# [EDA] practice by air polution of seoul data

"데이터 과학의 80%는 데이터 클리닝에 소비되고, 나머지 20%는 데이터 클리닝하는 시간을 불평하는데 쓰인다." - Kaggle 창립자

그만큼 데이터 전처리가 중요하다는 의미죠!!

#### 0. Before EDA

- Learning Pandas library
- knowing DataStructure
- Learning Visualization

is very helpful your studying ^^

## 1. Library & Data Load

import pandas as pd import numpy as np import matplotlib.pyplot as plt %matplotlib inline

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).

df = pd.read\_csv('<u>/content/drive/MyDrive</u>/한국분석/air\_pollution\_in\_seoul/AirPollutionSeoul/Measurement\_summary.csv')

#### 2. Data

#### 2.1 Data 구조

## '[참고] Pandas

- Series & DataFrame
- loc & iloc
- · groupby & sortby
- concat
- drop

# find row, column df.shape

**→** (647511, 11)

# look DataFrame
df.head()

<b>→</b>		Measurement date	Station code	Address	Latitude	Longitude	S02	N02	03	CO	PM10	PM2.5	
	0	2017-01-01 00:00	101	19, Jong-ro 35ga-gil, Jongno-gu, Seoul, Republ	37.572016	127.005008	0.004	0.059	0.002	1.2	73.0	57.0	ш
	1	2017-01-01 01:00	101	19, Jong-ro 35ga-gil, Jongno-gu, Seoul, Republ	37.572016	127.005008	0.004	0.058	0.002	1.2	71.0	59.0	
	2	2017-01-01 02:00	101	19, Jong-ro 35ga-gil, Jongno-gu, Seoul, Renubl	37.572016	127.005008	0.004	0.056	0.002	1.2	70.0	59.0	



		Station code	Latitude	Longitude	\$02	N02	03	CO	PM10	
	count	647511.000000	647511.000000	647511.000000	647511.000000	647511.000000	647511.000000	647511.000000	647511.000000	647511.0
	mean	113.000221	37.553484	126.989340	-0.001795	0.022519	0.017979	0.509197	43.708051	25.4
	std	7.211315	0.053273	0.078790	0.078832	0.115153	0.099308	0.405319	71.137342	43.9
	min	101.000000	37.452357	126.835151	-1.000000	-1.000000	-1.000000	-1.000000	-1.000000	-1.0
	25%	107.000000	37.517528	126.927102	0.003000	0.016000	0.008000	0.300000	22.000000	11.0
	50%	113.000000	37.544962	127.004850	0.004000	0.025000	0.021000	0.500000	35.000000	19.0
	75%	119.000000	37.584848	127.047470	0.005000	0.038000	0.034000	0.600000	53.000000	31.0
	max	125.000000	37.658774	127.136792	3.736000	38.445000	33.600000	71.700000	3586.000000	6256.0
4										▶

# dytype df.info()



<class 'pandas.core.frame.DataFrame'>
RangeIndex: 647511 entries, 0 to 647510 Data columns (total 11 columns):

Dutu	oordiiiio (totar ii	oo raiiiri	5, -	
#	Column	Non-Nu	II Count	Dtype
0	Measurement date	647511	non-null	object
1	Station code	647511	non-null	int64
2	Address	647511	non-null	object
3	Latitude	647511	non-null	float64
4	Longitude	647511	non-null	float64
5	S02	647511	non-null	float64
6	N02	647511	non-null	float64
7	03	647511	non-null	float64
8	CO	647511	non-null	float64
9	PM10	647511	non-null	float64
10	PM2.5	647511	non-null	float64
dtype	es: float64(8), in	t64(1),	object(2)	
	FA OL MO			

memory usage: 54.3+ MB

# find NA, NULL, NAN df.isnull().sum()



	0
Measurement date	0
Station code	0
Address	0
Latitude	0
Longitude	0
S02	0
NO2	0
03	0
CO	0
PM10	0
PM2.5	0
dtvne: int64	

2.2 column

- Latitude & Longitude
- S02(이산화 황)
- N02(이산화 질소)
- 03(오존)
- CO(일산화 탄소)
- PM10(미세먼지)
- PM2.5(초 미세먼지)

```
# 각 column별 value
df.nunique()
```

# station code, Adress, Latitude, Longtitude 가 25개의 다른 지역에서 조사함을 보여줌

```
₹
      Measurement date 25906
        Station code
          Address
                           25
          Latitude
                           25
         Longitude
                           25
            S02
                          186
            N02
                          132
            03
                          253
            СО
                          172
           PM10
                          551
           PM2.5
                          333
    dtvne: int64
```

### 3. Extracting Data

- 3.1 시간별 미세먼지 농도 변화 확인
- 3.2 미세먼지에 영항을 미치는 변인 확인
- 3.3 관측소 위치 Longtitude & Latitude

