


```
1 #Check for pytorch being installed  
2 import torch  
3 print(torch.__version__)
```

2.3.0+cu121

```
1 #Libraries  
2 import pandas as pd  
3 import os  
4 import functools  
5 import numpy as np  
6 from statsmodels.tsa.seasonal import STL  
7 import matplotlib.pyplot as plt  
8 import sklearn  
9 from sklearn.impute import KNNImputer  
10  
11 # Set the maximum number of rows and columns to display  
12 pd.set_option('display.max_rows', None) # None means unlimited  
13 pd.set_option('display.max_columns', None) # None means unlimited
```

```
1 #Create dfs to handle data  
2 US_df = pd.read_csv(r"C:\Users\User\Code\databook\data\US\CombinedData.csv")  
3  
4  
5 print(US_df)
```

		Date	Time	PM2.5	PM10	CO	NO2	O3	\
0	2013-01-01	2013-01-01	00:00:00	12.0	16.0	0.400	20.9	0.012	
1	2013-01-01	2013-01-01	01:00:00	16.0	18.0	0.400	23.2	0.021	
2	2013-01-01	2013-01-01	02:00:00	9.0	10.0	0.300	18.3	0.000	
3	2013-01-01	2013-01-01	03:00:00	8.0	10.0	0.300	16.9	0.019	
4	2013-01-01	2013-01-01	04:00:00	10.0	7.0	0.400	19.9	0.016	
5	2013-01-01	2013-01-01	05:00:00	13.0	12.0	0.400	24.1	0.009	
6	2013-01-01	2013-01-01	06:00:00	12.0	9.0	0.400	17.8	0.017	
7	2013-01-01	2013-01-01	07:00:00	9.9	6.0	0.300	22.2	0.013	
8	2013-01-01	2013-01-01	08:00:00	6.0	6.0	0.300	18.1	0.014	
9	2013-01-01	2013-01-01	09:00:00	6.0	7.0	0.200	14.2	0.023	
10	2013-01-01	2013-01-01	10:00:00	9.8	9.0	0.200	15.7	0.029	
11	2013-01-01	2013-01-01	11:00:00	7.5	8.0	0.200	12.7	0.032	
12	2013-01-01	2013-01-01	12:00:00	6.9	3.0	0.200	6.4	0.015	
13	2013-01-01	2013-01-01	13:00:00	15.8	3.0	0.200	6.3	0.038	
14	2013-01-01	2013-01-01	14:00:00	13.9	1.0	0.200	8.0	0.038	
15	2013-01-01	2013-01-01	15:00:00	16.9	7.0	0.200	8.7	0.037	
16	2013-01-01	2013-01-01	16:00:00	21.6	5.0	0.200	8.6	0.031	
17	2013-01-01	2013-01-01	17:00:00	26.4	17.0	0.200	11.5	0.023	
18	2013-01-01	2013-01-01	18:00:00	31.9	7.0	0.300	12.2	0.020	
19	2013-01-01	2013-01-01	19:00:00	30.9	16.0	0.300	11.1	0.018	
20	2013-01-01	2013-01-01	20:00:00	30.9	16.0	0.300	9.1	0.026	
21	2013-01-01	2013-01-01	21:00:00	32.9	9.0	0.300	9.9	0.025	
22	2013-01-01	2013-01-01	22:00:00	25.1	13.0	0.200	6.9	0.019	
23	2013-01-01	2013-01-01	23:00:00	22.1	10.0	0.300	9.2	0.001	
24	2013-01-02	2013-01-02	00:00:00	17.0	10.0	0.200	6.9	0.022	
25	2013-01-02	2013-01-02	01:00:00	15.0	7.0	0.200	6.7	0.005	
26	2013-01-02	2013-01-02	02:00:00	25.9	5.0	0.200	4.9	0.004	

27	2013-01-02	03:00:00	10.0	16.0	0.200	4.6	0.025
28	2013-01-02	04:00:00	31.6	3.0	0.200	5.1	0.008
29	2013-01-02	05:00:00	30.9	15.0	0.200	6.3	0.024
30	2013-01-02	06:00:00	25.9	18.0	0.200	8.3	0.023
31	2013-01-02	07:00:00	14.0	13.0	0.200	10.7	0.014
32	2013-01-02	08:00:00	21.0	18.0	0.300	10.1	0.013
33	2013-01-02	09:00:00	26.5	18.0	0.200	9.9	0.014
34	2013-01-02	10:00:00	16.0	17.0	0.200	7.9	0.025
35	2013-01-02	11:00:00	16.7	11.0	0.200	6.7	0.035
36	2013-01-02	12:00:00	9.0	10.0	0.200	6.0	0.017
37	2013-01-02	13:00:00	11.0	17.0	0.200	5.8	0.018
38	2013-01-02	14:00:00	12.0	10.0	0.200	6.7	0.030
39	2013-01-02	15:00:00	12.0	9.0	0.200	8.5	0.022
40	2013-01-02	16:00:00	13.0	10.0	0.300	10.8	0.027
41	2013-01-02	17:00:00	18.7	15.0	0.300	16.3	0.020
42	2013-01-02	18:00:00	17.7	14.0	0.300	21.3	0.017
43	2013-01-02	19:00:00	12.0	13.0	0.300	23.1	0.016
44	2013-01-02	20:00:00	12.0	19.0	0.400	29.6	0.002
45	2013-01-02	21:00:00	15.7	13.0	0.300	17.9	0.020
46	2013-01-02	22:00:00	18.9	16.0	0.300	15.5	0.022
47	2013-01-02	23:00:00	13.2	15.0	0.300	17.2	0.001
48	2013-01-03	00:00:00	13.0	18.0	0.300	15.9	0.000
49	2013-01-03	01:00:00	13.0	13.0	0.300	16.7	0.018
50	2013-01-03	02:00:00	20.0	14.0	0.300	16.2	0.017
51	2013-01-03	03:00:00	14.0	14.0	0.300	14.8	0.005
52	2013-01-03	04:00:00	14.0	16.0	0.200	19.3	0.010
53	2013-01-03	05:00:00	13.0	24.0	0.300	24.1	0.011
54	2013-01-03	06:00:00	26.4	25.0	0.400	29.5	0.002
55	2013-01-03	07:00:00	32.0	28.0	0.400	29.8	0.013
56	2013-01-03	08:00:00	23.0	28.0	0.400	33.7	0.005

```

1 # Function to handle datetime conversion and setting index
2 def preprocess_dataframe(df):
3
4     #Ensure that the DateTime is set to a processable format
5     df['DateTime'] = pd.to_datetime(df['DateTime'])
6     df.set_index('DateTime', inplace=True)      #set DateTime to index
7
8     # Convert 'No data' to NaN and change the data type to float so that we can use it in calculations
9     #df.replace('No data', np.nan, inplace=True)
10    df = df.apply(pd.to_numeric, errors='coerce')
11
12
13
14    return df

```

```

1 #Function to remove Outliers using InterQuartile Range method
2 def remove_outliers(df, column):
3
4     # Calculate Q1, Q3 and IQR
5     Q1 = df[column].quantile(0.25)
6     Q3 = df[column].quantile(0.75)
7     IQR = Q3 - Q1
8
9     # Define bounds for outliers
10    lower_bound = Q1 - 2.5 * IQR
11    upper_bound = Q3 + 3.5 * IQR

```

```
11     . . .
12     print(f"Working on {column}")
13
14     # Replace outliers with NaN
15     def handle_outliers(x):
16         original_x = x # store original value
17         if x < lower_bound or x > upper_bound or x < 0:
18             x = np.nan
19             print(f"Original: {original_x}, Modified: {x}") # Print
20         return x
21
22     df[column] = df[column].apply(handle_outliers)
23
24
25     return df
```

```
1 #Function to remove Outliers using STL method
2 def remove_outliers_STL(df, column, seasonal=49):
3
4     #Replace any negative values with NaN
5     df[column] = df[column].apply(lambda x: np.nan if x < 0 else x)
6
7     #Print length of column
8     print(f"The length of the {column} column is {len(df[column]).drop
9
10    # STL Decomposition
11    stl = STL(df[column], seasonal=seasonal , robust=True) #Seasonal
12    result = stl.fit()
13    df['residuals'] = result.resid
14
15    # Detect anomalies in residuals
16    threshold = 5 * np.std(df['residuals'].dropna()) # Calculate thr
17    df['stl_anomaly'] = np.abs(df['residuals']) > threshold # Identifi
18
19    print(f>Data being checked: {column}")
20    checking_df = df[[column,'stl_anomaly']]
21
22    #Checking quality of anomaly detection
23    anomalies = checking_df[df['stl_anomaly']== True]
24
25    print(anomalies)
26
27    # Replace anomalies with NaN
28    df.loc[df['stl_anomaly'], column] = np.nan # Replace detected an
29
30
31    #remove stl anomoly column so that it can be used again for the n
32    df.drop('stl_anomaly', axis=1, inplace=True)
33
34
35     return df
```

```
1 #Removing outliers from Data
```

```
1 #REMOVING OUTLIERS FROM DATA
2 def apply_outliers_removal(df, time_column, seasonal=49):
3     """
4     Parameters:
5         Input:
6             df - Dataframe to be processed by function
7             time_column - column that will be ignored by function.
8         """
9     for col in df.columns:
10         if col != time_column:
11             df = remove_outliers_STL(df, col, seasonal)
12     return df
13
```

```
1 #Function applying STL to each column of the dataset to fill in the missing values
2
3 def STL_fill_in_data(df, period):
4     """
5     Parameters:
6         Input: df - Dataframe who's data will be filled
7         period - The time period in which the pattern for the seasonal trend will be applied
8     """
9
10    for column in df.columns:
11        if df[column].isnull().any():
12            # Ensure the column is not all NaN and has enough data points
13            if df[column].dropna().shape[0] < period:
14                print(f"Not enough data points in {column} to apply STL decomposition")
15                continue
16
17            # Apply STL decomposition
18            stl = STL(df[column].interpolate(), seasonal=period)
19            res = stl.fit()
20
21            # Deseasonalize and impute
22            seasonal_component = res.seasonal
23            deseasonalised = df[column] - seasonal_component
24            deseasonalised_imputed = deseasonalised.interpolate(method='ffill')
25
26            # Recombine and fill missing values
27            df_imputed = deseasonalised_imputed + seasonal_component
28            imputed_indices = df[df[column].isnull()].index
29
30            # Ensure no negative values
31            df_imputed = df_imputed.apply(lambda x: max(x, 0))
32
33            df.loc[imputed_indices, column] = df_imputed[imputed_indices]
34
35    return df
36
```

```
1 def KNN_fill_in_data(df, n_neighbors=5, weights='distance'):
2     """
```

```
3     Impute missing data in each column of the dataset using the KNN m
4
5     Parameters:
6         df (pd.DataFrame): DataFrame whose data will be filled.
7         n_neighbors (int): Number of neighboring samples to use for imput
8
9     Returns:
10    pd.DataFrame: DataFrame with missing values imputed using KNN.
11    """
12    imputer = KNNImputer(n_neighbors=n_neighbors, weights = weights)
13
14    # Ensure all data is numeric for KNN imputer, converts non-numeri
15    df_numeric = df.apply(pd.to_numeric, errors='coerce')
16
17    # Apply KNN imputation
18    imputed_data = imputer.fit_transform(df_numeric)
19
20    # Create a DataFrame with the imputed data (NB changed from: colu
21    df_imputed = pd.DataFrame(imputed_data, columns=df_numeric.column
22
23    # Replace any resulting negative values with zero, assuming only
24    df_imputed = df_imputed.applymap(lambda x: max(x, 0))
25
26    return df_imputed
```

```
1 print(US_df.columns.to_list)

<bound method IndexOpsMixin.tolist of Index(['DateTime', 'PM2.5', 'PM10',
      'Temp'],
      dtype='object')>
```

```
1 #US data Cleaning
2 #Create dfs to handle data
3 US_df = pd.read_csv(r"C:\Users\User\Code\databook\data\US\CombinedDat
4
5 US_PM_df = US_df[['DateTime', 'PM2.5', 'PM10']]
6
7 US_pp_df = preprocess_dataframe(US_PM_df)
8 US_or_df = apply_outliers_removal(US_pp_df, 'DateTime')
9 US_or_df2 = US_or_df.copy(deep=True) #Creating a completely seperate
10 US_or_df2['DateTime'] = US_or_df2.index
11 US_K_fill = KNN_fill_in_data(US_or_df2, 26)
12
13
14 #print(US_fill_df)
```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.loc.html>
df['DateTime'] = pd.to_datetime(df['DateTime'])
The length of the PM2.5 column is 92054
Data being checked: PM2.5

```
Empty DataFrame
Columns: [PM2.5, stl_anomaly]
Index: []
The length of the PM10 column is 93064
Data being checked: PM10
Empty DataFrame
Columns: [PM10, stl_anomaly]
Index: []

-----
ValueError Traceback (most recent call last)
Cell In[101], line 11
      9 US_or_df2 = US_or_df.copy(deep=True) #Creating a completely
     10 #separate object so that the original dataframe can be kept
--> 11 US_or_df2['DateTime'] = US_or_df2.index
     12 US_K_fill = KNN_fill_in_data(US_or_df2, 26)
     13 #print(US_fill_df)

Cell In[100], line 21, in KNN_fill_in_data(df, n_neighbors, weights)
    18 imputed_data = imputer.fit_transform(df_numeric)
    19 # Create a DataFrame with the imputed data (NB changed from:
columns=df.columns )
--> 21 df_imputed = pd.DataFrame(imputed_data,
columns=df_numeric.columns, index=df.index)
    22 # Replace any resulting negative values with zero, assuming only
non-negative values are acceptable
    23 df_imputed = df_imputed.applymap(lambda x: max(x, 0))

File c:\Users\User\Code\databook\Lib\site-
packages\pandas\core\frame.py:827, in DataFrame.__init__(self, data,
index, columns, dtype, copy)
    816         mgr = dict_to_mgr(
    817             # error: Item "ndarray" of "Union[ndarray, Series,
Index]" has no
    818                 # attribute "name"
    819             (...),
    820                 copy=_copy,
    821             )
    822         else:
--> 823             mgr = ndarray_to_mgr(
    824                 data,
    825                 index,
    826                 columns,
    827                 dtype=dtype,
    828                 copy=copy,
    829                 typ=manager,
    830             )
    831         # For data is list-like, or Iterable (will consume into list)
    832     elif is_list_like(data):

File c:\Users\User\Code\databook\Lib\site-
packages\pandas\core\internals\construction.py:336, in
ndarray_to_mgr(values, index, columns, dtype, copy, typ)
    331 # _prep_ndarraylike ensures that values.ndim == 2 at this point
    332 index, columns = _get_axes(
    333     values.shape[0], values.shape[1], index=index, columns=columns
    334 )
    335 check_values_indices_shape_match(values, index, columns)
```

```
1 #UK data cleaning
2 UK_df = pd.read_csv(r"C:\Users\User\Code\databook\data\UK\AllData\UKF
3
4 UK_PM_df = UK_df[['DateTime', 'PM2.5', 'PM10']]
5 UK_pp_df = preprocess_dataframe(UK_PM_df)
6 print(UK_pp_df)
```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <https://pandas.pydata.org/pandas-doc>

```
df['DateTime'] = pd.to_datetime(df['DateTime'])
```

```
PM2.5      PM10
```

DateTime	PM2.5	PM10
2019-01-01 01:00:00	14.741	19.328
2019-01-01 02:00:00	11.408	16.705
2019-01-01 03:00:00	5.407	10.074
2019-01-01 04:00:00	5.503	10.326
2019-01-01 05:00:00	5.755	9.708
2019-01-01 06:00:00	5.527	9.518
2019-01-01 07:00:00	5.667	9.802
2019-01-01 08:00:00	5.512	8.935
2019-01-01 09:00:00	5.576	9.508
2019-01-01 10:00:00	5.780	9.294
2019-01-01 11:00:00	5.616	9.060
2019-01-01 12:00:00	4.002	7.053
2019-01-01 13:00:00	3.767	5.629
2019-01-01 14:00:00	3.788	6.083
2019-01-01 15:00:00	4.141	6.405
2019-01-01 16:00:00	4.679	7.407
2019-01-01 17:00:00	4.531	6.830
2019-01-01 18:00:00	6.571	11.988
2019-01-01 19:00:00	8.677	17.038
2019-01-01 20:00:00	8.935	16.323
2019-01-01 21:00:00	6.283	11.761
2019-01-01 22:00:00	5.826	10.900
2019-01-01 23:00:00	5.973	11.615
2019-01-02 00:00:00	5.973	11.855
2019-01-02 01:00:00	10.267	11.835
2019-01-02 02:00:00	8.309	9.584
2019-01-02 03:00:00	6.833	8.250
2019-01-02 04:00:00	4.404	5.042
2019-01-02 05:00:00	4.539	5.730
2019-01-02 06:00:00	5.783	7.555
2019-01-02 07:00:00	6.381	8.306
2019-01-02 08:00:00	7.630	9.329
2019-01-02 09:00:00	8.259	11.255
2019-01-02 10:00:00	11.094	16.608
2019-01-02 11:00:00	10.472	12.740
2019-01-02 12:00:00	10.675	13.998
2019-01-02 13:00:00	12.769	16.115
2019-01-02 14:00:00	8.520	10.795
2019-01-02 15:00:00	8.267	11.287
2019-01-02 16:00:00	6.920	9.032
2019-01-02 17:00:00	7.923	10.682
2019-01-02 18:00:00	11.163	13.740
2019-01-02 19:00:00	10.962	14.248

```
-----  
2019-01-02 20:00:00 9.688 13.743  
2019-01-02 21:00:00 9.557 13.740  
2019-01-02 22:00:00 10.375 13.145  
2019-01-02 23:00:00 10.576 13.163  
2019-01-03 00:00:00 10.300 12.880  
2019-01-03 01:00:00 15.151 18.780  
2019-01-03 02:00:00 15.061 19.070
```

```
1 #US data cleaning  
2 US_df = pd.read_csv(r"C:\Users\User\Code\databook\data\US\CombinedDat  
3  
4 US_PM_df = US_df[['DateTime', 'PM2.5', 'PM10']]  
5 US_pp_df = preprocess_dataframe(US_PM_df)  
6 US_or_df = apply_outliers_removal(US_pp_df, 'DateTime')  
7 US_or_df2 = US_or_df.copy(deep=True) #Creating a completely seperate
```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stable/indexing.html#inplace>
df['DateTime'] = pd.to_datetime(df['DateTime'])
The length of the PM2.5 column is 92054
Data being checked: PM2.5
Empty DataFrame
Columns: [PM2.5, stl_anomaly]
Index: []
The length of the PM10 column is 93064
Data being checked: PM10
Empty DataFrame
Columns: [PM10, stl_anomaly]
Index: []

```
1 US_or_df3 = US_or_df.copy(deep=True) #Creating a completely seperate  
2  
3 US_or_df3['DateTime'] = US_or_df3.index  
4  
5 US_or_df3.drop('residuals', axis=1, inplace=True)  
6
```

```
1 US_K_fill = KNN_fill_in_data(US_or_df3, 26)  
2 US_K_fill.drop('DateTime', axis=1, inplace=True)  
3 print(US_K_fill)
```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\3676177356.py:24: FutureWarning:
df_imputed = df_imputed.agg(applymap(lambda x: max(x, 0)))

DateTime	PM2.5	PM10
2013-01-01 00:00:00	12.000000	16.000000
2013-01-01 01:00:00	16.000000	18.000000
2013-01-01 02:00:00	9.000000	10.000000
2013-01-01 03:00:00	8.000000	10.000000
2013-01-01 04:00:00	10.000000	7.000000
2013-01-01 05:00:00	13.000000	12.000000
2013-01-01 06:00:00	12.000000	9.000000
2013-01-01 07:00:00	9.000000	6.000000

2013-01-01 07:00:00	0.000000	0.000000
2013-01-01 08:00:00	6.000000	6.000000
2013-01-01 09:00:00	6.000000	7.000000
2013-01-01 10:00:00	9.800000	9.000000
2013-01-01 11:00:00	7.500000	8.000000
2013-01-01 12:00:00	6.900000	3.000000
2013-01-01 13:00:00	15.800000	3.000000
2013-01-01 14:00:00	13.900000	1.000000
2013-01-01 15:00:00	16.900000	7.000000
2013-01-01 16:00:00	21.600000	5.000000
2013-01-01 17:00:00	26.400000	17.000000
2013-01-01 18:00:00	31.900000	7.000000
2013-01-01 19:00:00	30.900000	16.000000
2013-01-01 20:00:00	30.900000	16.000000
2013-01-01 21:00:00	32.900000	9.000000
2013-01-01 22:00:00	25.100000	13.000000
2013-01-01 23:00:00	22.100000	10.000000
2013-01-02 00:00:00	17.000000	10.000000
2013-01-02 01:00:00	15.000000	7.000000
2013-01-02 02:00:00	25.900000	5.000000
2013-01-02 03:00:00	10.000000	16.000000
2013-01-02 04:00:00	31.600000	3.000000
2013-01-02 05:00:00	30.900000	15.000000
2013-01-02 06:00:00	25.900000	18.000000
2013-01-02 07:00:00	14.000000	13.000000
2013-01-02 08:00:00	21.000000	18.000000
2013-01-02 09:00:00	26.500000	18.000000
2013-01-02 10:00:00	16.000000	17.000000
2013-01-02 11:00:00	16.700000	11.000000
2013-01-02 12:00:00	9.000000	10.000000
2013-01-02 13:00:00	11.000000	17.000000
2013-01-02 14:00:00	12.000000	10.000000
2013-01-02 15:00:00	12.000000	9.000000
2013-01-02 16:00:00	13.000000	10.000000
2013-01-02 17:00:00	18.700000	15.000000
2013-01-02 18:00:00	17.700000	14.000000
2013-01-02 19:00:00	12.000000	13.000000
2013-01-02 20:00:00	12.000000	19.000000
2013-01-02 21:00:00	15.700000	13.000000
2013-01-02 22:00:00	18.900000	16.000000
2013-01-02 23:00:00	13.200000	15.000000
2013-01-03 00:00:00	13.000000	18.000000
2013-01-03 01:00:00	13.000000	13.000000
2013-01-03 02:00:00	20.000000	14.000000
2013-01-03 03:00:00	14.000000	14.000000
2013-01-03 04:00:00	14.000000	16.000000
2013-01-03 05:00:00	13.000000	24.000000

```

1 #UK data cleaning
2 UK_df = pd.read_csv(r"C:\Users\User\Code\databook\data\UK\AllData\UKF
3
4 UK_PM_df = UK_df[['DateTime', 'PM2.5', 'PM10']]
5 UK_pp_df = preprocess_dataframe(UK_PM_df)
6 UK_or_df = apply_outliers_removal(UK_pp_df, 'DateTime')
7 UK_or_df2 = UK_or_df.copy(deep=True) #Creating a completely seperate
8 period = 167 #Using weekly period data (24 hours * 7 days)-1 1 hour
9 #UK_fill_df = STL_fill_in_data(UK_or_df2, period)
10
11

```

11

12 #print(UK_fill_df)

```
C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

```
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.loc.html
df['DateTime'] = pd.to_datetime(df['DateTime'])
The length of the PM2.5 column is 43337
Data being checked: PM2.5
Empty DataFrame
Columns: [PM2.5, stl_anomaly]
Index: []
The length of the PM10 column is 43336
Data being checked: PM10
Empty DataFrame
Columns: [PM10, stl_anomaly]
Index: []
```

```
1 UK_or_df3 = UK_or_df.copy(deep=True) #Creating a completely seperate
2
3 UK_or_df3['DateTime'] = UK_or_df3.index
4
5 UK_or_df3.drop('residuals', axis=1, inplace=True)
6
7
8
9 print(UK_or_df3)
```

DateTime	PM2.5	PM10	DateTime
2019-01-01 01:00:00	14.741	19.328	2019-01-01 01:00:00
2019-01-01 02:00:00	11.408	16.705	2019-01-01 02:00:00
2019-01-01 03:00:00	5.407	10.074	2019-01-01 03:00:00
2019-01-01 04:00:00	5.503	10.326	2019-01-01 04:00:00
2019-01-01 05:00:00	5.755	9.708	2019-01-01 05:00:00
2019-01-01 06:00:00	5.527	9.518	2019-01-01 06:00:00
2019-01-01 07:00:00	5.667	9.802	2019-01-01 07:00:00
2019-01-01 08:00:00	5.512	8.935	2019-01-01 08:00:00
2019-01-01 09:00:00	5.576	9.508	2019-01-01 09:00:00
2019-01-01 10:00:00	5.780	9.294	2019-01-01 10:00:00
2019-01-01 11:00:00	5.616	9.060	2019-01-01 11:00:00
2019-01-01 12:00:00	4.002	7.053	2019-01-01 12:00:00
2019-01-01 13:00:00	3.767	5.629	2019-01-01 13:00:00
2019-01-01 14:00:00	3.788	6.083	2019-01-01 14:00:00
2019-01-01 15:00:00	4.141	6.405	2019-01-01 15:00:00
2019-01-01 16:00:00	4.679	7.407	2019-01-01 16:00:00
2019-01-01 17:00:00	4.531	6.830	2019-01-01 17:00:00
2019-01-01 18:00:00	6.571	11.988	2019-01-01 18:00:00
2019-01-01 19:00:00	8.677	17.038	2019-01-01 19:00:00
2019-01-01 20:00:00	8.935	16.323	2019-01-01 20:00:00
2019-01-01 21:00:00	6.283	11.761	2019-01-01 21:00:00
2019-01-01 22:00:00	5.826	10.900	2019-01-01 22:00:00
2019-01-01 23:00:00	5.973	11.615	2019-01-01 23:00:00
2019-01-02 00:00:00	5.973	11.855	2019-01-02 00:00:00
2019-01-02 01:00:00	10.267	11.835	2019-01-02 01:00:00
2019-01-02 02:00:00	8.309	9.584	2019-01-02 02:00:00

2019-01-02 03:00:00	6.833	8.250	2019-01-02 03:00:00
2019-01-02 04:00:00	4.404	5.042	2019-01-02 04:00:00
2019-01-02 05:00:00	4.539	5.730	2019-01-02 05:00:00
2019-01-02 06:00:00	5.783	7.555	2019-01-02 06:00:00
2019-01-02 07:00:00	6.381	8.306	2019-01-02 07:00:00
2019-01-02 08:00:00	7.630	9.329	2019-01-02 08:00:00
2019-01-02 09:00:00	8.259	11.255	2019-01-02 09:00:00
2019-01-02 10:00:00	11.094	16.608	2019-01-02 10:00:00
2019-01-02 11:00:00	10.472	12.740	2019-01-02 11:00:00
2019-01-02 12:00:00	10.675	13.998	2019-01-02 12:00:00
2019-01-02 13:00:00	12.769	16.115	2019-01-02 13:00:00
2019-01-02 14:00:00	8.520	10.795	2019-01-02 14:00:00
2019-01-02 15:00:00	8.267	11.287	2019-01-02 15:00:00
2019-01-02 16:00:00	6.920	9.032	2019-01-02 16:00:00
2019-01-02 17:00:00	7.923	10.682	2019-01-02 17:00:00
2019-01-02 18:00:00	11.163	13.740	2019-01-02 18:00:00
2019-01-02 19:00:00	10.962	14.248	2019-01-02 19:00:00
2019-01-02 20:00:00	9.688	13.743	2019-01-02 20:00:00
2019-01-02 21:00:00	9.557	13.740	2019-01-02 21:00:00
2019-01-02 22:00:00	10.375	13.145	2019-01-02 22:00:00
2019-01-02 23:00:00	10.576	13.163	2019-01-02 23:00:00
2019-01-03 00:00:00	10.300	12.880	2019-01-03 00:00:00
2019-01-03 01:00:00	15.151	18.780	2019-01-03 01:00:00
2019-01-03 02:00:00	15.061	19.070	2019-01-03 02:00:00
2019-01-03 03:00:00	11.498	14.048	2019-01-03 03:00:00
2019-01-03 04:00:00	12.234	14.243	2019-01-03 04:00:00
2019-01-03 05:00:00	18.182	20.488	2019-01-03 05:00:00
2019-01-03 06:00:00	20.382	22.808	2019-01-03 06:00:00
2019-01-03 07:00:00	18.797	22.340	2019-01-03 07:00:00
2019-01-03 08:00:00	21.356	25.893	2019-01-03 08:00:00

