Linear Regression

In [94]:

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn import preprocessing,svm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
df=pd.read_csv(r"C:\Users\91949\Downloads\bottle.csv.zip")
df
```

C:\Users\91949\AppData\Local\Temp\ipykernel_13352\687492671.py:8: DtypeWa
rning: Columns (47,73) have mixed types. Specify dtype option on import o
r set low_memory=False.
 df=pd.read_csv(r"C:\Users\91949\Downloads\bottle.csv.zip")

Out[94]:

	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	Salnty	O2ml_L	STheta
0	1	1	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0000A-3	0	10.500	33.4400	NaN	25.64900
1	1	2	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0008A-3	8	10.460	33.4400	NaN	25.65600
2	1	3	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0010A-7	10	10.460	33.4370	NaN	25.65400
3	1	4	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0019A-3	19	10.450	33.4200	NaN	25.64300
4	1	5	054.0 056.0	19- 4903CR- HY-060- 0930- 05400560- 0020A-7	20	10.450	33.4210	NaN	25.64300
864858	34404	864859	093.4 026.4	20- 1611SR- MX-310- 2239- 09340264- 0000A-7	0	18.744	33.4083	5.805	23.87055
864859	34404	864860	093.4 026.4	20- 1611SR- MX-310- 2239- 09340264- 0002A-3	2	18.744	33.4083	5.805	23.87072
864860	34404	864861	093.4 026.4	20- 1611SR- MX-310- 2239- 09340264- 0005A-3	5	18.692	33.4150	5.796	23.88911
864861	34404	864862	093.4 026.4	20- 1611SR- MX-310- 2239- 09340264- 0010A-3	10	18.161	33.4062	5.816	24.01426

In [95]:	Cst_Cnt	Btl_Cnt	Sta_ID	Depth_ID	Depthm	T_degC	SaInty	O2ml_L	STheta
df.info(
				20-					
864862	34404	864863	093.4 026.4	1611SR- MX-310- 2239- 09340264- 0015A-3	15	17.533	33.3880	5.774	24.15297

864863 rows × 74 columns

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 864863 entries, 0 to 864862

Data colu	ımns (total	74 co.	lumns):
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#	Column (Cocal /	Non-Null Count	Dtype
		Non-Naii Counc	
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	O2ml_L	696201 non-null	float64
8	STheta	812174 non-null	float64
9	02Sat	661274 non-null	
10	Oxy_μmol/Kg	661268 non-null	
11	BtlNum	118667 non-null	
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		673755 non-null	float64
18	P_qual O qual	184676 non-null	float64
	SThtaq	65823 non-null	float64
19	•	217797 non-null	float64
20 21	O2Satq ChlorA	225272 non-null	float64
21			
22	Chlqua	639166 non-null	float64
23	Phaeop	225271 non-null	float64
24 25	Phaqua	639170 non-null	float64
25 26	PO4uM	413317 non-null	float64
26 27	PO4q	451786 non-null	float64
27	SiO3uM	354091 non-null	float64
28	SiO3qu	510866 non-null 337576 non-null	float64
29	NO2uM		float64
30 21	NO2q	529474 non-null 337403 non-null	float64 float64
31 32	NO3uM	529933 non-null	
	NO3q		float64
33	NH3uM	64962 non-null	float64
34 35	NH3q C14As1	808299 non-null 14432 non-null	float64 float64
		12760 non-null	
36	C14A1p		float64
37 20	C14A1q	848605 non-null	float64
38	C14As2	14414 non-null	float64 float64
39 40	C14A2p	12742 non-null 848623 non-null	float64
40 41	C14A2q DarkAs	22649 non-null	float64
		20457 non-null	
42	DarkAp		float64
43	DarkAq	840440 non-null	float64
44 45	MeanAs	22650 non-null	float64
45 46	MeanAp	20457 non-null	float64
46	MeanAq	840439 non-null	float64
47 49	IncTim	14437 non-null	object
48 40	LightP P Donth	18651 non-null	float64
49 50	R_Depth	864863 non-null	float64
50 51	R_TEMP	853900 non-null	float64
51	R_POTEMP	818816 non-null	float64
52 52	R_SALINITY	817509 non-null	float64
53 54	R_SIGMA	812007 non-null	float64
54	R_SVA	812092 non-null 818206 non-null	float64
55	R_DYNHT	010700 HOU-HATT	float64

56	R_02	696201 non-null	float64
57	R_02Sat	666448 non-null	float64
58	R_SIO3	354099 non-null	float64
59	R_P04	413325 non-null	float64
60	R_NO3	337 411 non-null	float64
61	R_NO2	337584 non-null	float64
62	R_NH4	64982 non-null	float64
63	R_CHLA	225276 non-null	float64
64	R_PHAEO	225275 non-null	float64
65	R_PRES	864863 non-null	int64
66	R_SAMP	122006 non-null	float64
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
72	pH1	84 non-null	float64
73	DIC Quality Comment	55 non-null	object

dtypes: float64(65), int64(5), object(4)

memory usage: 488.3+ MB

In [96]:

df.describe()

Out[96]:

	Cst_Cnt	Btl_Cnt	Depthm	T_degC	Salnty	С
count	864863.000000	864863.000000	864863.000000	853900.000000	817509.000000	696201.
mean	17138.790958	432432.000000	226.831951	10.799677	33.840350	3.:
std	10240.949817	249664.587269	316.050259	4.243825	0.461843	2.1
min	1.000000	1.000000	0.000000	1.440000	28.431000	-0.1
25%	8269.000000	216216.500000	46.000000	7.680000	33.488000	1.:
50%	16848.000000	432432.000000	125.000000	10.060000	33.863000	3.4
75%	26557.000000	648647.500000	300.000000	13.880000	34.196900	5.
max	34404.000000	864863.000000	5351.000000	31.140000	37.034000	11.

8 rows × 70 columns

```
In [97]:
```

```
df.isna().any()
Out[97]:
Cst_Cnt
                        False
Btl_Cnt
                        False
                        False
Sta_ID
Depth_ID
                        False
Depthm
                        False
                        . . .
TA1
                         True
TA2
                         True
pH2
                         True
pH1
                         True
DIC Quality Comment
                         True
Length: 74, dtype: bool
In [98]:
df.isnull().sum()
Out[98]:
Cst_Cnt
                             0
Btl_Cnt
                             0
Sta_ID
                             0
Depth_ID
                             0
Depthm
                             0
TA1
                        862779
TA2
                        864629
pH2
                        864853
pH1
                        864779
DIC Quality Comment
                        864808
Length: 74, dtype: int64
In [99]:
df=df[ [ 'Salnty', 'T_degC']]
df.columns=['Sal', 'Temp']
```

In [100]:

df.head(20)

Out[100]:

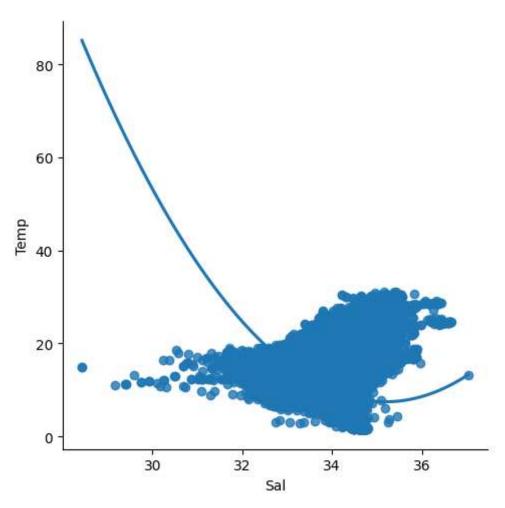
	Sal	Temp
0	33.440	10.50
1	33.440	10.46
2	33.437	10.46
3	33.420	10.45
4	33.421	10.45
5	33.431	10.45
6	33.440	10.45
7	33.424	10.24
8	33.420	10.06
9	33.494	9.86
10	33.510	9.83
11	33.580	9.67
12	33.640	9.50
13	33.689	9.32
14	33.847	8.76
15	33.860	8.71
16	33.876	8.53
17	NaN	8.45
18	33.926	8.26
19	33.980	7.96

In [101]:

```
sns.lmplot(x='Sal',y='Temp', data=df,order=2, ci=None)
```

Out[101]:

<seaborn.axisgrid.FacetGrid at 0x20b92f19190>



In [103]:

```
df.fillna(method='ffill',inplace=True)
```

C:\Users\91949\AppData\Local\Temp\ipykernel_13352\205861073.py:1: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

df.fillna(method='ffill',inplace=True)

In [104]:

```
x=np.array(df['Sal']).reshape(-1,1)
y=np.array(df[ 'Temp']).reshape(-1,1)
```

In [105]:

```
df.dropna(inplace=True)
```

C:\Users\91949\AppData\Local\Temp\ipykernel_13352\1552371080.py:1: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

df.dropna(inplace=True)

In [106]:

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
```

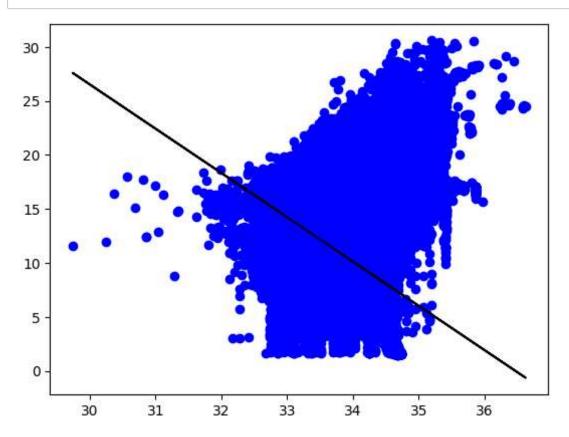
In [107]:

```
regr=LinearRegression()
regr.fit(x_train,y_train)
print(regr.score(x_test,y_test))
```

0.20498680913493517

In [109]:

```
y_pred=regr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test, y_pred, color='k')
plt.show()
```

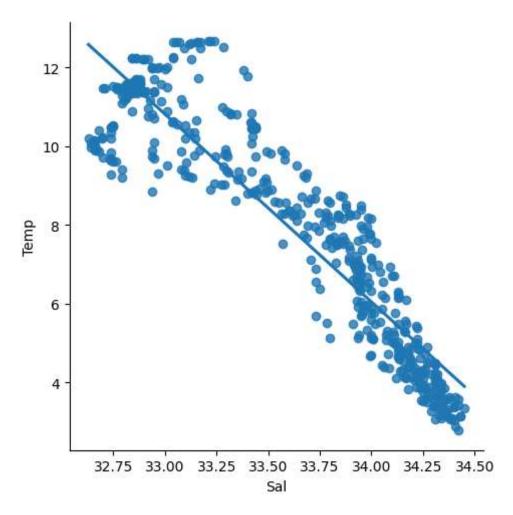


In [110]:

```
df500=df[:] [:500]
sns.lmplot(x='Sal',y='Temp', data=df500, order=1, ci=None)
```

Out[110]:

<seaborn.axisgrid.FacetGrid at 0x20b930a6c50>



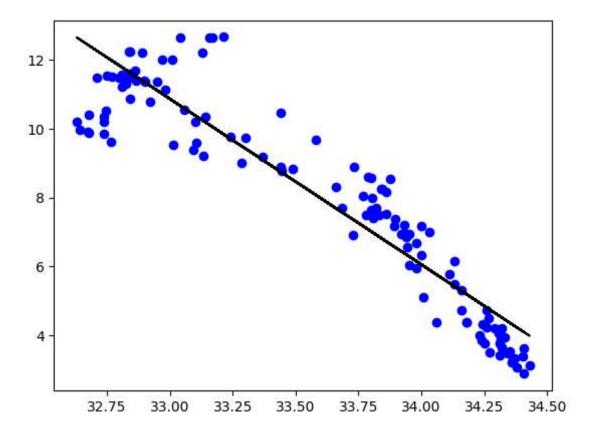
In [112]:

```
df500.fillna (method='ffill', inplace=True)
x=np.array(df500['Sal']).reshape(-1,1)
y=np.array(df500 [ 'Temp']).reshape(-1,1)
df500.dropna (inplace=True)
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
regr=LinearRegression ()
regr.fit(x_train,y_train)
print("Regression:",regr.score(x_test,y_test))
y_pred=regr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test,y_pred, color='k')
plt.show
```

Regression: 0.8679290669545694

Out[112]:

<function matplotlib.pyplot.show(close=None, block=None)>



In [115]:

```
#Evaluevating a model
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score
model=LinearRegression()
model.fit(x_train,y_train)
y_pred=model.predict(x_test)
r2=r2_score(y_test,y_pred)
print("R2_Score: ",r2)
```

R2_Score: 0.8679290669545694

conclusion:

Dataset we have taken is poor for linear model but with smaller data works well with linear model