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

## In [2]: x=pd.read\_csv(r"C:\Users\user\Downloads\\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[2]:

	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	054.0 056.0	19-4903CR- HY-060-0930-05400560-0020A-7	20	10.450	33.4210	
864858	34404	864859	093.4 026.4	20-1611SR- MX-310-2239-09340264-0000A-7	0	18.744	33.4083	5
864859	34404	864860	093.4 026.4	20-1611SR- MX-310-2239-09340264-0002A-3	2	18.744	33.4083	5
864860	34404	864861	093.4 026.4	20-1611SR- MX-310-2239-09340264-0005A-3	5	18.692	33.4150	5
864861	34404	864862	093.4 026.4	20-1611SR- MX-310-2239-09340264-0010A-3	10	18.161	33.4062	5
864862	34404	864863	093.4 026.4	20-1611SR- MX-310-2239-09340264-0015A-3	15	17.533	33.3880	5

864863 rows × 74 columns

In [3]: x=x.head(100)

## Out[3]:

	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	Salnty	O2ml_L
0	1	1	054.0 056.0	19-4903CR- HY-060-0930-05400560-0000A-3	0	10.50	33.440	NaN
1	1	2	054.0 056.0	19-4903CR- HY-060-0930-05400560-0008A-3	8	10.46	33.440	NaN
2	1	3	054.0 056.0	19-4903CR- HY-060-0930-05400560-0010A-7	10	10.46	33.437	NaN
3	1	4	054.0 056.0	19-4903CR- HY-060-0930-05400560-0019A-3	19	10.45	33.420	NaN
4	1	5	054.0 056.0	19-4903CR- HY-060-0930-05400560-0020A-7	20	10.45	33.421	NaN
95	4	96	050.0 095.0	19-4903CR- HY-061-1042-05000950-0019A-3	19	10.14	32.660	NaN
96	4	97	050.0 095.0	19-4903CR- HY-061-1042-05000950-0020A-7	20	10.14	32.657	NaN
97	4	98	050.0 095.0	19-4903CR- HY-061-1042-05000950-0030A-7	30	10.07	32.641	NaN
98	4	99	050.0 095.0	19-4903CR- HY-061-1042-05000950-0040A-3	40	9.97	NaN	NaN
99	4	100	050.0 095.0	19-4903CR- HY-061-1042-05000950-0050A-7	50	9.72	32.699	NaN

100 rows × 74 columns

## In [4]:

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 74 columns):

Data	columns	(total	74	columns):	
#	Column			Non-Null Count	Dtype
0	Cst_Cnt			100 non-null	int64
1	Btl_Cnt			100 non-null	int64
2	Sta_ID			100 non-null	object
3	Depth_ID	)		100 non-null	object
4	Depthm			100 non-null	int64
5	T_degC			100 non-null	float64
6	Salnty			98 non-null	float64
7	02ml_L			0 non-null	float64
8	STheta			98 non-null	float64
9	02Sat			0 non-null	float64
10	Oxy_µmol	./Kg		0 non-null	float64
11	Bt1Num	_		0 non-null	float64
12	RecInd			100 non-null	int64
13	T_prec			100 non-null	float64
14	T_qual			0 non-null	float64
15	S_prec			98 non-null	float64
16	S_qual			2 non-null	float64
17	P qual			100 non-null	float64
18	O_qual			100 non-null	float64
19	SThtaq			2 non-null	float64
20	02Satq			100 non-null	float64
21	ChlorA			0 non-null	float64
22	Chlqua			100 non-null	float64
23	Phaeop			0 non-null	float64
24	Phaqua			100 non-null	float64
25	PO4uM			0 non-null	float64
26	P04q			100 non-null	float64
27	SiO3uM			0 non-null	float64
28	Si03qu			100 non-null	float64
29	NO2uM			0 non-null	float64
30	NO2q			100 non-null	float64
31	NO3uM			0 non-null	float64
32	NO3q			100 non-null	float64
33	NH3uM			0 non-null	float64
34	NH3q			100 non-null	float64
35	C14As1			0 non-null	float64
36	C14A1p			0 non-null	float64
37	C14A1q			100 non-null	float64
38	C14As2			0 non-null	float64
39	C14A2p			0 non-null	float64
40	C14A2q			100 non-null	float64
41	DarkAs			0 non-null	float64
42	DarkAp			0 non-null	float64
43	DarkAq			100 non-null	float64
44	MeanAs			0 non-null	float64
45	MeanAp			0 non-null	float64
46	MeanAq			100 non-null	float64
47	IncTim			0 non-null	object
48	LightP			0 non-null	float64

```
49 R_Depth
                          100 non-null
                                          float64
                                          float64
50 R_TEMP
                          100 non-null
                          100 non-null
                                          float64
51 R POTEMP
52 R_SALINITY
                          98 non-null
                                          float64
53
    R_SIGMA
                          98 non-null
                                          float64
54
    R SVA
                          98 non-null
                                          float64
    R_DYNHT
55
                          100 non-null
                                          float64
56
    R_02
                          0 non-null
                                          float64
57
                          0 non-null
                                          float64
    R 02Sat
    R_SI03
                          0 non-null
                                          float64
58
59
    R_P04
                          0 non-null
                                          float64
                                          float64
60
    R NO3
                          0 non-null
61
    R NO2
                          0 non-null
                                          float64
    R_NH4
                          0 non-null
                                          float64
63
    R CHLA
                          0 non-null
                                          float64
                          0 non-null
                                          float64
64 R PHAEO
    R_PRES
                          100 non-null
                                          int64
65
                          0 non-null
                                          float64
66 R_SAMP
67
    DIC1
                          0 non-null
                                          float64
68
    DIC2
                          0 non-null
                                          float64
69
    TA1
                          0 non-null
                                          float64
70 TA2
                          0 non-null
                                          float64
71
    pH2
                          0 non-null
                                          float64
                          0 non-null
                                          float64
72 pH1
    DIC Quality Comment 0 non-null
                                          object
dtypes: float64(65), int64(5), object(4)
memory usage: 57.9+ KB
```

In [11]: d=x[['R\_PRES','RecInd','Depthm','Cst\_Cnt']]

Out[11]:	•				
out[II].		R_PRES	RecInd	Depthm	Cst_Cnt
	0	0	3	0	1
	1	8	3	8	1
	2	10	7	10	1
	3	19	3	19	1
	4	20	7	20	1
	95	19	3	19	4
	96	20	7	20	4
	97	30	7	30	4
	98	40	3	40	4
	99	50	7	50	4

100 rows × 4 columns

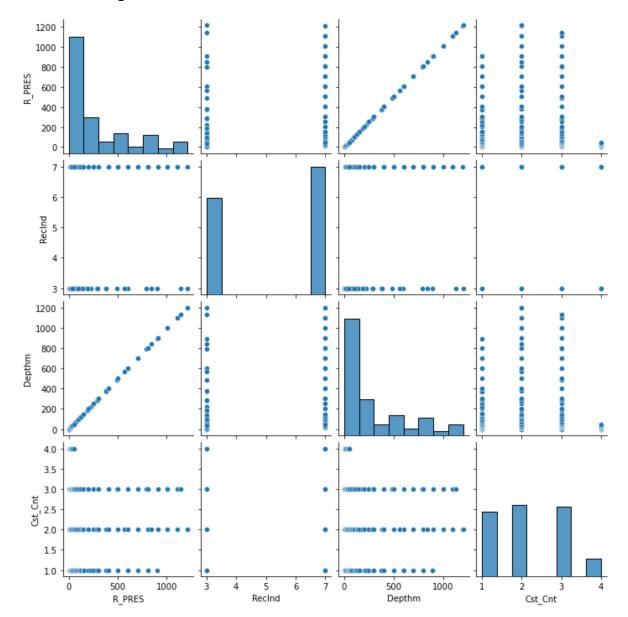
Out[12]:

	Cst_Cnt	Btl_Cnt	Depthm	T_degC	Salnty	O2ml_L	STheta	O2Sat	(
count	100.000000	100.000000	100.000000	100.000000	98.000000	0.0	98.000000	0.0	_
mean	2.180000	50.500000	302.130000	7.625300	33.621755	NaN	26.224337	NaN	
std	0.946872	29.011492	338.487296	2.413229	0.580959	NaN	0.767899	NaN	
min	1.000000	1.000000	0.000000	3.140000	32.630000	NaN	25.069000	NaN	
25%	1.000000	25.750000	45.250000	5.297500	33.159000	NaN	25.628750	NaN	
50%	2.000000	50.500000	150.000000	8.370000	33.810000	NaN	26.293000	NaN	
75%	3.000000	75.250000	500.000000	9.862500	34.112250	NaN	26.970250	NaN	
max	4.000000	100.000000	1203.000000	10.500000	34.450000	NaN	27.425000	NaN	

8 rows × 70 columns

In [13]:

Out[13]: <seaborn.axisgrid.PairGrid at 0x22d05ff3fa0>

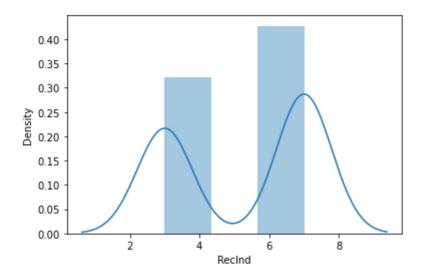


```
In [16]:
```

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)

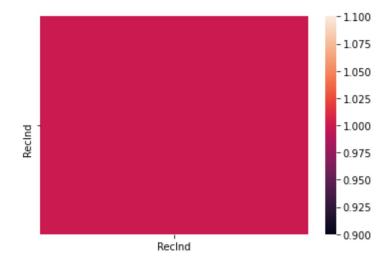
Out[16]: <AxesSubplot:xlabel='RecInd', ylabel='Density'>



In [17]:

In [18]:

Out[18]: <AxesSubplot:>



In [20]: x=x1[['RecInd']]

```
In [21]: # to split my dataset into traning and test date
         from sklearn.model_selection import train_test_split
In [22]: from sklearn.linear_model import LinearRegression
         lr=LinearRegression()
Out[22]: LinearRegression()
In [23]:
         8.881784197001252e-16
In [24]: coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
Out[24]:
                 Co-efficient
          RecInd
                        1.0
In [25]: prediction=lr.predict(x_test)
Out[25]: <matplotlib.collections.PathCollection at 0x22d08f7d9d0>
          7.0
          6.5
          6.0
          5.5
           5.0
          4.5
          4.0
          3.5
          3.0 -
               3.0
                    3.5
                         4.0
                                    5.0
                                         5.5
                                              6.0
                                                         7.0
In [26]: __
Out[26]: 1.0
In [27]:
Out[27]: 1.0
In [28]:
```

```
In [29]: rr=Ridge(alpha=10)
    rr.fit(x_train,y_train)

Out[29]: 0.9984798940655064

In [30]: la=Lasso(alpha=10)
Out[30]: Lasso(alpha=10)

In [31]:

Out[31]: -0.27840909090913

In []:
```