```

1 UK_K_fill = KNN_fill_in_data(UK_or_df3, 26)
2 UK_K_fill.drop('DateTime', axis=1, inplace=True)
3 print(UK_K_fill)

```

```
C:\Users\User\AppData\Local\Temp\ipykernel_8864\3676177356.py:24: FutureWarning:
df_imputed = df_imputed.agg(applymap(lambda x: max(x, 0)))
PM2.5           PM10
```

DateTime	PM2.5	PM10
2019-01-01 01:00:00	14.741000	19.328000
2019-01-01 02:00:00	11.408000	16.705000
2019-01-01 03:00:00	5.407000	10.074000
2019-01-01 04:00:00	5.503000	10.326000
2019-01-01 05:00:00	5.755000	9.708000
2019-01-01 06:00:00	5.527000	9.518000
2019-01-01 07:00:00	5.667000	9.802000
2019-01-01 08:00:00	5.512000	8.935000
2019-01-01 09:00:00	5.576000	9.508000
2019-01-01 10:00:00	5.780000	9.294000
2019-01-01 11:00:00	5.616000	9.060000
2019-01-01 12:00:00	4.002000	7.053000
2019-01-01 13:00:00	3.767000	5.629000
2019-01-01 14:00:00	3.788000	6.083000
2019-01-01 15:00:00	4.141000	6.405000
2019-01-01 16:00:00	4.679000	7.407000
2019-01-01 17:00:00	4.531000	6.830000
2019-01-01 18:00:00	6.571000	11.988000
2019-01-01 19:00:00	8.677000	17.038000
2019-01-01 20:00:00	8.025000	16.222000

2019-01-01 21:00:00	6.283000	11.761000
2019-01-01 22:00:00	5.826000	10.900000
2019-01-01 23:00:00	5.973000	11.615000
2019-01-02 00:00:00	5.973000	11.855000
2019-01-02 01:00:00	10.267000	11.835000
2019-01-02 02:00:00	8.309000	9.584000
2019-01-02 03:00:00	6.833000	8.250000
2019-01-02 04:00:00	4.404000	5.042000
2019-01-02 05:00:00	4.539000	5.730000
2019-01-02 06:00:00	5.783000	7.555000
2019-01-02 07:00:00	6.381000	8.306000
2019-01-02 08:00:00	7.630000	9.329000
2019-01-02 09:00:00	8.259000	11.255000
2019-01-02 10:00:00	11.094000	16.608000
2019-01-02 11:00:00	10.472000	12.740000
2019-01-02 12:00:00	10.675000	13.998000
2019-01-02 13:00:00	12.769000	16.115000
2019-01-02 14:00:00	8.520000	10.795000
2019-01-02 15:00:00	8.267000	11.287000
2019-01-02 16:00:00	6.920000	9.032000
2019-01-02 17:00:00	7.923000	10.682000
2019-01-02 18:00:00	11.163000	13.740000
2019-01-02 19:00:00	10.962000	14.248000
2019-01-02 20:00:00	9.688000	13.743000
2019-01-02 21:00:00	9.557000	13.740000
2019-01-02 22:00:00	10.375000	13.145000
2019-01-02 23:00:00	10.576000	13.163000
2019-01-03 00:00:00	10.300000	12.880000
2019-01-03 01:00:00	15.151000	18.780000
2019-01-03 02:00:00	15.061000	19.070000
2019-01-03 03:00:00	11.498000	14.048000
2019-01-03 04:00:00	12.234000	14.243000
2019-01-03 05:00:00	18.182000	20.488000
2019-01-03 06:00:00	20.382000	22.808000

```

1 #Columbia data cleaning
2 Columbia_df = pd.read_csv(r"C:\Users\User\Code\databook\data\Columbia
3
4 Columbia_PM_df = Columbia_df[['DateTime','PM2.5','PM10']]
5 Columbia_pp_df = preprocess_dataframe(Columbia_PM_df)
6 Columbia_or_df = apply_outliers_removal(Columbia_pp_df, 'DateTime')
7
8 Columbia_or_df3 = Columbia_or_df.copy(deep=True) #Creating a complete
9
10 Columbia_or_df3['DateTime'] = Columbia_or_df3.index
11
12 Columbia_or_df3.drop('residuals', axis=1, inplace=True)
13
14
15

```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.loc.html>

```
The length of the PM2.5 column is 46076
Data being checked: PM2.5
Empty DataFrame
Columns: [PM2.5, stl_anomaly]
Index: []
The length of the PM10 column is 47059
Data being checked: PM10
Empty DataFrame
Columns: [PM10, stl_anomaly]
Index: []
```

```
1 Columbia_K_fill = KNN_fill_in_data(Columbia_or_df3, 3)
2 Columbia_K_fill.drop('DateTime', axis=1, inplace=True)
3 print(Columbia_K_fill)
```

```
C:\Users\User\AppData\Local\Temp\ipykernel_8864\1136749358.py:24: FutureWarning:
df_imputed = df_imputed.agg(applymap(lambda x: max(x, 0)))
                                         ^~~~~~
PM2.5          PM10
DateTime
2016-01-01 01:00:00    28.800000   33.000000
2016-01-01 02:00:00    20.260024   16.000000
2016-01-01 03:00:00    16.000000   16.000000
2016-01-01 04:00:00    11.700000   16.000000
2016-01-01 05:00:00    9.100000   17.000000
2016-01-01 06:00:00    8.900000   10.000000
2016-01-01 07:00:00    7.000000   22.000000
2016-01-01 08:00:00    5.800000   12.000000
2016-01-01 09:00:00    3.500000   26.000000
2016-01-01 10:00:00    3.400000   40.000000
2016-01-01 11:00:00    6.400000   73.000000
2016-01-01 12:00:00    8.200000   63.000000
2016-01-01 13:00:00    8.300000   42.000000
2016-01-01 14:00:00    7.800000   32.000000
2016-01-01 15:00:00    6.600000   52.000000
2016-01-01 16:00:00    10.000000  34.000000
2016-01-01 17:00:00    10.900000  31.000000
2016-01-01 18:00:00    10.900000  38.000000
2016-01-01 19:00:00    13.700000  24.000000
2016-01-01 20:00:00    17.200000  34.000000
2016-01-01 21:00:00    18.300000  36.000000
2016-01-01 22:00:00    18.600000  45.000000
2016-01-01 23:00:00    16.700000  26.000000
2016-01-02 00:00:00    12.100000  16.000000
2016-01-02 01:00:00    9.100000   22.000000
2016-01-02 02:00:00    7.900000   25.000000
2016-01-02 03:00:00    7.500000   27.000000
2016-01-02 04:00:00    8.700000   35.000000
2016-01-02 05:00:00    12.400000  38.000000
2016-01-02 06:00:00    16.700000  34.000000
2016-01-02 07:00:00    18.400000  39.000000
2016-01-02 08:00:00    17.800000  35.000000
2016-01-02 09:00:00    13.100000  30.000000
2016-01-02 10:00:00    13.200000  33.000000
2016-01-02 11:00:00    14.300000  33.000000
2016-01-02 12:00:00    15.600000  47.000000
2016-01-02 13:00:00    13.800000  43.000000
2016-01-02 14:00:00    12.500000  52.000000
2016-01-02 15:00:00    13.200000  34.000000
```

2016-01-02 16:00:00	12.500000	31.000000
2016-01-02 17:00:00	13.200000	23.000000
2016-01-02 18:00:00	15.200000	33.000000
2016-01-02 19:00:00	17.800000	30.000000
2016-01-02 20:00:00	17.800000	35.000000
2016-01-02 21:00:00	17.000000	46.000000
2016-01-02 22:00:00	17.900000	42.000000
2016-01-02 23:00:00	18.000000	39.000000
2016-01-03 00:00:00	17.700000	29.000000
2016-01-03 01:00:00	13.400000	18.000000
2016-01-03 02:00:00	9.600000	40.000000
2016-01-03 03:00:00	7.800000	19.000000
2016-01-03 04:00:00	9.000000	38.000000
2016-01-03 05:00:00	11.000000	61.000000
2016-01-03 06:00:00	13.500000	66.000000

```

1 def check_data_intervals(df, time_column):
2     df[time_column] = pd.to_datetime(df[time_column])
3     #df = df.sort_values(by=time_column)
4     df['diff'] = df[time_column].diff().dropna()
5     print(df['diff'].describe())
6     unusual_intervals = df[df['diff'] != pd.Timedelta(hours=1)]
7     print("Unusual intervals:", unusual_intervals)
8
9

```

```

1 Australia_df = pd.read_csv(r"C:\Users\User\Code\databook\data\Australia.csv")
2
3 Australia_PM_df = Australia_df[['DateTime','PM2.5','PM10']]
4 #Australia_pp_df = preprocess_dataframe(Australia_PM_df)
5 check_data_intervals(Australia_PM_df, 'DateTime')

```

	count	35063		
	mean	0 days 01:00:00		
	std	10 days 14:54:48.344591630		
	min	-324 days +01:00:00		
	25%	0 days 01:00:00		
	50%	0 days 01:00:00		
	75%	0 days 01:00:00		
	max	30 days 01:00:00		
	Name: diff, dtype: object			
	Unusual intervals:			
0	2020-01-01 01:00:00	15.0	43.5	NaT
24	2020-02-01 01:00:00	21.9	37.4	30 days 01:00:00
48	2020-03-01 01:00:00	NaN	NaN	28 days 01:00:00
72	2020-04-01 01:00:00	NaN	NaN	30 days 01:00:00
96	2020-05-01 01:00:00	NaN	NaN	29 days 01:00:00
120	2020-06-01 01:00:00	NaN	NaN	30 days 01:00:00
144	2020-07-01 01:00:00	NaN	NaN	29 days 01:00:00
168	2020-08-01 01:00:00	23.1	40.1	30 days 01:00:00
192	2020-09-01 01:00:00	67.9	77.0	30 days 01:00:00
216	2020-10-01 01:00:00	-0.3	6.6	29 days 01:00:00
240	2020-11-01 01:00:00	43.4	82.2	30 days 01:00:00
264	2020-12-01 01:00:00	9.2	28.0	29 days 01:00:00
288	2020-01-13 01:00:00	38.6	51.4	-324 days +01:00:00
744	2020-01-02 01:00:00	NaN	NaN	-30 days +01:00:00
768	2020-02-02 01:00:00	8.6	37.7	30 days 01:00:00

792	2020-03-02	01:00:00	3.7	20.7	28 days	01:00:00
816	2020-04-02	01:00:00	NaN	NaN	30 days	01:00:00
840	2020-05-02	01:00:00	NaN	NaN	29 days	01:00:00
864	2020-06-02	01:00:00	10.9	14.6	30 days	01:00:00
888	2020-07-02	01:00:00	1.7	14.3	29 days	01:00:00
912	2020-08-02	01:00:00	-4.3	3.8	30 days	01:00:00
936	2020-09-02	01:00:00	-3.9	14.9	30 days	01:00:00
960	2020-10-02	01:00:00	-0.7	12.2	29 days	01:00:00
984	2020-11-02	01:00:00	1.9	8.5	30 days	01:00:00
1008	2020-12-02	01:00:00	1.4	7.9	29 days	01:00:00
1032	2020-02-13	01:00:00	NaN	NaN	-294 days	+01:00:00
1440	2020-01-03	01:00:00	NaN	NaN	-58 days	+01:00:00
1464	2020-02-03	01:00:00	NaN	NaN	30 days	01:00:00
1488	2020-03-03	01:00:00	NaN	NaN	28 days	01:00:00
1512	2020-04-03	01:00:00	4.8	10.8	30 days	01:00:00
1536	2020-05-03	01:00:00	3.8	17.4	29 days	01:00:00
1560	2020-06-03	01:00:00	4.1	8.9	30 days	01:00:00
1584	2020-07-03	01:00:00	2.9	14.4	29 days	01:00:00
1608	2020-08-03	01:00:00	5.2	11.7	30 days	01:00:00
1632	2020-09-03	01:00:00	2.2	5.4	30 days	01:00:00
1656	2020-10-03	01:00:00	2.3	7.4	29 days	01:00:00
1680	2020-11-03	01:00:00	3.9	13.4	30 days	01:00:00
1704	2020-12-03	01:00:00	-1.4	4.3	29 days	01:00:00
1728	2020-03-13	01:00:00	5.3	12.5	-266 days	+01:00:00
2184	2020-01-04	01:00:00	5.8	14.8	-88 days	+01:00:00
2208	2020-02-04	01:00:00	3.1	9.2	30 days	01:00:00
2232	2020-03-04	01:00:00	4.2	6.8	28 days	01:00:00
2256	2020-04-04	01:00:00	6.3	9.8	30 days	01:00:00
2280	2020-05-04	01:00:00	-2.0	-0.1	29 days	01:00:00
2304	2020-06-04	01:00:00	0.4	3.0	30 days	01:00:00
2328	2020-07-04	01:00:00	1.9	5.6	29 days	01:00:00
2352	2020-08-04	01:00:00	-3.3	-2.6	30 days	01:00:00
2376	2020-09-04	01:00:00	-2.0	4.2	30 days	01:00:00

```

1 Australia_df = pd.read_csv(r"C:\Users\User\Code\databook\data\Australia.csv")
2
3 Australia_PM_df = Australia_df[['DateTime','PM2.5','PM10']]
4 Australia_PM_df['DateTime'] = pd.to_datetime(Australia_PM_df['DateTime'])
5
6 # Check for NaT values
7 nat_present = Australia_PM_df['DateTime'].isna()
8
9 if nat_present.any():
10     print("NaT values found at index positions:")
11     print(nat_present[nat_present].index.tolist())
12 else:
13     print("No NaT values in the DateTime column.")

```

No NaT values in the DateTime column.

C:\Users\User\AppData\Local\Temp\ipykernel_8864\2916009012.py:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stable/indexing.html#inplace>
 Australia_PM_df['DateTime'] = pd.to_datetime(Australia_PM_df['DateTime'])

```
1 #Australia data cleaning
```

```
1 #Australia Data Cleaning
2 Australia_df = pd.read_csv(r"C:\Users\User\Code\databook\data\Australia.csv")
3
4 Australia_PM_df = Australia_df[['DateTime', 'PM2.5', 'PM10']]
5 Australia_pp_df = preprocess_dataframe(Australia_PM_df)
6 Australia_or_df = apply_outliers_removal(Australia_pp_df, 'DateTime')
7
8 Australia_or_df3 = Australia_or_df.copy(deep=True) #Creating a complete copy
9
10 Australia_or_df3['DateTime'] = Australia_or_df3.index
11
12 Australia_or_df3.drop('residuals', axis=1, inplace=True)
13
14
15
```

```
C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead
```

```
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy
df['DateTime'] = pd.to_datetime(df['DateTime'])
The length of the PM2.5 column is 28745
Data being checked: PM2.5
Empty DataFrame
Columns: [PM2.5, stl_anomaly]
Index: []
The length of the PM10 column is 30885
Data being checked: PM10
Empty DataFrame
Columns: [PM10, stl_anomaly]
Index: []
```

```
1 Australia_K_fill = KNN_fill_in_data(Australia_or_df3, 7)
2 Australia_K_fill.drop('DateTime', axis=1, inplace=True)
3 print(Australia_K_fill)
```

```
C:\Users\User\AppData\Local\Temp\ipykernel_8864\1136749358.py:24: FutureWarning: df_imputed = df_imputed.agg(lambda x: max(x, 0))
PM2.5          PM10
```

DateTime	PM2.5	PM10
2020-01-01 01:00:00	15.000000	43.500000
2020-01-01 02:00:00	6.400000	37.300000
2020-01-01 03:00:00	8.700000	37.000000
2020-01-01 04:00:00	8.800000	30.800000
2020-01-01 05:00:00	12.900000	35.700000
2020-01-01 06:00:00	14.500000	37.900000
2020-01-01 07:00:00	22.700000	45.700000
2020-01-01 08:00:00	22.500000	45.500000
2020-01-01 09:00:00	28.100000	56.300000
2020-01-01 10:00:00	8.400000	32.200000
2020-01-01 11:00:00	51.300000	82.700000
2020-01-01 12:00:00	86.800000	121.100000
2020-01-01 13:00:00	60.000000	79.200000
2020-01-01 14:00:00	74.500000	100.700000
2020-01-01 15:00:00	3.700000	22.500000
2020-01-01 16:00:00	32.700000	61.900000
...

2020-01-01 17:00:00	31.200000	55.500000
2020-01-01 18:00:00	36.100000	64.400000
2020-01-01 19:00:00	30.300000	65.900000
2020-01-01 20:00:00	23.563836	21.100000
2020-01-01 21:00:00	15.500000	39.600000
2020-01-01 22:00:00	10.000000	25.100000
2020-01-01 23:00:00	14.600000	33.700000
2020-01-02 00:00:00	21.000000	35.500000
2020-01-02 01:00:00	19.705430	36.646835
2020-01-02 02:00:00	20.017068	37.348439
2020-01-02 03:00:00	20.334125	37.838131
2020-01-02 04:00:00	20.596398	38.201776
2020-01-02 05:00:00	20.809777	38.483371
2020-01-02 06:00:00	17.088231	35.933152
2020-01-02 07:00:00	14.788878	32.282502
2020-01-02 08:00:00	14.654867	33.709768
2020-01-02 09:00:00	13.903879	32.423069
2020-01-02 10:00:00	13.862241	33.041111
2020-01-02 11:00:00	13.239883	32.953790
2020-01-02 12:00:00	11.919341	32.847382
2020-01-02 13:00:00	12.018798	32.919559
2020-01-02 14:00:00	12.165153	33.034577
2020-01-02 15:00:00	12.406932	33.244631
2020-01-02 16:00:00	12.912204	33.748106
2020-01-02 17:00:00	15.500000	37.000000
2020-01-02 18:00:00	13.600000	32.000000
2020-01-02 19:00:00	10.400000	32.700000
2020-01-02 20:00:00	10.400000	28.500000
2020-01-02 21:00:00	11.100000	30.300000
2020-01-02 22:00:00	10.500000	33.200000
2020-01-02 23:00:00	8.000000	34.200000
2020-01-03 00:00:00	8.700000	39.000000
2020-01-03 01:00:00	9.538199	34.889350
2020-01-03 02:00:00	9.768399	34.159670
2020-01-03 03:00:00	9.892505	33.826606
2020-01-03 04:00:00	9.972546	33.631402
2020-01-03 05:00:00	10.029163	33.502018
2020-01-03 06:00:00	10.071602	33.409601

```

1 #China data cleaning
2 China_df = pd.read_csv(r"C:\Users\User\Code\databook\data\China\China
3
4 China_PM_df = China_df[['DateTime','PM2.5','PM10']]
5 China_pp_df = preprocess_dataframe(China_PM_df)
6 China_or_df = apply_outliers_removal(China_pp_df, 'DateTime')
7
8 China_or_df3 = China_or_df.copy(deep=True) #Creating a completely sep
9
10 China_or_df3['DateTime'] = China_or_df3.index
11
12 China_or_df3.drop('residuals', axis=1, inplace=True)

```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <https://pandas.pydata.org/pandas-doc>
df['DateTime'] = pd.to_datetime(df['DateTime'])

```
The length of the PM2.5 column is 35040
Data being checked: PM2.5
          PM2.5  stl_anomaly
DateTime
2017-01-02 14:00:00    79.0      True
2017-01-02 15:00:00    73.0      True
2017-01-02 16:00:00    65.0      True
2017-01-03 17:00:00    59.0      True
2017-01-27 10:00:00   143.0      True
2017-02-06 07:00:00   164.0      True
2017-02-09 22:00:00   132.0      True
2017-02-09 23:00:00   131.0      True
2017-02-17 06:00:00   149.0      True
2017-02-17 07:00:00   152.0      True
2017-02-17 08:00:00   131.0      True
2017-02-17 09:00:00   130.0      True
2017-02-17 10:00:00   123.0      True
2017-03-10 22:00:00   157.0      True
2017-03-10 23:00:00   170.0      True
2017-03-11 00:00:00   160.0      True
2017-03-18 15:00:00   106.0      True
2017-03-21 21:00:00   118.0      True
2017-03-21 22:00:00   117.0      True
2017-03-21 23:00:00   126.0      True
2017-04-08 18:00:00   143.0      True
2017-10-29 09:00:00   129.0      True
2017-10-29 10:00:00   114.0      True
2017-10-29 11:00:00   118.0      True
2017-11-03 13:00:00   126.0      True
2017-11-03 14:00:00   151.0      True
2017-11-03 15:00:00   157.0      True
2017-11-03 16:00:00   128.0      True
2017-11-26 19:00:00   135.0      True
2017-11-26 20:00:00   128.0      True
2017-11-29 03:00:00   130.0      True
2017-11-29 04:00:00   119.0      True
2017-11-29 05:00:00   150.0      True
2017-11-29 06:00:00   109.0      True
2017-12-04 08:00:00   161.0      True
2017-12-04 09:00:00   166.0      True
2017-12-04 10:00:00   171.0      True
2017-12-04 11:00:00   174.0      True
2017-12-04 12:00:00   145.0      True
2017-12-04 13:00:00   136.0      True
2017-12-04 14:00:00   123.0      True
2017-12-04 16:00:00   137.0      True
2017-12-07 23:00:00   133.0      True
2017-12-08 00:00:00   139.0      True
2017-12-08 01:00:00   147.0      True
2017-12-08 02:00:00   136.0      True
2017-12-08 03:00:00   128.0      True
2017-12-16 11:00:00   160.0      True
```

```
1 China_K_fill = KNN_fill_in_data(China_or_df3, 26)
2 China_K_fill.drop('DateTime', axis=1, inplace=True)
3 print(China_K_fill)
```

```
C:\Users\User\AppData\Local\Temp\ipykernel_8864\3676177356.py:24: FutureWarning
df_imputed = df_imputed.annoyman(lambda x: max(x, 0))
```

		PM2.5	PM10
DateTime			
2017-01-02	00:00:00	164.000000	282.000000
2017-01-02	01:00:00	164.000000	282.000000
2017-01-02	02:00:00	164.000000	282.000000
2017-01-02	03:00:00	164.000000	282.000000
2017-01-02	04:00:00	160.000000	279.000000
2017-01-02	05:00:00	156.000000	276.000000
2017-01-02	06:00:00	156.000000	263.000000
2017-01-02	07:00:00	136.000000	232.000000
2017-01-02	08:00:00	126.000000	196.000000
2017-01-02	09:00:00	124.000000	202.000000
2017-01-02	10:00:00	135.000000	224.000000
2017-01-02	11:00:00	143.000000	245.000000
2017-01-02	12:00:00	148.000000	237.000000
2017-01-02	13:00:00	90.000000	141.000000
2017-01-02	14:00:00	132.061968	121.000000
2017-01-02	15:00:00	135.267405	123.000000
2017-01-02	16:00:00	133.500895	194.604084
2017-01-02	17:00:00	89.000000	138.000000
2017-01-02	18:00:00	168.000000	252.000000
2017-01-02	19:00:00	163.000000	257.000000
2017-01-02	20:00:00	164.000000	268.000000
2017-01-02	21:00:00	156.000000	266.000000
2017-01-02	22:00:00	155.000000	288.000000
2017-01-02	23:00:00	138.000000	266.000000
2017-01-03	00:00:00	127.000000	213.000000
2017-01-03	01:00:00	127.000000	218.000000
2017-01-03	02:00:00	137.000000	229.000000
2017-01-03	03:00:00	132.000000	218.000000
2017-01-03	04:00:00	139.000000	225.000000
2017-01-03	05:00:00	147.000000	234.000000
2017-01-03	06:00:00	142.000000	237.000000
2017-01-03	07:00:00	141.000000	233.000000
2017-01-03	08:00:00	148.000000	247.000000
2017-01-03	09:00:00	164.000000	259.000000
2017-01-03	10:00:00	168.000000	300.000000
2017-01-03	11:00:00	194.000000	289.000000
2017-01-03	12:00:00	180.000000	285.000000
2017-01-03	13:00:00	107.000000	194.000000
2017-01-03	14:00:00	110.000000	181.000000
2017-01-03	15:00:00	104.000000	179.000000
2017-01-03	16:00:00	84.000000	149.000000
2017-01-03	17:00:00	130.231167	220.700350
2017-01-03	18:00:00	89.000000	164.000000
2017-01-03	19:00:00	112.000000	215.000000
2017-01-03	20:00:00	154.000000	263.000000
2017-01-03	21:00:00	178.000000	292.000000
2017-01-03	22:00:00	204.000000	330.000000
2017-01-03	23:00:00	208.000000	338.000000
2017-01-04	00:00:00	195.000000	322.000000
2017-01-04	01:00:00	194.000000	309.000000
2017-01-04	02:00:00	180.000000	292.000000
2017-01-04	03:00:00	193.000000	294.000000
2017-01-04	04:00:00	190.000000	284.000000
2017-01-04	05:00:00	154.000000	255.000000

```
1 #India data cleaning
```

```
2 India_df = pd.read_csv(r'C:\users\user\code\databook\data\india\india.csv')
3
4 India_PM_df = India_df[['DateTime','PM2.5','PM10']]
5 India_pp_df = preprocess_dataframe(India_PM_df)
6 India_or_df = apply_outliers_removal(India_pp_df, 'DateTime')
7
8 India_or_df3 = India_or_df.copy(deep=True) #Creating a completely sep
9
10 India_or_df3['DateTime'] = India_or_df3.index
11
12 India_or_df3.drop('residuals', axis=1, inplace=True)
```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stable/indexing.html#inplace>
df['DateTime'] = pd.to_datetime(df['DateTime'])
The length of the PM2.5 column is 47817
Data being checked: PM2.5
Empty DataFrame
Columns: [PM2.5, stl_anomaly]
Index: []
The length of the PM10 column is 45771
Data being checked: PM10
Empty DataFrame
Columns: [PM10, stl_anomaly]
Index: []

```
1 India_K_fill = KNN_fill_in_data(India_or_df3, 26)
2 India_K_fill.drop('DateTime', axis=1, inplace=True)
3 print(India_K_fill)
```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\3676177356.py:24: FutureWarning:
df_imputed = df_imputed.applymap(lambda x: max(x, 0))

DateTime	PM2.5	PM10
2015-01-01 01:00:00	454.580000	935.180000
2015-01-01 02:00:00	440.440000	753.677272
2015-01-01 03:00:00	409.090000	731.805253
2015-01-01 04:00:00	436.120000	751.168685
2015-01-01 05:00:00	415.880000	976.990000
2015-01-01 06:00:00	384.160000	862.230000
2015-01-01 07:00:00	344.440000	731.830000
2015-01-01 08:00:00	337.980000	725.740000
2015-01-01 09:00:00	285.220000	656.930000
2015-01-01 10:00:00	258.420000	525.540000
2015-01-01 11:00:00	265.620000	548.820000
2015-01-01 12:00:00	260.090000	522.930000
2015-01-01 13:00:00	222.010000	408.620000
2015-01-01 14:00:00	202.140000	340.280000
2015-01-01 15:00:00	244.140000	459.010000
2015-01-01 16:00:00	211.510000	340.660000
2015-01-01 17:00:00	191.120000	257.400000
2015-01-01 18:00:00	218.440000	351.010000
2015-01-01 19:00:00	296.800000	600.950000
2015-01-01 20:00:00	336.430000	714.630000
2015-01-01 21:00:00	326.510000	601.810000

2015-01-01 21:00:00	320.040000	631.040000
2015-01-01 22:00:00	327.680000	732.590000
2015-01-01 23:00:00	335.180000	776.360000
2015-01-02 00:00:00	358.680000	860.320000
2015-01-02 01:00:00	444.440000	806.070000
2015-01-02 02:00:00	464.640000	856.680000
2015-01-02 03:00:00	344.050000	765.000000
2015-01-02 04:00:00	358.070000	797.090000
2015-01-02 05:00:00	271.410000	516.140000
2015-01-02 06:00:00	260.950000	467.810000
2015-01-02 07:00:00	192.950000	301.480000
2015-01-02 08:00:00	73.860000	115.400000
2015-01-02 09:00:00	52.540000	62.600000
2015-01-02 10:00:00	70.400000	41.870000
2015-01-02 11:00:00	96.840000	65.110000
2015-01-02 12:00:00	93.350000	36.250000
2015-01-02 13:00:00	208.770000	28.970000
2015-01-02 14:00:00	239.860000	52.680000
2015-01-02 15:00:00	187.960000	49.890000
2015-01-02 16:00:00	185.200000	64.700000
2015-01-02 17:00:00	157.230000	60.090000
2015-01-02 18:00:00	137.380000	99.630000
2015-01-02 19:00:00	50.680000	79.180000
2015-01-02 20:00:00	59.150000	92.420000
2015-01-02 21:00:00	54.250000	84.770000
2015-01-02 22:00:00	55.810000	87.200000
2015-01-02 23:00:00	49.810000	77.820000
2015-01-03 00:00:00	51.210000	80.010000
2015-01-03 01:00:00	50.690000	79.210000
2015-01-03 02:00:00	41.060000	64.160000
2015-01-03 03:00:00	45.080000	60.410000
2015-01-03 04:00:00	44.780000	58.670000
2015-01-03 05:00:00	51.440000	77.920000
2015-01-03 06:00:00	65.000000	121.090000

```

1 #Africa data cleaning
2 Africa_df = pd.read_csv(r"C:\Users\User\Code\databook\data\Africa\Pro
3
4 Africa_PM_df = Africa_df[['DateTime', 'PM2.5', 'PM10']]
5 Africa_pp_df = preprocess_dataframe(Africa_PM_df)
6 Africa_or_df = apply_outliers_removal(Africa_pp_df, 'DateTime')
7
8 Africa_or_df3 = Africa_or_df.copy(deep=True) #Creating a completely s
9
10 Africa_or_df3['DateTime'] = Africa_or_df3.index
11
12 Africa_or_df3.drop('residuals', axis=1, inplace=True)

```

The length of the PM2.5 column is 9408

Data being checked: PM2.5

Empty DataFrame

Columns: [PM2.5, stl_anomaly]

Index: []

The length of the PM10 column is 9408

Data being checked: PM10

Empty DataFrame

Columns: [PM10, stl_anomaly]

Index: []

```
1 Africa_K_fill = KNN_fill_in_data(Africa_or_df3, 26)
2 Africa_K_fill.drop('DateTime', axis=1, inplace=True)
3 print(Africa_K_fill)
```

	PM2.5	PM10
DateTime		
2019-11-21 08:00:00	27.210000	35.410000
2019-11-21 09:00:00	15.660000	22.920000
2019-11-21 10:00:00	12.200000	13.110000
2019-11-21 11:00:00	10.930000	11.060000
2019-11-21 12:00:00	9.610000	10.500000
2019-11-21 13:00:00	10.940000	11.190000
2019-11-21 14:00:00	13.800000	17.300000
2019-11-21 15:00:00	17.610000	21.110000
2019-11-21 16:00:00	33.100000	40.210000
2019-11-21 17:00:00	32.150000	40.230000
2019-11-21 18:00:00	37.290000	43.080000
2019-11-21 19:00:00	34.370000	46.300000
2019-11-21 20:00:00	33.970000	43.660000
2019-11-21 21:00:00	30.620000	40.910000
2019-11-21 22:00:00	20.900000	33.650000
2019-11-21 23:00:00	14.520000	17.850000
2019-11-22 00:00:00	11.970000	13.790000
2019-11-22 01:00:00	13.970000	18.930000
2019-11-22 02:00:00	17.240000	24.140000
2019-11-22 03:00:00	15.860000	21.010000
2019-11-22 04:00:00	10.080000	8.540000
2019-11-22 05:00:00	12.020000	13.560000
2019-11-22 06:00:00	12.330000	13.030000
2019-11-22 07:00:00	21.030000	31.580000
2019-11-22 08:00:00	20.360000	30.660000
2019-11-22 09:00:00	21.860000	27.950000
2019-11-22 10:00:00	20.840000	27.870000
2019-11-22 11:00:00	10.940000	9.540000
2019-11-22 12:00:00	12.380000	16.560000
2019-11-22 13:00:00	14.360000	24.350000
2019-11-22 14:00:00	15.100000	23.320000
2019-11-22 15:00:00	22.050000	34.840000
2019-11-22 16:00:00	35.050000	47.910000
2019-11-22 17:00:00	28.800000	41.650000
2019-11-22 18:00:00	40.030000	50.930000
2019-11-22 19:00:00	21.040000	34.170000
2019-11-22 20:00:00	8.310000	4.060000
2019-11-22 21:00:00	8.920000	3.730000
2019-11-22 22:00:00	15.180659	17.087233
2019-11-22 23:00:00	10.810000	4.630000
2019-11-23 00:00:00	14.927399	17.128448
2019-11-23 01:00:00	14.717289	17.345526
2019-11-23 02:00:00	10.520000	5.990000
2019-11-23 03:00:00	15.390000	22.770000
2019-11-23 04:00:00	21.510000	35.260000
2019-11-23 05:00:00	10.980000	12.290000
2019-11-23 06:00:00	20.670000	32.130000
2019-11-23 07:00:00	16.420000	27.390000
2019-11-23 08:00:00	13.890000	19.790000
2019-11-23 09:00:00	10.910000	14.720000
2019-11-23 10:00:00	8.340000	9.900000
2019-11-23 11:00:00	8.600000	9.170000

2019-11-23 11:00:00	8.690000	9.410000
2019-11-23 12:00:00	11.360000	9.180000
2019-11-23 13:00:00	10.630000	6.680000
2019-11-23 14:00:00	10.210000	9.240000
2019-11-23 15:00:00	9.970000	7.230000

```
1 #Mexico data cleaning
2 Mexico_df = pd.read_csv(r"C:\Users\User\Code\databook\data\Mexico\Pro
3
4 Mexico_PM_df = Mexico_df[['DateTime', 'PM2.5', 'PM10']]
5 Mexico_pp_df = preprocess_dataframe(Mexico_PM_df)
6 Mexico_or_df = apply_outliers_removal(Mexico_pp_df, 'DateTime')
7
8 Mexico_or_df3 = Mexico_or_df.copy(deep=True) #Creating a completely s
9
10 Mexico_or_df3['DateTime'] = Mexico_or_df3.index
11
12 Mexico_or_df3.drop('residuals', axis=1, inplace=True)
```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <https://pandas.pydata.org/pandas-doc>
df['DateTime'] = pd.to_datetime(df['DateTime'])

The length of the PM2.5 column is 42980

Data being checked: PM2.5

Empty DataFrame

Columns: [PM2.5, stl_anomaly]

Index: []

The length of the PM10 column is 43273

Data being checked: PM10

Empty DataFrame

Columns: [PM10, stl_anomaly]

Index: []

```
1 Mexico_K_fill = KNN_fill_in_data(Mexico_or_df3, 70)
2 Mexico_K_fill.drop('DateTime', axis=1, inplace=True)
3 print(Mexico_K_fill)
```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\3676177356.py:24: FutureWarning: df_imputed = df_imputed.agg(applymap(lambda x: max(x, 0)))
PM2.5 PM10

DateTime	PM2.5	PM10
2016-07-28 00:00:00	39.226000	59.960000
2016-07-28 01:00:00	59.496000	83.775000
2016-07-28 02:00:00	63.474000	90.966000
2016-07-28 03:00:00	46.661000	65.046000
2016-07-28 04:00:00	46.951000	78.580000
2016-07-28 05:00:00	37.223000	56.641000
2016-07-28 06:00:00	33.868000	54.605000
2016-07-28 07:00:00	35.657000	68.004000
2016-07-28 08:00:00	41.030000	79.380000
2016-07-28 09:00:00	43.069000	104.920000
2016-07-28 10:00:00	33.582000	76.111000
2016-07-28 11:00:00	34.113000	80.444000
2016-07-28 12:00:00	24.415000	52.781000

2016-07-28 13:00:00	19.982000	39.818000
2016-07-28 14:00:00	16.775000	39.510000
2016-07-28 15:00:00	22.981000	40.704000
2016-07-28 16:00:00	14.733000	35.359000
2016-07-28 17:00:00	17.797000	40.007000
2016-07-28 18:00:00	18.980000	43.497000
2016-07-28 19:00:00	25.034000	48.906000
2016-07-28 20:00:00	21.455000	39.172000
2016-07-28 21:00:00	16.398000	23.098000
2016-07-28 22:00:00	12.556000	28.059000
2016-07-28 23:00:00	16.372000	34.942000
2016-07-29 00:00:00	21.663000	35.453000
2016-07-29 01:00:00	13.123000	19.181000
2016-07-29 02:00:00	27.357000	38.772000
2016-07-29 03:00:00	52.205000	82.540000
2016-07-29 04:00:00	44.777000	81.392000
2016-07-29 05:00:00	42.552000	75.401000
2016-07-29 06:00:00	22.865000	33.700000
2016-07-29 07:00:00	27.923000	50.136000
2016-07-29 08:00:00	25.200000	40.539000
2016-07-29 09:00:00	25.766000	36.071000
2016-07-29 10:00:00	26.537000	47.159000
2016-07-29 11:00:00	21.228000	45.771000
2016-07-29 12:00:00	19.071000	37.639000
2016-07-29 13:00:00	8.050900	20.051000
2016-07-29 14:00:00	8.903300	20.880000
2016-07-29 15:00:00	10.428000	17.959000
2016-07-29 16:00:00	17.923000	47.622000
2016-07-29 17:00:00	15.975000	43.998000
2016-07-29 18:00:00	21.151000	42.334000
2016-07-29 19:00:00	17.085000	38.943000
2016-07-29 20:00:00	18.177000	33.995000
2016-07-29 21:00:00	15.981000	26.554000
2016-07-29 22:00:00	12.288000	25.812000
2016-07-29 23:00:00	11.446000	17.442000
2016-07-30 00:00:00	24.034000	35.559000
2016-07-30 01:00:00	12.076000	15.088000
2016-07-30 02:00:00	17.916000	29.970000
2016-07-30 03:00:00	7.823600	18.774000
2016-07-30 04:00:00	16.778000	34.918000
2016-07-30 05:00:00	16.165000	33.615000

```
1 #Korea data cleaning
2 Korea_df = pd.read_csv(r"C:\Users\User\Code\databook\data\Korea\Korea
3
4 Korea_PM_df = Korea_df[['DateTime','PM2.5','PM10']]
5 Korea_pp_df = preprocess_dataframe(Korea_PM_df)
6 Korea_or_df = apply_outliers_removal(Korea_pp_df, 'DateTime')
7
8 Korea_or_df3 = Korea_or_df.copy(deep=True) #Creating a completely sep
9
10 Korea_or_df3['DateTime'] = Korea_or_df3.index
11
12 Korea_or_df3.drop('residuals', axis=1, inplace=True)
```

C:\Users\User\AppData\Local\Temp\ipykernel_8864\110586271.py:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead

```
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.to\_datetime.html
df['DateTime'] = pd.to_datetime(df['DateTime'])
The length of the PM2.5 column is 25877
Data being checked: PM2.5
Empty DataFrame
Columns: [PM2.5, stl_anomaly]
Index: []
The length of the PM10 column is 25882
Data being checked: PM10
Empty DataFrame
Columns: [PM10, stl_anomaly]
Index: []
```

```
1 Korea_K_fill = KNN_fill_in_data(Korea_or_df3, 26)
2 Korea_K_fill.drop('DateTime', axis=1, inplace=True)
3 print(Korea_K_fill)
```

```
C:\Users\User\AppData\Local\Temp\ipykernel_8864\3676177356.py:24: FutureWarning:
df_imputed = df_imputed.agg(lambda x: max(x, 0))
          PM2.5           PM10
DateTime
2017-01-01 00:00:00    53.000000   68.000000
2017-01-01 01:00:00    63.000000   78.000000
2017-01-01 02:00:00    56.000000   72.000000
2017-01-01 03:00:00    46.000000   68.000000
2017-01-01 04:00:00    48.000000   66.000000
2017-01-01 05:00:00    44.000000   60.000000
2017-01-01 06:00:00    38.000000   67.000000
2017-01-01 07:00:00    44.000000   77.000000
2017-01-01 08:00:00    40.000000   73.000000
2017-01-01 09:00:00    42.000000   65.000000
2017-01-01 10:00:00    64.000000   67.000000
2017-01-01 11:00:00    60.000000   70.000000
2017-01-01 12:00:00    57.000000   66.000000
2017-01-01 13:00:00    53.000000   95.000000
2017-01-01 14:00:00    54.000000   88.000000
2017-01-01 15:00:00    46.000000   81.000000
2017-01-01 16:00:00    54.000000   87.000000
2017-01-01 17:00:00    48.000000   86.000000
2017-01-01 18:00:00    77.000000  106.000000
2017-01-01 19:00:00    75.000000   97.000000
2017-01-01 20:00:00    62.000000   91.000000
2017-01-01 21:00:00    57.000000   86.000000
2017-01-01 22:00:00    61.000000   90.000000
2017-01-01 23:00:00    59.000000  106.000000
2017-01-02 00:00:00    66.000000  101.000000
2017-01-02 01:00:00    74.000000  102.000000
2017-01-02 02:00:00    66.000000   95.000000
2017-01-02 03:00:00    62.000000   82.000000
2017-01-02 04:00:00    58.000000   86.000000
2017-01-02 05:00:00    58.000000   85.000000
2017-01-02 06:00:00    51.000000  102.000000
2017-01-02 07:00:00    72.000000   92.000000
2017-01-02 08:00:00    79.000000   93.000000
2017-01-02 09:00:00    56.000000   77.000000
2017-01-02 10:00:00    57.000000   97.000000
2017-01-02 11:00:00    59.000000  107.000000
```

2017-01-02 12:00:00	52.000000	90.000000
2017-01-02 13:00:00	48.000000	90.000000
2017-01-02 14:00:00	81.000000	117.000000
2017-01-02 15:00:00	120.000000	152.000000
2017-01-02 16:00:00	116.000000	159.000000
2017-01-02 17:00:00	121.000000	170.000000
2017-01-02 18:00:00	126.000000	183.000000
2017-01-02 19:00:00	131.000000	169.000000
2017-01-02 20:00:00	126.000000	192.000000
2017-01-02 21:00:00	111.000000	175.000000
2017-01-02 22:00:00	133.000000	159.000000
2017-01-02 23:00:00	111.729741	156.831689
2017-01-03 00:00:00	113.000000	155.000000
2017-01-03 01:00:00	127.000000	178.000000
2017-01-03 02:00:00	121.000000	170.000000
2017-01-03 03:00:00	92.000000	159.000000
2017-01-03 04:00:00	84.000000	114.000000
2017-01-03 05:00:00	87.000000	118.000000

```

1 #Spain data cleaning
2 Spain_df = pd.read_csv(r"C:\Users\User\Code\databook\data\Spain\Spain
3
4 Spain_PM_df = Spain_df[['DateTime','PM2.5','PM10']]
5 Spain_pp_df = preprocess_dataframe(Spain_PM_df)
6 Spain_or_df = apply_outliers_removal(Spain_pp_df, 'DateTime')
7
8 Spain_or_df3 = Spain_or_df.copy(deep=True) #Creating a completely sep
9
10 Spain_or_df3['DateTime'] = Spain_or_df3.index
11
12 Spain_or_df3.drop('residuals', axis=1, inplace=True)

```

The length of the PM2.5 column is 58691

Data being checked: PM2.5

Empty DataFrame

Columns: [PM2.5, stl_anomaly]

Index: []

The length of the PM10 column is 59024

Data being checked: PM10

Empty DataFrame

Columns: [PM10, stl_anomaly]

Index: []

```

1 Spain_K_fill = KNN_fill_in_data(Spain_or_df3, 26)
2 Spain_K_fill.drop('DateTime', axis=1, inplace=True)
3 print(Spain_K_fill)

```

C:\Users\User\AppData\Local\Temp\ipykernel_10772\3676177356.py:24: FutureWarning:

df_imputed = df_imputed.agg(applymap(lambda x: max(x, 0)))

PM2.5 PM10

DateTime

2015-01-01 00:00:00	45.000000	58.000000
2015-01-01 01:00:00	83.000000	99.000000
2015-01-01 02:00:00	72.000000	89.000000
2015-01-01 03:00:00	69.000000	92.000000
2015-01-01 04:00:00	78.000000	99.000000
2015-01-01 05:00:00	59.000000	79.000000

2015-01-01 06:00:00	49.000000	57.000000
2015-01-01 07:00:00	45.000000	52.000000
2015-01-01 08:00:00	38.000000	44.000000
2015-01-01 09:00:00	37.000000	44.000000
2015-01-01 10:00:00	45.000000	52.000000
2015-01-01 11:00:00	18.000000	19.000000
2015-01-01 12:00:00	12.000000	11.000000
2015-01-01 13:00:00	6.000000	4.000000
2015-01-01 14:00:00	16.000000	21.000000
2015-01-01 15:00:00	12.000000	15.000000
2015-01-01 16:00:00	9.000000	13.000000
2015-01-01 17:00:00	10.000000	19.000000
2015-01-01 18:00:00	10.000000	23.000000
2015-01-01 19:00:00	16.000000	25.000000
2015-01-01 20:00:00	19.000000	27.000000
2015-01-01 21:00:00	21.000000	29.000000
2015-01-01 22:00:00	17.000000	23.000000
2015-01-01 23:00:00	15.000000	21.000000
2015-01-02 00:00:00	17.000000	23.000000
2015-01-02 01:00:00	26.000000	42.000000
2015-01-02 02:00:00	17.000000	28.000000
2015-01-02 03:00:00	9.000000	11.000000
2015-01-02 04:00:00	7.000000	10.000000
2015-01-02 05:00:00	8.000000	14.000000
2015-01-02 06:00:00	11.000000	17.000000
2015-01-02 07:00:00	15.000000	21.000000
2015-01-02 08:00:00	18.000000	27.000000
2015-01-02 09:00:00	18.000000	34.000000
2015-01-02 10:00:00	13.000000	15.000000
2015-01-02 11:00:00	17.000000	22.000000
2015-01-02 12:00:00	19.403537	51.000000
2015-01-02 13:00:00	20.000000	33.000000
2015-01-02 14:00:00	17.000000	25.000000
2015-01-02 15:00:00	22.000000	32.000000
2015-01-02 16:00:00	22.000000	36.000000
2015-01-02 17:00:00	23.000000	44.000000
2015-01-02 18:00:00	26.000000	49.000000
2015-01-02 19:00:00	26.000000	48.000000
2015-01-02 20:00:00	31.000000	55.000000
2015-01-02 21:00:00	36.000000	61.000000
2015-01-02 22:00:00	53.000000	83.000000
2015-01-02 23:00:00	56.000000	78.000000
2015-01-03 00:00:00	22.000000	33.000000
2015-01-03 01:00:00	11.000000	17.000000
2015-01-03 02:00:00	7.000000	12.000000
2015-01-03 03:00:00	12.000000	22.000000
2015-01-03 04:00:00	10.000000	21.000000
2015-01-03 05:00:00	9.000000	16.000000

```
1 #Save PM Data of Spain data
2 Spain_K_fill.to_csv('C:/Users/User/Code/databook/data/Spain/SpainClea
```

```
1 Australia_PM10_Original_Mean = Australia_or_df['PM10'].mean()
2 print(f"PM10 Original Mean = {Australia_PM10_Original_Mean}")
3 Australia_PM10_Imp_Mean = Australia_K_fill['PM10'].mean()
4 print(f"PM10 Imputed Mean = {Australia_PM10_Imp_Mean}")
5 print(f"% Change in Mean = {(abs(Australia_PM10_Original_Mean-Austra
```

```
6
7 Austrailia_PM10_Original_STD = Australia_or_df['PM10'].std()
8 print(f"PM10 Original STD = {Austrailia_PM10_Original_STD}")
9 Australia_PM10_Imp_STD = Australia_K_fill['PM10'].std()
10 print(f"PM10 Imputed STD = {Australia_PM10_Imp_STD}")
11 print(f"% Change in STD = {(abs(Austrailia_PM10_Original_STD-Australi
12
13 Austrailia_PM25_Original_Mean = Australia_or_df['PM2.5'].mean()
14 print(f"PM25 Original Mean = {Austrailia_PM25_Original_Mean}")
15 Australia_PM25_Imp_Mean = Australia_K_fill['PM2.5'].mean()
16 print(f"PM2.5 Imputed Mean = {Australia_PM25_Imp_Mean}")
17 print(f"% Change in Mean = {(abs(Austrailia_PM25_Original_Mean-Austra
18
19 Austrailia_PM25_Original_STD = Australia_or_df['PM2.5'].std()
20 print(f"PM2.5 Original STD = {Austrailia_PM25_Original_STD}")
21 Australia_PM25_Imp_STD = Australia_K_fill['PM2.5'].std()
22 print(f"PM2.5 Imputed STD = {Australia_PM25_Imp_STD}")
23 print(f"% Change in STD = {(abs(Austrailia_PM25_Original_STD-Australi
24
```

```
PM10 Original Mean = 13.914885790059028
PM10 Imputed Mean = 15.49911605176426
% Change in Mean = 11.385147428497227
PM10 Original STD = 11.809938034057852
PM10 Imputed STD = 12.019752696767737
% Change in STD = 1.7765941032443617
PM25 Original Mean = 7.3767304315682445
PM2.5 Imputed Mean = 7.835602459208763
% Change in Mean = 6.220534041433931
PM2.5 Original STD = 9.779175130886165
PM2.5 Imputed STD = 9.700285300116004
% Change in STD = 0.8067125265094994
```

```
1 China_PM10_Original_Mean = China_or_df['PM10'].mean()
2 print(f"PM10 Original Mean = {China_PM10_Original_Mean}")
3 China_PM10_Imp_Mean = China_K_fill['PM10'].mean()
4 print(f"PM10 Imputed Mean = {China_PM10_Imp_Mean}")
5 print(f"% Change in Mean = {(abs(China_PM10_Original_Mean-China_PM10_
6 print("")
7 China_PM10_Original_STD = China_or_df['PM10'].std()
8 print(f"PM10 Original STD = {China_PM10_Original_STD}")
9 China_PM10_Imp_STD = China_K_fill['PM10'].std()
10 print(f"PM10 Imputed STD = {China_PM10_Imp_STD}")
11 print(f"% Change in STD = {(abs(China_PM10_Original_STD-China_PM10_Im
12 print("")
13 China_PM25_Original_Mean = China_or_df['PM2.5'].mean()
14 print(f"PM25 Original Mean = {China_PM25_Original_Mean}")
15 China_PM25_Imp_Mean = China_K_fill['PM2.5'].mean()
16 print(f"PM2.5 Imputed Mean = {China_PM25_Imp_Mean}")
17 print(f"% Change in Mean = {(abs(China_PM25_Original_Mean-China_PM25_
18 print("")
19 China_PM25_Original_STD = China_or_df['PM2.5'].std()
20 print(f"PM2.5 Original STD = {China_PM25_Original_STD}")
21 China_PM25_Imp_STD = China_K_fill['PM2.5'].std()
22 print(f"PM2.5 Imputed STD = {China_PM25_Imp_STD}")
```

```
-- 
23 print(f"% Change in STD = {((abs(China_PM25_Original_STD-China_PM25_Im
24

PM10 Original Mean = 76.79392950916481
PM10 Imputed Mean = 77.03470848032912
% Change in Mean = 0.31353906839155576

PM10 Original STD = 49.085042412905956
PM10 Imputed STD = 49.47699527566208
% Change in STD = 0.7985179262126181

PM25 Original Mean = 41.73384303119871
PM2.5 Imputed Mean = 41.92130506278232
% Change in Mean = 0.44918468554038277

PM2.5 Original STD = 30.38425585313831
PM2.5 Imputed STD = 30.73168817805868
% Change in STD = 1.1434616881837634
```

```
1 India_PM10_Original_Mean = India_or_df['PM10'].mean()
2 print(f"PM10 Original Mean = {India_PM10_Original_Mean}")
3 India_PM10_Imp_Mean = India_K_fill['PM10'].mean()
4 print(f"PM10 Imputed Mean = {India_PM10_Imp_Mean}")
5 print(f"% Change in Mean = {((abs(India_PM10_Original_Mean-India_PM10_
6 print(""))
7 India_PM10_Original_STD = India_or_df['PM10'].std()
8 print(f"PM10 Original STD = {India_PM10_Original_STD}")
9 India_PM10_Imp_STD = India_K_fill['PM10'].std()
10 print(f"PM10 Imputed STD = {India_PM10_Imp_STD}")
11 print(f"% Change in STD = {((abs(India_PM10_Original_STD-India_PM10_Im
12 print(""))
13 India_PM25_Original_Mean = India_or_df['PM2.5'].mean()
14 print(f"PM25 Original Mean = {India_PM25_Original_Mean}")
15 India_PM25_Imp_Mean = India_K_fill['PM2.5'].mean()
16 print(f"PM2.5 Imputed Mean = {India_PM25_Imp_Mean}")
17 print(f"% Change in Mean = {((abs(India_PM25_Original_Mean-India_PM25_"
18 print(""))
19 India_PM25_Original_STD = India_or_df['PM2.5'].std()
20 print(f"PM2.5 Original STD = {India_PM25_Original_STD}")
21 India_PM25_Imp_STD = India_K_fill['PM2.5'].std()
22 print(f"PM2.5 Imputed STD = {India_PM25_Imp_STD}")
23 print(f"% Change in STD = {((abs(India_PM25_Original_STD-India_PM25_Im
24
```

```
PM10 Original Mean = 241.20379937027016
PM10 Imputed Mean = 297.56168391489837
% Change in Mean = 23.365255726388305
```

```
PM10 Original STD = 149.41756469247497
PM10 Imputed STD = 134.05496529556518
% Change in STD = 10.281655592853793
```

```
PM25 Original Mean = 123.15807840447422
PM2.5 Imputed Mean = 171.5571606221127
% Change in Mean = 39.29834148490593
```

```
PM2.5 Original STD = 104.2813505783325
PM2.5 Imputed STD = 94.0248789940435
% Change in STD = 9.835384301610766
```

```
1 #print(US_fill_df)
2 #Save All US data
3 US_or_df.to_csv('C:/Users/User/Code/databook/data/US/CombinedData/USA'
4 US_fill_df.to_csv('C:/Users/User/Code/databook/data/US/CombinedData/U
5
```

```
1 #Save PM Data of US data
2 US_PM_df = US_fill_df[['PM2.5', 'PM10']]
3 US_PM_df.to_csv('C:/Users/User/Code/databook/data/US/CombinedData/USC
4 print(US_PM_df)
```

DateTime	PM2.5	PM10
2013-01-01 00:00:00	12.000000	16.000000
2013-01-01 01:00:00	16.000000	18.000000
2013-01-01 02:00:00	9.000000	10.000000
2013-01-01 03:00:00	8.000000	10.000000
2013-01-01 04:00:00	10.000000	7.000000
2013-01-01 05:00:00	13.000000	12.000000
2013-01-01 06:00:00	12.000000	9.000000
2013-01-01 07:00:00	9.900000	6.000000
2013-01-01 08:00:00	6.000000	6.000000
2013-01-01 09:00:00	6.000000	7.000000
2013-01-01 10:00:00	9.800000	9.000000
2013-01-01 11:00:00	7.500000	8.000000
2013-01-01 12:00:00	6.900000	3.000000
2013-01-01 13:00:00	15.800000	3.000000
2013-01-01 14:00:00	13.900000	1.000000
2013-01-01 15:00:00	16.900000	7.000000
2013-01-01 16:00:00	21.600000	5.000000
2013-01-01 17:00:00	26.400000	17.000000
2013-01-01 18:00:00	31.900000	7.000000
2013-01-01 19:00:00	30.900000	16.000000
2013-01-01 20:00:00	30.900000	16.000000
2013-01-01 21:00:00	32.900000	9.000000
2013-01-01 22:00:00	25.100000	13.000000
2013-01-01 23:00:00	22.100000	10.000000
2013-01-02 00:00:00	17.000000	10.000000
2013-01-02 01:00:00	15.000000	7.000000
2013-01-02 02:00:00	25.900000	5.000000
2013-01-02 03:00:00	10.000000	16.000000
2013-01-02 04:00:00	31.600000	3.000000
2013-01-02 05:00:00	30.900000	15.000000
2013-01-02 06:00:00	25.900000	18.000000
2013-01-02 07:00:00	14.000000	13.000000
2013-01-02 08:00:00	21.000000	18.000000
2013-01-02 09:00:00	26.500000	18.000000
2013-01-02 10:00:00	16.000000	17.000000
2013-01-02 11:00:00	16.700000	11.000000
2013-01-02 12:00:00	9.000000	10.000000
2013-01-02 13:00:00	11.000000	17.000000
2013-01-02 14:00:00	12.000000	10.000000
2013-01-02 15:00:00	12.000000	10.000000

2013-01-02 15:00:00	12.000000	9.000000
2013-01-02 16:00:00	13.000000	10.000000
2013-01-02 17:00:00	18.700000	15.000000
2013-01-02 18:00:00	17.700000	14.000000
2013-01-02 19:00:00	12.000000	13.000000
2013-01-02 20:00:00	12.000000	19.000000
2013-01-02 21:00:00	15.700000	13.000000
2013-01-02 22:00:00	18.900000	16.000000
2013-01-02 23:00:00	13.200000	15.000000
2013-01-03 00:00:00	13.000000	18.000000
2013-01-03 01:00:00	13.000000	13.000000
2013-01-03 02:00:00	20.000000	14.000000
2013-01-03 03:00:00	14.000000	14.000000
2013-01-03 04:00:00	14.000000	16.000000
2013-01-03 05:00:00	13.000000	24.000000
2013-01-03 06:00:00	26.400000	25.000000
2013-01-03 07:00:00	32.000000	28.000000

```
1 #Save PM Data of US data  
2 US_K_fill.to_csv('C:/Users/User/Code/databook/data/US/CombinedData/US
```

```
1 #Save PM Data of UK data  
2 UK_K_fill.to_csv('C:/Users/User/Code/databook/data/UK/AllData/UKClean
```

```
1 #Save PM Data of Columbia data  
2 Columbia_K_fill.to_csv('C:/Users/User/Code/databook/data/Columbia/All
```

```
1 #Save PM Data of Australia data  
2 Australia_K_fill.to_csv('C:/Users/User/Code/databook/data/Australia/A
```

```
1 #Save PM Data of China data  
2 China_K_fill.to_csv('C:/Users/User/Code/databook/data/China/ChinaClea
```

```
1 #Save PM Data of India data  
2 India_K_fill.to_csv('C:/Users/User/Code/databook/data/India/IndiaClea
```

```
1 #Save PM Data of Africa data  
2 Africa_K_fill.to_csv('C:/Users/User/Code/databook/data/Africa/AfricaC
```

```
1 #Save PM Data of Mexico data  
2 Mexico_K_fill.to_csv('C:/Users/User/Code/databook/data/Mexico/MexicoC
```

```
1 #Save PM Data of Korea data  
2 Korea_K_fill.to_csv('C:/Users/User/Code/databook/data/Korea/KoreaClea
```

```
1 #Select which data to plot:  
2 old_data = UK_or_df['PM2.5']  
3 new_data = UK_fill_df['PM2.5']  
4  
5 grouped_old = old_data.groupby(old_data.index.year)
```

```
5 grouped_new = new_data.groupby(new_data.index.year)
6 grouped_new = new_data.groupby(new_data.index.year)
7
8 print(old_data)
```

DateTime	
2019-01-01 01:00:00	14.741
2019-01-01 02:00:00	11.408
2019-01-01 03:00:00	5.407
2019-01-01 04:00:00	5.503
2019-01-01 05:00:00	5.755
2019-01-01 06:00:00	5.527
2019-01-01 07:00:00	5.667
2019-01-01 08:00:00	5.512
2019-01-01 09:00:00	5.576
2019-01-01 10:00:00	5.780
2019-01-01 11:00:00	5.616
2019-01-01 12:00:00	4.002
2019-01-01 13:00:00	3.767
2019-01-01 14:00:00	3.788
2019-01-01 15:00:00	4.141
2019-01-01 16:00:00	4.679
2019-01-01 17:00:00	4.531
2019-01-01 18:00:00	6.571
2019-01-01 19:00:00	8.677
2019-01-01 20:00:00	8.935
2019-01-01 21:00:00	6.283
2019-01-01 22:00:00	5.826
2019-01-01 23:00:00	5.973
2019-01-02 00:00:00	NaN
2019-01-02 01:00:00	10.267
2019-01-02 02:00:00	8.309
2019-01-02 03:00:00	6.833
2019-01-02 04:00:00	4.404
2019-01-02 05:00:00	4.539
2019-01-02 06:00:00	5.783
2019-01-02 07:00:00	6.381
2019-01-02 08:00:00	7.630
2019-01-02 09:00:00	8.259
2019-01-02 10:00:00	11.094
2019-01-02 11:00:00	10.472
2019-01-02 12:00:00	10.675
2019-01-02 13:00:00	12.769
2019-01-02 14:00:00	8.520
2019-01-02 15:00:00	8.267
2019-01-02 16:00:00	6.920
2019-01-02 17:00:00	7.923
2019-01-02 18:00:00	11.163
2019-01-02 19:00:00	10.962
2019-01-02 20:00:00	9.688
2019-01-02 21:00:00	9.557
2019-01-02 22:00:00	10.375
2019-01-02 23:00:00	10.576
2019-01-03 00:00:00	NaN
2019-01-03 01:00:00	15.151
2019-01-03 02:00:00	15.061
2019-01-03 03:00:00	11.498
2019-01-03 04:00:00	12.234
2019-01-03 05:00:00	18.182
.....

2019-01-03 06:00:00	20.382
2019-01-03 07:00:00	18.797
2019-01-03 08:00:00	21.356
2019-01-03 09:00:00	20.196

```
1 # Find negative values in the DataFrame
2 negative_mask = new_data < 0
3
4 # Print the negative values with their indices
5 print("Negative Values in DataFrame:")
6 print(new_data[negative_mask])
7
8 # Get indices of negative values
9 negative_indices = negative_mask.any(1)
10 indices_of_negative_values = new_data.index[negative_indices].tolist()
11
12 # Print indices where negative values are found
13 print("Indices with Negative Values:")
14 print(indices_of_negative_values)
```

Negative Values in DataFrame:

Series([], Name: PM2.5, dtype: float64)

TypeError

Traceback (most recent call

last)

Cell In[36], line 9

```
    6 print(new_data[negative_mask])
     8 # Get indices of negative values
----> 9 negative_indices = negative_mask.any(1)
    10 indices_of_negative_values =
new_data.index[negative_indices].tolist()
    12 # Print indices where negative values are found
```

```
1 stdPM25K30 = ((abs(6.669333-6.653877)/6.669333+(8.989437-6.540468)/8.
2
3 print(stdPM25K30)
```

18.005620292199904

```
1 stdPM25K25 = ((abs(6.669333-7.416088)/6.669333+(8.989437-6.371768)/8.
2
3 print(stdPM25K25)
```

19.272254554275175

```
1 stdPM25K20 = ((abs(6.669333-7.935228)/6.669333+(8.989437-6.496008)/8.
2
3 print(stdPM25K20)
```

19.73062315394634

```
1 stdPM25K26 = ((abs(6.669333-7.235088)/6.669333+abs(8.989437-6.410535)
2
```

```
3 print(stdPM25K26)
```

```
18.60411366250863
```

```
1 stdPM25K27 = ((abs(6.669333-7.070391)/6.669333+abs(8.989437-6.453523)
2
3 print(stdPM25K27)
```

```
18.11261140312003
```

```
1 stdPM10K30 = (((8.476048-7.755913)/8.476048+(12.351880- 8.563604)/12.
2
3 print(stdPM10K30)
```

```
22.23827295953838
```

```
1 stdPM10K25 = ((abs(8.476048-8.595012)/8.476048+(12.351880-8.429258)/1
2
3 print(stdPM10K25)
```

```
19.20593153026512
```

```
1 stdPM10K27 = ((abs(8.476048-8.212715)/8.476048+abs(12.351880-8.470781
2
3 print(stdPM10K27)
```

```
20.366802886415016
```

```
1 stdPM10K26 = ((abs(8.476048-8.395274)/8.476048+abs(12.351880-8.443440
2
3 print(stdPM10K26)
```

```
19.583395746807906
```

```
1 def PM25_Mean_calculator(mean2019, mean2020, mean2021, mean2022, mean
2
3     PM25_Mean = ((abs(7.860134-mean2019)/7.860134 + abs(8.610198-mean
4
5     print(PM25_Mean)
6     return(PM25_Mean)
```

```
1 mean2019 = 9.967921
2 mean2020 = 8.383908
3 mean2021 = 7.342154
4 mean2022 = 9.015066
5 mean2023 = 6.900610
6 P25_K25_mean = PM25_Mean_calculator(mean2019,mean2020, mean2021, mean
7
```

```
14.090394229770942
```

```
1 mean2019 = 9.903337
2 mean2020 = 8.504357
3 mean2021 = 7.367673
4 mean2022 = 8.939114
5 mean2023 = 7.019881
6 P25_K26_mean = PM25_Mean_calculator(mean2019,mean2020, mean2021, mean
7
13.05252001213361
```

```
1 mean2019 = 9.836766
2 mean2020 = 8.595274
3 mean2021 = 7.376906
4 mean2022 = 8.867190
5 mean2023 = 7.109005
6 P25_K27_mean = PM25_Mean_calculator(mean2019,mean2020, mean2021, mean
7
```

```
12.21089919693115
```

```
1 def PM10_Mean_calculator(mean2019, mean2020, mean2021, mean2022, mean
2
3     PM10_Mean = ((abs(12.786331-mean2019)/12.786331 + abs(14.121757-m
4
5     print(PM10_Mean)
6     return(PM10_Mean)
```

```
1 mean2019 = 14.705845
2 mean2020 = 12.710035
3 mean2021 = 10.940160
4 mean2022 = 13.436019
5 mean2023 = 11.067249
6 P10_K25_mean = PM10_Mean_calculator(mean2019,mean2020, mean2021, mean
7
```

```
12.44403820297543
```

```
1 mean2019 = 14.611628
2 mean2020 = 12.861032
3 mean2021 = 10.982364
4 mean2022 = 13.346716
5 mean2023 = 11.226019
6 P10_K26_mean = PM10_Mean_calculator(mean2019,mean2020, mean2021, mean
7
```

```
11.617175147850848
```

```
1 mean2019 = 14.522029
2 mean2020 = 12.980646
3 mean2021 = 11.006023
4 mean2022 = 13.260031
5 mean2023 = 11.334385
```

```
6 P10_K27_mean = PM10_Mean_calculator(mean2019,mean2020, mean2021, mean  
10.95526651785195
```

```
1 mean2019 = 14.192712  
2 mean2020 = 13.175501  
3 mean2021 = 11.009497  
4 mean2022 = 13.021303  
5 mean2023 = 11.546916  
6 P10_K30_mean = PM10_Mean_calculator(mean2019,mean2020, mean2021, mean  
9.427483614091189
```

```
1 Col_PM25_mean_change = ((abs(29.783682 - 28.821035)/29.783682+abs(28.  
2  
3 print(Col_PM25_mean_change)
```

```
1.1595619191829
```

```
1 omean1 = 37.525830  
2 imean1 = 35.910604  
3 omean2 = 33.046621  
4 imean2 = 31.605245  
5 omean3 = 33.320409  
6 imean3 = 31.828604  
7 omean4 = 28.522565  
8 imean4 = 27.943796  
9 omean5 = 34.553358  
10 imean5 = 33.720975  
11 omean6 = 34.001631  
12 imean6 = 31.986114  
13 n = 6  
14  
15 Percentage_Change = ((abs(iman1 - omean1)/omean1+abs(iman2- omean2)  
16 print(Percentage_Change)
```

```
3.918157802280703
```

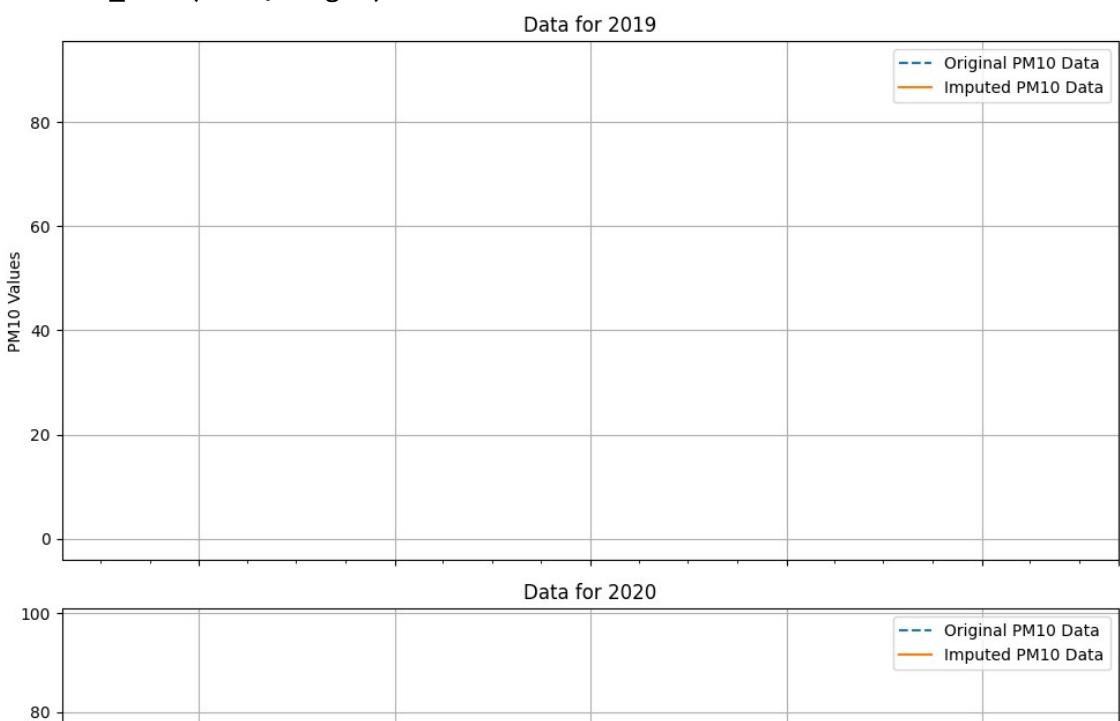
```
1 #Plot the data so that we can vizualise that the imputation does not  
2 #Plot the data by years so that you can see the patterns in the data  
3  
4 #Select which data to plot:  
5 old_data = UK_or_df['PM2.5']  
6 new_data = UK_K_fill['PM2.5']  
7  
8 #Group the data by year  
9 grouped_old = old_data.groupby(old_data.index.year)  
10 grouped_new = new_data.groupby(new_data.index.year)  
11  
12  
13  
14 # Set up the matplotlib figure and axes
```

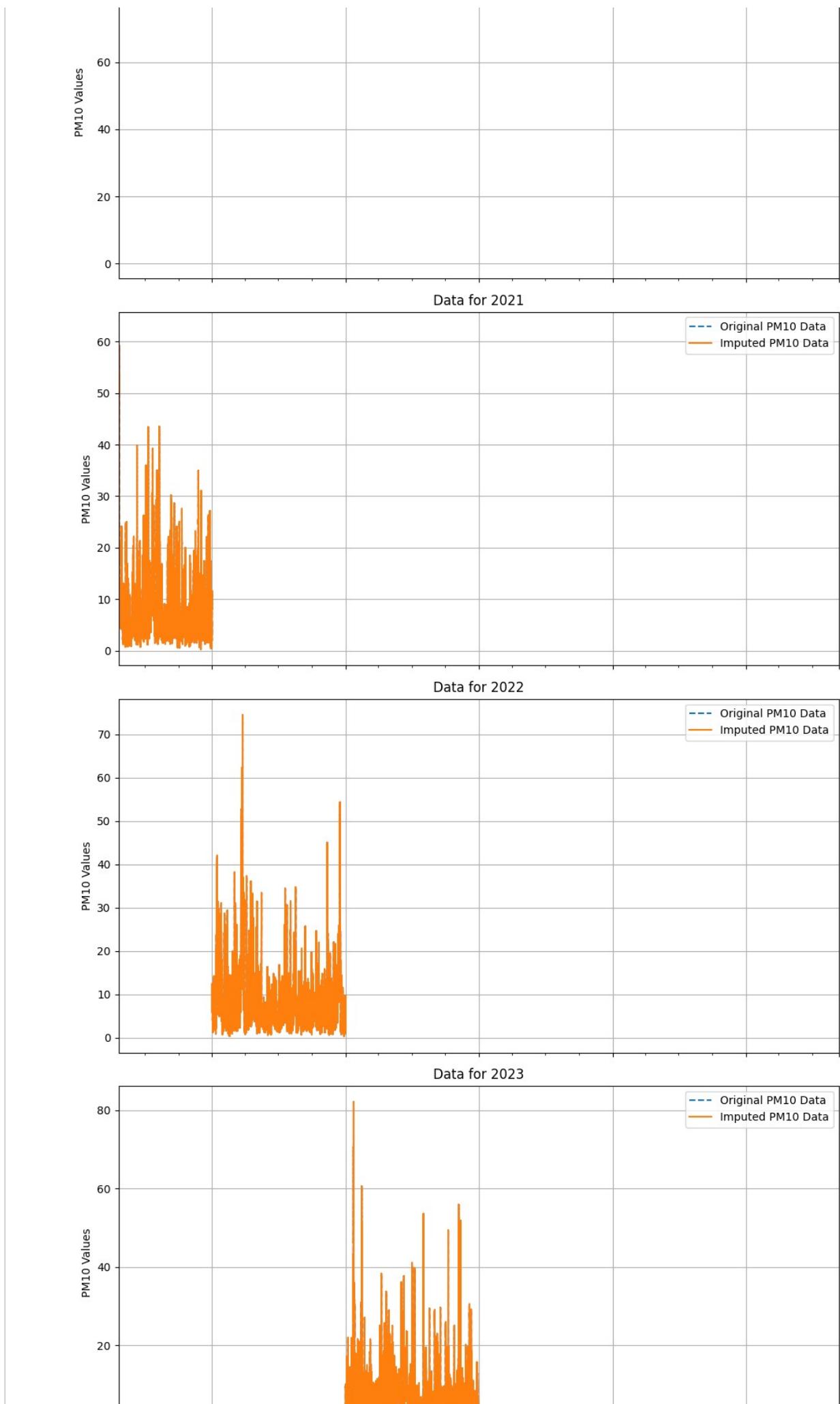
```
15 fig, axes = plt.subplots(nrows=ten(grouped_old), figsize=(10, 5 * ten
16
17
18 # Plot each group
19 for ((key1, group1), (key2, group2)), ax in zip(zip(grouped_old, grou
20     print("Plotting Old Data year:", key1)
21     print(group1.describe())
22     print("Plotting Imputed Data year:", key2)
23     print(group2.describe())
24
25     # Plot data from the first dataset
26     group1.plot(ax=ax, label='Original PM10 Data', linestyle='--')
27
28     # Plot data from the second dataset
29     group2.plot(ax=ax, label='Imputed PM10 Data') # Dashed line for
30
31     # Set the title to the current year from the first dataset's key
32     ax.set_title(f'Data for {key1}')
33     ax.set_ylabel('PM10 Values')
34     ax.grid(True)
35     ax.legend()
36
37 plt.tight_layout()
38 plt.show()
39
40
41
```

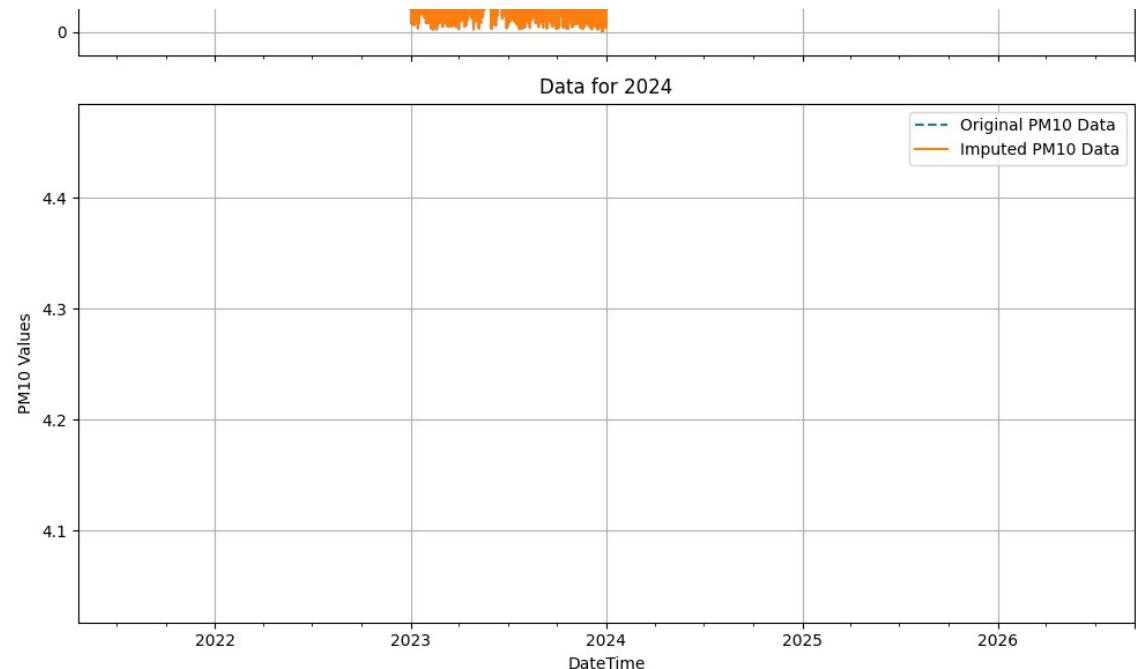
```
Plotting Old Data year: 2019
count    8509.000000
mean      9.535257
std       9.161261
min       0.413000
25%      4.130000
50%      6.337000
75%      11.175000
max      90.981000
Name: PM2.5, dtype: float64
Plotting Imputed Data year: 2019
count    8759.000000
mean      9.471212
std       9.068612
min       0.413000
25%      4.161500
50%      6.349000
75%      10.968500
max      90.981000
Name: PM2.5, dtype: float64
Plotting Old Data year: 2020
count    8681.000000
mean      7.916442
std       7.582349
min       0.319000
25%      3.575000
50%      5.449000
75%      9.155000
... 100000
```

```
    idx          ડાટાસ્ટોર  
Name: PM2.5, dtype: float64  
Plotting Imputed Data year: 2020  
count    8784.000000  
mean      7.966047  
std       7.579192  
min       0.319000  
25%      3.600750  
50%      5.483500  
75%      9.264000  
max      96.163000  
Name: PM2.5, dtype: float64  
Plotting Old Data year: 2021  
count    8703.000000  
mean      7.747758  
std       6.268321  
min       0.280000  
25%      3.763000  
50%      5.929000  
75%      9.600000  
max      62.623000  
Name: PM2.5, dtype: float64  
Plotting Imputed Data year: 2021  
count    8760.000000  
mean      7.758845  
std       6.282386  
min       0.280000  
25%      3.759750  
50%      5.929000  
75%      9.612500  
max      62.623000  
Name: PM2.5, dtype: float64  
Plotting Old Data year: 2022  
count    8694.000000  
mean      7.903948  
std       6.798462  
min       0.281000  
25%      3.841000  
50%      5.684500  
75%      9.517000  
max      74.470000  
Name: PM2.5, dtype: float64  
Plotting Imputed Data year: 2022  
count    8760.000000  
mean      7.894994  
std       6.778363  
min       0.281000  
25%      3.851750  
50%      5.706000  
75%      9.475500  
max      74.470000  
Name: PM2.5, dtype: float64  
Plotting Old Data year: 2023  
count    8749.000000  
mean      7.135204  
std       6.407699  
min       0.147000  
25%      3.490000  
50%
```

