# Weather Forecast

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#### Outline:

- > Project idea
- > Work plan
- > Time Series
- > AutoTS
- > PyCaret
- Best model
- > Issues
- > References

## Project Idea:

Weather forecasting for Dammam city, Saudi Arabia.





## Work Plan:

| Data collection        | Data Cleaning               | <b>EDA</b>         | Data reduction  | Feature and<br>Target<br>selection | Model Building & Deployment |
|------------------------|-----------------------------|--------------------|---|------------------------------------|-----------------------------|
| Collecting the dataset | Handling the missing Value. | Visualize the data | Dimensionality reduction: reduction the number of input variables in a dataset. | feature and split                  | AutoTS PyCaret              |

#### Data Collections:

Dataset : <a href="https://www.visualcrossing.com">https://www.visualcrossing.com</a>

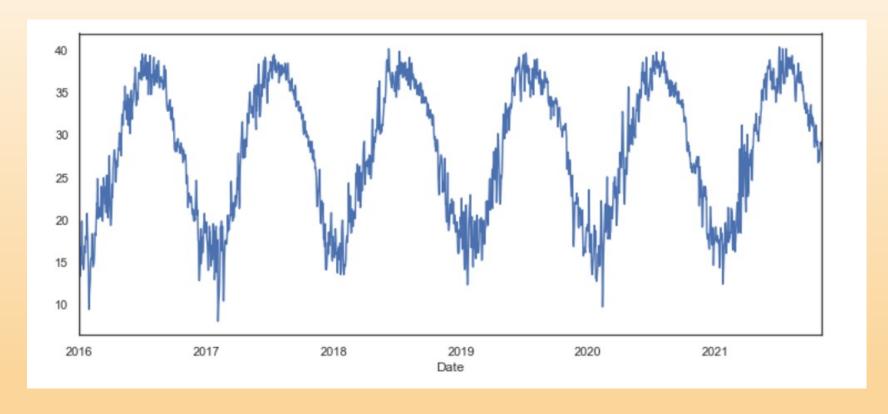
| Precipitation Snow Snow Depth Wind Speed Wind Direction Wind Gust |  |
|---|--|
| Visibility Cloud Cover Relative Humidity Conditions               |  |

# Data Cleaning:

- remove messy data
- manage missing values
- remove duplicates and outliers

## **EDA**

Temperature and Date plot



# Data Explorations:

• Heatmap

| emperature  | 1.0                 | 0.9                 | 1.0         | 0.5        | 0.1        | -0.2           | 0.0        | -0.4        | -0.6              | - 1.0 |
|---|---------------------|---------------------|-------------|------------|------------|----------------|------------|-------------|-------------------|-------|
| Temperature Minimum Temperature Maximum Temperature | 0.9                 | 1.0                 | 1.0         | 0.5        | 02         | -0.2           | -0.1       | -0.2        | -0.5              | - 0.8 |
| Temperature Minimum                                 | 1.0                 | 1.0                 | 1.0         | 0.5        | 0.1        | -0.2           | -0.0       | -0.3        | -0.6              | - 0.6 |
| Heat Index  | 0.5                 | 0.5                 | 0.5         | 1.0        | 0.0        | -0.1           | -0.1       | -0.2        | -0.1              | - 0.4 |
| Wind Speed  | 0.1                 | 02                  | 0.1         | 0.0        | 1.0        | 0.4            | -0.1       | 0.1         | -0.3              | - 0.2 |
| Wind Direction                                      | -0.2                | -0.2                | 0.2         | -0.1       | 0.4        | 1.0            | 0.1        | -0.1        | -0.2              | - 0.0 |
| Visibility  | 0.0                 | -0.1                | -0.0        | -0.1       | -0.1       | 0.1            | 1.0        | 0.2         | -0.3              | 0.2   |
| Cloud Cover   | -0.4                | -0.2                | 0.3         | -0.2       | 0.1        | -0.1           | -0.2       | 1.0         | 0.4               | 0.4   |
| Relative Humidity                                   | -0.6                | -0.5                | -0.6        | -0.1       | -0.3       | -0.2           | -0.3       | 0.4         | 1.0               | 0.6   |
| Reit  | Maximum Temperature | Minimum Temperature | Temperature | Heat Index | Wind Speed | Wind Direction | Visibility | Cloud Cover | Relative Humidity |       |

#### Time Series:

> Why time Series?

#### > What time Series?

- Time series is a sequence of observations recorded at regular time intervals.
- Time Series Analysis comprised methods for analyzing time series data in order to extract meaningful statistics and other characteristics of the data.
- Time Series forecasting which is the use of a model to predict future values based on previously observed values.

## > Importance:

Understanding past behavior and planning for the future

### > Components of a Time Series

- Trend
- Seasonality
- Cycles
- Irregular fluctuation

## > When not to use a time series analysis

- When the values are constant.
- Values in the form of a function

## **AutoTS Library**

- ➤ Why AutoTs?
- Why What AutoTS?

It is an open-source python library basically used to automate Time Series Forecasting.

