

Will It Rain Tommorow?

Predicting Rainfall in Australia with Neural Networks

Abhinav Raj
NA21PICS04

NEHRU ARTS AND SCIENCE COLLEGE KANHANGAD



Rainfall prediction is a critical task in meteorology with significant implications for **agriculture, water resource management, and disaster preparedness**. Traditional forecasting methods often struggle to handle the complexity and variability of weather patterns. In this case study, we explore a **data-driven approach** to **predict whether it will rain the next day (RainTomorrow)** using a **neural network model** trained on historical weather data from Australia. By leveraging modern machine learning techniques and extensive data preprocessing, this project aims to build a reliable and efficient rainfall prediction system that can support real-world decision-making.



Dataset Overview

[Dataset Link](#)

📍 Source: Australian Bureau of Meteorology

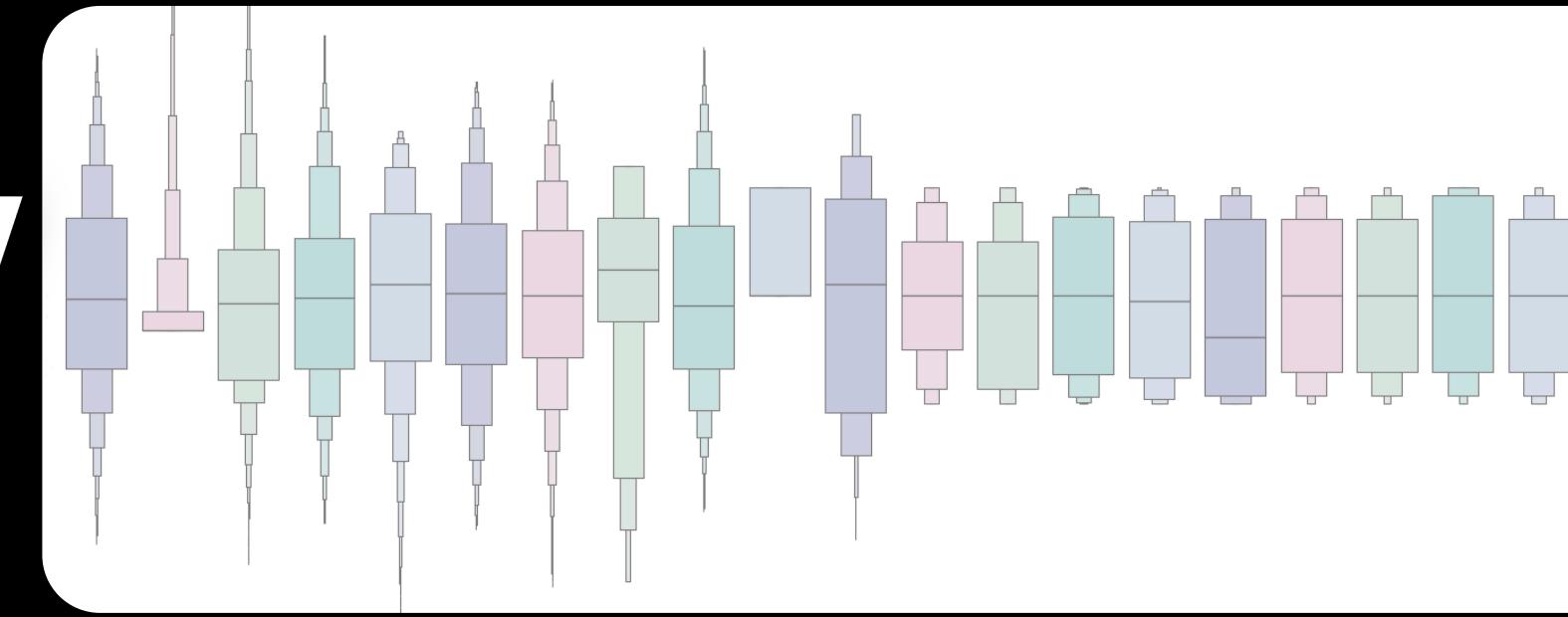
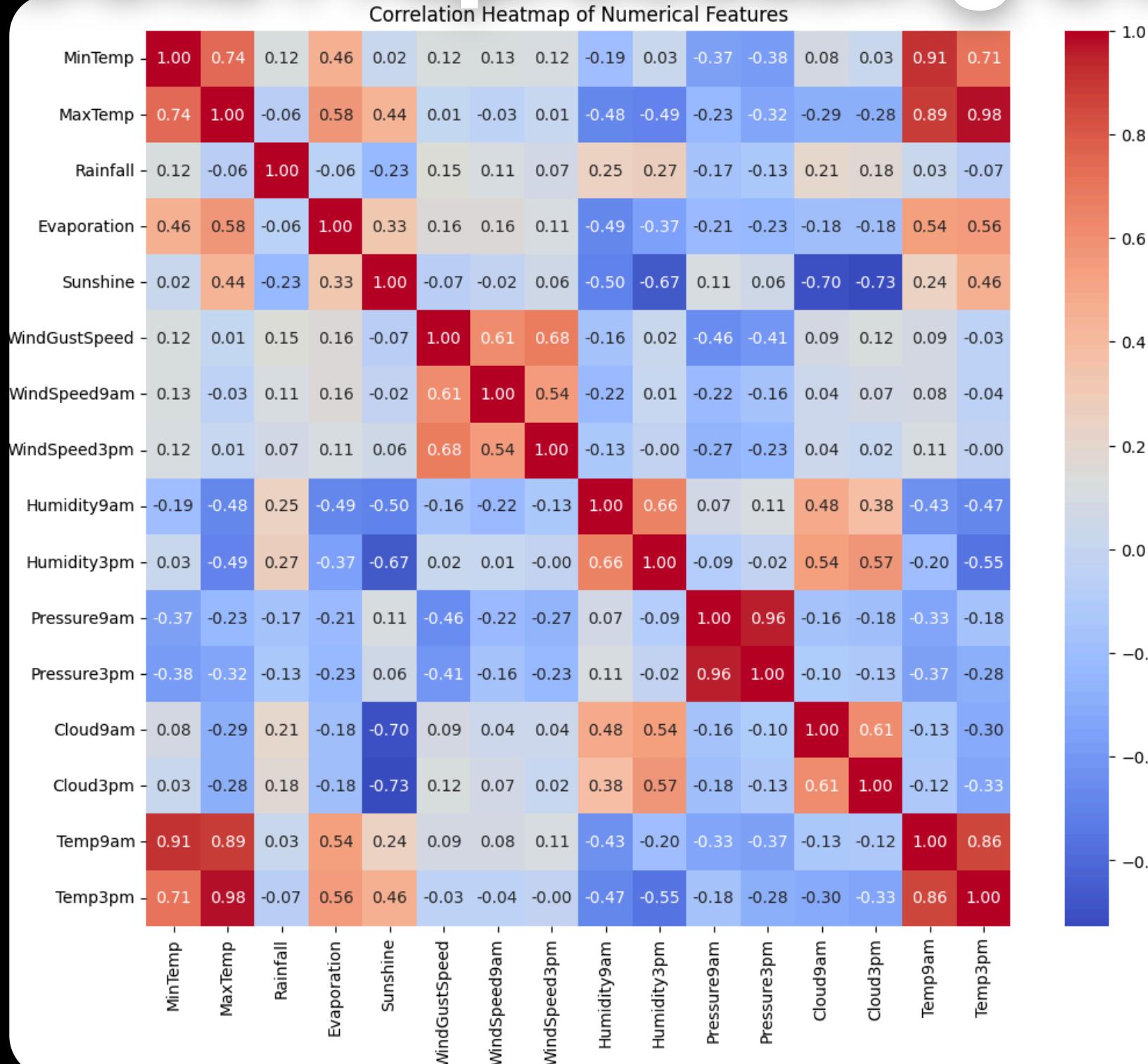
📅 Records: ~145,000 daily observations across various locations

