

Model Optimization and Tuning Phase Template

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| Date | 09 JULY 2024 |
| Team ID | 740024 |
| 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))

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· 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))

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· Prediction Evaluation using Random Regression
Mean Absolute Error: 1.6333333333333333
Mean Squared Error: 2.9666666666666667
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))

```

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```

-- Prediction Evaluation using Linear Regression
Mean Absolute Error: 0.31939441388921024
Mean Squared Error: 0.20665934429543611
Root Mean Squared Error: 0.4545988029602375
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))

```

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```

-- 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

```

```

> from sklearn.metrics import accuracy_score, f1_score, confusion_matrix, classification_report

[46]

```

```

confusion_matrix(y_test, prediction)

[47]

```

```

... array([[11,  1,  0],
          [ 2, 15,  0],
          [ 1,  0,  0]], dtype=int64)

```

```

accuracy_score(y_test, prediction)

[48]

```

```

... 0.8666666666666667

```

Performance Metrics Comparison Report (2 Marks):

| Model | Optimized Metric | Confusion Matrix |
|---------------|--|--|
| Decision Tree | <pre>Decision Tree Accuracy: 0.73 Decision Tree Classification Report: precision recall f1-score support 1: 0.65 0.93 0.76 14 2: 0.90 0.64 0.75 14 3: 0.00 0.00 0.00 2 accuracy: 0.73 macro avg: 0.52 0.52 0.50 30 weighted avg: 0.72 0.73 0.71 30</pre> | <pre>Confusion Matrix: [[9 4 1] [2 11 1] [0 0 3]]</pre> |
| Random Forest | <pre>Random Forest Accuracy: 0.83 Random Forest Classification Report: precision recall f1-score support 1: 0.87 0.93 0.90 14 2: 0.88 0.86 0.83 14 3: 0.00 0.00 0.00 2 accuracy: 0.83 macro avg: 0.56 0.60 0.57 30 weighted avg: 0.78 0.83 0.80 30</pre> | <pre>Confusion Matrix: [[10 3 1] [1 11 3] [0 0 2]]</pre> |

Final Model Selection Justification (2 Marks):

```
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)

0.8666666666666667
```