

## Model Development Phase Template

Date	06 JULY 2024
Team ID	739909
Project Name	Unlocking Silent Signals: Decoding Body Language With Mediapipe
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:

```
import numpy as np
from sklearn.metrics import accuracy_score, f1_score, confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.preprocessing import StandardScaler
from sklearn.pipeline import make_pipeline

# Example dataset (Replace with your actual dataset)
X = np.random.rand(100, 334 + 468*3 + 21*3 + 21*3) # 100 samples, feature size should match keypoints output
y = np.random.randint(0, 2, 100) # Binary classification example

# Split the data into training and test sets
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=42)

# Create an SVM pipeline with scaling and linear kernel
svm_pipeline_linear = make_pipeline(StandardScaler(), SVC(kernel='linear', random_state=42))

# Fit the SVM model with Linear kernel
svm_pipeline_linear.fit(x_train, y_train)

# Predict on test set with Linear kernel
y_pred_linear = svm_pipeline_linear.predict(x_test)
```

```
import numpy as np
from sklearn.metrics import accuracy_score, f1_score, confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler

# Example dataset (Replace with your actual dataset)
X = np.random.rand(100, 334 + 468*3 + 21*3 + 21*3) # 100 samples, feature size should match keypoints output
y = np.random.randint(0, 2, 100) # Binary classification example

# Split the data into training and test sets
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=42)

# Standardize the data
scaler = StandardScaler()
x_train_scaled = scaler.fit_transform(x_train)
x_test_scaled = scaler.transform(x_test)

# Create the Logistic Regression model
logreg_model = LogisticRegression(solver='liblinear', random_state=42)

# Fit the model
logreg_model.fit(x_train_scaled, y_train)

# Predict on test set
lr_yhat = logreg_model.predict(x_test_scaled)
```

```

import numpy as np
from sklearn.metrics import accuracy_score, f1_score, confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.linear_model import RidgeClassifier

# Example dataset (Replace with your actual dataset)
X = np.random.rand(100, 334 + 468*3 + 21*3 + 21*3) # 100 samples, feature size should match keypoints output
y = np.random.randint(0, 2, 100) # Binary classification example

# Split the data into training and test sets
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=42)

# Create the Ridge Classifier model with default parameters
ridge_model = RidgeClassifier(alpha=1.0, random_state=42)

# Fit the model
ridge_model.fit(x_train, y_train)

# Predict on test set
rc_yhat = ridge_model.predict(x_test)

```

```

import numpy as np
from sklearn.metrics import accuracy_score, f1_score, confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.ensemble import GradientBoostingClassifier

# Example dataset (Replace with your actual dataset)
X = np.random.rand(100, 334 + 468*3 + 21*3 + 21*3) # 100 samples, feature size should match keypoints output
y = np.random.randint(0, 2, 100) # Binary classification example

# Split the data into training and test sets
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=42)

# Create the Gradient Boosting Classifier model with default parameters
gb_model = GradientBoostingClassifier(random_state=42)

# Fit the model
gb_model.fit(x_train, y_train)

# Predict on test set
gb_yhat = gb_model.predict(x_test)

```

```

import numpy as np
from sklearn.metrics import accuracy_score, f1_score, confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier

# Example dataset (Replace with your actual dataset)
X = np.random.rand(100, 334 + 468*3 + 21*3 + 21*3) # 100 samples, feature size should match keypoints output
y = np.random.randint(0, 2, 100) # Binary classification example

# Split the data into training and test sets
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=42)

# Create the Random Forest Classifier model with default parameters
rf_model = RandomForestClassifier(random_state=42)

# Fit the model
rf_model.fit(x_train, y_train)

# Predict on test set
rf_yhat = rf_model.predict(x_test)

```

## Model Validation and Evaluation Report:

Model	Classification Report	F1 Score	Confusion Matrix
SVM	<pre> SVM Classification Report:       precision    recall  f1-score   support        0       0.44      0.88      0.59        17       1       0.67      0.17      0.28        23   accuracy          0.48        40  macro avg       0.55      0.53      0.43        40  weighted avg    0.57      0.47      0.41        40           </pre>	27%	<pre> svm_cm_linear = confusion_matrix(y_test, y_pred_linear) print(f"Confusion Matrix:\n(svm_cm_linear)")  Confusion Matrix: [[15  2]  [19  4]]           </pre>
Logistic Regression	<pre> Logistic Regression Classification Report:       precision    recall  f1-score   support        0       0.42      0.53      0.47        19       1       0.44      0.33      0.38        21   accuracy          0.42        40  macro avg       0.43      0.43      0.42        40  weighted avg    0.43      0.42      0.42        40           </pre>	37%	<pre> lr_cm = confusion_matrix(y_test, lr_yhat) print(f"Logistic Regression Confusion Matrix:\n(lr_cm)")  Logistic Regression Confusion Matrix: [[10  9]  [14  7]]           </pre>
Ridge Classifier	<pre> Ridge Classifier Classification Report:       precision    recall  f1-score   support        0       0.44      0.40      0.42        20       1       0.45      0.50      0.48        20   accuracy          0.45        40  macro avg       0.45      0.45      0.45        40  weighted avg    0.45      0.45      0.45        40           </pre>	47%	<pre> rc_cm = confusion_matrix(y_test, rc_yhat) print(f"Ridge Classifier Confusion Matrix:\n(rc_cm)")  Ridge Classifier Confusion Matrix: [[ 8 12]  [10 10]]           </pre>
Gradient Boosting Classifier	<pre> Gradient Boosting Classifier Classification Report:       precision    recall  f1-score   support        0       0.27      0.40      0.32        15       1       0.50      0.36      0.42        25   accuracy          0.38        40  macro avg       0.39      0.38      0.37        40  weighted avg    0.41      0.38      0.38        40           </pre>	41%	<pre> gb_cm = confusion_matrix(y_test, gb_yhat) print(f"Gradient Boosting Classifier Confusion Matrix:\n(gb_cm)")  Gradient Boosting Classifier Confusion Matrix: [[ 6  9]  [16  9]]           </pre>
RandomForest Classifier	<pre> Random Forest Classifier Classification Report:       precision    recall  f1-score   support        0       0.00      0.00      0.00        16       1       0.60      1.00      0.75        24   accuracy          0.60        40  macro avg       0.30      0.50      0.37        40  weighted avg    0.36      0.60      0.45        40           </pre>	74%	<pre> rf_cm = confusion_matrix(y_test, rf_yhat) print(f"Random Forest Confusion Matrix:\n(rf_cm)")  Random Forest Confusion Matrix: [[ 0 16]  [ 0 24]]           </pre>