Deena 20104016

In [3]: import pandas as pd
import numpy as np
import matplotlib.pyplot as pp

In [5]: df = pd.read_csv("9_bottle.csv")

C:\ProgramData\Anaconda3\lib\site-packages\IPython\core\interactiveshell.py:3
165: DtypeWarning: Columns (47,73) have mixed types.Specify dtype option on i
mport or set low_memory=False.

has_raised = await self.run_ast_nodes(code_ast.body, cell_name,

Out[5]:

	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	Salnty	O2n
0	1	1	054.0 056.0	19-4903CR- HY-060-0930-05400560-0000A-3	0	10.500	33.4400	
1	1	2	054.0 056.0	19-4903CR- HY-060-0930-05400560-0008A-3	8	10.460	33.4400	
2	1	3	054.0 056.0	19-4903CR- HY-060-0930-05400560-0010A-7	10	10.460	33.4370	
3	1	4	054.0 056.0	19-4903CR- HY-060-0930-05400560-0019A-3	19	10.450	33.4200	
4	1	5	054.0	19-4903CR-	20	10 450	33 4210	

In [6]:

Out[6]:

	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	Salnty	O2ml_L	s
0	1	1	054.0 056.0	19-4903CR- HY-060-0930-05400560-0000A-3	0	10.50	33.440	NaN	2
1	1	2	054.0 056.0	19-4903CR- HY-060-0930-05400560-0008A-3	8	10.46	33.440	NaN	2
2	1	3	054.0 056.0	19-4903CR- HY-060-0930-05400560-0010A-7	10	10.46	33.437	NaN	2
3	1	4	054.0 056.0	19-4903CR- HY-060-0930-05400560-0019A-3	19	10.45	33.420	NaN	2
4	1	5	054.0 056.0	19-4903CR- HY-060-0930-05400560-0020A-7	20	10.45	33.421	NaN	2

5 rows × 74 columns

In [7]:

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 864863 entries, 0 to 864862
Data columns (total 74 columns):

Data	columns (total 74	ł columns):	
#	Column	Non-Null Count	Dtype
0	Cst_Cnt	864863 non-null	int64
1	Btl_Cnt	864863 non-null	int64
2	Sta_ID	864863 non-null	object
3	Depth_ID	864863 non-null	object
4	Depthm	864863 non-null	int64
5	T_degC	853900 non-null	float64
6	Salnty	817509 non-null	float64
7	02ml_L	696201 non-null	float64
8	STheta	812174 non-null	float64
9	02Sat	661274 non-null	float64
10	Oxy_μmol/Kg	661268 non-null	float64
11	Bt1Num	118667 non-null	float64
12	RecInd	864863 non-null	int64
13	T_prec	853900 non-null	float64
14	T_qual	23127 non-null	float64
15	S_prec	817509 non-null	float64
16	S_qual	74914 non-null	float64
17	P_qual	673755 non-null	float64
18	0 qual	184676 non-null	float64
19	SThtaq	65823 non-null	float64
20	02Satq	217797 non-null	float64
21	ChlorA	225272 non-null	float64
22	Chlqua	639166 non-null	float64
23	Phaeop	225271 non-null	float64
24	Phaqua	639170 non-null	float64
25	PO4uM	413317 non-null	float64
26	PO4q	451786 non-null	float64
27	SiO3uM	354091 non-null	float64
28	SiO3qu	510866 non-null	float64
28 29	NO2uM	337576 non-null	float64
30		529474 non-null	float64
31	NO2q NO3uM	337403 non-null	float64
32	NO3q NH3uM	529933 non-null 64962 non-null	float64
33	NH3q	808299 non-null	float64
34 25	•		float64
35	C14As1	14432 non-null	float64
36	C14A1p	12760 non-null	float64
37	C14A1q	848605 non-null	float64
38	C14As2	14414 non-null	float64
39	C14A2p	12742 non-null	float64
40	C14A2q	848623 non-null	float64
41	DarkAs	22649 non-null	float64
42	DarkAp	20457 non-null	float64
43	DarkAq	840440 non-null	float64
44	MeanAs	22650 non-null	float64
45	MeanAp	20457 non-null	float64
46	MeanAq	840439 non-null	float64
47	IncTim	14437 non-null	object
48	LightP	18651 non-null	float64

