```
In [21]:
          import pandas as pd
          import matplotlib.pyplot as plt
          from sklearn.linear_model import LinearRegression as LR
          from sklearn.model_selection import train_test_split as TTS
          from sklearn.metrics import mean_squared_error as MSE,mean_absolute_error as MAE,r2_score as R
          # Load dataset
 In [5]:
          df=pd.read_csv(r'D:\Intern\SkillCraft Intern\House pred.csv')
 In [6]: | df.head()
 Out[6]:
              bedrooms bathrooms sqft_living sqft_lot floors waterfront view condition sqft_above sqft_basement yr_buil
                     3
                                                                                    3
           0
                              1.50
                                        1340
                                                7912
                                                         1.5
                                                                    0
                                                                          0
                                                                                            1340
                                                                                                             0
                                                                                                                   195
           1
                     5
                              2.50
                                        3650
                                                9050
                                                         2.0
                                                                    0
                                                                          4
                                                                                    5
                                                                                            3370
                                                                                                           280
                                                                                                                   192
           2
                     3
                              2.00
                                        1930
                                               11947
                                                         1.0
                                                                    0
                                                                          0
                                                                                    4
                                                                                            1930
                                                                                                             0
                                                                                                                   196
           3
                     3
                              2.25
                                        2000
                                                8030
                                                         1.0
                                                                    0
                                                                          0
                                                                                    4
                                                                                            1000
                                                                                                          1000
                                                                                                                   196
                     4
                              2.50
                                        1940
                                               10500
                                                         1.0
                                                                    0
                                                                          0
                                                                                    4
                                                                                            1140
                                                                                                           800
                                                                                                                   197
 In [7]: df.tail()
 Out[7]:
                           bathrooms sqft_living sqft_lot floors waterfront view condition sqft_above sqft_basement
           4595
                        3
                                 1.75
                                           1510
                                                   6360
                                                           1.0
                                                                       0
                                                                             0
                                                                                       4
                                                                                               1510
                                                                                                                0
           4596
                        3
                                 2.50
                                           1460
                                                   7573
                                                           2.0
                                                                       0
                                                                             0
                                                                                       3
                                                                                               1460
                                                                                                                0
           4597
                        3
                                 2.50
                                           3010
                                                   7014
                                                                       0
                                                                             0
                                                                                       3
                                                                                               3010
                                                                                                                0
                                                           2.0
           4598
                        4
                                 2.00
                                           2090
                                                   6630
                                                           1.0
                                                                       0
                                                                             0
                                                                                       3
                                                                                               1070
                                                                                                             1020
           4599
                        3
                                 2.50
                                           1490
                                                   8102
                                                           2.0
                                                                       0
                                                                             0
                                                                                       4
                                                                                               1490
                                                                                                                0
 In [8]:
          df.isnull().sum()
 Out[8]: bedrooms
                             0
          bathrooms
                             0
          sqft_living
                             0
          sqft_lot
                             0
                             0
          floors
          waterfront
                             0
          view
                             0
          condition
                             0
          sqft_above
                             0
          sqft basement
                             0
          yr_built
                             0
          price
                             0
```

dtype: int64

```
In [9]: df.describe()
```

Out[9]:

		bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condition	
_	count	4600.000000	4600.000000	4600.000000	4.600000e+03	4600.000000	4600.000000	4600.000000	4600.000000	4
	mean	3.400870	2.160815	2139.346957	1.485252e+04	1.512065	0.007174	0.240652	3.451739	1
	std	0.908848	0.783781	963.206916	3.588444e+04	0.538288	0.084404	0.778405	0.677230	
	min	0.000000	0.000000	370.000000	6.380000e+02	1.000000	0.000000	0.000000	1.000000	
	25%	3.000000	1.750000	1460.000000	5.000750e+03	1.000000	0.000000	0.000000	3.000000	1
	50%	3.000000	2.250000	1980.000000	7.683000e+03	1.500000	0.000000	0.000000	3.000000	1
	75%	4.000000	2.500000	2620.000000	1.100125e+04	2.000000	0.000000	0.000000	4.000000	2
	max	9.000000	8.000000	13540.000000	1.074218e+06	3.500000	1.000000	4.000000	5.000000	ĉ

```
In [11]: # Split into Indepent and Dependent Values
    x=df.drop(['price'],axis=1)
    y=df['price']
```

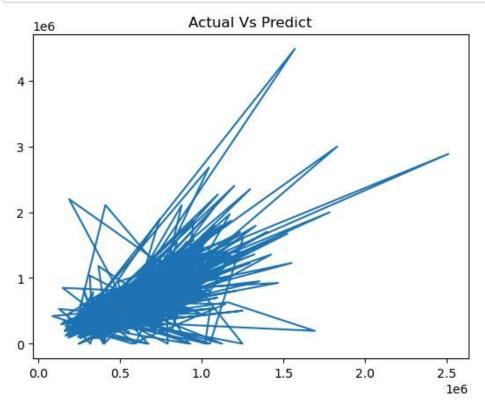
```
In [12]: # Data Split Into Train And Test
x_train,x_test,y_train,y_test=TTS(x,y,test_size=0.3,random_state=203)
```

```
In [13]: lm=LR()
lm.fit(x_train,y_train)
```

```
Out[13]: v LinearRegression
LinearRegression()
```

```
In [14]: y_pred=lm.predict(x_test)
```

```
In [20]: # Visualize
plt.plot(y_pred,y_test)
plt.title('Actual Vs Predict')
plt.show()
```



```
In [22]: # evalute its Perfomance Using Metrics
    mse=MSE(y_pred,y_test)
    mae=MAE(y_pred,y_test)
    rs=RS(y_pred,y_test)
    print(f"Mean Squared Error = {mse}")
    print(f"Mean Absolute Error = {mae}")
    print(f"R2 Score = {rs}")
```

Mean Squared Error = 63187196340.68652 Mean Absolute Error = 158397.51460542338 R2 Score = -0.04511278779316674

```
In [23]: #Enter the New Data to Predict import numpy as np
```

```
In [27]: new_data=np.array([[3,2,3400,7910,2,1,0,3,1300,200,2000]])
print(f"Predicted House Price = {lm.predict(new_data)[0]}")
```

Predicted House Price = 934260.4190417998

C:\Users\nikhil\anaconda3\Lib\site-packages\sklearn\base.py:464: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names warnings.warn(