Assignment #06- Python Pandas

Name: Soumyadeep Ganguly

Reg. No.: 24MDT0082

M.Sc. Data Science

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

Q1. Read the data file NutAverage.xlsx into a DataFrame and perform the following tasks:

```
In [2]: df = pd.read_excel('Pandas/NutAverage.xlsx')
        df.head()
Out[2]:
           Day Count
                       NH4-N
                                NO2-N NO3-N
                                                       TN
        0
                   1 6.915879 2.885372 7.457832 35.834969
                  47 6.344965 2.852123 5.696753 36.359106
        1
        2
                  78 4.964745 2.090747 2.167375 40.719987
        3
                 116 4.361492 2.301630 1.653266 24.931194
                 143 3.980372 1.419541 0.233538 36.234797
```

Find the null values, if any, in the data set and fill the null values with the method of your choice

Add a column DIN (stands for Dissolved Inorganic Nitrogen) to this DataFrame, where DIN = NH4-N+NO2-N+NO3-N.

```
In [4]: df.columns
```

```
Out[4]: Index(['Day Count', 'NH4-N', 'NO2-N', 'NO3-N', 'TN'], dtype='object')
In [5]: df['DIN'] = df['NH4-N'] + df['NO2-N'] + df['NO3-N']
df
```

Out[5]: -		Day Count	NH4-N	NO2-N	NO3-N	TN	DIN
	0	1	6.915879	2.885372	7.457832	35.834969	17.259083
	1	47	6.344965	2.852123	5.696753	36.359106	14.893841
	2	78	4.964745	2.090747	2.167375	40.719987	9.222866
	3	116	4.361492	2.301630	1.653266	24.931194	8.316388
	4	143	3.980372	1.419541	0.233538	36.234797	5.633451
	5	181	4.814007	1.416273	0.185584	36.269086	6.415864
	6	210	5.774826	2.250251	1.034297	27.557018	9.059374
	7	236	4.439287	1.977844	0.482635	45.557639	6.899766
	8	270	2.394753	2.289396	1.439277	28.102278	6.123425
	9	298	1.956891	1.685278	0.831846	30.017141	4.474016
	10	332	3.965562	2.330494	1.613106	31.488240	7.909163
	11	364	4.521629	2.305632	1.670597	41.095589	8.497858

Add another column DON (Dissolved Organic Nitrogen) to this DataFrame, where DON = TN - DIN.

```
In [6]: df['DON'] = df['TN'] - df['DIN']
df
```

Out[6]:		Day Count	NH4-N	NO2-N	NO3-N	TN	DIN	DON
	0	1	6.915879	2.885372	7.457832	35.834969	17.259083	18.575886
	1	47	6.344965	2.852123	5.696753	36.359106	14.893841	21.465266
	2	78	4.964745	2.090747	2.167375	40.719987	9.222866	31.497121
	3	116	4.361492	2.301630	1.653266	24.931194	8.316388	16.614806
	4	143	3.980372	1.419541	0.233538	36.234797	5.633451	30.601346
	5	181	4.814007	1.416273	0.185584	36.269086	6.415864	29.853222
	6	210	5.774826	2.250251	1.034297	27.557018	9.059374	18.497643
	7	236	4.439287	1.977844	0.482635	45.557639	6.899766	38.657873
	8	270	2.394753	2.289396	1.439277	28.102278	6.123425	21.978853
	9	298	1.956891	1.685278	0.831846	30.017141	4.474016	25.543126
	10	332	3.965562	2.330494	1.613106	31.488240	7.909163	23.579077
	11	364	4.521629	2.305632	1.670597	41.095589	8.497858	32.597731

Add a row named Averages which contains the averages of each of NH4-N, NO2-N, NO3-N, TN, DIN, DON