In [8]:

```
49
   R_Depth
                         864863 non-null float64
50 R_TEMP
                         853900 non-null float64
                         818816 non-null float64
51 R POTEMP
52 R_SALINITY
                         817509 non-null float64
53
    R_SIGMA
                         812007 non-null float64
                         812092 non-null float64
54
    R SVA
55
    R_DYNHT
                         818206 non-null float64
                         696201 non-null float64
56
    R_02
57
                         666448 non-null float64
    R 02Sat
                         354099 non-null float64
58
    R_SIO3
59
                         413325 non-null float64
    R_P04
60
    R NO3
                         337411 non-null float64
                         337584 non-null float64
61
    R NO2
    R_NH4
                         64982 non-null
                                          float64
                         225276 non-null float64
63
    R CHLA
                         225275 non-null float64
64
   R PHAEO
                         864863 non-null int64
65
    R_PRES
                         122006 non-null float64
66
    R_SAMP
67
    DIC1
                         1999 non-null
                                          float64
68
    DIC2
                         224 non-null
                                          float64
69
    TA1
                         2084 non-null
                                          float64
70
    TA2
                         234 non-null
                                          float64
71
    pH2
                         10 non-null
                                          float64
                         84 non-null
                                          float64
72 pH1
    DIC Quality Comment 55 non-null
                                          object
dtypes: float64(65), int64(5), object(4)
memory usage: 488.3+ MB
```

In [12]: df1=df.dropna(axis=1)

Out[12]:

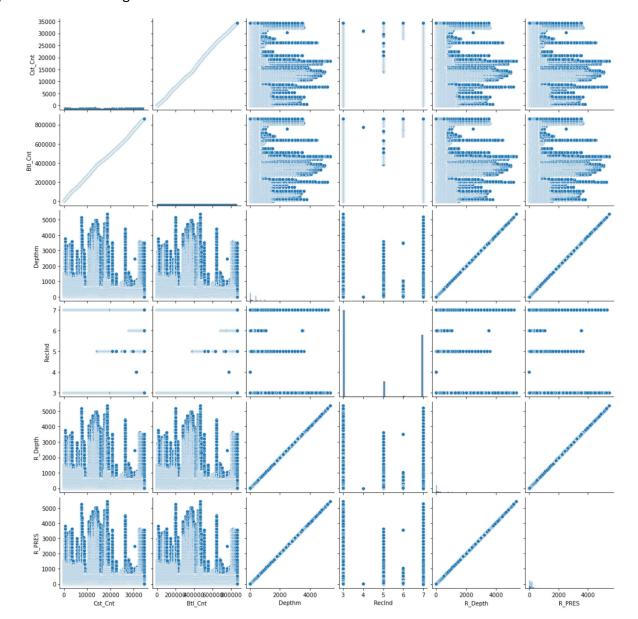
	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	RecInd	R_Depth	R_F
0	1	1	054.0 056.0	19-4903CR- HY-060-0930-05400560-0000A-3	0	3	0.0	
1	1	2	054.0 056.0	19-4903CR- HY-060-0930-05400560-0008A-3	8	3	8.0	
2	1	3	054.0 056.0	19-4903CR- HY-060-0930-05400560-0010A-7	10	7	10.0	
3	1	4	054.0 056.0	19-4903CR- HY-060-0930-05400560-0019A-3	19	3	19.0	
4	1	5	054.0 056.0	19-4903CR- HY-060-0930-05400560-0020A-7	20	7	20.0	
864858	34404	864859	093.4 026.4	20-1611SR- MX-310-2239-09340264-0000A-7	0	7	0.0	
864859	34404	864860	093.4 026.4	20-1611SR- MX-310-2239-09340264-0002A-3	2	3	2.0	
864860	34404	864861	093.4 026.4	20-1611SR- MX-310-2239-09340264-0005A-3	5	3	5.0	
864861	34404	864862	093.4 026.4	20-1611SR- MX-310-2239-09340264-0010A-3	10	3	10.0	
864862	34404	864863	093.4 026.4	20-1611SR- MX-310-2239-09340264-0015A-3	15	3	15.0	
864863 rows × 8 columns								

864863 rows × 8 columns

```
In [23]:
```

In [13]:

Out[13]: <seaborn.axisgrid.PairGrid at 0x2848002d340>



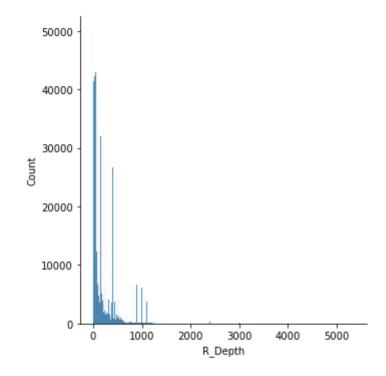
In [14]:

Out[14]:

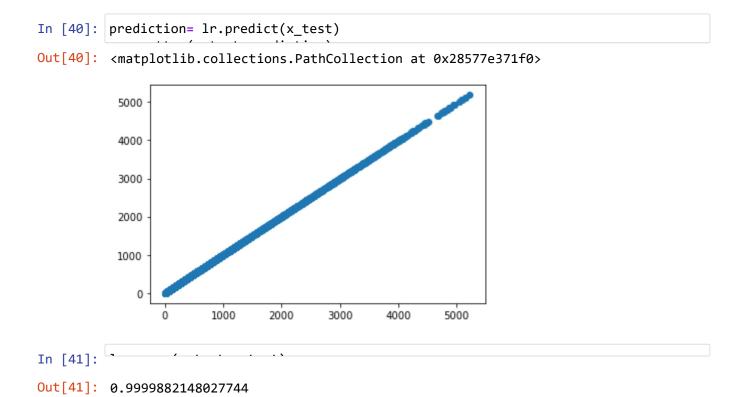
	Cst_Cnt	Btl_Cnt	Depthm	RecInd	R_Depth	R_PRE
count	864863.000000	864863.000000	864863.000000	864863.000000	864863.000000	864863.00000
mean	17138.790958	432432.000000	226.831951	4.700273	226.832495	228.39569
std	10240.949817	249664.587267	316.050259	1.877428	316.050007	319.45670
min	1.000000	1.000000	0.000000	3.000000	0.000000	0.00000
25%	8269.000000	216216.500000	46.000000	3.000000	46.000000	46.00000
50%	16848.000000	432432.000000	125.000000	3.000000	125.000000	126.00000
75%	26557.000000	648647.500000	300.000000	7.000000	300.000000	302.00000
max	34404.000000	864863.000000	5351.000000	7.000000	5351.000000	5458.00000

In [15]:

Out[15]: <seaborn.axisgrid.FacetGrid at 0x284b084a820>



```
In [16]:
Out[16]: <AxesSubplot:>
                                                              - 1.0
            Cst_Cnt -
                                                              - 0.8
            Btl Cnt -
                                                              - 0.6
            Depthm
                                                              - 0.4
             RecInd
                                                              0.2
           R_Depth
                                                               0.0
            R_PRES
                   Cst_Cnt Btl_Cnt Depthm RecInd R_Depth R_PRES
In [36]: x = df1[['Cst_Cnt', 'Btl_Cnt', 'Depthm', 'RecInd', 'R_Depth']]
In [37]: # to split my dataset into training and test data
          from sklearn.model_selection import train_test_split
In [38]: | from sklearn.linear_model import LinearRegression
          lr = LinearRegression()
Out[38]: LinearRegression()
In [39]:
          coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
Out[39]:
                    Co-efficient
                      -0.000167
            Cst_Cnt
                      0.000007
            Btl_Cnt
            Depthm
                      -0.738240
            RecInd
                      -0.018757
                      1.749083
           R_Depth
```



8 of 8