

# Data Analysis

## Linear Regressions Analysis (USA\_HOUSING)

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In [57]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
from sklearn.model_selection import train_test_split
```

In [2]:

```
US_house = pd.read_csv("./USA_Housing.csv")
```

In [3]:

```
print(f"Data Shape: {US_house.shape}")
```

Data Shape: (5000, 7)

In [4]:

```
print(US_house.columns)
```

```
Index(['Avg. Area Income', 'Avg. Area House Age', 'Avg. Area Number of
Rooms',
      'Avg. Area Number of Bedrooms', 'Area Population', 'Price', 'Add
ress'],
      dtype='object')
```

**first five rows of the data set**

In [5]:

```
US_house.head(5)
```

Out[5]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Addre
0	79545.458574	5.682861	7.009188	4.09	23086.800503	1.059034e+06	208 Michael Ferry A 674\nLaurabury, M 3701
1	79248.642455	6.002900	6.730821	3.09	40173.072174	1.505891e+06	188 Johnson Vie Suite 079\nLa Kathleen, CA
2	61287.067179	5.865890	8.512727	5.13	36882.159400	1.058988e+06	9127 Elizabe Stravenue\nDanieltow WI 06482
3	63345.240046	7.188236	5.586729	3.26	34310.242831	1.260617e+06	USS Barnett\nFPO / 448
4	59982.197226	5.040555	7.839388	4.23	26354.109472	6.309435e+05	USNS Raymond\nFF AE 093

Description & Info

In [6]:

```
print(US_house.description)
```

```
-----  
----  
AttributeError                                Traceback (most recent call 1  
ast)  
<ipython-input-6-8181bf280ed1> in <module>  
----> 1 print(US_house.description)  
  
~/.local/share/virtualenvs/401-YopfmMgc/lib/python3.7/site-packages/panda  
das/core/generic.py in __getattr__(self, name)  
    5177         if self._info_axis._can_hold_identifiers_and_holds_  
name(name):  
    5178             return self[name]  
-> 5179         return object.__getattr__(self, name)  
    5180  
    5181     def __setattr__(self, name, value):  
  
AttributeError: 'DataFrame' object has no attribute 'description'
```

In [7]:

```
print(US_house.info)
```

```
<bound method DataFrame.info of
Age  Avg. Area Number of Rooms  \
0      79545.458574      5.682861      7.009188
1      79248.642455      6.002900      6.730821
2      61287.067179      5.865890      8.512727
3      63345.240046      7.188236      5.586729
4      59982.197226      5.040555      7.839388
...
4995    60567.944140      7.830362      6.137356
4996    78491.275435      6.999135      6.576763
4997    63390.686886      7.250591      4.805081
4998    68001.331235      5.534388      7.130144
4999    65510.581804      5.992305      6.792336

      Avg. Area Number of Bedrooms  Area Population      Price  \
0              4.09      23086.800503  1.059034e+06
1              3.09      40173.072174  1.505891e+06
2              5.13      36882.159400  1.058988e+06
3              3.26      34310.242831  1.260617e+06
4              4.23      26354.109472  6.309435e+05
...
4995          3.46      22837.361035  1.060194e+06
4996          4.02      25616.115489  1.482618e+06
4997          2.13      33266.145490  1.030730e+06
4998          5.44      42625.620156  1.198657e+06
4999          4.07      46501.283803  1.298950e+06

      Address
0  208 Michael Ferry Apt. 674\nLaurabury, NE 3701...
1  188 Johnson Views Suite 079\nLake Kathleen, CA...
2  9127 Elizabeth Stravenue\nDanielstown, WI 06482...
3  USS Barnett\nFPO AP 44820
4  USNS Raymond\nFPO AE 09386
...
4995  USNS Williams\nFPO AP 30153-7653
4996  PSC 9258, Box 8489\nAPO AA 42991-3352
4997  4215 Tracy Garden Suite 076\nJoshualand, VA 01...
4998  USS Wallace\nFPO AE 73316
4999  37778 George Ridges Apt. 509\nEast Holly, NV 2...

[5000 rows x 7 columns]>
```

## Comparisons Between Ave. Income & Price of House

While Test Size is 20%

In [8]:

```
X = US_house['Avg. Area Income'].values  
y = US_house['Price'].values
```

In [9]:

```
X
```

Out[9]:

```
array([79545.45857432, 79248.64245483, 61287.06717866, ...,  
       63390.6868855 , 68001.33123509, 65510.58180367])
```

In [14]:

```
X=X.reshape(-1,1)  
X
```

Out[14]:

```
array([[79545.45857432],  
       [79248.64245483],  
       [61287.06717866],  
       ...,  
       [63390.6868855 ],  
       [68001.33123509],  
       [65510.58180367]])
```

In [15]:

```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.8,test_size=0.2,random_state=100)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shape {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```

```
X_train shape (4000, 1)
y_train shape (4000,)
X_test shape (1000, 1)
y_test shape (1000,)
[ 800809.13168613 1620949.4369106 1219637.36651886 1172133.48684149
  864483.75849638 1153135.22029401 1178272.29747728 821859.06572785
 1266947.26397664 1636559.24060426 1085494.82011087 995783.16248321
 1096938.42790984 1702528.79401861 1443027.26306137 826306.14792423
 2115010.70359761 1112286.04638504 1094069.79812154 905328.7691457
 782656.08041957 1341722.83596915 2119176.26193328 1543966.42644823
 717213.26879895 1538039.5989439 1823498.40687621 964596.74983993
 1593331.70871016 877247.24536775 1111108.50121858 1313304.58773627
 1123753.73420143 1376714.67419885 1988563.99326237 1874415.10704969
 1732196.21679656 1502443.22975094 1078779.49709872 705100.51155545
 1595620.55713363 1814462.35050116 1610577.49664142 1012262.71165943
 1212939.95351759 934408.71428136 1363086.90453256 1450996.03817929
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 1223777.49762374 1100152.08776636 1179440.83225322 1891398.25582299
 889113.23886125 671960.64489301 716771.00574888 1046030.11422825
 902520.9400336 1795093.03400894 1146637.54384348 1131225.75970326
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 1253609.76404087 1412626.51874178 1031367.58851648 903657.56254793
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 853750.65294879 973068.55703495 1046442.63447414 676738.71554521
 1496724.38060648 1729392.20416728 664978.87377689 1820189.53280409
 839638.46096435 1022709.69550666 1128403.36605923 927163.81194845
 980141.21910767 1196996.68108546 1445527.13666544 901881.74268624
 1552915.16448986 1598593.7326674 1057252.5826211 909781.46676822
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 1392793.81424375 828120.95498882 1604207.68427662 1942640.26532281
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 1136448.40876046 799124.84915754 1617405.42473352 1925615.84080549
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 1280199.2946554 1228810.74504552 685880.32127708 1220591.00604205
 1329273.22764986 401148.56879138 1134273.42452143 1428247.07418468
 1539465.41967739 831762.7909034 1116730.86697073 1131698.21419536
 1212205.34065262 2249122.54133519 1299991.95081201 1142264.24767841
 1671350.14009336 1053338.5923718 1456969.57682195 1832230.09273646
 1390377.37493619 1237902.82007375 1784260.84872194 814784.24572801]
```

