## Wendy Dharmawan - 2206059591

Proyek berikut merupakan proyek yang dibuat untuk course Analisis Data dengan Python, yang membahas data wrangling (mengambil dan membersihkan data), EDA (exploratory data analysis), dan data visualization menggunakan library numpy, pandas, dan matplotlib pada python.

## Menentukan Pertanyaan Bisnis

- Seberapa besar persentase korelasi kadar O3 terhadap suhu kota Guanyuan pada tahun 2013-2017?
- Bagaimana perbandingan kadar SO2, NO2, dan CO antara kota Guanyuan dan Gucheng pada tahun 2013-2017?

## Import Semua Packages/Library yang Digunakan

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

## **Data Wrangling**

## Gathering Data

## Memuat tabel kota Guanyuan

```
quanyuan df =
pd.read csv("https://raw.githubusercontent.com/marceloreis/HTI/master/
PRSA Data 20130301-20170228/PRSA Data Guanyuan 20130301-20170228.csv")
guanyuan df.head()
{"summary":"{\n \"name\": \"guanyuan_df\",\n \"rows\": 35064,\n
                  \"fields\": [\n
                                                \"properties\":
          \"dtype\": \"number\",\n
                                        \"std\": 10122,\n
\"min\": 1,\n
                   \"max\": 35064,\n
                                           \"num_unique_values\":
               \"samples\": [\n
35064,\n
                                                        32601,\n
                                       6071,\n
8847\n
             ],\n
                        \"semantic_type\": \"\",\n
\"description\": \"\"\n
                                  },\n
                                         {\n
                                                  \"column\":
                           }\n
\"descrip
\"year\",\n \p
                                         \"dtype\": \"number\",\n
               \"properties\": {\n
                   \"min\": 2013,\n
                                         \"max\": 2017,\n
\"num unique values\": 5,\n
                                 \"samples\": [\n
                                                         2014,\n
                                        \"semantic_type\": \"\",\n
2017,\n
                2015\n
\"description\": \"\"\n
                           }\n
                                  },\n
                                                  \"column\":
\"month\",\n \"properties\": {\n
\"std\": 3,\n \"min\": 1,\n
                                          \"dtype\": \"number\",\n
                                       \"max\": 12,\n
\"samples\": [\n
                                  \"semantic_type\": \"\",\n
\"description\": \"\"\n
                                                  \"column\":
                           }\n
                                  },\n
```

```
\"day\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 8,\n \"min\": 1,\n \"max\": 31,\n
\"hour\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 6,\n \"min\": 0,\n \"max\": 23,\n
\"num_unique_values\": 24,\n \"samples\": [\n 8,\n 16,\n 0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n {\n \"column\": \"PM2.5\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 80.93349731219654,\n \"min\": 2.0,\n \"max\": 680.0 \n \n \\"num unique values\": 557 \n \\"comples\": \"\"
680.0,\n \"num_unique_values\": 557,\n \"samples\": [\n 140.0,\n 14.3,\n 129.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"PM10\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 91.57370871803138,\n \"mip\": 2.0,\n \"max\": 000.0\"
 \"min\": 2.0,\n \"max\": 999.0,\n \"num_unique_values\":
630,\n \"samples\": [\n 305.4,\n 45.0,\n 605.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n },\n {\n \"column\": \"S02\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 23.600366633044654,\n \"min\": 1.0,\n \"max\":
293.0,\n \"num_unique_values\": 280,\n \"samples\": [\n 75.0,\n 33.0,\n 142.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n \\"dtype\": \"number\",\n \"std\": 35.15085745143049,\n \"min\": 2.0 \n \"max\": 270.0 \
\"max\": 10000.0,\n \"num_unique_values\": 117,\n \"samples\": [\n 4599.0,\n 700.0,\n 200.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n },\n {\n \"column\": \"03\",\n \"properties\": {\n \"03\",\n \"column\": \"03\",\n \"properties\": {\n \"03\",\n \"03\",\n \"properties\": \"\"
\"dtype\": \"number\",\n \"std\": 57.43698266437783,\n
\"min\": 0.2142,\n \"max\": 415.0,\n
\"num_unique_values\": 767,\n \"samples\": [\n 20.7774,\n 38.9844,\n 56.5488\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n },\n {\n \"column\": \"TEMP\",\n \"properties\": {\n \"}
\"dtype\": \"number\",\n \"std\": 11.399096949947065,\n \"min\": -16.8,\n \"max\": 40.5,\n
\"num_unique_values\": 967,\n \"samples\": [\n 18.8,\n -10.3,\n -1.575\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n {\n
```

```
\"column\": \"PRES\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 10.404046577854823,\n \"min\":
985.9,\n \"max\": 1042.0,\n \"num unique values\": 600,\
n \"samples\": [\n 998.6,\n 1029.9,\n 1024.25\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n }\n {\n \"column\":
\"DEWP\",\n\\"properties\": {\n\\"dtype\": \\"std\": 13.688896009288912,\n\\"min\": -35.3,\n
                                              \"dtype\": \"number\",\n
\"max\": 28.5,\n \"num_unique_values\": 604,\n \"samples\": [\n -2.1,\n 27.2,\n -30.2\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
       },\n {\n \"column\": \"RAIN\",\n \"properties\":
}\n
{\n \"dtype\": \"number\",\n \"std\":
0.9100559139821632,\n\"min\": 0.0,\n\\"max\": 72.5,\n
\mbox{"num\_unique\_values}": 127,\n \ \"samples\": [\n 2.1,\n]
\"description\": \"\"n }\n },\n {\n
                                                        \"column\":
\"wd\",\n \"properties\": {\n \"dtype\": \"category\",\n
\"num_unique_values\": 16,\n \"samples\": [\n \"NNW\",\n \"N\",\n \"E\"\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
n },\n {\n \"column\": \"WSPM\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 1.2040710727929371,\n
\"min\": 0.0,\n \"max\": 11.2,\n \"num_unique_values\":
91,\n \"samples\": [\n 1.9,\n
                                                         0.7, n
\"samples\":
[\n \"Guanyuan\"\n ],\n \"semantic_type\":
\"\",\n \"description\": \"\"\n }\n ]\
n}","type":"dataframe","variable_name":"guanyuan_df"}
```

