## **DATA COLLECTION**

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

```
In [2]: # To Import Dataset
sd=pd.read_csv(r"c:\Users\user\Downloads\bottle.csv")
sd
```

	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	SaInty	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	
5240	173	5241	069.0 144.0	19- 4905CR- HY-125- 1712- 06901440- 0279A-3	279	7.86	33.870	3.81	26.411	57.0	
5241	173	5242	069.0 144.0	19- 4905CR- HY-125- 1712- 06901440- 0300A-7	300	7.52	33.896	3.55	26.481	52.7	
5242	173	5243	069.0 144.0	19- 4905CR- HY-125- 1712- 06901440- 0371A-3	371	6.54	33.930	2.67	26.642	38.8	
5243	173	5244	069.0 144.0	19- 4905CR- HY-125- 1712- 06901440- 0400A-7	400	6.25	33.951	2.32	26.697	33.5	

Out[2]:

5244	173	52		9.0 HY-12 4.0 17 <i>1</i> 069014	25- 12- 10-	500 5.	44 34.0	)27 1	.30 26.8	358 1	8.
245 rows	s × 7	1 colur	nns								
		' top	5 rows								
Cst_Cı	nt B	tl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	SaInty	O2ml_L	STheta	O2Sat	
)	1	1	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0000A-3	0	10.50	33.440	NaN	25.649	NaN	
	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	
ı	1	5	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560-	20	10.45	33.421	NaN	25.643	NaN	
	to disd. head( Cst_C	to display d.head()  Cst_Cnt B	245 rows × 71 colur to display top d.head()  Cst_Cnt Btl_Cnt  1 1  1 2  1 1  1 4	245 rows × 71 columns  to display top 5 rows d.head()  Cst_Cnt Btl_Cnt Sta_ID  1 1 054.0 056.0  1 2 054.0 056.0  1 3 054.0 056.0	173 5245 069.0 HY-12 174 0500/  245 rows × 71 columns  254.0 Depth_ID  265.0 Depth_ID  379-4903CR-HY-060-0930-05400560-0000A-3  379-4903CR-HY-060-0930-05400560-0010A-7  381 1 4 054.0 HY-060-0930-05400560-0010A-7  391-4903CR-HY-060-0930-05400560-0019A-3  41 1 5 054.0 HY-060-0930-05400560-0019A-3  41 1 5 054.0 HY-060-0930-05400560-0019A-3	173 5243 144.0 1712- 06901440- 0500A-7  245 rows × 71 columns  to display top 5 rows 1.head()  Cst_Cnt Btl_Cnt Sta_ID Depth_ID Depthm  19- 4903CR- HY-060- 0000A-3  1 2 054.0 HY-060- 0000A-3  1 1 2 056.0 0930- 05400560- 00008A-3  1 1 3 054.0 HY-060- 00008A-3  1 1 4 054.0 HY-060- 0010A-7  1 1 4 054.0 HY-060- 0010A-7  1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	1 1 2 054.0 056.0 0930- 05400560- 0000A-7  1 1 3 054.0 056.0 0930- 05400560- 0000A-3  1 1 4 054.0 0930- 05400560- 0000A-3  1 1 3 054.0 0930- 05400560- 0010A-7  1 1 3 054.0 0930- 05400560- 0010A-7  1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	1 1 2 054.0 HY-060- 0000A-3 1 2 054.0 HY-060- 0000A-3 1 3 054.0 HY-060- 0000A-3 1 4 054.0 HY-060- 0000A-3 1 4 054.0 HY-060- 0000A-3 1 5 054.0 HY-060- 0000A-3 1 5 054.0 HY-060- 0000A-3 1 5 054.0 HY-060- 0000A-3 1 1 5 054.0 HY-060- 0010A-7 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	173 5245 069.0 HY-125- 1712- 500 5.44 34.027 1 245 rows × 71 columns  246 rows × 71 columns  247 rows × 71 columns  248 rows × 71 columns  258 rows × 71 columns  258 rows × 71 columns  259 rows × 71 columns  250 rows × 71 columns  260 rows × 71 columns  270 rows × 71 columns  271 rows × 71 columns  272 rows × 71 columns  273 rows × 71 columns  274 rows × 71 columns  275 rows × 71 columns	244 173 5245 069.0 HY-125- 500 5.44 34.027 1.30 26.6 06901440- 0500A-7  245 rows × 71 columns  to display top 5 rows 1.head()  Cst_Cnt Btl_Cnt Sta_ID Depth_ID Depthm T_degC Salnty O2ml_L STheta  1 054.0 HY-060- 0930- 05400560- 0000A-3  1 1 2 054.0 HY-060- 0930- 05400560- 0000A-3  1 1 2 054.0 HY-060- 0930- 05400560- 0000A-3  1 1 3 054.0 HY-060- 0930- 05400560- 0000A-3  1 1 3 054.0 HY-060- 0930- 05400560- 0000A-3  1 1 3 054.0 HY-060- 0930- 05400560- 0010A-7  1 1 4 054.0 HY-060- 0930- 05400560- 0019A-3  1 1 5 054.0 HY-060- 0930- 05600560- 0019A-3	1

Cst\_Cnt Btl\_Cnt Sta\_ID Depth\_ID Depthm T\_degC SaInty O2ml\_L STheta O2Sat ...

# DATA CLEANING AND PRE\_PROCESSING

In [4]: | sd.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5245 entries, 0 to 5244
Data columns (total 71 columns):

	·	11 /1 CO14IIII3):	
#	Column	Non-Null Count	Dtype
0	Cst_Cnt	5245 non-null	int64
1	Btl_Cnt	5245 non-null	int64
2	Sta_ID	5245 non-null	object
3	Depth_ID	5245 non-null	object
4	Depthm	5245 non-null	int64
5	T_degC	5225 non-null	float64
6	Salnty	5092 non-null	float64
7	O2m1_L	3051 non-null	float64
8	STheta	5077 non-null	float64
9	02Sat	2954 non-null	float64
10	Oxy_µmol/Kg	2954 non-null	float64
11	BtlNum	0 non-null	float64
12	RecInd	5245 non-null	int64
13	T_prec	5225 non-null	float64
14	T_qual	52 non-null	float64
15	S_prec	5092 non-null	float64
16	S_qual	239 non-null	float64
17	P_qual	5245 non-null	int64
18	O_qual	2198 non-null	float64
19	SThtaq	283 non-null	float64
20	02Satq	2362 non-null	float64
21	ChlorA	0 non-null	float64
22	Chlqua	5245 non-null	int64
23	Phaeop	0 non-null	float64
24	Phaqua	5245 non-null	int64
25	PO4uM	1236 non-null	float64
26	PO4q	4009 non-null	float64
27	SiO3uM	0 non-null	float64
28	SiO3qu	5245 non-null	int64
29	NO2uM	0 non-null	float64
30	NO2q	5245 non-null	int64
31	NO3uM	0 non-null	float64
32	NO3q	5245 non-null	int64
33	NH3uM	0 non-null	float64
34	NH3q	5245 non-null	int64
		0 non-null	
35	C14As1		float64
36 27	C14A1p	0 non-null	float64
37	C14A1q	5245 non-null	int64
38	C14As2	0 non-null	float64
39	C14A2p	0 non-null	float64
40	C14A2q	5245 non-null	int64
41	DarkAs	0 non-null	float64
42	DarkAp	0 non-null	float64
43	DarkAq	5245 non-null	int64
44	MeanAs	0 non-null	float64
45	MeanAp	0 non-null	float64
46	MeanAq	5245 non-null	int64
47	IncTim	0 non-null	float64
48	LightP	0 non-null	float64
49	R_Depth	5245 non-null	int64
50	R_TEMP	5225 non-null	float64
51	R_POTEMP	5019 non-null	float64

52	R_SALINITY	56	992 non-null	float64
53	R_SIGMA	49	962 non-null	float64
54	R_SVA	49	962 non-null	float64
55	R_DYNHT	56	31 non-null	float64
56	R_02	36	951 non-null	float64
57	R_02Sat	29	930 non-null	float64
58	R_SIO3	0	non-null	float64
59	R_P04	12	236 non-null	float64
60	R_NO3	0	non-null	float64
61	R_NO2	0	non-null	float64
62	R_NH4	0	non-null	float64
63	R_CHLA	0	non-null	float64
64	R_PHAEO	0	non-null	float64
65	R_PRES	52	245 non-null	int64
66	R_SAMP	0	non-null	float64
67	DIC1	0	non-null	float64
68	DIC2	0	non-null	float64
69	TA1	0	non-null	float64
70	TA2	0	non-null	float64
dtvn	es • float64/5	21	int64(17)	object(2)

dtypes: float64(52), int64(17), object(2)

memory usage: 2.8+ MB

In [5]: # to display summary of statistics
sd.describe()

