

# Climate Factors Forecast

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# Overview: Climate Change in the World and in Singapore

In the world:

- Long-lasting change of the temperature and weather patterns.
- Many human activities contribute to climate change.
- Has become a very challenging global issue.

In Singapore:

- Experiencing more frequent extreme weather.
- Impacted Singapore's temperature and precipitation patterns.



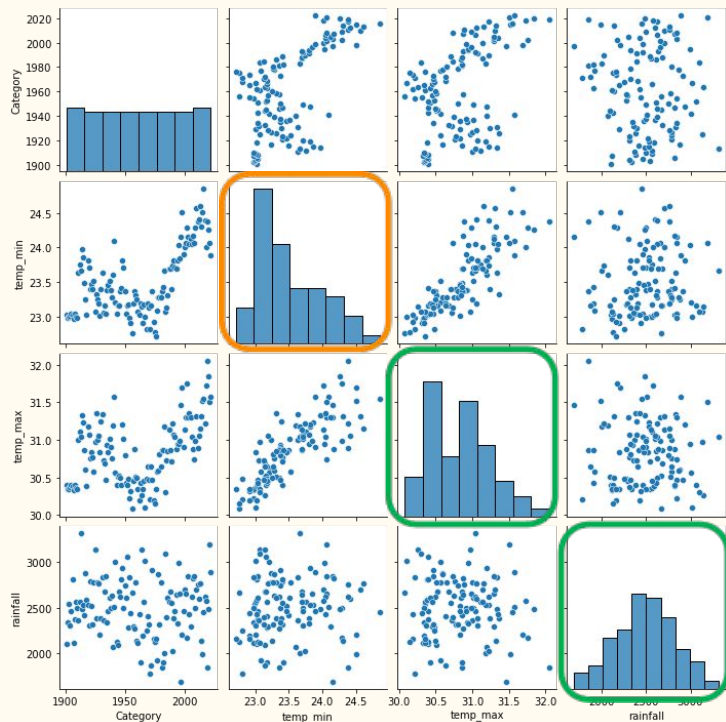
# Introduction: Why temperature and rainfall in Singapore?

- Singapore is located near the equator, very warm all year round.
- Singapore is also close to the ocean, with frequent precipitation.
- Due to these features, we wish to study the changes in precipitation, maximum temperature and minimum temperature in Singapore from 1901 to 2022.
- Temperature changes are the main manifestation of climate change
- Climate change can also lead to changes in precipitation patterns.

# Data Description

Variables	description	type	e.g.
Category	Year	integer	1901, 1902,..., 2022
temp_min	Minimum temperature for the year	numeric (float)	23.03, 22.98,..., 23.89
temp_max	Maximum temperature for the year	numeric (float)	30.4, 30.34,...,31.58
rainfall	Total amount of rain collected for the year (mm)	numeric (float)	2106.77, 2334.47,..., 2894.53

