

Predicting Rainfall in Australia

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PROJECT GROUP #3

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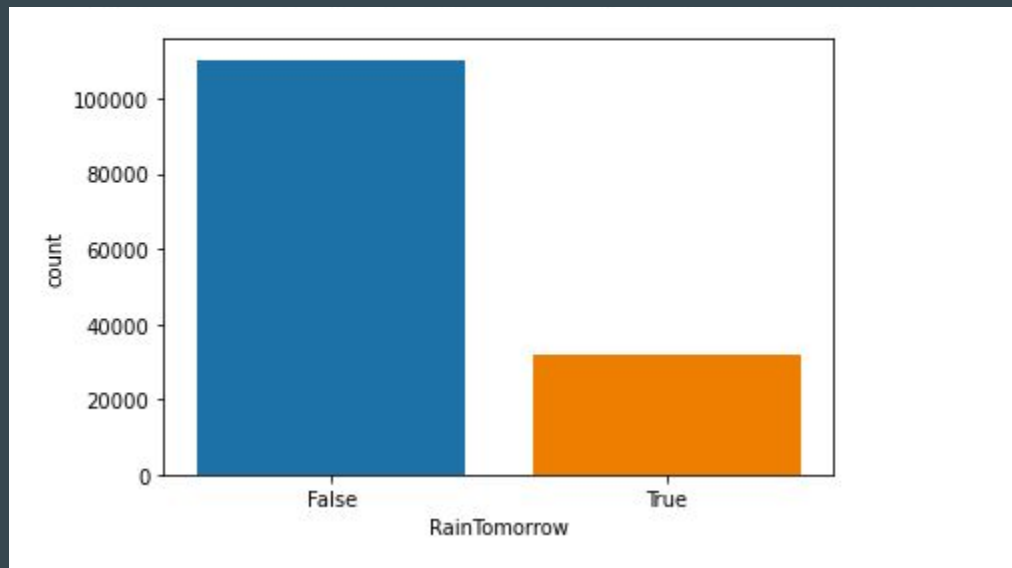
Introduction

- The aim of the project is to predict if it will rain in Australia.
- The target audience is the general public of Australia.
- EDA (Exploratory Data Analysis) was performed to visualize various important trends.
- A logistic regression model was trained to predict rain.
- The model was evaluated using various classification metrics.

