

Your name:

Assignment Name: CA01 - Data Cleaning and Exploration of India Air Quality

## ▼ Program Initialization Section

Enter your import packages here

```
# import packages
import pandas as pd
import numpy as np
```

## ▼ Data File Reading Section

Write code to read in data from external sources here

```
#read datasets
df = pd.read_csv('/content/data.csv',na_values=0, encoding='cp1252')
df
```

```
/usr/local/lib/python3.6/dist-packages/IPython/core/interactiveshell.py:2718: DtypeWarning:
interactivity=interactivity, compiler=compiler, result=result)
```

|   | stn_code | sampling_date         | state             | location  | agency | type                                     | so2 | no:  |
|---|----------|-----------------------|-------------------|-----------|--------|--|-----|------|
| 0 | 150      | February -<br>M021990 | Andhra<br>Pradesh | Hyderabad | NaN    | Residential,<br>Rural and<br>other Areas | 4.8 | 17.4 |
| 1 | 151      | February -<br>M021990 | Andhra<br>Pradesh | Hyderabad | NaN    | Industrial<br>Area                       | 3.1 | 7.0  |
| 2 | 152      | February -<br>M021990 | Andhra<br>Pradesh | Hyderabad | NaN    | Residential,<br>Rural and<br>other Areas | 6.2 | 28.5 |
|   |          |                       |                   |           |        | Residential                              |     |      |

## ▼ Initial Data Investigation Section

### Summarized details

Generate descriptive statistics that summarize the central tendency, dispersion, and shape of a dataset's distribution, excluding NaN values.

Steps:

1. Statistical Description of data (data.describe)
2. Display number of total rows and columns of the dataset (data.shape)
3. Display number of non-null values for each column (data.count)
4. Display number of null values for each column (sum of data.isnull)
5. Display range, column, number of non-null objects of each column, datatype and memory usage (data.info)
6. Display Top 10 and Bottom 10 records (head and tail)

```
# Your code for this section here ...
df.describe()
```

|              | so2           | no2           | rspm          | spm           | pm2_5       |
|--------------|---------------|---------------|---------------|---------------|-------------|
| <b>count</b> | 400221.000000 | 418724.000000 | 394754.000000 | 197142.000000 | 9314.000000 |
| <b>mean</b>  | 10.853091     | 25.858009     | 109.043969    | 222.141944    | 40.791467   |
| <b>std</b>   | 11.177910     | 18.486613     | 74.791245     | 150.863648    | 30.832525   |

df.shape

(435742, 13)

so2 0.000000 22.000000 90.000000 100.000000 32.000000

df.count()

```

stn_code          291665
sampling_date     435739
state             435742
location          435739
agency            286261
type              430349
so2               400221
no2               418724
rspm              394754
spm               197142
location_monitoring_station 408251
pm2_5              9314
date              435735
dtype: int64

```

df.isnull().sum()

```

stn_code          144077
sampling_date         3
state                0
location             3
agency              149481
type                 5393
so2                  35521
no2                  17018
rspm                 40988
spm                 238600
location_monitoring_station 27491
pm2_5                426428
date                  7
dtype: int64

```

df.info()

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 435742 entries, 0 to 435741
Data columns (total 13 columns):
#   Column              Non-Null Count  Dtype
---  -

```

```

0    stn_code                291665 non-null object
1    sampling_date           435739 non-null object
2    state                   435742 non-null object
3    location                435739 non-null object
4    agency                  286261 non-null object
5    type                    430349 non-null object
6    so2                     400221 non-null float64
7    no2                     418724 non-null float64
8    rspm                    394754 non-null float64
9    spm                     197142 non-null float64
10   location_monitoring_station 408251 non-null object
11   pm2_5                   9314 non-null float64
12   date                    435735 non-null object
dtypes: float64(5), object(8)
memory usage: 43.2+ MB

```

```
df.head(10)
```

|   | stn_code | sampling_date      | state          | location  | agency | type                               | so2 | no2  | rspm | spm |
|---|----------|--------------------|----------------|-----------|--------|------------------------------------|-----|------|------|-----|
| 0 | 150      | February - M021990 | Andhra Pradesh | Hyderabad | NaN    | Residential, Rural and other Areas | 4.8 | 17.4 | NaN  | NaN |
| 1 | 151      | February - M021990 | Andhra Pradesh | Hyderabad | NaN    | Industrial Area                    | 3.1 | 7.0  | NaN  | NaN |
| 2 | 152      | February - M021990 | Andhra Pradesh | Hyderabad | NaN    | Residential, Rural and other Areas | 6.2 | 28.5 | NaN  | NaN |
| 3 | 150      | March - M031990    | Andhra Pradesh | Hyderabad | NaN    | Residential, Rural and other Areas | 6.3 | 14.7 | NaN  | NaN |
| 4 | 151      | March - M031990    | Andhra Pradesh | Hyderabad | NaN    | Industrial Area                    | 4.7 | 7.5  | NaN  | NaN |
| 5 | 152      | March - M031990    | Andhra Pradesh | Hyderabad | NaN    | Residential                        | 6.2 | 28.5 | NaN  | NaN |
| 6 | 150      | March - M031990    | Andhra Pradesh | Hyderabad | NaN    | Residential, Rural and other Areas | 6.3 | 14.7 | NaN  | NaN |
| 7 | 151      | March - M031990    | Andhra Pradesh | Hyderabad | NaN    | Industrial Area                    | 4.7 | 7.5  | NaN  | NaN |
| 8 | 152      | March - M031990    | Andhra Pradesh | Hyderabad | NaN    | Residential                        | 6.2 | 28.5 | NaN  | NaN |
| 9 | 150      | March - M031990    | Andhra Pradesh | Hyderabad | NaN    | Residential, Rural and other Areas | 6.3 | 14.7 | NaN  | NaN |

```
df.tail(10)
```

|        | stn_code | sampling_date | state       | location | agency                                    | type  | so2  | no2  | r |
|--------|----------|---------------|-------------|----------|---|-------|------|------|---|
| 435732 | SAMP     | 09-12-15      | West Bengal | ULUBERIA | West Bengal State Pollution Control Board | RIRUO | 22.0 | 50.0 | 1 |
| 435733 | SAMP     | 12-12-15      | West Bengal | ULUBERIA | West Bengal State Pollution Control       | RIRUO | 34.0 | 61.0 | 1 |

## ▼ Cleansing the dataset

### Dropping of less valued columns:

1. stn\_code, agency, sampling\_date, location\_monitoring\_agency do not add much value to the dataset in terms of information. Therefore, we can drop those columns.
2. Dropping rows where no date is available.

```

435735    SAMP    18-12-15    West Bengal    ULUBERIA    RIRUO    RIRUO    17.0    44.0    1
# Cleaning up the data

#dropping columns that aren't required

# ... your code here

# dropping rows where no date is available

# ... your code here

#data columns:  stn_code  sampling_date  state  location  agency  type  so2  no2  rspm  spm  locat
#Drop stn_code, agency, sampling_date, location_monitoring_agency
#New data: state  location  type  so2  no2  rspm  spm  pm2_5  date

df = df.loc[:,['state', 'location', 'type', 'so2', 'no2', 'rspm', 'spm', 'pm2_5', 'date']]

# displaying final columns (data.columns)
df.columns

# ... your code here
#Drop columns where there is NaN in Date

#Doesn't work
#df['date'].dropna(how = "all", axis = 0,inplace = True)
df = df[df['date'].notnull()]

```

```
#No null values in those columns
```

```
df.isnull().sum()
```

```
state      0
location   0
type      5390
so2       35518
no2       17015
rspm      40985
spm       238593
pm2_5     426421
date       0
dtype: int64
```

```
df
```

|        | state          | location  | type                               | so2  | no2  | rspm  | spm | pm2_5 | date       |
|--------|----------------|-----------|------------------------------------|------|------|-------|-----|-------|------------|
| 0      | Andhra Pradesh | Hyderabad | Residential, Rural and other Areas | 4.8  | 17.4 | NaN   | NaN | NaN   | 1990-02-01 |
| 1      | Andhra Pradesh | Hyderabad | Industrial Area                    | 3.1  | 7.0  | NaN   | NaN | NaN   | 1990-02-01 |
| 2      | Andhra Pradesh | Hyderabad | Residential, Rural and other Areas | 6.2  | 28.5 | NaN   | NaN | NaN   | 1990-02-01 |
| 3      | Andhra Pradesh | Hyderabad | Residential, Rural and other Areas | 6.3  | 14.7 | NaN   | NaN | NaN   | 1990-03-01 |
| 4      | Andhra Pradesh | Hyderabad | Industrial Area                    | 4.7  | 7.5  | NaN   | NaN | NaN   | 1990-03-01 |
| ...    | ...            | ...       | ...                                | ...  | ...  | ...   | ... | ...   | ...        |
| 435734 | West Bengal    | ULUBERIA  | RIRUO                              | 20.0 | 44.0 | 148.0 | NaN | NaN   | 2015-12-15 |
| 435735 | West Bengal    | ULUBERIA  | RIRUO                              | 17.0 | 44.0 | 131.0 | NaN | NaN   | 2015-12-18 |

## ▼ Changing the types to uniform format:

Notice that the 'type' column has values such as 'Industrial Area' and 'Industrial Areas'—both actually mean the same, so let's remove such type of stuff and make it uniform. Replace the 'type' values with standard codes as follows:

```
types = { "Residential": "R", "Residential and others": "RO", "Residential, Rural and other Areas": "RRO", "Industrial Area": "I", "Industrial Areas": "I", "Industrial": "I", "Sensitive Area": "S", "Sensitive Areas": "S", "Sensitive": "S", np.nan: "RRO" }
```

```
data.type = data.type.replace(types)
```

```
# ... Your code here
#df['type'].replace({ "Residential": "R", "Residential and others": "RO", "Residential, Rural and other /
types = { "Residential": "R", "Residential and others": "RO", "Residential, Rural and other /
df.type = df.type.replace(types)
```

```
df
```

|               | state          | location  | type  | so2  | no2  | rspm  | spm | pm2_5 | date       |
|---------------|----------------|-----------|-------|------|------|-------|-----|-------|------------|
| <b>0</b>      | Andhra Pradesh | Hyderabad | RRO   | 4.8  | 17.4 | NaN   | NaN | NaN   | 1990-02-01 |
| <b>1</b>      | Andhra Pradesh | Hyderabad | I     | 3.1  | 7.0  | NaN   | NaN | NaN   | 1990-02-01 |
| <b>2</b>      | Andhra Pradesh | Hyderabad | RRO   | 6.2  | 28.5 | NaN   | NaN | NaN   | 1990-02-01 |
| <b>3</b>      | Andhra Pradesh | Hyderabad | RRO   | 6.3  | 14.7 | NaN   | NaN | NaN   | 1990-03-01 |
| <b>4</b>      | Andhra Pradesh | Hyderabad | I     | 4.7  | 7.5  | NaN   | NaN | NaN   | 1990-03-01 |
| ...           | ...            | ...       | ...   | ...  | ...  | ...   | ... | ...   | ...        |
| <b>435734</b> | West Bengal    | ULUBERIA  | RIRUO | 20.0 | 44.0 | 148.0 | NaN | NaN   | 2015-12-15 |
| <b>435735</b> | West Bengal    | ULUBERIA  | RIRUO | 17.0 | 44.0 | 131.0 | NaN | NaN   | 2015-12-18 |
| <b>435736</b> | West Bengal    | ULUBERIA  | RIRUO | 18.0 | 45.0 | 140.0 | NaN | NaN   | 2015-12-21 |
| <b>435737</b> | West Bengal    | ULUBERIA  | RIRUO | 22.0 | 50.0 | 143.0 | NaN | NaN   | 2015-12-24 |
| <b>435738</b> | West Bengal    | ULUBERIA  | RIRUO | 20.0 | 46.0 | 171.0 | NaN | NaN   | 2015-12-29 |

435735 rows × 9 columns

```
# Display top 10 records after codification of 'types'
# ... Your code here
#
df.head(10)
```

|   | state          | location  | type | so2 | no2  | rspm | spm | pm2_5 | date       |
|---|----------------|-----------|------|-----|------|------|-----|-------|------------|
| 0 | Andhra Pradesh | Hyderabad | RRO  | 4.8 | 17.4 | NaN  | NaN | NaN   | 1990-02-01 |

## ▼ Creating a year column

To view the trend over a period of time, we need year values for each row and also when you see in most of the values in date column only has 'year' value. So, let's create a new column holding year values. Convert the column to 'datetime' type and extract the year to populate the new column.

Display Top 5 records after the conversion.

```
6 Andhra Pradesh Hyderabad RRO 5.4 17.1 NaN NaN NaN 1990-04-01
```

```
# ... Your code here
#convert to datetime
df['date']= pd.to_datetime(df['date'])
#Extract year to populate new column year
df['year'] = df['date'].dt.year
```

df

|        | state          | location  | type  | so2  | no2  | rspm  | spm | pm2_5 | date       | year |
|--------|----------------|-----------|-------|------|------|-------|-----|-------|------------|------|
| 0      | Andhra Pradesh | Hyderabad | RRO   | 4.8  | 17.4 | NaN   | NaN | NaN   | 1990-02-01 | 1990 |
| 1      | Andhra Pradesh | Hyderabad | I     | 3.1  | 7.0  | NaN   | NaN | NaN   | 1990-02-01 | 1990 |
| 2      | Andhra Pradesh | Hyderabad | RRO   | 6.2  | 28.5 | NaN   | NaN | NaN   | 1990-02-01 | 1990 |
| 3      | Andhra Pradesh | Hyderabad | RRO   | 6.3  | 14.7 | NaN   | NaN | NaN   | 1990-03-01 | 1990 |
| 4      | Andhra Pradesh | Hyderabad | I     | 4.7  | 7.5  | NaN   | NaN | NaN   | 1990-03-01 | 1990 |
| ...    | ...            | ...       | ...   | ...  | ...  | ...   | ... | ...   | ...        | ...  |
| 435734 | West Bengal    | ULUBERIA  | RIRUO | 20.0 | 44.0 | 148.0 | NaN | NaN   | 2015-12-15 | 2015 |
| 435735 | West Bengal    | ULUBERIA  | RIRUO | 17.0 | 44.0 | 131.0 | NaN | NaN   | 2015-12-18 | 2015 |

## ▼ Handling Missing Values

The column such as SO2, NO2, rspm, spm, pm2\_5 are the ones which contribute much to our analysis. So, we need to remove null from those columns to avoid inaccuracy in the prediction. We



use the Imputer from sklearn.preprocessing to fill the missing values in every column with the

```
# define columns of importance, which shall be used regularly (COLS = ....)
# invoke SimpleImputer to fill missing values using 'mean' as the replacement strategy
# Display data.info after the transformation
# Display that there are no more missing values in the dataset

# ... your code here
#remove null values from SO2, NO2, rspm, spm, pm2_5
#fill missing value with imputer from sklearn
from sklearn.impute import SimpleImputer
#imp = SimpleImputer(missing_values=-1, strategy='mean')
dfnew = df.loc[:,['so2','no2', 'rspm', 'spm', 'pm2_5']]
imp = SimpleImputer(missing_values = np.nan, strategy='mean')
#Fit then transform
imp.fit(dfnew)
dftransformed = imp.transform(dfnew)
dftransformednew = imp.transform(dftransformed)
#Dataframe turned into nested list arrays
#now convert it back into it being a dataframe
dfnew = pd.DataFrame(dftransformednew)
dfnew.columns = ['so2','no2', 'rspm', 'spm', 'pm2_5']

#Drop 'so2','no2', 'rspm', 'spm', 'pm2_5' in original dataframe
df = df.loc[:,['state', 'location', 'type', 'date', 'year']]
result = pd.concat([df, dfnew], axis=1)

#Figure out a way to do this without dropping columns !!! DO LATER
#result2 = pd.merge(df, dfnew, how='right')
```

## ▼ Statewise Grouping of so2, no2, rspm, spm values

Calculate median values of so2, no2, rspm, spm for each state and display in (a) as table (b) bar chart, with values sorted in ascending order. Separate section for each of the component. Use matplotlib().

```
result
check = result.groupby('state')['so2'].agg(['median'])
```

## ▼ so2 status

```
# ... Your code here
import matplotlib.pyplot as plt
dfso2 = result.groupby('state')['so2'].agg(['median'])
```

```
#sort ascending table then barplot
dfso2 = dfso2.sort_values(by='median')
dfso2['state'] = dfso2.index
dfso2.state.count()

n = 34
ind = np.arange(n)
width = 0.35
dfso2med = dfso2['median']

plt.bar(ind, dfso2med, width)

plt.xticks(ind+width / 2, dfso2.index, rotation=90)
```

```
([<matplotlib.axis.XTick at 0x7f911dd87630>,
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Text(0, 0, 'Delhi'),
Text(0, 0, 'Assam'),
Text(0, 0, 'Puducherry'),
Text(0, 0, 'Daman & Diu'),
Text(0, 0, 'Chandigarh'),
Text(0, 0, 'West Bengal'),
Text(0, 0, 'Dadra & Nagar Haveli'),
Text(0, 0, 'Karnataka'),
```

▼ no2 status

```
Text(0, 0, 'Manipur'),

# ... Your code here
dfno2 = result.groupby('state')['no2'].agg(['median'])
#Sort ascending table then barplot
dfno2 = dfno2.sort_values(by='median')
dfno2['state'] = dfno2.index
dfno2.state.count()

n = 34
ind = np.arange(n)
width = 0.35
dfno2med = dfno2['median']

plt.bar(ind, dfno2med, width)

plt.xticks(ind+width / 2, dfno2.index, rotation=90)
```

```
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Text(0, 0, 'Odisha'),
Text(0, 0, 'Daman & Diu'),
Text(0, 0, 'Dadra & Nagar Haveli'),
Text(0, 0, 'Chandigarh'),
Text(0, 0, 'Manipur'),
Text(0, 0, 'Madhya Pradesh'),
Text(0, 0, 'Andhra Pradesh'),
Text(0, 0, 'Tamil Nadu'),
Text(0, 0, 'Karnataka'),
Text(0, 0, 'Chhattisgarh'),
Text(0, 0, 'Telangana')])
```

## ▼ rspm status

```
Test(0, 0, 'Redacted')

# ... Your code here
dfrspm = result.groupby('state')['rspm'].agg(['median'])
#Sort ascending table then barplot
dfrspm = dfrspm.sort_values(by='median')
dfrspm['state'] = dfrspm.index
dfrspm.state.count()

n = 34
ind = np.arange(n)
width = 0.35
dfrspmmed = dfrspm['median']

plt.bar(ind, dfrspmmed, width)

plt.xticks(ind+width / 2, dfrspm.index, rotation=90)
```

```
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Text(0, 0, 'Daman & Diu'),
Text(0, 0, 'Bihar')]
```

```
Text(0, 0, 'Sikkim'),
```

## ▼ spm status

```
Text(0, 0, 'Jammu & Kashmir' ),

# ... Your code here
dfspm = result.groupby('state')['spm'].agg(['median'])
#Sort ascending table then barplot
dfspm = dfspm.sort_values(by='median')
dfspm['state'] = dfspm.index
dfspm.state.count()

n = 34
ind = np.arange(n)
width = 0.35
dfspmmed = dfspm['median']

plt.bar(ind, dfspmmed, width)

plt.xticks(ind+width / 2, dfspm.index, rotation=90)
```



```
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Text(0, 0, 'Jammu & Kashmir'),
Text(0, 0, 'Himachal Pradesh'),
Text(0, 0, 'Haryana'),
```

```
Text(0, 0, 'Gujarat'),
Text(0, 0, 'Goa'),
Text(0, 0, 'Dadra & Nagar Haveli'),
Text(0, 0, 'Chhattisgarh'),
```

## ▼ What is the yearly trend in a particular state, say 'Andhra Pradesh'?

Create a new dataframe containing the NO2, SO2, rspm, and spm data regarding state 'Andhra Pradesh' only and group it by 'year'. Display top 5 records after.

```
text(0, 0, 'uttaranchal '))
```

result

|        | state          | location  | type  | date       | year   | so2  | no2  | rspm       | spm        |     |
|--------|----------------|-----------|-------|------------|--------|------|------|------------|------------|-----|
| 0      | Andhra Pradesh | Hyderabad | RRO   | 1990-02-01 | 1990.0 | 4.8  | 17.4 | 109.044278 | 222.141944 | 40. |
| 1      | Andhra Pradesh | Hyderabad | I     | 1990-02-01 | 1990.0 | 3.1  | 7.0  | 109.044278 | 222.141944 | 40. |
| 2      | Andhra Pradesh | Hyderabad | RRO   | 1990-02-01 | 1990.0 | 6.2  | 28.5 | 109.044278 | 222.141944 | 40. |
| 3      | Andhra Pradesh | Hyderabad | RRO   | 1990-03-01 | 1990.0 | 6.3  | 14.7 | 109.044278 | 222.141944 | 40. |
| 4      | Andhra Pradesh | Hyderabad | I     | 1990-03-01 | 1990.0 | 4.7  | 7.5  | 109.044278 | 222.141944 | 40. |
| ...    | ...            | ...       | ...   | ...        | ...    | ...  | ...  | ...        | ...        | ... |
| 435734 | West Bengal    | ULUBERIA  | RIRUO | 2015-12-15 | 2015.0 | 20.0 | 46.0 | 171.000000 | 222.141944 | 40. |
| 435735 | West Bengal    | ULUBERIA  | RIRUO | 2015-12-18 | 2015.0 | NaN  | NaN  | NaN        | NaN        | NaN |

dfandhranew

|        | no2       | so2       | rspm       | spm        | year   |
|--------|-----------|-----------|------------|------------|--------|
|        | mean      | mean      | mean       | mean       |        |
| year   |           |           |            |            |        |
| 1990.0 | 15.296552 | 8.868966  | 109.044278 | 165.984875 | 1990.0 |
| 1991.0 | 18.827778 | 10.811111 | 109.044278 | 154.503943 | 1991.0 |
| 1992.0 | 36.013978 | 19.766738 | 109.044278 | 210.643678 | 1992.0 |
| 1993.0 | 15.961111 | 8.544444  | 109.044278 | 222.141944 | 1993.0 |
| 1994.0 | 18.273016 | 10.760317 | 109.044278 | 178.677526 | 1994.0 |
| 1995.0 | 33.531868 | 17.225275 | 109.044278 | 167.293513 | 1995.0 |
| 1996.0 | 36.589482 | 19.861099 | 109.044278 | 180.345095 | 1996.0 |
| 1997.0 | 38.163330 | 20.339507 | 109.044278 | 177.785714 | 1997.0 |
| 1998.0 | 26.409459 | 11.491892 | 109.044278 | 175.351351 | 1998.0 |
| 1999.0 | 19.887963 | 14.384259 | 109.044278 | 197.509259 | 1999.0 |
| 2000.0 | 23.882019 | 13.856973 | 109.044278 | 140.935185 | 2000.0 |
| 2001.0 | 27.105714 | 12.000952 | 109.044278 | 142.790476 | 2001.0 |
| 2002.0 | 24.976190 | 7.506667  | 109.044278 | 85.866667  | 2002.0 |
| 2004.0 | 31.430507 | 7.338767  | 85.384298  | 170.814559 | 2004.0 |
| 2005.0 | 30.081095 | 6.370926  | 79.468608  | 220.257282 | 2005.0 |

```
# ... Your code here
```

```
dfandhra = result.loc[result.state == 'Andhra Pradesh', ['no2', 'so2', 'rspm', 'spm', 'year']]
dfandhranew = dfandhra.groupby('year').agg(['mean'])
```

```
# Display yearly trend graph (year vs. value) in pairs: (a) so2 and no2 (b) rspm and spm.
# So, you will display TWO graphs altogether.
```

```
# ... Your code here
```

```
so2mean = dfandhranew['so2']
no2mean = dfandhranew['no2']
```

```
plt.plot(so2mean, ls='--', color='red', marker='s', markerfacecolor='white', label='so2')
plt.plot(no2mean, ls='-.', color='g', marker='o', markerfacecolor='grey', label='no2')
plt.legend()
plt.title('Mean NO2 vs SO2 for Andhra Pradesh')
plt.xlabel('Year')
plt.ylabel('Gas Level')
plt.show()
```