## Memuat tabel kota Gucheng

```
gucheng_df =
pd.read_csv("https://raw.githubusercontent.com/marceloreis/HTI/master/
PRSA_Data_20130301-20170228/PRSA_Data_Gucheng_20130301-20170228.csv")
gucheng_df.head()

{"summary":"{\n \"name\": \"gucheng_df\",\n \"rows\": 35064,\n
\"fields\": [\n {\n \"column\": \"No\",\n \"properties\":
{\n \"dtype\": \"number\",\n \"std\": 10122,\n
\"min\": 1,\n \"max\": 35064,\n \"num_unique_values\":
35064,\n \"samples\": [\n 6071,\n 32601,\n
8847\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n {\n \"column\":
\"year\",\n \"properties\": {\n \"dtype\": \"number\",\n
\"std\": 1,\n \"min\": 2013,\n \"max\": 2017,\n
\"num_unique_values\": 5,\n \"samples\": [\n 2014,\n
```

```
\"num_unique_values\": 12,\n \"samples\": [\n 1,\
12,\n 3\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n
                                                                                                                                                },\n {\n \"column\":
\"day\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 8,\n \"min\": 1,\n \"max\": 31,\n
                                                                                                                                                  \"samples\": [\n 28,\n
\"num_unique_values\": 31,\n
                                                                                                                                                 \"semantic_type\": \"\",\n
16,\n 24\n ],\n
\"samples\": [\n
                                                                                                                                                                                                                                                     8,\n
                                                                                                                                               \"semantic_type\": \"\",\n
\"PM2.5\",\n \"properties\": {\n \"dtype\": \"number\",\n
\"std\": 82.79644460750218,\n \"min\": 2.0,\n \"max\":
770.0,\n \"num_unique_values\": 577,\n \"samples\": [\n 250.0,\n 444.0,\n 628.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"PM10\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 96.7426255992567,\n \"min\": 2.0,\n \"mi
\"min\": 2.0,\n \"max\": 994.0,\n \"num_unique_values\":
660,\n \"samples\": [\n 496.0,\n 502.0,\n 293.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n },\n {\n \"column\": \"S02\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 21.204526026319193,\n \"min\": 0.2856,\n
\"max\": 500.0,\n \"num_unique_values\": 243,\n \"samples\": [\n 62.0,\n 12.0,\n 10.8528\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n },\n {\n \"column\": \"N02\",\n \"properties\": {\}
n \"dtype\": \"number\",\n \"std\": 36.47386009283308,\n \"min\": 2.0,\n \"max\": 276.0,\n \"num_unique_values\":
367,\n \"samples\": [\n 56.2522,\n 94.0,\n 30.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n }\n {\n \"column\":
\"CO\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 1208.957772258247,\n \"min\": 100.0,\n \"max\": 10000.0,\n \"num_unique_values\": 116,\n \"samples\": [\
n 6700.0,\n 1000.0,\n 6900.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"03\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 57.01958700323256,\n \"min\": 0.2142,\n \"max\": 450.0,\n \""arms | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\" | 1000.0\"
\"num_unique_values\": 804,\n \"samples\": [\n 98.1036,\n 20.5632,\n 110.9556\n
                                                                                                                                                                                                                                          ],\n
```

```
\"dtype\": \"number\",\n \"std\": 11.292857008072325,\n \"min\": -15.6,\n \"max\": 41.6,\n
\"num_unique_values\": 995,\n \"samples\": [\n -8.02,\
n -15.5,\n 17.75\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
n },\n {\n \"column\": \"PRES\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 10.103255921223534,\n
\"min\": 984.0,\n \"max\": 1038.1,\n
\"num unique values\": 592,\n \"samples\": [\n
1034.6,\n 1015.3333333333,\n 1013.0\n
\"semantic_type\": \"\",\n \"description\": \"\"\n }\

n },\n {\n \"column\": \"DEWP\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 13.782990710571664,\n \"min\": -34.6,\n \"max\": 27.4,\n
\"num_unique_values\": 594,\n \"samples\": [\n -16.0,\n 15.1,\n -1.2\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"RAIN\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 0.838653860679221,\n
\"min\": 0.0,\n \"max\": 41.9,\n \"num unique values\":
125,\n \"samples\": [\n 9.3,\n 3.4,\n 12.8\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n {\n \"column\":
\"wd\",\n \"properties\": {\n \"dtype\": \"category\",\n
[\n \"Gucheng\"\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n ]\
n}","type":"dataframe","variable_name":"gucheng_df"}
```

## Assessing Data

## Menilai data Guanyuan

```
Menghitung jumlah data yang tidak lengkap
```

```
guanyuan_df.info()
guanyuan_df.isna().sum()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 35064 entries, 0 to 35063
Data columns (total 18 columns):
              Non-Null Count Dtype
     Column
0
              35064 non-null int64
     No
 1
              35064 non-null int64
     year
 2
              35064 non-null int64
     month
 3
     day
              35064 non-null int64
4
     hour
              35064 non-null int64
5
     PM2.5
              34448 non-null float64
 6
     PM10
              34635 non-null float64
 7
              34590 non-null float64
     S02
 8
     N02
              34405 non-null float64
 9
     C0
              33311 non-null float64
 10
    03
              33891 non-null float64
 11
    TEMP
              35044 non-null float64
              35044 non-null float64
 12
    PRES
 13
    DEWP
              35044 non-null float64
14
    RAIN
              35044 non-null float64
15
    wd
              34983 non-null
                              object
16
    WSPM
              35050 non-null float64
     station 35064 non-null
17
                              object
dtypes: float64(11), int64(5), object(2)
memory usage: 4.8+ MB
              0
No
              0
year
month
              0
              0
day
hour
              0
PM2.5
            616
PM10
            429
S02
            474
N02
            659
C0
           1753
03
           1173
TEMP
             20
             20
PRES
             20
DEWP
RAIN
             20
             81
wd
WSPM
             14
station
              0
dtype: int64
```

Terlihat bahwa terdapat beberapa data yang kosong pada column PM2.5, PM10, SO22, NO2, CO, O3, TEMP, PRES, DEWP, RAIN, wd dan WSPM

#### Menampilkan parameter statistik tabel Guanyuan

```
guanyuan df.describe()
{"summary":"{\n \"name\": \"guanyuan_df\",\n \"rows\": 8,\n
\ '' fields \ '': [\n\\"column\\":\"No\\",\n\\"properties\\":
{\n \"dtype\": \"number\",\n \"std\":
12636.930448670224,\n\\"min\": 1.0,\n\\"max\": 35064.0,\
n \"num_unique_values\": 6,\n \"samples\": [\n 35064.0,\n 17532.5,\n 26298.25\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"year\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 11807.362916902399,\n \"min\": 1.1772134318198135,\n \"max\": 35064.0,\n \"num_unique_values\": 8,\n \"samples\": [\n
                                     2015.0,\n 35064.0\n
2014.662559890486,\n
\"semantic type\": \"\",\n \"description\": \"\"\n
n },\n {\n \"column\": \"month\",\n \"properties\": {\
           \"dtype\": \"number\",\n \"std\": 12394.775690880602,\
n
n \"min\": 1.0,\n \"max\": 35064.0,\n \"num_unique_values\": 8,\n \"samples\": [\n 6.522929500342231,\n 7.0,\n 35064.0\n
                                                                             ],\n
\"semantic type\": \"\",\n \"description\": \"\"\n
                                                                               }\
n },\n {\n \"column\": \"day\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 12391.770525834587,\n
\"min\": 1.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 15.729637234770705,\n 16.0,\n 35064.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n }\n \\"n \\"column\": \"hour\",\n \"properties\": \\"dtype\": \"number\",\n \"std\": 12393.163442376766,\n
\"min\": 0.0,\n \"max\": 35064.0,\n \"num_unique_values\": 7,\n \"samples\": [\n 35064.0,\n 11.5,\n 17.25\n ],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"PM2.5\",\n \"properties\": {\
           \"dtype\": \"number\",\n \"std\": 12128.523448483706,\
n
n
n \"min\": 2.0,\n \"max\": 34448.0,\n \"num_unique_values\": 8,\n \"samples\": [\n 82.93337203901532,\n 59.0,\n 34448.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                              ],\n
                                                                               }\
n },\n {\n \"column\": \"PM10\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12174.91689123717,\n
\"min\": 2.0,\n \"max\": 34635.0,\n
\"dtype\": \"number\",\n \"std\": 12211.212534718983,\n \"min\": 1.0,\n \"max\": 34590.0,\n
```

```
\"num_unique_values\": 8,\n \"samples\": [\n 17.59094149754264,\n 8.0,\n 34590.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
             },\n {\n \"column\": \"NO2\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 12137.76628206752,\n
\"min\": 2.0,\n \"max\": 34405.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 57.901642517076006,\n 51.0,\n 34405.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                                                                                                                ],\n
n },\n {\n \"column\": \"CO\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 11451.576920854763,\n \"min\": 100.0,\n \"max\": 33311.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
1271.294377232746,\n 900.0,\n 33311.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n }\n {\n \"column\": \"03\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 11949.820184258542,\n \"min\": 0.2142,\n \"max\": 33891.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
n },\n {\n \"column\": \"TEMP\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12385.411047634085,\n
\"min\": -16.8,\n \"max\": 35044.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 13.584607417568478,\n 14.5,\n 35044.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n \,\n \"column\": \"PRES\",\n \"properties\": {\n \}\n \\"column\": \"\"\n \\"\"\n \\"\n \\"\"\n \\"\"\n \\"\"\n \\"\"\n \\"\"\n \\"\n \\"\n \\"\"\n \\"\"\n \\"\n \\"\"\n \\"\n \\"\n \\"\n \\"\n \\"\"\n \\"\n \\"\n
\"dtype\": \"number\",\n \"std\": 12087.68535182901,\n \"min\": 10.404046577854823,\n \"max\": 35044.0,\n \"num_unique_values\": 8,\n \"samples\": [\n 1011.8469198246108,\n 1011.4,\n 35044.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                                                                                 35044.0\n ],\
12388.863098113037,\n \"min\": -35.3,\n \"max\": 35044.0,\n \"num_unique_values\": 8,\n \"samples\": [\n
\"num_unique_values\": 5,\n \"samples\": [\n 0.06742095651181372,\n 72.5,\n 0.9100559139821632\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
12391.106770540331,\n\"min\": 0.0,\n\\"max\": 35050.0,\
                   \"num_unique_values\": 8,\n \"samples\": [\n
```

Dapat dilihat terdapat banyak data kosong pada kolom rain yang ditandai dengan median dan quartil bawah / atasnya bernilai 0

#### Menampilkan jumlah data duplikat pada tabel Guanyuan

```
print("Jumlah duplikasi: ", guanyuan_df.duplicated().sum())
Jumlah duplikasi: 0
```

## Menilai data Gucheng

Menghitung jumlah data yang tidak lengkap

```
gucheng df.info()
gucheng df.isna().sum()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 35064 entries, 0 to 35063
Data columns (total 18 columns):
#
     Column
              Non-Null Count Dtype
- - -
 0
              35064 non-null
     No
                              int64
 1
     vear
              35064 non-null int64
 2
              35064 non-null int64
     month
 3
              35064 non-null int64
     day
 4
              35064 non-null int64
     hour
 5
              34418 non-null float64
     PM2.5
 6
     PM10
              34683 non-null float64
 7
     S02
              34557 non-null float64
 8
     N02
              34396 non-null float64
 9
     C0
              33663 non-null float64
 10
    03
              34335 non-null float64
 11
    TEMP
              35013 non-null float64
              35014 non-null float64
 12
    PRES
 13
    DEWP
              35013 non-null float64
 14
              35021 non-null float64
    RAIN
 15
    wd
              34905 non-null object
 16
    WSPM
              35022 non-null float64
     station 35064 non-null
 17
                              obiect
dtypes: float64(11), int64(5), object(2)
memory usage: 4.8+ MB
              0
No
year
              0
              0
month
              0
day
```

```
0
hour
PM2.5
             646
PM10
             381
S02
             507
N02
             668
C0
            1401
             729
03
              51
TEMP
              50
PRES
DEWP
              51
RAIN
              43
wd
             159
              42
WSPM
station
               0
dtype: int64
```

Terlihat bahwa terdapat beberapa data yang kosong pada column PM2.5, PM10, SO22, NO2, CO, O3, TEMP, PRES, DEWP, RAIN, wd dan WSPM

#### Menampilkan parameter statistik tabel Gucheng

```
qucheng df.describe()
{"summary":"{\n \"name\": \"gucheng df\",\n \"rows\": 8,\n
\fields: [\n \"column\\": \"No\",\n
                                                \"properties\":
          \"dtype\": \"number\",\n \"std\":
12636.930448670224,\n \"min\": 1.0,\n
                                               \"max\": 35064.0.\
n \"num unique values\": 6,\n
                                        \"samples\": [\n
                 17532.5,\n
                                     26298.25\n ],\n
35064.0,\n
\"semantic_type\": \"\",\n \"description\": \"\"\n
    },\n {\n \"column\": \"year\",\n
                                              \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 11807.362916902399,\n
\"min\": 1.1772134318198135,\n
                               \mbox{"max}": 35064.0,\n
                                \"samples\": [\n
\"num unique values\": 8,\n
2014.662559890486,\n
                           2015.0,\n
                                             35064.0\n
                                                             ],\n
\"semantic type\": \"\",\n
                          \"description\": \"\"\n
    },\n {\n \"column\": \"month\",\n \"properties\": {\
        \"dtype\": \"number\",\n \"std\": 12394.775690880602,\
n
        \"min\": 1.0,\n \"max\": 35064.0,\n que values\": 8,\n \"samples\": [\n
\"num unique values\": 8,\n
                           7.0,\n
6.522929500342231,\n
                                          35064.0\n
                                                          ],\n
\"semantic_type\": \"\",\n
                              \"description\": \"\"\n
                                                          }\
           {\n \"column\": \"day\",\n \"properties\": {\n
    },\n
\"dtype\": \"number\",\n
                            \"std\": 12391.770525834587,\n
\"min\": 1.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n
                                \"samples\": [\n
                            16.0, n
15.729637234770705,\n
                                            35064.0\n
15./29637234770705,\n 16.0,\n 35064.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n
n },\n {\n \"column\": \"hour\",\n \"properties\": \{\n
\"dtype\": \"number\",\n \"std\": 12393.163442376766,\n
```

```
\"min\": 0.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 7,\n \"samples\": [\n 35064.0,\
n 11.5,\n 17.25\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"PM2.5\",\n \"properties\": {\
n \"dtype\": \"number\",\n \"std\": 12113.69298356956,\n
\"min\": 2.0,\n\\"max\": 34418.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 83.85208902318554,\n 60.0,\n 34418.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"PM10\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12189.609191626367,\n \"""
 \"min\": 2.0,\n \"max\": 34683.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 118.86197849090333,\n 99.0,\n 34683.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"S02\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12190.368028002727,\n \"min\": 0.2856,\n \"max\": 34557.0,\n
\"num unique values\": 8,\n \"samples\": [\n
n },\n {\n \"column\": \"CO\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 11558.282691821732,\n
 \"min\": 100.0,\n \"max\": 33663.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 1323.9744229569558,\n 900.0,\n 33663.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"03\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12104.622867372267,\n \"min\": 0.2142,\n \"max\": 34335.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 57.694878835008005,\n 45.0,\n 34335.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"TEMP\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 12374.275383198543,\n
 \"min\": -15.6,\n \"max\": 35013.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 13.864524478028411,\n 14.8,\n 35013.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n \\n \\"column\": \"PRES\",\n \"properties\": {\n \}\n \\\"column\": \"\"\n \\"\"\n \\"\n \\"\"\n \\"\"\n \\"\n \\"\"\n \\"\"\n \\"\"\n \\"\"\n \\"\"\n \\"\"\n \\"\n \\"\"\n \\"\"\n \\"\n \\"\"\n \\"\n \\"\"\n \\"\"\n \\"\n \\"\n \\"\n \\"\"\n \\"\n \\\"\n \\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\\"\n \\"\n \\\"\n \\
\"dtype\": \"number\",\n\\"std\": 12077.960428885428,\n\\"min\": 10.103255921223534,\n\\"max\": 35014.0,\n
```

```
\"num unique values\": 8,\n
                                 \"samples\": [\n
1008.8295917481486,\n
                             1008.5, n
                                               35014.0\n
                                                                ],\
        \"semantic type\": \"\",\n
                                         \"description\": \"\"\n
                      \"column\": \"DEWP\",\n
}\n
              {\n
                                                  \"properties\":
                                         \"std\":
          \"dtype\": \"number\",\n
{\n
12378.039705764257,\n
                           \"min\": -34.6,\n
                                                   \"max\":
                 \"num unique values\": 8,\n
                                                  \"samples\": [\n
35013.0,\n
2.6104418358895276,\n
                             3.0, n
                                                             ],\n
                                             35013.0\n
\"semantic type\": \"\",\n
                                \"description\": \"\"\n
                                                            }\
    },\n {\n
                    \"column\": \"RAIN\",\n
                                                \"properties\": {\n
\"dtype\": \"number\",\n
                              \"std\": 12379.640032154402,\n
                     \"max\": 35021.0,\n
\"min\": 0.0,\n
\"num unique_values\": 5,\n
                                 \"samples\": [\n
0.06445275691727821,\n
                                               0.838653860679221\n
                              41.9,\n
],\n
           \"semantic_type\": \"\",\n
                                           \"description\": \"\"\n
                      \"column\": \"WSPM\",\n
}\n
                                                  \"properties\":
      },\n
              {\n
          \"dtype\": \"number\",\n \"std\":
{\n
12381.243654409724,\n
                            \"min\": 0.0,\n
                                                 \"max\": 35022.0,\
        \"num unique values\": 8,\n
                                          \"samples\": [\n
1.3433099194791847.\n
                             1.0, n
                                             35022.0\n
                                                             ],\n
\"semantic type\": \"\",\n
                               \"description\": \"\"\n
                                                            }\
    }\n ]\n}","type":"dataframe"}
```

## Menampilkan jumlah data duplikat pada tabel Gucheng

```
print("Jumlah duplikasi: ", gucheng_df.duplicated().sum())
Jumlah duplikasi: 0
```

Tidak terdapat data duplikat pada tabel tersebut

Dapat dilihat terdapat banyak data kosong pada kolom rain yang ditandai dengan median dan quartil bawah / atasnya bernilai 0

## Cleaning Data

Karena kedua data berupa time series, maka data yang dianalisa harus utuh sehingga tidak dapat menggunakan drop. Maka untuk mengatasi data yang hilang digunakan metode interpolation

```
guanyuan_df.interpolate(method='linear',limit_direction='forward',inpl
ace = True)
gucheng_df.interpolate(method='linear',limit_direction='forward',inpla
ce = True)
```

Tidak ada data yang duplikat pada kedua tabel. Data yang tidak lengkap hanya terdapat pada kolom rain yang tidak digunakan dalam analisis data.

# Exploratory Data Analysis (EDA)

## Explore ...

Melihat rangkuman parameter statistik menggunakan describe

```
guanyuan df.describe(include="all")
{"summary":"{\n \"name\": \"guanyuan df\",\n \"rows\": 11,\n
{\n \"dtype\": \"number\",\n \"std\":
12636.930448670224,\n\\"min\": 1.0,\n\\"max\": 35064.0,\
n \"num_unique_values\": 6,\n \"samples\": [\n 35064.0,\n 17532.5,\n 26298.25\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
n },\n {\n \"column\": \"year\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 11807.362916902399,\n
2015.0,\n
                                                                                            35064.0\n
2014.662559890486,\n
\"semantic_type\": \"\",\n \"description\": \"\"\n
n },\n {\n \"column\": \"month\",\n \"properties\": {\
                   \"dtype\": \"number\",\n \"std\": 12394.775690880602,\
n \"min\": 1.0,\n \"max\": 35064.0,\n \"num_unique_values\": 8,\n \"samples\": [\n 6.522929500342231,\n 7.0,\n 35064.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                                                                                        ],\n
                                                                                                                                        }\
n },\n {\n \"column\": \"day\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12391.770525834587,\n
\"min\": 1.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 15.729637234770705,\n 16.0,\n 35064.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n },\n {\n \"column\": \"hour\",\n \"properties\": {\n \"},\n \"column\": \"hour\",\n \"properties\": {\n \"},\n \"hour\",\n \"
\"dtype\": \"number\",\n \"std\": 12393.163442376766,\n
\"min\": 0.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 7,\n \"samples\": [\n 35064.0,\
n 11.5,\n 17.25\n ],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n
n },\n {\n \"column\": \"PM2.5\",\n \"properties\": {\
                   \"dtype\": \"number\",\n \"std\": 12346.322972172586,\
n
n \"min\": 2.0,\n \"max\": 35064.0,\n \"num_unique_values\": 8,\n \"samples\": [\n 82.89752167465207,\n 59.0,\n 35064.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                                                                                        ],\n
                                                                                                                                        }\
n },\n {\n \"column\": \"PM10\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 12326.480225532672,\n \"min\": 2.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
```

```
n },\n {\n \"column\": \"S02\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12378.788944684795,\n
\"min\": 1.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
\"dtype\": \"number\",\n \"std\": 12370.723959287741,\n \"min\": 2.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 58.139297945185945,\n 51.0,\n 35064.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"CO\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12054.999033976908,\n \"min\": 100.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 1258.3270448323067,\n 900.0,\n 35064.0\n ],\\"semantic_type\": \"\",\n \"description\": \"\"\n }\\n \\"n \\"n \\"column\": \"03\",\n \"properties\": \\"dtype\": \"number\",\n \"std\": 12364.728431222053,\n \"min\": 0.2142,\n \"max\": 35064.0,\n \\"number\",\n \"max\": 35064.0,\n
                                                                         ],\n
\n },\n {\n \"column\": \"DEWP\",\n \"properties\":
{\n \"dtype\": \"number\",\n \"std\":
12395.934225953106,\n \"min\": -35.3,\n \"max\": 35064.0,\n \"num_unique_values\": 8,\n \"samples\": [\n
```

```
\"std\": 12396.0565081987,\n \"min\": 0.0,\n \"max\":
35064.0,\n \"num_unique_values\": 8,\n \"samples\": [\n
gucheng df.describe(include="all")
 {"summary":"{\n \"name\": \gucheng_df\",\n \"rows\": 11,\n}
{\n \"dtype\": \"number\",\n \"std\":
 12636.930448670224,\n\\"min\": 1.0,\n\\"max\": 35064.0,\
n \"num_unique_values\": 6,\n \"samples\": [\n 35064.0,\n 17532.5,\n 26298.25\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"year\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 11807.362916902399,\n
\"min\": 1.1772134318198135,\n \"max\": 35064.0,\n \"num_unique_values\": 8,\n \"samples\": [\n 2014.662559890486,\n 2015.0,\n 35064.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n }
n },\n {\n \"column\": \"month\",\n \"properties\": {\
n \"dtype\": \"number\",\n \"std\": 12394.77569088066

n \"min\": 1.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
6.522929500342231,\n 7.0,\n 35064.0\n ],\r
\"semantic_type\": \"\",\n \"description\": \"\"\n \\"nroperties\": 4
                   \"dtype\": \"number\",\n \"std\": 12394.775690880602,\
n },\n {\n \"column\": \"day\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 12391.770525834587,\n
\"min\": 1.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
15.729637234770705,\n 16.0,\n 35064.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n }\n {\n \"column\": \"hour\",\n \"properties\": {\n \"},\n \"column\": \"hour\",\n \"properties\": {\n \"},\n \"hour\",\n 
\"dtype\": \"number\",\n \"std\": 12393.163442376766,\n
\"min\": 0.0,\n \"max\": 35064.0,\n \"num_unique_values\": 7,\n \"samples\": [\n 35064.0,\n \ 11.5,\n 17.25\n ],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                                                                                                   }\
```

```
n },\n {\n \"column\": \"PM2.5\",\n \"properties\": {\
           \"dtype\": \"number\",\n \"std\": 12342.006141102263,\
n \"min\": 2.0,\n \"max\": 35064.0,\n \"num_unique_values\": 8,\n \"samples\": [\n
84.07480179101073,\n 60.0,\n 35064.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n }\n \"column\": \"PM10\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 12324.153392354208,\n \"min\": 2.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 119.26156741957564,\n 100.0,\n 35064.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                                    ],\n
n },\n {\n \"column\": \"S02\",\n \"properties\": {\n\"dtype\": \"number\",\n \"std\": 12369.478387183306,\n
\"min\": 0.2856,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
15.705820887519966,\n 7.0,\n 35064.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"N02\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12363.682565745099,\n \"""
\"min\": 2.0,\n \"max\": 35044.0,\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
\"dtype\": \"number\",\n \"std\": 12362.27111519826,\n \"min\": 0.2142,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 58.000000866986085,\n 45.0,\n 35064.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"TEMP\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 12392.306924014538,\n
\"min\": -15.6,\n \"max\": 35064.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
n },\n {\n \"column\": \"PRES\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 12095.630322460285,\n
\"min\": 10.104301856861493,\n\\"max\": 35064.0,\n\\"num_unique_values\": 8,\n\\"samples\": [\n\\1008.8356926040858,\n\\1008.5,\n\\35064.0\n\\]
n\\"semantic_type\": \"\",\n\\"description\": \"\"\n\\
        },\n {\n \"column\": \"DEWP\",\n \"properties\":
}\n
```

```
{\n \"dtype\": \"number\",\n \"std\":
12396.07612358495,\n \"min\": -34.6,\n 35064.0,\n \"num_unique_values\": 8,\n
                                            \"max\":
                                            \"samples\": [\n
                                 35064.0\n
],\n
                                                     }\
    \"dtype\": \"number\",\n \"std\": 12394.842847229575,\n
\"min\": 0.0,\n \"max\": 35064.0,\n
\"num_unique_values\": 5,\n
0.06437371663244354,\n 41
                             \"samples\": [\n
         663244354,\n 41.9,\n 0.838142490923598\n \"semantic_type\": \"\",\n \"description\": \"\"\n
                                        0.838142490923598\n
],\n
     \ \,\n \"column\": \"wd\",\n \"properties\": {\n
}\n
\"dtype\": \"category\",\n \"num_unique_values\": 4,\n
                                 _\"3949\",\n
                    16,\n
\"samples\": [\n
                        \"semantic_type\": \"\",\n
\"34905\"\n
                ],\n
\"description\": \"\"\n
                       }\n },\n {\n \"column\":
\"WSPM\",\n \"properties\": {\n \"dtype\": \"num\"std\": 12396.092878501477,\n \"min\": 0.0,\n
                                     \"dtype\": \"number\",\n
                                                   \"max\":
35064.0,\n \"num unique values\": 8,\n
                                            \"samples\": [\n
                    1.3436672940908054,\n
                                       35064.0\n
\"semantic type\": \"\",\n \"description\": \"\"\n
    \"dtype\": \"category\",\n
                                      \"num unique values\":
{\n
        \"samples\": [\n \"35064\",\n \"n \"semantic_type\": \"\",\n
3,\n
                                                  1, n
\"Gucheng\"\n
                           \"semantic type\": \"\",\n
\"description\": \"\"\n
                       }\n }\n ]\n}","type":"dataframe"}
```

## Mengelompokkan kadar rata-rata gas O3 kota Gucheng per tahun

```
gucheng_df.groupby(by="year")["03"].mean()

year
2013    55.781492
2014    60.253139
2015    60.779538
2016    59.522883
2017    28.924788
Name: 03, dtype: float64
```

## Mengelompokkan rata-rata suhu kota Gucheng per tahun

```
gucheng_df.groupby(by="year")["TEMP"].mean()

year
2013    15.932823
2014    14.431661
2015    14.027483
2016    13.547228
2017    0.373600
Name: TEMP, dtype: float64
```

Terlihat bahwa kadar O3 tertinggi ada pada tahun 2015, sedangkan, sedangkan suhu rata-rata tertinggi berada pada tahun 2013. Kadar O3 dan suhu terrendah ada pada tahun 2017

# Mengelompokkan produksi rata-rata Gas SO2, NO2, dan CO Guanyuan berdasarkan tahun

```
guanyuan df.groupby(by="year")["S02"].mean()
year
2013
        21.859392
2014
        23.133562
2015
        14.073579
2016
        11.484517
        21.259181
2017
Name: SO2, dtype: float64
guanyuan_df.groupby(by="year")["N02"].mean()
year
2013
        58.866182
2014
        64.343179
2015
        54.921958
2016
        53.459984
2017
        64.920904
Name: NO2, dtype: float64
guanyuan_df.groupby(by="year")["C0"].mean()
year
2013
        1132,745030
2014
        1362.357306
2015
        1313.704338
        1168.169399
2016
        1482.768362
2017
Name: CO, dtype: float64
```

# Mengelompokkan produksi rata-rata Gas SO2, NO2, dan CO Gucheng berdasarkan tahun

```
gucheng_df.groupby(by="year")["S02"].mean()

year
2013    17.508021
2014    21.290103
2015    13.338265
2016    10.204121
2017    20.587924
Name: S02, dtype: float64

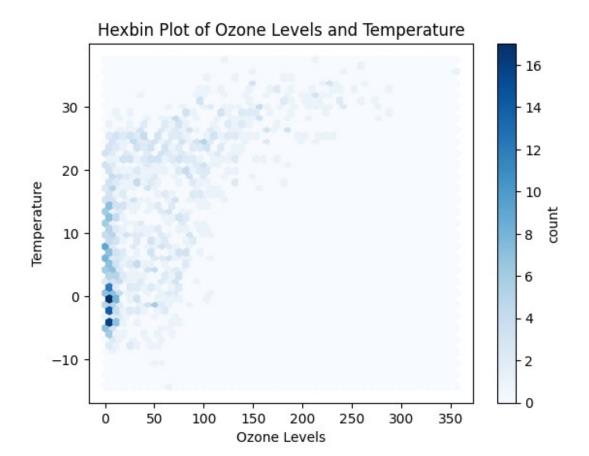
gucheng_df.groupby(by="year")["N02"].mean()
```

```
year
2013
       61.514691
2014
       62.424463
2015
       49.588271
2016
       49.777265
2017
       62.319915
Name: NO2, dtype: float64
gucheng df.groupby(by="year")["C0"].mean()
year
       1341.740264
2013
2014
       1442.163242
2015
       1323.144977
2016
       1170.952869
2017
       1804.837571
Name: CO, dtype: float64
```

# Visualization & Explanatory Analysis

## Pertanyaan 1:

```
sampled_data = gucheng_df.sample(n=1000)
plt.hexbin(sampled_data['03'],sampled_data['TEMP'], gridsize=50,
cmap='Blues')
plt.colorbar(label='count')
plt.title('Hexbin Plot of Ozone Levels and Temperature')
plt.xlabel('Ozone Levels')
plt.ylabel('Temperature')
plt.show()
```



```
temperature = sampled_data['TEMP']
ozone_levels = sampled_data['03']

correlation_coefficient = np.corrcoef( ozone_levels,temperature)[0, 1]

print("Correlation Coefficient:", correlation_coefficient)

Correlation Coefficient: 0.6300191856439685
```

Dari scatterplot dan nilai koefisien, maka bisa ditarik kesimpulan bahwa terdapat moderate positive correlation antara kadar Ozon dan Temperatur di kota tersebut. Hal ini berarti dengan adanya peningkatan kadar ozon, maka juga diikuti dengan peningkatan suhu udara dan sebaliknya.