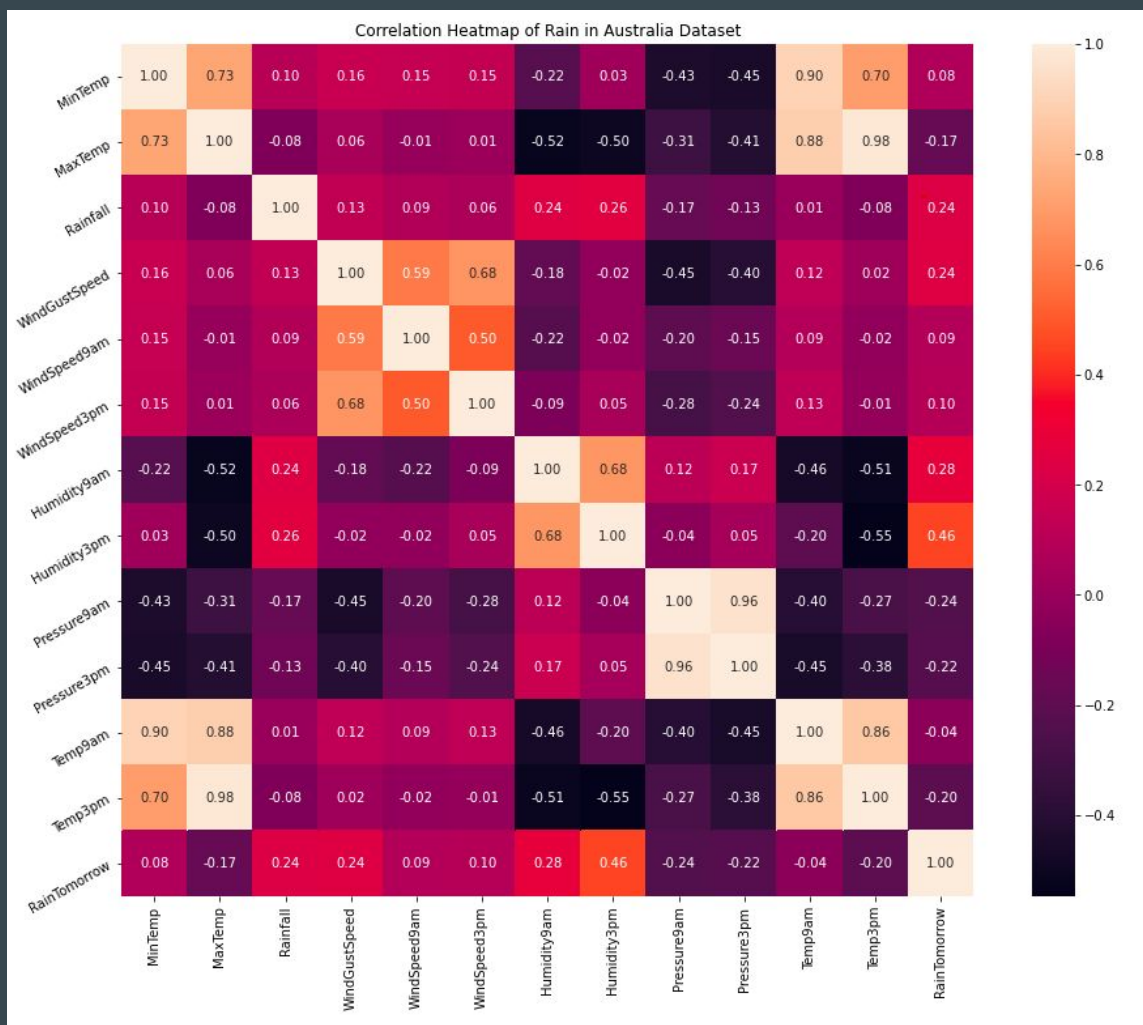
About the Dataset

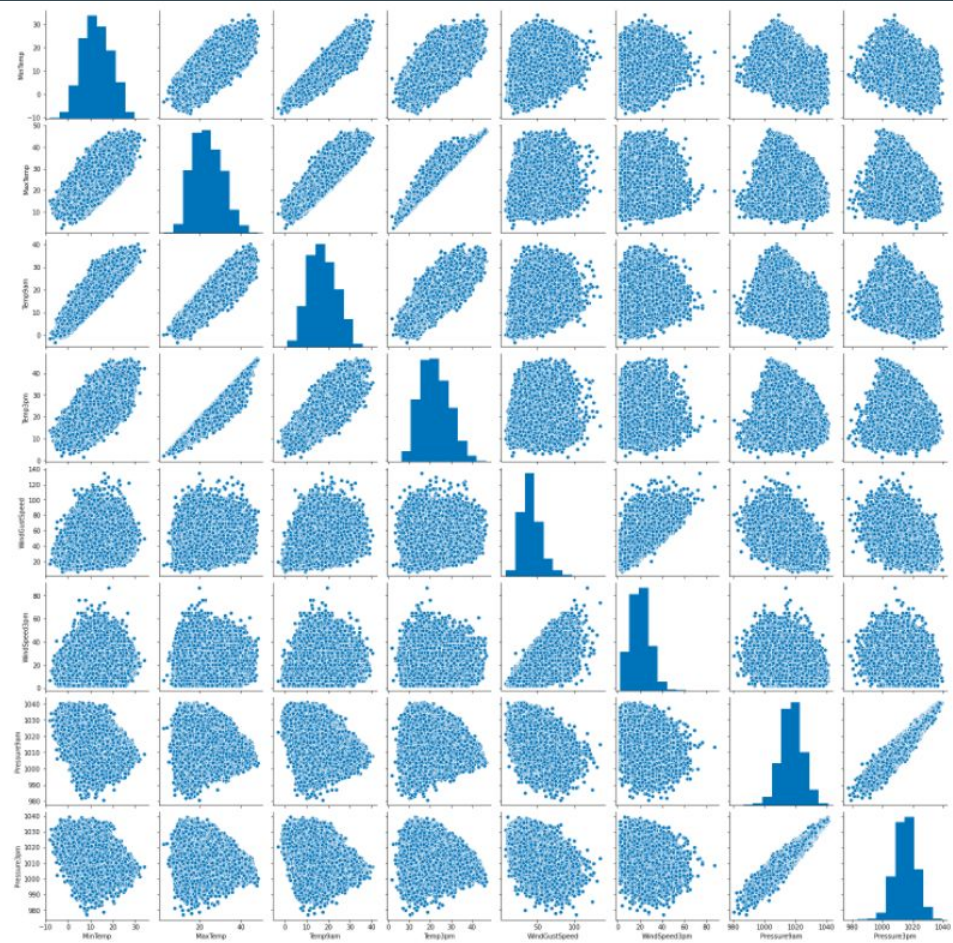
- This dataset contains about 10 years of daily weather observations from numerous Australian weather stations.
- It consists of 24 columns and 142193 rows. It has numerical, categorical and boolean data.
- It is an unbalanced dataset, as the target column RainTomorrow has 110316 "No" entries and 31877 "Yes" entries.

