Final Project Report(Milestone 1 to 5)

Milestone 1 : Data Source

https://www.kaggle.com/c/zillow-prize-1 (https://www.kaggle.com/c/zillow-prize-1)

Description

There are two data sets with over 1 million records each and 58 columns. properties_2016 and properties_2017 datasets contain data for each year. The data we will use for this project will be a small sample of the master data.

The two datasets are linked by parcleid.

I transactions dataset, the trabsaction date shows the date the property was sold and logerror is the log10(estimated price - price sold).

Properties dataset has the physical information about the properities. The columns on the properties dataset will have to be renamed. Subsets of data can be used to group by region, and other features such as number of bedrooms, square footage, etc.

In [321]:

```python
# Load Libraries
import pandas as pd
import matplotlib.pyplot as plt
import xlrd
import numpy as np
# Load Data
transactions_2016 = "Data/transactions_2016.json"
transactions_2017 = "Data/transactions_2017.json"

properties_2016  =  "Data/properties_2016.csv"
properties_2017  =  "Data/properties_2017.csv"
data_dictionary = "Data/data_dictionary.xlsx"

transactions_2016 = pd.read_json(transactions_2016)
transactions_2017 = pd.read_json(transactions_2017)
properties_2016 = pd.read_csv(properties_2016)
properties_2017 = pd.read_csv(properties_2017)
data_dictionary = pd.read_excel(data_dictionary)
```

```
c:\users\safar\documents\github\safarie1103\bellevue university\courses\d
sc540\venv\lib\site-packages\IPython\core\interactiveshell.py:3063: Dtype
Warning: Columns (50) have mixed types.Specify dtype option on import or
set low_memory=False.
  interactivity=interactivity, compiler=compiler, result=result)
c:\users\safar\documents\github\safarie1103\bellevue university\courses\d
sc540\venv\lib\site-packages\IPython\core\interactiveshell.py:3063: Dtype
Warning: Columns (23,50) have mixed types.Specify dtype option on import
or set low_memory=False.
  interactivity=interactivity, compiler=compiler, result=result)
```

In [322]:

```python
transactions_2016.head()
```

Out[322]:

	parcelid	logerror	transactiondate
0	11016594	0.0276	2016-01-01
1	14366692	-0.1684	2016-01-01
2	12098116	-0.0040	2016-01-01
3	12643413	0.0218	2016-01-02
4	14432541	-0.0050	2016-01-02

In [323]:
```
properties_2016.head()
```

Out[323]:

	Unnamed: 0	parcelid	airconditioningtypeid	architecturalstyletypeid	basements
0	0	10754147	NaN	NaN	N
1	1	10759547	NaN	NaN	N
2	2	10843547	NaN	NaN	N
3	3	10859147	NaN	NaN	N
4	4	10879947	NaN	NaN	N

5 rows × 59 columns

In [324]:
```
print(len(properties_2016.columns))
print(properties_2016.columns)
```

```
59
Index(['Unnamed: 0', 'parcelid', 'airconditioningtypeid',
       'architecturalstyletypeid', 'basementsqft', 'bathroomcnt', 'bedroo
mcnt',
       'buildingclasstypeid', 'buildingqualitytypeid', 'calculatedbathnb
r',
       'decktypeid', 'finishedfloor1squarefeet',
       'calculatedfinishedsquarefeet', 'finishedsquarefeet12',
       'finishedsquarefeet13', 'finishedsquarefeet15', 'finishedsquarefee
t50',
       'finishedsquarefeet6', 'fips', 'fireplacecnt', 'fullbathcnt',
       'garagecarcnt', 'garagetotalsqft', 'hashottuborspa',
       'heatingorsystemtypeid', 'latitude', 'longitude', 'lotsizesquarefe
et',
       'poolcnt', 'poolsizesum', 'pooltypeid10', 'pooltypeid2', 'pooltype
id7',
       'propertycountylandusecode', 'propertylandusetypeid',
       'propertyzoningdesc', 'rawcensustractandblock', 'regionidcity',
       'regionidcounty', 'regionidneighborhood', 'regionidzip', 'roomcn
t',
       'storytypeid', 'threequarterbathnbr', 'typeconstructiontypeid',
       'unitcnt', 'yardbuildingsqft17', 'yardbuildingsqft26', 'yearbuil
t',
       'numberofstories', 'fireplaceflag', 'structuretaxvaluedollarcnt',
       'taxvaluedollarcnt', 'assessmentyear', 'landtaxvaluedollarcnt',
       'taxamount', 'taxdelinquencyflag', 'taxdelinquencyyear',
       'censustractandblock'],
      dtype='object')
```

In [325]:
```
print(len(properties_2017.columns))
print(properties_2017.columns)
```

```
59
Index(['Unnamed: 0', 'parcelid', 'airconditioningtypeid',
       'architecturalstyletypeid', 'basementsqft', 'bathroomcnt', 'bedroo
mcnt',
       'buildingclasstypeid', 'buildingqualitytypeid', 'calculatedbathnb
r',
       'decktypeid', 'finishedfloor1squarefeet',
       'calculatedfinishedsquarefeet', 'finishedsquarefeet12',
       'finishedsquarefeet13', 'finishedsquarefeet15', 'finishedsquarefee
t50',
       'finishedsquarefeet6', 'fips', 'fireplacecnt', 'fullbathcnt',
       'garagecarcnt', 'garagetotalsqft', 'hashottuborspa',
       'heatingorsystemtypeid', 'latitude', 'longitude', 'lotsizesquarefe
et',
       'poolcnt', 'poolsizesum', 'pooltypeid10', 'pooltypeid2', 'pooltype
id7',
       'propertycountylandusecode', 'propertylandusetypeid',
       'propertyzoningdesc', 'rawcensustractandblock', 'regionidcity',
       'regionidcounty', 'regionidneighborhood', 'regionidzip', 'roomcn
t',
       'storytypeid', 'threequarterbathnbr', 'typeconstructiontypeid',
       'unitcnt', 'yardbuildingsqft17', 'yardbuildingsqft26', 'yearbuil
t',
       'numberofstories', 'fireplaceflag', 'structuretaxvaluedollarcnt',
       'taxvaluedollarcnt', 'assessmentyear', 'landtaxvaluedollarcnt',
       'taxamount', 'taxdelinquencyflag', 'taxdelinquencyyear',
       'censustractandblock'],
      dtype='object')
```

In [326]:
```
print(len(transactions_2016.columns))
print(transactions_2016.columns)
```

```
3
Index(['parcelid', 'logerror', 'transactiondate'], dtype='object')
```

In [327]:
```
print(len(transactions_2017.columns))
print(transactions_2017.columns)
```

```
3
Index(['parcelid', 'logerror', 'transactiondate'], dtype='object')
```

In [328]:

```
data_dictionary.head()
```

Out[328]:

	Feature	Description
0	'airconditioningtypeid'	Type of cooling system present in the home (i...
1	'architecturalstyletypeid'	Architectural style of the home (i.e. ranch, ...
2	'basementsqft'	Finished living area below or partially below...
3	'bathroomcnt'	Number of bathrooms in home including fractio...
4	'bedroomcnt'	Number of bedrooms in home

Milestone 2 : Cleaning/formatting flat file sources

We will first combine the properties_2016 and properties_2017 and calle the result properties. We will also combine the two transactions datasets.

In [329]:

```
properties = pd.concat([properties_2016,properties_2017],axis=0)
print(properties_2016.shape)
print(properties_2017.shape)
print(properties.shape)
```

```
(20000, 59)
(20000, 59)
(40000, 59)
```

In [330]:

```
transactions = pd.concat([transactions_2016,transactions_2017],axis=0)
print(properties_2016.shape)
print(properties_2017.shape)
print(properties.shape)
```

```
(20000, 59)
(20000, 59)
(40000, 59)
```

In [331]:

```
properties.columns
```

Out[331]:

```
Index(['Unnamed: 0', 'parcelid', 'airconditioningtypeid',
       'architecturalstyletypeid', 'basementsqft', 'bathroomcnt', 'bedroo
mcnt',
       'buildingclasstypeid', 'buildingqualitytypeid', 'calculatedbathnb
r',
       'decktypeid', 'finishedfloor1squarefeet',
       'calculatedfinishedsquarefeet', 'finishedsquarefeet12',
       'finishedsquarefeet13', 'finishedsquarefeet15', 'finishedsquarefee
t50',
       'finishedsquarefeet6', 'fips', 'fireplacecnt', 'fullbathcnt',
       'garagecarcnt', 'garagetotalsqft', 'hashottuborspa',
       'heatingorsystemtypeid', 'latitude', 'longitude', 'lotsizesquarefe
et',
       'poolcnt', 'poolsizesum', 'pooltypeid10', 'pooltypeid2', 'pooltype
id7',
       'propertycountylandusecode', 'propertylandusetypeid',
       'propertyzoningdesc', 'rawcensustractandblock', 'regionidcity',
       'regionidcounty', 'regionidneighborhood', 'regionidzip', 'roomcn
t',
       'storytypeid', 'threequarterbathnbr', 'typeconstructiontypeid',
       'unitcnt', 'yardbuildingsqft17', 'yardbuildingsqft26', 'yearbuil
t',
       'numberofstories', 'fireplaceflag', 'structuretaxvaluedollarcnt',
       'taxvaluedollarcnt', 'assessmentyear', 'landtaxvaluedollarcnt',
       'taxamount', 'taxdelinquencyflag', 'taxdelinquencyyear',
       'censustractandblock'],
      dtype='object')
```

Get rid of the Unamed column.

```
In [332]: properties = properties.loc[:, ~properties.columns.str.contains('^Unname
          d')]
          properties.columns
```

```
Out[332]: Index(['parcelid', 'airconditioningtypeid', 'architecturalstyletypeid',
                 'basementsqft', 'bathroomcnt', 'bedroomcnt', 'buildingclasstypei
          d',
                 'buildingqualitytypeid', 'calculatedbathnbr', 'decktypeid',
                 'finishedfloor1squarefeet', 'calculatedfinishedsquarefeet',
                 'finishedsquarefeet12', 'finishedsquarefeet13', 'finishedsquarefee
          t15',
                 'finishedsquarefeet50', 'finishedsquarefeet6', 'fips', 'fireplacec
          nt',
                 'fullbathcnt', 'garagecarcnt', 'garagetotalsqft', 'hashottuborsp
          a',
                 'heatingorsystemtypeid', 'latitude', 'longitude', 'lotsizesquarefe
          et',
                 'poolcnt', 'poolsizesum', 'pooltypeid10', 'pooltypeid2', 'pooltype
          id7',
                 'propertycountylandusecode', 'propertylandusetypeid',
                 'propertyzoningdesc', 'rawcensustractandblock', 'regionidcity',
                 'regionidcounty', 'regionidneighborhood', 'regionidzip', 'roomcn
          t',
                 'storytypeid', 'threequarterbathnbr', 'typeconstructiontypeid',
                 'unitcnt', 'yardbuildingsqft17', 'yardbuildingsqft26', 'yearbuil
          t',
                 'numberofstories', 'fireplaceflag', 'structuretaxvaluedollarcnt',
                 'taxvaluedollarcnt', 'assessmentyear', 'landtaxvaluedollarcnt',
                 'taxamount', 'taxdelinquencyflag', 'taxdelinquencyyear',
                 'censustractandblock'],
                dtype='object')
```

Rename column names in properties dataset.

