

# HousePricing

January 31, 2021

## 1 Housing Prices: Can you predict the home sales prices in Melbourne?

```
[1]: import warnings
warnings.filterwarnings('ignore')
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

### 1.1 Import the Data

```
[2]: # Remove redundant/unique information
dfHousePrices = pd.read_csv("house_prices.csv", usecols = ['Suburb', 'Rooms',
→ 'Type', 'Price', 'Method', 'Seller', 'Distance', 'Bedroom', 'Bathroom',
→ 'Car', 'Landsize', 'BuildingArea', 'YearBuilt', 'CouncilArea', 'RegionName',
→ 'PropertyCount', 'Postcode'])
dfHousePrices.head()
```

```
[2]:
```

	Suburb	Rooms	Type	Price	Method	Seller	Distance	Postcode	\
0	Abbotsford	2	h	1480000	S	Biggin	2.5	3067	
1	Abbotsford	2	h	1035000	S	Biggin	2.5	3067	
2	Abbotsford	3	h	1465000	SP	Biggin	2.5	3067	
3	Abbotsford	3	h	850000	PI	Biggin	2.5	3067	
4	Abbotsford	4	h	1600000	VB	Nelson	2.5	3067	

	Bedroom	Bathroom	Car	Landsize	BuildingArea	YearBuilt	CouncilArea	\
0	2	1	1.0	202	NaN	NaN	Yarra	
1	2	1	0.0	156	79.0	1900.0	Yarra	
2	3	2	0.0	134	150.0	1900.0	Yarra	
3	3	2	1.0	94	NaN	NaN	Yarra	
4	3	1	2.0	120	142.0	2014.0	Yarra	

	RegionName	PropertyCount
0	Northern Metropolitan	4019
1	Northern Metropolitan	4019
2	Northern Metropolitan	4019

```
3 Northern Metropolitan      4019
4 Northern Metropolitan      4019
```

```
[3]: dfHousePrices.shape
dfHousePrices.describe()
```

```
[3]:
```

	Rooms	Price	Distance	Postcode	Bedroom \
count	13580.000000	1.358000e+04	13580.000000	13580.000000	13580.000000
mean	2.937997	1.075684e+06	10.137776	3105.301915	2.914728
std	0.955748	6.393107e+05	5.868725	90.676964	0.965921
min	1.000000	8.500000e+04	0.000000	3000.000000	0.000000
25%	2.000000	6.500000e+05	6.100000	3044.000000	2.000000
50%	3.000000	9.030000e+05	9.200000	3084.000000	3.000000
75%	3.000000	1.330000e+06	13.000000	3148.000000	3.000000
max	10.000000	9.000000e+06	48.100000	3977.000000	20.000000

	Bathroom	Car	Landsize	BuildingArea	YearBuilt \
count	13580.000000	13518.000000	13580.000000	7130.000000	8205.000000
mean	1.534242	1.610075	558.416127	151.967650	1964.684217
std	0.691712	0.962634	3990.669241	541.014538	37.273762
min	0.000000	0.000000	0.000000	0.000000	1196.000000
25%	1.000000	1.000000	177.000000	93.000000	1940.000000
50%	1.000000	2.000000	440.000000	126.000000	1970.000000
75%	2.000000	2.000000	651.000000	174.000000	1999.000000
max	8.000000	10.000000	433014.000000	44515.000000	2018.000000

	PropertyCount
count	13580.000000
mean	7454.417378
std	4378.581772
min	249.000000
25%	4380.000000
50%	6555.000000
75%	10331.000000
max	21650.000000

## 1.2 Clean the Data

```
[4]: import math
# We need to fill empty values on year built, built area and car
median_year_built = math.floor(dfHousePrices.YearBuilt.median())
median_year_built
dfHousePrices.YearBuilt = dfHousePrices.YearBuilt.fillna(median_year_built)
```

```
[5]: median_car = math.floor(dfHousePrices.Car.median())
median_car
dfHousePrices.Car = dfHousePrices.Car.fillna(median_car)
```

```
[6]: median_building_area = math.floor(dfHousePrices.BuildingArea.median())
median_building_area
dfHousePrices.BuildingArea = dfHousePrices.BuildingArea.
↳fillna(median_building_area)
```

```
[7]: # For council area I don't want discrepancies with other location data, so rows
↳with empty cells will get dropped
dfHousePrices = dfHousePrices[dfHousePrices['CouncilArea'].notna()]
```

```
[8]: # Delete rows with 0 as landsize
dfHousePrices = dfHousePrices[dfHousePrices.Landsize != 0]
```

```
[9]: # Check data to see if we all empty cells are filled
dfHousePrices.describe(include='all')
```

```
[9]:
```

	Suburb	Rooms	Type	Price	Method	Seller	\
count	10272	10272.000000	10272	1.027200e+04	10272	10272	
unique	305	NaN	3	NaN	5	237	
top	Reservoir	NaN	h	NaN	S	Nelson	
freq	301	NaN	8033	NaN	7017	1239	
mean	NaN	3.068536	NaN	1.142861e+06	NaN	NaN	
std	NaN	0.899147	NaN	6.493972e+05	NaN	NaN	
min	NaN	1.000000	NaN	1.310000e+05	NaN	NaN	
25%	NaN	3.000000	NaN	7.100000e+05	NaN	NaN	
50%	NaN	3.000000	NaN	9.675000e+05	NaN	NaN	
75%	NaN	4.000000	NaN	1.400000e+06	NaN	NaN	
max	NaN	10.000000	NaN	9.000000e+06	NaN	NaN	

	Distance	Postcode	Bedroom	Bathroom	Car	\
count	10272.000000	10272.000000	10272.000000	10272.000000	10272.000000	
unique	NaN	NaN	NaN	NaN	NaN	
top	NaN	NaN	NaN	NaN	NaN	
freq	NaN	NaN	NaN	NaN	NaN	
mean	10.402434	3102.092971	3.038162	1.570678	1.659073	
std	5.578720	91.127321	0.916483	0.707189	0.985341	
min	0.000000	3000.000000	0.000000	0.000000	0.000000	
25%	6.700000	3044.000000	2.000000	1.000000	1.000000	
50%	9.850000	3081.000000	3.000000	1.000000	2.000000	
75%	13.000000	3146.000000	4.000000	2.000000	2.000000	
max	47.400000	3977.000000	20.000000	8.000000	10.000000	

	Landsize	BuildingArea	YearBuilt	CouncilArea	\
count	10272.000000	10272.000000	10272.000000	10272	
unique	NaN	NaN	NaN	33	
top	NaN	NaN	NaN	Moreland	
freq	NaN	NaN	NaN	1030	
mean	647.930199	141.715375	1965.026967	NaN	

std	4548.807775	101.695480	30.162665	NaN
min	1.000000	0.000000	1196.000000	NaN
25%	258.000000	126.000000	1960.000000	NaN
50%	516.000000	126.000000	1970.000000	NaN
75%	670.000000	137.000000	1970.000000	NaN
max	433014.000000	6791.000000	2018.000000	NaN

	RegionName	PropertyCount
count	10272	10272.000000
unique	8	NaN
top	Southern Metropolitan	NaN
freq	3303	NaN
mean	NaN	7304.301889
std	NaN	4392.989470
min	NaN	249.000000
25%	NaN	4019.000000
50%	NaN	6482.000000
75%	NaN	9704.000000
max	NaN	21650.000000

```
[10]: # Run this commands to check for typos on categorical data
unique_types_suburb = set(dfHousePrices['Suburb'].unique())
unique_types_suburb
unique_types_seller = set(dfHousePrices['Seller'].unique())
unique_types_seller
unique_types_council_area = set(dfHousePrices['CouncilArea'].unique())
unique_types_council_area
unique_types_region_name = set(dfHousePrices['RegionName'].unique())
unique_types_region_name
unique_types_type = set(dfHousePrices['Type'].unique())
unique_types_type
# Fortunately, there are no typos and unique values are correct
```

```
[10]: {'h', 't', 'u'}
```

```
[11]: dfHousePrices.dtypes
```

```
[11]: Suburb          object
Rooms             int64
Type              object
Price             int64
Method            object
Seller            object
Distance          float64
Postcode          int64
Bedroom           int64
Bathroom          int64
```

```

Car                float64
Landsize           int64
BuildingArea       float64
YearBuilt          float64
CouncilArea        object
RegionName         object
PropertyCount      int64
dtype: object

```

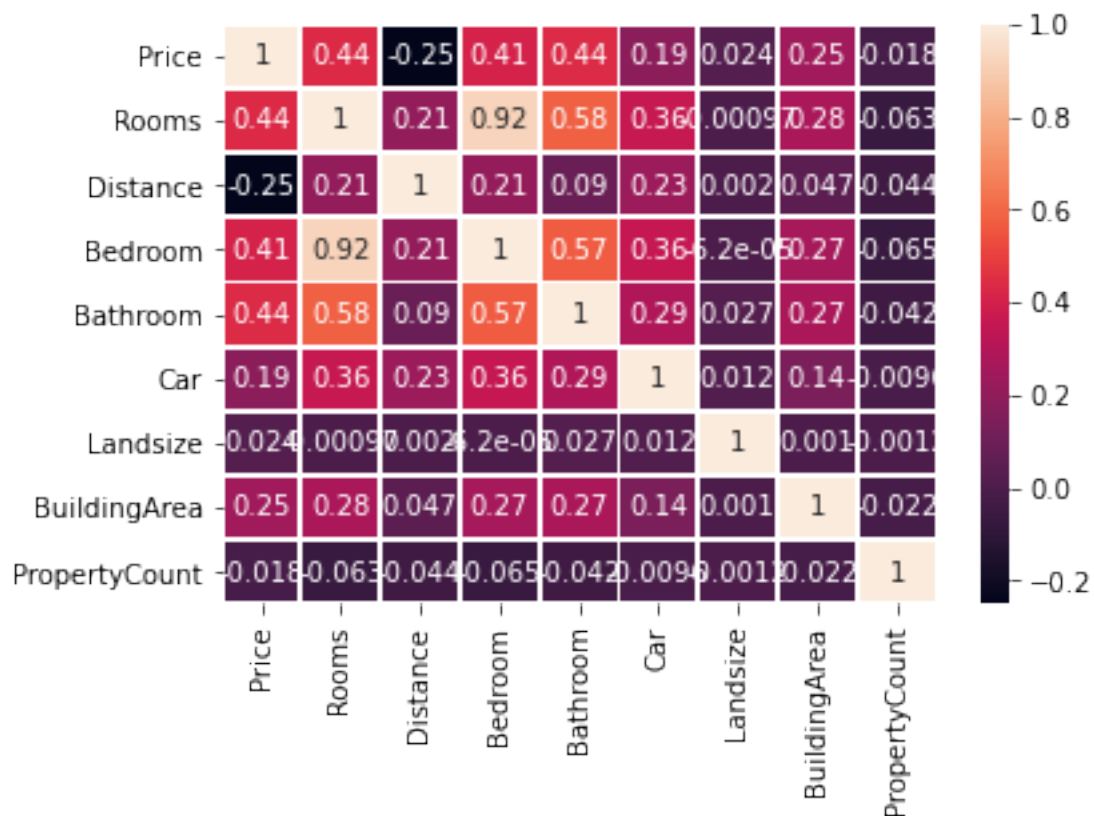
### 1.3 Exploratory Data Analysis - 1st Part

```

[12]: # Find correlations between continous variables (IV and DV)
X = dfHousePrices[['Price', 'Rooms', 'Distance', 'Bedroom', 'Bathroom', 'Car', 'Landsize', 'BuildingArea', 'PropertyCount']]
sns.heatmap(X.corr(),annot=True,lw=1)

```

[12]: <AxesSubplot:>



## 1.4 Build the model - 1st Part

```
[13]: X = dfHousePrices[['Rooms', 'Distance', 'Bathroom', 'Car', 'Landsize',  
↳ 'BuildingArea']]
```

```
[14]: Y = dfHousePrices['Price']
```

```
[15]: from sklearn.model_selection import train_test_split  
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.4,  
↳ random_state=101)  
print(X_train.shape)  
print(X_test.shape)  
print(y_train.shape)  
print(y_test.shape)
```

```
(6163, 6)
```

```
(4109, 6)
```

```
(6163,)
```

```
(4109,)
```

```
[16]: from sklearn.linear_model import LinearRegression  
model = LinearRegression()  
model.fit(X_train,y_train)
```

```
[16]: LinearRegression()
```

```
[17]: coeff_parameter = pd.DataFrame(model.coef_,X.columns,columns=['Coefficient'])  
coeff_parameter
```

```
[17]:
```

	Coefficient
Rooms	236505.665574
Distance	-43529.127991
Bathroom	225562.279396
Car	52646.923733
Landsize	2.376645
BuildingArea	531.643953

```
[18]: # Print the intercept  
print(model.intercept_)
```

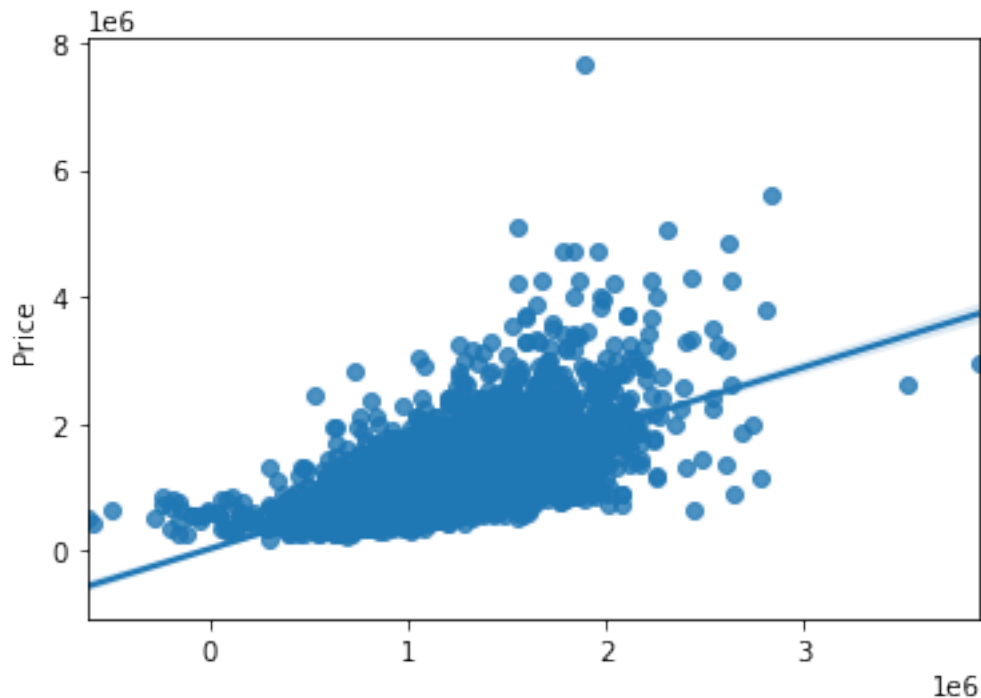
```
360564.808286401
```

```
[19]: predictions = model.predict(X_test)  
predictions
```

```
[19]: array([1591207.329957 , 1675408.08757592, 1520493.42450145, ...,  
1757506.19761488, 1099322.57289164, 1310125.88500423])
```

```
[20]: sns.regplot(x=predictions, y=y_test)
```

```
[20]: <AxesSubplot:ylabel='Price'>
```



```
[21]: import statsmodels.api as sm
X_train_Sm= sm.add_constant(X_train)
X_train_Sm= sm.add_constant(X_train)
ls=sm.OLS(y_train,X_train_Sm).fit()
print(ls.summary())
```

#### OLS Regression Results

```
=====
Dep. Variable:          Price    R-squared:                0.370
Model:                  OLS      Adj. R-squared:           0.369
Method:                 Least Squares    F-statistic:           601.6
Date:                   Sun, 31 Jan 2021    Prob (F-statistic):      0.00
Time:                   13:40:34    Log-Likelihood:         -90039.
No. Observations:       6163    AIC:                    1.801e+05
Df Residuals:           6156    BIC:                    1.801e+05
Df Model:                6
Covariance Type:        nonrobust
=====
```

```
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
```

const	3.606e+05	2.59e+04	13.910	0.000	3.1e+05	4.11e+05
Rooms	2.365e+05	9785.812	24.168	0.000	2.17e+05	2.56e+05
Distance	-4.353e+04	1280.785	-33.986	0.000	-4.6e+04	-4.1e+04
Bathroom	2.256e+05	1.2e+04	18.868	0.000	2.02e+05	2.49e+05
Car	5.265e+04	7333.730	7.179	0.000	3.83e+04	6.7e+04
Landsize	2.3766	1.175	2.023	0.043	0.074	4.679
BuildingArea	531.6440	61.678	8.620	0.000	410.733	652.555

```
=====
Omnibus:                 3458.236    Durbin-Watson:                 1.971
Prob(Omnibus):            0.000    Jarque-Bera (JB):            79899.249
Skew:                     2.217    Prob(JB):                     0.00
Kurtosis:                 20.073    Cond. No.                     2.28e+04
=====
```

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

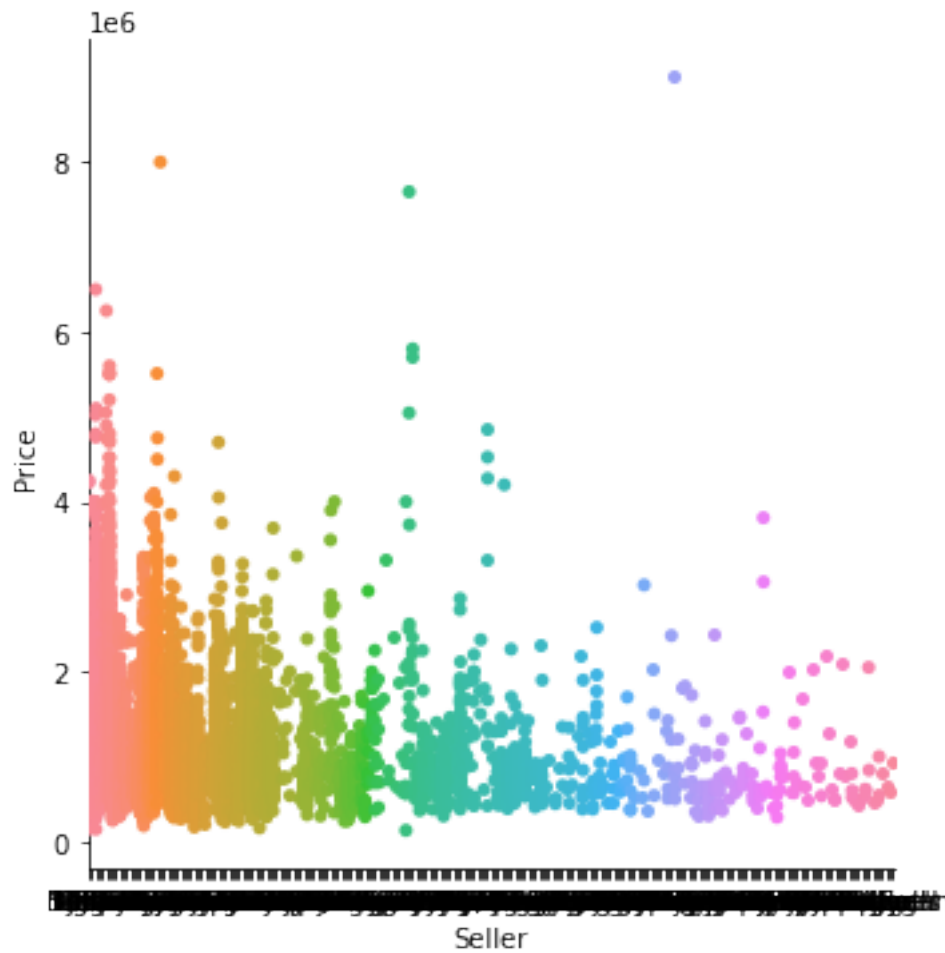
[2] The condition number is large, 2.28e+04. This might indicate that there are strong multicollinearity or other numerical problems.

## 1.5 Exploratory Data Analysis - 2nd Part

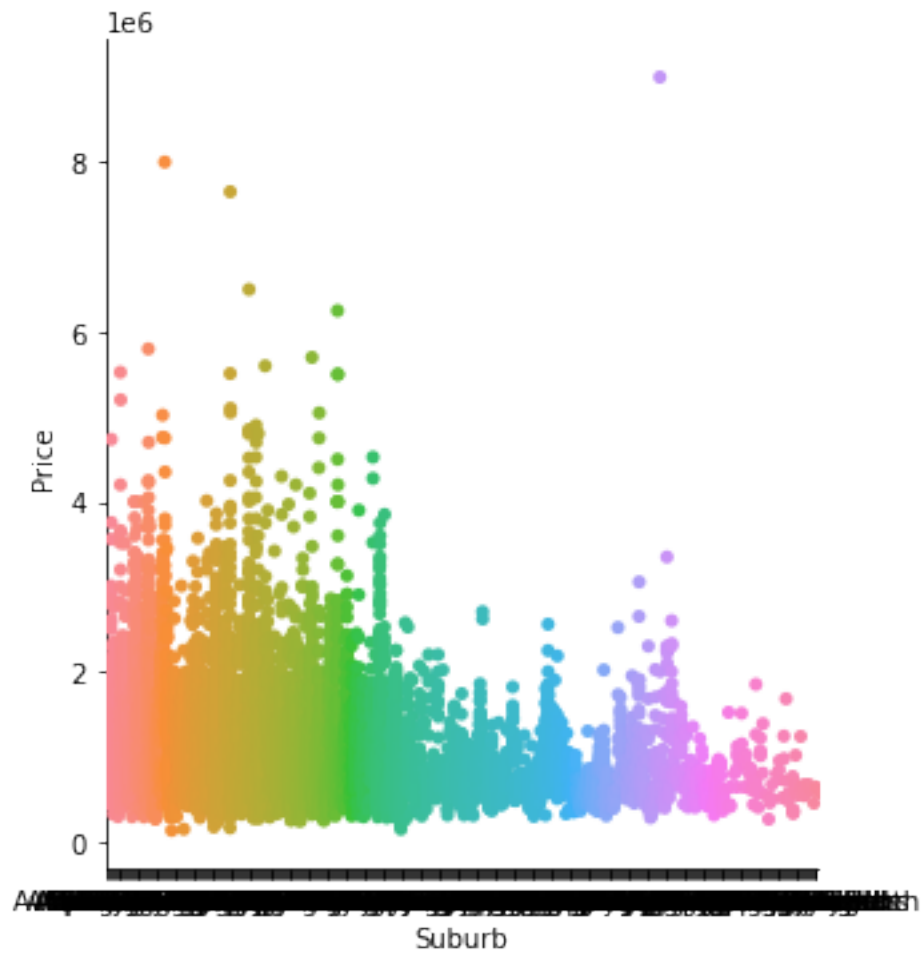
```
[22]: sns.catplot(x="Seller", y="Price", data=dfHousePrices, height=5)
```

```
[22]: <seaborn.axisgrid.FacetGrid at 0x7f829bdd4970>
```



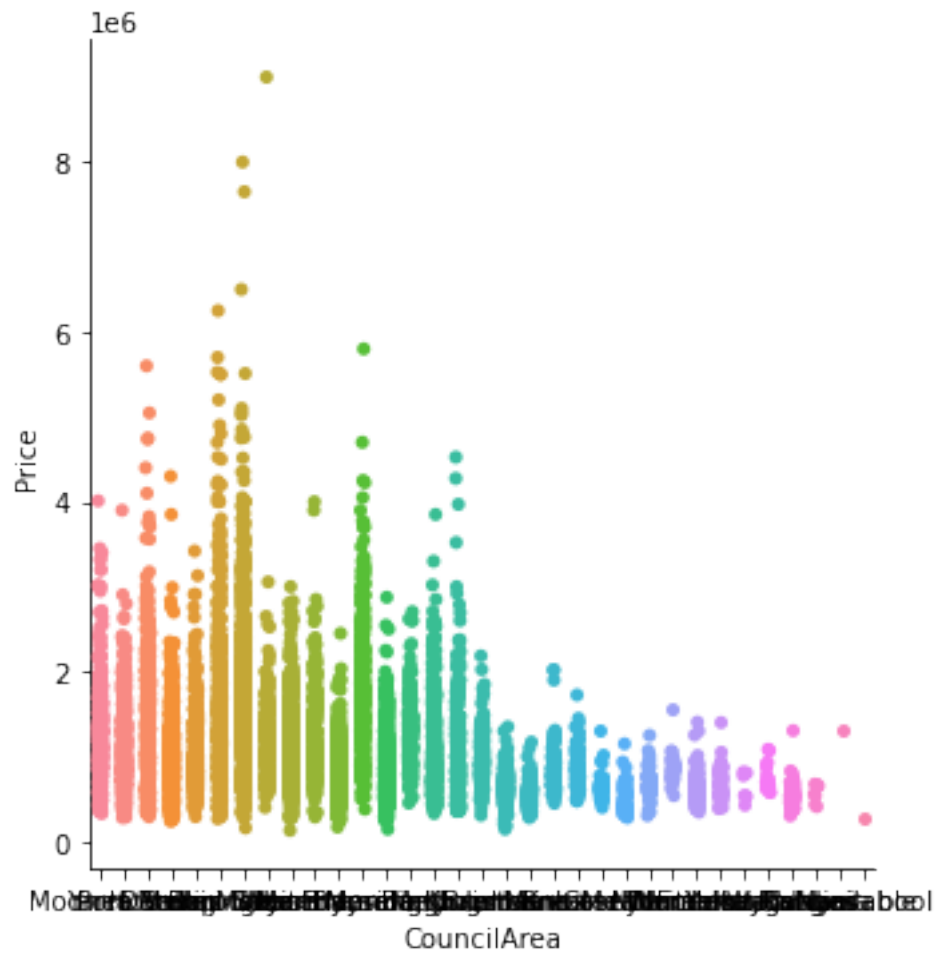


```
[23]: sns_plot = sns.catplot(x="Suburb", y="Price", data=dfHousePrices, height=5)  
      # sns_plot.savefig("output.png")
```



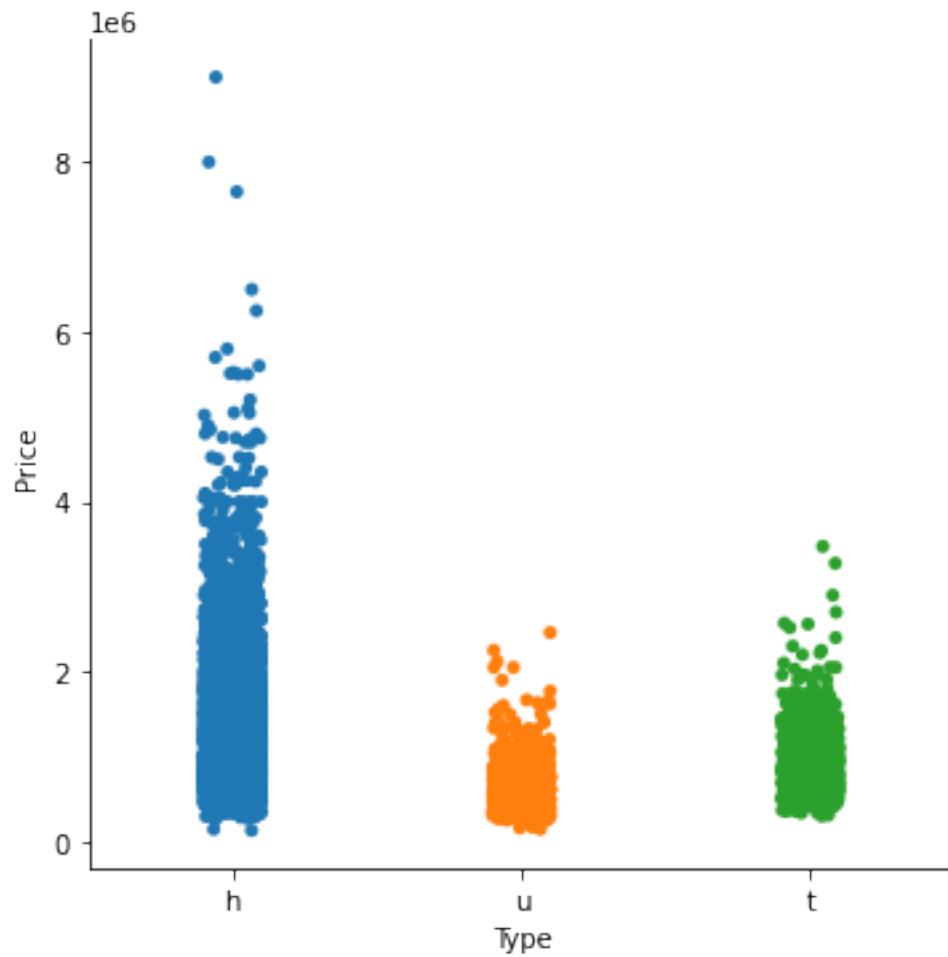
```
[24]: sns.catplot(x="CouncilArea", y="Price", data=dfHousePrices, height=5)
```

```
[24]: <seaborn.axisgrid.FacetGrid at 0x7f829be7a2b0>
```



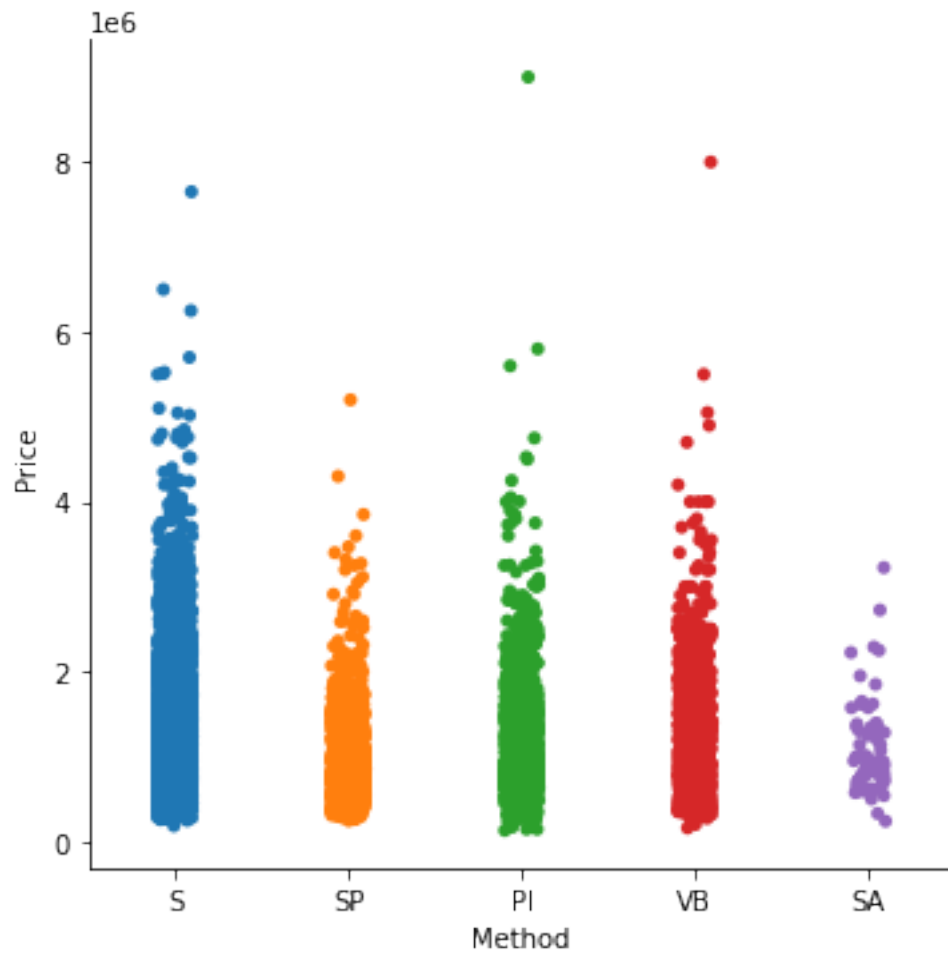
```
[25]: sns.catplot(x="Type", y="Price", data=dfHousePrices, height=5)
```

```
[25]: <seaborn.axisgrid.FacetGrid at 0x7f829fe0f130>
```



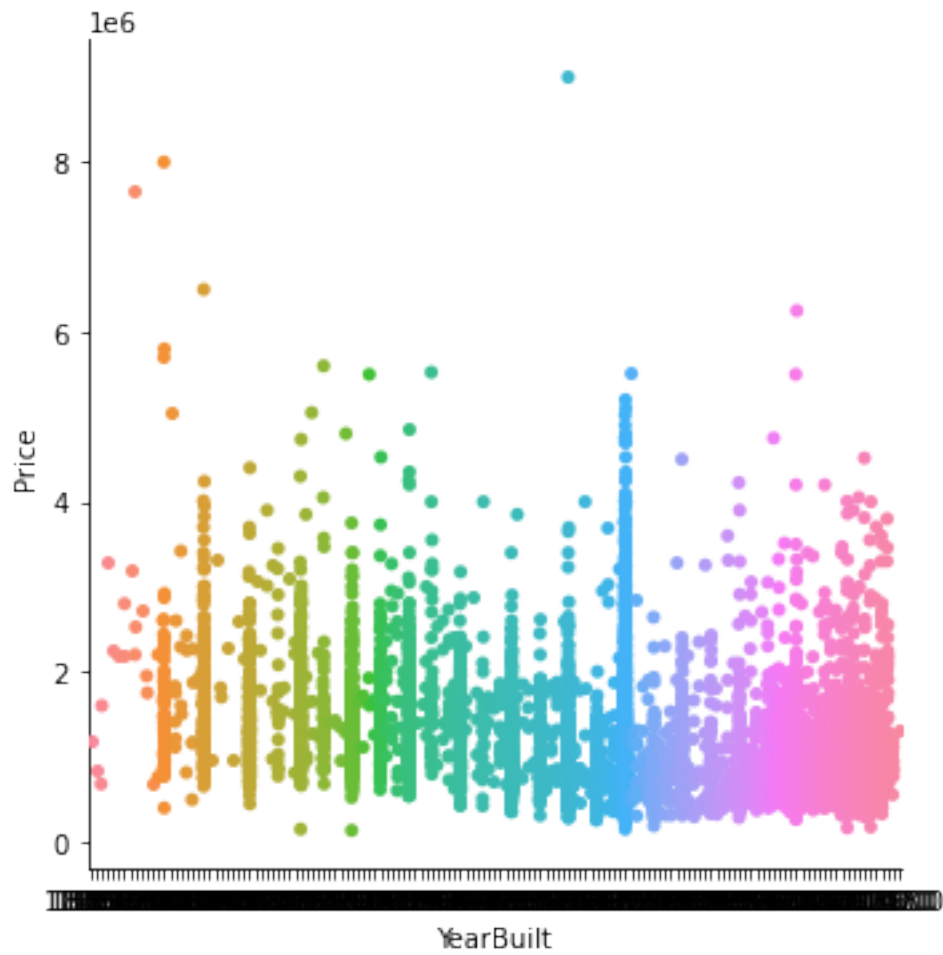
```
[26]: sns.catplot(x="Method", y="Price", data=dfHousePrices, height=5)
```

```
[26]: <seaborn.axisgrid.FacetGrid at 0x7f829cdbe790>
```



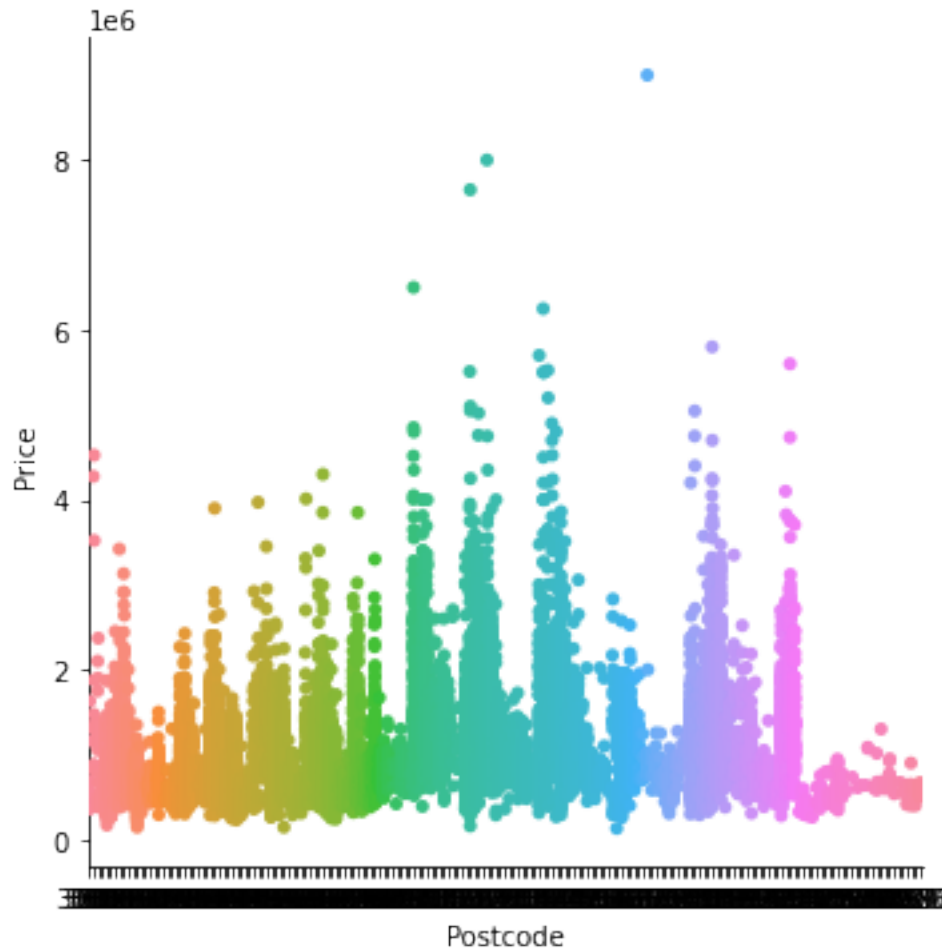
```
[27]: sns.catplot(x="YearBuilt", y="Price", data=dfHousePrices, height=5)
```

```
[27]: <seaborn.axisgrid.FacetGrid at 0x7f82a0a38310>
```



```
[28]: sns.catplot(x="Postcode", y="Price", data=dfHousePrices, height=5)
```

```
[28]: <seaborn.axisgrid.FacetGrid at 0x7f82a0a382b0>
```



```
[29]: #dfHousePrices['Suburb'] = dfHousePrices['Suburb'].astype('category')
#dfHousePrices['Type'] = dfHousePrices['Type'].astype('category')
#dfHousePrices['Method'] = dfHousePrices['Method'].astype('category')
#dfHousePrices['Seller'] = dfHousePrices['Seller'].astype('category')
#dfHousePrices['CouncilArea'] = dfHousePrices['CouncilArea'].astype('category')
#dfHousePrices['RegionName'] = dfHousePrices['RegionName'].astype('category')
```

## 1.6 Build the model - 2nd Part

```
[30]: # List independent variables
X = dfHousePrices[['Rooms', 'Distance', 'Bathroom', 'Car', 'Landsize',
↳ 'BuildingArea', 'PropertyCount', 'Type', 'Method', 'Seller', 'CouncilArea',
↳ 'RegionName']]
# Add dummy vars for cat vars
X = pd.get_dummies(data=X, drop_first=True)
# New dataframe
X.head()
```

```

[30]: Rooms Distance Bathroom Car Landsize BuildingArea PropertyCount \
0      2      2.5          1 1.0      202      126.0      4019
1      2      2.5          1 0.0      156      79.0      4019
2      3      2.5          2 0.0      134      150.0      4019
3      3      2.5          2 1.0       94      126.0      4019
4      4      2.5          1 2.0      120      142.0      4019

Type_t Type_u Method_S ... CouncilArea_Wyndham CouncilArea_Yarra \
0      0      0          1 ...              0              1
1      0      0          1 ...              0              1
2      0      0          0 ...              0              1
3      0      0          0 ...              0              1
4      0      0          0 ...              0              1

CouncilArea_Yarra Ranges RegionName_Eastern Victoria \
0                  0              0
1                  0              0
2                  0              0
3                  0              0
4                  0              0

RegionName_Northern Metropolitan RegionName_Northern Victoria \
0                  1              0
1                  1              0
2                  1              0
3                  1              0
4                  1              0

RegionName_South-Eastern Metropolitan RegionName_Southern Metropolitan \
0                  0              0
1                  0              0
2                  0              0
3                  0              0
4                  0              0

RegionName_Western Metropolitan RegionName_Western Victoria
0                  0              0
1                  0              0
2                  0              0
3                  0              0
4                  0              0

[5 rows x 288 columns]

```

```

[31]: # Dependent variable
Y = dfHousePrices['Price']
Y

```



```
[31]: 0      1480000
      1      1035000
      2      1465000
      3       850000
      4      1600000
      ...
     12206    1050000
     12208    1285000
     12210     525300
     12211     750000
     12212    2450000
      Name: Price, Length: 10272, dtype: int64
```

```
[32]: # Create the test set
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.4,
    ↪ random_state=101)
print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(6163, 288)
(4109, 288)
(6163,)
(4109,)
```

```
[33]: # Perform the regression
model = LinearRegression()
model.fit(X_train,y_train)
```

```
[33]: LinearRegression()
```

```
[34]: # Print the intercept
print(model.intercept_)
```

```
595512.4755647061
```

```
[35]: # Print all other parameters
coeff_parameter = pd.DataFrame(model.coef_,X.columns,columns=['Coefficient'])
coeff_parameter
coeff_parameter.to_csv (r'/Users/alejandrogleson/Downloads/Coefficients.csv ')
```

```
[36]: predictions = model.predict(X_test)
      predictions
```

```
[36]: array([1403836.26372661, 1325754.30293877, 1221592.32567192, ...,
        1564572.91262582,  940903.69314997, 1046661.74272207])
```

```
[37]: X_train_Sm= sm.add_constant(X_train)
X_train_Sm= sm.add_constant(X_train)
ls=sm.OLS(y_train,X_train_Sm).fit()
print(ls.summary())
```

#### OLS Regression Results

```
=====
Dep. Variable:          Price    R-squared:                0.647
Model:                  OLS      Adj. R-squared:            0.632
Method:                 Least Squares    F-statistic:          43.38
Date:                  Sun, 31 Jan 2021    Prob (F-statistic):      0.00
Time:                  13:41:18    Log-Likelihood:        -88251.
No. Observations:      6163    AIC:                  1.770e+05
Df Residuals:          5912    BIC:                  1.787e+05
Df Model:              250
Covariance Type:       nonrobust
=====
```

```
=====
                                coef    std err          t
P>|t|    [0.025    0.975]
-----
const                                5.955e+05    4.16e+05     1.432
0.152    -2.2e+05    1.41e+06
Rooms                                1.66e+05    8395.134    19.776
0.000     1.5e+05    1.82e+05
Distance                           -4.078e+04    2538.180   -16.069
0.000   -4.58e+04   -3.58e+04
Bathroom                           1.612e+05    9845.656    16.372
0.000     1.42e+05     1.8e+05
Car                                5.364e+04    5894.641     9.099
0.000     4.21e+04    6.52e+04
Landsize                             1.6659         0.914     1.822
0.068     -0.126         3.458
BuildingArea                         370.5476     47.550     7.793
0.000     277.331     463.764
PropertyCount                       -1.7820         1.607    -1.109
0.268     -4.933         1.369
Type_t                             -3.535e+05     2e+04   -17.699
0.000   -3.93e+05   -3.14e+05
Type_u                             -4.371e+05    1.81e+04   -24.153
0.000   -4.73e+05   -4.02e+05
Method_S                           1.016e+05    1.73e+04     5.875
0.000     6.77e+04    1.36e+05
Method_SA                           8.764e+04     6.9e+04     1.271
0.204   -4.75e+04    2.23e+05
Method_SP                           5.968e+04    2.28e+04     2.612
```

0.009	1.49e+04	1.04e+05			
Method_VB			2.413e+04	2.46e+04	0.979
0.328	-2.42e+04	7.24e+04			
Seller_AS			3.62e+04	5.04e+05	0.072
0.943	-9.52e+05	1.02e+06			
Seller_Abercromby's			9.141e+05	4.61e+05	1.983
0.047	1.06e+04	1.82e+06			
Seller_Ace			-9.826e+04	5.82e+05	-0.169
0.866	-1.24e+06	1.04e+06			
Seller_Alexkarbon			-1.767e+05	4.27e+05	-0.414
0.679	-1.01e+06	6.6e+05			
Seller_Allens			9.008e-08	6.02e-07	0.150
0.881	-1.09e-06	1.27e-06			
Seller_Anderson			3.008e+04	5.04e+05	0.060
0.952	-9.59e+05	1.02e+06			
Seller_Appleby			-4.748e+05	5.85e+05	-0.812
0.417	-1.62e+06	6.72e+05			
Seller_Aquire			-9.531e+04	5.88e+05	-0.162
0.871	-1.25e+06	1.06e+06			
Seller_Ascend			-7.818e+04	5.82e+05	-0.134
0.893	-1.22e+06	1.06e+06			
Seller_Ash			-1.759e-07	3.03e-06	-0.058
0.954	-6.12e-06	5.77e-06			
Seller_Asset			4080.7008	5.87e+05	0.007
0.994	-1.15e+06	1.16e+06			
Seller_Assisi			3.405e-07	1.53e-06	0.222
0.824	-2.66e-06	3.34e-06			
Seller_Australian			-2.313e+05	5.83e+05	-0.397
0.691	-1.37e+06	9.11e+05			
Seller_Barlow			6.609e+04	4.31e+05	0.154
0.878	-7.78e+05	9.1e+05			
Seller_Barry			-8.681e+04	4.13e+05	-0.210
0.833	-8.96e+05	7.22e+05			
Seller_Bayside			-2.988e+04	5.05e+05	-0.059
0.953	-1.02e+06	9.6e+05			
Seller_Bekdon			-1.788e+05	4.41e+05	-0.405
0.685	-1.04e+06	6.86e+05			
Seller_Beller			-4.624e+05	4.46e+05	-1.038
0.299	-1.34e+06	4.11e+05			
Seller_Bells			-8939.5364	4.21e+05	-0.021
0.983	-8.35e+05	8.17e+05			
Seller_Better			-1.987e+05	5.82e+05	-0.342
0.733	-1.34e+06	9.42e+05			
Seller_Biggin			-8.958e+04	4.14e+05	-0.216
0.829	-9.01e+05	7.22e+05			
Seller_Blue			6.719e+05	5.82e+05	1.155
0.248	-4.69e+05	1.81e+06			
Seller_Bowman			5.403e+04	5.87e+05	0.092

0.927	-1.1e+06	1.21e+06			
Seller_Brace			-1.575e+05	5.82e+05	-0.271
0.787	-1.3e+06	9.83e+05			
Seller_Brad			-5.527e+04	4.14e+05	-0.134
0.894	-8.66e+05	7.56e+05			
Seller_Buckingham			-4.739e+04	4.23e+05	-0.112
0.911	-8.77e+05	7.82e+05			
Seller_Bullen			1.858e+05	5.82e+05	0.319
0.749	-9.55e+05	1.33e+06			
Seller_Burnham			6.289e+04	4.23e+05	0.149
0.882	-7.67e+05	8.93e+05			
Seller_Buxton			-1.01e+05	4.13e+05	-0.244
0.807	-9.12e+05	7.1e+05			
Seller_Buxton/Find			1.396e+05	5.83e+05	0.239
0.811	-1e+06	1.28e+06			
Seller_C21			-4.596e+04	4.21e+05	-0.109
0.913	-8.72e+05	7.8e+05			
Seller_Caine			9.009e+05	4.52e+05	1.992
0.046	1.45e+04	1.79e+06			
Seller_Calder			4.404e+04	5.82e+05	0.076
0.940	-1.1e+06	1.18e+06			
Seller_Carter			-7.291e+04	4.8e+05	-0.152
0.879	-1.01e+06	8.68e+05			
Seller_Castran			7.693e+05	4.35e+05	1.769
0.077	-8.34e+04	1.62e+06			
Seller_Cayzer			2.051e+04	4.26e+05	0.048
0.962	-8.14e+05	8.55e+05			
Seller_Century			-9.898e-08	7.98e-07	-0.124
0.901	-1.66e-06	1.47e-06			
Seller_Chambers			-8.403e+04	4.45e+05	-0.189
0.850	-9.57e+05	7.89e+05			
Seller_Changing			-1.99e+05	5.83e+05	-0.342
0.733	-1.34e+06	9.43e+05			
Seller_Charlton			-9.49e+04	4.76e+05	-0.199
0.842	-1.03e+06	8.39e+05			
Seller_Chisholm			-8.553e+04	4.2e+05	-0.204
0.839	-9.08e+05	7.37e+05			
Seller_Christopher			-2.187e+05	4.61e+05	-0.474
0.635	-1.12e+06	6.85e+05			
Seller_Clairmont			1.745e-07	8.61e-07	0.203
0.840	-1.51e-06	1.86e-06			
Seller_Collins			1.403e+05	4.22e+05	0.333
0.739	-6.87e+05	9.67e+05			
Seller_Community			-4188.8962	5.87e+05	-0.007
0.994	-1.16e+06	1.15e+06			
Seller_Compton			-3.02e+05	5.82e+05	-0.519
0.604	-1.44e+06	8.39e+05			
Seller_Conquest			2084.0956	5.82e+05	0.004

0.997	-1.14e+06	1.14e+06			
Seller_Considine			-1.581e+05	4.33e+05	-0.366
0.715	-1.01e+06	6.9e+05			
Seller_Coventry			-5.226e+05	5.82e+05	-0.897
0.369	-1.66e+06	6.19e+05			
Seller_Craig			6.391e+04	5.82e+05	0.110
0.913	-1.08e+06	1.2e+06			
Seller_Crane			-2.36e+05	5.81e+05	-0.406
0.685	-1.37e+06	9.03e+05			
Seller_D'Aprano			-4.738e+04	4.61e+05	-0.103
0.918	-9.5e+05	8.56e+05			
Seller_Daniel			-6.407e-08	1.29e-07	-0.497
0.619	-3.17e-07	1.88e-07			
Seller_Darren			5.08e+04	4.25e+05	0.119
0.905	-7.83e+05	8.84e+05			
Seller_David			-1.798e+05	5.82e+05	-0.309
0.757	-1.32e+06	9.61e+05			
Seller_Del			-8.312e+04	5.07e+05	-0.164
0.870	-1.08e+06	9.12e+05			
Seller_Dingle			-2.287e+05	4.38e+05	-0.522
0.601	-1.09e+06	6.29e+05			
Seller_Dixon			-1.125e-07	3.82e-06	-0.029
0.977	-7.61e-06	7.38e-06			
Seller_Domain			-4.771e+05	5.04e+05	-0.946
0.344	-1.47e+06	5.12e+05			
Seller_Douglas			-5.119e+04	4.17e+05	-0.123
0.902	-8.69e+05	7.66e+05			
Seller_Edward			-1.643e+05	4.38e+05	-0.375
0.708	-1.02e+06	6.94e+05			
Seller_Elite			7.327e+05	5.82e+05	1.260
0.208	-4.07e+05	1.87e+06			
Seller_Eview			-5.592e+04	4.2e+05	-0.133
0.894	-8.79e+05	7.67e+05			
Seller_FN			-1.28e+05	4.37e+05	-0.293
0.770	-9.85e+05	7.29e+05			
Seller_First			-6.039e+04	4.76e+05	-0.127
0.899	-9.93e+05	8.72e+05			
Seller_Fletchers			-2.11e+04	4.14e+05	-0.051
0.959	-8.33e+05	7.91e+05			
Seller_Fletchers/One			-1.149e-07	1.06e-06	-0.109
0.913	-2.19e-06	1.96e-06			
Seller_Frank			8.815e+04	4.26e+05	0.207
0.836	-7.47e+05	9.24e+05			
Seller_Free			-1.025e+04	5.84e+05	-0.018
0.986	-1.16e+06	1.14e+06			
Seller_GL			-2.892e+04	4.33e+05	-0.067
0.947	-8.78e+05	8.2e+05			
Seller_Gardiner			-3.772e+04	5.82e+05	-0.065

0.948	-1.18e+06	1.1e+06			
Seller_Garvey			6.157e+04	5.81e+05	0.106
0.916	-1.08e+06	1.2e+06			
Seller_Gary			-1.118e+05	4.16e+05	-0.269
0.788	-9.27e+05	7.03e+05			
Seller_Grantham			-1.341e-08	1.16e-06	-0.012
0.991	-2.29e-06	2.26e-06			
Seller_Greg			1.476e+05	4.15e+05	0.355
0.722	-6.66e+05	9.62e+05			
Seller_Gunn&Co			-1.272e+04	4.46e+05	-0.028
0.977	-8.88e+05	8.62e+05			
Seller_HAR			-9.269e+04	4.25e+05	-0.218
0.827	-9.25e+05	7.4e+05			
Seller_Hall			1.403e+06	4.54e+05	3.093
0.002	5.14e+05	2.29e+06			
Seller_Ham			5.909e+04	5.83e+05	0.101
0.919	-1.08e+06	1.2e+06			
Seller_Harcourts			-1.145e+05	4.15e+05	-0.276
0.783	-9.28e+05	7e+05			
Seller_Harrington			-2.189e+05	4.31e+05	-0.508
0.611	-1.06e+06	6.26e+05			
Seller_Haughton			-2.551e+05	4.27e+05	-0.598
0.550	-1.09e+06	5.81e+05			
Seller_Hayeswinckle			-1.838e+05	5.83e+05	-0.315
0.753	-1.33e+06	9.6e+05			
Seller_Hodges			-9.753e+04	4.16e+05	-0.235
0.815	-9.13e+05	7.18e+05			
Seller_Holland			-2.18e+05	4.76e+05	-0.458
0.647	-1.15e+06	7.14e+05			
Seller_Homes			-1.567e+05	5.81e+05	-0.269
0.788	-1.3e+06	9.83e+05			
Seller_Hooper			3.628e+05	5.04e+05	0.720
0.472	-6.25e+05	1.35e+06			
Seller_Hoskins			1.119e+04	4.77e+05	0.023
0.981	-9.25e+05	9.47e+05			
Seller_Hunter			-1.252e+05	4.53e+05	-0.277
0.782	-1.01e+06	7.62e+05			
Seller_Iconek			-1.684e+04	4.78e+05	-0.035
0.972	-9.53e+05	9.19e+05			
Seller_J			-1.636e+05	4.41e+05	-0.371
0.711	-1.03e+06	7.01e+05			
Seller_JMRE			-2.455e+05	5.05e+05	-0.486
0.627	-1.24e+06	7.45e+05			
Seller_JRW			-2.774e-08	7.58e-07	-0.037
0.971	-1.51e-06	1.46e-06			
Seller_Jas			2.656e+04	4.15e+05	0.064
0.949	-7.87e+05	8.41e+05			
Seller_Jason			5.584e+04	4.43e+05	0.126

0.900	-8.12e+05	9.24e+05			
Seller_Jellis			8.767e+04	4.13e+05	0.212
0.832	-7.22e+05	8.97e+05			
Seller_Jim			3.244e+05	5.93e+05	0.547
0.584	-8.38e+05	1.49e+06			
Seller_Joe			4.831e-07	8.95e-07	0.540
0.589	-1.27e-06	2.24e-06			
Seller_Johnston			-7.017e+05	5.81e+05	-1.207
0.227	-1.84e+06	4.38e+05			
Seller_Kay			4.201e+05	4.17e+05	1.008
0.313	-3.97e+05	1.24e+06			
Seller_Kaye			-2.366e-07	3.01e-06	-0.079
0.937	-6.14e-06	5.66e-06			
Seller_Keatings			2.148e-07	1.19e-06	0.181
0.857	-2.11e-06	2.54e-06			
Seller_Kelly			1.05e+06	5.82e+05	1.805
0.071	-9.04e+04	2.19e+06			
Seller_Ken			-1.542e+04	5.06e+05	-0.030
0.976	-1.01e+06	9.77e+05			
Seller_LITTLE			-3.071e+05	4.25e+05	-0.723
0.470	-1.14e+06	5.25e+05			
Seller_LJ			-3.327e+04	4.26e+05	-0.078
0.938	-8.69e+05	8.02e+05			
Seller_LLC			5.528e+05	5.85e+05	0.945
0.345	-5.94e+05	1.7e+06			
Seller_Langwell			-5.586e+05	5.82e+05	-0.960
0.337	-1.7e+06	5.82e+05			
Seller_Leading			-1.288e+04	5.05e+05	-0.026
0.980	-1e+06	9.77e+05			
Seller_Leased			5.051e+04	5.81e+05	0.087
0.931	-1.09e+06	1.19e+06			
Seller_Leeburn			4.909e+05	5.05e+05	0.973
0.331	-4.98e+05	1.48e+06			
Seller_Leyton			-6.374e-08	3.85e-06	-0.017
0.987	-7.61e-06	7.48e-06			
Seller_Lindellas			6.603e+05	4.38e+05	1.508
0.132	-1.98e+05	1.52e+06			
Seller_Love			-1.052e+05	4.17e+05	-0.253
0.801	-9.22e+05	7.11e+05			
Seller_Lucas			3.264e+05	5.82e+05	0.560
0.575	-8.15e+05	1.47e+06			
Seller_Luxton			-1.105e+05	5.81e+05	-0.190
0.849	-1.25e+06	1.03e+06			
Seller_M.J			6.247e-09	5.05e-07	0.012
0.990	-9.83e-07	9.95e-07			
Seller_MICM			-2.098e+05	4.53e+05	-0.463
0.643	-1.1e+06	6.79e+05			
Seller_Maddison			-2.244e+05	5.04e+05	-0.445

0.656	-1.21e+06	7.64e+05			
Seller_Mandy			-1.894e+05	5.05e+05	-0.375
0.708	-1.18e+06	8.01e+05			
Seller_Marshall			2.492e+05	4.13e+05	0.603
0.547	-5.61e+05	1.06e+06			
Seller_Mason			-5.829e+05	5.13e+05	-1.137
0.255	-1.59e+06	4.22e+05			
Seller_Matthew			-2.842e+04	5.05e+05	-0.056
0.955	-1.02e+06	9.61e+05			
Seller_Max			1.499e+05	5.08e+05	0.295
0.768	-8.45e+05	1.15e+06			
Seller_McDonald			3.75e+04	4.24e+05	0.089
0.929	-7.93e+05	8.68e+05			
Seller_McGrath			9396.4741	4.15e+05	0.023
0.982	-8.04e+05	8.22e+05			
Seller_McLennan			4.241e+05	5.88e+05	0.721
0.471	-7.28e+05	1.58e+06			
Seller_Melbourne			2.281e+05	4.51e+05	0.505
0.613	-6.57e+05	1.11e+06			
Seller_Metro			-1.141e+05	5.87e+05	-0.194
0.846	-1.26e+06	1.04e+06			
Seller_Miles			7.806e+04	4.16e+05	0.188
0.851	-7.37e+05	8.93e+05			
Seller_Millership			-3.988e+04	4.63e+05	-0.086
0.931	-9.47e+05	8.67e+05			
Seller_Mindacom			4.292e+05	6.08e+05	0.706
0.480	-7.63e+05	1.62e+06			
Seller_Mitchell			4.446e+05	5.83e+05	0.762
0.446	-6.99e+05	1.59e+06			
Seller_Moonee			-2.482e+05	4.22e+05	-0.588
0.556	-1.08e+06	5.79e+05			
Seller_Morleys			-4.431e+05	5.82e+05	-0.761
0.447	-1.58e+06	6.98e+05			
Seller_Morrison			-8.076e+04	4.3e+05	-0.188
0.851	-9.24e+05	7.63e+05			
Seller_Naison			1.314e-07	1.05e-06	0.125
0.901	-1.93e-06	2.19e-06			
Seller_Nardella			-4.181e+05	5.82e+05	-0.718
0.473	-1.56e+06	7.24e+05			
Seller_Nelson			-3.811e+04	4.13e+05	-0.092
0.926	-8.47e+05	7.71e+05			
Seller_New			-2.107e+05	5.82e+05	-0.362
0.717	-1.35e+06	9.3e+05			
Seller_Nguyen			1.92e-07	9.46e-07	0.203
0.839	-1.66e-06	2.05e-06			
Seller_Nicholls			-2.438e+05	5.95e+05	-0.410
0.682	-1.41e+06	9.23e+05			
Seller_Nicholson			-1.449e+05	5.04e+05	-0.288



0.774	-1.13e+06	8.43e+05			
Seller_Nick			1.346e+05	4.25e+05	0.317
0.751	-6.98e+05	9.67e+05			
Seller_Noel			-9.946e+04	4.15e+05	-0.240
0.811	-9.13e+05	7.14e+05			
Seller_North			2.059e+05	5.81e+05	0.354
0.723	-9.34e+05	1.35e+06			
Seller_O'Brien			1.854e+04	4.23e+05	0.044
0.965	-8.1e+05	8.47e+05			
Seller_O'Donoghues			-9.175e+04	4.41e+05	-0.208
0.835	-9.56e+05	7.73e+05			
Seller_Oak			8.476e+04	5.83e+05	0.145
0.884	-1.06e+06	1.23e+06			
Seller_Obrien			2.472e+05	4.78e+05	0.517
0.605	-6.89e+05	1.18e+06			
Seller_One			-5.586e+04	4.75e+05	-0.118
0.906	-9.86e+05	8.75e+05			
Seller_Only			2.397e+05	5.18e+05	0.463
0.643	-7.75e+05	1.25e+06			
Seller_Oriental			-1.216e+05	5.83e+05	-0.209
0.835	-1.26e+06	1.02e+06			
Seller_Owen			3.396e-07	1.99e-06	0.171
0.865	-3.56e-06	4.24e-06			
Seller_PRDNationwide			1.506e+05	4.86e+05	0.310
0.756	-8.01e+05	1.1e+06			
Seller_Pagan			-1.564e+05	4.52e+05	-0.346
0.729	-1.04e+06	7.3e+05			
Seller_Parkes			-3.254e+04	4.76e+05	-0.068
0.946	-9.67e+05	9.02e+05			
Seller_Parkinson			-2.669e+05	5.82e+05	-0.458
0.647	-1.41e+06	8.75e+05			
Seller_Paul			9.33e+04	4.52e+05	0.207
0.836	-7.92e+05	9.78e+05			
Seller_Peake			-3.839e-07	1.66e-06	-0.232
0.817	-3.63e-06	2.87e-06			
Seller_Peter			-9.821e+04	4.29e+05	-0.229
0.819	-9.4e+05	7.44e+05			
Seller_Philip			-4.709e+04	4.21e+05	-0.112
0.911	-8.72e+05	7.78e+05			
Seller_Pride			-1.35e+05	4.61e+05	-0.293
0.770	-1.04e+06	7.69e+05			
Seller_Prime			2.769e-07	8.57e-07	0.323
0.747	-1.4e-06	1.96e-06			
Seller_Private/Tiernan's			8.567e+04	5.82e+05	0.147
0.883	-1.05e+06	1.23e+06			
Seller_Prof.			-1.536e+05	4.41e+05	-0.348
0.728	-1.02e+06	7.11e+05			
Seller_Property			-2.972e+04	5.81e+05	-0.051

0.959	-1.17e+06	1.11e+06			
Seller_Propertyau			1.969e-07	2.99e-07	0.658
0.510	-3.9e-07	7.83e-07			
Seller_Purplebricks			-1.286e+05	4.22e+05	-0.305
0.760	-9.55e+05	6.98e+05			
Seller_R&H			-2.146e-08	3.82e-07	-0.056
0.955	-7.71e-07	7.28e-07			
Seller_RE			1.033e+05	5.04e+05	0.205
0.838	-8.85e+05	1.09e+06			
Seller_REMAX			4.648e+04	5.1e+05	0.091
0.927	-9.53e+05	1.05e+06			
Seller_RT			1.606e+05	4.15e+05	0.387
0.699	-6.54e+05	9.75e+05			
Seller_RW			-5.979e+04	4.18e+05	-0.143
0.886	-8.79e+05	7.59e+05			
Seller_Raine			-1.144e+05	4.16e+05	-0.275
0.783	-9.29e+05	7e+05			
Seller_Raine&Horne			-6.804e+05	5.83e+05	-1.168
0.243	-1.82e+06	4.62e+05			
Seller_Ray			-1.338e+05	4.13e+05	-0.324
0.746	-9.43e+05	6.75e+05			
Seller_Re			-2.796e+05	5.04e+05	-0.554
0.579	-1.27e+06	7.09e+05			
Seller_Real			4.349e+05	5.82e+05	0.748
0.455	-7.05e+05	1.58e+06			
Seller_Red			4.221e+05	5.05e+05	0.837
0.403	-5.67e+05	1.41e+06			
Seller_Redina			1.078e+05	5.82e+05	0.185
0.853	-1.03e+06	1.25e+06			
Seller_Reed			6.795e-08	5.99e-07	0.113
0.910	-1.11e-06	1.24e-06			
Seller_Reliance			4.557e+04	5.09e+05	0.089
0.929	-9.53e+05	1.04e+06			
Seller_Rendina			-3.359e+04	4.2e+05	-0.080
0.936	-8.58e+05	7.9e+05			
Seller_Ristic			-9.754e+04	5.06e+05	-0.193
0.847	-1.09e+06	8.94e+05			
Seller_Rodney			-5.057e+05	4.76e+05	-1.062
0.288	-1.44e+06	4.28e+05			
Seller_Ross			-1.46e+06	5.83e+05	-2.502
0.012	-2.6e+06	-3.16e+05			
Seller_Rounds			-4.439e+05	5.83e+05	-0.761
0.447	-1.59e+06	7e+05			
Seller_Ryder			-1.765e+05	5.94e+05	-0.297
0.766	-1.34e+06	9.88e+05			
Seller_S&L			-1.828e+04	4.52e+05	-0.040
0.968	-9.04e+05	8.67e+05			
Seller_Schroeder			3.995e+05	5.85e+05	0.683

0.494	-7.46e+05	1.55e+06			
Seller_Scott			-7.011e-09	1.67e-08	-0.420
0.674	-3.97e-08	2.57e-08			
Seller_Sell			3.1e+05	5.82e+05	0.532
0.594	-8.32e+05	1.45e+06			
Seller_Sotheby's			1.293e+06	4.61e+05	2.804
0.005	3.89e+05	2.2e+06			
Seller_Steveway			2.16e-07	1.24e-06	0.174
0.862	-2.22e-06	2.65e-06			
Seller_Stockdale			-1.448e+05	4.15e+05	-0.349
0.727	-9.59e+05	6.69e+05			
Seller_Sweeney			-5.291e+04	4.15e+05	-0.128
0.899	-8.66e+05	7.6e+05			
Seller_Sweeney/Advantage			-1.693e-07	7.28e-07	-0.233
0.816	-1.6e-06	1.26e-06			
Seller_TRUE			-6.37e-08	8.17e-07	-0.078
0.938	-1.67e-06	1.54e-06			
Seller_Thomas			-4.147e+05	5.82e+05	-0.713
0.476	-1.56e+06	7.26e+05			
Seller_Thomson			-2.445e+05	4.27e+05	-0.572
0.567	-1.08e+06	5.93e+05			
Seller_Tiernan's			-1.005e-07	4.6e-07	-0.218
0.827	-1e-06	8.02e-07			
Seller_Tim			-2.873e+05	4.45e+05	-0.645
0.519	-1.16e+06	5.86e+05			
Seller_Trimson			1.015e+04	4.52e+05	0.022
0.982	-8.76e+05	8.97e+05			
Seller_Triwest			-7.826e+04	5.06e+05	-0.155
0.877	-1.07e+06	9.15e+05			
Seller_U			-1.084e+05	4.69e+05	-0.231
0.817	-1.03e+06	8.1e+05			
Seller_Upper			8.438e-08	1.54e-07	0.546
0.585	-2.18e-07	3.87e-07			
Seller_VICProp			8.118e+05	5.04e+05	1.609
0.108	-1.77e+05	1.8e+06			
Seller_Veitch			-1.387e-07	1.4e-07	-0.991
0.322	-4.13e-07	1.36e-07			
Seller_Vic			-8.712e+04	5.82e+05	-0.150
0.881	-1.23e+06	1.05e+06			
Seller_Victory			8.304e-08	3.76e-07	0.221
0.825	-6.54e-07	8.2e-07			
Seller_Village			6.4e+04	4.16e+05	0.154
0.878	-7.52e+05	8.8e+05			
Seller_W.B.			-9739.7169	5.05e+05	-0.019
0.985	-1e+06	9.81e+05			
Seller_WHITEFOX			1.379e+05	5.82e+05	0.237
0.813	-1e+06	1.28e+06			
Seller_Walsh			1.086e-07	2.72e-07	0.399

0.690	-4.25e-07	6.42e-07			
Seller_Walshe			-2.609e+04	4.41e+05	-0.059
0.953	-8.9e+05	8.38e+05			
Seller_Weast			6.674e-08	4.56e-07	0.146
0.884	-8.27e-07	9.6e-07			
Seller_Weda			1.927e+05	4.76e+05	0.405
0.686	-7.4e+05	1.13e+06			
Seller_Weston			-1.876e+05	5.82e+05	-0.322
0.747	-1.33e+06	9.53e+05			
Seller_Westside			-3.947e+05	5.06e+05	-0.780
0.435	-1.39e+06	5.97e+05			
Seller_White			-2.922e-08	3.09e-08	-0.945
0.344	-8.98e-08	3.14e-08			
Seller_Whiting			-3.294e+05	4.61e+05	-0.715
0.475	-1.23e+06	5.74e+05			
Seller_William			-2.518e+05	4.62e+05	-0.546
0.585	-1.16e+06	6.53e+05			
Seller_Williams			-1.906e+04	4.18e+05	-0.046
0.964	-8.38e+05	8e+05			
Seller_Wilson			-5.953e+04	4.42e+05	-0.135
0.893	-9.25e+05	8.06e+05			
Seller_Win			-2.194e+05	4.62e+05	-0.475
0.635	-1.13e+06	6.87e+05			
Seller_Woodards			-7.268e+04	4.14e+05	-0.175
0.861	-8.85e+05	7.4e+05			
Seller_Xynergy			3.683e-09	4.96e-09	0.742
0.458	-6.05e-09	1.34e-08			
Seller_YPA			-1.094e+05	4.14e+05	-0.264
0.791	-9.2e+05	7.01e+05			
Seller_Zahn			-3.961e-09	8.56e-09	-0.463
0.643	-2.07e-08	1.28e-08			
Seller_buyMyplace			-1.164e+05	5.04e+05	-0.231
0.817	-1.11e+06	8.72e+05			
Seller_hockingstuart			-1.138e+05	4.13e+05	-0.276
0.783	-9.23e+05	6.95e+05			
Seller_hockingstuart/Barry			5.551e+04	5.81e+05	0.096
0.924	-1.08e+06	1.19e+06			
Seller_iProperty			-9.421e+04	5.83e+05	-0.162
0.872	-1.24e+06	1.05e+06			
Seller_iSell			1.352e+05	4.6e+05	0.294
0.769	-7.68e+05	1.04e+06			
Seller_iTRAK			-8.852e+04	4.56e+05	-0.194
0.846	-9.81e+05	8.04e+05			
CouncilArea_Bayside			8.728e+05	6.17e+04	14.157
0.000	7.52e+05	9.94e+05			
CouncilArea_Boroondara			6.791e+05	5.64e+04	12.043
0.000	5.69e+05	7.9e+05			
CouncilArea_Brimbank			-9.05e+04	8.47e+04	-1.068

0.285	-2.57e+05	7.56e+04			
CouncilArea_Cardinia			3.406e+05	2.81e+05	1.213
0.225	-2.1e+05	8.91e+05			
CouncilArea_Casey			1.944e+05	1.43e+05	1.359
0.174	-8.61e+04	4.75e+05			
CouncilArea_Darebin			1.242e+05	6.03e+04	2.061
0.039	6052.889	2.42e+05			
CouncilArea_Frankston			5.906e+05	1.24e+05	4.778
0.000	3.48e+05	8.33e+05			
CouncilArea_Glen Eira			4.921e+05	5.81e+04	8.471
0.000	3.78e+05	6.06e+05			
CouncilArea_Greater Dandenong			7.637e+04	1.17e+05	0.654
0.513	-1.53e+05	3.05e+05			
CouncilArea_Hobsons Bay			8.495e+04	8.93e+04	0.952
0.341	-9e+04	2.6e+05			
CouncilArea_Hume			-6.181e+04	7.28e+04	-0.849
0.396	-2.05e+05	8.09e+04			
CouncilArea_Kingston			4.28e+05	7.12e+04	6.011
0.000	2.88e+05	5.68e+05			
CouncilArea_Knox			2.398e+05	7.74e+04	3.100
0.002	8.82e+04	3.92e+05			
CouncilArea_Macedon Ranges			6.248e+05	2.9e+05	2.155
0.031	5.64e+04	1.19e+06			
CouncilArea_Manningham			1.588e+05	4.66e+04	3.409
0.001	6.75e+04	2.5e+05			
CouncilArea_Maribyrnong			-6.788e+04	8.74e+04	-0.776
0.438	-2.39e+05	1.04e+05			
CouncilArea_Maroondah			3.076e+05	7.72e+04	3.985
0.000	1.56e+05	4.59e+05			
CouncilArea_Melbourne			1.862e+05	6.46e+04	2.882
0.004	5.95e+04	3.13e+05			
CouncilArea_Melton			-1.843e+05	1.12e+05	-1.644
0.100	-4.04e+05	3.55e+04			
CouncilArea_Monash			4.125e+05	5.58e+04	7.387
0.000	3.03e+05	5.22e+05			
CouncilArea_Moonsee Valley			1.295e+05	8.04e+04	1.612
0.107	-2.8e+04	2.87e+05			
CouncilArea_Moorabool			0	0	nan
nan	0	0			
CouncilArea_Moreland			-6638.8805	5.75e+04	-0.116
0.908	-1.19e+05	1.06e+05			
CouncilArea_Nillumbik			4.254e+04	1.05e+05	0.406
0.685	-1.63e+05	2.48e+05			
CouncilArea_Port Phillip			4.941e+05	6.3e+04	7.843
0.000	3.71e+05	6.18e+05			
CouncilArea_Stonnington			6.852e+05	5.96e+04	11.491
0.000	5.68e+05	8.02e+05			
CouncilArea_Unavailable			0	0	nan

nan	0	0			
CouncilArea_Whitehorse			3.469e+05	4.88e+04	7.112
0.000	2.51e+05	4.43e+05			
CouncilArea_Whittlesea			2.655e+04	7.42e+04	0.358
0.721	-1.19e+05	1.72e+05			
CouncilArea_Wyndham			-1.335e+05	9.97e+04	-1.338
0.181	-3.29e+05	6.21e+04			
CouncilArea_Yarra			2.188e+05	6.43e+04	3.405
0.001	9.28e+04	3.45e+05			
CouncilArea_Yarra Ranges			9.894e+04	2.08e+05	0.475
0.635	-3.09e+05	5.07e+05			
RegionName_Eastern Victoria			2.56e+05	1.79e+05	1.427
0.154	-9.57e+04	6.08e+05			
RegionName_Northern Metropolitan			-1.485e+05	5.46e+04	-2.721
0.007	-2.56e+05	-4.15e+04			
RegionName_Northern Victoria			2.239e+05	1.55e+05	1.447
0.148	-7.95e+04	5.27e+05			
RegionName_South-Eastern Metropolitan			-2.656e+04	6.17e+04	-0.430
0.667	-1.48e+05	9.44e+04			
RegionName_Southern Metropolitan			-1.65e+05	4.4e+04	-3.747
0.000	-2.51e+05	-7.87e+04			
RegionName_Western Metropolitan			-1.694e+05	7.96e+04	-2.129
0.033	-3.25e+05	-1.34e+04			
RegionName_Western Victoria			2.142e+05	1.73e+05	1.241
0.215	-1.24e+05	5.52e+05			
=====					
Omnibus:	3689.264	Durbin-Watson:	1.997		
Prob(Omnibus):	0.000	Jarque-Bera (JB):	127784.945		
Skew:	2.295	Prob(JB):	0.00		
Kurtosis:	24.830	Cond. No.	7.69e+20		
=====					

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[2] The smallest eigenvalue is 7.65e-31. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

### 1.6.1 Alejandro Gleason Méndez - ag77698