



## **Model Development Phase Template**

Date	21 July 2024
Team ID	739717
Project Title	Unlocking Silent Signals :Decoding Body Language with Mediapipe
Maximum Marks	10 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 a summary and training and validation performance metrics for multiple models, presented through respective screenshots.

## **Initial Model Training Code (5 marks):**

Paste the screenshot of the model traning code

## **Model Validation and Evaluation Report (5 marks):**

Model	Summary	Training and Validation Performance Metrics
Model 1	Gradient Boosting is the grouping of Gradient descent and Boosting. In gradient boosting, each new model minimizes the loss function from its predecessor using the Gradient Descent Method. This procedure continues until a more optimal estimate of the target variable has been achieved	Train Machine Learning Classification Model  [70]. from sklaum.njepiline import make.pjusline from sklaum.ninear.model import Endodredicalar from sklaum.ninear.model import Endodredicalar from sklaum.ninear.model import Endodredicalar from sklaum.ninear.model import Endodredicalsifier, GradlentBoostingClassifier  [80]: pipelines. v[    "!- imake.pjueline(StandardScalar(), ingisticRegression()),





Logistic Regression is used for • StandardScaler binary classification tasks, predicting - LogisticRegression outcomes like spam vs. non-spam or from sklearn.linear\_model import LogisticRegression disease vs. no disease, and it LogisticRegression provides probabilities of class LogisticRegression(max\_iter=1000, solver='saga') Model 2 membership. It also helps in fit\_models = {}
for algo, pipeline in pipelines.items():
 model = pipeline.fit(X\_train, y\_train)
 fit\_models[algo] = model understanding feature importance and serves as a strong baseline fit models['rc'].predict(X test) model due to its simplicity and Evaluate and serialize model from sklearn.metrics import accuracy\_score # Accuracy metrics import pickle interpretability. Train Machine Learning Classification Model The Random Forest Classifier is ideal because it combines the strength of multiple decision trees, offering high accuracy, robust Model 3 performance against overfitting, and [41]: dict\_keys(['lr', 'rc', 'rf', 'gb']) the ability to handle large and [42]: from sklearn.linear\_model import LogisticRegression complex datasets, e<sub>I</sub> suring reliable - LogisticRegression predictions in varied LogisticRegression(max\_iter=1000) shipping scenarios. Training model using Scikit Learn The file "coords.csv" into a Pandas Read in collected data and process DataFrame. It then splits the data into training and testing sets using the train\_test\_split function from Model 4 2 Happy 0.461510 0.595861 -1.264079 0.999784 0.481376 0.513183 -1.798528 0.999603 0.498597 \_ -0.000800 0.0 0.528513 0.512119 0.034527 Scikit-learn. The df.head() and 3 Huppy 0.463032 0.599466 -1.253284 0.999765 0.481420 0.515575 -1.181970 0.999566 0.498603 \_ -0.001560 0.0 0.529767 0.519925 0.032815 0.0 0.534700 df.tail() functions are used to display the first and last few rows of the DataFrame, respectively. 4591 Fight 0.684455 0.511044 -0.065705 0.999905 0.706319 0.433798 -0.828467 0.999967 0.722609 \_ -0.003345 0.0 0.747559 0.434463 0.029979 0.0 0.75255