- > Features of Auto-TS library:
- it finds the optimal time series forecasting model using genetic programming optimization.
- It trains different models such as ARIMA, SARIMAX, ensemble machine learning models and FB prophet-based models, with all possible hyperparameter configurations, and cross-validation.
- Save time.

```
Leaderboard with best model on top of list:

name rmse

ML 0.581661

Prophet 1.981375

auto_SARIMAX inf
```

#### Test

|            | yhat      |
|------------|-----------|
| Date       |           |
| 2021-06-01 | 37.459557 |
| 2021-06-02 | 37.868336 |
| 2021-06-03 | 37.782017 |
| 2021-06-04 | 38.101112 |
| 2021-06-05 | 38.014118 |
| • • •      | • • •     |
| 2021-10-27 | 27.396517 |
| 2021-10-28 | 27.971649 |
| 2021-10-29 | 27.316912 |
| 2021-10-30 | 28.535494 |
| 2021-10-31 | 28.644131 |

#### Forecast

|            | Temperature |
|------------|-------------|
| 2021-11-01 | 27.059781   |
| 2021-11-02 | 26.789482   |
| 2021-11-03 | 26.685752   |
| 2021-11-04 | 26.546830   |
| 2021-11-05 | 26.393519   |
|            |             |
| 2022-02-24 | 19.491261   |
| 2022-02-25 | 19.661698   |
| 2022-02-26 | 19.766187   |
| 2022-02-27 | 19.717202   |
| 2022-02-28 | 19.761517   |
|            |             |

### PyCaret:

- ➤ Why PyCaret?
- > What PyCaret

**PyCaret** is an open-source, low-code machine learning library in Python that automates machine learning workflows.

- > Features of PyCaret:
- Low-code library.
- Fast and efficient
- Wrap several machine learning libraries and frameworks such as: scikit-learn, XGBoost, LightGBM, CatBoost, Ray, and a few more.

**compare\_models** function trains and evaluates 30+ algorithms from ARIMA to XGboost and more.

# PyCaret:

 After comparing the model. it will use the best model to make the test and the predictions.

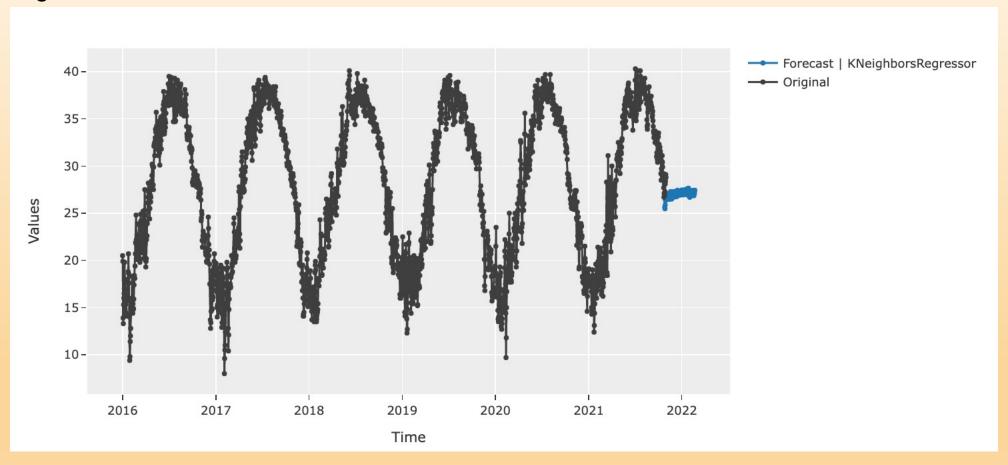
#### Forecast

| _          |         |  |
|------------|---------|--|
| 2021-11-01 | 28.8047 |  |
| 2021-11-02 | 28.2070 |  |
| 2021-11-03 | 28.4894 |  |
| 2021-11-04 | 28.6512 |  |
| 2021-11-05 | 28.4924 |  |
|            |         |  |
| 2022-02-24 | 28.1176 |  |
| 2022-02-25 | 28.3764 |  |
| 2022-02-26 | 28.9585 |  |
| 2022-02-27 | 28,1221 |  |
|            |         |  |
| 2022-02-28 | 28.3809 |  |

|                 | Model   | MAE    | RMSE   | MAPE   | SMAPE  | R2       | TT (Sec) |
|-----------------|---|--------|--------|--------|--------|----------|----------|
| knn_cds_dt      | K Neighbors w/ Cond. Deseasonalize & Detrending                   | 0.8290 | 1.0016 | 0.0286 | 0.0280 | -0.7753  | 0.0533   |
| xgboost_cds_dt  | Extreme Gradient Boosting w/ Cond. Deseasonalize & Detrending     | 1.2196 | 1.4305 | 0.0422 | 0.0410 | -2.3161  | 0.5767   |
| lightgbm_cds_dt | Light Gradient Boosting w/ Cond. Deseasonalize & Detrending       | 1.2228 | 1.4194 | 0.0425 | 0.0411 | -2.0787  | 0.5733   |
| et_cds_dt       | Extra Trees w/ Cond. Deseasonalize & Detrending                   | 1.2252 | 1.4329 | 0.0421 | 0.0412 | -2.9778  | 0.1967   |
| rf_cds_dt       | Random Forest w/ Cond. Deseasonalize & Detrending                 | 1.2312 | 1.4531 | 0.0425 | 0.0414 | -2.5234  | 0.3167   |
| lasso_cds_dt    | Lasso w/ Cond. Deseasonalize & Detrending                         | 1.2730 | 1.4901 | 0.0443 | 0.0428 | -2.3231  | 0.0300   |
| auto_arima      | Auto ARIMA  | 1.2743 | 1.5249 | 0.0441 | 0.0428 | -2.7926  | 52.1333  |
| omp_cds_dt      | Orthogonal Matching Pursuit w/ Cond. Deseasonalize & Detrending   | 1.3054 | 1.5100 | 0.0455 | 0.0439 | -2.3823  | 0.0267   |
| theta           | Theta Forecaster  | 1.3142 | 1.5289 | 0.0458 | 0.0442 | -2.4600  | 0.0600   |
| en_cds_dt       | Elastic Net w/ Cond. Deseasonalize & Detrending                   | 1.3200 | 1.5204 | 0.0459 | 0.0443 | -2.4753  | 0.0300   |
| naive           | Naive Forecaster  | 1.3238 | 1.5191 | 0.0462 | 0.0445 | -2.3995  | 0.0133   |
| lr_cds_dt       | Linear w/ Cond. Deseasonalize & Detrending                        | 1.3255 | 1.5824 | 0.0460 | 0.0445 | -3.0008  | 0.0267   |
| ridge_cds_dt    | Ridge w/ Cond. Deseasonalize & Detrending                         | 1.3255 | 1.5823 | 0.0460 | 0.0445 | -3.0000  | 0.0267   |
| br_cds_dt       | Bayesian Ridge w/ Cond. Deseasonalize & Detrending                | 1.3257 | 1.5816 | 0.0460 | 0.0445 | -2.9914  | 0.0267   |
| lar_cds_dt      | Least Angular Regressor w/ Cond. Deseasonalize & Detrending       | 1.3262 | 1.5850 | 0.0460 | 0.0445 | -3.0224  | 0.0400   |
| exp_smooth      | Exponential Smoothing   | 1.3398 | 1.5459 | 0.0468 | 0.0450 | -2.5090  | 0.2100   |
| ets             | ETS   | 1.3397 | 1.5458 | 0.0468 | 0.0450 | -2.5080  | 0.1833   |
| gbr_cds_dt      | Gradient Boosting w/ Cond. Deseasonalize & Detrending             | 1.3650 | 1.5654 | 0.0475 | 0.0458 | -2.6584  | 0.2433   |
| llar_cds_dt     | Lasso Least Angular Regressor w/ Cond. Deseasonalize & Detrending | 1.3967 | 1.5532 | 0.0468 | 0.0469 | -7.8857  | 0.0267   |
| polytrend       | Polynomial Trend Forecaster                                       | 1.4032 | 1.5582 | 0.0470 | 0.0471 | -8.1916  | 0.0100   |
| huber_cds_dt    | Huber w/ Cond. Deseasonalize & Detrending                         | 1.4347 | 1.6695 | 0.0497 | 0.0480 | -3.3516  | 0.0433   |
| ada_cds_dt      | AdaBoost w/ Cond. Deseasonalize & Detrending                      | 1.4633 | 1.6655 | 0.0508 | 0.0490 | -3.3477  | 0.1233   |
| snaive          | Seasonal Naive Forecaster   | 1.5571 | 1.9411 | 0.0537 | 0.0516 | -5.7614  | 0.0167   |
| arima           | ARIMA   | 1.6360 | 1.9891 | 0.0561 | 0.0543 | -6.7084  | 0.1133   |
| grand_means     | Grand Means Forecaster  | 2.1250 | 2.2323 | 0.0692 | 0.0723 | -25.7169 | 0.0133   |
| dt_cds_dt       | Decision Tree w/ Cond. Deseasonalize & Detrending                 | 2.1936 | 2.4497 | 0.0738 | 0.0741 | -22.8405 | 0.0333   |
| par_cds_dt      | Passive Aggressive w/ Cond. Deseasonalize & Detrending            | 2.4624 | 2.7991 | 0.0825 | 0.0836 | -24.3785 | 0.0267   |

# Best model:

According to MAE is KNN



### Issues:

- Dataset
- > Model



#### References:

• <a href="https://www.visualcrossing.com/weather/weather-data-services#/editDataDefinition">https://www.visualcrossing.com/weather/weather-data-services#/editDataDefinition</a>

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- https://www.analyticsvidhya.com/blog/2021/04/aut omate-time-series-forecasting-using-auto-ts/
- https://towardsdatascience.com/announcingpycarets-new-time-series-module-b6e724d4636c