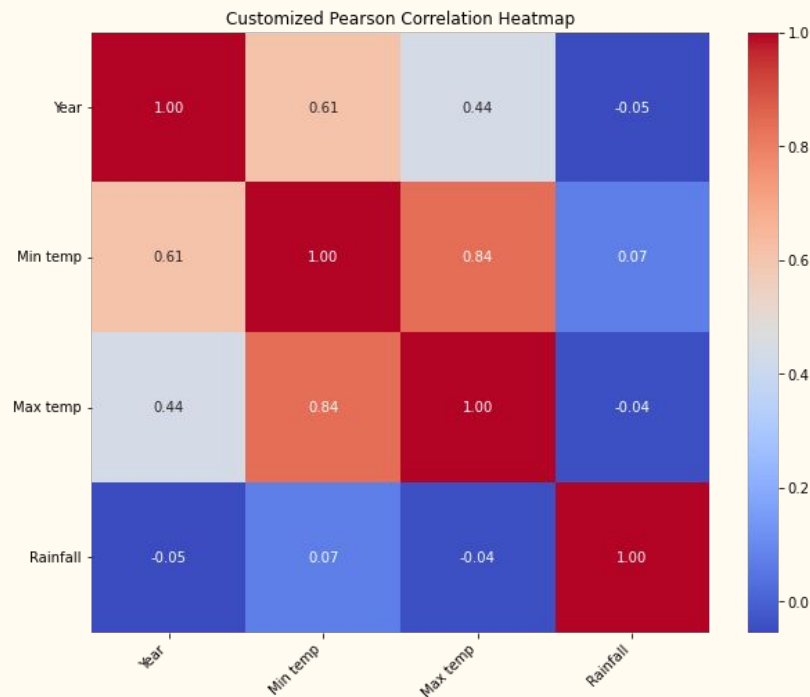
# Data Exploration



## Distribution:

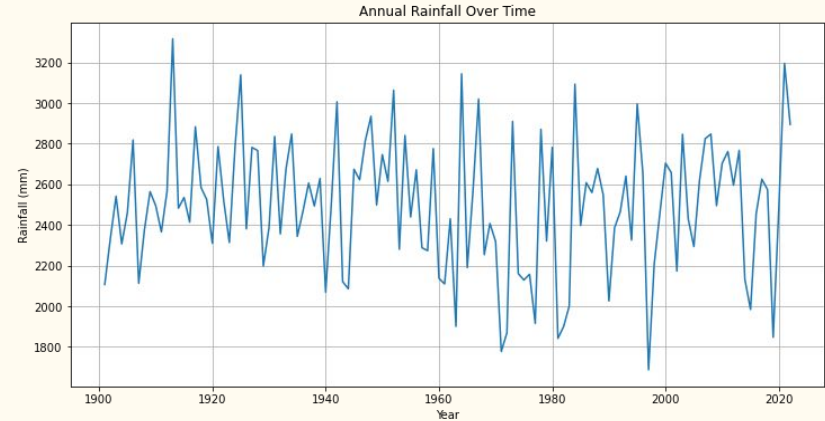
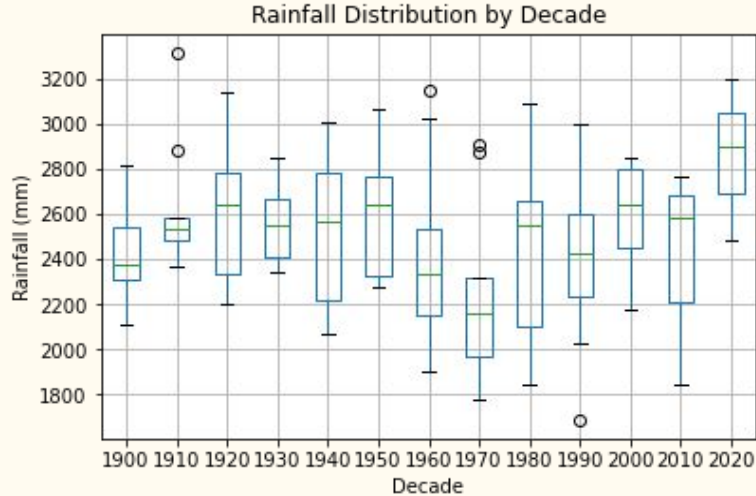
- Minimum temperature is more **right-skewed**.
- Maximum temperature and rainfall resemble more of a **normal distribution**.

# Data Exploration



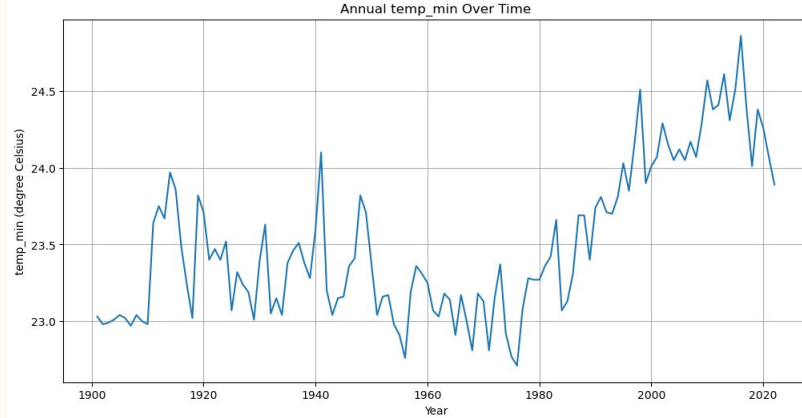
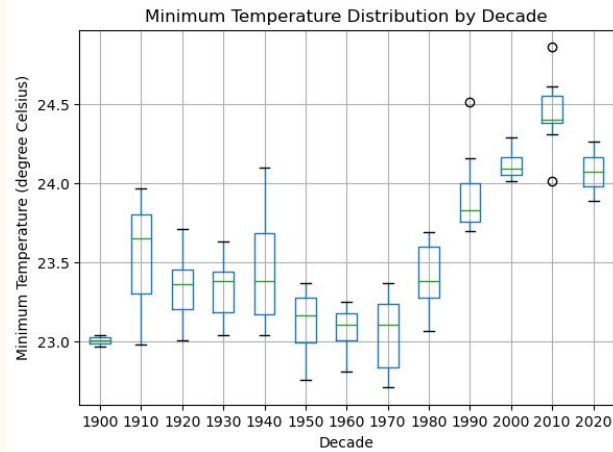
- Strong correlation between minimum temperature and maximum temperature
- Relatively strong correlation can also be observed between rainfall and minimum temperature
- Weak/no correlation between rainfall and the other variables

