



Model Optimization and Tuning Phase Template

Date	09 JULY 2024	
Team ID	740660	
Project Title	Evolving efficient classification patterns in Lymphography	
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):

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```
y_pred = svr.predict(x_test)
   print("Prediction Evaluation using SVR Regression")
   print('Mean Absolute Error:', mean_absolute_error(y_test, y_pred))
   print('Mean Squared Error:', mean_squared_error(y_test, y_pred))
   print('Root Mean Squared Error:', np.sqrt(mean_squared_error(y_test, y_pred)))
   print('R-squared:', r2_score(y_test, y_pred))
Prediction Evaluation using SVR Regression
Mean Absolute Error: 0.7461813805059471
Mean Squared Error: 0.857300991971709
Root Mean Squared Error: 0.9259054984023526
R-squared: -1.8682932816897333
   y_pred = dt.predict(x_test)
   print("Prediction Evaluation using Random Regression")
   print('Mean Absolute Error:', mean_absolute_error(y_test, y_pred))
   print('Mean Squared Error:', mean_squared_error(y_test, y_pred))
   print('Root Mean Squared Error:', np.sqrt(mean_squared_error(y_test, y_pred)))
   print('R-squared:', r2_score(y_test, y_pred))
Prediction Evaluation using Random Regression
Mean Absolute Error: 1.63333333333333333
Mean Squared Error: 2.966666666666667
Root Mean Squared Error: 1.7224014243685084
R-squared: -8.92565055762082
```

```
# Assuming 'x test' is available in the environment and is a pandas DataFrame or a NumPy array.
    y pred = linReg.predict(x test) # Predict on the entire x test dataset
    print("Prediction Evaluation using Linear Regression")
    print('Mean Absolute Error:', mean_absolute_error(y_test, y_pred))
    print('Mean Squared Error:', mean_squared_error(y_test, y_pred))
    print('Root Mean Squared Error:', np.sqrt(mean_squared_error(y_test, y_pred)))
    print('R-squared:', r2_score(y_test, y_pred))
 Prediction Evaluation using Linear Regression
 Mean Absolute Error: 0.31939441380921024
 Mean Squared Error: 0.20665934429543611
 Root Mean Squared Error: 0.4545980029602375
 R-squared: 0.30857468451341064
    y_pred = lassoReg.predict(x_test)
    print("Prediction Evaluation using lasso Regression")
    print('Mean Absolute Error:', mean_absolute_error(y_test, y_pred))
    print('Mean Squared Error:', mean_squared_error(y_test, y_pred))
    print('Root Mean Squared Error:', np.sqrt(mean_squared_error(y_test, y_pred)))
    print('R-squared:', r2_score(y_test, y_pred))
 Prediction Evaluation using lasso Regression
 Mean Absolute Error: 0.6559753131806499
 Mean Squared Error: 0.7000312610728252
 Root Mean Squared Error: 0.8366787083898007
 R-squared: -1.3421120258942114
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        from sklearn.metrics import accuracy_score,f1_score,confusion_matrix,classification_report
       confusion matrix(y test,prediction)
    array([[11, 1, 0],
           [ 2, 15, 0],
[ 1, 0, 0]], dtype=int64)
       accuracy_score(y_test,prediction)
    0.866666666666667
```

Model	Optimized Metric	Confusion Matrix
Decision Tree	Decision Tree Accuracy: 0.73 Decision Tree Classification Report:	Confusion Matrix: [[9 4 1] [2 11 1] [0 0 3]
Dandom Forest	Random Forest Accuracy: 0.83 Random Forest Classification Report:	Confusion Matrix: [[10 3 1] [1 11 3] [0 0 2]]

Final Model Selection Justification (2 Marks):