In [7]: df.loc['Average'] = [np.mean(df[i]) for i in df.columns] Out[7]: **Day Count** NH4-N NO2-N NO3-N TN DIN DON 0 1.000000 2.885372 7.457832 35.834969 17.259083 18.575886 6.915879 5.696753 47.000000 6.344965 2.852123 36.359106 14.893841 21.465266 2 78.000000 4.964745 2.090747 2.167375 40.719987 9.222866 31.497121 116.000000 4.361492 2.301630 1.653266 24.931194 8.316388 16.614806 143.000000 3.980372 1.419541 0.233538 36.234797 5.633451 30.601346 181.000000 4.814007 1.416273 0.185584 36.269086 6.415864 29.853222 210.000000 5.774826 2.250251 18.497643 1.034297 27.557018 9.059374 236.000000 4.439287 1.977844 0.482635 45.557639 6.899766 38.657873 270.000000 2.394753 2.289396 6.123425 21.978853 1.439277 28.102278 298.000000 1.956891 1.685278 0.831846 30.017141 4.474016 25.543126 332.000000 3.965562 2.330494 1.613106 31.488240 7.909163 23.579077 364.000000 4.521629 2.305632 1.670597 41.095589 8.497858 32.597731 4.536201 2.150382 2.038842 34.513920 8.725425 25.788496 189.666667

Describe characteristics of the DataFrame

in [8]:	<pre>df.describe()</pre>							
)ut[8]:		Day Count	NH4-N	NO2-N	NO3-N	TN	DIN	DON
	count	13.000000	13.000000	13.000000	13.000000	13.000000	13.000000	13.000000
	mean	189.666667	4.536201	2.150382	2.038842	34.513920	8.725425	25.788496
	std	110.867288	1.373646	0.453143	2.144153	5.956099	3.596171	6.530319
	min	1.000000	1.956891	1.416273	0.185584	24.931194	4.474016	16.614806
2	25%	116.000000	3.980372	1.977844	0.831846	30.017141	6.415864	21.465266
	50%	189.666667	4.521629	2.250251	1.613106	35.834969	8.316388	25.543126
	75%	270.000000	4.964745	2.305632	2.038842	36.359106	9.059374	30.601346
	max	364.000000	6.915879	2.885372	7.457832	45.557639	17.259083	38.657873

Plot all the data (except the Day Count column) using the area plot, line plot and box plot of DataFrame. (use Subplots where ever appropriate for better visualisation of the data)

```
In [9]: df.columns

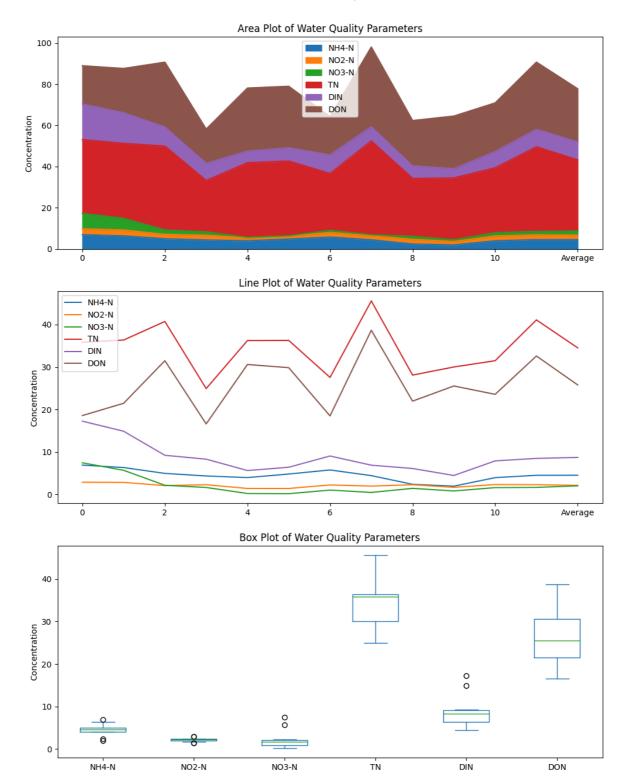
Out[9]: Index(['Day Count', 'NH4-N', 'NO2-N', 'NO3-N', 'TN', 'DIN', 'DON'], dtype='obje ct')

In [10]: df2 = df[['NH4-N', 'NO2-N', 'NO3-N', 'TN', 'DIN', 'DON']]
    fig, axs = plt.subplots(3, 1, figsize=(12, 16))
    df2.plot.area(ax=axs[0])
    axs[0].set_title('Area Plot of Water Quality Parameters')
    axs[0].set_ylabel('Concentration')

df2.plot(ax=axs[1])
    axs[1].set_title('Line Plot of Water Quality Parameters')
    axs[1].set_ylabel('Concentration')

df2.plot.box(ax=axs[2])
    axs[2].set_title('Box Plot of Water Quality Parameters')
    axs[2].set_ylabel('Concentration')
```

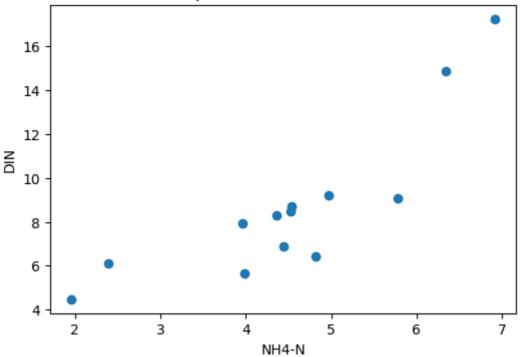
Out[10]: Text(0, 0.5, 'Concentration')



Compare the NH4-N vs DIN composition graphically using a scatter plot.

```
In [11]: fig, ax = plt.subplots(figsize=(6,4))
    ax.scatter(df['NH4-N'], df['DIN'])
    ax.set_xlabel('NH4-N')
    ax.set_ylabel('DIN')
    ax.set_title('Comparison of NH4-N and DIN')
    plt.show()
```

Comparison of NH4-N and DIN



Q2. Read the data file PhythoBiomass.xlsx into a DataFrame and perform the following tasks:

 <pre>df = pd.read_excel('Pandas/PythoBiomass.xlsx') df.head()</pre>	

Out[12]:		Days	Cyanophyceans	Chlorophyceae	Total Biomass
	0	1	0.554035	0.340955	4.775824
	1	47	0.409126	0.446749	4.536462
	2	78	0.606581	0.210896	4.131376
	3	116	0.308334	1.301525	3.597625
	4	143	0.828900	0.352965	2.867716

Find the null values, if any, in the data set and fill these null values with the method of your choice

```
Out[14]: Days 0
Cyanophyceans 0
Chlorophyceae 0
Total Biomass 0
dtype: int64
```

Add a column Others which list the biomass of other phytoplankton groups obtained by subtracting TotalBiomass with the sum of the biomass of Cyanophycean and Chlorophyceae.

```
In [15]: df = df.rename(columns={'Total Biomass':'Total Biomass'})
          df.columns
In [16]:
Out[16]: Index(['Days', 'Cyanophyceans', 'Chlorophyceae', 'Total Biomass'], dtype='objec
          t')
          df['Others'] = df['Total Biomass'] - (df['Cyanophyceans']-df['Chlorophyceae'])
Out[17]:
              Days
                    Cyanophyceans Chlorophyceae Total Biomass
                                                                     Others
           0
                  1
                           0.554035
                                           0.340955
                                                         4.775824 4.562744
           1
                 47
                           0.409126
                                           0.446749
                                                         4.536462 4.574084
           2
                 78
                           0.606581
                                           0.210896
                                                         4.131376 3.735692
           3
                116
                           0.308334
                                           1.301525
                                                         3.597625 4.590816
           4
                143
                           0.828900
                                           0.352965
                                                         2.867716 2.391781
           5
                181
                           0.822262
                                           1.327444
                                                         3.179547 3.684730
           6
               210
                                                         7.895237 5.760359
                           3.303263
                                           1.168384
           7
                236
                          35.462698
                                           0.739803
                                                        40.195265 5.472369
           8
               270
                                                         9.079613 5.735321
                           3.882161
                                           0.537869
           9
                298
                           0.575795
                                           0.399925
                                                          3.351554 3.175684
                                                         9.098784 9.145511
          10
                332
                           1.276101
                                           1.322828
          11
                364
                           1.127914
                                           0.344519
                                                        10.457813 9.674417
```

Describe the characteristics of the DataFrame.

```
In [18]: df.describe()
```

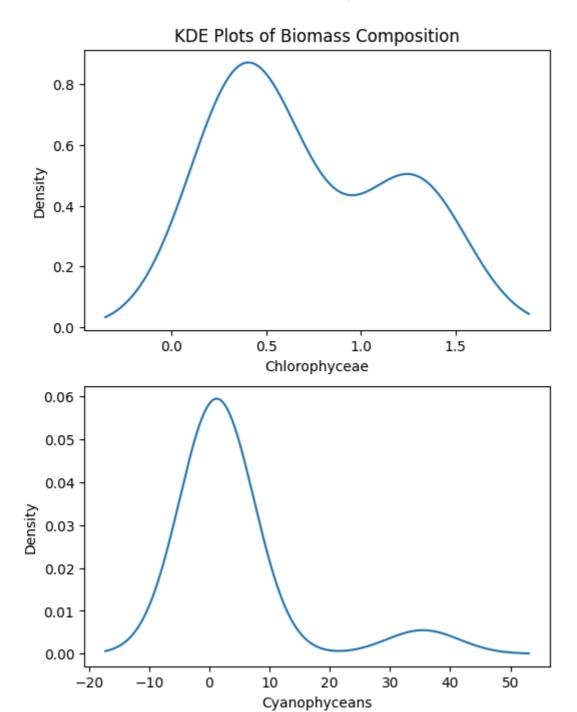
Out[18]:		Days	Cyanophyceans	Chlorophyceae	Total Biomass	Others
	count	12.000000	12.000000	12.000000	12.000000	12.000000
	mean	189.666667	4.096431	0.707822	8.597235	5.208626
	std	115.797106	9.944160	0.443034	10.307117	2.212660
	min	1.000000	0.308334	0.210896	2.867716	2.391781
	25%	106.500000	0.570355	0.350853	3.536107	3.722951
	50%	195.500000	0.825581	0.492309	4.656143	4.582450
	75%	277.000000	1.782892	1.201670	9.084406	5.741581
	max	364.000000	35.462698	1.327444	40.195265	9.674417

Plot the biomass composition of each group using a barh and kde plot.

KDE Plots

```
In [19]: fig, axs = plt.subplots(2,1, figsize=(6, 8))
    df['Chlorophyceae'].plot.kde(ax=axs[0])
    axs[0].set_xlabel('Chlorophyceae')
    df['Cyanophyceans'].plot.kde(ax=axs[1])
    axs[1].set_xlabel('Cyanophyceans')
    axs[0].set_title("KDE Plots of Biomass Composition")
```

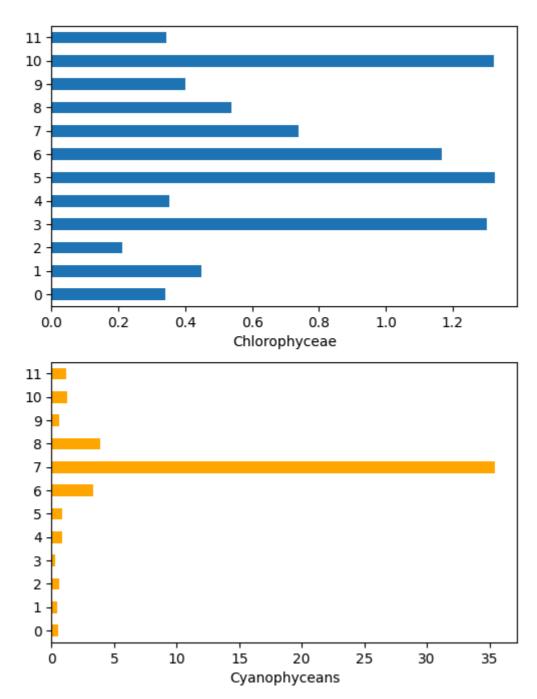
Out[19]: Text(0.5, 1.0, 'KDE Plots of Biomass Composition')



Barh Plots

```
In [20]: fig, axs = plt.subplots(2,1, figsize=(6, 8))
    df['Chlorophyceae'].plot.barh(ax=axs[0], x=np.arange(len(df['Chlorophyceae'])))
    axs[0].set_xlabel('Chlorophyceae')
    df['Cyanophyceans'].plot.barh(ax=axs[1], x=np.arange(len(df['Cyanophyceans'])),
    axs[1].set_xlabel('Cyanophyceans')
```

Out[20]: Text(0.5, 0, 'Cyanophyceans')