1020098.04138186	1280910.189933	999814.88981408	1167421.73902776
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599504.0192867	1309922.85417634	1279160.98758547	1290784.49085966
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1550036.518714	989643.03138901	867044.95964192	1022781.17074291
1276259.0183221	1363623.04420837	1224454.54742604	921396.97525415
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1641473.66234136	1656080.08908926	935061.30929776	879356.36604829
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1242316.9909078	1236633.08656358	1070318.81489593	1162469.88691582
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1299430.1759554	1591188.34848729	1431507.62342797	1296146.93885966
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1277780.51220943	1501622.1254996	1560693.09827186	1453974.50595087
622449.64522892	652991.10542482	1440246.32987309	1256086.50705911
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891468.21327911	1299126.82985549	1060193.78588461	778836.92290454
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```

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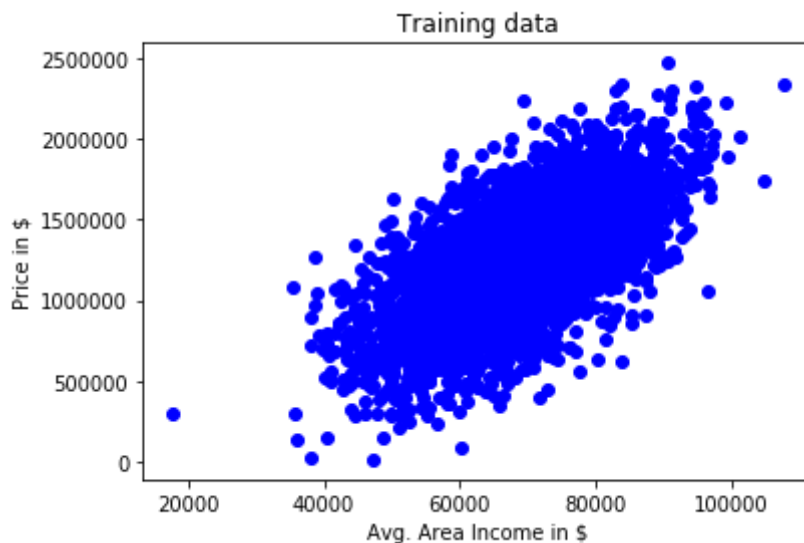
```

In [16]:

```

%matplotlib inline
plt.scatter(x_train,y_train,color='blue')
plt.xlabel('Avg. Area Income in $' )
plt.ylabel('Price in $')
plt.title('Training data')
plt.show()

```



In [17]:

```
lm = LinearRegression()
lm.fit(x_train,y_train)
y_predict = lm.predict(x_test)
print(f"Train accuracy {round(lm.score(x_train,y_train)*100,2)} %")
print(f"Test accuracy {round(lm.score(x_test,y_test)*100,2)} %")
```

Train accuracy 40.09 %

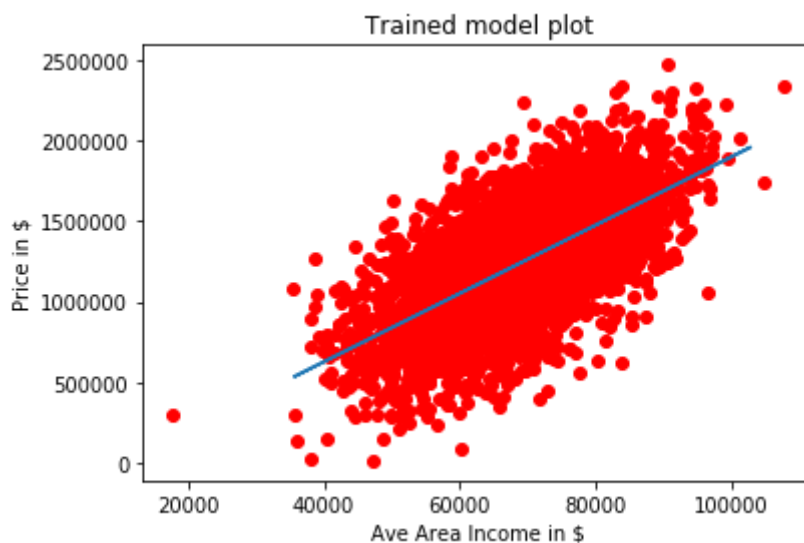
Test accuracy 44.23 %

In [18]:

```
plt.scatter(x_train,y_train,color='red')
plt.plot(x_test,y_predict)
plt.xlabel("Ave Area Income in $")
plt.ylabel("Price in $")
plt.title("Trained model plot")
plt.plot
```

Out[18]:

```
<function matplotlib.pyplot.plot(*args, scalex=True, scaley=True, data=
None, **kwargs)>
```



## Comparisons Between Ave. Income & Price of House

While Test Size is 10%

In [19]:

```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.9,test_size=0.1,random_state=500)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shape {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```

```
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y_train shape (4500,)
X_test shape (500, 1)
y_test shape (500,)
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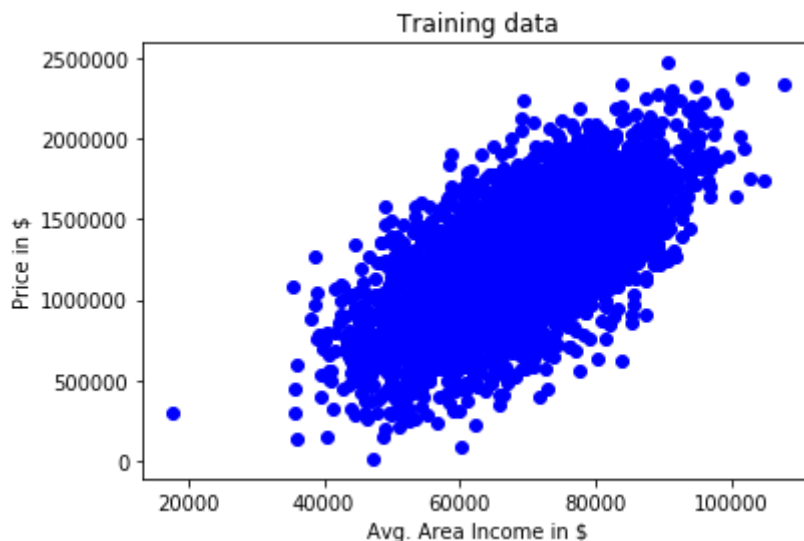
```

In [20]:

```

%matplotlib inline
plt.scatter(x_train,y_train,color='blue')
plt.xlabel('Avg. Area Income in $' )
plt.ylabel('Price in $')
plt.title('Training data')
plt.show()

```



In [21]:

```

lm = LinearRegression()
lm.fit(x_train,y_train)
y_predict = lm.predict(x_test)
print(f"Train accuracy {round(lm.score(x_train,y_train)*100,2)} %")
print(f"Test accuracy {round(lm.score(x_test,y_test)*100,2)} %")