# Data Exploration



- Some outliers:
  - Year 1913  $\Rightarrow$  High rainfall exceeding 3300mm
  - Year 1997  $\Rightarrow$  Low rainfall of less than 1800mm
- Rainfall hovers in the range of 2200 to 2800mm

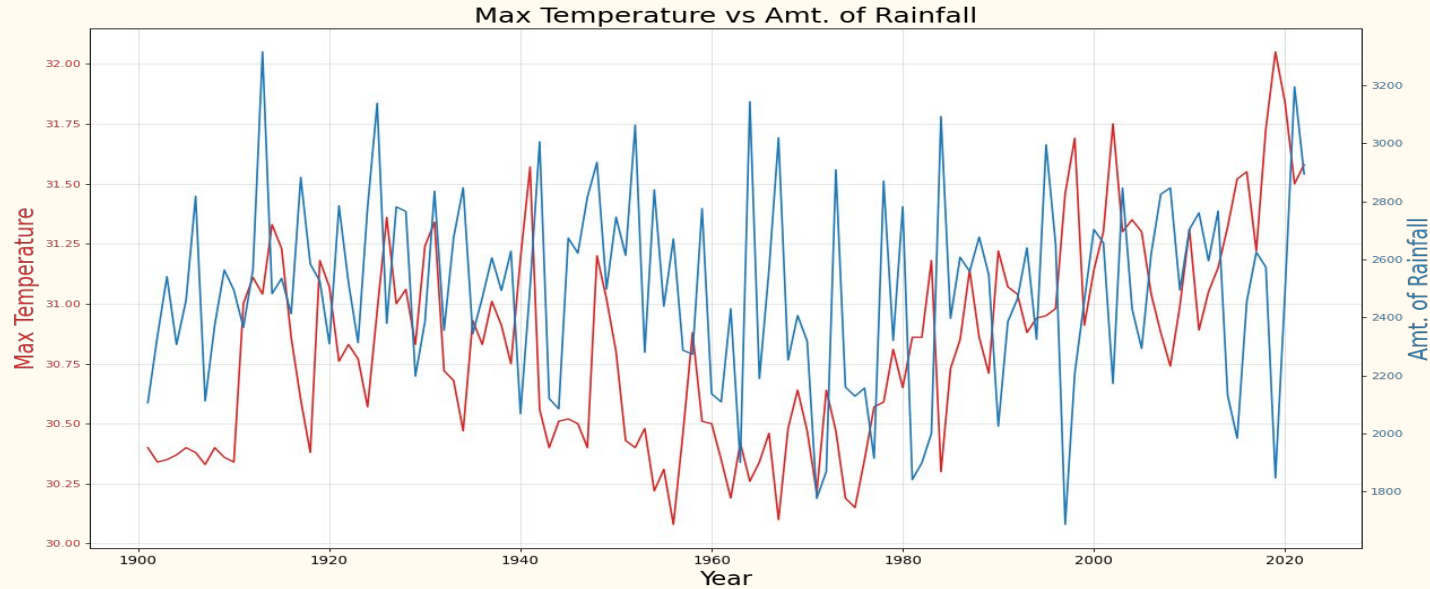
# Data Exploration



- Range of the temperatures seems relatively big in these decades:
  - 1910-1920, 1940-1950, and 1970-1980
- But short range of temperatures in the experienced in 1900-1910 decade
- More outliers and higher temperatures experienced in the recent decades



# Data Exploration



- Decreasing trend for both maximum temperature and rainfall in from year 1940 to 1960.
- Unusual years (e.g. 1934) which has high rainfall but low maximum temperature

# Methodology

- Two analysis methods explored.
- First method-Machine Learning Model
  - Year as predictor and Climate Factors as response
  - Linear Regression Model
  - Support Vector Machine
- Second method-Time Series Analysis
  - ARIMA (Autoregressive Integrated Moving Average) model



# Methodology

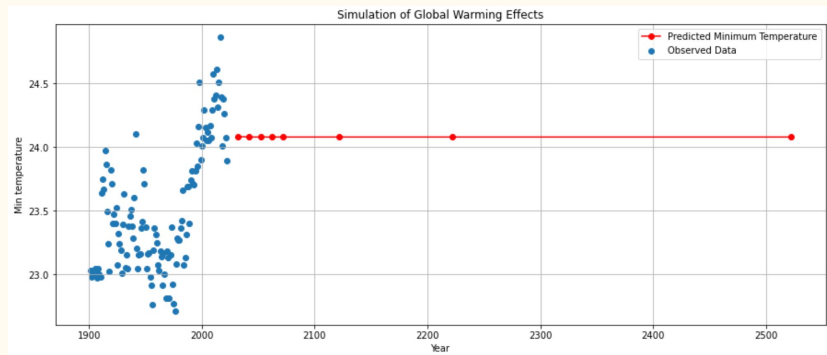
- For the machine learning models, the dataset is split into training and testing sets.
- Models are built using the train set and evaluated on the test set.
- The evaluation is done using MSE (Root Mean Squared Error), and R-Squared.
- The chosen model (linear regression or support vector machine) is used to forecast/predict the climate factor over the future years.

# Methodology

- For the time series analysis we make use of the ARIMA model.
- Before implementing the model, initial analysis such testing whether the data is stationary or not, in addition initial model parameters were also determined by plotting the ACF and PACF plots.
- The parameters determined are,  $p$ , the number of autoregressive terms in the model,  $d$ , is the number of differentiations applied on the time series values and  $q$ , refers to the number of moving average terms in the model.
- The models implemented are evaluated using MSE (Mean Squared Error) and the residual plots of the climate factors.

# Results-Forecasting

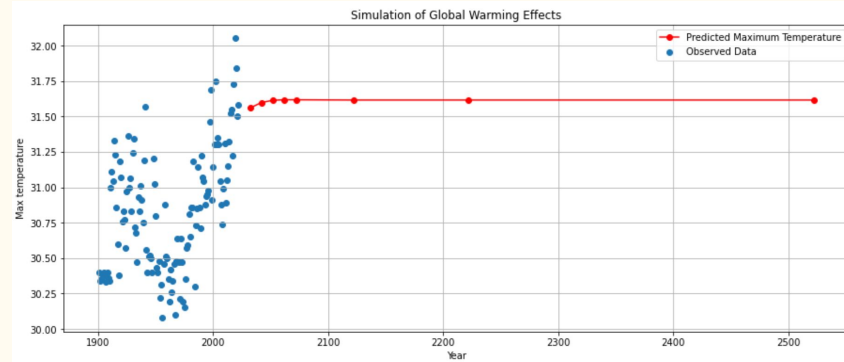
## 1. min\_temp



ARIMA forecast

For the Time Series ARIMA model, we chose AutoRegressive Order=2, Differencing Order=1, and Moving Average Order=9, and the predicted future minimum temperature has a smooth trend and is expected to remain at 24.1 degrees Celsius for a long time.

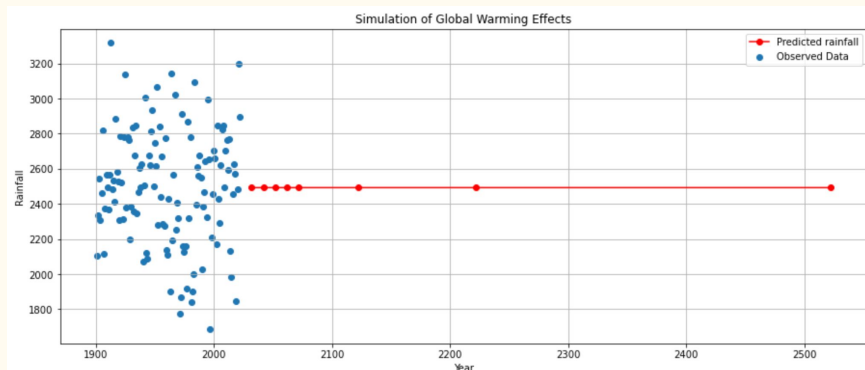
## 2. max\_temp



ARIMA forecast

For the Time Series ARIMA model, we chose AutoRegressive Order=2, Differencing Order=1, and Moving Average Order=6, and the predicted future maximum temperature has a smooth trend and is expected to remain at 31.6 degrees Celsius for a long time.