```
    75%      8.463000
    max     82.144000
Name: PM2.5, dtype: float64
Plotting Imputed Data year: 2023
count    8760.000000
mean     7.132763
std      6.404382
min      0.147000
25%      3.490000
50%      5.288500
75%      8.463000
max     82.144000
Name: PM2.5, dtype: float64
Plotting Old Data year: 2024
count    1.000
mean     4.251
std      NaN
min      4.251
25%      4.251
50%      4.251
75%      4.251
max     4.251
Name: PM2.5, dtype: float64
Plotting Imputed Data year: 2024
count    1.000
mean     4.251
std      NaN
min      4.251
25%      4.251
50%      4.251
75%      4.251
max     4.251
Name: PM2.5, dtype: float64
c:\Users\User\Code\databook\Lib\site-packages\pandas\plotting\_matplotlib\
    ax.set_xlim(left, right)
c:\Users\User\Code\databook\Lib\site-packages\pandas\plotting\_matplotlib\
    ax.set_xlim(left, right)
```







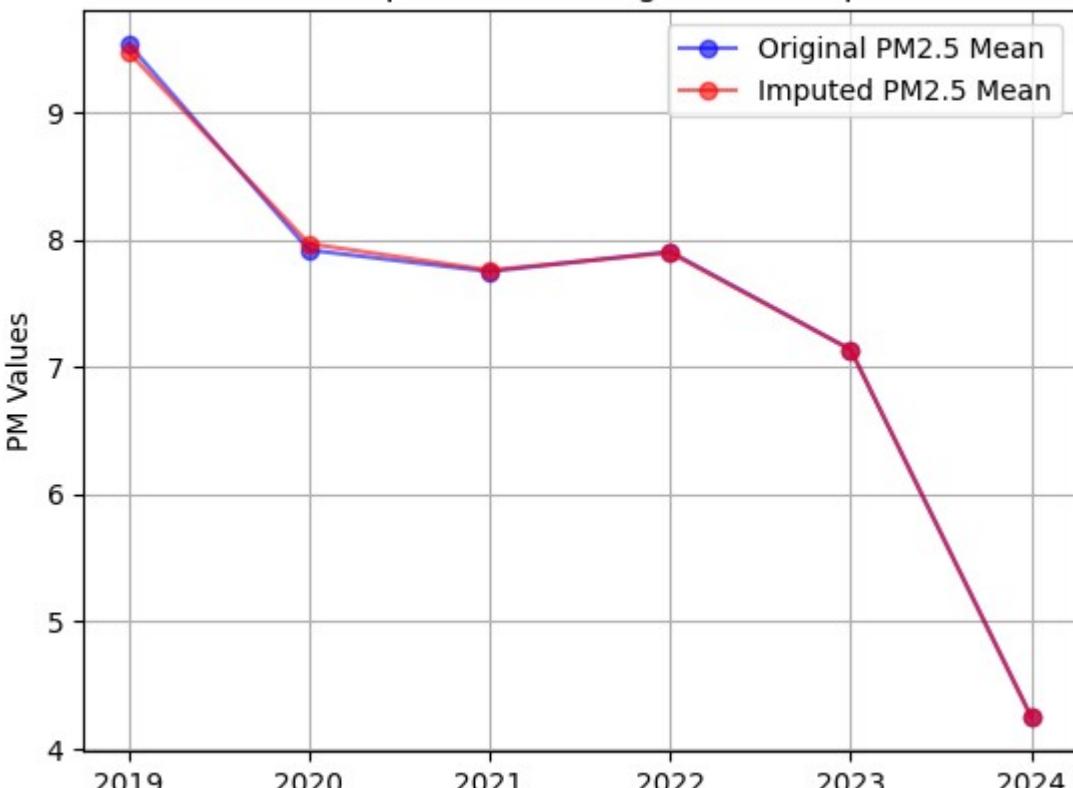
```
1 #function to print out and plot the statistics of the dataframe
2 def PM_imputation_stats_calc(df_orginal, df_impute, input_data):
3
4     #Calculate Overall Statistics of original and Imputed Data
5     #print(f"Original {input_data} Data Mean = {df_orginal[input_data]}
6     #print(f"Imputed {input_data} Data Mean = {df_impute[input_data].s
7     print(f"Percentage Change of {input_data} Overall Mean = {(abs(df_
8
9     #print(f"Original {input_data} Data Std = {df_orginal[input_data]
10    #print(f"Imputed {input_data} Data Std = {df_impute[input_data].s
11    print(f"Percentage Change of {input_data} Overall Std = {(abs(df_
12
13
14    grouped_original = df_orginal.groupby(df_orginal.index.year)
15    grouped_imputed = df_impute.groupby(df_impute.index.year)
16
17    #PCalcuate mean and std statistics
18    group_original_mean = grouped_original[input_data].mean()
19    grouped_imputed_mean = grouped_imputed[input_data].mean()
20    group_original_std = grouped_original[input_data].std()
21    grouped_imputed_std = grouped_imputed[input_data].std()
22
23
```

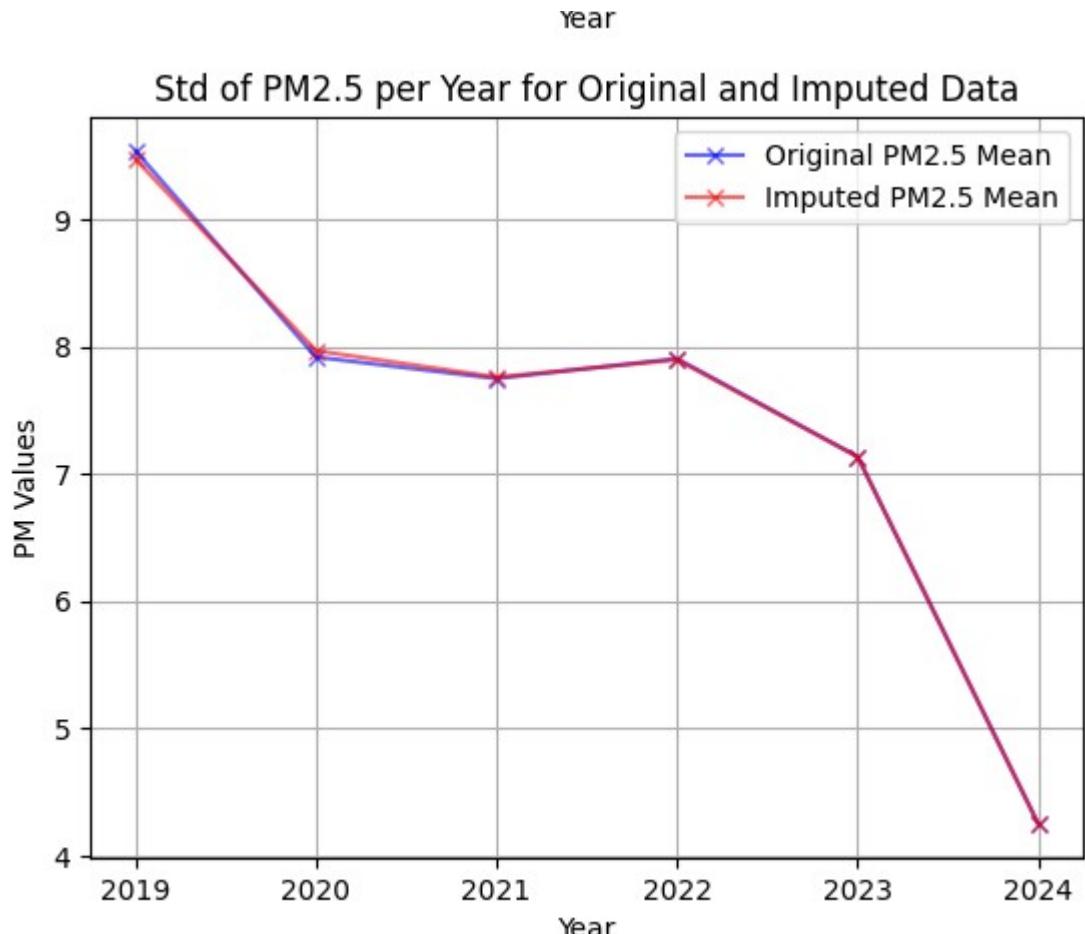
```
24     #Plotting the means
25     fig, ax = plt.subplots()
26
27     ax.plot(group_original_mean.index, group_original_mean, color='blue')
28     ax.plot(grouped_imputed_mean.index, grouped_imputed_mean, color='red')
29
30     plt.xlabel('Year')
31     plt.ylabel('PM Values')
32     plt.title(f'Mean of {input_data} per Year for Original and Imputed')
33     plt.legend()
34     plt.grid(True)
35     plt.show()
36
37     #Plotting the standard deviation
38     fig, ax = plt.subplots()
39
40     ax.plot(group_original_mean.index, group_original_mean, color='blue')
41     ax.plot(grouped_imputed_mean.index, grouped_imputed_mean, color='red')
42
43     plt.xlabel('Year')
44     plt.ylabel('PM Values')
45     plt.title(f'Std of {input_data} per Year for Original and Imputed')
46     plt.legend()
47     plt.grid(True)
48     plt.show()
```

```
1 PM_imputation_stats_calc(UK_or_df, UK_K_fill, 'PM2.5')
2 PM_imputation_stats_calc(UK_or_df, UK_K_fill, 'PM10')
```

Percentage Change of PM2.5 Overall Mean = 0.05605355130964996
Percentage Change of PM2.5 Overall Std = 0.23795888178094704

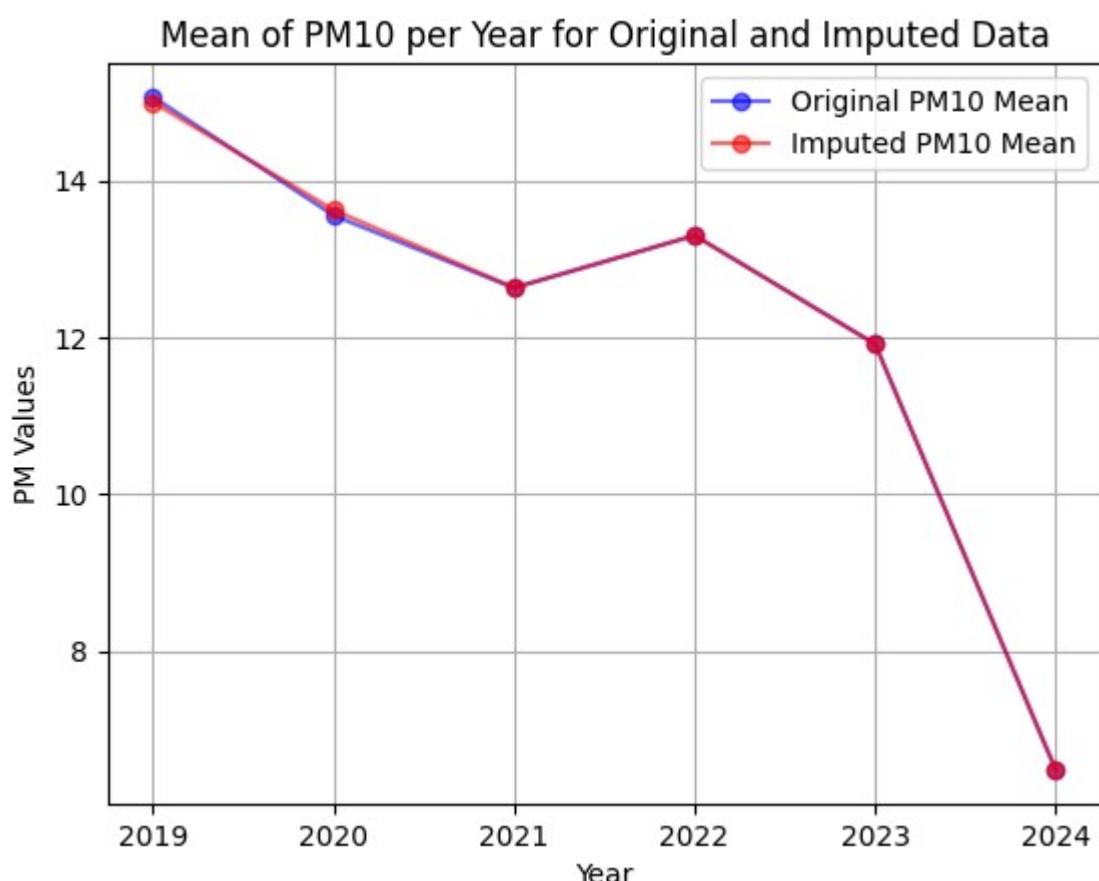
Mean of PM2.5 per Year for Original and Imputed Data





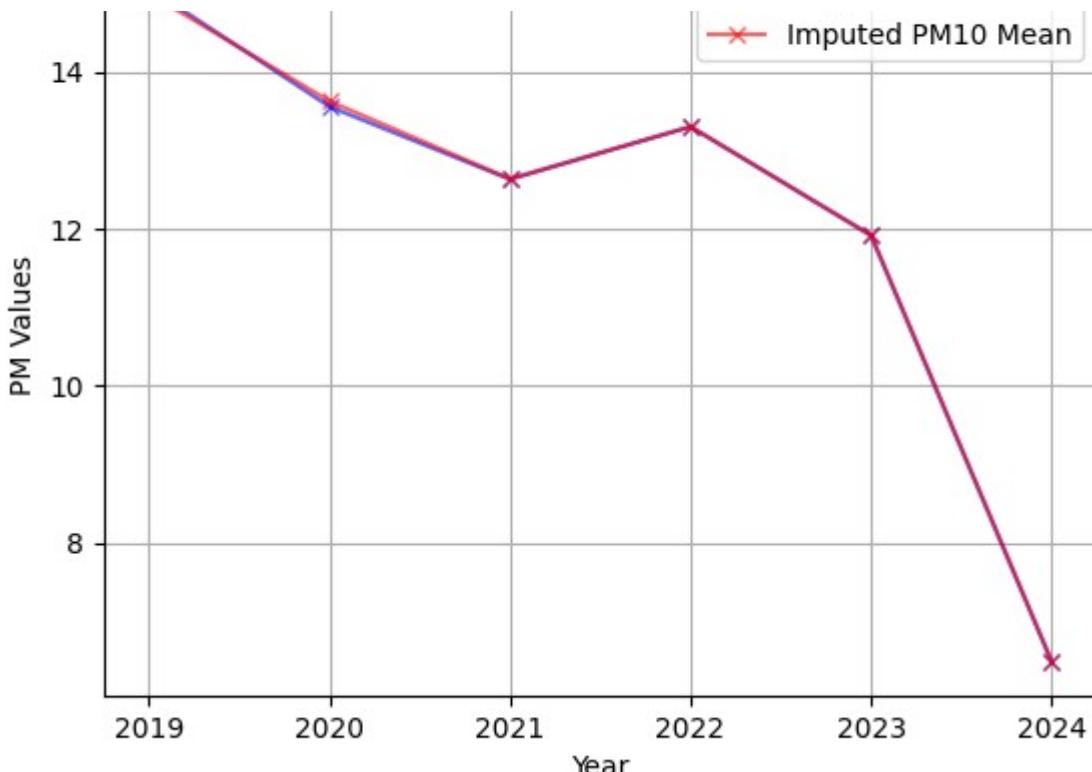
Percentage Change of PM10 Overall Mean = 0.08028171681506606

Percentage Change of PM10 Overall Std = 0.32061487495746716



Std of PM10 per Year for Original and Imputed Data



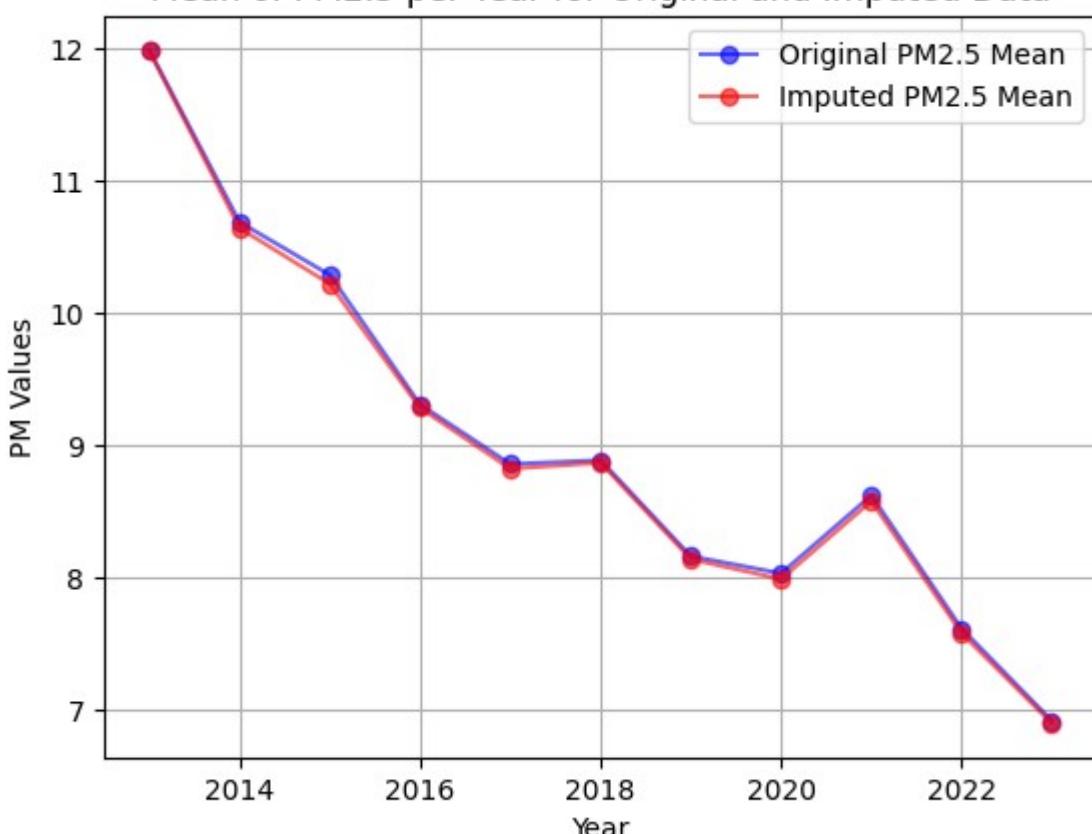


```
1 PM_imputation_stats_calc(US_or_df, US_K_fill, 'PM2.5')
2 PM_imputation_stats_calc(US_or_df, US_K_fill, 'PM10')
```

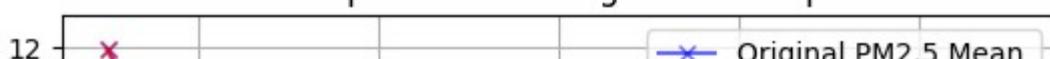
Percentage Change of PM2.5 Overall Mean = 0.427891717051312

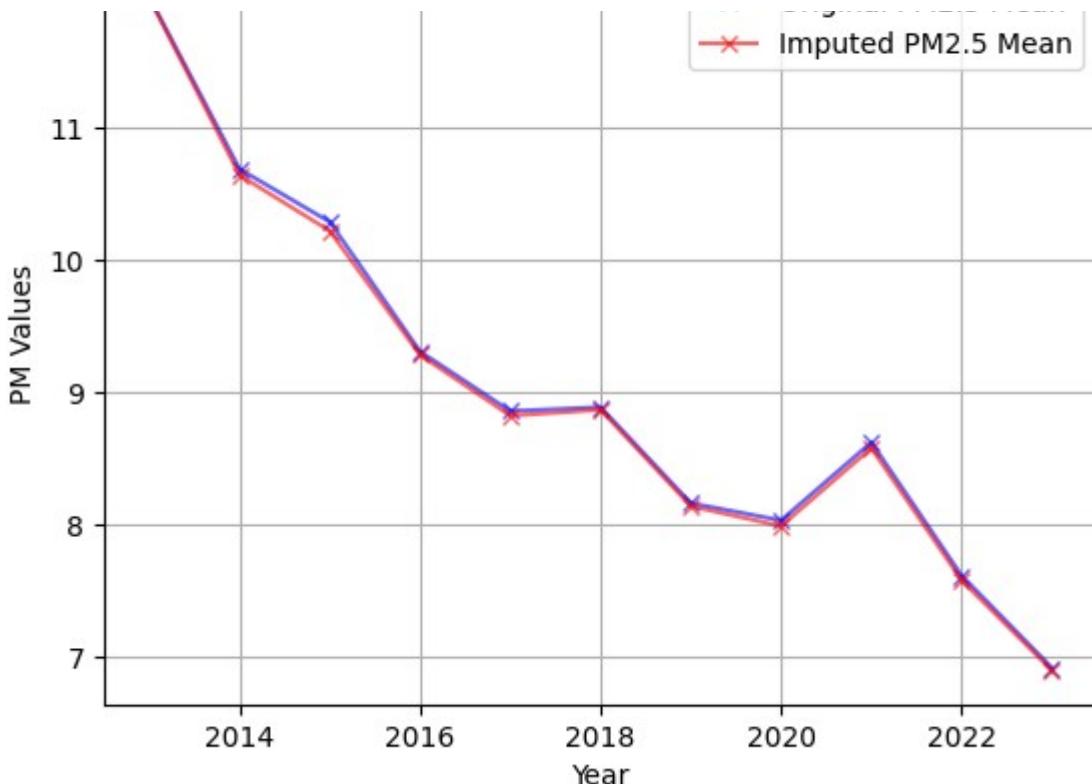
Percentage Change of PM2.5 Overall Std = 0.5778582720704228

Mean of PM2.5 per Year for Original and Imputed Data



Std of PM2.5 per Year for Original and Imputed Data

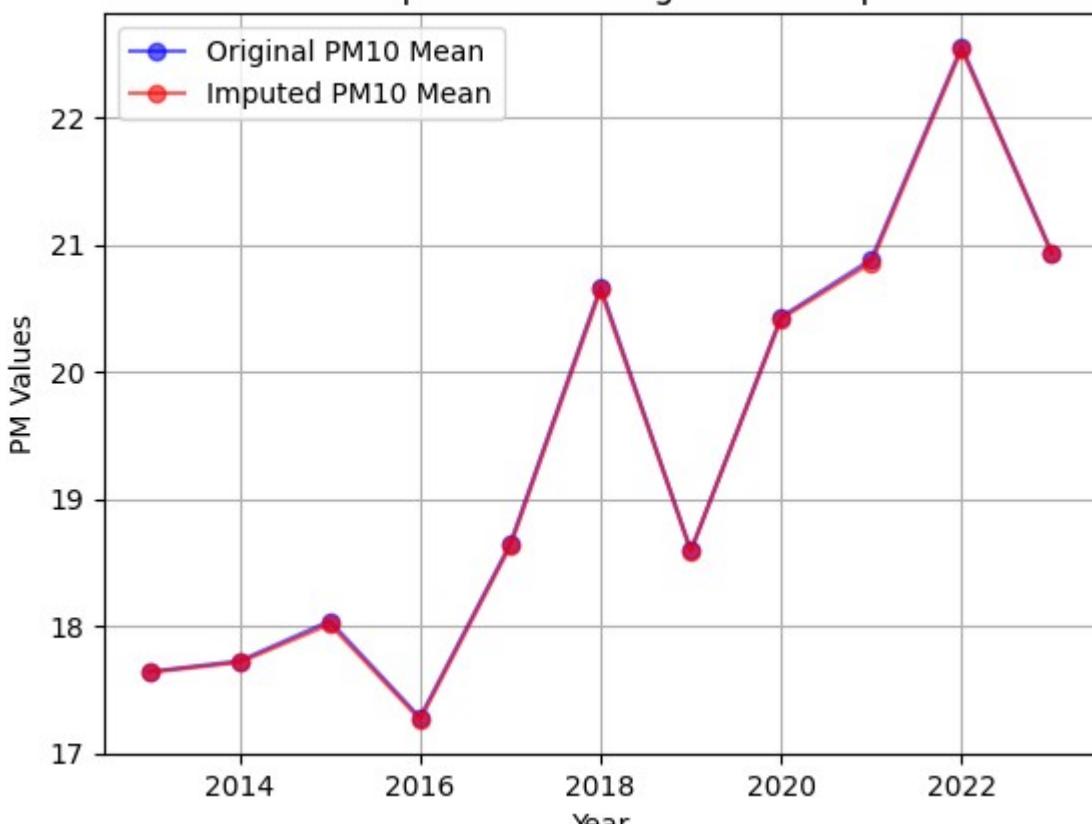




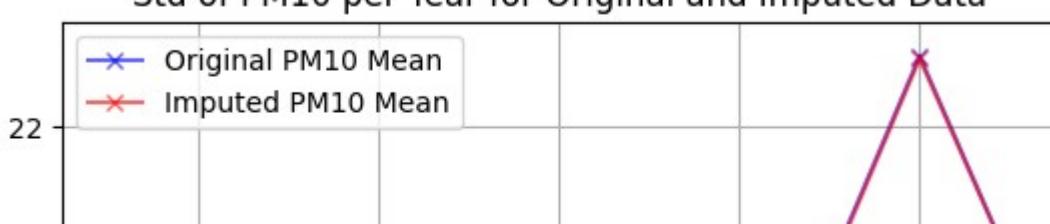
Percentage Change of PM10 Overall Mean = 0.0611882319009041

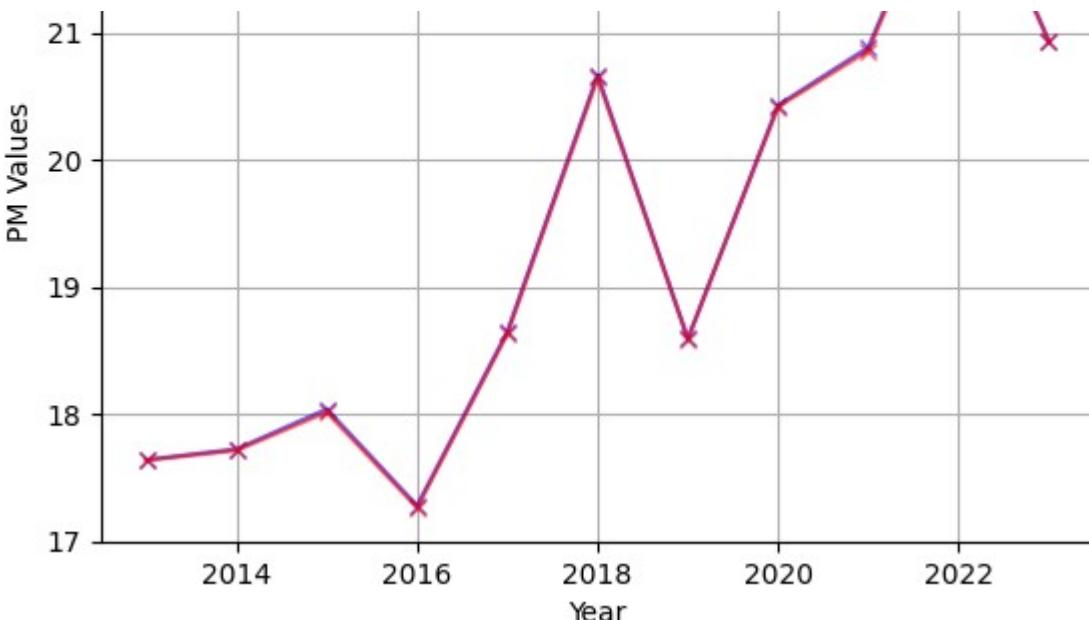
Percentage Change of PM10 Overall Std = 0.1673138123306832

Mean of PM10 per Year for Original and Imputed Data



Std of PM10 per Year for Original and Imputed Data



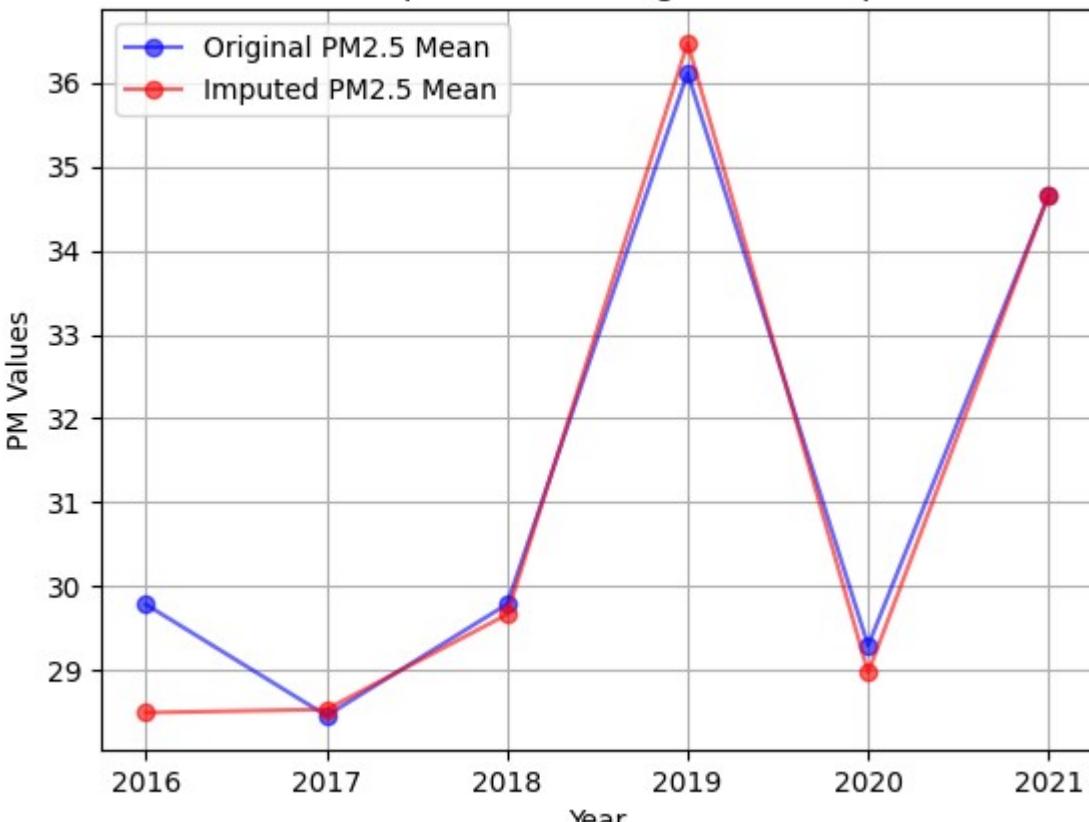