```

Train accuracy 40.97 %

Test accuracy 40.23 %

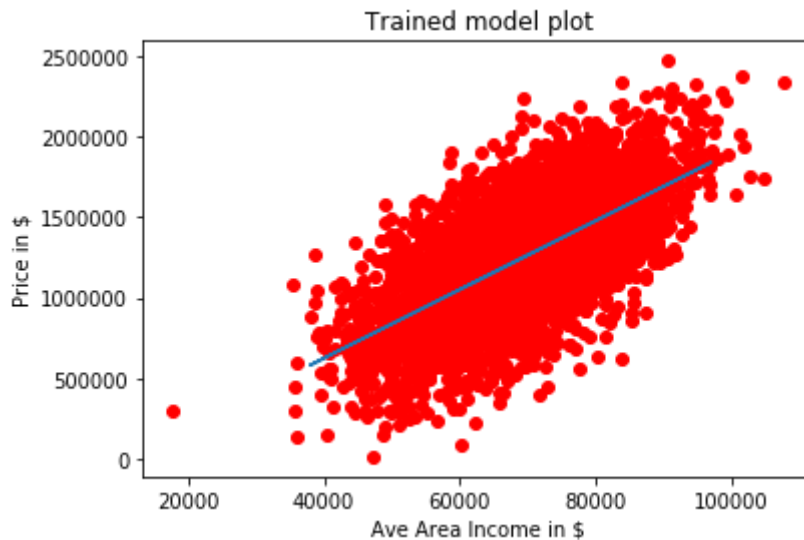


In [22]:

```
plt.scatter(x_train,y_train,color='red')
plt.plot(x_test,y_predict)
plt.xlabel("Ave Area Income in $")
plt.ylabel("Price in $")
plt.title("Trained model plot")
plt.plot
```

Out[22]:

```
<function matplotlib.pyplot.plot(*args, scalex=True, scaley=True, data=
None, **kwargs)>
```



## Comparisons Between Ave. Income & Price of House

### While Test Size is 30%

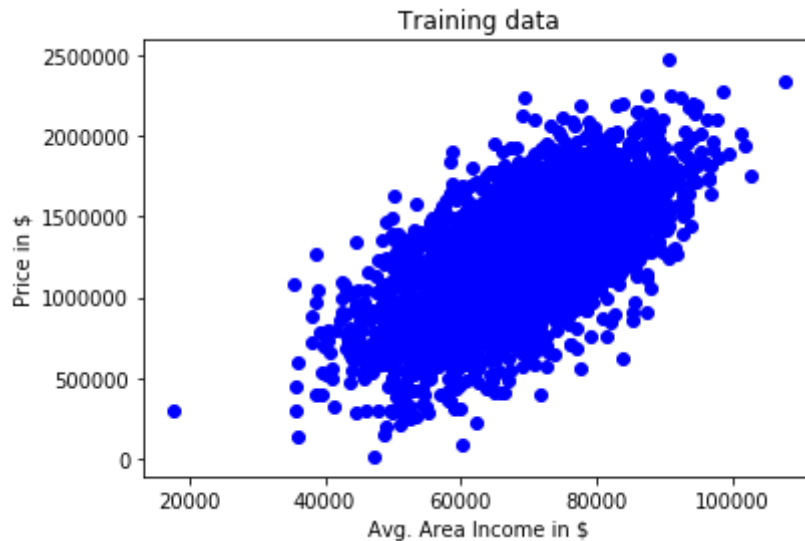
In [23]:

```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.7,test_size=0.
3,random_state=100000)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shape {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```

```
X_train shape (3500, 1)
y_train shape (3500,)
X_test shape (1500, 1)
y_test shape (1500,)
[1525533.40662551 1369006.11488144 1096912.03579858 ... 1154917.3936116
4
1214262.69927372 1023595.11287068]
```

In [24]:

```
%matplotlib inline
plt.scatter(x_train,y_train,color='blue')
plt.xlabel('Avg. Area Income in $' )
plt.ylabel('Price in $')
plt.title('Training data')
plt.show()
```



In [25]:

```
lm = LinearRegression()
lm.fit(x_train,y_train)
y_predict = lm.predict(x_test)
print(f"Train accuracy {round(lm.score(x_train,y_train)*100,2)} %")
print(f"Test accuracy {round(lm.score(x_test,y_test)*100,2)} %")
```

Train accuracy 40.03 %

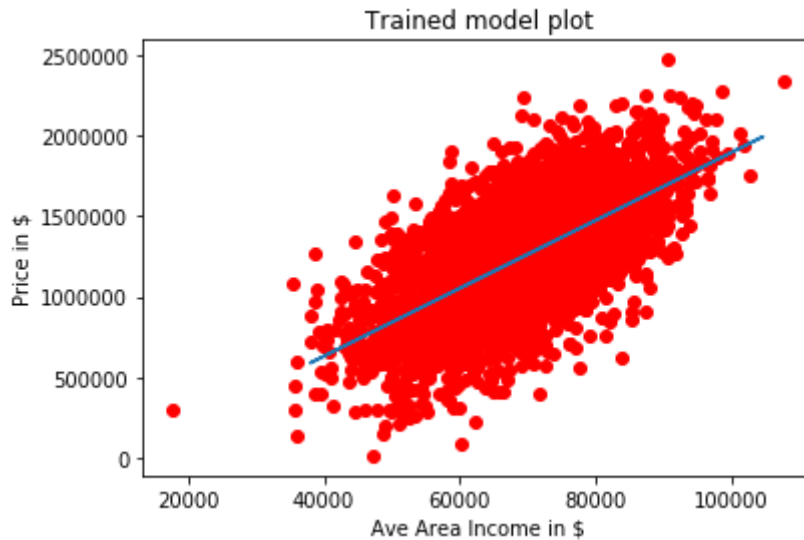
Test accuracy 42.94 %

In [26]:

```
plt.scatter(x_train,y_train,color='red')
plt.plot(x_test,y_predict)
plt.xlabel("Ave Area Income in $")
plt.ylabel("Price in $")
plt.title("Trained model plot")
plt.plot
```

Out[26]:

```
<function matplotlib.pyplot.plot(*args, scalex=True, scaley=True, data=
None, **kwargs)>
```



## Comparisons Price of House & Area Population

While Test Size is 10%

In [27]:

```
X = US_house['Area Population'].values
y = US_house['Price'].values
```

In [28]:

```
X
```

Out[28]:

```
array([23086.80050269, 40173.07217364, 36882.1593997 , ...,
       33266.14548965, 42625.62015585, 46501.28380314])
```

In [29]:

```
X=X.reshape(-1,1)  
X
```

Out[29]:

```
array([[23086.80050269],  
       [40173.07217364],  
       [36882.1593997 ],  
       ...,  
       [33266.14548965],  
       [42625.62015585],  
       [46501.28380314]])
```

In [30]:

```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.9,test_size=0.1,random_state=100)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shap {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```

```
X_train shape (4500, 1)
y_train shape (4500,)
X_test shap (500, 1)
y_test shape (500,)
[ 800809.13168613 1620949.4369106 1219637.36651886 1172133.48684149
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```

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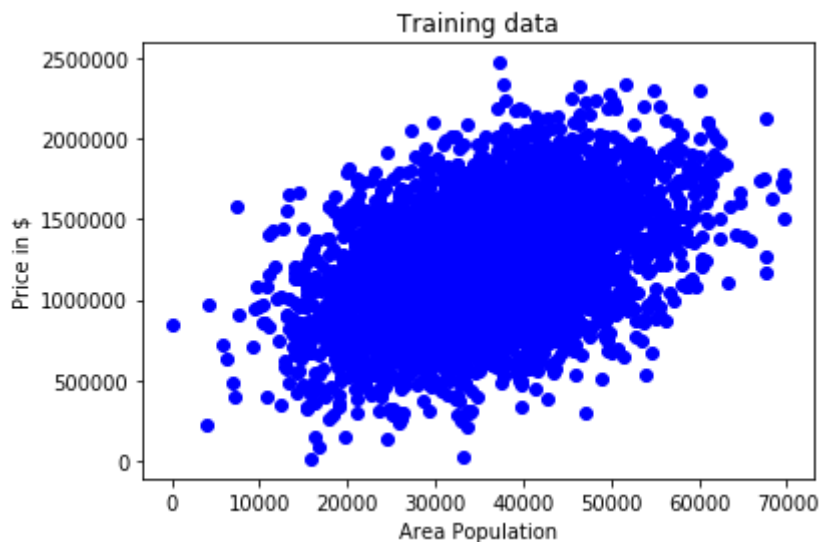
```

