



Model Optimization and Tuning Phase

Date	07 July 2024
Team ID	739899
Project Title	SmartLender – Envisioning Success: Predicting University Scores With Machine Learning
Maximum Marks	2Marks

Performance Metrics Comparison Report (2 Marks):

A Performance Metrics Comparison Report systematically evaluates the effectiveness of various machine learning models or algorithms by comparing key metrics such as MAE,MSE,R-Square. This report highlights the strengths and weaknesses of each model, providing insights into their performance on different datasets or tasks. By presenting a clear, detailed analysis, the report aids in selecting the most suitable model for deployment, ensuring optimal performance in realworld applications.









```
[56] # 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))
  Frediction Evaluation using Linear Regression
       Mean Absolute Error: 0,9264657671450711
      Mean Squared Error: 1.7890643253785259
       Root Mean Squared Error: 1.337559092294066
       R-squared: 0.7439493774592185
       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: 0.9352851280381133
      Mean Squared Error: 1.7857764808364731
       Root Mean Squared Error: 1.3363294806433303
       R-squared: 0.7444199332854502
     y_pred = svr.predict(x_test)
      print("Prediction Evaluation using support vector 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))
 Frediction Evaluation using support vector Regression
      Mean Absolute Error: 0.5454340693726399
      Mean Squared Error: 1.2411917299771091
      Root Mean Squared Error: 1,1140878466158355
      R-squared: 0.8223608225568596
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[59] y_pred = dt.predict(x_test)
      print("Prediction Evaluation using Decision 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 Decision Regression
      Mean Absolute Error: 5.264475724040743e-15
      Mean Squared Error: 2.7561365735867205e-28
      Root Mean Squared Error: 1.6601616106833456e-14
      R-squared: 1.0
```





```
y_pred = rf.predict(x_test)
print("Prediction Evaluation using Random Regression")
print('Mean Absolute Error:', mean_absolute_error(y_test, w_nead))
print('Mean Squared Error:', mean_squared_error(y_test, y_ Loading...
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: 0.010686590909099649
Mean Squared Error: 0.0009053592244319952
Root Mean Squared Error: 0.0300891878327082
R-squared: 0.9998704251212489