linear regression model: 1)Problem statement:how best fit the dataset?

```
In [4]: 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
```

2)Reading data

```
In [5]: df=pd.read_csv(r"C:\Users\Jayadeep\Downloads\bottle.csv.zip")
df
```

C:\Users\Jayadeep\AppData\Local\Temp\ipykernel\_26344\2871314872.py:1: DtypeWarning: Columns (47,73) have mixed type
s. Specify dtype option on import or set low\_memory=False.
 df=pd.read\_csv(r"C:\Users\Jayadeep\Downloads\bottle.csv.zip")

### Out[5]:

|        | Cst_Cnt | Btl_Cnt | Sta_ID         | Depth_ID   | Depthm | T_degC | Salnty  | O2ml_L | STheta   | O2Sat  | <br>R_PHAEO | R_PRES | R_SAMP | DIC1 | DIC2 |
|--------|---------|---------|----------------|--|--------|--------|---------|--------|----------|--------|-------------|--------|--------|------|------|
| 0      | 1       | 1       | 054.0<br>056.0 | 19-<br>4903CR-<br>HY-060-<br>0930-<br>05400560-<br>0000A-3 | 0      | 10.500 | 33.4400 | NaN    | 25.64900 | NaN    | <br>NaN     | 0      | NaN    | NaN  | NaN  |
| 1      | 1       | 2       | 054.0<br>056.0 | 19-<br>4903CR-<br>HY-060-<br>0930-<br>05400560-<br>0008A-3 | 8      | 10.460 | 33.4400 | NaN    | 25.65600 | NaN    | <br>NaN     | 8      | NaN    | NaN  | NaN  |
| 2      | 1       | 3       | 054.0<br>056.0 | 19-<br>4903CR-<br>HY-060-<br>0930-<br>05400560-<br>0010A-7 | 10     | 10.460 | 33.4370 | NaN    | 25.65400 | NaN    | <br>NaN     | 10     | NaN    | NaN  | NaN  |
| 3      | 1       | 4       | 054.0<br>056.0 | 19-<br>4903CR-<br>HY-060-<br>0930-<br>05400560-<br>0019A-3 | 19     | 10.450 | 33.4200 | NaN    | 25.64300 | NaN    | <br>NaN     | 19     | NaN    | NaN  | NaN  |
| 4      | 1       | 5       | 054.0<br>056.0 | 19-<br>4903CR-<br>HY-060-<br>0930-<br>05400560-<br>0020A-7 | 20     | 10.450 | 33.4210 | NaN    | 25.64300 | NaN    | <br>NaN     | 20     | NaN    | NaN  | NaN  |
|        |         |         |                |  |        |        |         |        |          |        | <br>        |        |        |      |      |
| 864858 | 34404   | 864859  | 093.4<br>026.4 | 20-<br>1611SR-<br>MX-310-<br>2239-<br>09340264-<br>0000A-7 | 0      | 18.744 | 33.4083 | 5.805  | 23.87055 | 108.74 | <br>0.18    | 0      | NaN    | NaN  | NaN  |

|        | Cst_Cnt | Btl_Cnt | Sta_ID         | Depth_ID   | Depthm | T_degC | Salnty  | O2ml_L | STheta   | O2Sat  | <br>R_PHAEO | R_PRES | R_SAMP | DIC1 | DIC2 |
|--------|---------|---------|----------------|--|--------|--------|---------|--------|----------|--------|-------------|--------|--------|------|------|
| 864859 | 34404   | 864860  | 093.4<br>026.4 | 20-<br>1611SR-<br>MX-310-<br>2239-<br>09340264-<br>0002A-3 | 2      | 18.744 | 33.4083 | 5.805  | 23.87072 | 108.74 | <br>0.18    | 2      | 4.0    | NaN  | NaN  |
| 864860 | 34404   | 864861  | 093.4<br>026.4 | 20-<br>1611SR-<br>MX-310-<br>2239-<br>09340264-<br>0005A-3 | 5      | 18.692 | 33.4150 | 5.796  | 23.88911 | 108.46 | <br>0.18    | 5      | 3.0    | NaN  | NaN  |
| 864861 | 34404   | 864862  | 093.4<br>026.4 | 20-<br>1611SR-<br>MX-310-<br>2239-<br>09340264-<br>0010A-3 | 10     | 18.161 | 33.4062 | 5.816  | 24.01426 | 107.74 | <br>0.31    | 10     | 2.0    | NaN  | NaN  |
| 864862 | 34404   | 864863  | 093.4<br>026.4 | 20-<br>1611SR-<br>MX-310-<br>2239-<br>09340264-<br>0015A-3 | 15     | 17.533 | 33.3880 | 5.774  | 24.15297 | 105.66 | <br>0.61    | 15     | 1.0    | NaN  | NaN  |

864863 rows × 74 columns

```
In [8]: df=df[['Salnty','T_degC']]
df.columns=['sal','temp']
```

In [9]: df.head(15)

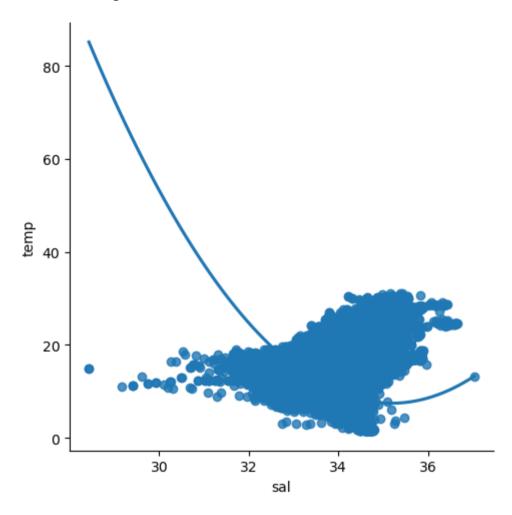
Out[9]:

|    | 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  |

3)Exploing the data scatter-plottting the data scatter

```
In [11]: sns.lmplot(x='sal',y='temp',data=df,order=2,ci=None)
```

Out[11]: <seaborn.axisgrid.FacetGrid at 0x1e9f42e9220>



# In [12]: df.describe()

#### Out[12]:

|       | sal           | temp          |
|-------|---------------|---------------|
| count | 817509.000000 | 853900.000000 |
| mean  | 33.840350     | 10.799677     |
| std   | 0.461843      | 4.243825      |
| min   | 28.431000     | 1.440000      |
| 25%   | 33.488000     | 7.680000      |
| 50%   | 33.863000     | 10.060000     |
| 75%   | 34.196900     | 13.880000     |
| max   | 37.034000     | 31.140000     |

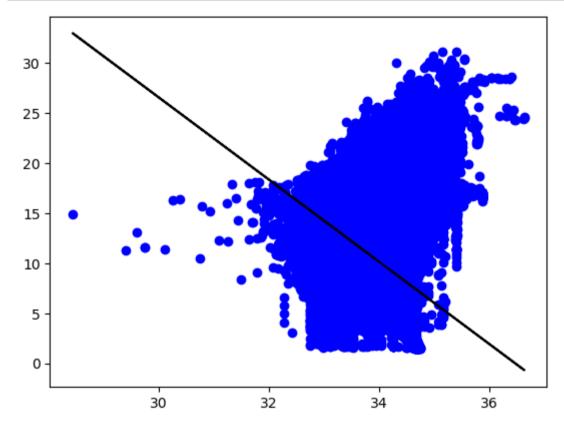
## In [13]: df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 864863 entries, 0 to 864862
Data columns (total 2 columns):
# Column Non-Null Count Dtype
--- 0 sal 817509 non-null float64
1 temp 853900 non-null float64
dtypes: float64(2)
memory usage: 13.2 MB
```

4)Data cleaning-Eliminating nan or missing i/p numbers

```
In [15]: df.fillna(method='ffill',inplace=True)
         C:\Users\Jayadeep\AppData\Local\Temp\ipykernel 26344\4116506308.py:1: SettingWithCopyWarning:
         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#returnin
         g-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a-view-versu
         s-a-copy)
           df.fillna(method='ffill',inplace=True)
         5)Training our model
In [16]: x=np.array(df['sal']).reshape(-1,1)
         y=np.array(df['temp']).reshape(-1,1)
In [17]: df.dropna(inplace=True)
         C:\Users\Jayadeep\AppData\Local\Temp\ipykernel 26344\1379821321.py:1: SettingWithCopyWarning:
         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#returnin
         g-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a-view-versu
         s-a-copy)
           df.dropna(inplace=True)
In [18]: x train,x test,y train,y test=train test split(x,y,test size=0.25)
         #splitting data into train and test
         regr=LinearRegression()
         regr.fit(x train,y train)
         print(regr.score(x test,y test))
         0.20364539273594318
         6)Exploring our results
```

```
In [19]: y_pred=regr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test,y_pred,color='k')
plt.show()
```



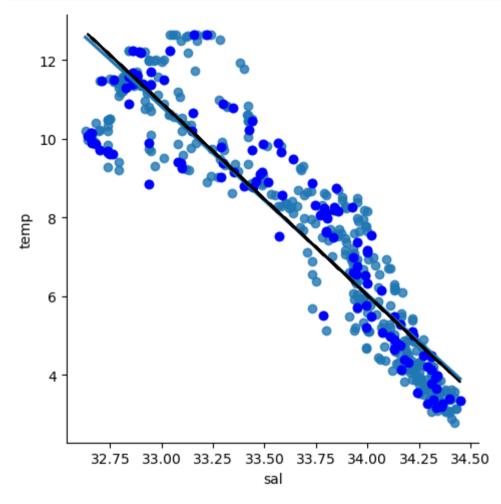
7)working with smaller dataset

```
In [20]: df500=df[:][:500]
sns.lmplot(x="sal",y="temp",data=df500,order=1,ci=None)
```

Out[20]: <seaborn.axisgrid.FacetGrid at 0x1e9f4071d30>

Regression: 0.8173107904985126

```
In [26]: y_pred=regr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test,y_pred,color='k')
plt.show()
```



8)Evaluation of model

```
In [27]: from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score
```

# In [28]: #train model model=LinearRegression() model.fit(x\_train,y\_train) #Evaluation the model on the test set y\_pred=model.predict(x\_test) r2=r2\_score(y\_test,y\_pred) print("R2 score:",r2)

R2 score: 0.8173107904985126

conclusion: Data set we have taken is poor for this model, but smaller data is suitable for linear model