## ∨ 3.1 시간별 미세먼지 농도 변화 확인

```
from datetime import datetime
df['Measurement date'] = df['Measurement date'].astype('datetime64[ns]')
df['hour'] = df.loc[:, "Measurement date"].dt.hour
df = df.drop('Measurement date', axis=1)
df.hour.head(24)
\overline{2}
     KeyError
                                                Traceback (most recent call last)
     /usr/local/lib/python3.10/dist-packages/pandas/core/indexes/base.py in get_loc(self, key)
        3790
     -> 3791
                         return self._engine.get_loc(casted_key)
        3792
                     except KeyError as err
     index.pyx in pandas._libs.index.IndexEngine.get_loc()
     index.pyx in pandas._libs.index.IndexEngine.get_loc()
     pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
     pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
     KevError: 'Measurement date'
     The above exception was the direct cause of the following exception:
     KeyError
                                                Traceback (most recent call last)
                                            2 frames
     /usr/local/lib/python3.10/dist-packages/pandas/core/indexes/base.py in get_loc(self, key)
        3796
        3797
                             raise InvalidIndexError(kev)
      -> 3798
                         raise KeyError(key) from err
        3799
                     except TypeError:
        3800
                         # If we have a listlike key, _check_indexing_error will raise
     KeyError: 'Measurement date
```

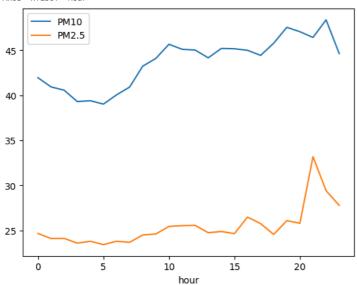
 $\label{data} $$ data = df.groupby('hour', as_index=False).agg({'SO2':'mean', 'NO2':'mean', 'O3':'mean', 'C0':'mean', 'PM10':'mean', 'PM2.5':'mean'}) $$ data = df.groupby('hour', as_index=False).agg({'SO2':'mean', 'NO2':'mean', 'O3':'mean', 'C0':'mean', 'PM10':'mean', 'PM2.5':'mean'}) $$ data = df.groupby('hour', as_index=False).agg({'SO2':'mean', 'NO2':'mean', 'O3':'mean', 'C0':'mean', 'PM10':'mean', 'PM2.5':'mean'}) $$ data = df.groupby('hour', as_index=False).agg({'SO2':'mean', 'NO2':'mean', 'O3':'mean', 'C0':'mean', 'PM10':'mean', 'PM2.5':'mean'}) $$ data = df.groupby('hour', as_index=False).agg({'SO2':'mean', 'NO2':'mean', 'C0':'mean', 'C0':'mean', 'PM10':'mean', 'PM2.5':'mean'}) $$ data = df.groupby('hour', as_index=False).agg({'SO2':'mean', 'NO2':'mean', 'C0':'mean', 'C0':'mean'$ 

<del>_</del> →		hour	S02	N02	03	CO	PM10	PM2.5	
	0	0	-0.001909	0.024584	0.011626	0.529891	41.944368	24.651817	11.
	1	1	-0.002126	0.021533	0.012065	0.524424	40.927181	24.091507	+//
	2	2	-0.002047	0.019609	0.012448	0.516841	40.558962	24.103003	
	3	3	-0.002203	0.018209	0.012198	0.509860	39.303351	23.577686	
	4	4	-0.002267	0.018247	0.011033	0.507101	39.394380	23.785717	
	5	5	-0.002303	0.020649	0.008226	0.515338	39.018339	23.413360	
	6	6	-0.001988	0.025026	0.005753	0.536940	40.044229	23.788876	
	7	7	-0.001966	0.027204	0.005840	0.562405	40.911421	23.670649	
	8	8	-0.001507	0.028721	0.008698	0.576376	43.231577	24.484415	
	9	9	-0.001398	0.027778	0.014076	0.559335	44.089890	24.593978	
	10	10	-0.001183	0.024440	0.019282	0.529488	45.661971	25.433010	
	11	11	-0.001360	0.020498	0.022907	0.500881	45.108788	25.520666	
	12	12	-0.001286	0.018103	0.028270	0.482626	45.022000	25.553704	
	13	13	-0.001212	0.016675	0.031950	0.470215	44.155556	24.737778	
	14	14	-0.001377	0.016059	0.034195	0.459807	45.197106	24.870575	
	15	15	-0.001595	0.016763	0.034715	0.438539	45.164152	24.636552	
	16	16	-0.001363	0.018260	0.032921	0.446770	44.994926	26.470481	
	17	17	-0.001417	0.021212	0.029132	0.455848	44.426963	25.748778	
	18	18	-0.001553	0.024476	0.023835	0.481205	45.778767	24.546031	
	19	19	-0.001558	0.026721	0.019516	0.507660	47.546279	26.075595	
	20	20	-0.001681	0.027166	0.016499	0.522186	47.059484	25.781436	
	21	21	-0.001718	0.027119	0.014502	0.528154	46.420701	33.179113	
	22	22	-0.001829	0.027196	0.012546	0.531633	48.375496	29.401208	
	23	23	-0.004239	0.024149	0.009250	0.526959	44.636799	27.775472	

# 미세먼지 농도변화 Hour data.plot(x='hour', y=['PM10', 'PM2.5'])

<Axes: xlabel='hour'>

4



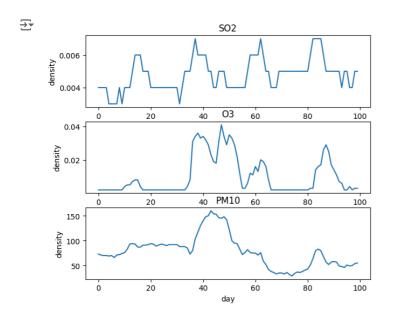
• 요인은 미세먼지를 수집하는 장치의 위치에 따라 다름!

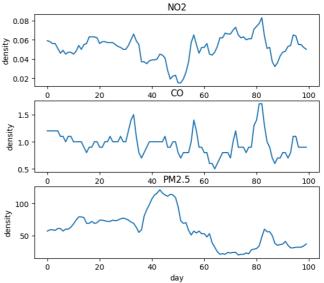
## ∨ 3.2 미세먼지에 영항을 미치는 변인 확인

```
# 데이터 시각화
plt.figure(figsize=(15, 10))

for i in range(4, 10):
    y = df.iloc[:, i]
    plt.subplot(5, 2, i-3)
    plt.title(y.name)
    plt.xlabel('day')
    plt.ylabel('density')
    plt.plot(y[:100])
```

# 03, !N02 와 관계가 있음!





# 3.3 관측소 위치 Longtitude & Latitude

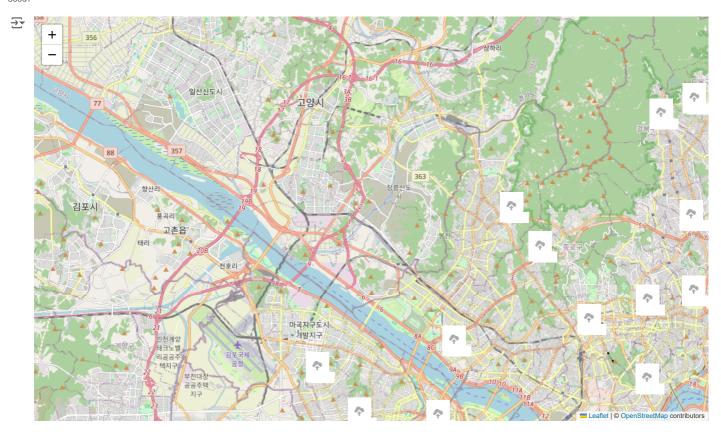
```
# 위도 경도 DataFrame
location = df.groupby('Station code')['PM10'].agg([np.mean])
location['Latitude'] = df['Latitude'].unique() # 절대 이렇게 코드짜면 안되요!
location['Longitude'] = df['Longitude'].unique()
location.head()
```

<ipython-input-17-72f461a0b840>:2: FutureWarning: The provided callable <function mean at 0x7ea1ea393d90> is currently using SeriesGroupBy.mean. location = df.groupby('Station code')['PM10'].agg([np.mean])

	mean	Latitude	Longitude
Station code			
101	37.965605	37.572016	127.005008
102	37.970469	37.564263	126.974676
103	35.539183	37.540033	127.004850
104	42.328468	37.609823	126.934848
105	41.437737	37.593742	126.949679

import folium
seoul = folium.Map(location=[37.55138077230307, 126.98712254969668], zoom\_start=12)
for i in range(len(location)):

marker = folium.Marker([location.iloc[i,1], location.iloc[i,2]], icon=folium.lcon(popup=str(location.index[i]), color='blue', icon='glyphicon gly folium.Marker([37.55195608145124, 127.07362532752212], icon=folium.lcon(popup='Sejoing Univ', color='red', icon='glyphicon glyphicon-home')).add\_to(sepul



```
import folium
from folium.plugins import MarkerCluster
def color_select(x):
    if x >= 45:
        return 'red'
    elif x >= 40:
        return 'yellow'
    else:
        return 'blue'
```

#### # Map

seoul = folium.Map(location=[37.55138077230307, 126.98712254969668], zoom\_start=12)

#### # Circle

for i in range(len(location)):

# 관측소

folium.Circle(location=[location.iloc[i,1], location.iloc[i,2]], radius = location.iloc[i, 0]\*20, color=color\_select(location.iloc[i,0]),fill\_col # Marker / Sejong Univ.

folium.Marker([37.55195608145124, 127.07362532752212], icon=folium.lcon(popup='Sejoing Univ.', color='red', icon='glyphicon glyphicon-home')).add\_to(sequil

