DAY 9:

Bottle Dataset

In [1]:

```
#to import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [2]:

```
df=pd.read_csv(r"E:\Dataset\9_bottle.csv")[0:500]
df
```

C:\ProgramData\Anaconda3\lib\site-packages\IPython\core\interactiveshell.p
y:3165: DtypeWarning: Columns (47,73) have mixed types.Specify dtype optio
n on import or set low_memory=False.
has_raised = await self.run_ast_nodes(code_ast.body, cell_name,

Out[2]:

	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	Salnty	O2ml_L	STheta	O2Sat
0	1	1	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0000A-3	0	10.50	33.440	NaN	25.649	NaN
1	1	2	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0008A-3	8	10.46	33.440	NaN	25.656	NaN
2	1	3	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0010A-7	10	10.46	33.437	NaN	25.654	NaN
3	1	4	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0019A-3	19	10.45	33.420	NaN	25.643	NaN
4	1	5	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0020A-7	20	10.45	33.421	NaN	25.643	NaN
495	16	496	063.3 058.0	19- 4903CR- HY-065- 1030- 06330580- 0700A-7	700	4.90	34.269	NaN	27.114	NaN
496	16	497	063.3 058.0	19- 4903CR- HY-065- 1030- 06330580- 0792A-3	792	4.50	34.310	NaN	27.191	NaN
497	16	498	063.3 058.0	19- 4903CR- HY-065- 1030- 06330580- 0800A-7	800	4.48	34.311	NaN	27.194	NaN
498	16	499	063.3 058.0	19- 4903CR- HY-065- 1030- 06330580- 0900A-7	900	4.21	34.319	NaN	27.230	NaN

In [30st_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	Salnty	O2ml_L	STheta	O2Sat
df.info()									
499 16	500	063.3 058.0	19- 4903CR- HY-065- 1030- 06330580- 1000A-7	1000	3.95	34.329	NaN	27.265	NaN

500 rows × 74 columns

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 500 entries, 0 to 499
Data columns (total 74 columns):

Data	columns (total 74	columns):	
#	Column	Non-Null Count	Dtype
0	Cst_Cnt	500 non-null	int64
1	Btl_Cnt	500 non-null	int64
2	Sta_ID	500 non-null	object
3	Depth_ID	500 non-null	object
4	Depthm	500 non-null	int64
	•		
5	T_degC	499 non-null	float64
6	Salnty	494 non-null	float64
7	02ml_L	0 non-null	float64
8	STheta	493 non-null	float64
9	02Sat	0 non-null	float64
10	Oxy_μmol/Kg	0 non-null	float64
11	BtlNum	0 non-null	float64
12	RecInd	500 non-null	int64
13	T_prec	499 non-null	float64
14	T_qual	4 non-null	float64
15	S_prec	494 non-null	float64
16	S_qual	10 non-null	float64
17	P_qual	500 non-null	float64
18	O_qual	500 non-null	float64
19	SThtaq	14 non-null	float64
20	02Satq	500 non-null	float64
21	ChlorA	0 non-null	float64
22	Chlqua	500 non-null	float64
23	Phaeop	0 non-null	float64
24	•	500 non-null	float64
	Phaqua		
25	PO4uM	0 non-null	float64
26	PO4q	500 non-null	float64
27	SiO3uM	0 non-null	float64
28	Si03qu	500 non-null	float64
29	NO2uM	0 non-null	float64
30	NO2q	500 non-null	float64
31	NO3uM	0 non-null	float64
32	NO3q	500 non-null	float64
33	NH3uM	0 non-null	float64
34	NH3q	500 non-null	float64
35	C14As1	0 non-null	float64
36	C14A1p	0 non-null	float64
37	C14A1q	500 non-null	float64
38	C14As2	0 non-null	float64
39	C14A2p	0 non-null	float64
40	C14A2q	500 non-null	float64
41	DarkAs	0 non-null	float64
42	DarkAp	0 non-null	float64
43	DarkAq	500 non-null	float64
44	MeanAs	0 non-null	float64
45	MeanAp	0 non-null	float64
46	MeanAq	500 non-null	float64
	•		
47 49	IncTim	0 non-null	object
48	LightP	0 non-null	float64
49	R_Depth	500 non-null	float64
50	R_TEMP	499 non-null	float64
51	R_POTEMP	495 non-null	float64
52	R_SALINITY	494 non-null	float64
53	R_SIGMA	486 non-null	float64
54	R_SVA	486 non-null	float64
55	R_DYNHT	500 non-null	float64