```
1 PM_imputation_stats_calc(Columbia_or_df, Columbia_K_fill, 'PM2.5')
2 PM_imputation_stats_calc(Columbia_or_df, Columbia_K_fill, 'PM10')
```

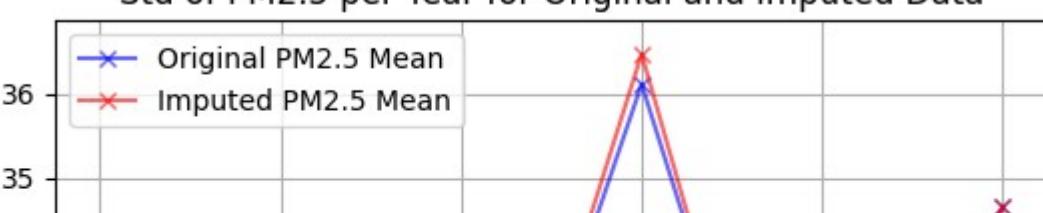
Percentage Change of PM2.5 Overall Mean = 0.7321924247741135

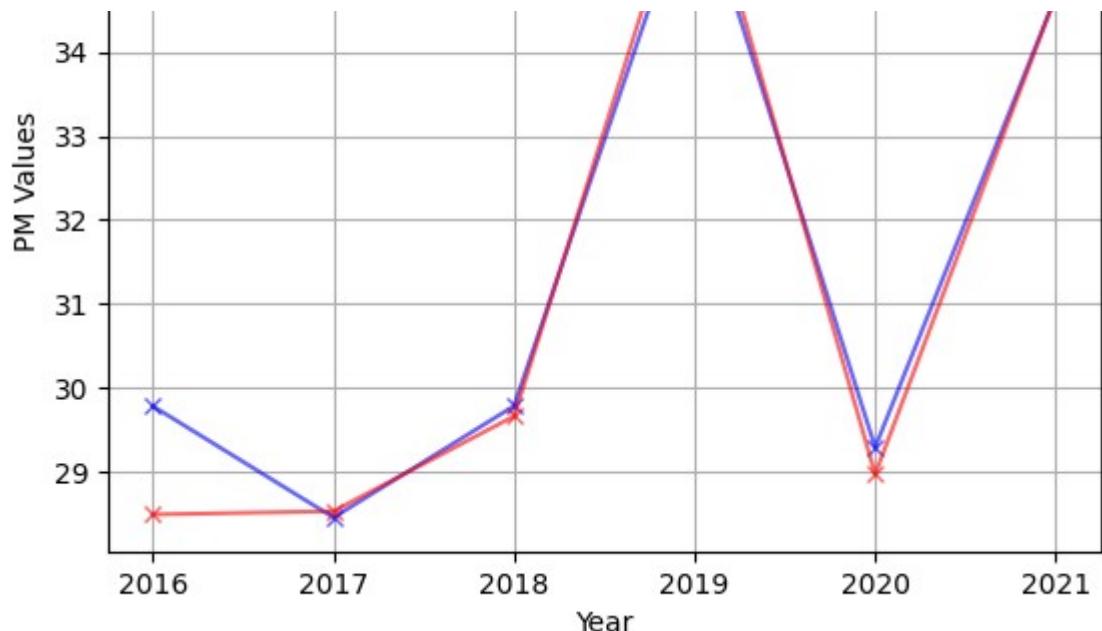
Percentage Change of PM2.5 Overall Std = 1.5793313433095268

Mean of PM2.5 per Year for Original and Imputed Data



Std of PM2.5 per Year for Original and Imputed Data

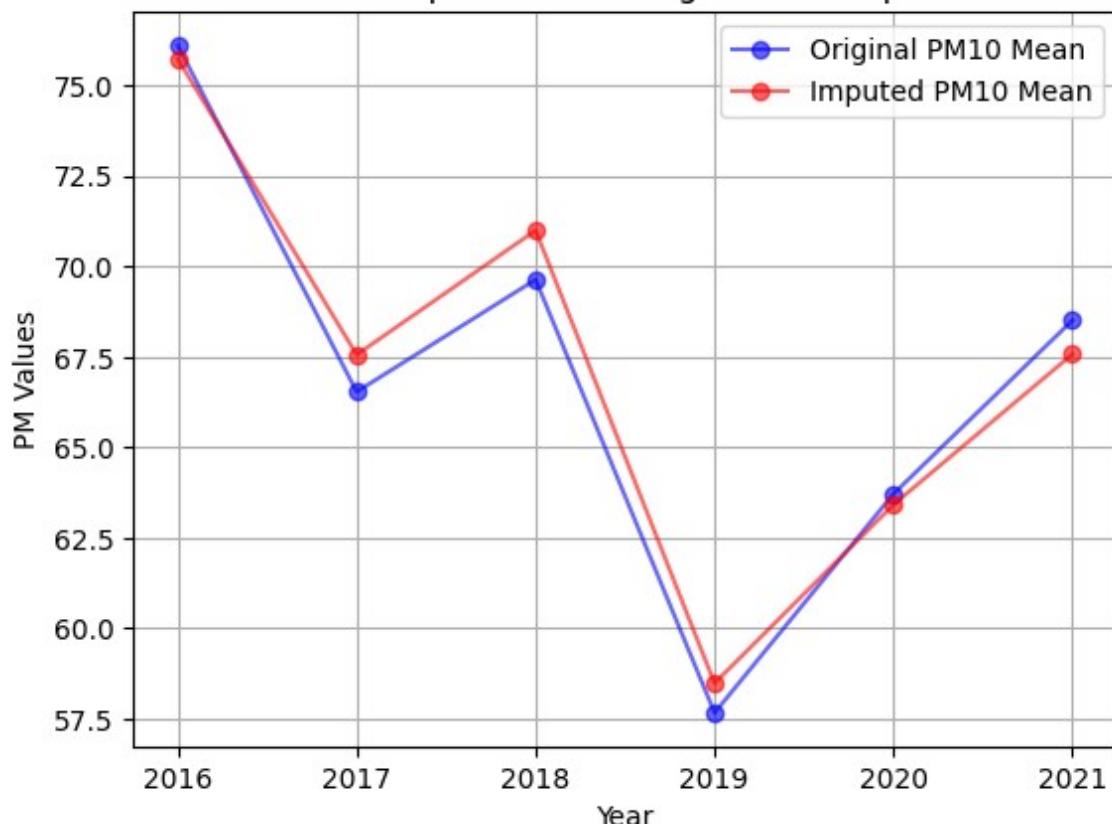




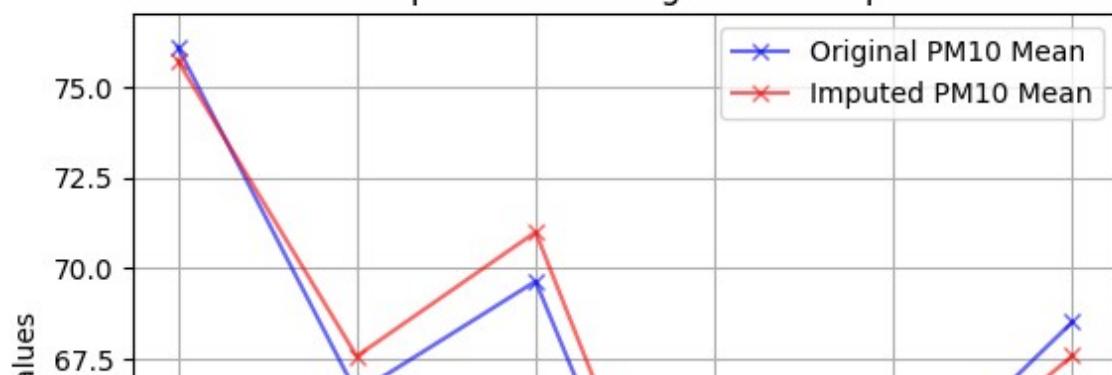
Percentage Change of PM10 Overall Mean = 0.4651727282316991

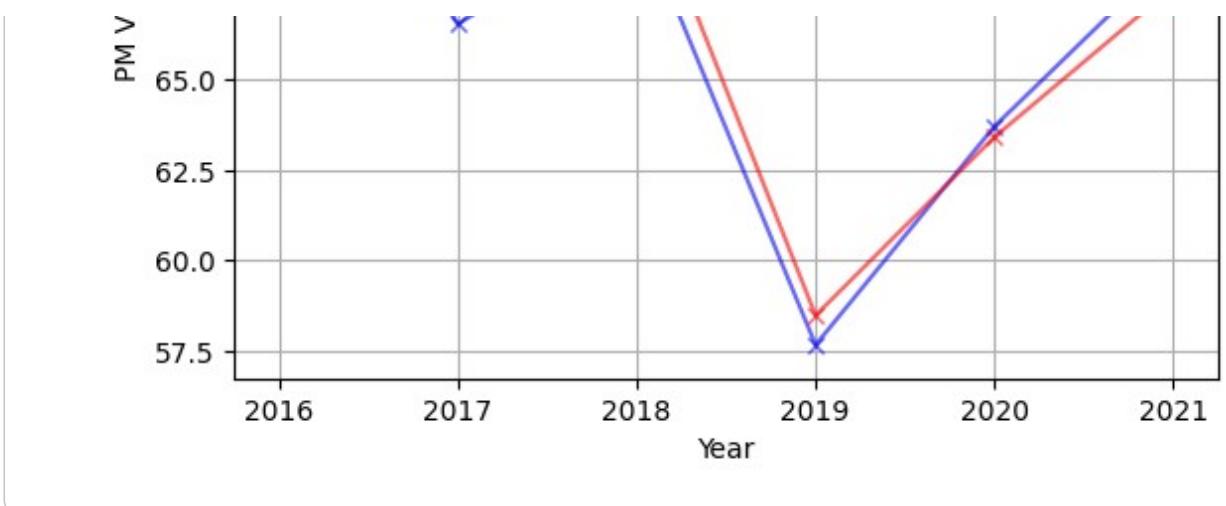
Percentage Change of PM10 Overall Std = 2.281018247802094

Mean of PM10 per Year for Original and Imputed Data



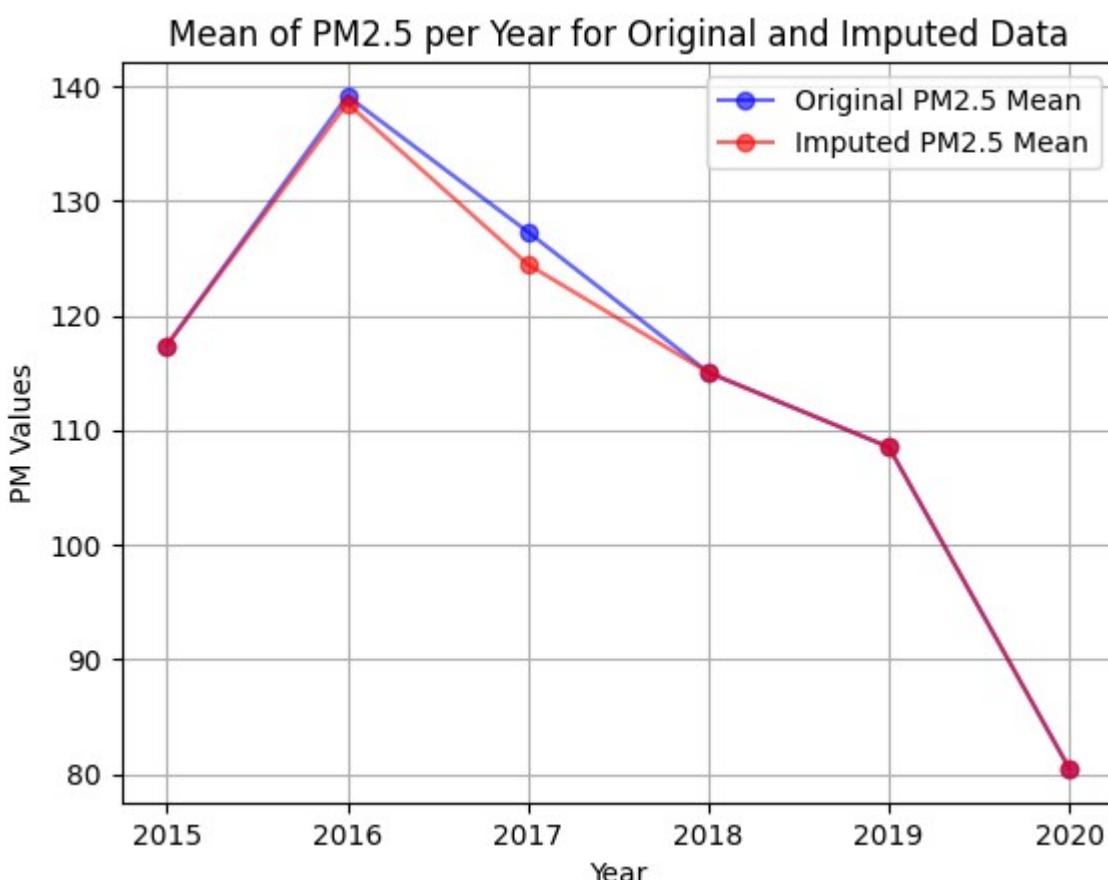
Std of PM10 per Year for Original and Imputed Data

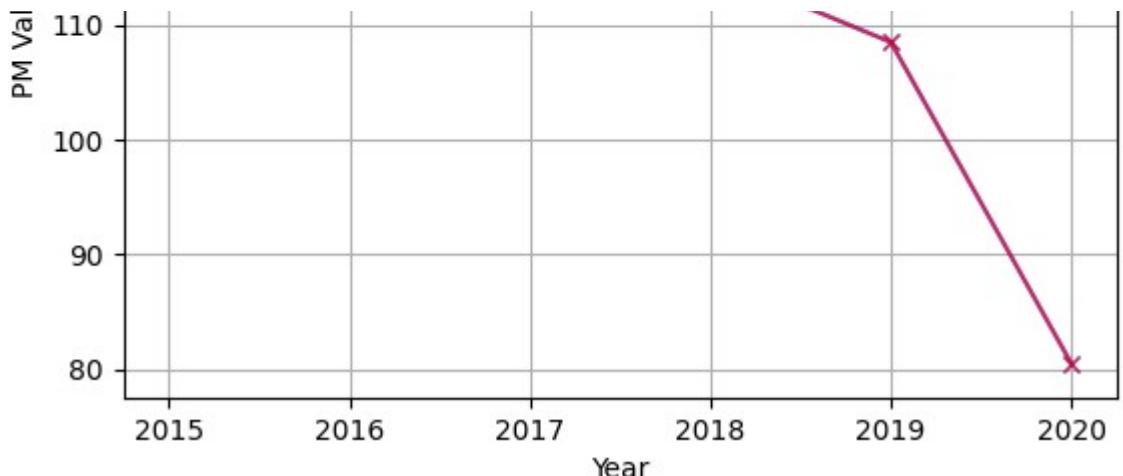




```
1 PM_imputation_stats_calc(India_or_df, India_K_fill, 'PM2.5')
2 PM_imputation_stats_calc(India_or_df, India_K_fill, 'PM10')
```

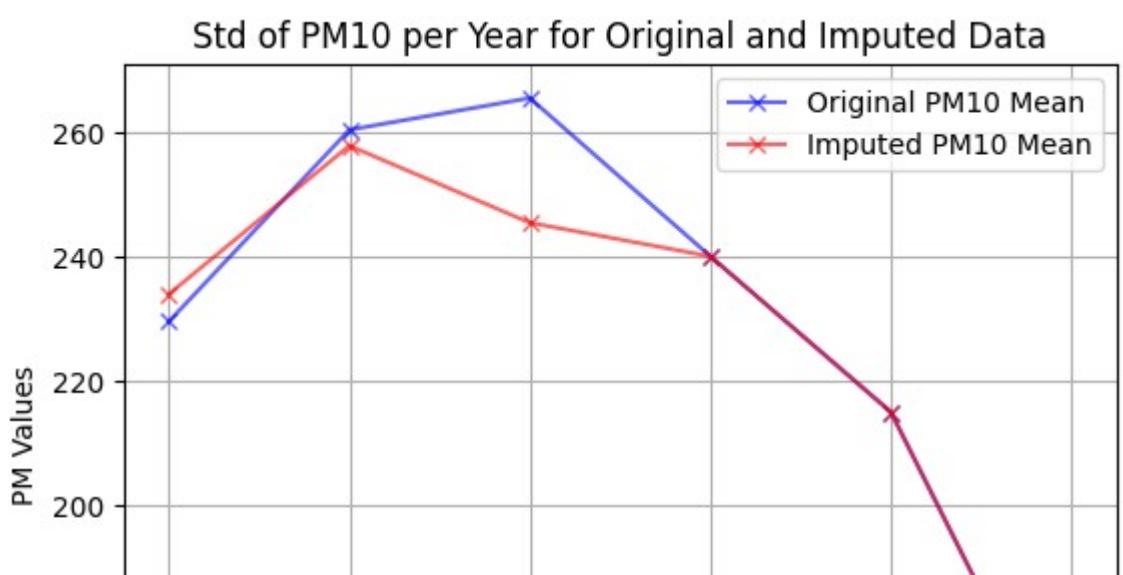
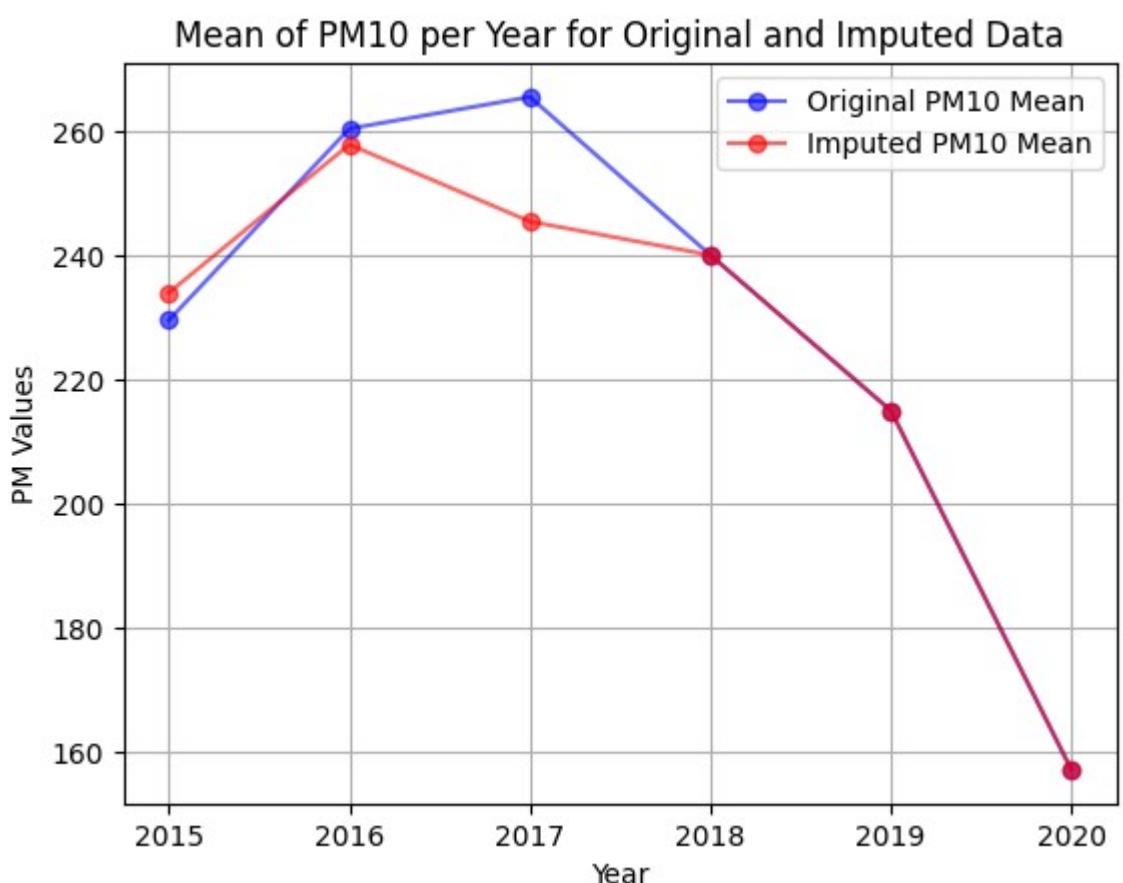
Percentage Change of PM2.5 Overall Mean = 0.4604876644418227
Percentage Change of PM2.5 Overall Std = 0.14906950836784796

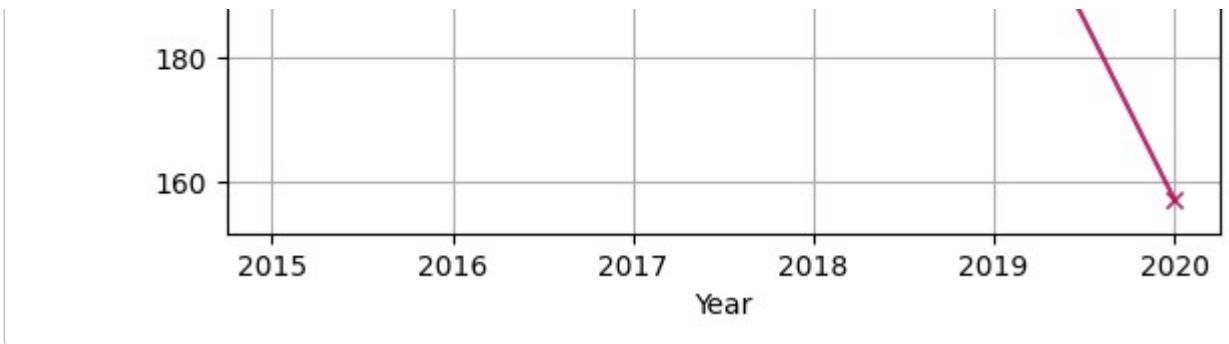




Percentage Change of PM10 Overall Mean = 0.8385303779176518

Percentage Change of PM10 Overall Std = 0.5125678445818023



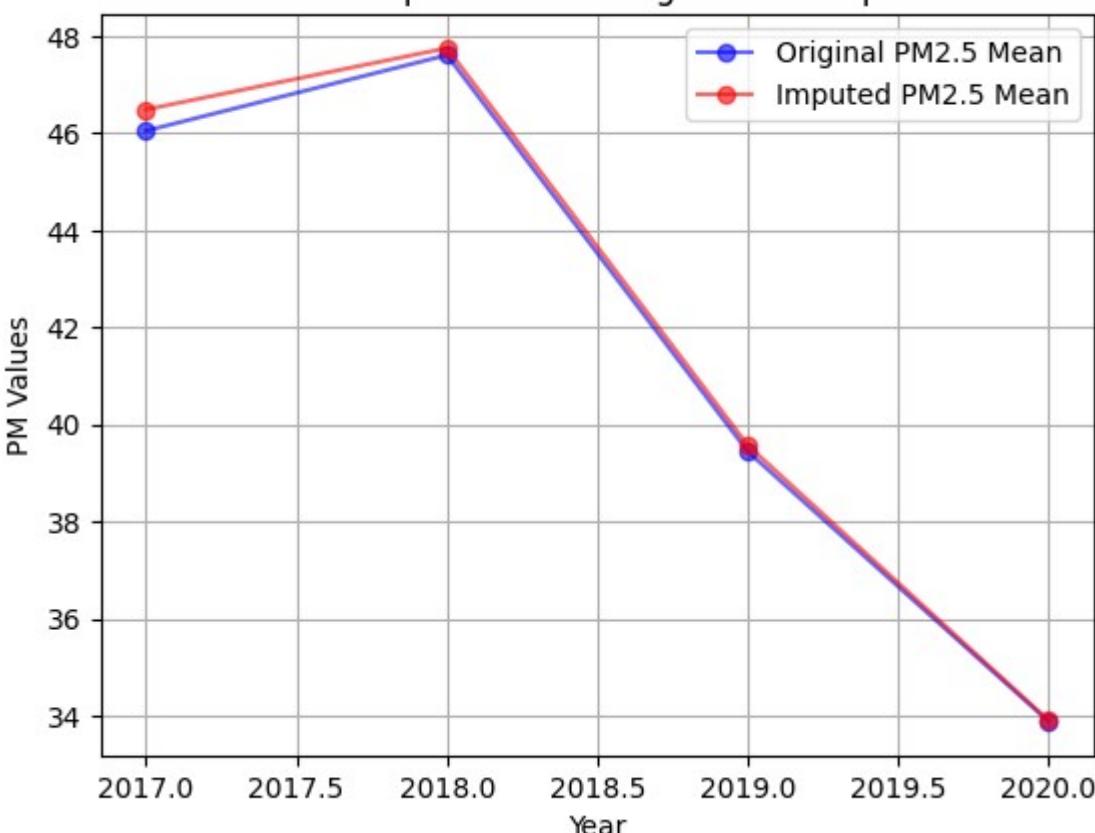


```
1 PM_imputation_stats_calc(China_or_df, China_K_fill, 'PM2.5')
2 PM_imputation_stats_calc(China_or_df, China_K_fill, 'PM10')
```

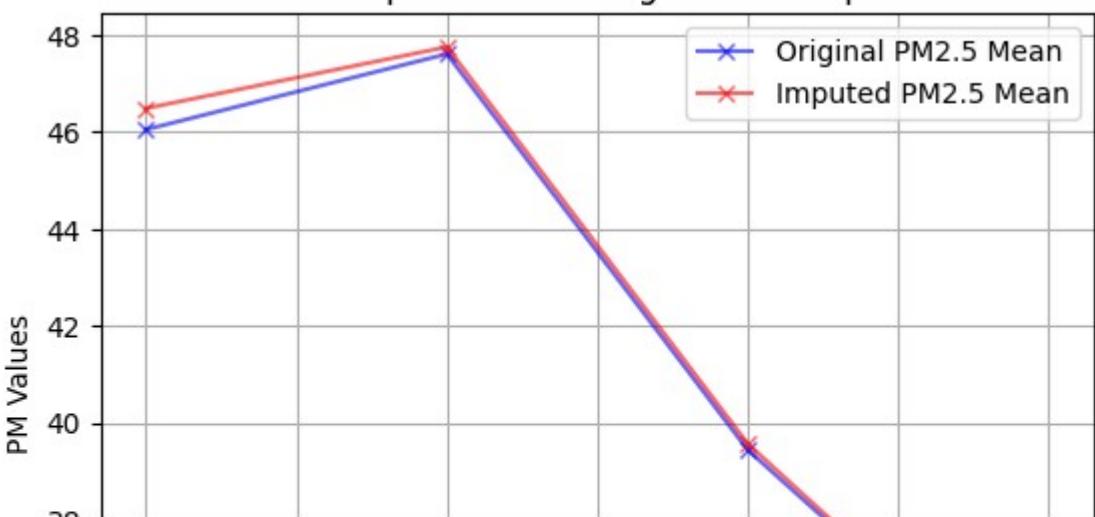
Percentage Change of PM2.5 Overall Mean = 0.44918468554038277

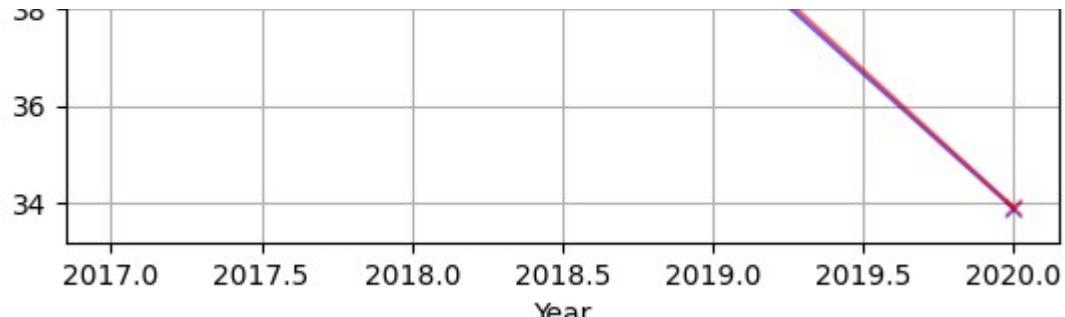
Percentage Change of PM2.5 Overall Std = 1.1434616881837634

Mean of PM2.5 per Year for Original and Imputed Data



Std of PM2.5 per Year for Original and Imputed Data

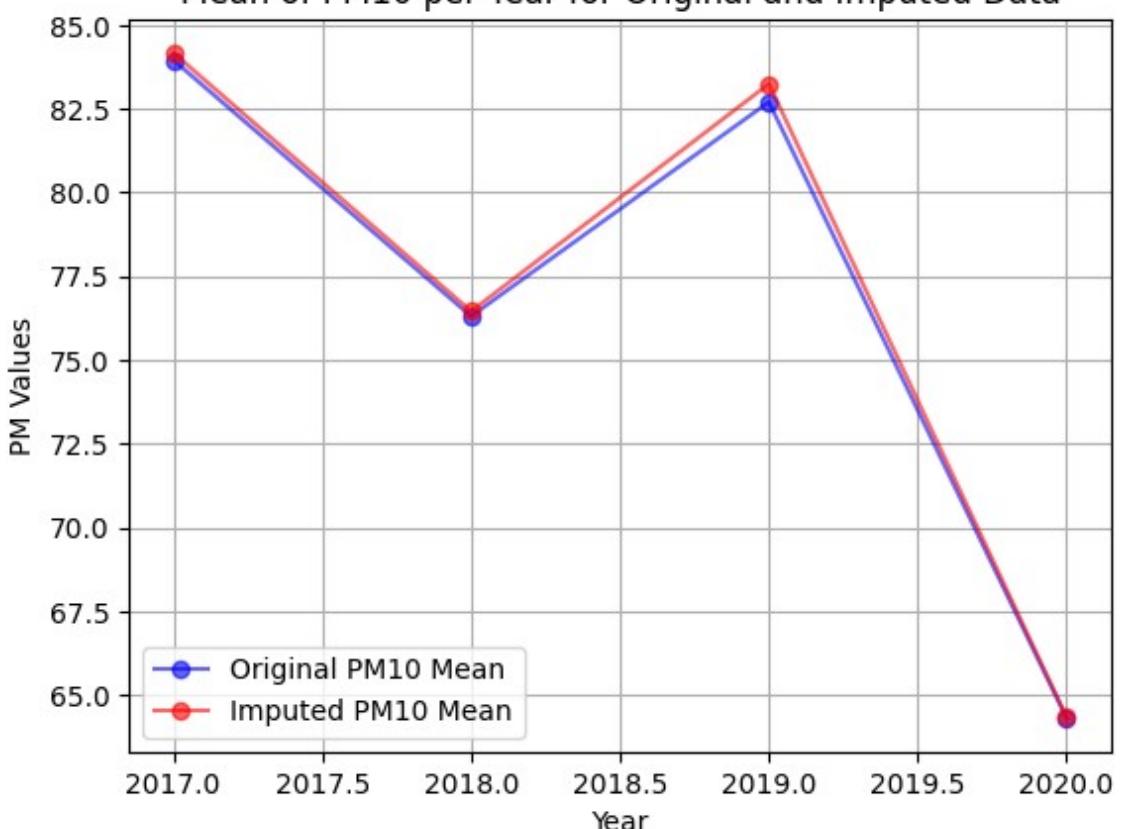




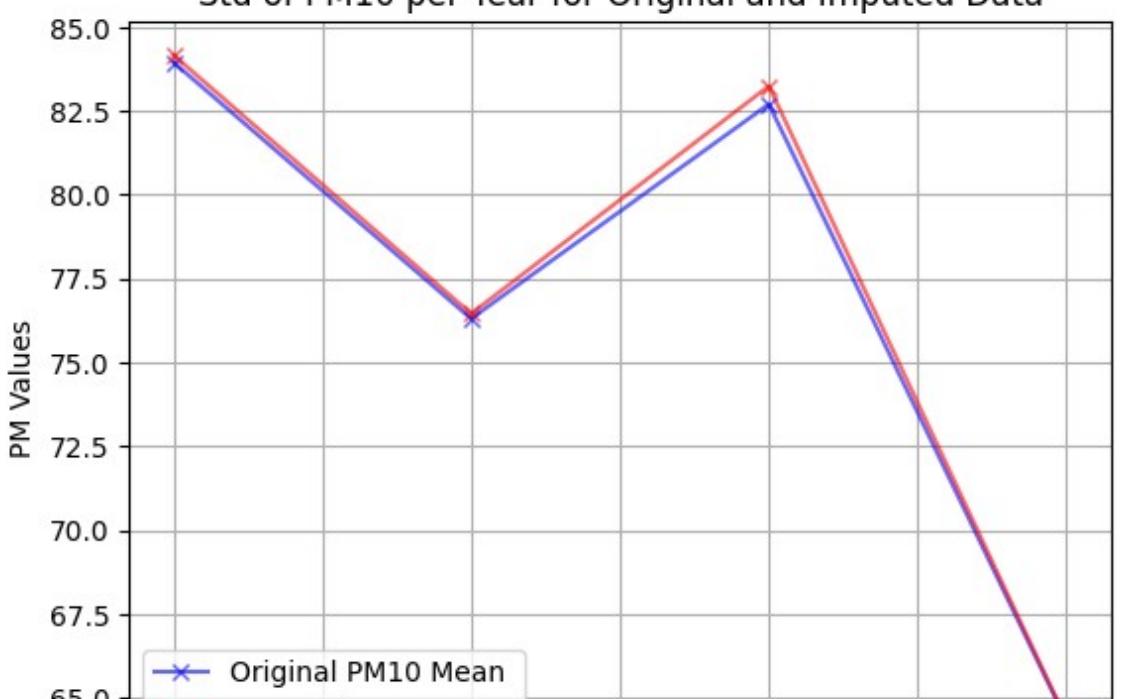
Percentage Change of PM10 Overall Mean = 0.31353906839155576

Percentage Change of PM10 Overall Std = 0.7985179262126181

Mean of PM10 per Year for Original and Imputed Data



Std of PM10 per Year for Original and Imputed Data



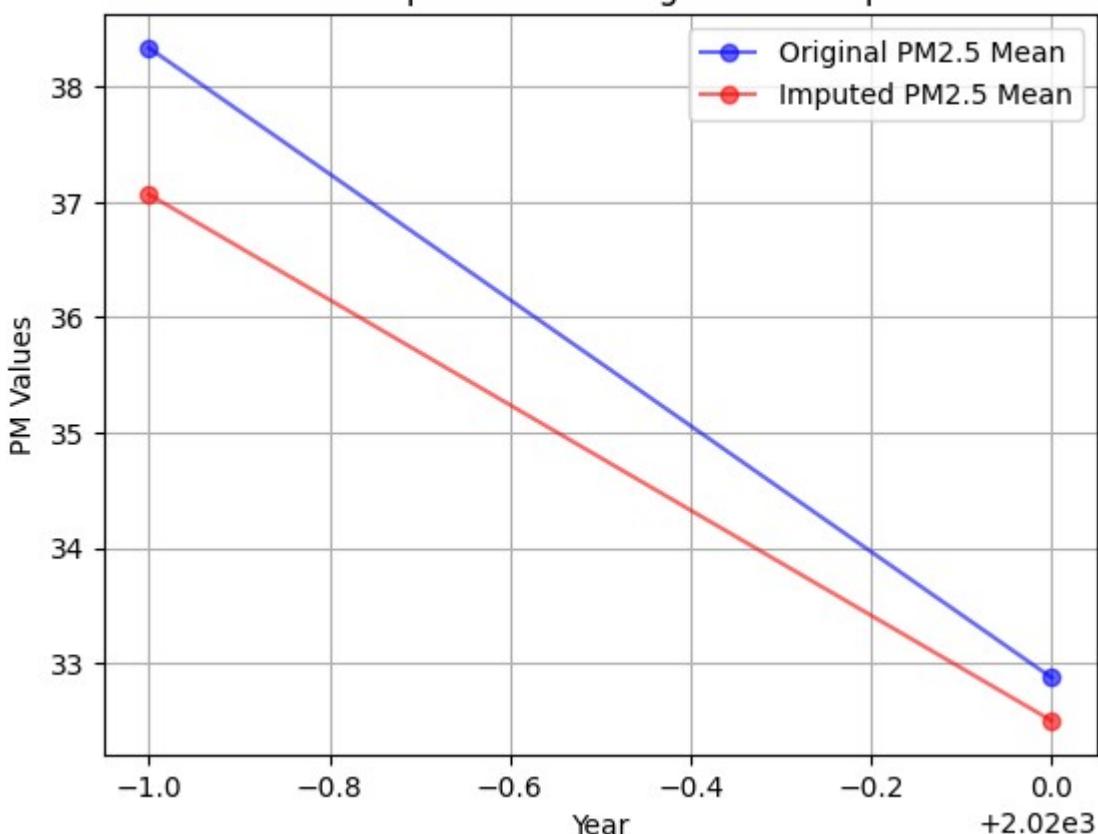


```
1 PM_imputation_stats_calc(Africa_or_df, Africa_K_fill, 'PM2.5')
2 PM_imputation_stats_calc(Africa_or_df, Africa_K_fill, 'PM10')
```

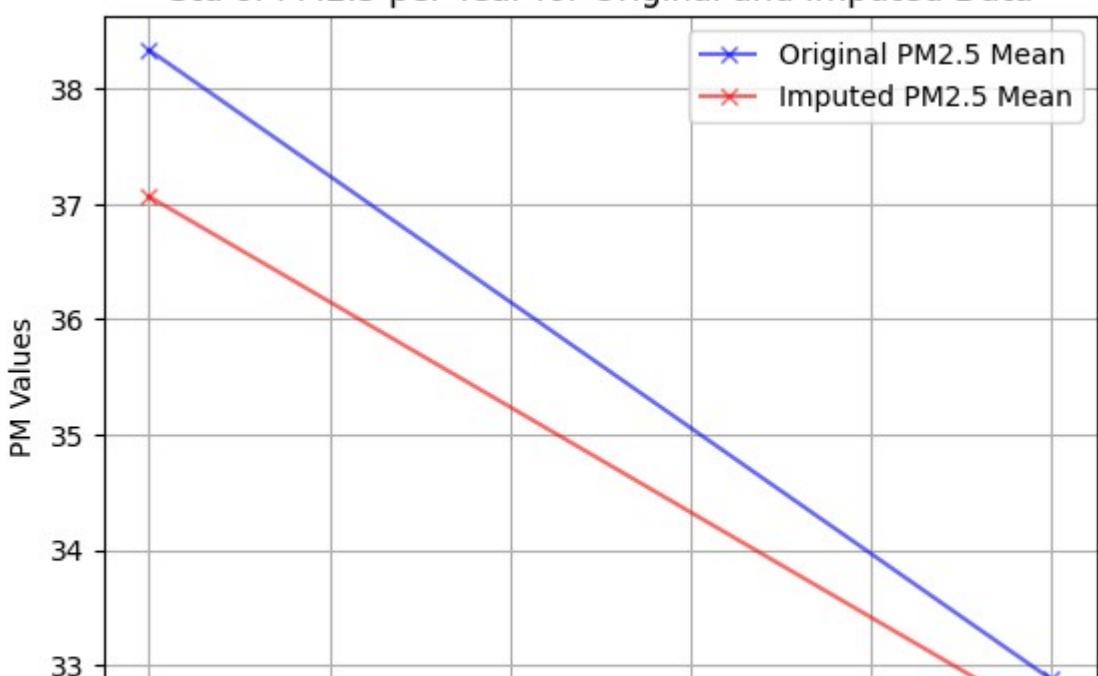
Percentage Change of PM2.5 Overall Mean = 1.1964887519314449

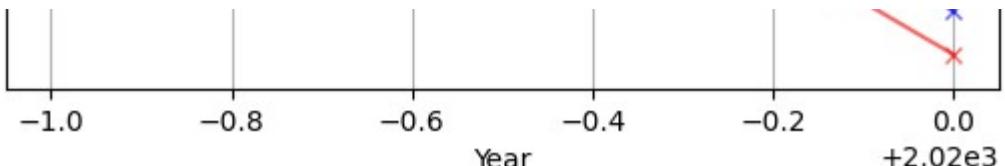
Percentage Change of PM2.5 Overall Std = 1.1644457091283062

Mean of PM2.5 per Year for Original and Imputed Data



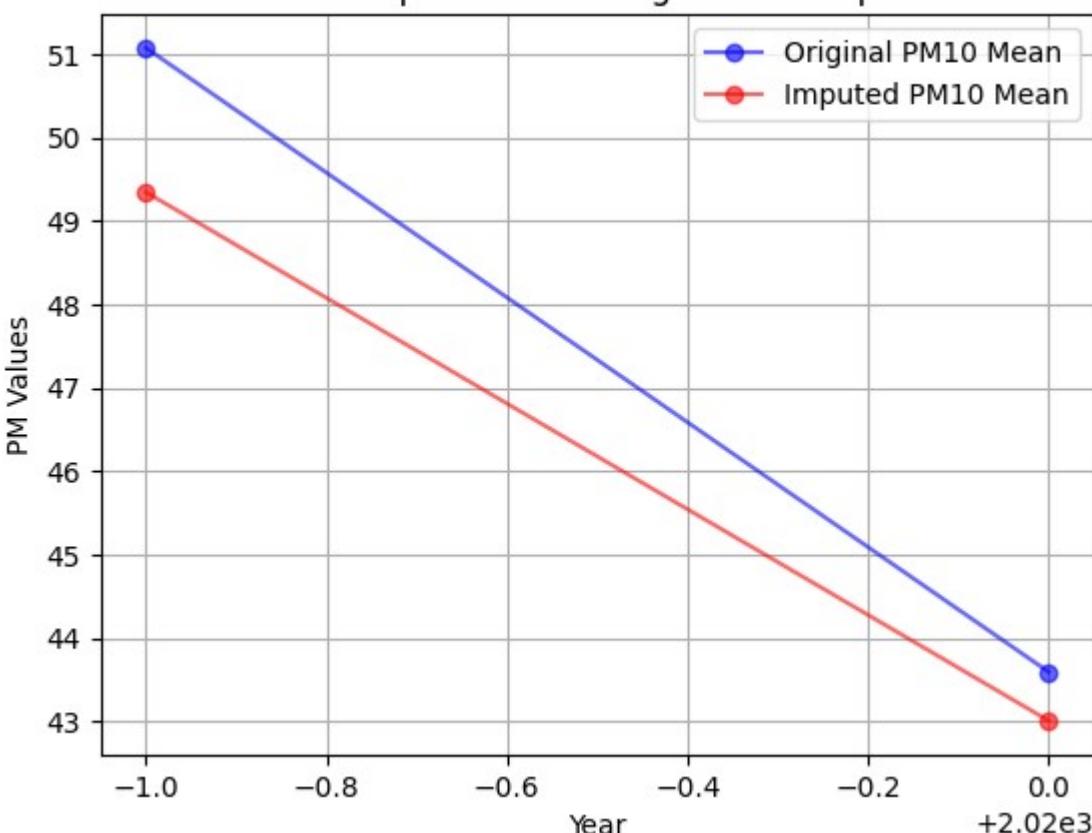
Std of PM2.5 per Year for Original and Imputed Data



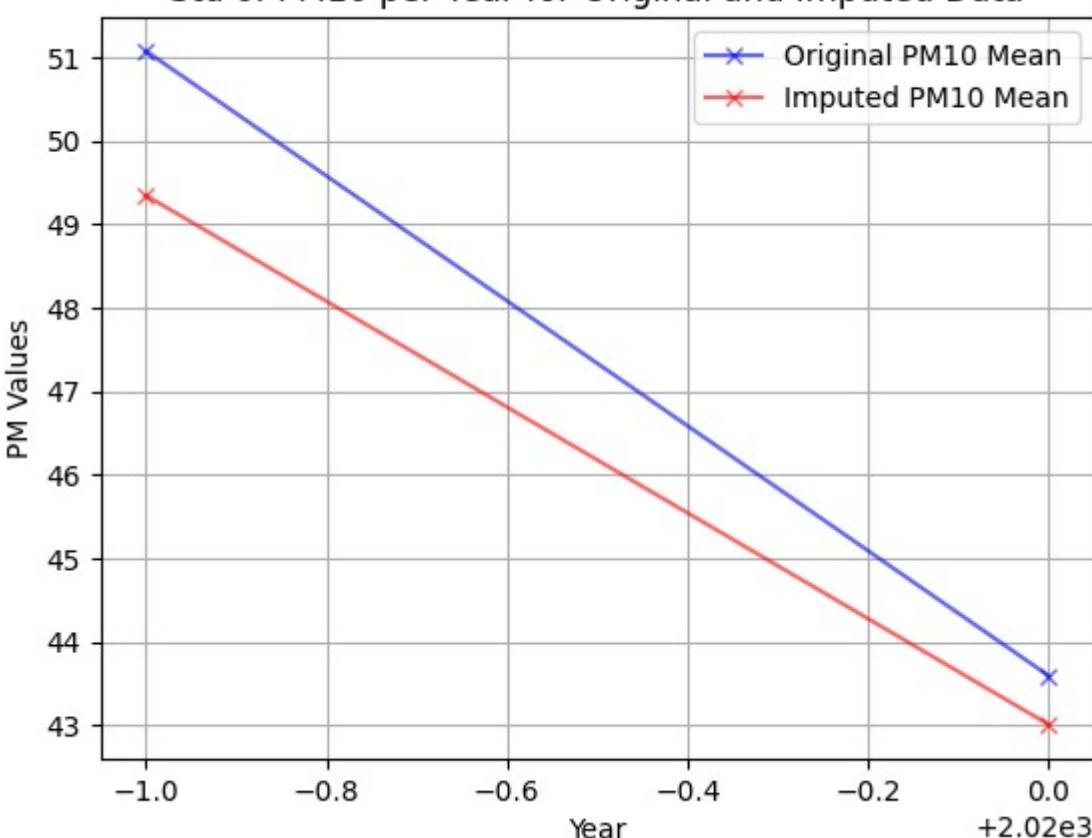


Percentage Change of PM10 Overall Mean = 1.3827971946344795
Percentage Change of PM10 Overall Std = 0.9716531546598578

Mean of PM10 per Year for Original and Imputed Data



Std of PM10 per Year for Original and Imputed Data

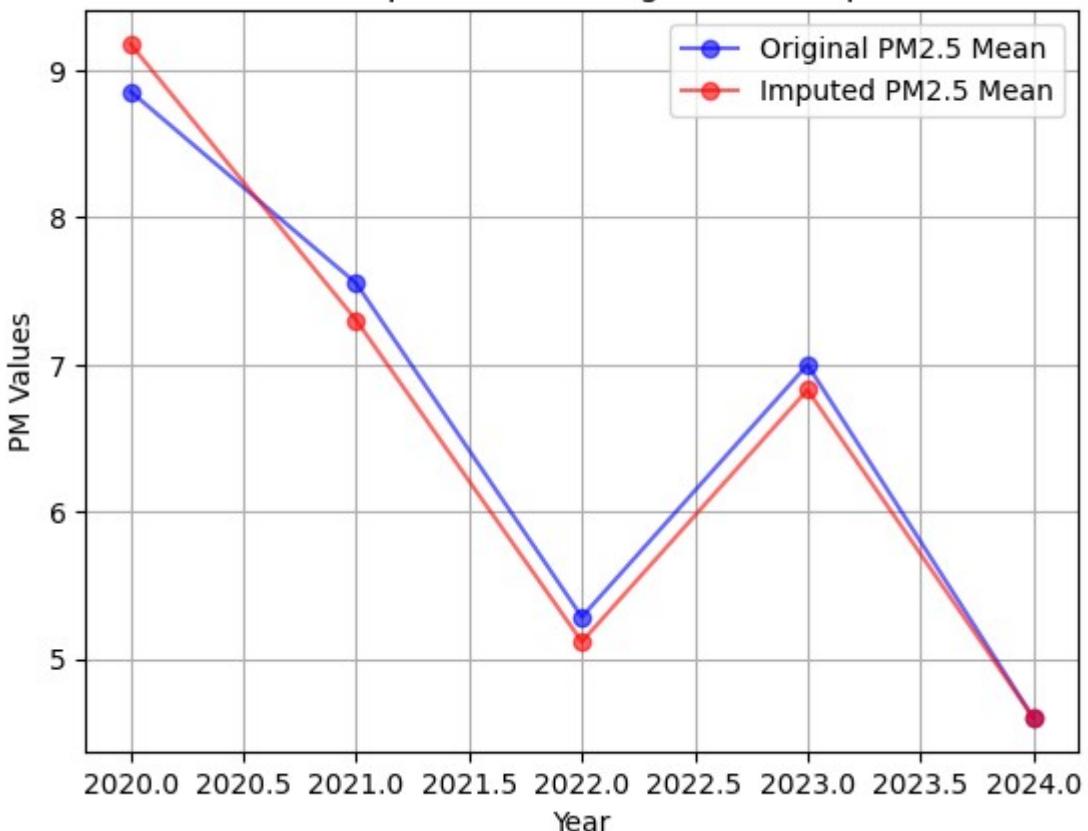


```
1 PM_imputation_stats_calc(Australia_or_df, Australia_K_fill, 'PM2.5')
2 PM_imputation_stats_calc(Australia_or_df, Australia_K_fill, 'PM10')
```

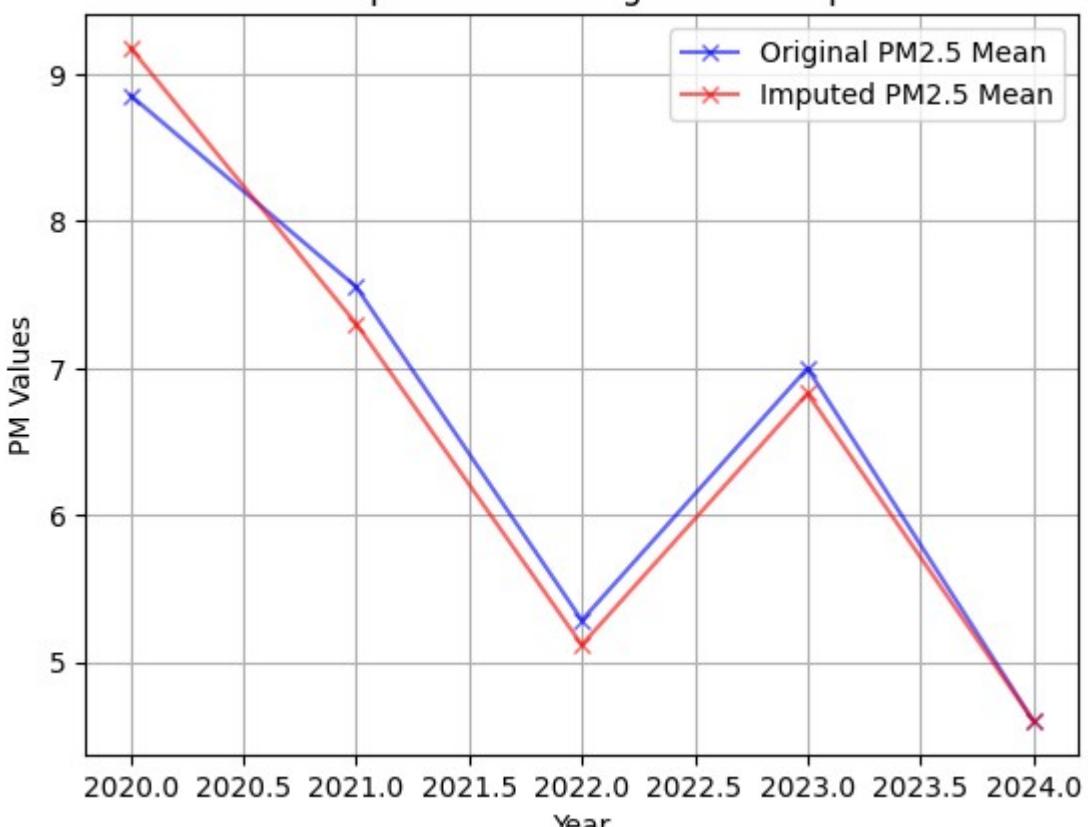
Percentage Change of PM2.5 Overall Mean = 0.42956348108133624

Percentage Change of PM2.5 Overall Std = 1.7468248260242891

Mean of PM2.5 per Year for Original and Imputed Data

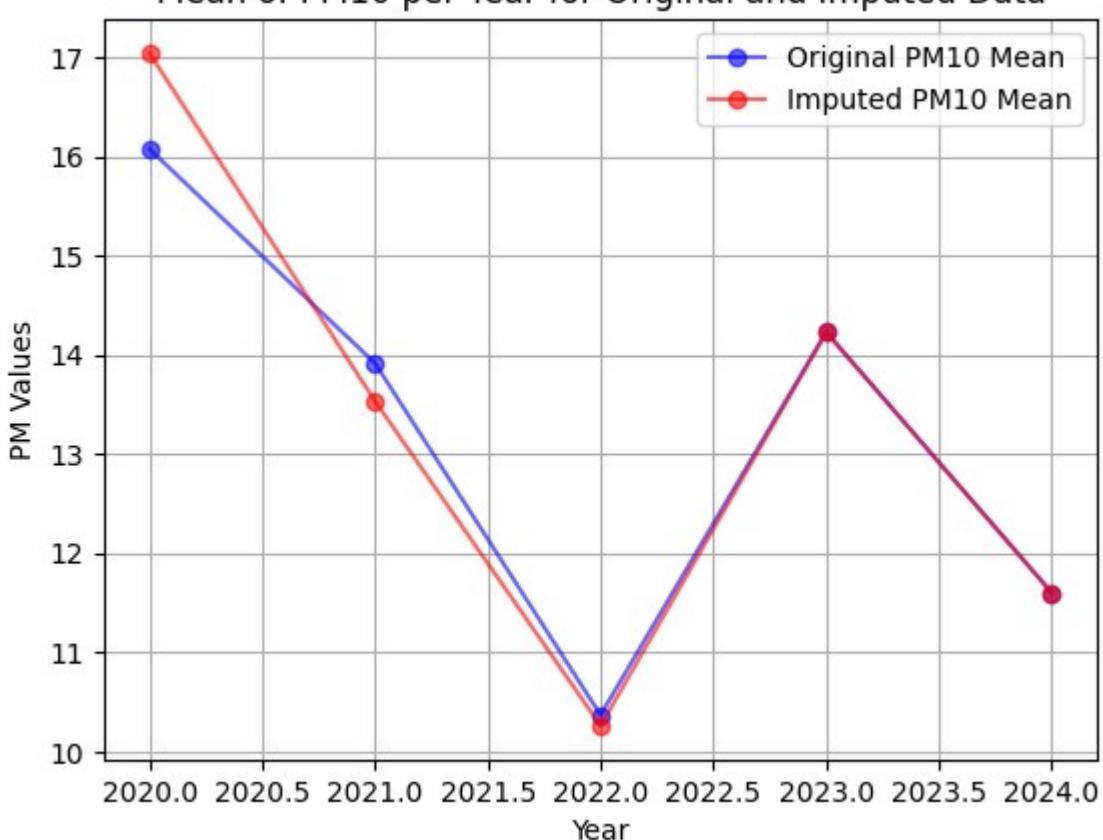


Std of PM2.5 per Year for Original and Imputed Data

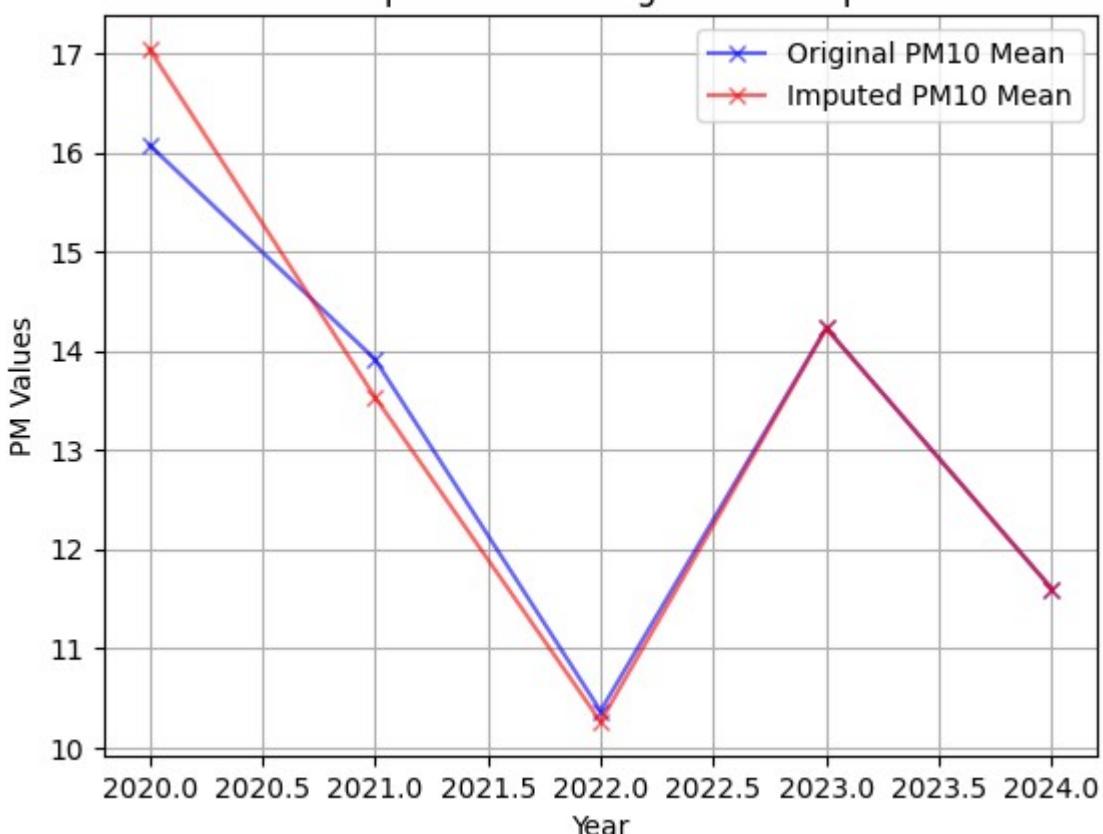


Percentage Change of PM10 Overall Mean = 1.3097138800643722
Percentage Change of PM10 Overall Std = 2.4864330336273626

Mean of PM10 per Year for Original and Imputed Data

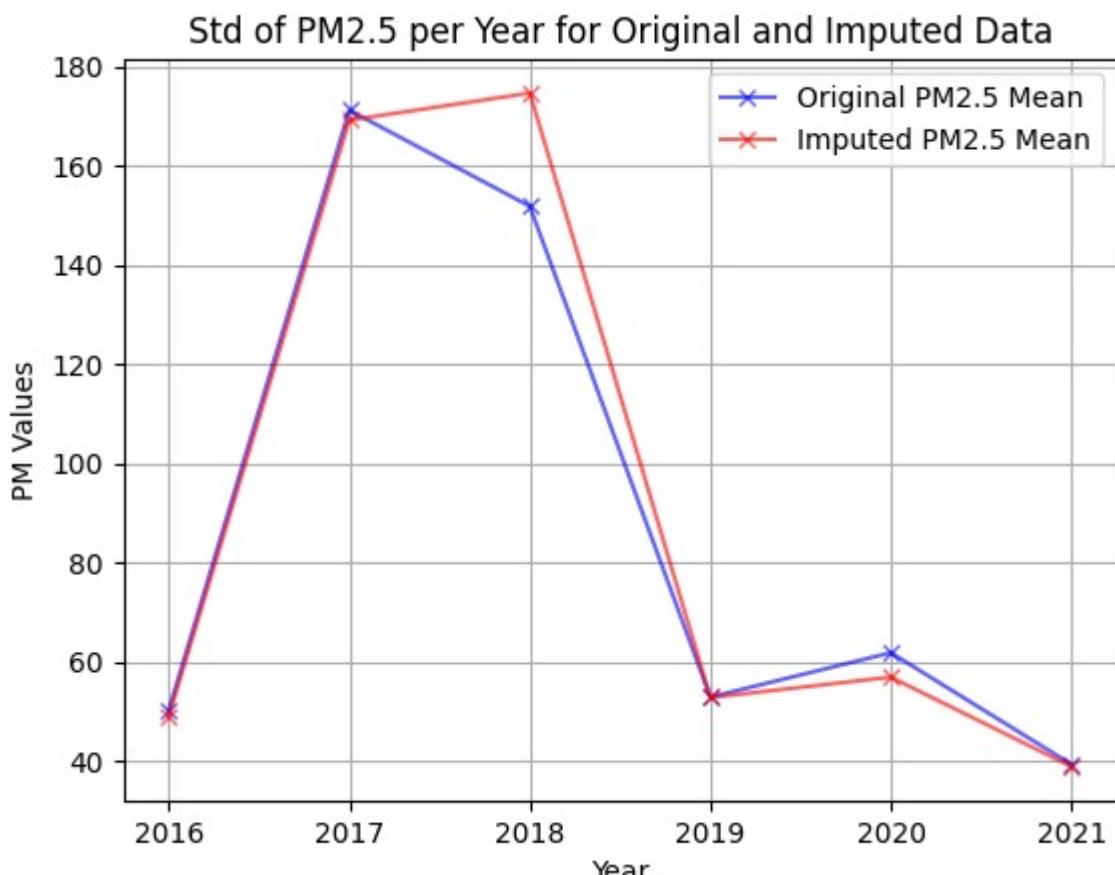
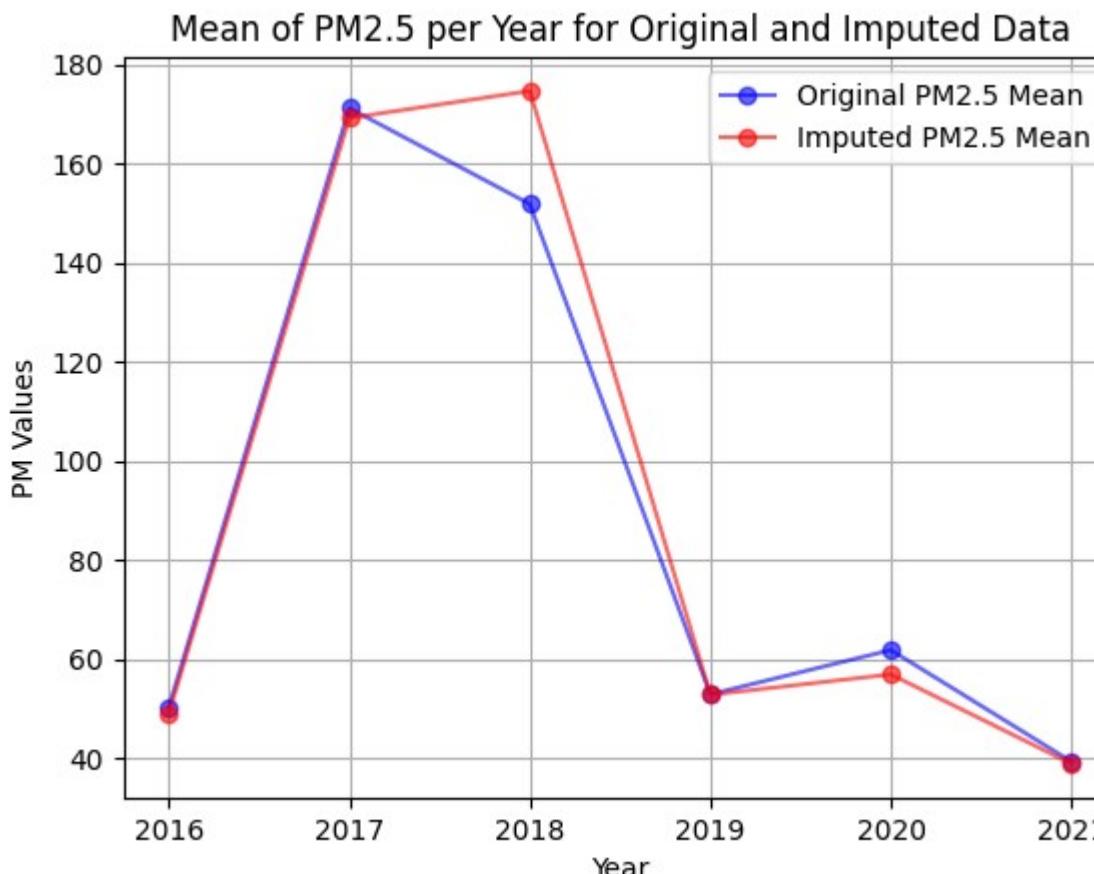


Std of PM10 per Year for Original and Imputed Data



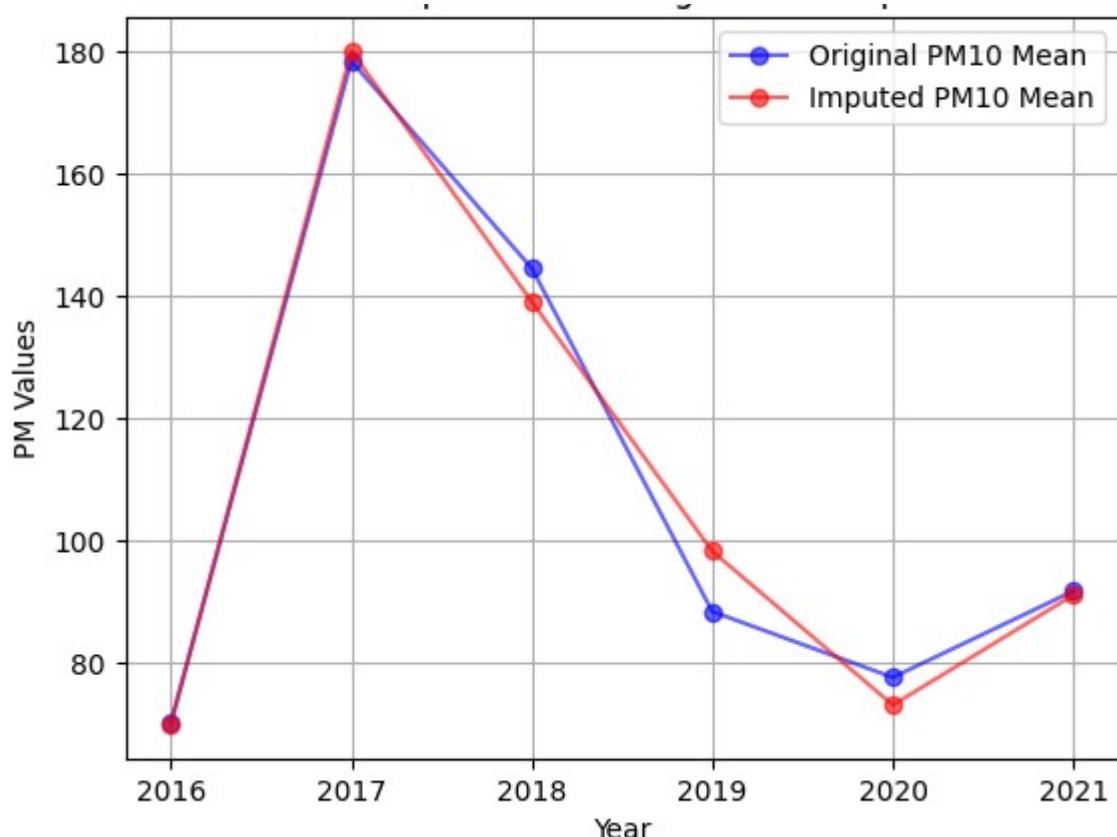
```
1 PM_imputation_stats_calc(Mexico_or_df, Mexico_K_fill, 'PM2.5')
2 PM_imputation stats calc(Mexico or df, Mexico K fill, 'PM10')
```

Percentage Change of PM2.5 Overall Mean = 2.184979860893057
Percentage Change of PM2.5 Overall Std = 0.9820005967992084

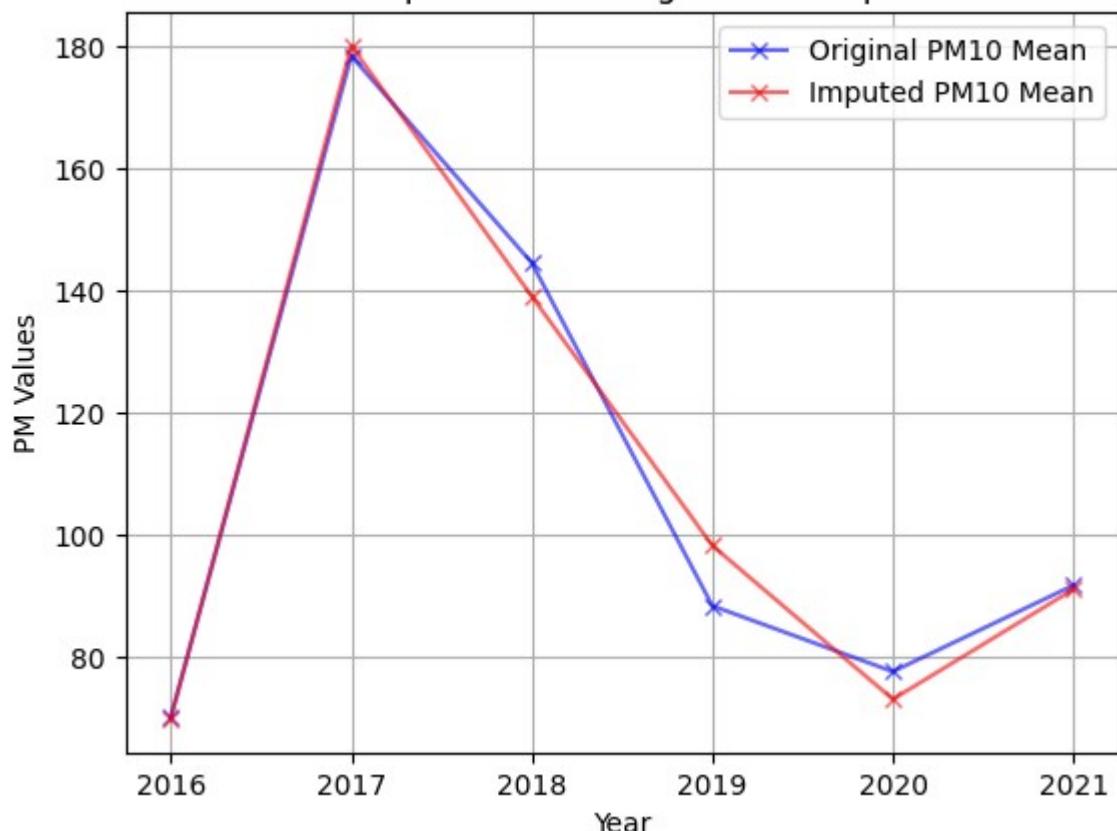


Percentage Change of PM10 Overall Mean = 0.7996667665438505
Percentage Change of PM10 Overall Std = 2.329818276887337

Mean of PM10 per Year for Original and Imputed Data



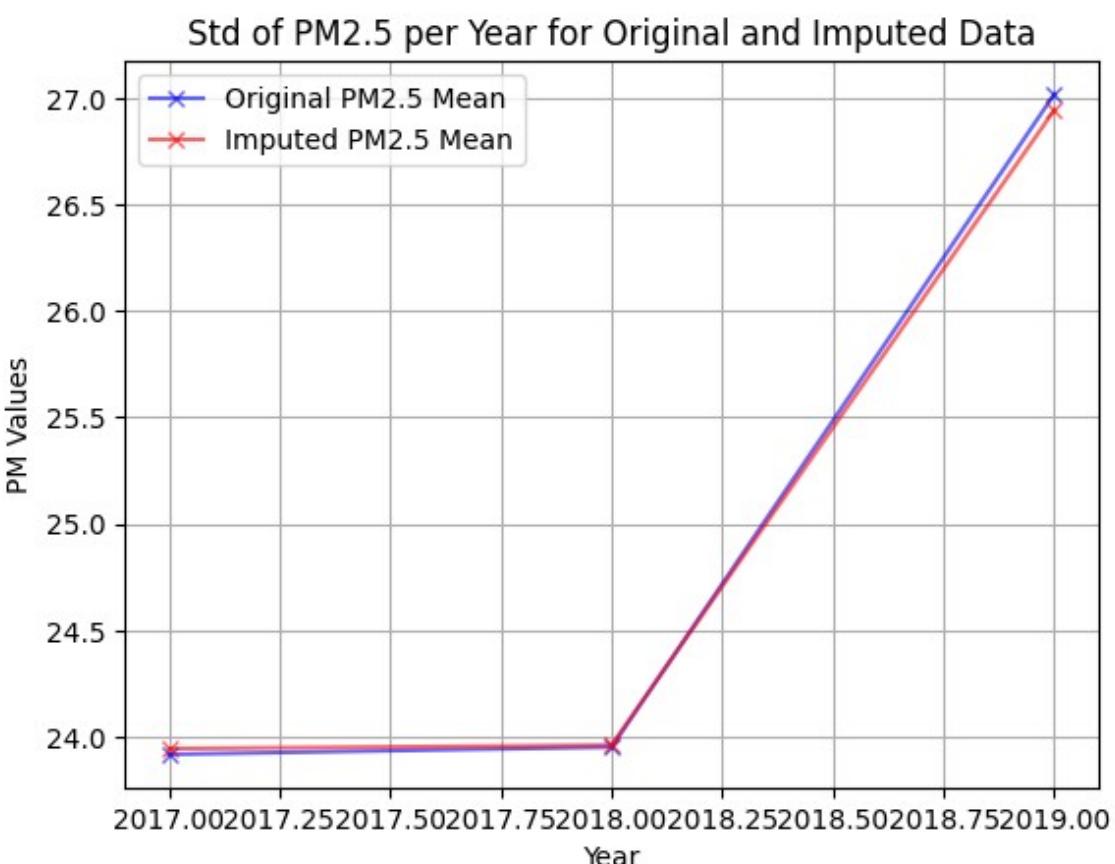
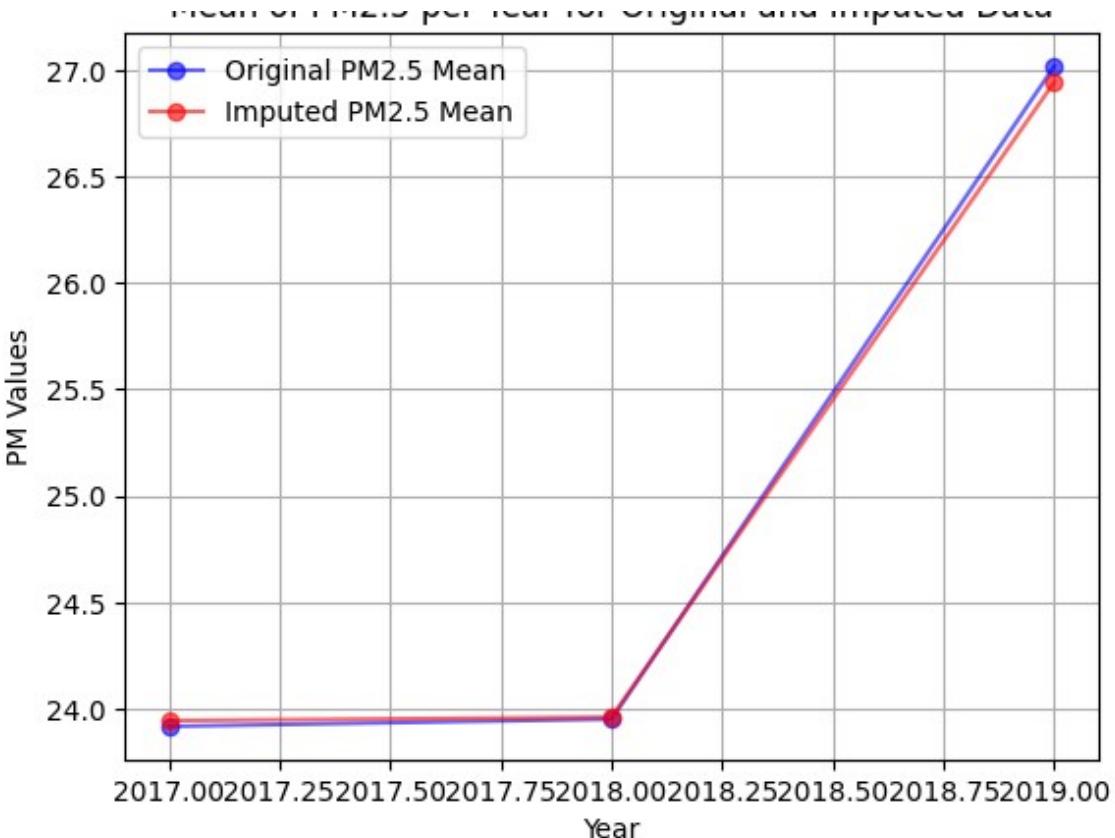
Std of PM10 per Year for Original and Imputed Data



```
1 PM_imputation_stats_calc(Korea_or_df, Korea_K_fill, 'PM2.5')
2 PM_imputation_stats_calc(Korea_or_df, Korea_K_fill, 'PM10')
```

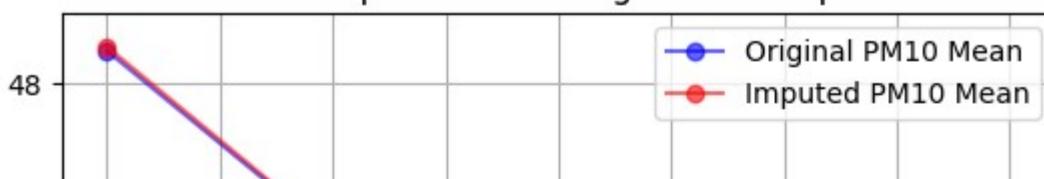
Percentage Change of PM2.5 Overall Mean = 0.07101317992289206
Percentage Change of PM2.5 Overall Std = 0.5712802022892692

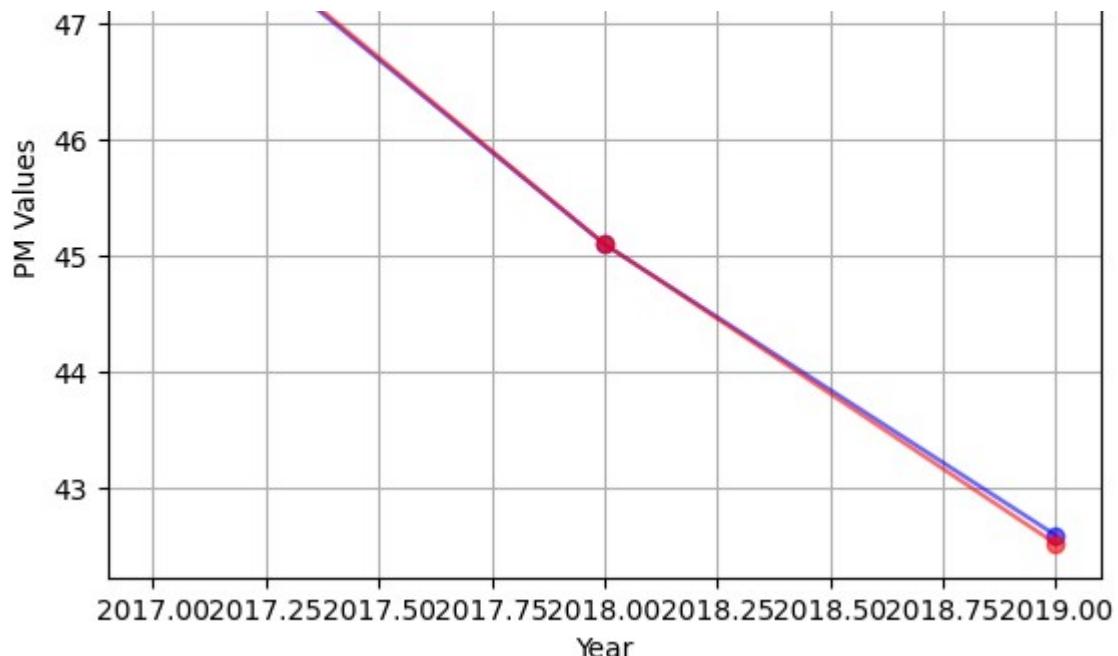
Mean of PM2.5 per Year for Original and Imputed Data



Percentage Change of PM10 Overall Mean = 0.11246764906690922
Percentage Change of PM10 Overall Std = 0.5719494283271769

Mean of PM10 per Year for Original and Imputed Data





Std of PM10 per Year for Original and Imputed Data

