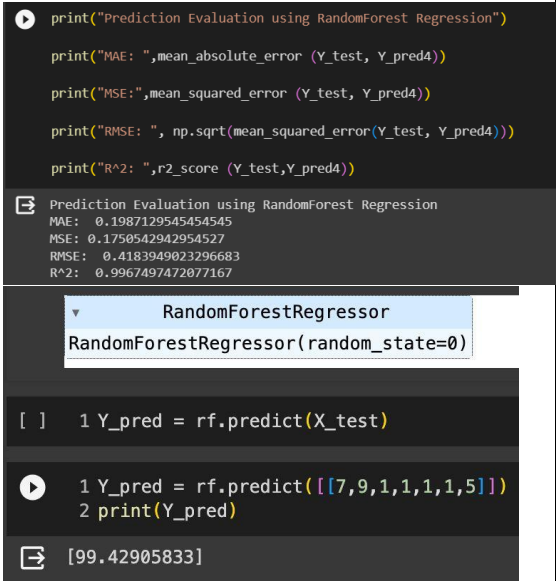
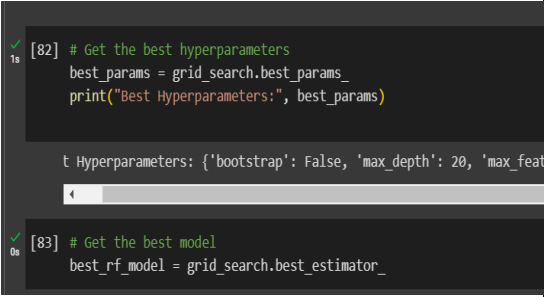


## Project Development Phase Model Performance Test

Date	14 November 2023
Team ID	591606
Project Name	ENVISIONING SUCCESS: Predicting University Scores using Machine Learning
Maximum Marks	10 Marks

### Model Performance Testing:

S.No.	Parameter	Values	Screenshot
1.	Metrics	<b>Regression Model:</b> MAE - , MSE - , RMSE - , R2 score -	<p>After performing all the Regression Models we found the best model as Random Forest Regression with best MAE, MSE, RMSE, R2 Scores</p>  <p>The screenshot shows a Jupyter Notebook cell with the following code and output:</p> <pre>print("Prediction Evaluation using RandomForest Regression") print("MAE: ",mean_absolute_error (Y_test, Y_pred4)) print("MSE:",mean_squared_error (Y_test, Y_pred4)) print("RMSE: ", np.sqrt(mean_squared_error(Y_test, Y_pred4))) print("R^2: ",r2_score (Y_test,Y_pred4))</pre> <p>The output displays the following metrics:</p> <pre>Prediction Evaluation using RandomForest Regression MAE:  0.1987129545454545 MSE:  0.1750542942954527 RMSE:  0.4183949023296683 R^2:  0.9967497472077167</pre> <p>Below the metrics, the code defines a <code>RandomForestRegressor</code> and makes a prediction:</p> <pre>RandomForestRegressor RandomForestRegressor(random_state=0)  1 Y_pred = rf.predict(X_test)  1 Y_pred = rf.predict([[7,9,1,1,1,1,5]]) 2 print(Y_pred)</pre> <p>The final output of the prediction is <code>[99.42905833]</code>.</p>
2.	Tune the Model	Hyperparameter Tuning - Validation Method -	<p>Using GridSearch for random forest model</p>  <p>The screenshot shows a Jupyter Notebook cell with the following code and output:</p> <pre>[82] # Get the best hyperparameters best_params = grid_search.best_params_ print("Best Hyperparameters:", best_params)</pre> <p>The output displays the best hyperparameters:</p> <pre>t Hyperparameters: {'bootstrap': False, 'max_depth': 20, 'max_feat</pre> <p>Below the hyperparameters, the code gets the best model:</p> <pre>[83] # Get the best model best_rf_model = grid_search.best_estimator_</pre>

			<div><div><div><div><div><div></div><div>✓</div><div>0s</div></div><div>[77] # Evaluate the model on the test set y_pred = best_rf_model.predict(X_test) mse = mean_squared_error(Y_test, Y_pred4) print("Mean Squared Error on Test Set:", mse)</div></div></div><div>Mean Squared Error on Test Set: 0.1750542942954527</div><div><div><div><div><div><div></div><div>✓</div><div>0s</div></div><div>[78] y_pred = best_rf_model.predict(X_test) mae = mean_absolute_error(Y_test, Y_pred4) print("Mean absolute error on Test Set:", mae)</div></div></div><div>Mean absolute error on Test Set: 0.1987129545454545</div><div><div><div><div><div><div></div><div>✓</div><div>0s</div></div><div>[79] y_pred = best_rf_model.predict(X_test) rmse = np.sqrt(mean_squared_error(Y_test, Y_pred4)) print("Root Mean Squared Error on Test Set:", mae)</div></div></div><div>Root Mean Squared Error on Test Set: 0.1987129545454545</div><div><div><div><div><div><div></div><div>✓</div><div>0s</div></div><div><div><div></div><div></div></div><div>y_pred = best_rf_model.predict(X_test) r2 = r2_score(Y_test, Y_pred4) print("r2 score on Test Set:", r2)</div></div></div><div><div><div></div><div></div></div><div>r2 score on Test Set: 0.9967497472077167</div></div></div></div></div></div></div></div></div></div></div>
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