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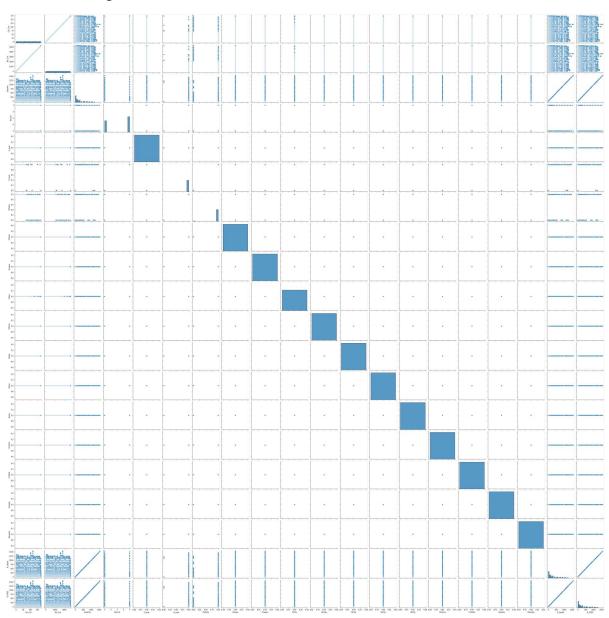
	Cst_Cnt	BtI_Cnt	Depthm	T_degC	Salnty	O2ml_L	SThet
count	5245.000000	5245.000000	5245.000000	5225.000000	5092.000000	3051.000000	5077.00000
mean	86.191992	2623.000000	345.400572	8.993144	33.827093	3.129839	26.11996
std	49.742962	1514.245412	356.298143	3.902631	0.492518	2.259226	0.92301
min	1.000000	1.000000	0.000000	2.700000	32.520000	0.210000	23.56800
25%	43.000000	1312.000000	54.000000	5.430000	33.490000	0.670000	25.11700
50%	86.000000	2623.000000	200.000000	8.620000	33.890000	2.990000	26.27900
75%	129.000000	3934.000000	600.000000	12.230000	34.255000	5.490000	26.99800
max	173.000000	5245.000000	1547.000000	19.760000	34.700000	6.630000	27.62300

8 rows × 69 columns

### **EDA** and visualization

In [97]: sns.pairplot(sd1)

Out[97]: <seaborn.axisgrid.PairGrid at 0x2ed22dca9a0>

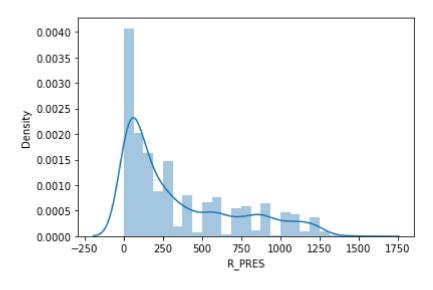


```
In [98]: sns.distplot(sd['R_PRES'])
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: Fut ureWarning: `distplot` is a deprecated function and will be removed in a futu re version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for hi stograms).

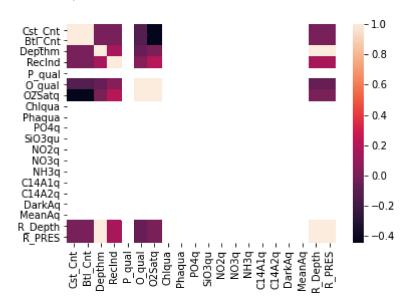
warnings.warn(msg, FutureWarning)

Out[98]: <AxesSubplot:xlabel='R\_PRES', ylabel='Density'>



In [99]: |sns.heatmap(sd1.corr())

Out[99]: <AxesSubplot:>



## TO TRAIN THE MODEL \_MODEL BUILDING

we are goint train Liner Regression model; we need to split out the data into two varibles x and y where x is independent on x (output) and y is dependent on x(output) adress coloumn as it is not required our model

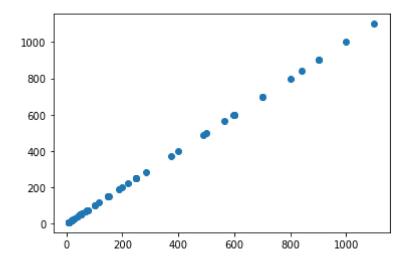
```
In [112]: coeff= pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff
```

#### Out[112]:

```
Co-efficient
Cst_Cnt 2.697449e-13
 Btl_Cnt -8.438516e-15
 RecInd -1.186997e-14
 P_qual 0.000000e+00
 O_qual 0.000000e+00
 O2Satq 0.000000e+00
 Chlqua 0.000000e+00
Phaqua 0.000000e+00
  PO4q
         0.000000e+00
 SiO3qu 0.000000e+00
  NO2q 0.000000e+00
  NO3q
         0.000000e+00
  NH3q 0.000000e+00
C14A1q 0.000000e+00
C14A2q 0.000000e+00
 DarkAq 0.000000e+00
MeanAq 0.000000e+00
R_Depth 1.000000e+00
R_PRES
         2.710715e-14
```

```
In [113]: prediction = lr.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[113]: <matplotlib.collections.PathCollection at 0x2ed3b811e50>



In [114]:	<pre>print(lr.score(x_test,y_test))</pre>							
	1.0							
In [ ]:								
In [ ]:								