In [333]:

```
properties = properties.rename(columns=
                        {
  'parcelid':'parcelid',
  'yearbuilt':'build_year',
  'basementsqft':'area_basement',
  'yardbuildingsqft17':'area_patio',
  'yardbuildingsqft26':'area_shed',
  'poolsizesum':'area_pool',
  'lotsizesquarefeet':'area_lot',
  'garagetotalsqft':'area_garage',
  'finishedfloor1squarefeet':'area_firstfloor_finished',
  'calculatedfinishedsquarefeet':'area_total_calc',
  'finishedsquarefeet6':'area_base',
  'finishedsquarefeet12':'area_live_finished',
  'finishedsquarefeet13':'area_liveperi_finished',
  'finishedsquarefeet15':'area_total_finished',
  'finishedsquarefeet50':'area_unknown',
  'unitcnt': 'num_unit',
  'numberofstories': 'num_story',
  'roomcnt':'num_room',
  'bathroomcnt':'num_bathroom',
  'bedroomcnt':'num_bedroom',
  'calculatedbathnbr':'num_bathroom_calc',
  'fullbathcnt':'num_bath',
  'threequarterbathnbr':'num_75_bath',
  'fireplacecnt':'num_fireplace',
  'poolcnt': 'num_pool',
  'garagecarcnt':'num_garage',
  'regionidcounty':'region_county',
  'regionidcity':'region_city',
  'regionidzip':'region_zip',
  'regionidneighborhood':'region_neighbor',
  'taxvaluedollarcnt':'tax_total',
  'structuretaxvaluedollarcnt':'tax_building',
  'landtaxvaluedollarcnt':'tax_land',
  'taxamount':'tax_property',
  'assessmentyear':'tax_year',
  'taxdelinquencyflag':'tax_delinquency',
  'taxdelinquencyyear':'tax_delinquency_year',
  'propertyzoningdesc':'zoning_property',
  'propertylandusetypeid':'zoning_landuse',
  'propertycountylandusecode':'zoning_landuse_county',
  'fireplaceflag':'flag_fireplace',
  'hashottuborspa':'flag_tub',
  'buildingqualitytypeid':'quality',
  'buildingclasstypeid':'framing',
  'typeconstructiontypeid':'material',
  'decktypeid':'deck',
  'storytypeid':'story',
  'heatingorsystemtypeid':'heating',
  'airconditioningtypeid':'aircon',
  'architecturalstyletypeid':'architectural_style'
})
```

```
In [334]: properties.columns
```

```
Out[334]: Index(['parcelid', 'aircon', 'architectural_style', 'area_basement',
                 'num_bathroom', 'num_bedroom', 'framing', 'quality',
                 'num_bathroom_calc', 'deck', 'area_firstfloor_finished',
                 'area_total_calc', 'area_live_finished', 'area_liveperi_finished',
                 'area_total_finished', 'area_unknown', 'area_base', 'fips',
                 'num_fireplace', 'num_bath', 'num_garage', 'area_garage', 'flag_tu
          b',
                 'heating', 'latitude', 'longitude', 'area_lot', 'num_pool', 'area_
          pool',
                 'pooltypeid10', 'pooltypeid2', 'pooltypeid7', 'zoning_landuse_coun
          ty',
                 'zoning_landuse', 'zoning_property', 'rawcensustractandblock',
                 'region_city', 'region_county', 'region_neighbor', 'region_zip',
                 'num_room', 'story', 'num_75_bath', 'material', 'num_unit',
                 'area_patio', 'area_shed', 'build_year', 'num_story', 'flag_firepl
          ace',
                 'tax_building', 'tax_total', 'tax_year', 'tax_land', 'tax_propert
          y',
                 'tax_delinquency', 'tax_delinquency_year', 'censustractandblock'],
                dtype='object')
```

```
In [335]: # Check new column names
          properties[['num_bedroom','num_bathroom']]
```

Out[335]:

	num_bedroom	num_bathroom
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0
...	...	...
19995	2.0	1.0
19996	5.0	3.0
19997	8.0	5.0
19998	4.0	2.0
19999	2.0	1.0

40000 rows × 2 columns

Rename column names in transactions dataset.

```
In [336]: transactions = transactions.rename(columns={'parcelid':'parcelid','date':
          'transactiondate'})
```

```
In [337]: transactions.columns
```

```
Out[337]: Index(['parcelid', 'logerror', 'transactiondate'], dtype='object')
```

Check out the new columns

```
In [338]: transactions[['parcelid','transactiondate']]
```

Out[338]:

	parcelid	transactiondate
0	11016594	2016-01-01
1	14366692	2016-01-01
2	12098116	2016-01-01
3	12643413	2016-01-02
4	14432541	2016-01-02
...	...	...
77608	10833991	2017-09-20
77609	11000655	2017-09-20
77610	17239384	2017-09-21
77611	12773139	2017-09-21
77612	12826780	2017-09-25

167888 rows × 2 columns

```
In [339]: propertiesAndTransactions = pd.merge(properties,transactions,on='parcelid')
```

check out the merge

```
In [340]: propertiesAndTransactions[['parcelid','num_bedroom','transactiondate','logerror']].head()
```

Out[340]:

	parcelid	num_bedroom	transactiondate	logerror
0	17054981	4.0	2017-06-15	-0.013099
1	17054981	4.0	2017-06-15	-0.013099
2	17055743	3.0	2017-07-26	0.073985
3	17055743	3.0	2017-07-26	0.073985
4	17068109	3.0	2017-07-28	0.071886

let's take of missings

In [341]:

```
column_names = propertiesAndTransactions.columns
print('sum\n', propertiesAndTransactions.isnull()[column_names].sum())
```

```
sum
 parcelid                       0
aircon                      1485
architectural_style         2234
area_basement               2234
num_bathroom                   0
num_bedroom                    0
framing                     2234
quality                      705
num_bathroom_calc             26
deck                        2214
area_firstfloor_finished    2000
area_total_calc                9
area_live_finished           102
area_liveperi_finished      2234
area_total_finished         2145
area_unknown                2000
area_base                   2230
fips                           0
num_fireplace               1982
num_bath                      26
num_garage                  1593
area_garage                 1593
flag_tub                    2192
heating                      752
latitude                       0
longitude                      0
area_lot                     216
num_pool                    1708
area_pool                   2206
pooltypeid10                2216
pooltypeid2                 2210
pooltypeid7                 1732
zoning_landuse_county          0
zoning_landuse                 0
zoning_property              678
rawcensustractandblock         0
region_city                   42
region_county                  0
region_neighbor             1186
region_zip                     2
num_room                       0
story                       2234
num_75_bath                 1984
material                    2234
num_unit                     679
area_patio                  2137
area_shed                   2234
build_year                    11
num_story                   1792
flag_fireplace              2234
tax_building                   6
tax_total                      0
tax_year                       0
tax_land                       0
tax_property                   0
tax_delinquency             2166
```

```
tax_delinquency_year        2166
censustractandblock            8
logerror                       0
transactiondate                0
dtype: int64
```

In [342]:

```
print('mean\n', propertiesAndTransactions.isnull()[column_names].mean())
```

```
mean
 parcelid                    0.000000
aircon                      0.664727
architectural_style         1.000000
area_basement               1.000000
num_bathroom                0.000000
num_bedroom                 0.000000
framing                     1.000000
quality                     0.315577
num_bathroom_calc           0.011638
deck                        0.991047
area_firstfloor_finished    0.895255
area_total_calc             0.004029
area_live_finished          0.045658
area_liveperi_finished      1.000000
area_total_finished         0.960161
area_unknown                0.895255
area_base                   0.998209
fips                        0.000000
num_fireplace               0.887198
num_bath                    0.011638
num_garage                  0.713071
area_garage                 0.713071
flag_tub                    0.981200
heating                     0.336616
latitude                    0.000000
longitude                   0.000000
area_lot                    0.096688
num_pool                    0.764548
area_pool                   0.987466
pooltypeid10                0.991943
pooltypeid2                 0.989257
pooltypeid7                 0.775291
zoning_landuse_county       0.000000
zoning_landuse              0.000000
zoning_property             0.303491
rawcensustractandblock      0.000000
region_city                 0.018800
region_county               0.000000
region_neighbor             0.530886
region_zip                  0.000895
num_room                    0.000000
story                       1.000000
num_75_bath                 0.888093
material                    1.000000
num_unit                    0.303939
area_patio                  0.956580
area_shed                   1.000000
build_year                  0.004924
num_story                   0.802149
flag_fireplace              1.000000
tax_building                0.002686
tax_total                   0.000000
tax_year                    0.000000
tax_land                    0.000000
tax_property                0.000000
tax_delinquency             0.969561
```

```
tax_delinquency_year        0.969561
censustractandblock         0.003581
logerror                    0.000000
transactiondate             0.000000
dtype: float64
```

Let's look at columns woth more than 80% missing values

In [343]:

```
propertiesAndTransactions.isnull()[column_names].sum()
# this shows columns and the number of NaN's.Note parcelID has no missing
values.
```

```
Out[343]: parcelid                       0
          aircon                      1485
          architectural_style         2234
          area_basement               2234
          num_bathroom                   0
          num_bedroom                    0
          framing                     2234
          quality                      705
          num_bathroom_calc             26
          deck                        2214
          area_firstfloor_finished    2000
          area_total_calc                9
          area_live_finished           102
          area_liveperi_finished      2234
          area_total_finished         2145
          area_unknown                2000
          area_base                   2230
          fips                           0
          num_fireplace               1982
          num_bath                      26
          num_garage                  1593
          area_garage                 1593
          flag_tub                    2192
          heating                      752
          latitude                       0
          longitude                      0
          area_lot                     216
          num_pool                    1708
          area_pool                   2206
          pooltypeid10                2216
          pooltypeid2                 2210
          pooltypeid7                 1732
          zoning_landuse_county          0
          zoning_landuse                 0
          zoning_property              678
          rawcensustractandblock         0
          region_city                   42
          region_county                  0
          region_neighbor             1186
          region_zip                     2
          num_room                       0
          story                       2234
          num_75_bath                 1984
          material                    2234
          num_unit                     679
          area_patio                  2137
          area_shed                   2234
          build_year                    11
          num_story                   1792
          flag_fireplace              2234
          tax_building                   6
          tax_total                      0
          tax_year                       0
          tax_land                       0
          tax_property                   0
          tax_delinquency             2166
          tax_delinquency_year        2166
```

```
censustractandblock          8
logerror                     0
transactiondate              0
dtype: int64
```

Make a list of columns with moe than 80% missing data

```
In [344]: remove_columns = propertiesAndTransactions.columns[propertiesAndTransacti
          ons.isnull().mean() > .8]
          print(remove_columns)
```

```
Index(['architectural_style', 'area_basement', 'framing', 'deck',
       'area_firstfloor_finished', 'area_liveperi_finished',
       'area_total_finished', 'area_unknown', 'area_base', 'num_fireplac
e',
       'flag_tub', 'area_pool', 'pooltypeid10', 'pooltypeid2', 'story',
       'num_75_bath', 'material', 'area_patio', 'area_shed', 'num_story',
       'flag_fireplace', 'tax_delinquency', 'tax_delinquency_year'],
      dtype='object')
```

Drop the columns

```
In [345]: propertiesAndTransactions = propertiesAndTransactions.drop(columns = remo
          ve_columns)
```

Check results

```
In [346]: print(len(propertiesAndTransactions.columns))
          print(propertiesAndTransactions.columns)
```

```
37
Index(['parcelid', 'aircon', 'num_bathroom', 'num_bedroom', 'quality',
       'num_bathroom_calc', 'area_total_calc', 'area_live_finished', 'fip
s',
       'num_bath', 'num_garage', 'area_garage', 'heating', 'latitude',
       'longitude', 'area_lot', 'num_pool', 'pooltypeid7',
       'zoning_landuse_county', 'zoning_landuse', 'zoning_property',
       'rawcensustractandblock', 'region_city', 'region_county',
       'region_neighbor', 'region_zip', 'num_room', 'num_unit', 'build_ye
ar',
       'tax_building', 'tax_total', 'tax_year', 'tax_land', 'tax_propert
y',
       'censustractandblock', 'logerror', 'transactiondate'],
      dtype='object')
```

Check results

```
In [347]: print(len(propertiesAndTransactions.columns))
          print(propertiesAndTransactions.columns)
```

```
37
Index(['parcelid', 'aircon', 'num_bathroom', 'num_bedroom', 'quality',
       'num_bathroom_calc', 'area_total_calc', 'area_live_finished', 'fip
s',
       'num_bath', 'num_garage', 'area_garage', 'heating', 'latitude',
       'longitude', 'area_lot', 'num_pool', 'pooltypeid7',
       'zoning_landuse_county', 'zoning_landuse', 'zoning_property',
       'rawcensustractandblock', 'region_city', 'region_county',
       'region_neighbor', 'region_zip', 'num_room', 'num_unit', 'build_ye
ar',
       'tax_building', 'tax_total', 'tax_year', 'tax_land', 'tax_propert
y',
       'censustractandblock', 'logerror', 'transactiondate'],
      dtype='object')
```

Let's check the missing values mean

In [348]:

```
print('mean\n', propertiesAndTransactions.isnull()[propertiesAndTransacti
ons.columns].mean())
# we see the means to all be below 80%.
```

```
mean
 parcelid                 0.000000
aircon                   0.664727
num_bathroom             0.000000
num_bedroom              0.000000
quality                  0.315577
num_bathroom_calc        0.011638
area_total_calc          0.004029
area_live_finished       0.045658
fips                     0.000000
num_bath                 0.011638
num_garage               0.713071
area_garage              0.713071
heating                  0.336616
latitude                 0.000000
longitude                0.000000
area_lot                 0.096688
num_pool                 0.764548
pooltypeid7              0.775291
zoning_landuse_county    0.000000
zoning_landuse           0.000000
zoning_property          0.303491
rawcensustractandblock   0.000000
region_city              0.018800
region_county            0.000000
region_neighbor          0.530886
region_zip               0.000895
num_room                 0.000000
num_unit                 0.303939
build_year               0.004924
tax_building             0.002686
tax_total                0.000000
tax_year                 0.000000
tax_land                 0.000000
tax_property             0.000000
censustractandblock      0.003581
logerror                 0.000000
transactiondate          0.000000
dtype: float64
```

Are there any duplicate?

In [349]:

```
propertiesAndTransactions[propertiesAndTransactions.duplicated(keep=False)]
# There are no duplocate rows; however, there are duplicate parcelIDs and corresponding latitude and Longitude.
```

Out[349]:

	parcelid	aircon	num_bathroom	num_bedroom	quality	num_bathroom_calc	are

0 rows × 37 columns

In [350]:

```
propertiesAndTransactions
```

Out[350]:

	parcelid	aircon	num_bathroom	num_bedroom	quality	num_bathroom_cal
0	17054981	NaN	5.0	4.0	NaN	5.
1	17054981	NaN	5.0	4.0	NaN	5.
2	17055743	NaN	2.0	3.0	NaN	2.
3	17055743	NaN	2.0	3.0	NaN	2.
4	17068109	NaN	1.5	3.0	NaN	1.
...	...	...	...	...	...	..
2229	11769554	NaN	3.0	4.0	4.0	3.
2230	11778756	NaN	2.0	7.0	7.0	2.
2231	11778756	NaN	2.0	7.0	4.0	2.
2232	11779780	1.0	2.0	2.0	10.0	2.
2233	11779780	1.0	2.0	2.0	11.0	2.

2234 rows × 37 columns

In [351]:

```
# Write scraped data to a file for safe keeps and also to avoid rescraping during development
propertiesAndTransactions.to_csv("data/propertiesAndTransactions.csv")
```

The two datasets have been merged, columns with more than 80% missing values were removed. The final dataset 'propertiesAndTransactions' will be used in the next milestone.

Milestone 3. Webscaraping Data Source

Description

Using webscraping techniques, we will use 'latitude', 'longitude' from properties dataset to access properties and get current data for those locations. The property description of homes in given region will be stored into a dataset with as many features as in properties dataset we can grab. This dataset can then be used to do some price comparision between properties in 2016 and 2017. Getting data from years prior(say 10 years), we will be able to create trend charts and see market fluctuations.

In [222]:

```
# Build a table consisiting of the parcelID, latitude and longitude of the properties.
# This table will be used to get data from www.trulia.com by web scraping

LonLat = pd.DataFrame(propertiesAndTransactions[['parcelid','latitude','longitude']])
LonLat
```

Out[222]:

```
      parcelid  latitude   longitude
   0  17054981  34449407  -119254052
   1  17054981  34449407  -119254052
   2  17055743  34454169  -119237898
   3  17055743  34454169  -119237898
   4  17068109  34365693  -119448392
 ...       ...       ...         ...
2229  11769554  34006415  -118246669
2230  11778756  34050678  -118282732
2231  11778756  34050678  -118282732
2232  11779780  34045100  -118261000
2233  11779780  34045100  -118261000

2234 rows × 3 columns
```

In [223]:

```
# We will remove duplicate parcelIDs here since we are only interested in
comparable values near each parcelID.
LonLat = LonLat.sort_values('parcelid', ascending=False)
LonLat = LonLat.drop_duplicates()
LonLat.reset_index(drop=True)
LonLat
```

Out[223]:

	parcelid	latitude	longitude
1761	17299670	34186100	-118767000
107	17296734	34174051	-118757031
1758	17294231	34153879	-118839561
1756	17293716	34152179	-118851454
1427	17292856	34125457	-118891074
...	...	...	...
112	10726315	34184300	-118657000
110	10725532	34196000	-118658000
1767	10722858	34195746	-118624097
108	10722336	34199100	-118633000
1763	10719731	34206094	-118620655

1096 rows × 3 columns

In [224]:

```
print('sum\n', LonLat.isnull()[['parcelid','latitude','longitude']].sum
())
```

```
sum
 parcelid     0
latitude     0
longitude    0
dtype: int64
```

In [225]:

```
# This dictionary is used to return state code. trulia requires the state
code rather than state name
us_state_abbrev = {
    'Alabama': 'AL',
    'Alaska': 'AK',
    'American Samoa': 'AS',
    'Arizona': 'AZ',
    'Arkansas': 'AR',
    'California': 'CA',
    'Colorado': 'CO',
    'Connecticut': 'CT',
    'Delaware': 'DE',
    'District of Columbia': 'DC',
    'Florida': 'FL',
    'Georgia': 'GA',
    'Guam': 'GU',
    'Hawaii': 'HI',
    'Idaho': 'ID',
    'Illinois': 'IL',
    'Indiana': 'IN',
    'Iowa': 'IA',
    'Kansas': 'KS',
    'Kentucky': 'KY',
    'Louisiana': 'LA',
    'Maine': 'ME',
    'Maryland': 'MD',
    'Massachusetts': 'MA',
    'Michigan': 'MI',
    'Minnesota': 'MN',
    'Mississippi': 'MS',
    'Missouri': 'MO',
    'Montana': 'MT',
    'Nebraska': 'NE',
    'Nevada': 'NV',
    'New Hampshire': 'NH',
    'New Jersey': 'NJ',
    'New Mexico': 'NM',
    'New York': 'NY',
    'North Carolina': 'NC',
    'North Dakota': 'ND',
    'Northern Mariana Islands':'MP',
    'Ohio': 'OH',
    'Oklahoma': 'OK',
    'Oregon': 'OR',
    'Pennsylvania': 'PA',
    'Puerto Rico': 'PR',
    'Rhode Island': 'RI',
    'South Carolina': 'SC',
    'South Dakota': 'SD',
    'Tennessee': 'TN',
    'Texas': 'TX',
    'Utah': 'UT',
    'Vermont': 'VT',
    'Virgin Islands': 'VI',
    'Virginia': 'VA',
    'Washington': 'WA',
    'West Virginia': 'WV',
    'Wisconsin': 'WI',
    'Wyoming': 'WY'
}

abbrev_us_state = dict(map(reversed, us_state_abbrev.items()))
```

In [43]:

```
import urllib.request
import urllib.parse
import urllib.error
import json
from bs4 import BeautifulSoup
from urllib.request import Request, urlopen
import geopy
from geopy.geocoders import Nominatim

def create_url(city,state,zipcode):
    # Creating trulia URL based on the filter.

    url = "https://www.trulia.com/" + state + "/" + city + "/" + zipcode
    return url

def get_response(url):
    ret = None
    try:
        for i in range(5):
            response = requests.get(url, headers={'User-Agent': 'Mozilla/
5.0'})
            print("status code received:", response.status_code)
            if (response.status_code != 200):
                return None
            else:
                return response
    except:
        print('exception in get_response')
        return None

def GetCityStateZip(lat,lon):
    lat = lat/10**6
    lon = lon/10**6
    geolocator = Nominatim(timeout=5)
    #print(location.raw)
    try:
        location = geolocator.reverse((lat, lon))
        city = location.raw['address']['city']
        state = us_state_abbrev[location.raw['address']['state']]
        zipcode = location.raw['address']['postcode'].split('-')[0]
    except:
        city = ""
        state = ""
        zipcode = ""

    return city,state,zipcode
```

In [44]:

```
def GetComp(parcelId,latitude,longitude):
    city,state,zipcode = GetCityStateZip(latitude,longitude)
    #print(parcelId,latitude,longitude)
    #print("city=", city)
    #print("state=", state)
    #print("zipcode=",zipcode)

    emptylistings_json = {}
    emptylistings_json['parcelId'] = {0:parcelId}
    emptylistings_json['price'] = {0:np.nan}
    emptylistings_json['bedrooms'] = {0:np.nan}
    emptylistings_json['bathrooms'] = {0:np.nan}
    emptylistings_json['floorSpace'] = {0:np.nan}
    emptylistings_json['region'] = {0:np.nan}

    if (city == "" or state == "" or state == ""):
        return(pd.DataFrame(emptylistings_json))

    url = create_url(city,state,zipcode)

    #req = Requests(url, headers={'User-Agent': 'Mozilla/5.0'})
    #webpage = urlopen(req).read()
    #soup = BeautifulSoup(webpage, 'html.parser')

    response = get_response(url)
    #print(response.text)
    if not response:
        print("Failed to fetch the page, please check `response.html` to see the response received from zillow.com.")
        return(pd.DataFrame(emptylistings_json))

    soup = BeautifulSoup(response.text, 'html.parser')

    html = soup.prettify('utf-8')

    details = {}
    parcels = {}
    listings_json = {}
    index = 0

    for price in  soup.findAll('div',attrs={'data-testid': 'property-price'}):
        details.update({index:price.text.strip()})
        parcels.update({index:parcelId})
        index = index + 1

    listings_json['parcelId'] = {}
    listings_json['parcelId']  = parcels
    listings_json['price'] = {}
    listings_json['price']  = details
    #print(listings_json['price'])



    details = {}
    index = 0
    for bedroom  in  soup.findAll('div',attrs={'data-testid': 'property-beds'}):
        details.update({index:bedroom.text.strip()})
        index = index + 1

    listings_json['bedrooms'] = {}
    listings_json['bedrooms']  = details
    #print(listings_json)



    details = {}
    index = 0
    for bathroom  in  soup.findAll('div',attrs={'data-testid': 'property-baths'}):
        details.update({index:bathroom.text.strip()})
        index = index + 1

    listings_json['bathrooms'] = {}
    listings_json['bathrooms']  = details
    #print(listings_json)



    details = {}
    index = 0
    for floorSpace  in  soup.findAll('div',attrs={'data-testid': 'property-floorSpace'}):
        details.update({index:floorSpace.text.strip()})
        index = index + 1

    listings_json['floorSpace'] = {}
    listings_json['floorSpace']  = details
    #print(listings_json)



    details = {}
    index = 0
    for region  in  soup.findAll('div',attrs={'data-testid': 'property-region'}):
        details.update({index:region.text.strip()})
        index = index + 1

    listings_json['region'] = {}
    listings_json['region']  = details
    #print(listings_json)

    #listings_table = pd.DataFrame()

    #with open('house_details.json', 'w') as outfile:
    #    json.dump(listings_json, outfile, indent=4)
    #listings_table = pd.read_json("house_details.json")
    return pd.DataFrame(listings_json)
```

In [45]:
```
LonLat[:5]
```

Out[45]:

	parcelid	latitude	longitude
1761	17299670	34186100	-118767000
107	17296734	34174051	-118757031
1758	17294231	34153879	-118839561
1756	17293716	34152179	-118851454
1427	17292856	34125457	-118891074

Here we get 20 compare properties for the parcelIDs. Note that a parcelID from propertiesAndTransactions table may have one ore more comps near it's latitude and longitude. This process sometime times out. We have taken care to continue collecting even after such exceptions.

In [ ]:
```
comp_listing_table = pd.DataFrame(columns={'parcelid','price','bedrooms',
'bathrooms','floorSpace','region'})

dfs = []
for index, row in LonLat[:20].iterrows():
    parcelId = row['parcelid']
    latitude = row['latitude']
    longitude = row['longitude']
    #print(parcelId,latitude,longitude)
    Temp_listing_table = GetComp(parcelId,latitude,longitude)
    #print(Temp_listing_table.shape)
    dfs.append(Temp_listing_table)
    #print(Temp_listing_table)


comp_listing_table = pd.concat(dfs, ignore_index=True)
```

In [47]:
```
print(comp_listing_table)
```

```
    parcelId  price  bedrooms  bathrooms  floorSpace  region
0   17299670    NaN       NaN        NaN         NaN     NaN
1   17296734    NaN       NaN        NaN         NaN     NaN
2   17294231    NaN       NaN        NaN         NaN     NaN
3   17293716    NaN       NaN        NaN         NaN     NaN
4   17292856    NaN       NaN        NaN         NaN     NaN
5   17291231    NaN       NaN        NaN         NaN     NaN
6   17290419    NaN       NaN        NaN         NaN     NaN
7   17290104    NaN       NaN        NaN         NaN     NaN
8   17289398    NaN       NaN        NaN         NaN     NaN
9   17287986    NaN       NaN        NaN         NaN     NaN
10  17285909    NaN       NaN        NaN         NaN     NaN
11  17283891    NaN       NaN        NaN         NaN     NaN
12  17283162    NaN       NaN        NaN         NaN     NaN
13  17280385    NaN       NaN        NaN         NaN     NaN
14  17276736    NaN       NaN        NaN         NaN     NaN
15  17276290    NaN       NaN        NaN         NaN     NaN
16  17275763    NaN       NaN        NaN         NaN     NaN
17  17275640    NaN       NaN        NaN         NaN     NaN
18  17274552    NaN       NaN        NaN         NaN     NaN
19  17273670    NaN       NaN        NaN         NaN     NaN
```

In [48]:
```
comp_listing_table.isnull()[comp_listing_table.columns].sum()
```

Out[48]:
```
parcelId       0
price         20
bedrooms      20
bathrooms     20
floorSpace    20
region        20
dtype: int64
```

In [398]:
```
comp_listing_table = comp_listing_table.dropna()
```

In [399]:
```
comp_listing_table.isnull()[comp_listing_table.columns].sum()
```

Out[399]:
```
parcelId      0
price         0
bedrooms      0
bathrooms     0
floorSpace    0
region        0
dtype: int64
```

In [400]:
```
comp_listing_table.shape
```

Out[400]:
```
(467, 6)
```

In [236]:

```
comp_listing_table
```

Out[236]:

	Unnamed: 0	parcelId	price	bedrooms	bathrooms	floorSpace	region
0	0	17294231	14999000.0	7	13.0	14073	Newbury Park, Thousand Oaks, CA
1	1	17294231	1450000.0	4	3.0	2568	Westlake Village, CA
2	2	17294231	1225000.0	4	3.0	2745	Westlake Village, CA
3	3	17294231	9990000.0	7	10.0	12656	Newbury Park, Thousand Oaks, CA
4	4	17294231	1150000.0	5	4.0	2393	Westlake Village, CA
...	...	...	...	...	...	...	...
462	462	17273670	897000.0	4	3.0	3259	Newbury Park, Thousand Oaks, CA
463	463	17273670	680000.0	4	3.0	2096	Newbury Park, Thousand Oaks, CA
464	464	17273670	569000.0	3	3.0	1550	Newbury Park, Thousand Oaks, CA
465	465	17273670	830000.0	3	3.0	2243	Newbury Park, Thousand Oaks, CA
466	466	17273670	999900.0	5	4.0	3780	Newbury Park, Thousand Oaks, CA

467 rows × 7 columns

prepare the dataset

In [228]:

```
comp_listing_table = comp_listing_table.loc[:, ~comp_listing_table.column
s.str.contains('^Unnamed')]
```

In [229]:

```
comp_listing_table['price']= comp_listing_table['price'].replace('[\$,]',
'', regex=True).astype(float)
comp_listing_table
```

```
c:\users\safar\documents\github\safarie1103\bellevue university\courses\d
sc540\venv\lib\site-packages\ipykernel_launcher.py:1: SettingWithCopyWarn
ing:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  """Entry point for launching an IPython kernel.
```

Out[229]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region
0	17294231	14999000.0	7bd	13ba	14,073 sqft	Newbury Park, Thousand Oaks, CA
1	17294231	1450000.0	4bd	3ba	2,568 sqft	Westlake Village, CA
2	17294231	1225000.0	4bd	3ba	2,745 sqft	Westlake Village, CA
3	17294231	9990000.0	7bd	10ba	12,656 sqft	Newbury Park, Thousand Oaks, CA
4	17294231	1150000.0	5bd	4ba	2,393 sqft	Westlake Village, CA
...	...	...	...	...	...	...
462	17273670	897000.0	4bd	3ba	3,259 sqft	Newbury Park, Thousand Oaks, CA
463	17273670	680000.0	4bd	3ba	2,096 sqft	Newbury Park, Thousand Oaks, CA
464	17273670	569000.0	3bd	3ba	1,550 sqft	Newbury Park, Thousand Oaks, CA
465	17273670	830000.0	3bd	3ba	2,243 sqft	Newbury Park, Thousand Oaks, CA
466	17273670	999900.0	5bd	4ba	3,780 sqft	Newbury Park, Thousand Oaks, CA

467 rows × 6 columns

```
In [230]: comp_listing_table['bedrooms']= comp_listing_table['bedrooms'].replace('bd', '', regex=True).astype(int)
          comp_listing_table
```

```
c:\users\safar\documents\github\safarie1103\bellevue university\courses\dsc540\venv\lib\site-packages\ipykernel_launcher.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  """Entry point for launching an IPython kernel.
```

Out[230]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region
0	17294231	14999000.0	7	13ba	14,073 sqft	Newbury Park, Thousand Oaks, CA
1	17294231	1450000.0	4	3ba	2,568 sqft	Westlake Village, CA
2	17294231	1225000.0	4	3ba	2,745 sqft	Westlake Village, CA
3	17294231	9990000.0	7	10ba	12,656 sqft	Newbury Park, Thousand Oaks, CA
4	17294231	1150000.0	5	4ba	2,393 sqft	Westlake Village, CA
...	...	...	...	...	...	...
462	17273670	897000.0	4	3ba	3,259 sqft	Newbury Park, Thousand Oaks, CA
463	17273670	680000.0	4	3ba	2,096 sqft	Newbury Park, Thousand Oaks, CA
464	17273670	569000.0	3	3ba	1,550 sqft	Newbury Park, Thousand Oaks, CA
465	17273670	830000.0	3	3ba	2,243 sqft	Newbury Park, Thousand Oaks, CA
466	17273670	999900.0	5	4ba	3,780 sqft	Newbury Park, Thousand Oaks, CA

467 rows × 6 columns

```
In [231]: comp_listing_table['bathrooms']= comp_listing_table['bathrooms'].replace(
          'ba', '', regex=True).astype(float)
```

```
c:\users\safar\documents\github\safarie1103\bellevue university\courses\d
sc540\venv\lib\site-packages\ipykernel_launcher.py:1: SettingWithCopyWarn
ing:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  """Entry point for launching an IPython kernel.
```

Out[231]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region
0	17294231	14999000.0	7	13.0	14,073 sqft	Newbury Park, Thousand Oaks, CA
1	17294231	1450000.0	4	3.0	2,568 sqft	Westlake Village, CA
2	17294231	1225000.0	4	3.0	2,745 sqft	Westlake Village, CA
3	17294231	9990000.0	7	10.0	12,656 sqft	Newbury Park, Thousand Oaks, CA
4	17294231	1150000.0	5	4.0	2,393 sqft	Westlake Village, CA
...	...	...	...	...	...	...
462	17273670	897000.0	4	3.0	3,259 sqft	Newbury Park, Thousand Oaks, CA
463	17273670	680000.0	4	3.0	2,096 sqft	Newbury Park, Thousand Oaks, CA
464	17273670	569000.0	3	3.0	1,550 sqft	Newbury Park, Thousand Oaks, CA
465	17273670	830000.0	3	3.0	2,243 sqft	Newbury Park, Thousand Oaks, CA
466	17273670	999900.0	5	4.0	3,780 sqft	Newbury Park, Thousand Oaks, CA

467 rows × 6 columns

In [232]:
```
comp_listing_table['floorSpace'] = comp_listing_table['floorSpace'].repla
ce('sqft', '', regex=True).replace(',','',regex=True).astype(np.int64)
comp_listing_table.columns
```

```
c:\users\safar\documents\github\safarie1103\bellevue university\courses\d
sc540\venv\lib\site-packages\ipykernel_launcher.py:1: SettingWithCopyWarn
ing:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  """Entry point for launching an IPython kernel.
```

Out[232]:
```
Index(['parcelId', 'price', 'bedrooms', 'bathrooms', 'floorSpace', 'regio
n'], dtype='object')
```

In [233]:
```
# Write scraped data to a file for safe keeps and also to avoid rescrapin
g during development
comp_listing_table.to_csv("data/comp_listing_table.csv")
```

In [239]:
```
# Read
comp_listing_table = pd.read_csv("data/comp_listing_table.csv")
```

In [240]:
```
comp_listing_table
```

Out[240]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region
0	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA
1	17294231	1450000	4	3.0	2568	Westlake Village, CA
2	17294231	1225000	4	3.0	2745	Westlake Village, CA
3	17294231	9990000	7	10.0	12656	Newbury Park, Thousand Oaks, CA
4	17294231	1150000	5	4.0	2393	Westlake Village, CA
...	...	...	...	...	...	...
462	17273670	897000	4	3.0	3259	Newbury Park, Thousand Oaks, CA
463	17273670	680000	4	3.0	2096	Newbury Park, Thousand Oaks, CA
464	17273670	569000	3	3.0	1550	Newbury Park, Thousand Oaks, CA
465	17273670	830000	3	3.0	2243	Newbury Park, Thousand Oaks, CA
466	17273670	999900	5	4.0	3780	Newbury Park, Thousand Oaks, CA

467 rows × 6 columns

now that we have our comp table built let's do some comparisons

We'll grab a property from propertiesAndTransactions and query the comp table.

In [56]:

```
# THis table has duplicates and NaNs removed so it is a subset of the propertiesAndTransactions table.
LonLat
```

Out[56]:

	parcelid	latitude	longitude
1761	17299670	34186100	-118767000
107	17296734	34174051	-118757031
1758	17294231	34153879	-118839561
1756	17293716	34152179	-118851454
1427	17292856	34125457	-118891074
...	...	...	...
112	10726315	34184300	-118657000
110	10725532	34196000	-118658000
1767	10722858	34195746	-118624097
108	10722336	34199100	-118633000
1763	10719731	34206094	-118620655

1096 rows × 3 columns

In [57]:

```
propertiesAndTransactions
```

Out[57]:

	parcelid	aircon	num_bathroom	num_bedroom	quality	num_bathroom_cal
0	17054981	NaN	5.0	4.0	NaN	5.(
1	17054981	NaN	5.0	4.0	NaN	5.(
2	17055743	NaN	2.0	3.0	NaN	2.(
3	17055743	NaN	2.0	3.0	NaN	2.(
4	17068109	NaN	1.5	3.0	NaN	1.!
...	...	...	...	...	...	..
2229	11769554	NaN	3.0	4.0	4.0	3.(
2230	11778756	NaN	2.0	7.0	7.0	2.(
2231	11778756	NaN	2.0	7.0	4.0	2.(
2232	11779780	1.0	2.0	2.0	10.0	2.(
2233	11779780	1.0	2.0	2.0	11.0	2.(

2234 rows × 37 columns

In [58]:

```
# Notice the duplicates
selected_parcelid = propertiesAndTransactions['parcelid'] == 17294231
propertiesAndTransactions[selected_parcelid]
```

Out[58]:

	parcelid	aircon	num_bathroom	num_bedroom	quality	num_bathroom_cal
1758	17294231	NaN	2.0	3.0	NaN	2.(
1759	17294231	NaN	2.0	3.0	NaN	2.(

2 rows × 37 columns

In [59]:

```
selected_parcelid = comp_listing_table['parcelId'] == 17294231
comp_listing_table[selected_parcelid]
```

Out[59]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region
0	17294231	14999000.0	7	13.0	14073	Newbury Park, Thousand Oaks, CA
1	17294231	1450000.0	4	3.0	2568	Westlake Village, CA
2	17294231	1225000.0	4	3.0	2745	Westlake Village, CA
3	17294231	9990000.0	7	10.0	12656	Newbury Park, Thousand Oaks, CA
4	17294231	1150000.0	5	4.0	2393	Westlake Village, CA
5	17294231	525000.0	2	3.0	1440	Westlake Village, CA
6	17294231	1499000.0	5	5.0	3804	Westlake Village, CA
7	17294231	1099000.0	4	3.0	2300	Westlake Village, CA
8	17294231	919000.0	4	2.0	1838	Westlake Village, CA
9	17294231	3195000.0	3	3.0	2543	Westlake Village, CA
10	17294231	1875000.0	5	5.0	4431	Westlake Village, CA
11	17294231	9900000.0	5	7.0	8095	Lake Sherwood, CA
12	17294231	1250000.0	4	3.0	3012	Westlake Village, CA
13	17294231	1799999.0	4	4.0	2106	Westlake Village, CA
14	17294231	640000.0	2	2.0	1231	Westlake Village, CA
15	17294231	1080000.0	4	2.0	2371	Westlake Village, CA
16	17294231	1289000.0	3	3.0	2222	Lake Sherwood, CA
17	17294231	3450000.0	5	6.0	5954	Thousand Oaks, CA
18	17294231	1049000.0	4	3.0	2538	Westlake Village, CA
19	17294231	5495000.0	7	9.0	9304	Thousand Oaks, CA
20	17294231	2995000.0	5	6.0	5421	Westlake Village, CA
21	17294231	1499000.0	4	3.0	2920	Thousand Oaks, CA
22	17294231	1449000.0	4	4.0	3013	Lake Sherwood, CA
23	17294231	765000.0	2	2.0	1508	Westlake Village, CA
24	17294231	1599000.0	3	3.0	2282	Westlake Village, CA
25	17294231	2399000.0	5	4.0	4724	Westlake Village, CA
26	17294231	2975000.0	4	3.0	4075	Westlake Village, CA
27	17294231	988000.0	4	3.0	2412	Westlake Village, CA
28	17294231	4750000.0	6	6.0	7470	Thousand Oaks, CA
29	17294231	3950000.0	5	5.0	5466	Thousand Oaks, CA

data from API

Description

Googlemap API and matplotlib or equivalant will be used to locate properties by zipcode and display them on the map of the Unites States. We will convert 'longitude' and 'latitude' columns in properties dataset to zip code and use the zipcode in the API call.We will show the density of homes sold in various regions in the dataset. We will also show the properties we extracted using

In [60]:
```
propertiesAndTransactions
```

Out[60]:

	parcelid	aircon	num_bathroom	num_bedroom	quality	num_bathroom_cal
0	17054981	NaN	5.0	4.0	NaN	5.
1	17054981	NaN	5.0	4.0	NaN	5.
2	17055743	NaN	2.0	3.0	NaN	2.
3	17055743	NaN	2.0	3.0	NaN	2.
4	17068109	NaN	1.5	3.0	NaN	1.
...	...	...	...	...	...	..
2229	11769554	NaN	3.0	4.0	4.0	3.
2230	11778756	NaN	2.0	7.0	7.0	2.
2231	11778756	NaN	2.0	7.0	4.0	2.
2232	11779780	1.0	2.0	2.0	10.0	2.
2233	11779780	1.0	2.0	2.0	11.0	2.

2234 rows × 37 columns

In [61]:
```
# Notice the duplicates
selected_parcelid = propertiesAndTransactions['parcelid'] == 17294231
propertiesAndTransactions[selected_parcelid]
```

Out[61]:

	parcelid	aircon	num_bathroom	num_bedroom	quality	num_bathroom_cal
1758	17294231	NaN	2.0	3.0	NaN	2.
1759	17294231	NaN	2.0	3.0	NaN	2.

2 rows × 37 columns

```
In [62]: selected_parcelid = comp_listing_table['parcelId'] == 17294231
         comp_listing_table[selected_parcelid]
```

Out[62]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region
0	17294231	14999000.0	7	13.0	14073	Newbury Park, Thousand Oaks, CA
1	17294231	1450000.0	4	3.0	2568	Westlake Village, CA
2	17294231	1225000.0	4	3.0	2745	Westlake Village, CA
3	17294231	9990000.0	7	10.0	12656	Newbury Park, Thousand Oaks, CA
4	17294231	1150000.0	5	4.0	2393	Westlake Village, CA
5	17294231	525000.0	2	3.0	1440	Westlake Village, CA
6	17294231	1499000.0	5	5.0	3804	Westlake Village, CA
7	17294231	1099000.0	4	3.0	2300	Westlake Village, CA
8	17294231	919000.0	4	2.0	1838	Westlake Village, CA
9	17294231	3195000.0	3	3.0	2543	Westlake Village, CA
10	17294231	1875000.0	5	5.0	4431	Westlake Village, CA
11	17294231	9900000.0	5	7.0	8095	Lake Sherwood, CA
12	17294231	1250000.0	4	3.0	3012	Westlake Village, CA
13	17294231	1799999.0	4	4.0	2106	Westlake Village, CA
14	17294231	640000.0	2	2.0	1231	Westlake Village, CA
15	17294231	1080000.0	4	2.0	2371	Westlake Village, CA
16	17294231	1289000.0	3	3.0	2222	Lake Sherwood, CA
17	17294231	3450000.0	5	6.0	5954	Thousand Oaks, CA
18	17294231	1049000.0	4	3.0	2538	Westlake Village, CA
19	17294231	5495000.0	7	9.0	9304	Thousand Oaks, CA
20	17294231	2995000.0	5	6.0	5421	Westlake Village, CA
21	17294231	1499000.0	4	3.0	2920	Thousand Oaks, CA
22	17294231	1449000.0	4	4.0	3013	Lake Sherwood, CA
23	17294231	765000.0	2	2.0	1508	Westlake Village, CA
24	17294231	1599000.0	3	3.0	2282	Westlake Village, CA
25	17294231	2399000.0	5	4.0	4724	Westlake Village, CA
26	17294231	2975000.0	4	3.0	4075	Westlake Village, CA
27	17294231	988000.0	4	3.0	2412	Westlake Village, CA
28	17294231	4750000.0	6	6.0	7470	Thousand Oaks, CA
29	17294231	3950000.0	5	5.0	5466	Thousand Oaks, CA

Milestone 4. Data from API

Description

Googlemaps API is used to get additional information for parcelIDs in LonLat table built in Milestone 3. We will get the geometric coordinates for a given parcel, latitude and longitude of that parcel. Googlemaps returns various corrdinates sorrounding the given coordinates such as nw/sw

In [182]:

```
# This is a sample code and does not pertain to this project. We will try
to implement a function s
import googlemaps
from datetime import datetime

with open('../APIkeys/APIkeys.json') as f:
    keys = json.load(f)
    key = keys['googlemaps']['key']

gmaps = googlemaps.Client(key=key)
```

Some testing and exploration of the interface

In [183]:

```
# Geocoding an address
geocode_result = gmaps.geocode('1600 Amphitheatre Parkway, Mountain View,
CA')

print(geocode_result[0]['geometry'])
```

```
{'location': {'lat': 37.4223106, 'lng': -122.0846328}, 'location_type':
'ROOFTOP', 'viewport': {'northeast': {'lat': 37.42365958029151, 'lng': -1
22.0832838197085}, 'southwest': {'lat': 37.42096161970851, 'lng': -122.08
59817802915}}}
```

In [184]:

```
print(geocode_result[0]['geometry']['viewport']['northeast']['lat'])
```

```
37.42365958029151
```

In [185]:

```
# Get a sample
reverse_geocode_result = gmaps.reverse_geocode((40.714224, -73.961452))
```

In [186]:

```
# print result
print(reverse_geocode_result)
```

```
[{'access_points': [], 'address_components': [{'long_name': '279', 'short
_name': '279', 'types': ['street_number']}, {'long_name': 'Bedford Avenu
e', 'short_name': 'Bedford Ave', 'types': ['route']}, {'long_name': 'Will
iamsburg', 'short_name': 'Williamsburg', 'types': ['neighborhood', 'polit
ical']}, {'long_name': 'Brooklyn', 'short_name': 'Brooklyn', 'types': ['p
olitical', 'sublocality', 'sublocality_level_1']}, {'long_name': 'Kings C
ounty', 'short_name': 'Kings County', 'types': ['administrative_area_leve
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y_level_1']}, {'long_name': 'Kings County', 'short_name': 'Kings County',
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litical']}, {'long_name': 'United States', 'short_name': 'US', 'types':
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rea_level_2', 'political']}, {'long_name': 'New York', 'short_name': 'N
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ounty', 'short_name': 'Kings County', 'types': ['administrative_area_leve
l_2', 'political']}, {'long_name': 'New York', 'short_name': 'NY', 'type
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```

```
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ative_area_level_1', 'political']}, {'long_name': 'United States', 'short
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ame': 'NY', 'types': ['administrative_area_level_1', 'political']}, {'lon
g_name': 'United States', 'short_name': 'US', 'types': ['country', 'polit
ical']}], 'formatted_address': 'Long Island, New York, USA', 'geometry':
{'bounds': {'northeast': {'lat': 41.1612401, 'lng': -71.85620109999999},
'southwest': {'lat': 40.5429789, 'lng': -74.0419497}}, 'location': {'la
t': 40.789142, 'lng': -73.13496099999999}, 'location_type': 'APPROXIMAT
E', 'viewport': {'northeast': {'lat': 41.1612401, 'lng': -71.856201099999
99}, 'southwest': {'lat': 40.5429789, 'lng': -74.0419497}}}, 'place_id':
'ChIJy6Xu4VRE6IkRGA2UhmH59x0', 'types': ['establishment', 'natural_featur
e']}, {'access_points': [], 'address_components': [{'long_name': 'New Yor
k', 'short_name': 'NY', 'types': ['administrative_area_level_1', 'politic
al']}, {'long_name': 'United States', 'short_name': 'US', 'types': ['coun
try', 'political']}], 'formatted_address': 'New York, USA', 'geometry':
```

```
{'bounds': {'northeast': {'lat': 45.015861, 'lng': -71.777491}, 'southwes
t': {'lat': 40.4773991, 'lng': -79.7625901}}, 'location': {'lat': 43.2994
285, 'lng': -74.21793260000001}, 'location_type': 'APPROXIMATE', 'viewpor
t': {'northeast': {'lat': 45.015861, 'lng': -71.777491}, 'southwest': {'l
at': 40.4773991, 'lng': -79.7625901}}}, 'place_id': 'ChIJqaUj8fBLzEwRZ5UY
3sHGz90', 'types': ['administrative_area_level_1', 'political']}, {'acces
s_points': [], 'address_components': [{'long_name': 'United States', 'sho
rt_name': 'US', 'types': ['country', 'political']}], 'formatted_address':
'United States', 'geometry': {'bounds': {'northeast': {'lat': 71.5388001,
'lng': -66.885417}, 'southwest': {'lat': 18.7763, 'lng': 170.5957}}, 'loc
ation': {'lat': 37.09024, 'lng': -95.712891}, 'location_type': 'APPROXIMA
TE', 'viewport': {'northeast': {'lat': 71.5388001, 'lng': -66.885417}, 's
outhwest': {'lat': 18.7763, 'lng': 170.5957}}}, 'place_id': 'ChIJCzYy5IS1
6lQRQrfeQ5K5Oxw', 'types': ['country', 'political']}]
```