Visualizations

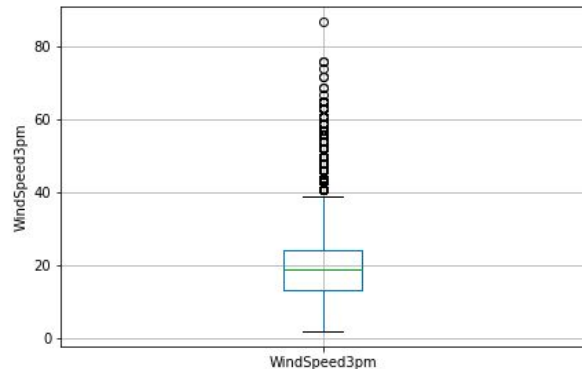
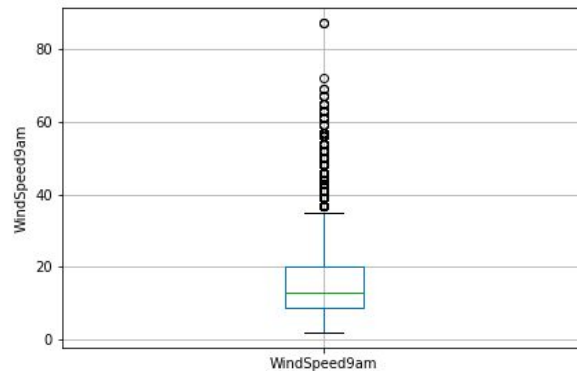
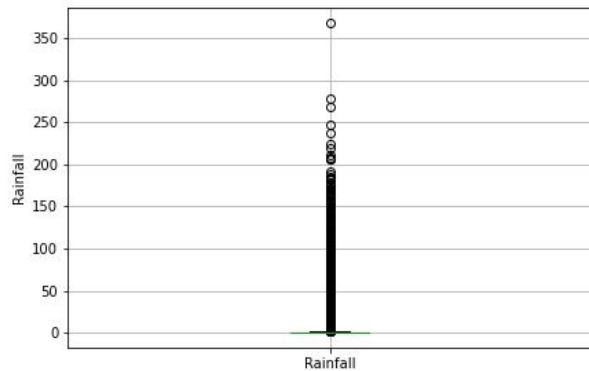


Visualizing the Target Variable





```
Text(0, 0.5, 'WindSpeed3pm')
```



Checking for Outliers

Data Preprocessing

- The dataset is stored in BigQuery.
- The preprocessing is done using Jupyter Notebooks in AI Platform
- The datatypes of the columns are corrected.
- The columns that have less than 60% of data are dropped.
- The rows that consist of missing values in any of the columns are dropped.
- Outliers are handled using top-coding approach.
- Categorical data is encoded using One Hot Encoding as Logistic Regression cannot handle categorical data.

Machine Learning Model

- A Logistic Regression model is trained using BigQuery ML.
- The query used to create the model is :

```
CREATE OR REPLACE MODEL
```

```
`thematic-flash-266714.australia_weather.model`
```

```
OPTIONS
```

```
( model_type="logistic_reg",  
  input_label_cols=["RainTomorrow"] ) AS
```

```
SELECT * FROM
```

```
`thematic-flash-266714.australia_weather.preprocessed`
```

Aggregate metrics ?