### 3. rainfall



ARIMA forecast

For the Time Series ARIMA model, we chose AutoRegressive Order=1, Differencing Order=1, and Moving Average Order=2, and the predicted future total amount of rain collected for the year has a smooth trend and is expected to remain at 2500 mm just like the prediction of support vector regression.

# Time series method-results

MSE:

models	min_temp	max_temp	rainfall
ARIMA regression	<b>0.051</b>	<b>0.070</b>	117640.904

ARIMA Parameters:

indicators	ACF/PACF	ARIMA parameters
min_temp	ACF: 9th-order trailing PACF: 2nd-order truncation	ARIMA (p=2, d=1, q=9)
max_temp	ACF: 6th-order trailing PACF: 2nd-order truncation	ARIMA (p=2, d=1, q=6)
rainfall	ACF: 1st-order truncation PACF: 1st-order truncation	ARIMA (p=1, d=1, q=2)



# Machine learning method-results

Minimum temperature:

Models	R-squared	MSE
Linear regression	-5.943	0.440
Linear SVR	-5.627	0.420
Non-linear SVR (using RBF kernel)	-0.065	0.067

Maximum temperature:

Models	R-squared	MSE
Linear regression	-2.650	0.358
Linear SVR	-2.640	0.357
Non-linear SVR (using RBF kernel)	-2.481	0.342

Rainfall:

Models	R-squared	MSE
Linear regression	-0.102	140704.223
Linear SVR	-0.062	135581.193
Non-linear SVR (using RBF kernel)	-0.017	129825.557

# Conclusion - short run effect

Analysis of 5-year average:

**average minimum temperature projected to rise by 0.038 degree celsius**

**average maximum temperature projected to fall by 0.10 degree celsius**

**Significance:**

→ decrease in maximum temperature could be **attributed to the La Niña effect** present in 2020-2022, which may lead to temporarily lower temperature.

→ Given the **high level of humidity in Singapore**, temperature increase will be acutely felt.

→ increase in temperature can **negatively impact cognitive performance**(Brink et al.,2020), which may affect productivity and economic growth.

# Conclusion - short run effect

The average rainfall is projected to fall by 4% in 5 years

## **Significance:**

→ risk to Singapore's already **limited water supply**

→ may pose risk of **inflationary pressures** in future

# Conclusion - proposed action

To deal with temperature increase:

- Need for **stop-gap measures** to help residents deal with rising temperatures in their day-to-day activities.
  - there will likely be a need for greater heat insulation and flexibility in dressing in schools and workplaces.

To deal with decrease in rainfall:

- **diversify its water sources**, including **importing more from other countries**.

Overall, it is clear that **increased government expenditure** is required to help residents manage the effects of climate change in the short term.

# Conclusion - long run effect

**Average minimum temperature is projected to decrease by 0.20 degree celsius.**

**Average maximum temperature is projected to increase by 0.07 degree celsius.**

**average rainfall is projected to decrease by 0.12%.**

## **Significance:**

→ rise in temperature and **increased temperature volatility;**

→ temperature changes projected for next decade are more adverse than 5-year projections.

→ rainfall change projected for next decade is more moderate compared to the 5-year projection.

However, weather changes may become more pronounced with the increasing environmental damage brought on by expected rise in military efforts worldwide, such as in the Russia-Ukraine War and Israel-Palestinian conflict.

# Conclusion - proposed actions

- **Stop-gap measures insufficient; systemic change required**
- Green Plan 2030: restore nature in its landscape
  - increased vegetation can help to reduce surface temperature and mitigate temperature changes in the next decade.

# Future Works

- There is scope to **reflect greater complexity in how climate change data evolves over time** in machine learning models such as linear regression, SVR models.
  - engineered features such as moving average, **lagged predictors** in SVR, and the **El Niño and La Niña effects can be incorporated** into our models to improve predictions.
- Last but not least, there is merit in performing **cross-sectional multivariate analysis**,
  - e.g. by using measures of pollution as predictors to rainfall and maximum and minimum temperature to establish relationships between different predictors and the response variables.