

Model Optimization and Tuning Phase Template

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Team ID	SWTID1720527361
Project Title	TrafficTellgence: Advanced Traffic Volume Estimation with Machine Learning
Maximum Marks	10 Marks

Model Optimization and Tuning Phase

The Model Optimization and Tuning Phase involves refining machine learning models for peak performance. It includes optimized model code, fine-tuning hyperparameters, comparing performance metrics, and justifying the final model selection for enhanced predictive accuracy and efficiency.

Hyperparameter Tuning Documentation (6 Marks):

Model	Tuned Hyperparameters	Optimal Values
XG Boost	<pre> model=xgb.XGBRegressor() parameters={ 'max_depth': [3, 5, 8], 'min_child_weight': [1, 3, 5], 'eta': [0.1, 0.3, 0.5], 'subsample': [0.6, 0.8, 1], 'colsample_bytree': [0.6, 0.8, 1] } </pre>	<pre> y_pred=clf.predict(x_test) print("Best Score: ", r2_score(y_test, y_pred)) clf.best_params_ </pre> <p>Best Score: 0.9676877994811365</p> <pre> {'colsample_bytree': 1, 'eta': 0.3, 'max_depth': 8, 'min_child_weight': 1, 'subsample': 1} </pre>

Random Forest Regressor	<pre>#model Initialization regressor = RandomForestRegressor() #Parameters parameters={ 'n_estimators':[20, 50, 100], 'bootstrap':[True, False] }</pre>	<pre>y_pred=clf.predict(x_test) print("Best Score: ", r2_score(y_test, y_pred)) print("Best Values: ", clf.best_params_) Best Score: 0.9556679960267289 Best Values: {'bootstrap': True, 'n_estimators': 100}</pre>
Polynomial Regression	<pre>model=LinearRegression() parameters={ 'fit_intercept':[True, False], 'positive':[True, False] }</pre>	<pre>y_pred=clf.predict(x_test) print("Best Score: ", r2_score(y_test, y_pred)) print("Best Values: ", clf.best_params_) Best Score: 0.7686065818544895 Best Values: {'fit_intercept': True, 'positive': False}</pre>
SVR	<pre>model=SVR() parameters={ 'C': [0.1, 1, 10], 'kernel': ['linear', 'rbf'], 'gamma': [0.1, 1, 10], 'epsilon': [0.1, 0.5, 1] }</pre>	<pre>y_pred=clf.predict(x_test) print("Best Score: ", r2_score(y_test, y_pred)) clf.best_params_ Best Score: 0.6402522031519096 {'kernel': 'rbf', 'gamma': 10, 'epsilon': 0.5, 'C': 10}</pre>

Performance Metrics Comparison Report (2 Marks):

Model	Baseline Metric	Optimized Metric
XG Boost	<pre>from sklearn.metrics import mean_squared_error, r2_score, mean_absolute_error print("Mean Square Error: ", mean_squared_error(y_test, y_pred)) print("Mean Absolute Error: ", mean_absolute_error(y_test, y_pred)) print("R-square Score: ", r2_score(y_test, y_pred)) Mean Square Error: 120958.54825379612 Mean Absolute Error: 228.5786688810355 R-square Score: 0.9563201748182905</pre>	<pre>y_pred=clf.predict(x_test) print("Mean Square Error: ", mean_squared_error(y_test, y_pred)) print("Mean Absolute Error: ", mean_absolute_error(y_test, y_pred)) print("R-square Score: ", r2_score(y_test, y_pred)) Mean Square Error: 87733.82381560856 Mean Absolute Error: 199.1112719822503 R-square Score: 0.9676877994811365</pre>

Random Forest Regressor	<pre>from sklearn.metrics import mean_squared_error, r2_score, mean_absolute_error print("Mean Square Error: ", mean_squared_error(y_test, y_pred)) print("Mean Absolute Error: ", mean_absolute_error(y_test, y_pred)) print("R-square Score: ", r2_score(y_test, y_pred))</pre> <p>Mean Square Error: 198008.8816750678 Mean Absolute Error: 280.9588373446771 R-square Score: 0.9277791335225944</p>	<pre>y_pred=clf.predict(x_test) print("Mean Square Error: ", mean_squared_error(y_test, y_pred)) print("Mean Absolute Error: ", mean_absolute_error(y_test, y_pred)) print("R-square Score: ", r2_score(y_test, y_pred))</pre> <p>Mean Square Error: 120881.52201975712 Mean Absolute Error: 219.65486158265864 R-square Score: 0.9556679960267289</p>
Polynomial Regression	<pre>from sklearn.metrics import mean_squared_error, r2_score, mean_absolute_error print("Mean Square Error: ", mean_squared_error(y_test, y_pred)) print("Mean Absolute Error: ", mean_absolute_error(y_test, y_pred)) print("R-square Score: ", r2_score(y_test, y_pred))</pre> <p>Mean Square Error: 646496.8829842781 Mean Absolute Error: 588.6552844192978 R-square Score: 0.7605639174654056</p>	<pre>y_pred=clf.predict(x_test) print("Mean Square Error: ", mean_squared_error(y_test, y_pred)) print("Mean Absolute Error: ", mean_absolute_error(y_test, y_pred)) print("R-square Score: ", r2_score(y_test, y_pred))</pre> <p>Mean Square Error: 629758.8109993833 Mean Absolute Error: 591.4450058414657 R-square Score: 0.7686065818544895</p>
SVR	<pre>from sklearn.metrics import mean_squared_error, r2_score, mean_absolute_error print("Mean Square Error: ", mean_squared_error(y_test, y_pred)) print("Mean Absolute Error: ", mean_absolute_error(y_test, y_pred)) print("R-square Score: ", r2_score(y_test, y_pred))</pre> <p>Mean Square Error: 2104039.9111552383 Mean Absolute Error: 1256.7067758496808 R-square Score: 0.23504232546490522</p>	<pre>y_pred=clf.predict(x_test) print("Mean Square Error: ", mean_squared_error(y_test, y_pred)) print("Mean Absolute Error: ", mean_absolute_error(y_test, y_pred)) print("R-square Score: ", r2_score(y_test, y_pred))</pre> <p>Mean Square Error: 974323.6860184855 Mean Absolute Error: 760.1124997263398 R-square Score: 0.6402522031519096</p>

Final Model Selection Justification (2 Marks):

Final Model	Reasoning
XG Boost	This model had highest R2-Score before optimization and also it has highest R2-Score after optimization of 96.8%. It is selected for its highest performance among all other mode after hypertuning.