In [31]:

```

%matplotlib inline
plt.scatter(x_train,y_train,color='blue')
plt.xlabel('Area Population')
plt.ylabel('Price in $')
plt.title('Training data')
plt.show()

```



In [32]:

```

lm = LinearRegression()
lm.fit(x_train,y_train)
y_predict = lm.predict(x_test)
print(f"Train accuracy {round(lm.score(x_train,y_train)*100,2)} %")
print(f"Test accuracy {round(lm.score(x_test,y_test)*100,2)} %")

```

```

Train accuracy 17.11 %
Test accuracy 12.9 %

```

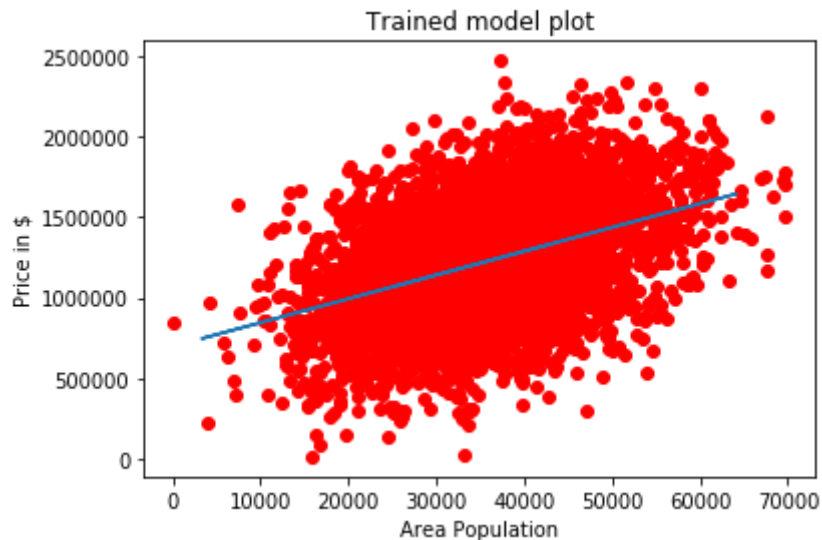


In [33]:

```
plt.scatter(x_train,y_train,color='red')
plt.plot(x_test,y_predict)
plt.xlabel("Area Population")
plt.ylabel("Price in $")
plt.title("Trained model plot")
plt.plot
```

Out[33]:

```
<function matplotlib.pyplot.plot(*args, scalex=True, scaley=True, data=
None, **kwargs)>
```



## Comparisons Price of House & Area Population

While Test Size is 20%

In [34]:

```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.8,test_size=0.2,random_state=1000)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shap {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```

```
X_train shape (4000, 1)
y_train shape (4000,)
X_test shape (1000, 1)
y_test shape (1000,)
[1277380.52860853  849153.12252829 1224397.42724927 2016910.73970733
 956435.26333485  979568.62854125 1852584.52301333 1246410.34057798
1419345.60730258 1273629.12431665 1942640.26532281 1270297.54493105
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```

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1623100.83987563	627650.40257098	1161232.65746201	1210184.01761757
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802846.12229537	1643291.72247874	1163741.65830081	1166750.31285964
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1356062.10449866	1358213.90612806	1404435.50461839	1156786.0838811
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988175.00739231	1046721.97617325	1011909.69443135	1410010.99400056
1341069.15980991	718874.67976529	1363086.90453256	1245070.33791015
714166.39780611	1340343.85651884	964596.74983993	864483.75849638
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1892623.54590422	914231.29230679	1315666.87392322	476971.45594277
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743999.8191602	1845630.07082289	997448.72807699	1794214.42474246
799207.70427504	757943.22282057	912585.62116925	1152145.9120119
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1591234.77340678	1654965.29571208	813415.12806775	1175781.41757006
1699952.81444788	496359.97079138	1031737.38481276	987026.37074102

```

1080624.49716572 1608726.68054645 1702528.79401861 1511526.92112864
1592768.24182693 1637841.26200846 1410814.98469241 708045.57492055
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619664.25116439 1375467.27099677 1406865.49456075 1388596.243304
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1646033.41018364 991398.82186552 889113.23886125 832691.17813085
1213530.84986587 772111.9721021 1285923.73421191 829794.82329338
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483986.10899253 1222277.09102041 1101120.25874283 1737759.05002945
1795093.03400894 1048818.21025016 940162.72447779 1775874.76010401
1706291.99055932 1522083.94415309 1317234.04287959 1433542.02127752
1276448.79204297 1134273.42452143 1529281.84011934 1128403.36605923
573434.66423432 1131040.70232425 1133139.92617632 1636414.89207332]

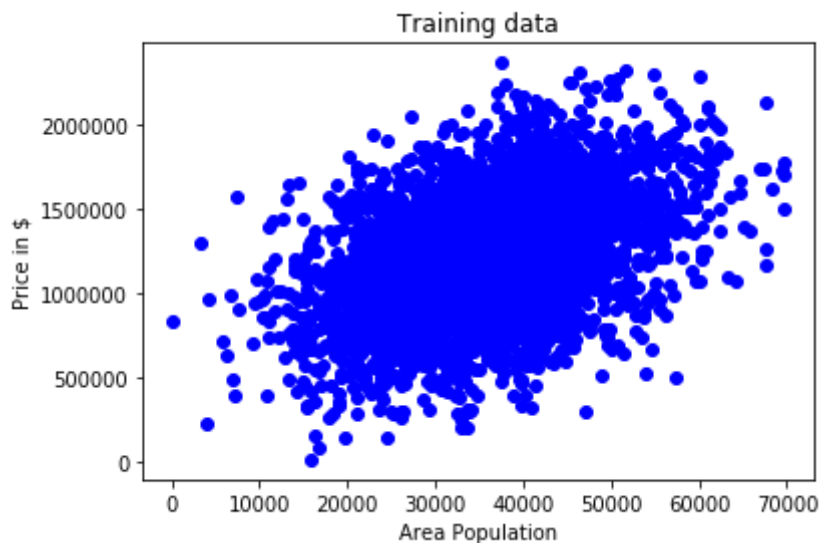
```

In [35]:

```

%matplotlib inline
plt.scatter(x_train,y_train,color='blue')
plt.xlabel('Area Population')
plt.ylabel('Price in $')
plt.title('Training data')
plt.show()

```



In [36]:

```
lm = LinearRegression()
lm.fit(x_train,y_train)
y_predict = lm.predict(x_test)
print(f"Train accuracy {round(lm.score(x_train,y_train)*100,2)} %")
print(f"Test accuracy {round(lm.score(x_test,y_test)*100,2)} %")
```

Train accuracy 17.33 %

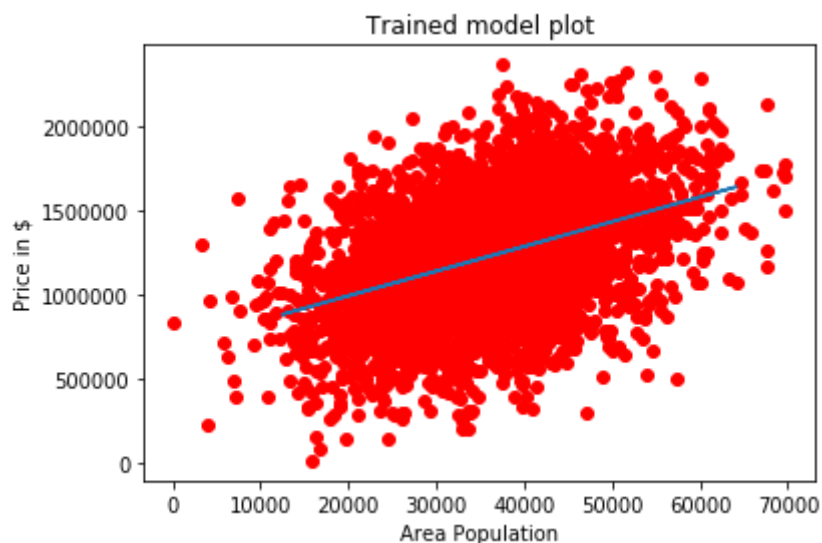
Test accuracy 14.17 %

In [37]:

```
plt.scatter(x_train,y_train,color='red')
plt.plot(x_test,y_predict)
plt.xlabel("Area Population")
plt.ylabel("Price in $")
plt.title("Trained model plot")
plt.plot
```

Out[37]:

```
<function matplotlib.pyplot.plot(*args, scalex=True, scaley=True, data=
None, **kwargs)>
```



## Comparisons Price of House & Area Population

**While Test Size is 30%**



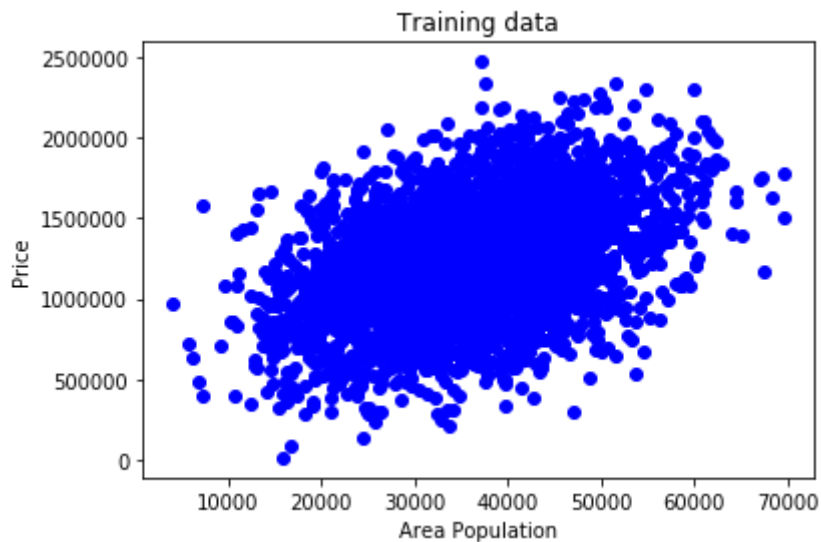
In [38]:

```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.7,test_size=0.3,random_state=100)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shap {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```

```
X_train shape (3500, 1)
y_train shape (3500,)
X_test shap (1500, 1)
y_test shape (1500,)
[ 800809.13168613 1620949.4369106 1219637.36651886 ... 1647216.5936644
 5
 412057.44010889 942508.96223867]
```

In [39]:

```
%matplotlib inline
plt.scatter(x_train,y_train,color='blue')
plt.xlabel('Area Population')
plt.ylabel('Price')
plt.title('Training data')
plt.show()
```



In [40]:

```
lm = LinearRegression()
lm.fit(x_train,y_train)
y_predict = lm.predict(x_test)
print(f"Train accuracy {round(lm.score(x_train,y_train)*100,2)} %")
print(f"Test accuracy {round(lm.score(x_test,y_test)*100,2)} %")
```

```
Train accuracy 16.72 %
Test accuracy 16.62 %
```

In [ ]:

```
plt.scatter(x_train,y_train,color='red')
plt.plot(x_test,y_predict)
plt.xlabel("Area Population")
plt.ylabel("Price in $")
plt.title("Trained model plot")
plt.plot
```

## Comparisons Ave. Income & Ave. Area of Bedrooms

### While Test Size is 10%

In [41]:

```
X = US_house['Avg. Area Income'].values
y = US_house['Avg. Area Number of Rooms'].values
```

In [42]:

X

Out[42]:

```
array([ 79545.45857432,  79248.64245483,  61287.06717866, ...,
        63390.6868855 ,  68001.33123509,  65510.58180367])
```

In [43]:

```
X=X.reshape(-1,1)
X
```

Out[43]:

```
array([[ 79545.45857432],
       [ 79248.64245483],
       [ 61287.06717866],
       ...,
       [ 63390.6868855 ],
       [ 68001.33123509],
       [ 65510.58180367]])
```

In [44]:

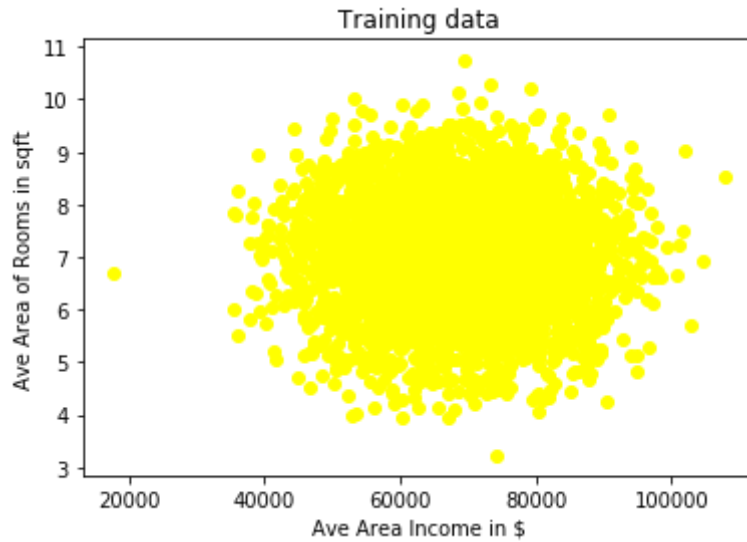
```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.9,test_size=0.1,random_state=1000)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shap {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```

```
X_train shape (4500, 1)
y_train shape (4500,)
X_test shap (500, 1)
y_test shape (500,)
[7.67098283 6.02452319 7.65651167 8.76661274 6.46252334 6.52222048
 8.50065931 6.03645802 7.13125578 7.22911211 7.88847287 5.96140678
 5.16486548 6.65816348 8.56001888 8.54176845 7.3182145 7.316551
 8.46146827 6.9096769 7.74683647 6.15220352 8.8615771 7.8410663
 8.4513549 8.38320811 7.80189707 7.47067235 7.67923057 6.73400688
 8.00644913 8.59824688 5.74405984 5.58679444 5.0427468 6.12940457
 8.84066791 7.12398966 7.97987782 7.50381655 6.13419293 7.87459883
 7.23704281 6.31106491 6.66953171 7.47284985 6.81364961 7.16132931
 7.54589692 5.65425637 6.79627685 6.5413757 7.03385222 8.50933677
 7.24990524 8.08412347 7.73486203 7.8452404 6.42408542 6.50394588
 6.73255173 7.27878258 7.87896295 7.29698406 7.03666701 8.25904432
 7.64896367 4.98021947 6.18284258 4.9673884 5.56435288 7.40068467
 8.00541729 6.9682825 5.78918815 6.11019169 6.94214348 6.91430889
 7.42768912 6.87615033 8.28908486 5.07967061 7.51138304 6.02020121
 7.08507236 6.04232838 5.71544106 6.68668613 5.01279289 8.91606569
 7.41078451 7.51820433 6.95202667 8.1052283 6.47600734 8.69531316
 5.22798817 6.809013 6.25567488 6.13548984 7.6987646 6.36475668
 8.30330169 6.70105955 6.9225152 5.67113634 4.67913425 7.13315074
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 6.36827869 7.57964206 7.79586819 7.4804088 7.18252748 7.4843437
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6.19461305 6.33024341]
```

In [45]:

```
%matplotlib inline
plt.scatter(x_train,y_train,color='yellow')
plt.xlabel('Ave Area Income in $')
plt.ylabel('Ave Area of Rooms in sqft')
plt.title('Training data')
plt.show()
```



In [46]:

```
lm = LinearRegression()
lm.fit(x_train,y_train)
y_predict = lm.predict(x_test)
print(f"Train accuracy {round(lm.score(x_train,y_train)*100,2)} %")
print(f"Test accuracy {round(lm.score(x_test,y_test)*100,2)} %")
```

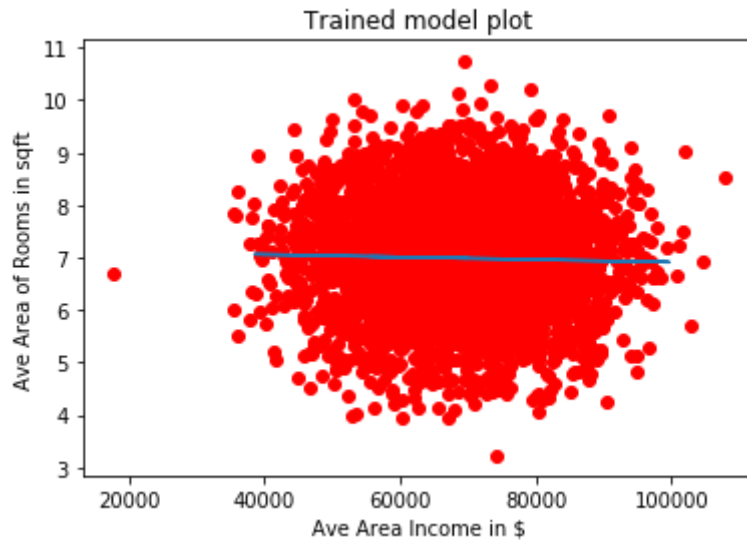
```
Train accuracy 0.07 %
Test accuracy -0.74 %
```

In [47]:

```
plt.scatter(x_train,y_train,color='red')
plt.plot(x_test,y_predict)
plt.xlabel('Ave Area Income in $')
plt.ylabel('Ave Area of Rooms in sqft')
plt.title("Trained model plot")
plt.plot
```

Out[47]:

```
<function matplotlib.pyplot.plot(*args, scalex=True, scaley=True, data=
None, **kwargs)>
```



## Comparisons Ave. Income & Ave. Area of Bedrooms

While Test Size is 20%

In [48]:

```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.8,test_size=0.2,random_state=1000)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shap {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```



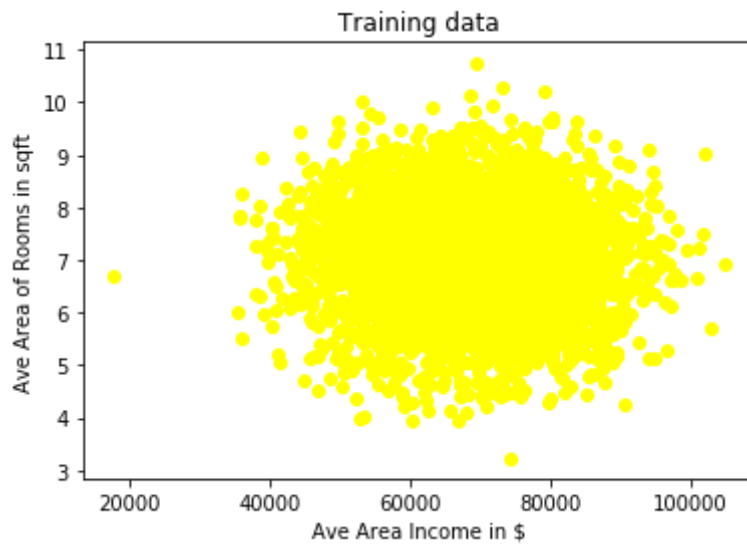
```
X_train shape (4000, 1)
y_train shape (4000,)
X_test shap (1000, 1)
y_test shape (1000,)
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5.34702706	5.56465345	6.28524674	8.00300675	8.95830017	6.91920414
6.49389932	6.46664525	6.28086382	8.51860818	7.98751262	7.66560184
8.31760044	7.06757467	4.81162263	5.62615918	6.45875588	7.02558811
5.55228986	6.47999368	7.60210723	6.36520575	6.9961141	8.54464981
7.07443434	7.31190652	7.87740359	6.86299437	6.2512391	6.99904497
7.81172447	5.52894386	5.91706446	6.45381582	8.4092963	7.546758
8.15953758	5.83173897	8.77666081	5.09352674	7.06620874	8.27198289
8.06591976	7.89947708	7.6934021	7.16078798	7.35488319	7.52135853
6.80887893	7.65416271	7.9095452	6.12246523	7.69016918	7.89415042
7.38555504	8.0662199	7.98951567	6.02916087	7.14649989	6.80707758
6.16415145	6.99769954	7.99781793	7.62644406	7.36045528	6.37606022
7.57044711	7.84737579	7.99904764	5.60480342	7.98194165	6.39488032
7.56019912	6.88291844	8.65734264	7.48193032	6.85153583	7.35028508
6.61798687	6.1005507	7.30991938	6.69199648	7.39790527	6.61175592
7.64873058	6.9713317	6.54525691	7.04743534	6.62495129	8.29983644
7.32537954	5.86554403	7.2825213	6.02929154	7.37561603	6.86452971
6.66988841	6.3811622	7.21482412	6.53579637	5.93570931	6.08728421
7.64833492	6.20372882	5.7853584	8.13509842	8.8227894	6.6449069
6.21178319	6.23031238	5.80740148	7.46488551	8.12645815	5.9086012
5.66960027	4.33488449	6.41531242	7.38360641	7.23666116	4.76590308
7.42787307	5.92121599	7.12902984	6.63822317	6.13370418	6.90932658
5.00383637	8.5217319	5.03642892	9.0442532	6.59697306	9.70880301
6.1745778	6.90580601	6.57716039	6.29980726	4.78838021	6.66879092
4.58083992	5.09691746	7.10931161	8.0190146	9.25740438	7.50690335
7.27229441	8.44111904	6.26282588	7.02272248	7.84524962	7.46282132
8.69272343	5.56934016	5.37021133	7.43854426	7.61959376	6.03009799
8.48231689	9.79489831	7.2043689	5.31076127	6.61601171	6.01062793
6.11002147	7.59132805	9.08657347	7.39157655	j	

In [49]:

```
%matplotlib inline
plt.scatter(x_train,y_train,color='yellow')
plt.xlabel('Ave Area Income in $')
plt.ylabel('Ave Area of Rooms in sqft')
plt.title('Training data')
plt.show()
```



In [50]:

```
lm = LinearRegression()
lm.fit(x_train,y_train)
y_predict = lm.predict(x_test)
print(f"Train accuracy {round(lm.score(x_train,y_train)*100,2)} %")
print(f"Test accuracy {round(lm.score(x_test,y_test)*100,2)} %")
```

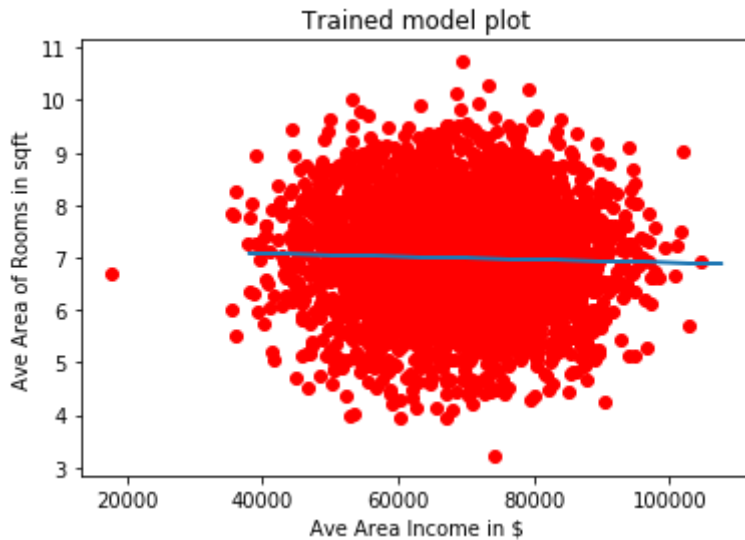
```
Train accuracy 0.1 %
Test accuracy -0.63 %
```

In [51]:

```
plt.scatter(x_train,y_train,color='red')
plt.plot(x_test,y_predict)
plt.xlabel('Ave Area Income in $')
plt.ylabel('Ave Area of Rooms in sqft')
plt.title("Trained model plot")
plt.plot
```

Out[51]:

```
<function matplotlib.pyplot.plot(*args, scalex=True, scaley=True, data=
None, **kwargs)>
```



## Comparisons Ave. Income & Ave. Area of Bedrooms

### Polynomial Regression Method

While Test Size is 20%

In [58]:

```
X = US_house['Avg. Area Income'].values
y = US_house['Avg. Area Number of Rooms'].values
```

In [59]:

```
X
```

Out[59]:

```
array([79545.45857432, 79248.64245483, 61287.06717866, ...,  
       63390.6868855 , 68001.33123509, 65510.58180367])
```

In [60]:

```
X=X.reshape(-1,1)  
X
```

Out[60]:

```
array([[79545.45857432],  
       [79248.64245483],  
       [61287.06717866],  
       ...,  
       [63390.6868855 ],  
       [68001.33123509],  
       [65510.58180367]])
```

In [61]:

```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.8,test_size=0.2,random_state=1000)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shap {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```

```
X_train shape (4000, 1)
y_train shape (4000,)
X_test shape (1000, 1)
y_test shape (1000,)
[7.67098283 6.02452319 7.65651167 8.76661274 6.46252334 6.52222048
 8.50065931 6.03645802 7.13125578 7.22911211 7.88847287 5.96140678
 5.16486548 6.65816348 8.56001888 8.54176845 7.3182145 7.316551
 8.46146827 6.9096769 7.74683647 6.15220352 8.8615771 7.8410663
 8.4513549 8.38320811 7.80189707 7.47067235 7.67923057 6.73400688
 8.00644913 8.59824688 5.74405984 5.58679444 5.0427468 6.12940457
 8.84066791 7.12398966 7.97987782 7.50381655 6.13419293 7.87459883
 7.23704281 6.31106491 6.66953171 7.47284985 6.81364961 7.16132931
 7.54589692 5.65425637 6.79627685 6.5413757 7.03385222 8.50933677
 7.24990524 8.08412347 7.73486203 7.8452404 6.42408542 6.50394588
 6.73255173 7.27878258 7.87896295 7.29698406 7.03666701 8.25904432
 7.64896367 4.98021947 6.18284258 4.9673884 5.56435288 7.40068467
 8.00541729 6.9682825 5.78918815 6.11019169 6.94214348 6.91430889
 7.42768912 6.87615033 8.28908486 5.07967061 7.51138304 6.02020121
 7.08507236 6.04232838 5.71544106 6.68668613 5.01279289 8.91606569
 7.41078451 7.51820433 6.95202667 8.1052283 6.47600734 8.69531316
 5.22798817 6.809013 6.25567488 6.13548984 7.6987646 6.36475668
 8.30330169 6.70105955 6.9225152 5.67113634 4.67913425 7.13315074
 7.51863612 6.33728559 7.60001598 6.67780569 9.26945262 8.36606762
 6.36827869 7.57964206 7.79586819 7.4804088 7.18252748 7.4843437
 4.99468979 6.22641197 7.51882743 6.49496655 6.70312502 7.19144435
 7.96045754 6.11432319 8.36090689 7.347288 6.72864277 8.56260323
 7.02125235 6.40013174 7.10384015 8.53996217 6.70264287 5.82540547
 7.42789161 7.40487042 7.8059037 7.15721495 7.88384041 6.97930778
 6.56120428 7.35139752 5.58803088 8.27338677 9.46876637 5.64536314
 6.123813 7.49269975 5.72300868 6.57165542 6.5900353 5.94839209
 7.39019749 6.7656924 7.13895771 6.65298344 8.76548129 5.91117641
 8.20329527 7.48132111 5.34530334 6.64519128 7.34936292 6.88139814
 6.92040927 7.27610229 5.82970368 6.69072862 7.08838588 7.03947718
 8.35354165 7.61038621 7.07359289 5.58104069 6.26153528 8.15737889
 8.31777468 8.39948615 7.31484847 6.97512803 7.2123397 7.55440649
 6.46959427 8.36254785 7.04682565 6.34607205 6.31145247 5.57104279
 6.50961621 7.7753357 4.41438087 7.24286605 7.11986208 6.31790688
 9.0381508 4.52542103 8.70042833 6.99074475 6.8242861 7.77004618
 6.57835207 7.00912859 5.17862848 9.71572267 7.795375 7.77504924
 8.1391394 6.47952731 8.08735663 7.37114309 6.08284131 7.38476365
 6.08432247 6.56122047 6.94676293 6.62377571 8.87724004 7.00371544
 8.44100893 7.00480304 5.83241183 5.40178683 6.88540385 8.32989418
 6.44643681 6.1435452 6.23116777 7.70961765 8.4173232 7.57062216
 7.92637224 6.67325638 6.22945854 8.00591367 7.70392045 6.64968387
 7.50827158 6.38247203 5.93561862 6.70551375 6.24885445 5.55074587
 8.9044547 6.50254083 6.98885176 8.33541863 6.49953359 5.79358981
 5.05585786 6.97315103 4.40683594 6.74487309 8.14775959 6.54983602
 6.00565649 7.17691728 4.99375978 5.29682702 7.54290995 7.59452876
 6.964616 6.84471227 7.45618504 5.4485801 5.37016214 6.10451244
 5.65297424 8.36657044 6.85184183 7.28033734 8.21060212 7.93153954
 5.81643633 8.40436325 7.18867902 5.58090594 8.20087222 6.66712651
 5.30079846 6.05324041 6.9590556 4.80508098 6.35421425 6.15177134
 7.56686536 8.67226397 7.14283552 7.43211427 6.41477457 9.50083282
 6.32261086 6.47694199 7.3491215 7.71668807 8.33620873 7.84432552
 5.60489361 8.10567172 5.73698335 7.11149677 8.11400139 5.82073184
 5.70811919 6.25044768 7.78776442 7.02340256 5.75716509 6.74859037
 6.90882967 5.89844743 8.21977112 7.33696283 7.56542067 4.63516111]
```



6.48468492	7.73709568	6.61443902	6.6758618	6.16541401	6.68683478
7.28833775	5.8415651	7.42409316	7.54040385	7.74179279	7.19223774
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7.42577752	8.45736227	7.55551872	6.21089502	7.84230638	8.47992133
7.21535475	8.06565992	6.78415351	5.3729923	6.15970522	7.20450454
8.34052442	7.55361501	8.60845681	6.04598402	6.38836263	5.91195308
7.21225427	5.81482265	7.16650165	8.05950533	4.70501539	6.96576449
8.07378651	5.89226871	7.85960445	6.00436502	8.19728942	5.42884667
7.19335183	8.51229511	8.32777647	7.68529206	6.63759752	7.59526393
5.68706258	5.71553335	7.52809801	5.38299815	6.8339559	7.50127508
6.76612964	6.29181815	7.93551076	5.82897549	5.63531521	6.59459228
7.21365908	7.21726847	7.31212466	7.29172807	5.33226457	5.69808252
6.41219784	7.05219498	5.72161852	6.26259715	8.61042055	6.9963361
7.48109117	6.77498642	6.97442448	7.1952462	6.82668629	6.40733952
7.68134202	9.14124198	6.32480683	7.79782478	6.965111	6.35862933
7.7344186	7.01525297	7.08983177	6.47103518	7.78813838	7.38712583
7.8395525	6.97168639	7.3729051	4.46053037	7.90367488	6.36072222
6.50983678	8.42582069	6.28721317	8.04075452	8.73721569	7.54543845
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7.35417683	7.05392715	6.37640125	5.45706093	7.11665503	6.91567232
7.0516388	8.07858448	7.73860842	7.25829112	7.08332635	5.66025312
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6.28035876	6.95671919	6.69528101	7.21196293	8.12688859	7.08015374
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6.564423	4.97716915	6.87237955	6.54460275	6.40694457	6.86779281
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4.7844429	7.91617166	6.77424335	8.02045005	6.4641121	6.15918296
8.02096551	7.46513643	6.71511073	4.83609078	4.04932055	7.05110773
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6.70105939	7.44083232	7.74984345	5.80750953	5.24030089	6.94059533
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6.16287142	6.40284391	8.28040387	8.76701435	5.78336239	7.70059735
6.46512886	6.5851681	7.40181464	5.90694008	6.65046209	8.44072615
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6.1780777	6.93275635	8.03562315	4.83942722	8.80204405	5.93628687
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5.34702706	5.56465345	6.28524674	8.00300675	8.95830017	6.91920414
6.49389932	6.46664525	6.28086382	8.51860818	7.98751262	7.66560184
8.31760044	7.06757467	4.81162263	5.62615918	6.45875588	7.02558811
5.55228986	6.47999368	7.60210723	6.36520575	6.9961141	8.54464981
7.07443434	7.31190652	7.87740359	6.86299437	6.2512391	6.99904497
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8.15953758	5.83173897	8.77666081	5.09352674	7.06620874	8.27198289
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7.38555504	8.0662199	7.98951567	6.02916087	7.14649989	6.80707758
6.16415145	6.99769954	7.99781793	7.62644406	7.36045528	6.37606022
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7.56019912	6.88291844	8.65734264	7.48193032	6.85153583	7.35028508
6.61798687	6.1005507	7.30991938	6.69199648	7.39790527	6.61175592
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6.66988841	6.3811622	7.21482412	6.53579637	5.93570931	6.08728421
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6.21178319	6.23031238	5.80740148	7.46488551	8.12645815	5.9086012
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7.42787307	5.92121599	7.12902984	6.63822317	6.13370418	6.90932658
5.00383637	8.5217319	5.03642892	9.0442532	6.59697306	9.70880301
6.1745778	6.90580601	6.57716039	6.29980726	4.78838021	6.66879092
4.58083992	5.09691746	7.10931161	8.0190146	9.25740438	7.50690335
7.27229441	8.44111904	6.26282588	7.02272248	7.84524962	7.46282132
8.69272343	5.56934016	5.37021133	7.43854426	7.61959376	6.03009799
8.48231689	9.79489831	7.2043689	5.31076127	6.61601171	6.01062793
6.11002147	7.59132805	9.08657347	7.39157655	j	

In [65]:

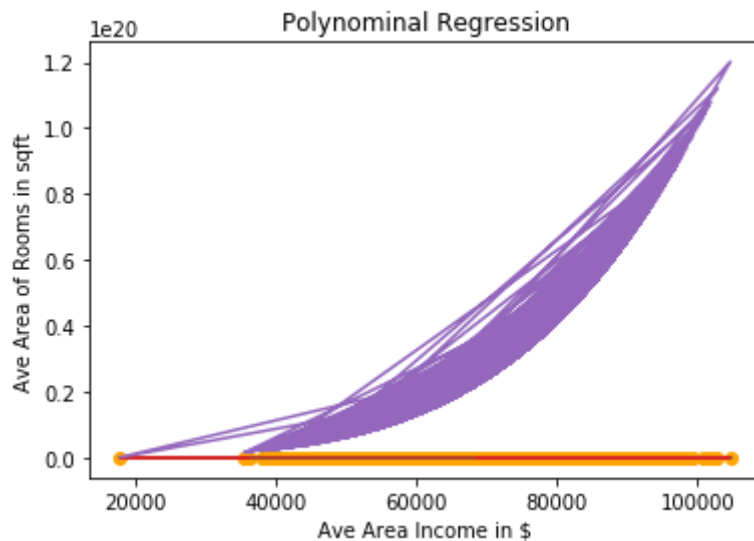
```
poly = PolynomialFeatures(degree = 4)
X_poly = poly.fit_transform(x_train)
poly.fit(X_poly,y_train)
```

Out[65]:

```
PolynomialFeatures(degree=4, include_bias=True, interaction_only=False,
                  order='C')
```

In [70]:

```
%matplotlib inline
plt.scatter(x_train,y_train,color='orange')
plt.plot(x_train, poly.fit_transform(x_train))
plt.xlabel('Ave Area Income in $')
plt.ylabel('Ave Area of Rooms in sqft')
plt.title('Polynominal Regression')
plt.show()
```



In [ ]:

In [ ]: