#### Predict Rain in Australia.

by Y. Kostrov



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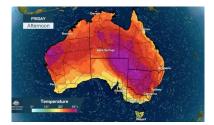
Overview

**Business Problem** 

Data

Modeling
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Overview



The purpose of this project is to use weather data set from Kaggle to predict rainfall for the next day, based on the data about today's weather.



#### **Business Problem**

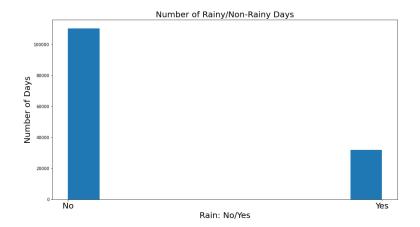
- Predicting rainy weather for the next day is a very important task.
- Usually weather is predicted by using complicated deterministic models involving partial differential equations.
- ► I will suggest a model that predicts weather by using Machine Learning.



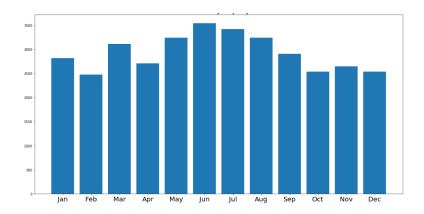
# Data Used in the Project

- ► This data set contains about 10 years of daily weather observations from many locations across Australia.
- RainTomorrow is the target variable to predict. It means did it rain the next day, Yes or No? This column is Yes if the rain for that day was 1mm or more.

# Number of Rainy and Sunny days.



# Rainy and Sunny days by Month.



I have built the following two classifiers

- Logistic Regression Classifier
- Random Forest Classifier
- KNeighbors Classifier
- Support Vector Machines Classifier
- XG Boost Classifier
- Naive Bayes Classifier



#### **Explanation of Recall**

- ► I used recall as a metric to select the base model and optimize by later on.
- Recall is defined as:

$$\mathsf{Recall} = \frac{\mathsf{True\ Positive}}{\mathsf{True\ Positive} + \mathsf{False\ Negative}} = \frac{\mathsf{True\ Positive}}{\mathsf{Total\ Actual\ Positive}}$$



## **Explanation of Recall**

Metrics

- Recall calculates how many of the Actual Positives our model captures by marking it as Positive (True Positive).
- Thus Recall is a better model metric when there is a high cost associated with False Negative.
- ▶ In our case False Negative is predicting "No Rain" when there is a "Rain Tomorrow".



#### Explanation of Recall

- For instance, in rain prediction.
- If it rains tomorrow (Actual Positive) is predicted as no rain tomorrow (Predicted Negative), then the person who relies on the prediction will be really upset since being unprepared for bad weather.

#### **Explanation of Precision**

- There is a secondary metric I will be watching, called "precision".
- Precision is defined as:

$$\mathsf{Precision} = \frac{\mathsf{True\ Positive}}{\mathsf{True\ Positive} + \mathsf{False\ Positive}} = \frac{\mathsf{True\ Positive}}{\mathsf{Total\ Predicted\ Positive}}$$

#### **Explanation of Precision**

- Precision describes how precise/accurate your model is out of those predicted positive, how many of them are actual positive.
- Precision is a good measure when we worry about the costs of False Positive.



Metrics

## **Explanation of Precision**

- In our rain prediction, a false positive means "No Rain" tomorrow (actual negative) has been identified as "Rain" tomorrow.
- It is not that bad, since a person will carry an umbrella or rain coat for nothing.

#### How Well Baseline Models Performed



#### Conclusions



## Ways to Improve the Project

- ▶ I would like to optimize the code.
- Learn more about weather and related data.



# THE END THANK YOU!