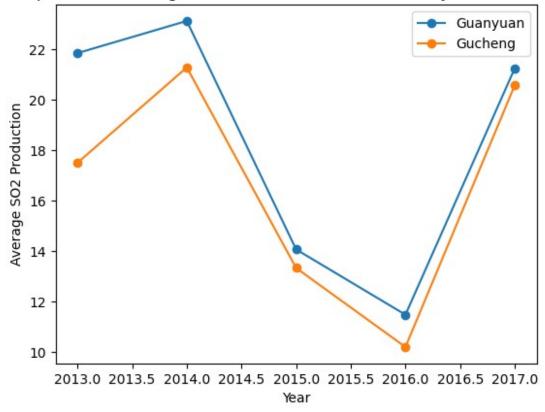
## Pertanyaan 2:

```
guanyuan_so2_mean = guanyuan_df.groupby(by="year")["S02"].mean()
gucheng_so2_mean = gucheng_df.groupby(by="year")["S02"].mean()

plt.plot(guanyuan_so2_mean.index, guanyuan_so2_mean.values,
label='Guanyuan', marker='o')
plt.plot(gucheng_so2_mean.index, gucheng_so2_mean.values,
label='Gucheng', marker='o')
```

```
plt.xlabel('Year')
plt.ylabel('Average SO2 Production')
plt.title('Comparison of Average SO2 Production Between Guanyuan and Gucheng')
plt.legend()
plt.show()
```

## Comparison of Average SO2 Production Between Guanyuan and Gucheng



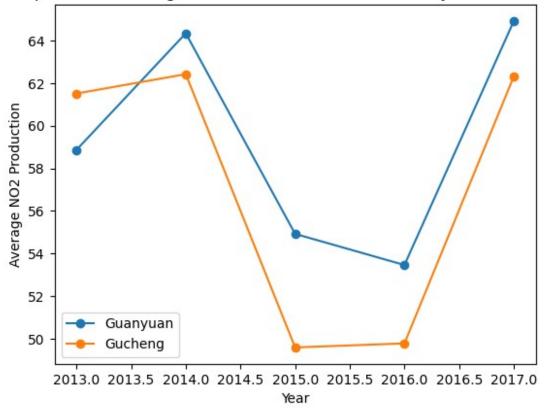
```
guanyuan_no2_mean = guanyuan_df.groupby(by="year")["NO2"].mean()
gucheng_no2_mean = gucheng_df.groupby(by="year")["NO2"].mean()

plt.plot(guanyuan_no2_mean.index, guanyuan_no2_mean.values,
label='Guanyuan', marker='o')
plt.plot(gucheng_no2_mean.index, gucheng_no2_mean.values,
label='Gucheng', marker='o')

plt.xlabel('Year')
plt.ylabel('Average NO2 Production')
plt.title('Comparison of Average NO2 Production Between Guanyuan and Gucheng')
plt.legend()
```

```
plt.show()
```

### Comparison of Average NO2 Production Between Guanyuan and Gucheng



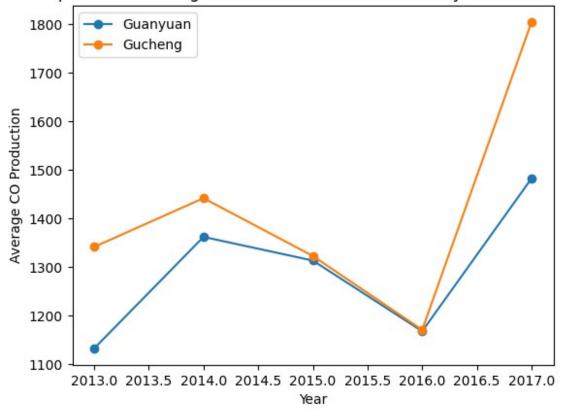
```
guanyuan_co_mean = guanyuan_df.groupby(by="year")["CO"].mean()
gucheng_co_mean = gucheng_df.groupby(by="year")["CO"].mean()

plt.plot(guanyuan_co_mean.index, guanyuan_co_mean.values,
label='Guanyuan', marker='o')
plt.plot(gucheng_co_mean.index, gucheng_co_mean.values,
label='Gucheng', marker='o')

plt.xlabel('Year')
plt.ylabel('Average CO Production')
plt.title('Comparison of Average CO Production Between Guanyuan and Gucheng')
plt.legend()

plt.show()
```

#### Comparison of Average CO Production Between Guanyuan and Gucheng



Dari ketiga grafik, dapat dilihat perbandingan produksi gas polutan rata-rata per tahun dari kedua kota. Kesimpulan yang bisa ditarik dari ketiga grafik adalah sebagai berikut:

- Untuk gas SO2, kota Guanyuan memiliki produksi rata-rata yang lebih banyak dibanding Gucheng.
- Untuk Gas NO2, kota Gucheng memiliki produksi rata-rata yang lebih banyak dibanding Guanyuan pada 2013, namun menurun dan lebih rendah dibanding Guanyuan untuk tahun-tahun selanjutnya
- Untuk Gas CO, kota Gucheng memiliki produksi yang lebih banyak dibanding Guanyuan pada tahun 2013-2015, dan menurun pada tahun 2015-2016 hingga mendekat dengan produksi rata-rata Guanyuan, sebelum menjulang tinggi pada tahun 2016

Perbandingan kualitas udara pada kota tersebut dapat dihubungkan pada kandungan gas polutan yang diukur, sehingga dapat menjadi referensi dalam menjaga kualitas udara kedua kota tersebut.

## Conclusion

- Terdapat korelasi sebesar 63% diantara kadar ozon dan suhu udara kota Guanyuan
- Rata-rata kadar gas SO2 dan CO pada 2016 di Guanyuan lebih besar dibanding Gucheng, sedangkan rata-rata kadar NO2 dan CO pada 2013-2015 lebih banyak dibanding Guanyuan