```
float64
56 R 02
                          0 non-null
 57
    R 02Sat
                          0 non-null
                                           float64
                          0 non-null
                                           float64
58
    R SIO3
59
    R P04
                          0 non-null
                                          float64
    R NO3
                          0 non-null
                                          float64
    R_N02
                          0 non-null
                                          float64
61
    R NH4
                          0 non-null
                                          float64
    R_CHLA
                          0 non-null
                                          float64
63
    R PHAEO
                          0 non-null
                                          float64
    R PRES
                          500 non-null
                                          int64
65
    R SAMP
                          0 non-null
                                          float64
66
    DIC1
                          0 non-null
                                          float64
67
                                          float64
68
    DIC2
                          0 non-null
                                          float64
69
    TA1
                          0 non-null
70
    TA2
                          0 non-null
                                          float64
                                          float64
71
    pH2
                          0 non-null
72 pH1
                          0 non-null
                                          float64
73 DIC Quality Comment 0 non-null
                                          object
dtypes: float64(65), int64(5), object(4)
```

memory usage: 289.2+ KB

In [4]:

```
df.columns
```

Out[4]:

```
Index(['Cst_Cnt', 'Btl_Cnt', 'Sta_ID', 'Depth_ID', 'Depthm', 'T_degC',
         'Salnty', 'O2ml_L', 'STheta', 'O2Sat', 'Oxy_µmol/Kg', 'BtlNum', 'RecInd', 'T_prec', 'T_qual', 'S_prec', 'S_qual', 'P_qual', 'O_qua
1',
         'SThtag', 'O2Satg', 'ChlorA', 'Chlqua', 'Phaeop', 'Phaqua', 'PO4u
Μ',
         'PO4q', 'SiO3uM', 'SiO3qu', 'NO2uM', 'NO2q', 'NO3uM', 'NO3q', 'NH3u
Μ',
         'NH3q', 'C14As1', 'C14A1p', 'C14A1q', 'C14As2', 'C14A2p', 'C14A2q',
         'DarkAs', 'DarkAp', 'DarkAq', 'MeanAs', 'MeanAp', 'MeanAq', 'IncTi
m',
         'LightP', 'R_Depth', 'R_TEMP', 'R_POTEMP', 'R_SALINITY', 'R_SIGMA',
        'R_SVA', 'R_DYNHT', 'R_O2', 'R_O2Sat', 'R_SIO3', 'R_PO4', 'R_NO3', 'R_NO2', 'R_NH4', 'R_CHLA', 'R_PHAEO', 'R_PRES', 'R_SAMP', 'DIC1',
         'DIC2', 'TA1', 'TA2', 'pH2', 'pH1', 'DIC Quality Comment'],
       dtvpe='object')
```

Linear Regression

```
In [5]:
```

```
x=df[['Cst_Cnt', 'Btl_Cnt', 'Depthm', 'RecInd', 'P_qual', 'O_qual', 'O2Satq']]
y=df[ 'R_PRES']
```

```
In [6]:
```

```
# to split my dataset into test and train data
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
```

In [7]:

```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

Out[7]:

LinearRegression()

In [8]:

```
print(lr.score(x_test,y_test))
```

0.9999989086841241

In [9]:

```
lr.score(x_train,y_train)
```

Out[9]:

0.9999989178354033

Ridge Regression

```
In [10]:
```

```
from sklearn.linear_model import Ridge,Lasso
```

In [11]:

```
rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
rr.score(x_test,y_test)
```

Out[11]:

0.9999987917054648

Lasso Regression

```
In [12]:
```

```
la=Lasso(alpha=10)
la.fit(x_train,y_train)
```

Out[12]:

Lasso(alpha=10)

In [13]:
<pre>la.score(x_test,y_test)</pre>
Out[13]:
0.999986153960675
In []: