Problem statement

predicting the house price in USA. To create a model to help him estimate of what the house would sell for.

To display top 10 rows

```
In [3]: 1 df.head(10)
```

Out[3]:

	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
0	22000	78	87	91	91	88	98	94	100	100	100	100	93
1	22001	79	71	81	72	73	68	59	69	59	60	61	67
2	22002	66	65	70	74	78	86	87	96	88	82	90	86
3	22003	60	58	54	61	54	57	64	62	72	63	72	76
4	22004	99	95	96	93	97	89	92	98	91	98	95	88
5	22005	41	36	35	28	35	36	27	26	19	22	27	31
6	22006	47	50	47	57	62	64	71	75	85	87	85	89
7	22007	84	74	70	68	58	59	56	56	64	70	67	59
8	22008	74	64	58	57	53	51	47	45	42	43	34	24
9	22009	87	81	73	74	71	63	53	45	39	43	46	38

Data Cleaning And Pre-Processing

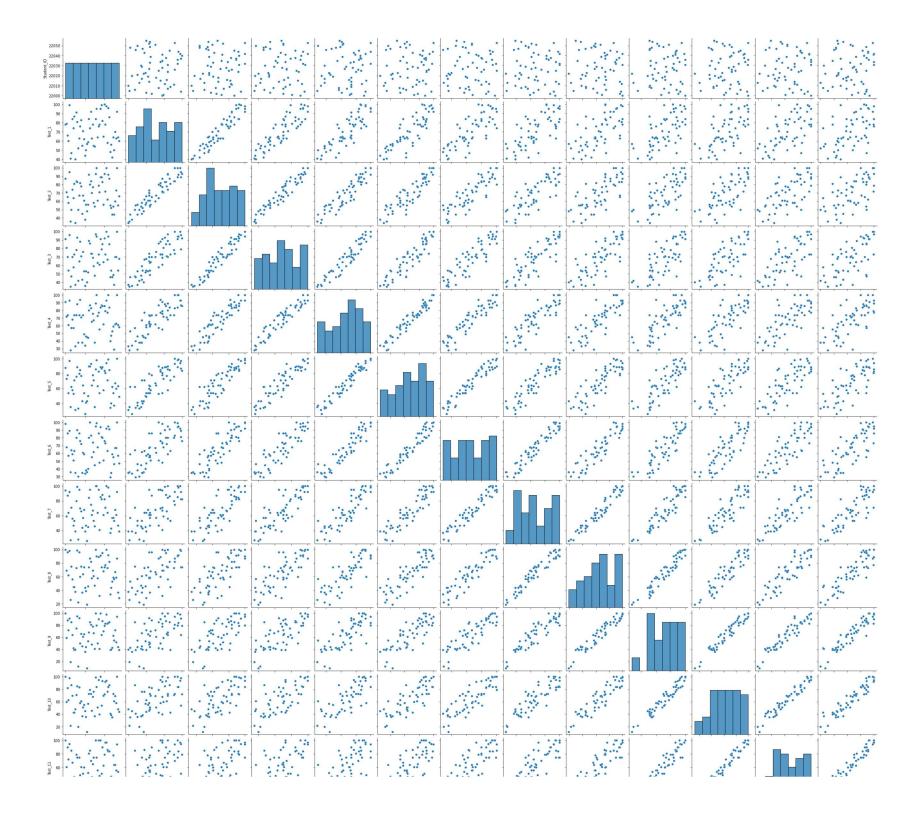
```
In [4]:
         1 df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 56 entries, 0 to 55
        Data columns (total 13 columns):
                         Non-Null Count Dtype
             Column
             Student_ID 56 non-null
                                         int64
                         56 non-null
                                         int64
         1
             Test 1
             Test_2
                         56 non-null
                                         int64
                         56 non-null
                                         int64
             Test 3
                         56 non-null
                                         int64
             Test 4
                         56 non-null
                                         int64
             Test 5
                         56 non-null
                                         int64
             Test 6
            Test 7
                         56 non-null
                                         int64
             Test 8
                         56 non-null
                                         int64
                         56 non-null
             Test 9
                                         int64
         10 Test 10
                         56 non-null
                                         int64
         11 Test 11
                         56 non-null
                                         int64
         12 Test 12
                         56 non-null
                                         int64
        dtypes: int64(13)
        memory usage: 5.8 KB
```

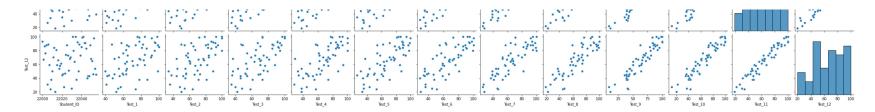
```
In [5]:
           1 # Display the statistical summary
           2 df.describe()
Out[5]:
                                                                                                                                         Т
                   Student ID
                                   Test 1
                                              Test 2
                                                          Test_3
                                                                     Test_4
                                                                                Test 5
                                                                                            Test 6
                                                                                                       Test_7
                                                                                                                   Test 8
                                                                                                                              Test 9
                    56.000000
                                56.000000
                                           56.000000
                                                       56.000000
                                                                  56.000000
                                                                              56.000000
                                                                                         56.000000
                                                                                                    56.000000
                                                                                                                56.000000
                                                                                                                            56.000000
                                                                                                                                       56.0
          count
           mean 22027.500000
                                70.750000
                                           69.196429
                                                       68.089286
                                                                  67.446429
                                                                              67.303571
                                                                                         66.000000
                                                                                                    66.160714
                                                                                                                65.303571
                                                                                                                            64.392857
                                                                                                                                       64.2
                    16.309506
                                                       18.838333
                                                                              20.746890
                                                                                                    21.427914
                                                                                                                22.728372
                                                                                                                                       22.5
             std
                                17.009356
                                           17.712266
                                                                  19.807179
                                                                                         21.054043
                                                                                                                            23.211814
                 22000.000000
                                           34.000000
                                                                  28.000000
                                                                              26.000000
                                                                                         29.000000
                                                                                                                19.000000
                                                                                                                            9.000000
                                                                                                                                       12.0
                                40.000000
                                                       35.000000
                                                                                                    26.000000
                 22013.750000
                                57.750000
                                           55.750000
                                                       53.000000
                                                                  54.500000
                                                                              53.750000
                                                                                         50.250000
                                                                                                    47.000000
                                                                                                                45.750000
                                                                                                                            44.000000
                                                                                                                                       45.7
            50%
                 22027.500000
                                           68.500000
                                                       70.000000
                                                                              69.000000
                                                                                         65.500000
                                                                                                                67.500000
                                                                                                                            65.500000
                                                                                                                                       65.5
                                70.500000
                                                                  71.500000
                                                                                                    64.000000
                22041.250000
                                           83.250000
                                                                              85.250000
                                                                                         83.750000
                                                                                                                           84.250000
                                                                                                                                       83.2
            75%
                                84.000000
                                                       85.000000
                                                                  84.000000
                                                                                                    85.250000
                                                                                                                83.250000
            max 22055.000000
                               100.000000 100.000000
                                                      100.000000
                                                                 100.000000
                                                                            100.000000
                                                                                        100.000000
                                                                                                   100.000000
                                                                                                               100.000000
                                                                                                                          100.000000
                                                                                                                                     100.0
           1 # To display the col headings
In [6]:
             df.columns
Out[6]: 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')
               cols=df.dropna(axis=1)
In [7]:
              cols.columns
In [8]:
Out[8]: 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')
```

EDA and Visualization

```
In [9]: 1 sns.pairplot(cols)
```

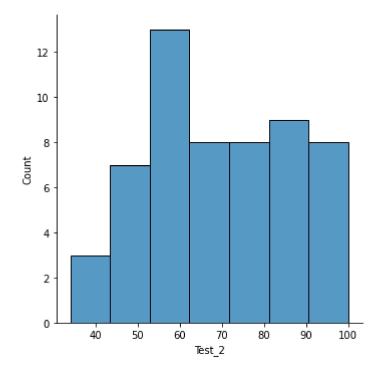
Out[9]: <seaborn.axisgrid.PairGrid at 0x132f1124b20>





In [11]: | 1 | sns.displot(df['Test_2'])

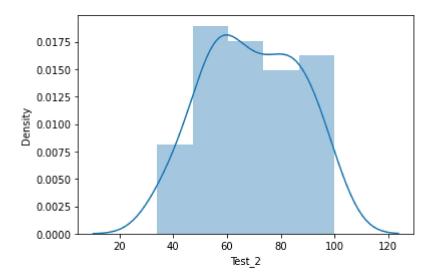
Out[11]: <seaborn.axisgrid.FacetGrid at 0x132ed32d580>



C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `distplot` is a dep recated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[12]: <AxesSubplot:xlabel='Test_2', ylabel='Density'>



Out[13]:

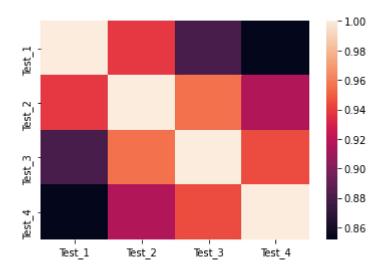
	Test_1	Test_2	Test_3	Test_4
0	78	87	91	91
1	79	71	81	72
2	66	65	70	74
3	60	58	54	61
4	99	95	96	93
5	41	36	35	28
6	47	50	47	57
7	84	74	70	68
8	74	64	58	57
9	87	81	73	74
10	40	34	37	33
11	91	84	78	74
12	81	83	93	88
13	52	50	42	38
14	63	67	65	74
15	76	82	88	94
16	83	78	71	71
17	55	45	43	38
18	71	67	76	74
19	62	61	53	49
20	44	38	36	34
21	50	56	53	46
22	57	48	40	45
23	59	56	52	44
24	84	92	89	80
25	74	80	86	87

	Test_1	Test_2	Test_3	Test_4
26	92	84	74	83
27	63	70	74	65
28	78	77	69	76
29	55	58	59	67
30	54	54	48	38
31	84	93	97	89
32	95	100	94	100
33	64	61	63	73
34	76	79	73	77
35	78	71	61	55
36	95	89	91	84
37	99	89	79	87
38	82	83	85	86
39	65	56	64	62
40	100	93	92	86
41	78	72	73	79
42	98	100	100	93
43	58	62	67	77
44	96	92	94	100
45	86	87	85	84
46	48	55	46	40
47	56	52	54	47
48	42	44	46	53
49	64	54	49	59
50	50	44	37	29
51	70	60	70	62
52	63	73	70	63

	Test_1	Test_2	Test_3	Test_4
53	92	100	100	100
54	64	55	54	61
55	60	66	68	58

```
In [14]: 1 sns.heatmap(df1.corr())
```

Out[14]: <AxesSubplot:>



To train the model - MODEL BUILD

Going to train linear regression model; We split our data into 2 variables x and y where x is independent var(input) and y is dependent on x(output), we could ignore address col as it is not required for our model

To split the dataset into test data

```
In [17]:
           1 # importing lib for splitting test data
           2 from sklearn.model_selection import train_test_split
           1 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
In [18]:
In [19]:
           1 from sklearn.linear_model import LinearRegression
           3 lr=LinearRegression()
           4 lr.fit(x_train,y_train)
Out[19]: LinearRegression()
In [20]:
           1 print(lr.intercept )
         [-4.26325641e-14]
In [21]:
           1 print(lr.score(x_test,y_test))
         1.0
In [22]:
           1 coeff=pd.DataFrame(lr.coef_)
           2 coeff
Out[22]:
                         1
                                    2
                                                3
```

0 1.0 3.025439e-16 2.716402e-16 -1.050204e-16

```
In [28]:
           1 pred = lr.predict(x_test)
           plt.scatter(y_test,pred)
Out[28]: <matplotlib.collections.PathCollection at 0x132fb394250>
          100
           90
           80
           70
           60
           50
           40
                            60
                                   70
                                          80
                                                 90
                                                       100
                      50
           1 from sklearn.linear_model import Ridge,Lasso
In [24]:
In [25]:
           1 rr=Ridge(alpha=10)
           2 rr.fit(x_train,y_train)
Out[25]: Ridge(alpha=10)
In [26]:
           1 rr.score(x_test,y_test)
Out[26]: 0.999987927916574
In [27]:
           1 la=Lasso(alpha=10)
           2 la.fit(x_train,y_train)
Out[27]: Lasso(alpha=10)
In [29]:
           1 la.score(x_test,y_test)
Out[29]: 0.9985030373688533
```

In []: 1