A real estate agent want help to predict the house price for regions in Usa.he gave us the dataset to work on to use linear Regression model.Create a model that helps him to estimate

Data Collection

```
#import Libraries
 In [1]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
In [78]:
          #import the dataset
          data=pd.read csv(r"C:\Users\user\Desktop\Vicky\17 student marks.csv")[0:500]
In [79]:
          #to display top 10 rows
          data.head()
Out[79]:
              Student_ID Test_1 Test_2 Test_3 Test_4 Test_5 Test_6 Test_7 Test_8 Test_9 Test_10 1
           0
                  22000
                                   87
                                                        88
                                                               98
                                                                             100
                                                                                    100
                                                                                            100
           1
                  22001
                            79
                                   71
                                          81
                                                 72
                                                        73
                                                               68
                                                                      59
                                                                             69
                                                                                     59
                                                                                             60
           2
                  22002
                                                 74
                            66
                                   65
                                          70
                                                        78
                                                               86
                                                                      87
                                                                             96
                                                                                     88
                                                                                             82
                  22003
                                                 61
                                                                                             63
                            60
                                   58
                                          54
                                                        54
                                                               57
                                                                      64
                                                                             62
                                                                                     72
                  22004
                            99
                                                 93
                                                        97
                                                                      92
                                                                              98
                                                                                     91
                                                                                             98
                                   95
                                          96
                                                               89
```

```
In [80]:
          #to display null values
          data.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 56 entries, 0 to 55
          Data columns (total 13 columns):
           #
                Column
                             Non-Null Count
                                               Dtype
                                               ----
           0
                Student_ID
                             56 non-null
                                               int64
                Test_1
                             56 non-null
                                               int64
           1
                             56 non-null
           2
                Test 2
                                               int64
           3
                Test_3
                             56 non-null
                                               int64
                             56 non-null
           4
                Test_4
                                               int64
           5
                Test 5
                             56 non-null
                                               int64
           6
                             56 non-null
                                               int64
                Test_6
           7
                Test_7
                             56 non-null
                                               int64
           8
                Test_8
                             56 non-null
                                               int64
           9
                Test_9
                             56 non-null
                                               int64
               Test 10
                             56 non-null
           10
                                               int64
           11
               Test 11
                             56 non-null
                                               int64
               Test 12
                             56 non-null
           12
                                               int64
          dtypes: int64(13)
          memory usage: 5.8 KB
In [81]: data.shape
Out[81]: (56, 13)
In [82]: #to display summary of statistics
          data.describe()
Out[82]:
                    Student_ID
                                  Test_1
                                             Test_2
                                                        Test_3
                                                                   Test_4
                                                                              Test_5
                                                                                         Test_6
           count
                    56.000000
                               56.000000
                                           56.000000
                                                     56.000000
                                                                56.000000
                                                                           56.000000
                                                                                      56.000000
           mean 22027.500000
                               70.750000
                                           69.196429
                                                                67.446429
                                                                           67.303571
                                                                                                 6
                                                     68.089286
                                                                                      66.000000
                               17.009356
                                                                           20.746890
             std
                     16.309506
                                          17.712266
                                                     18.838333
                                                                19.807179
                                                                                      21.054043
             min 22000.000000
                               40.000000
                                          34.000000
                                                     35.000000
                                                                28.000000
                                                                           26.000000
                                                                                      29.000000
            25% 22013.750000
                                          55.750000
                               57.750000
                                                     53.000000
                                                                           53.750000
                                                                54.500000
                                                                                      50.250000
            50% 22027.500000
                               70.500000
                                           68.500000
                                                     70.000000
                                                                71.500000
                                                                           69.000000
                                                                                      65.500000
                                                                                                 (
            75% 22041.250000
                               84.000000
                                          83.250000
                                                     85.000000
                                                                84.000000
                                                                           85.250000
                                                                                      83.750000
                                                                                                 {
                                                               100.000000 100.000000
            max 22055.000000 100.000000 100.000000
                                                    100.000000
                                                                                     100.000000
                                                                                                1(
          #to display columns name
In [83]:
          data.columns
Out[83]: Index(['Student_ID', 'Test_1', 'Test_2', 'Test_3', 'Test_4', 'Test_5',
                  'Test_6', 'Test_7', 'Test_8', 'Test_9', 'Test_10', 'Test_11',
                  'Test 12'],
                 dtype='object')
```

In [45]: data.fillna(value=5)

Out[45]:

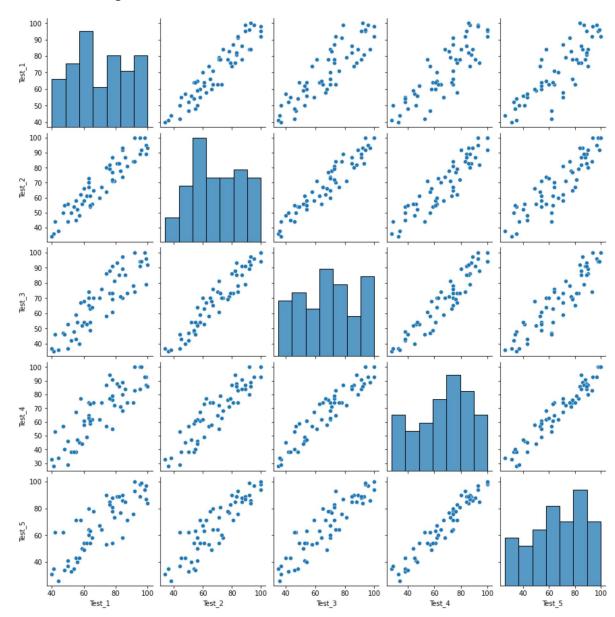
	ld	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa
145	146	6.7	3.0	5.2	2.3	Iris-virginica
146	147	6.3	2.5	5.0	1.9	Iris-virginica
147	148	6.5	3.0	5.2	2.0	Iris-virginica
148	149	6.2	3.4	5.4	2.3	Iris-virginica
149	150	5.9	3.0	5.1	1.8	Iris-virginica

150 rows × 6 columns

In [84]: data1=data[['Test_1', 'Test_2', 'Test_3', 'Test_4', 'Test_5']]

In [85]: sns.pairplot(data1)

Out[85]: <seaborn.axisgrid.PairGrid at 0x1257cd70d90>

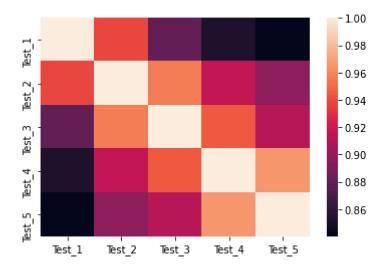


EDA and Visualization

In [86]: #sns.distplot(data['Co2-Emissions'])

```
In [87]: sns.heatmap(data1.corr())
```

Out[87]: <AxesSubplot:>



To train the model

we are going to train the linear regression model; We need to split the two variable x and y where x in independent variable (input) and y is dependent of x(output) so we could ignore address columns as it is not requires for our model

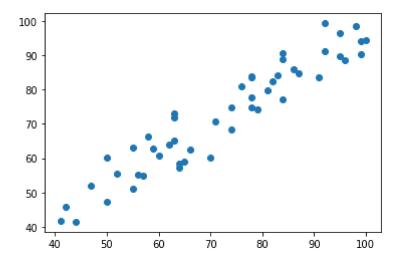
```
In [115]: coeff = pd.DataFrame(lr.coef_,x.columns,columns=["Co-efficient"])
coeff
```

Out[115]:

	Co-efficient
Test_2	1.246687
Test_3	-0.180468
Test_4	-0.067759
Test_5	0.168814
Test_6	- 0.084408
Test_7	-0.359143
Test_8	0.083592
Test_9	0.188246
Test_10	-0.066289
Test_11	-0.117327
Test_12	0.089670

```
In [116]: prediction = lr.predict(x_train)
plt.scatter(y_train,prediction)
```

Out[116]: <matplotlib.collections.PathCollection at 0x1257fe1a430>



```
In [117]: lr.score(x_test,y_test)
```

Out[117]: 0.5198538480792071

```
In [118]: lr.score(x_train,y_train)
```

Out[118]: 0.9015962733027583

```
In [119]: from sklearn.linear_model import Ridge,Lasso
```

```
In [120]: | rr=Ridge(alpha=10)
         rr.fit(x_train,y_train)
         rr.score(x_test,y_test)
Out[120]: 0.5279347842005535
In [121]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
         la.score(x_test,y_test)
Out[121]: 0.6895001439973342
In [122]: from sklearn.linear_model import ElasticNet
         en= ElasticNet()
         en.fit(x_train,y_train)
Out[122]: ElasticNet()
In [123]: print(en.coef_)
         0.0307049
                      0.1560741 -0.04507002 -0.03903421 0.01253983]
In [124]: | print(en.intercept_)
         9.192973877887766
In [125]:
         prediction = en.predict(x test)
         prediction
Out[125]: array([57.62294502, 70.82109935, 61.87583691, 76.45798479, 37.45064913,
                60.8023169 ])
```

```
In [126]: | print(en.score(x test,y train))
          ValueError
                                                     Traceback (most recent call last)
          <ipython-input-126-48d1f0543252> in <module>
          ----> 1 print(en.score(x test,y train))
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\base.py in score(self, X,
          y, sample_weight)
              552
                           from .metrics import r2_score
              553
                           y_pred = self.predict(X)
           --> 554
                           return r2_score(y, y_pred, sample_weight=sample_weight)
              555
              556
                       def _more_tags(self):
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inn
          er_f(*args, **kwargs)
               61
                               extra_args = len(args) - len(all_args)
               62
                               if extra args <= 0:</pre>
           ---> 63
                                   return f(*args, **kwargs)
               64
               65
                               # extra_args > 0
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\ regression.py in
          r2_score(y_true, y_pred, sample_weight, multioutput)
              674
                       -3.0
              675
          --> 676
                      y type, y true, y pred, multioutput = check reg targets(
                           y true, y pred, multioutput)
              677
              678
                       check_consistent_length(y_true, y_pred, sample_weight)
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\ regression.py in
          _check_reg_targets(y_true, y_pred, multioutput, dtype)
               86
                           the dtype argument passed to check array.
               87
           ---> 88
                       check consistent length(y true, y pred)
                       y true = check array(y true, ensure 2d=False, dtype=dtype)
               89
                       y pred = check array(y pred, ensure 2d=False, dtype=dtype)
               90
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in che
          ck consistent length(*arrays)
              260
                       uniques = np.unique(lengths)
                       if len(uniques) > 1:
              261
                           raise ValueError("Found input variables with inconsistent num
          --> 262
          bers of"
                                            " samples: %r" % [int(1) for 1 in lengths])
              263
              264
          ValueError: Found input variables with inconsistent numbers of samples: [50,
          6]
In [127]: from sklearn import metrics
```