Log loss ? 0.3638

ROC AUC ? 0.8663

Score threshold

Positive class threshold ? 0.5542

Positive class true

Negative class false

Precision 0.7913

Recall 0.3882

Accuracy ? 0.8400

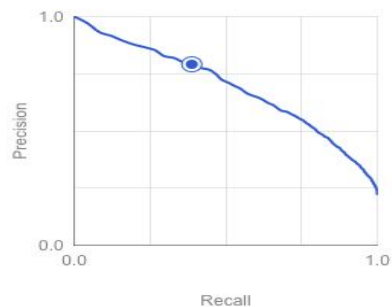
F1 score ? 0.5208

Confusion matrix

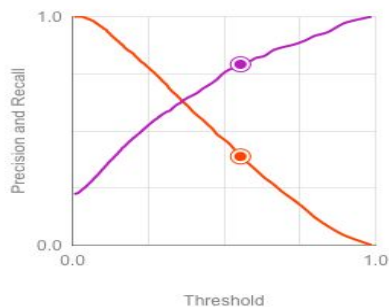
Actual labels	Predicted labels	
	true	false
true	38.82%	61.18%
false	2.96%	97.04%

Use this slider above to see which score threshold works best for your model.

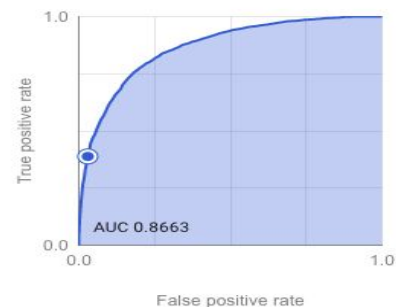
Precision-Recall curve



Precision and Recall vs Threshold



ROC curve

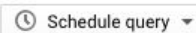
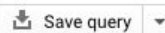


Predictions Using the Model

- We use the following query to evaluate the model

```
SELECT * FROM ML.PREDICT (MODEL `thematic-flash-266714.australia_weather.log_reg_model`,  
  
SELECT * EXCEPT(YEAR) FROM  
  
`thematic-flash-266714.australia_weather.preprocessed_new`  
  
WHERE YEAR < 2011 ))
```

```
1 SELECT * FROM ML.PREDICT (MODEL `thematic-flash-266714.australia_weather.log_reg_model`, (  
2   SELECT * EXCEPT(YEAR) FROM  
3   `thematic-flash-266714.australia_weather.preprocessed_new`  
4   WHERE YEAR < 2011 ))
```



This query will process 92.3 MB when run.



Query results



Query complete (2.8 sec elapsed, 92.3 MB processed)

Job information

Results

JSON

Execution details

Row	predicted_RainTomorrow	predicted_RainTomorrow_probs.label	predicted_RainTomorrow_probs.prob	MinTemp	MaxTemp	Rainfall	WindGustSpeed	WindSpeed9am	WindSpeed3pm	Hi
1	false	true	0.12808793434238738	-0.8	12.9	3.2	24.0	6.0	6.0	
		false	0.8719120656576126							
2	true	true	0.5297993431692399	20.2	25.3	2.2	43.0	2.0	13.0	
		false	0.47020065683076007							
3	false	true	0.48164343216357963	16.2	21.0	3.2	65.0	44.0	41.0	
		false	0.5183565678364204							
4	false	true	0.2119057304407178	18.7	25.4	1.2	39.0	19.0	20.0	
		false	0.7880942695592822							
5	false	true	0.369439295879749	18.5	26.4	3.2	37.0	6.0	11.0	
		false	0.630560704120251							
6	false	true	0.37659448765017495	15.0	21.7	3.2	52.0	30.0	24.0	
		false	0.623405512349825							
7	false	true	0.4393126975143989	1.4	12.9	1.6	26.0	9.0	9.0	

Rows per page:

100

1 - 100 of 27902

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Evaluating the Model

- We use the following query to evaluate the model

```
SELECT * FROM ML.EVALUATE(MODEL `thematic-flash-266714.australia_weather.log_reg_model`)
```

1 <code>SELECT * FROM ML.EVALUATE (MODEL `thematic-flash-266714.australia_weather.log_reg_model`)</code>						
<div><div> Run</div><div> Save query</div><div> Save view</div><div> Schedule query</div><div> More</div></div>						
<div>Query results<div> SAVE RESULTS</div><div> EXPLORE DATA</div></div>						
Query complete (0.2 sec elapsed, 0 B processed)						
<div>Job information<div>Results</div>JSONExecution details</div>						
Row	precision	recall	accuracy	f1_score	log_loss	roc_auc
1	0.7544738725841088	0.4612691466083151	0.8456711442298265	0.5725149375339489	0.36384063927514204	0.8663436563436564

Conclusion

- The accuracy of the model is 0.85 and precision is 0.75 which is good for a classification model.
- But the recall and f1-score is low, which implies that the model needs improvement.
- The performance of the model suffers due to the unbalanced dataset.

THANK YOU!