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# Purpose

To compare between the use of SARIMA and SARIMAX model in weather forecasting using Python.



### Data Overview

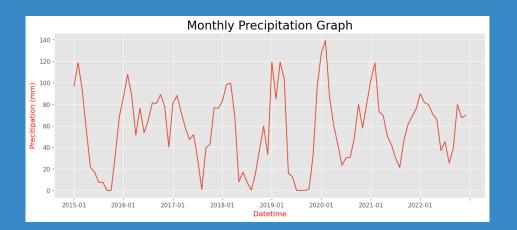
| Weather data have 2913 rows and time from 2015 to 2022 *red blocked are the selected variables |                            |       |                      |  |  |  |
|--|----------------------------|-------|----------------------|--|--|--|
| NAME   | Station Name               | WDSP  | Windspeed (m/s)      |  |  |  |
| Latitude   | Latitude of Station        | PRCP  | Precipitation (Inch) |  |  |  |
| Longitude  | Longitude of station       | DEWP  | Dewpoint (Farenheit) |  |  |  |
| DATE   | Date the data recorded     | SNDP  | Snowdepth            |  |  |  |
| SLP  | Slope                      | VISIB | Visibility           |  |  |  |
| TEMP   | Temperature<br>(Farenheit) | Etc.  | Many more            |  |  |  |

Data source: https://www.ncei.noaa.gov/



# Data Preprocessing

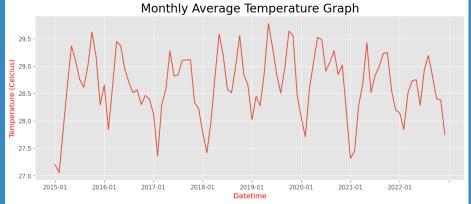




#### **Time Series**

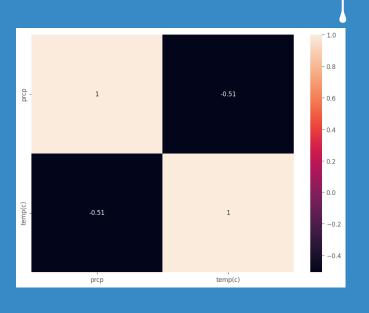
The precipitation is **high at the beginning and final of year**, but **low at the middle of the year**. The
season is 12 months, so **s = 12**.

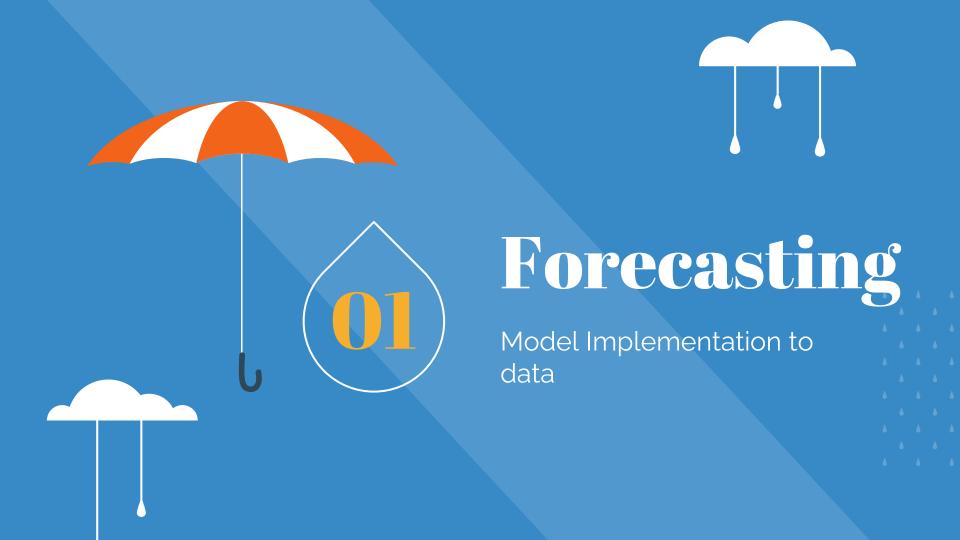






The correation between precipitation and temperature is high but negative (-0.51). That makes the relation between them is inverse. So, if the **precipitation is high, then the temperature is low**, vice versa. These 2 variables selected because they have high correlation.





#### **Need to Differentiate?**

Time series chart beside is showed data after first and second differencing, but **the differencing don't need to be done** because the data is already stationer (p-value < 0.05)

ADF Statistic: -6.549938 p-value: 0.000000 Critical Values:

> 1%: -3.505 5%: -2.894 10%: -2.584



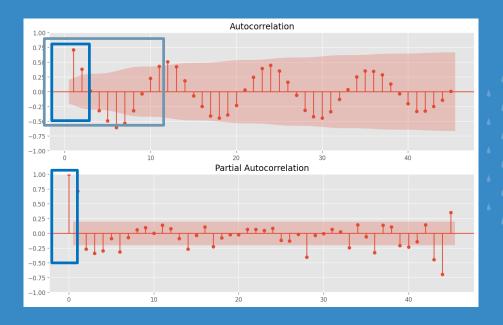
#### **Parameter?**

From ACF (Autocorrelation) and PACF (Partial Autocorrelation) Chart, selected:

- p = 1 or 2
- q = 1 or 2
- d = 0 (not differentiate)
- s = 12

I use SARIMAX (1,0,1)x(1,0,1,12) then, because this is the best model after tried many models.





#### Parameter?

After that, I train the data and I select:

- p = 1
- q = 1

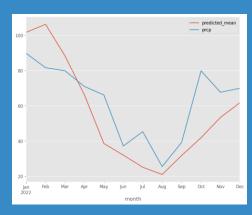
Because the model have the lowest p-value and AIC value.



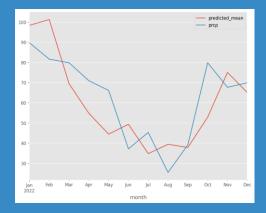
| SARIMAX Results                                 |                           |               |            |      |       |            |          |  |
|---|---------------------------|---------------|------------|------|-------|------------|----------|--|
| Dep. Variable:                                  | prcp No. Observations: 84 |               |            |      |       |            |          |  |
| Model:  | SARIMAX(                  | 1, 0, 1)      | x(1, 0, 1, | 12)  | Log   | Likelihood | -384.588 |  |
| Date:   | Tue, 11 Ap                | r <b>2023</b> |            |      |       | AIC        | 779.177  |  |
| Time:   | 00:04:31                  |               |            |      |       | BIC        | 791.331  |  |
| Sample:   | 01-01-2015                | 5             |            |      |       | HQIC       | 784.062  |  |
|   | - 12-01-202               | 21            |            |      |       |            |          |  |
| Covariance Type: opg                            |                           |               |            |      |       |            |          |  |
| coef  | std err                   | z             | P> z       | 0.02 | 25    | 0.975]     |          |  |
| <b>ar.L1</b> 0.7935                             | 0.080                     | 9.888         | 0.000 0    | 636  | C     | ).951      |          |  |
| <b>ma.L1</b> -0.1159                            | 0.150                     | -0.77!        | 0.438 -    | .409 | ) (   | 0.177      |          |  |
| ar.S.L12 0.9989                                 | 0.050                     | 19.90         | 2 0.000 C  | 901  | 1     | 1.097      |          |  |
| ma.S.L12 -0.9523                                | 3 1.082                   | -0.880        | 0.379 -    | .072 | 2 1   | 1.168      |          |  |
| <b>sigma2</b> 422.09                            | 39 425.396                | 0.992         | 0.321 -    | 11.6 | 566 1 | 1255.854   |          |  |
| Ljung-Box (L1) (Q): 0.20 Jarque-Bera (JB): 2.89 |                           |               |            |      |       |            |          |  |
| Prob(Q):  | 0.65                      | P             | rob(JB):   | 0    | ).24  |            |          |  |
| Heteroskedastici                                | <b>ty (H):</b> 0.95       |               | Skew:      | -(   | 0.03  |            |          |  |
| Prob(H) (two-si                                 | <b>ded):</b> 0.88         | K             | urtosis:   | 3    | 3.91  |            |          |  |
|   |                           |               |            |      |       |            |          |  |
|   |                           |               |            |      |       |            |          |  |







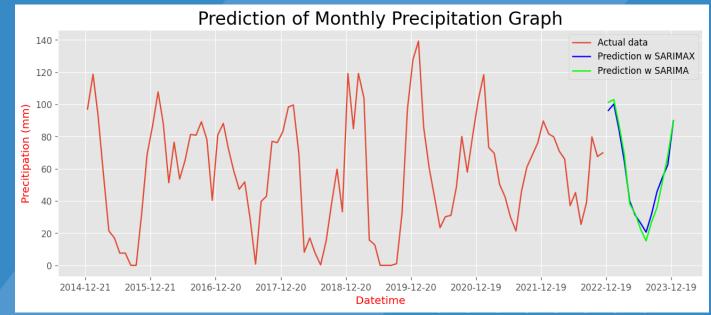
Without exogenuos variable (temperature) (SARIMA)



With exogenuos variable (temperature) (SARIMAX)

## **Forecasting**



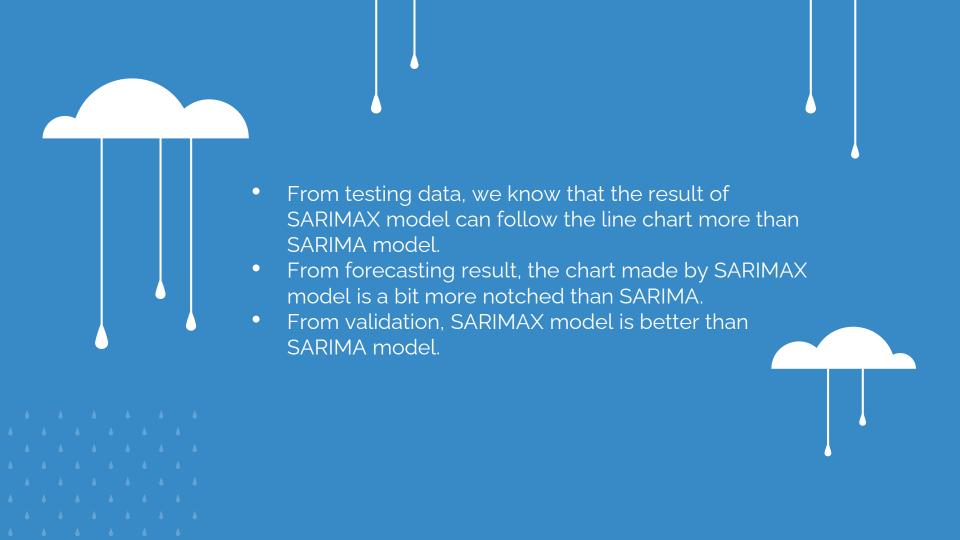


Forecasting with SARIMA & SARIMAX models

## Validation

|         | Root Mean Error<br>(RMSE) | Mean Absolute Error<br>(MAE) | Mean Absolute<br>Percent Error<br>(MAPE) |
|---------|---------------------------|------------------------------|--|
| SARIMA  | 17.85                     | 14.58                        | 23.18                                    |
| SARIMAX | 14.60                     | 12.80                        | 22.37                                    |





# Thanks

Link to this code: <a href="https://github.com/ahyaramdha/precip-forecast">https://github.com/ahyaramdha/precip-forecast</a>