🎯 Target Variable: RainTomorrow (Yes / No)

📊 Features Include:

- Temperature (Min, Max)
- Rainfall
- Humidity (9am, 3pm)
- Wind Speed & Direction
- Atmospheric Pressure
- Date & Location

Data Preprocessing Overview



- Removed rows with missing RainTomorrow; encoded as 0/1
- Balanced classes using bootstrapped oversampling
- Parsed Date → cyclic day, month (sine/cosine)
- Imputed missing values (median for numeric, mode for categorical)
- Removed outliers using IQR
- Encoded:
 - Location: label / one-hot
 - Wind directions: degree → sine/cosine
- Scaled numeric features (z-score)
- Created two datasets (22 & 69 features); stratified 80:20 split

Model Architecture & Training Overview

```
model = Sequential()
model.add(Dense(64, kernel_initializer='he_normal', activation='relu', input_dim=22))
model.add(BatchNormalization())

model.add(Dense(32, kernel_initializer='he_normal', activation='relu'))
model.add(BatchNormalization())

model.add(Dense(16, kernel_initializer='he_normal', activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.1))
```

⚙️ Training Configuration

- Optimizer: Adam (LR = 0.00009)
- Loss: Binary Crossentropy
- Metrics: Accuracy, AUC
- Validation Split: 20%
- Early Stopping: Patience = 20, Δ = 0.001
- Max Epochs: 150
- Batch Size: 32

Used BatchNormalization and Dropout to improve generalization.

Trained both versions: Label Encoded & One-Hot Encoded.

Visualized training metrics (accuracy & AUC) to monitor performance.

Streamlit App

Rain Tomorrow Prediction

Predict whether it will rain tomorrow in Australia using weather features and a trained neural network model. Please provide the following input values:

Select Date: 2025/06/10

Select Location: Albany

Min Temperature (°C): 10.00

Rainfall (mm): 0.00

Did it rain today? No Yes

Gust Direction: N

Wind Direction at 9 AM: N

Wind Direction at 3 PM: N

Wind Gust Speed (km/h): 30.00

Wind Speed at 3 PM (km/h): 20.00

Humidity at 9 AM (%): 0

Humidity at 3 PM (%): 0

Pressure at 9 AM (hPa): 1015.00

Cloud Cover at 3 PM (%): 0

Temperature at 3 PM (°C): 20.00

Prediction Result

Will it rain tomorrow? Yes

Model Confidence: 69.40%

A simple web app built with Streamlit lets users input weather data and instantly predicts whether it will rain tomorrow using the trained ANN model. It features a clean UI, real-time inference, and visual feedback.

Conclusion

✓ Developed a Neural Network model to predict RainTomorrow using Australian weather data.

🔍 Applied robust preprocessing:

- Handled missing values
- Normalized numerical features
- Categorical and cyclical encoding

⚖️ Tackled class imbalance using balanced class weights

🧪 Compared encoding methods:

- Label Encoding: Moderate performance
- One-Hot Encoding: Best performance with 82% accuracy

📈 Achieved high recall ($\approx 89\%$) for predicting rainy days — crucial for real-world utility

💡 Highlighted the importance of feature representation and data balancing in classification tasks



Thank You