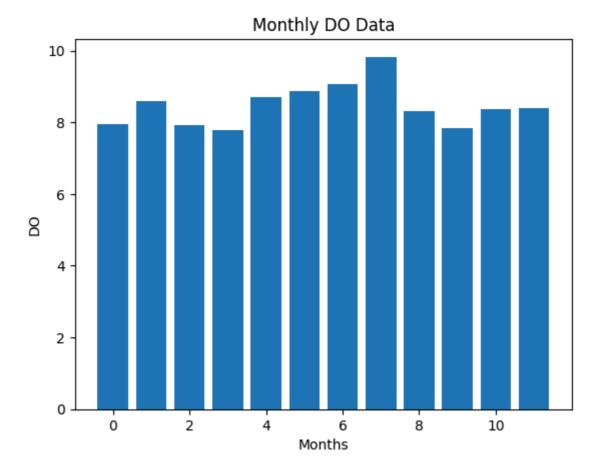
Q3. Read the data file DOData.csv into a DataFrame and perform the following tasks:

```
In [21]: df = pd.read_csv('Pandas/DOData.csv')
    df2 = pd.read_excel('Pandas/NutAverage.xlsx')
    df
```

Out[21]:		Days	DO
	0	1	7.96
	1	47	8.60
	2	78	7.92
	3	116	7.78
	4	143	8.70
	5	181	8.87
	6	210	9.07
	7	236	9.83
	8	270	8.32
	9	298	7.85
	10	332	8.38
	11	365	8.40

Plot the monthly DO data using a bar plot.

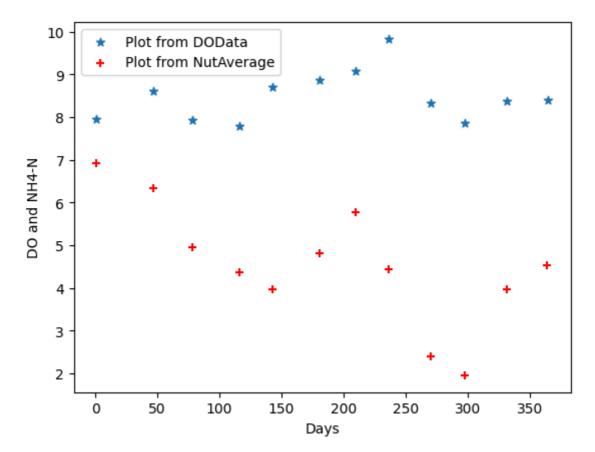
```
In [22]: plt.bar(np.arange(len(df['Days'])), df['DO'])
   plt.title('Monthly DO Data')
   plt.xlabel('Months')
   plt.ylabel('DO')
   plt.show()
```



Plot the monthly DO vs NH4 (from NutAverage.xlsx) as a scatter plot.

```
In [23]: fig, axs = plt.subplots()
   axs.scatter(df['Days'],df['DO'], label="Plot from DOData", marker='*')
   axs.scatter(df2['Day Count'],df2['NH4-N'], color="red", label="Plot from NutAver
   axs.set_ylabel('DO and NH4-N')
   axs.set_xlabel('Days')
   axs.legend()
```

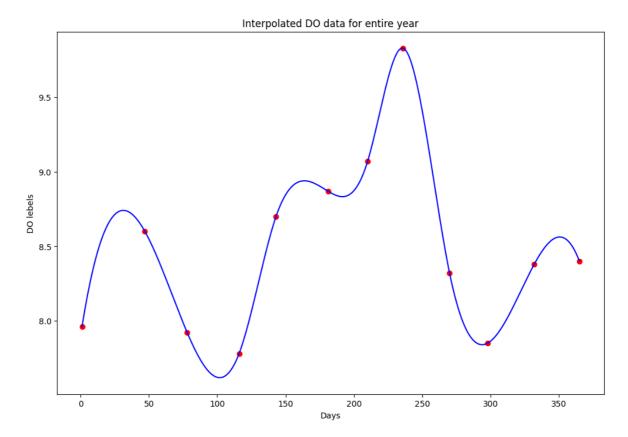
Out[23]: <matplotlib.legend.Legend at 0x2006705f020>



Construct an interpolating polynomial to estimate the DO for the entire year starting from day 1 to day 365. Visualise the interpolated and monthly data (with monthly data plotted as points).

```
In [32]: from scipy.interpolate import interp1d

In [38]: ip = interp1d(df['Days'], df['DO'], kind='cubic')
    fig, axs = plt.subplots(figsize=(12,8))
    days = np.arange(1,366)
    axs.plot(df['Days'], df['DO'], 'or', label='Monthly Data')
    axs.plot(days, ip(days), '-b', label='Interpolated Data')
    axs.set_title('Interpolated DO data for entire year')
    axs.set_xlabel('Days')
    axs.set_ylabel('DO lebels')
    plt.show()
```



In []: