Machine Learning

Linear Regression



Mounting Google Drive

from google.colab import drive
drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

Importing some important libraries

```
import pandas as pd
import numpy as np
import warnings
warnings.filterwarnings('ignore')
```

Loading and Reading Dataset

df = pd.read_csv('/content/drive/MyDrive/Weather.csv')
df.head()

| | STA | Date | Precip | WindGustSpd | MaxTemp | MinTemp | MeanTemp | Snowfall | Р |
|---------------------|-------|----------|--------|-------------|-----------|-----------|-----------|----------|---|
| 0 | 10001 | 7/1/1942 | 1.016 | NaN | 25.55556 | 22.22222 | 23.888889 | 0.0 | |
| 1 | 10001 | 7/2/1942 | 0 | NaN | 28.888889 | 21.666667 | 25.55556 | 0.0 | |
| 2 | 10001 | 7/3/1942 | 2.54 | NaN | 26.111111 | 22.22222 | 24.44444 | 0.0 | |
| 3 | 10001 | 7/4/1942 | 2.54 | NaN | 26.666667 | 22.22222 | 24.44444 | 0.0 | |
| 4 | 10001 | 7/5/1942 | 0 | NaN | 26.666667 | 21.666667 | 24.44444 | 0.0 | |
| 5 rows x 30 columns | | | | | | | | | |
| 4 | | | | | | | | | • |

df.tail()

| | STA | Date | Precip | WindGustSpd | MaxTemp | MinTemp | MeanTemp | Snow |
|----------|--------|------------|--------|-------------|-----------|-----------|-----------|------|
| 119035 | 82506 | 12/27/1945 | 0 | NaN | 28.333333 | 18.333333 | 23.333333 | |
| 119036 | 82506 | 12/28/1945 | 9.906 | NaN | 29.44444 | 18.333333 | 23.888889 | |
| 119037 | 82506 | 12/29/1945 | 0 | NaN | 28.333333 | 18.333333 | 23.333333 | |
| 119038 | 82506 | 12/30/1945 | 0 | NaN | 28.333333 | 18.333333 | 23.333333 | |
| 119039 | 82506 | 12/31/1945 | 0 | NaN | 29.44444 | 17.222222 | 23.333333 | |
| 5 rows x | 30 col | limns | | | | | | • |
| 1 | | | | | | | | |

df.shape

(119040, 30)

df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 119040 entries, 0 to 119039 Data columns (total 30 columns):

| Jaca | COTAINIUS (COT | at so corumns): | |
|------|-----------------|-----------------|---------|
| # | Column | Non-Null Count | Dtype |
| | | | |
| 0 | STA | 119040 non-null | int64 |
| 1 | Date | 119040 non-null | object |
| 2 | Precip | 119040 non-null | object |
| 3 | WindGustSpd | 532 non-null | float64 |
| 4 | MaxTemp | 119040 non-null | float64 |
| 5 | MinTemp | 119040 non-null | float64 |
| 6 | MeanTemp | 119040 non-null | float64 |
| 7 | Snowfall | 117877 non-null | obiect |

```
PoorWeather 34237 non-null
                  119040 non-null int64
       10 MO
                       119040 non-null int64
                      119040 non-null int64
       11 DA
       12 PRCP
                      117108 non-null object
                      533 non-null
                                      float64
       13 DR
                      532 non-null float64
118566 non-null float64
       14 SPD
       15 MAX
                      118572 non-null float64
118542 non-null float64
       16 MIN
       17 MEA
       18 SNF
                      117877 non-null object
       19 SND
                       5563 non-null
       20 FT
                      0 non-null
       21 FB
                       0 non-null
                                       float64
                       0 non-null
                                       float64
       22 FTI
                       525 non-null
       23 PGT
                                       float64
       24 TSHDSBRSGF 34237 non-null object
                       0 non-null
                                       float64
       25 SD3
       26 RHX
                       0 non-null
                                       float64
       27 RHN
                       0 non-null
                                       float64
       28 RVG
                       0 non-null
                                       float64
       29 WTE
                       0 non-null
                                      float64
      dtypes: float64(19), int64(4), object(7)
      memory usage: 27.2+ MB
Based on Minimum Temperature we have to predict Maximum Temperature.
Define independent variable :
  # Define independent variable
  x = df['MinTemp'].values.reshape(-1,1)
      array([[22.2222222],
             Γ21.666666677.
             [22.2222222],
             [18.33333333],
             Γ18.33333333
             [17.2222222]])
Define dependent variable :
```

Define dependent variable

Train_Test_Split :

Splitting

▼ To check →

To check
x_train.shape

(83328, 1)

y = df['MaxTemp'].values.reshape(-1,1)

Used to split the dataset to train and test

train_size, random_state can be exclude

Splitting at 70:30 ratio

Train_Test_Split : Used to split the dataset to train and test

sklearn => Scikit Learn

x_train, x_test, y_train, y_test = train_test_split(x,y,test_size = 0.3, train_size = 0.7, random_state = 1)

from sklearn.model_selection import train_test_split

0.7 * 119040

```
# 0.7 * 119040
  y_train.shape
       (83328, 1)
  x_test.shape
                    # 0.3 * 119040
       (35712, 1)
                    # 0.3 * 119040
  y_test.shape
       (35712, 1)
  df.shape[0]
       119040

    Importing our first Machine Learning Model

  # Importing our first Machine Learning Model
  from sklearn.linear_model import LinearRegression

    Create an object of the Linear Regression Model

  # Create an object of the Linear Regression Model
  model = LinearRegression()
▼ Train the model → .fit()
  # Train the model : .fit()
  model.fit(x_train, y_train)
       LinearRegression()
Predict the model :
  # Predict the model
  y_pred = model.predict(x_test)
  y_pred
       array([[20.88972865],
              [28.56451499],
              [31.63442954],
              [31.12277711],
              [28.56451499],
              [31.12277711]])
▼ R2_Score (R Square) →
  # R2_Score (R Square) : model.score(x_test, y_test)
  r2 = model.score(x_test, y_test)
  model.score(x_train, y_train) # It's not R2_Score
       0.773970250594516

→ Accuracy →

  # Accuracy
  r2 * 100
       76.5801933965467
```

Another method to calculate R2_Score :

```
# Another method to calculate R2_Score
from sklearn.metrics import r2_score
r2_score(y_test, y_pred)
0.765801933965467
```

Adjusted R2_Score = 1 - [(1 - R²) * (N - 1) / (N - p - 1)]

```
where,
R² = R2_Score or R Square
N = Total number of samples or Total number of rows
p = Total number of training or test size of independet features(variable)
# Adjusted R2_Score
adj_r2 = 1 - ((1-r2)*(len(df)-1)) / (len(df)-len(x_test)-1)
adj_r2
```

$ilde{\ }$ Save as an Excel File o

0.6654301296976397

```
# Save as an excel file
y_pred = pd.DataFrame(y_pred)  # First change y_pred to a DataFrame
max_temp_pred = y_pred.to_csv('MaxTemp Prediction.csv')
```