion='entropy', random_state=0)

y_train_pred = des1.predict(x_train)
y_test_pred = des1.predict(x_test)

#Confusion Matrix For Training Data With Decision Tree
confusion_matrix(y_train , y_train_pred)
array([[24548,  0],
       [ 0, 14005]], dtype=int64)

#Accuracy For Training Data With Decision Tree
accuracy_score(y_train,y_train_pred)*100
100.0

```

```

Model Building with Logistic Regression

from sklearn.linear_model import LogisticRegression
log1 = LogisticRegression()
log1
LogisticRegression()
log1.fit(x_train, y_train)
LogisticRegression()

from sklearn.metrics import confusion_matrix, accuracy_score, classification_report
y_train_pred = log1.predict(x_train)
y_test_pred = log1.predict(x_test)

# Confusion Matrix For Training Data with Training Data
confusion_matrix(y_train, y_train_pred)
array([[1915, 842],
       [4375, 9839]], dtype=int64)

# Accuracy For Training Data with Logistic Regression
accuracy_score(y_train, y_train_pred)*100
74.55969704043784

```

Model Validation and Evaluation Report:

Model	Classification Report	Accuracy	Confusion Matrix
Random forest classifier	<pre> Model Building With Random Forest Classifier In [53]: from sklearn.ensemble import RandomForestClassifier In [54]: rf = RandomForestClassifier(criterion = 'entropy', random_state = 0) In [55]: rf Out[55]: RandomForestClassifier(criterion='entropy', random_state=0) In [56]: rf.fit(x_train,y_train) Out[56]: RandomForestClassifier(criterion='entropy', random_state=0) In [57]: y_train_pred = rf.predict(x_train) In [58]: y_test_pred = rf.predict(x_test) In [59]: #Confusion Matrix For Training Data with Random Forest Classifier In [60]: confusion_matrix(y_train , y_train_pred) Out[60]: array([[24548, 0], [1, 14805]], dtype=int64) In [61]: #Accuracy For Training Data with Random Forest Classifier In [62]: accuracy_score(y_train,y_train_pred)*100 Out[62]: 99.99740616813219 In [63]: #Classification Report For Training Data with Random Forest Classifier In [64]: print(classification_report(y_train,y_train_pred)) precision recall f1-score support 0 1.00 1.00 1.00 24548 1 1.00 1.00 1.00 14805 accuracy macro avg 1.00 1.00 1.00 38553 weighted avg 1.00 1.00 1.00 38553 </pre>	69%	<pre> In [83]: confusion_matrix(y_test, y_test_pred) Out[83]: array([[8915, 1476], [1388, 4744]], dtype=int64) </pre>
Decision tree	<pre> Model Building With Decision Tree from sklearn.tree import DecisionTreeClassifier De1 = DecisionTreeClassifier(criterion = 'entropy', random_state = 0) De1.fit(x_train,y_train) DecisionTreeClassifier(criterion='entropy', random_state=0) y_train_pred = De1.predict(x_train) y_test_pred = De1.predict(x_test) #Confusion Matrix For Training Data with Decision Tree confusion_matrix(y_train , y_train_pred) array([[24548, 0], [0, 14805]], dtype=int64) #Accuracy For Training Data with Decision Tree accuracy_score(y_train,y_train_pred)*100 100.0 </pre>	64%	<pre> In [83]: confusion_matrix(y_test, y_test_pred) Out[83]: array([[8407, 1054], [1042, 4196]], dtype=int64) </pre>

Logistic Regression

Model Building with Logistic Regression

```
from sklearn.linear_model import LogisticRegression

logit = LogisticRegression()

logit.fit(x_train, y_train)
logit.predict(x_test)

from sklearn.metrics import confusion_matrix, accuracy_score, classification_report

y_train_pred = logit.predict(x_train)
y_test_pred = logit.predict(x_test)

# Confusion Matrix for Training Data with Training Data
confusion_matrix(y_train, y_train_pred)
array([[1915, 543],
       [ 4379, 9939]], dtype=int64)

# Accuracy for Training Data with Logistic Regression
accuracy_score(y_train, y_train_pred)*100
74.55969704043784
```

76.4%

```
In [83]: confusion_matrix(y_test, y_test_pred)
Out[83]: array([[5487, 1924],
               [1942, 4198]], dtype=int64)
```

Gradient Boosting



	print(confusion_matrix(y_train, y_train_pred))	print(accuracy_score(y_train, y_train_pred))
Decision Tree	<pre>array([[1915, 543], [4379, 9939]], dtype=int64)</pre>	74%
KNN	<pre>array([[1915, 543], [4379, 9939]], dtype=int64)</pre>	64%
Gradient Boosting	<pre>array([[1915, 543], [4379, 9939]], dtype=int64)</pre>	75%

75%



	print(confusion_matrix(y_train, y_train_pred))	print(accuracy_score(y_train, y_train_pred))
Decision Tree	<pre>array([[1915, 543], [4379, 9939]], dtype=int64)</pre>	74%
KNN	<pre>array([[1915, 543], [4379, 9939]], dtype=int64)</pre>	64%
Gradient Boosting	<pre>array([[1915, 543], [4379, 9939]], dtype=int64)</pre>	75%