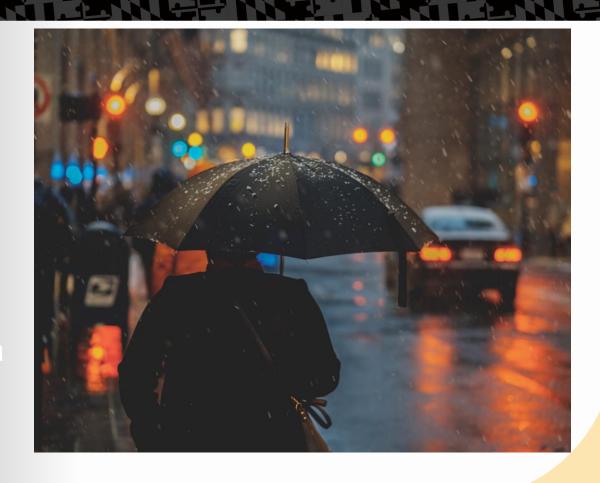


Analysis and Prediction of Rain fall

By Sivaji Ganesh Konka





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Introduction & Abstract

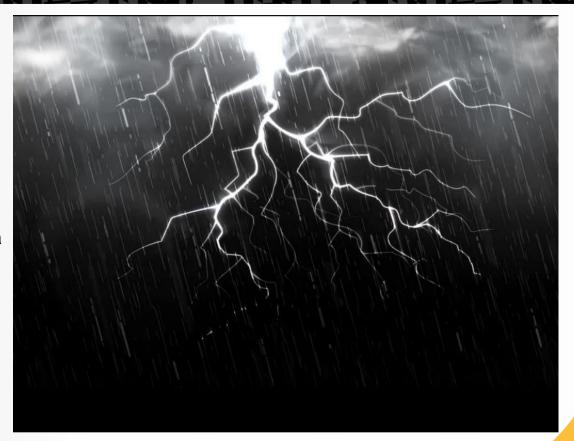


Introduction:

Rainfall prediction is an important part of weather forecasting. It involves using multiple methods and strategies to forecast the amount of precipitation that will fall in a given location.

Also, Rainfall prediction accuracy has improved significantly over the years as a result of technological improvements and the development of sophisticated models and algorithms.

This information is essential for various sectors like agriculture, water management, disaster management, and more.





Abstract

Rainfall forecast is influenced by several factors, including atmospheric pressure, humidity, wind patterns, and temperature.

In general, data must be collected from a variety of sources, including meteorological stations, satellites, and radar systems, in order to predict rainfall. then, the data is analyzed using statistical models and machine learning algorithms to identify patterns and trends that can be used to forecast future rainfall events.

In this Project, we are trying to develop a predictive model on the Rainfall Dataset in order to forecast whether or not it will rain in Australia tomorrow. The collection of data contains around ten years of daily weather measurements from numerous places throughout Australia.





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Data Source & Methodology



Data Source:

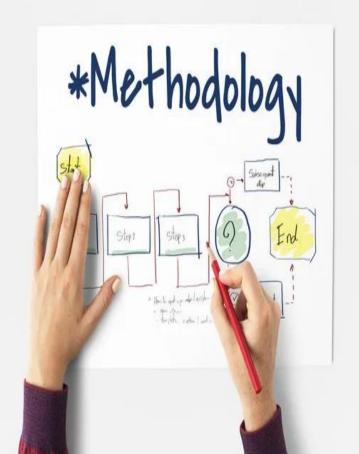
- Here the data is collected from Australian Government Bureau of Metrology.
- Data source: <u>http://www.bom.gov.au/climate/dwo/</u> and http://www.bom.gov.au/climate/data.
- Definitions adapted from <u>http://www.bom.gov.au/climate/dwo/IDCJDW0</u> <u>000.shtml</u>
- Data consist of 145460 rows and 23 columns
- Copyright Commonwealth of Australia 2010, Bureau of Meteorology

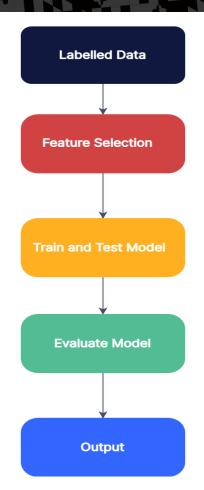
Date	Location	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	 Humidity9am	Humidity3pm
0 2008- 12-01	Albury	13.4	22.9	0.6	NaN	NaN	W	44.0	W	 71.0	22.0
1 2008- 12-02	Albury	7.4	25.1	0.0	NaN	NaN	WNW	44.0	NNW	 44.0	25.0
2 2008- 12-03	Albury	12.9	25.7	0.0	NaN	NaN	WSW	46.0	W	 38.0	30.0

Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm (RainToday	RainTomorrow
1007.7	1007.1	8.0	NaN	16.9	21.8	No	No
1010.6	1007.8	NaN	NaN	17.2	24.3	No	No
1007.6	1008.7	NaN	2.0	21.0	23.2	No	No

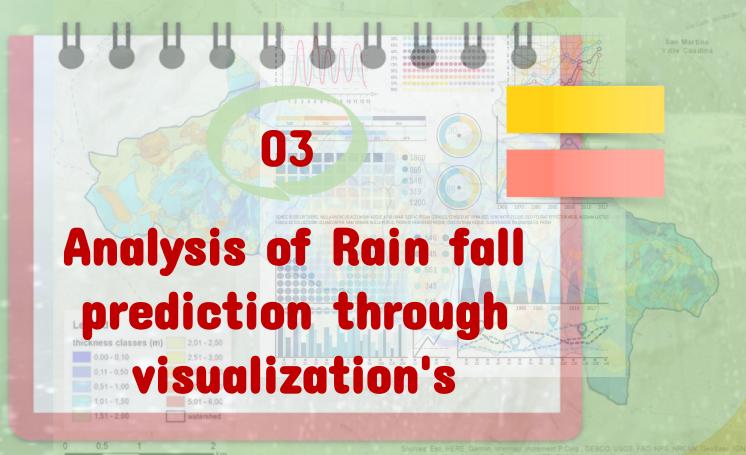












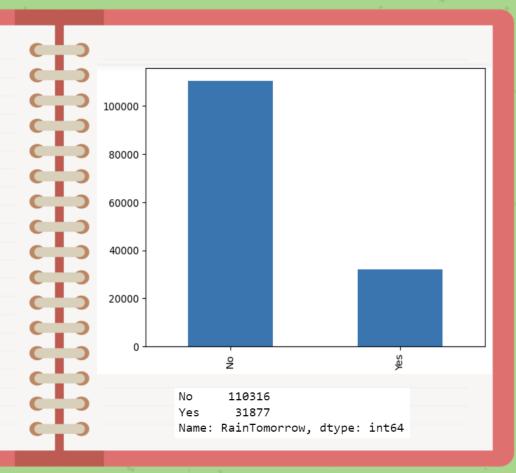
Sources: Esri, HERE, Garmin, Intermap Increment P Corp., GEBCO, USGS, FAO, NPS, NRCAM, GeoBase, IC Kadaster NL, Ordnance Surrey, Esri Japan, METI, Esri China (Hong Kong), (c) OpenStreeMap contributors, at the GIS User Community.



- ✓ Plotting Distribution of null values in the data set using pictorial representation.
- ✓ The columns: Evaporation, Sunshine, Cloud9am, cloud 3pm contribute highest number of null values in this data set.

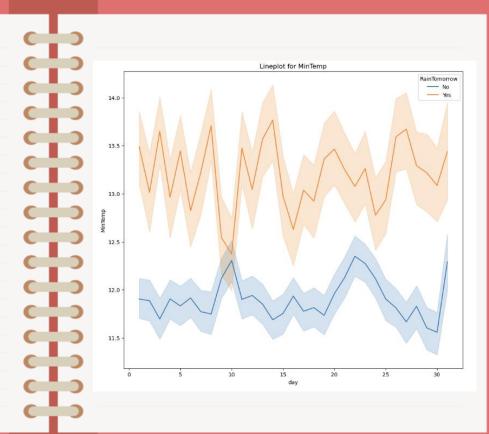
100012 109104 118196

- ✓ Plotting the instances of the target variable, that is rain tomorrow.
- ✓ As we can see we have more number of "Nos" in the given data



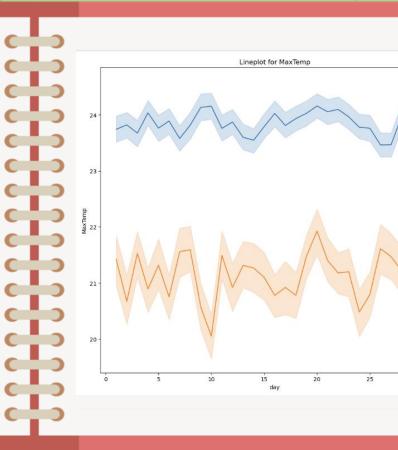


- ✓ Here we can see the correlation between the target variable and minimum temperature
- ✓ The graph depicts that when there is minimum temperature, we can expect the rain fall as per the data.



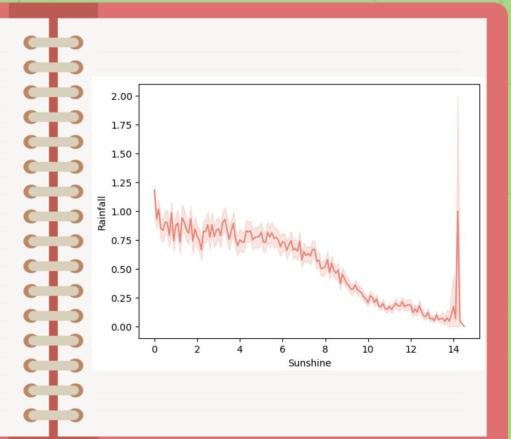


- ✓ Here we can see the correlation between the target variable and maximum temperature
- ✓ The graph depicts that when there is maximum temperature, we cannot expect the rain fall as per the data.



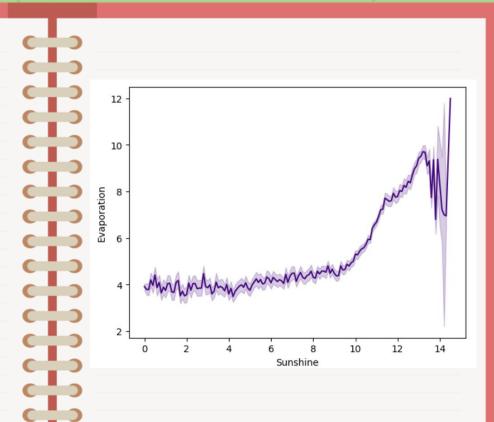


- ✓ Plotting Here we can see the correlation between rainfall and sunshine.
- ✓ The graph depits that when there is rainfall the sunshine gradually decreases.

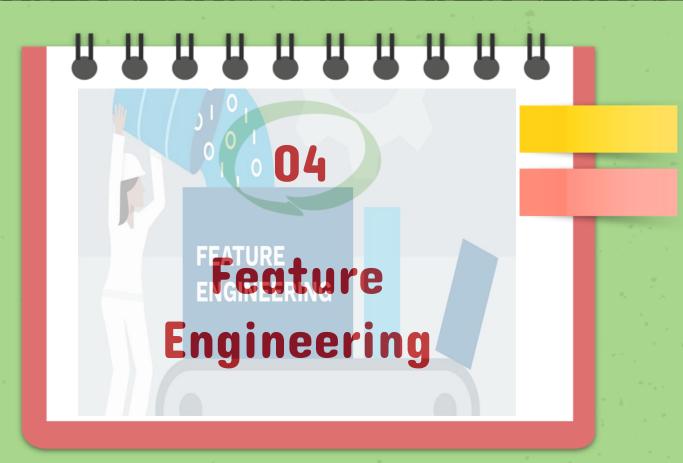




- ✓ we can see the correlation between Evaporation and sunshine.
- ✓ The graph depicts that when there is gradual increase in sunshine the amount of evaporation also increases.





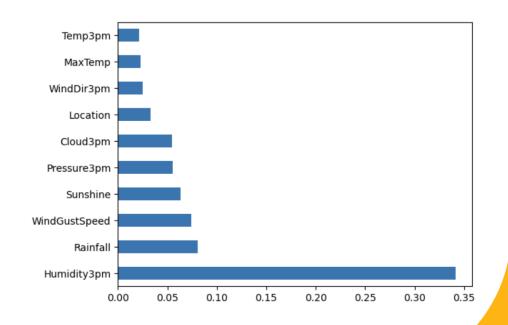




Feature Selection in Machine learning:

 We have approached this method of feature selection using XGB Classifier

 Here are the important features that have more impact on the target variable.







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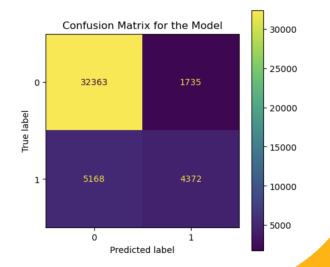
Model Prediction Algorithms



Logistic Regression

- Logistic regression is a machine learning approach that is used for binary classification. It is not, as the name implies, a set of rules for regression problems with nonstop results. Instead, logistic regression is the preferred method for binary classification. This gives a discrete binary value between 0 and 1. Logistic regression estimates probabilities using its core logistic feature to assess the relationship between dependent variables and one or more independent variables. These probabilities are then transformed into binary numbers for an estimation.
- ✓ Its range is fixed to be between zero and one. Additionally, R performs well when there is a large gap between the groups. here we have gained accuracy for LR is 84%.

		precision	recall	T1-score	support	
	0	0.86	0.95	0.90	34098	
	1	0.72	0.46	0.56	9540	
accura	су			0.84	43638	
macro a	ı∨g	0.79	0.70	0.73	43638	
weighted a	ı∨g	0.83	0.84	0.83	43638	





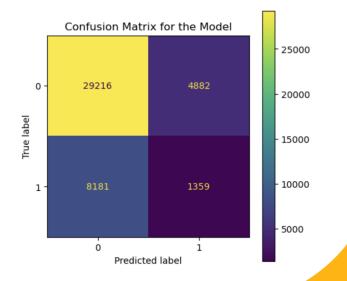
Random Forest

- ✓ Random forest is an algorithm used in behavior analysis that is built on modeling assumptions and wei decision trees. It consists of several decision trees representing a different example of the classification of data input into a random forest.
- ✓ Here we achieved Accuracy of 86% for Random Forest.

0	0.87	0.96	0.91	34075
1	0.77	0.50	0.61	9563
accuracy			0.86	43638
macro avg	0.82	0.73	0.76	43638
ighted avg	0.85	0.86	0.85	43638

recall f1-score

precision

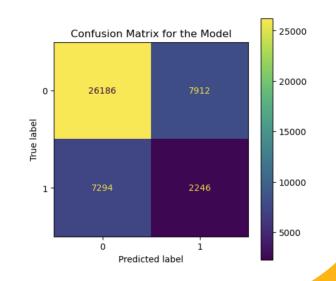




Decision Tree

- ✓ Decision trees are models that begin with root split into branches at nodes representing the predictor variables using the rules.
- ✓ Each node Is then divided into multiple nodes. The goal is to use the decision tree to divide the data space into dense and thin regions.
- ✓ As the number of splits increases the time required to build the tree also increases. And here we achieved Accuracy of 78% for Decision tree .

	precision	recall	f1-score	support
0	0.87	0.85	0.86	34075
1	0.51	0.54	0.52	9563
accuracy			0.78	43638
macro avg	0.69	0.70	0.69	43638
weighted avg	0.79	0.78	0.79	43638

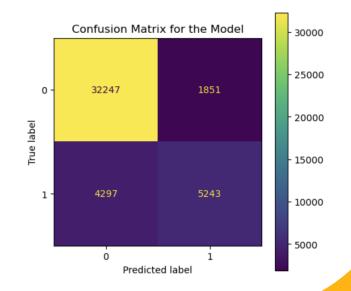




XGB Classifier

- XGBoost stands for extreme Gradient Boosting weighted avg and it's an open-source implementation of the gradient boosted trees algorithm. It has been one of the most popular machine learning techniques in Kaggle competitions, due to its prediction power and ease of use. It is a supervised learning algorithm that can be used for regression or classification tasks.
- ✓ we achieved Accuracy of 78% for GB Classifier

support	f1-score	recall	precision	
34098	0.91	0.95	0.88	0
9540	0.63	0.55	0.74	1
43638	0.86			accuracy
43638	0.77	0.75	0.81	macro avg
43638	0 85	9 86	0 85	ighted avg

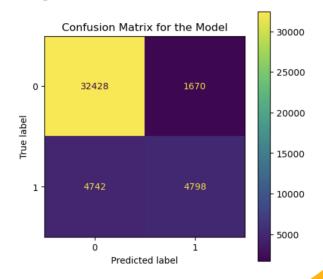




GB Classifier

- ✓ Gradient boosting is a machine learning technique used in regression and classification tasks, among others. It gives a prediction model in the form of an ensemble of weak prediction models, which are typically decision weit trees.
- ✓ When a decision tree is the weak learner, the resulting algorithm is called gradient-boosted trees; it usually outperforms random forest.
- A gradient-boosted trees model is built in a stage-wise fashion as in other boosting methods, but it generalizes the other methods by allowing optimization of an arbitrary differentiable loss function.
- ✓ we achieved Accuracy of 78% for GB Classifier

support	f1-score	recall	precision	
34098	0.91	0.95	0.87	0
9540	0.60	0.50	0.74	1
43638	0.85			accuracy
43638	0.75	0.73	0.81	macro avg
43638	0.84	0.85	0.84	ighted avg





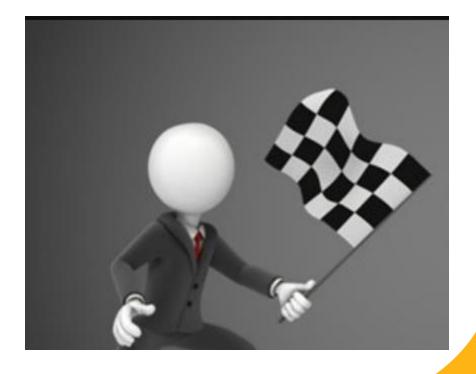


Conclusion



Conclusion:

Based upon the prediction results, we conclude that our prediction is correct, and the factors which we have considered have significant impact in predicting the Rainfall.







Future Scope

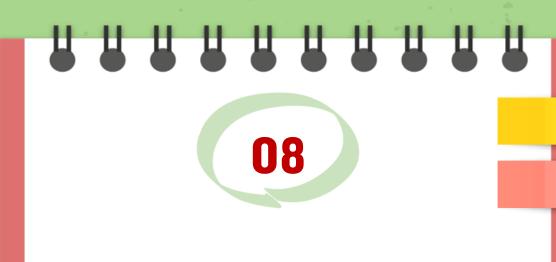


Future Scope

The proposed model can be inferred to predict the rainfall. If relevant data is available, the model can also be trained to take additional parameters. The model, with minor modifications, can be integrated into simulation software to better predict the rainfall and give some emergency warning about the weather. So that people can take necessary precautions.







References



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