

## Model Development Phase Template

Date	15 March 2024
Team ID	XXXXXX
Project Title	Predictive Modeling for Fleet Fuel Management using Machine Learning
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:

```
# Initialize and train a RandomForestRegressor
rf = RandomForestRegressor(random_state=42)
rf.fit(X_train, y_train)
y_pred_rf = rf.predict(X_test)
```

```
# Initialize HistGradientBoostingRegressor
hgb_reg = HistGradientBoostingRegressor(random_state=42)
hgb_reg.fit(X_train, y_train)
y_pred_hgb = hgb_reg.predict(X_test)
```

## Model Validation and Evaluation Report:

Model	Classification Report	Accuracy	Confusion Matrix																																													
Random Forest Regressor	<pre>[67]: # Calculate the R2 score print(f"Random Forest Regressor R² Score: {r2_rf}") print(f"Random Forest Regressor Mean Squared Error: {mse}") print(f"Random Forest Regressor Mean Absolute Error: {mae}")  Random Forest Regressor R² Score: 0.5889485334633298 Random Forest Regressor Mean Squared Error: 0.3442496727207985 Random Forest Regressor Mean Absolute Error: 0.42898686939886877</pre>	60	<pre>print("Classification Report:") print(class_report)  print("Confusion Matrix:") print(conf_matrix)</pre> <table><tr><th colspan="5">Classification Report:</th></tr><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>0</td><td>0.77</td><td>0.72</td><td>0.74</td><td>32</td></tr><tr><td>1</td><td>0.41</td><td>0.44</td><td>0.43</td><td>27</td></tr><tr><td>2</td><td>0.34</td><td>0.33</td><td>0.34</td><td>30</td></tr><tr><td>3</td><td>0.59</td><td>0.61</td><td>0.60</td><td>28</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.53</td><td>117</td></tr><tr><td>macro avg</td><td>0.53</td><td>0.53</td><td>0.53</td><td>117</td></tr><tr><td>weighted avg</td><td>0.53</td><td>0.53</td><td>0.53</td><td>117</td></tr></table> <pre>Confusion Matrix: [[23  7  0  2]  [ 4 12 10  1]  [ 3  8 10  9]  [ 0  2  9 17]]</pre>	Classification Report:						precision	recall	f1-score	support	0	0.77	0.72	0.74	32	1	0.41	0.44	0.43	27	2	0.34	0.33	0.34	30	3	0.59	0.61	0.60	28	accuracy			0.53	117	macro avg	0.53	0.53	0.53	117	weighted avg	0.53	0.53	0.53	117
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HistGradient Boosting Regressor	<pre># Calculate the R2 score for the best model r2_best_gbr = r2_score(y_test, y_pred_best_gbr) mse = mean_squared_error(y_test, y_pred_best_gbr) mae = mean_absolute_error(y_test, y_pred_best_gbr) print('Best GradientBoosting R2 Score:', r2_best_gbr) print('Best Hyperparameters:', grid_search.best_params_)  Best GradientBoosting R2 Score: -9.383642721589119 Best Hyperparameters: {'learning_rate': 0.01, 'max_depth': 5, 'min_samples_leaf': 1, 'min_sam  # Calculate the R2 score for the best model print(r2_best_gbr) print(mse) print(mae)  -9.383642721589119 13.220675735887355 3.5036213550286974</pre>	53	<pre>print("Classification Report:") print(class_report)  print("Confusion Matrix:") print(conf_matrix)</pre> <table><tr><th colspan="5">Classification Report:</th></tr><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>0</td><td>0.77</td><td>0.72</td><td>0.74</td><td>32</td></tr><tr><td>1</td><td>0.41</td><td>0.44</td><td>0.43</td><td>27</td></tr><tr><td>2</td><td>0.34</td><td>0.33</td><td>0.34</td><td>30</td></tr><tr><td>3</td><td>0.59</td><td>0.61</td><td>0.60</td><td>28</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.53</td><td>117</td></tr><tr><td>macro avg</td><td>0.53</td><td>0.53</td><td>0.53</td><td>117</td></tr><tr><td>weighted avg</td><td>0.53</td><td>0.53</td><td>0.53</td><td>117</td></tr></table> <pre>Confusion Matrix: [[23  7  0  2]  [ 4 12 10  1]  [ 3  8 10  9]  [ 0  2  9 17]]</pre>	Classification Report:						precision	recall	f1-score	support	0	0.77	0.72	0.74	32	1	0.41	0.44	0.43	27	2	0.34	0.33	0.34	30	3	0.59	0.61	0.60	28	accuracy			0.53	117	macro avg	0.53	0.53	0.53	117	weighted avg	0.53	0.53	0.53	117
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