```
In [187]: # Explore reply
          print(reverse_geocode_result[0]['geometry'])
```

```
{'location': {'lat': 40.71423350000001, 'lng': -73.9613686}, 'location_ty
pe': 'ROOFTOP', 'viewport': {'northeast': {'lat': 40.71558248029151, 'ln
g': -73.9600196197085}, 'southwest': {'lat': 40.71288451970851, 'lng': -7
3.96271758029151}}}
```

```
In [69]: # We will parse the geometry part
```

Now, we will implement on 20 records of the lonlat table. Notice that googlemap return error code 400 for invalid lon/lat values. Care has been taken to avoid recording NaN's in the table in such circumstance.

In [188]:

```
Geographic_Location_Coordinates = pd.DataFrame(columns={'parcelID','lat',
'lng','loc_type','view_NW_lat','view_NW_lng','view_NW_lat','view_NW_lng'
})

dfs = []
for index, row in LonLat[:20].iterrows():
    parcelId = row['parcelid']
    latitude = row['latitude']/10**6
    longitude = row['longitude']/10**6
    #print(parcelId,latitude,longitude)
    try:
        reverse_geocode_result = gmaps.reverse_geocode((latitude, longitu
de))
        #print(reverse_geocode_result[0]['geometry'])
        for item in reverse_geocode_result:
            lat = item['geometry']['location']['lat']
            lng = item['geometry']['location']['lng']
            loc_type = item['geometry']['location_type']
            view_NW_lat = item['geometry']['viewport']['northeast']['lat'
]
            view_NW_lng = item['geometry']['viewport']['northeast']['lng'
]
            view_NW_lat = item['geometry']['viewport']['southwest']['lat'
]
            view_NW_lng = item['geometry']['viewport']['southwest']['lng'
]
            dfs.append(
                {
                    'parcelID' : parcelId,
                    'lat': lat ,
                    'lng' : lng,
                    'loc_type': loc_type,
                    'view_NW_lat' : view_NW_lat,
                    'view_NW_lng' : view_NW_lng,
                    'view_NW_lat' : view_NW_lat,
                    'view_NW_lng' : view_NW_lng
                })
    except:
        continue


Geographic_Location_Coordinates = pd.DataFrame(dfs)
print(Geographic_Location_Coordinates)
```

```
     parcelID        lat         lng             loc_type  view_NW_lat  \
0    17299670  34.186396 -118.766827              ROOFTOP    34.185047   
1    17299670  34.186270 -118.766494  RANGE_INTERPOLATED    34.184921   
2    17299670  34.186411 -118.766587    GEOMETRIC_CENTER    34.185062   
3    17299670  34.188033 -118.760611          APPROXIMATE    34.167911   
4    17299670  34.370488 -119.139064          APPROXIMATE    33.163493   
..        ...        ...         ...                  ...          ...   
160  17273670  34.183616 -118.943432          APPROXIMATE    34.178342   
161  17273670  34.181067 -118.947042          APPROXIMATE    34.135933   
162  17273670  34.370488 -119.139064          APPROXIMATE    33.163493   
163  17273670  36.778261 -119.417932          APPROXIMATE    32.528832   
164  17273670  37.090240  -95.712891          APPROXIMATE    18.776300   

     view_NW_lng  
0    -118.768176  
1    -118.767843  
2    -118.767936  
3    -118.789393  
4    -119.636302  
..           ...  
160  -118.950291  
161  -119.007712  
162  -119.636302  
163  -124.482003  
164   170.595700  

[165 rows x 6 columns]
```

Milestone Conclusion

We now have three tables from their respective sources. All three tables are linked by parcelID. The relationship betwen propertiesandtransactions table, comp_listing_table, and the new table Geographic_Location_Coordinates is one-to-many.

Milestone 5. Merging the data and storing in a database/visualizing data

Description

We will store tables fron previous milestones in sqlite and make queries from them using parcelID as index. We will also provide visulization of the stored data.

In [352]:

```
import sqlite3
```

In [353]:

```
conn.close()
sqlite_file = 'Data/DSC540_EdrisSafari_FinalProject.sqlite'
conn = sqlite3.connect(sqlite_file)
```

In [355]:

```
propertiesAndTransactions[['latitude','longitude']].head()
```

Out[355]:

	latitude	longitude
0	34449407	-119254052
1	34449407	-119254052
2	34454169	-119237898
3	34454169	-119237898
4	34365693	-119448392

In [356]:

```
propertiesAndTransactions['latitude'] = propertiesAndTransactions['latitude']/10**6
propertiesAndTransactions['longitude'] = propertiesAndTransactions['longitude']/10**6
propertiesAndTransactions['abs_logerror'] = propertiesAndTransactions['logerror'].abs()

propertiesAndTransactions.to_sql('propertiesAndTransactions', conn, if_exists='replace', index=False)

propertiesAndTransactions = pd.read_sql_query("SELECT * from propertiesAndTransactions", conn)

propertiesAndTransactions[['latitude','longitude','abs_logerror']].head()
```

Out[356]:

	latitude	longitude	abs_logerror
0	34.449407	-119.254052	0.013099
1	34.449407	-119.254052	0.013099
2	34.454169	-119.237898	0.073985
3	34.454169	-119.237898	0.073985
4	34.365693	-119.448392	0.071886

In [357]:

```
comp_listing_table.to_sql('comp_listing_table', conn, if_exists='replace', index=False)
comp_listing_table.head()
```

Out[357]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region
0	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA
1	17294231	1450000	4	3.0	2568	Westlake Village, CA
2	17294231	1225000	4	3.0	2745	Westlake Village, CA
3	17294231	9990000	7	10.0	12656	Newbury Park, Thousand Oaks, CA
4	17294231	1150000	5	4.0	2393	Westlake Village, CA

In [358]:

```
Geographic_Location_Coordinates.to_sql('Geographic_Location_Coordinates',
conn, if_exists='replace', index=False)
Geographic_Location_Coordinates.head()
```

Out[358]:

	parcelID	lat	lng	loc_type	view_NW_lat	view_NW_ln
0	17299670	34.186396	-118.766827	ROOFTOP	34.185047	-118.76817
1	17299670	34.186270	-118.766494	RANGE_INTERPOLATED	34.184921	-118.76784
2	17299670	34.186411	-118.766587	GEOMETRIC_CENTER	34.185062	-118.76793
3	17299670	34.188033	-118.760611	APPROXIMATE	34.167911	-118.78939
4	17299670	34.370488	-119.139064	APPROXIMATE	33.163493	-119.63630

In [359]:

```
!jupyter nbextension enable --py --sys-prefix widgetsnbextension
```

```
Enabling notebook extension jupyter-js-widgets/extension...
      - Validating: ok
```

In [360]:

```
!jupyter nbextension enable --py --sys-prefix gmaps
```

```
Enabling notebook extension jupyter-gmaps/extension...
      - Validating: ok
```

In [361]:

```
import gmaps

with open('../APIkeys/APIkeys.json') as f:
    keys = json.load(f)
    key = keys['googlemaps']['key']

gmaps.configure(api_key=key) # Fill in with your API key
```

Heatmap shows absolute log error in regions in the properties and transactions table

In [363]:

```
locations = propertiesAndTransactions[['latitude', 'longitude']]
weights = propertiesAndTransactions['abs_logerror']
fig = gmaps.figure(map_type="HYBRID")
fig.add_layer(gmaps.heatmap_layer(locations, weights=weights))
fig
```

In [372]:

```
ParcelID_17294231 =  pd.read_sql_query("SELECT * from propertiesandtransa
ctions where parcelID = '17294231' LIMIT 1", conn)
ParcelID_17294231[['parcelid','latitude','longitude','abs_logerror']]
```

Out[372]:

	parcelid	latitude	longitude	abs_logerror
0	17294231	34.153879	-118.839561	0.013219

In [373]:

```
lat = ParcelID_17294231['latitude'][0]
lon = ParcelID_17294231['longitude'][0]
cen = (pd.to_numeric(lat),pd.to_numeric(lon))
print(cen)
```

```
(34.153879, -118.839561)
```

In [374]:

```
gmaps.figure(center=cen,zoom_level=18)
```

In [377]:

```
ParcelID_17294231 =  pd.read_sql_query("SELECT * from Geographic_Location
_Coordinates where Geographic_Location_Coordinates.parcelid = '17294231'"
, conn)
ParcelID_17294231.head()
```

Out[377]:

	parcelID	lat	lng	loc_type	view_NW_lat	view_NW_ln
0	17294231	34.154077	-118.839494	ROOFTOP	34.152716	-118.84084
1	17294231	34.154298	-118.839583	ROOFTOP	34.152949	-118.84093
2	17294231	34.153681	-118.839965	RANGE_INTERPOLATED	34.152332	-118.84131
3	17294231	34.153147	-118.840481	GEOMETRIC_CENTER	34.151801	-118.84183
4	17294231	34.138463	-118.894631	APPROXIMATE	34.104268	-118.99458

In [378]:

```
ParcelID_17294231 =  pd.read_sql_query("SELECT * from comp_listing_table
 where comp_listing_table.parcelid = '17294231'", conn)
ParcelID_17294231.head()
```

Out[378]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region
0	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA
1	17294231	1450000	4	3.0	2568	Westlake Village, CA
2	17294231	1225000	4	3.0	2745	Westlake Village, CA
3	17294231	9990000	7	10.0	12656	Newbury Park, Thousand Oaks, CA
4	17294231	1150000	5	4.0	2393	Westlake Village, CA

In [384]:

```
comp_and_geo_table =  pd.read_sql_query("SELECT * from comp_listing_table,Geographic_Location_Coordinates where Geographic_Location_Coordinates.parcelID = comp_listing_table.parcelid", conn)
comp_and_geo_table.head()
```

Out[384]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region	parcelID	
0	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA	17294231	34.
1	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA	17294231	34.
2	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA	17294231	34.
3	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA	17294231	34.
4	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA	17294231	34.

Heatmap of home prices

In [385]:

```
locations = comp_and_geo_table[['lat', 'lng']]
weights = comp_and_geo_table['price']
fig = gmaps.figure()
fig.add_layer(gmaps.heatmap_layer(locations, weights=weights))
fig
```

this table shows comparable prices, number of bed and bathrooms., etc. while properties and transactions table does not have a sale or sold price(only estimate error), we can decipher from tax rate.

In [386]:

```
comp_and_propandtrans_table =  pd.read_sql_query("SELECT * from comp_list
ing_table,propertiesAndTransactions where propertiesAndTransactions.parce
lID = comp_listing_table.parcelid", conn)
comp_and_propandtrans_table.head()
```

Out[386]:

	parcelId	price	bedrooms	bathrooms	floorSpace	region	parcelid	airc
0	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA	17294231	N
1	17294231	14999000	7	13.0	14073	Newbury Park, Thousand Oaks, CA	17294231	N
2	17294231	1450000	4	3.0	2568	Westlake Village, CA	17294231	N
3	17294231	1450000	4	3.0	2568	Westlake Village, CA	17294231	N
4	17294231	1225000	4	3.0	2745	Westlake Village, CA	17294231	N

5 rows × 44 columns

In [391]:

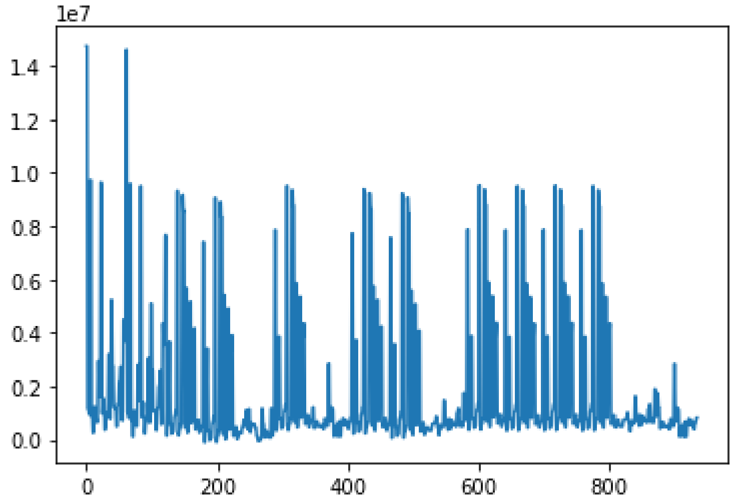
```
comp_and_propandtrans_table['comp_diff'] =  comp_and_propandtrans_table[
'price'] - comp_and_propandtrans_table['tax_building']
print(comp_and_propandtrans_table[['price','tax_building','comp_diff']])
```

```
        price  tax_building   comp_diff
0    14999000      265152.0  14733848.0
1    14999000      261170.0  14737830.0
2     1450000      265152.0   1184848.0
3     1450000      261170.0   1188830.0
4     1225000      265152.0    959848.0
..        ...           ...         ...
929    569000      170000.0    399000.0
930    830000      172592.0    657408.0
931    830000      170000.0    660000.0
932    999900      172592.0    827308.0
933    999900      170000.0    829900.0

[934 rows x 3 columns]
```

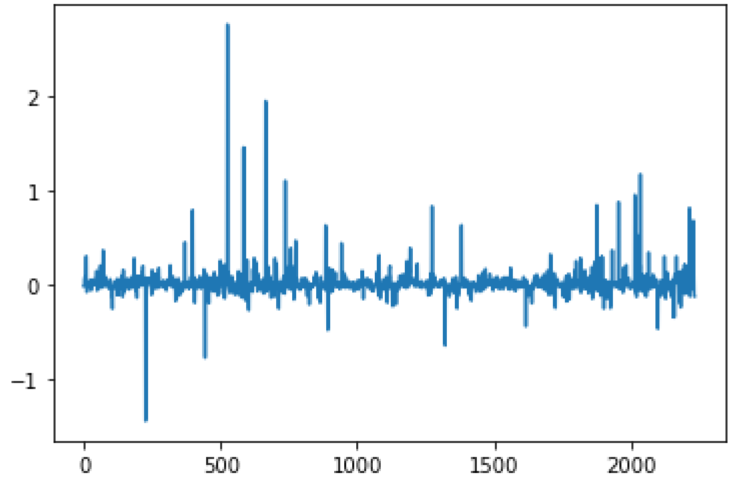
In [392]:
```
# Scatter plot of comp_diff
plt.plot(comp_and_propandtrans_table.comp_diff)
```

Out[392]:
```
[<matplotlib.lines.Line2D at 0x182411f0>]
```



In [393]:
```
# Scatter plot of logerror
plt.plot(propertiesAndTransactions.logerror)
```

Out[393]:
```
[<matplotlib.lines.Line2D at 0x10c90630>]
```

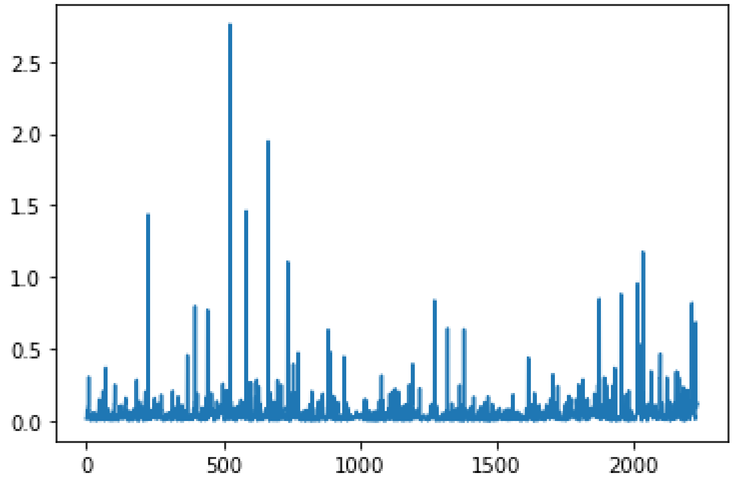


In [394]:

```
# Scatter plot of abs_logerror
plt.plot(propertiesAndTransactions.abs_logerror)
```

Out[394]:

```
[<matplotlib.lines.Line2D at 0x16ebbed0>]
```

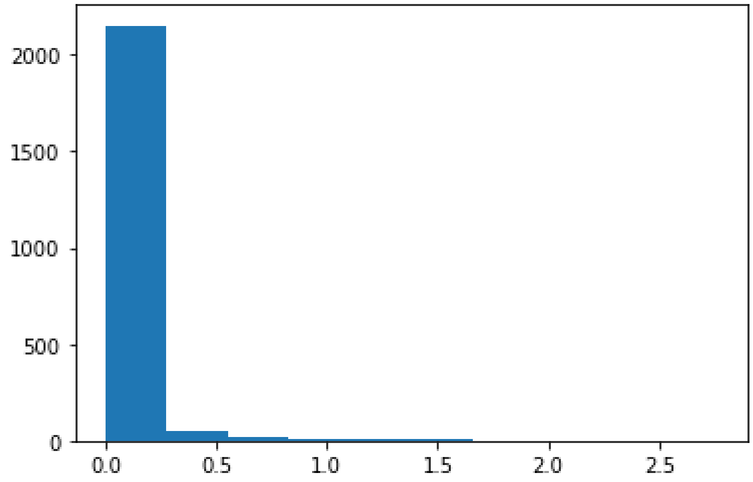


In [399]:

```
plt.hist(propertiesAndTransactions.abs_logerror)
```

Out[399]:

```
(array([2.152e+03, 4.800e+01, 1.400e+01, 8.000e+00, 4.000e+00, 4.000e+00,
        0.000e+00, 2.000e+00, 0.000e+00, 2.000e+00]),
 array([0.    , 0.2758, 0.5516, 0.8274, 1.1032, 1.379 , 1.6548, 1.9306,
        2.2064, 2.4822, 2.758 ]),
 <a list of 10 Patch objects>)
```

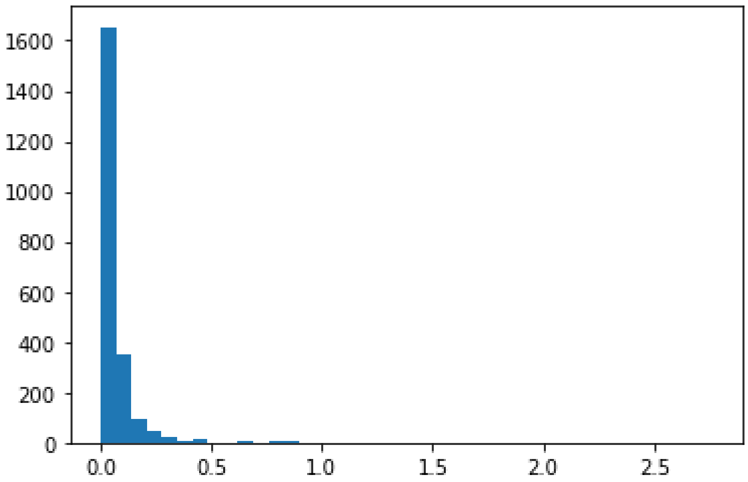


In [400]:

```
plt.hist(propertiesAndTransactions.abs_logerror,bins=40)
```

Out[400]:

```
(array([1654.,  352.,   98.,   48.,   24.,   10.,   12.,    2.,    0.,
           8.,    0.,    6.,    6.,    2.,    0.,    0.,    2.,    2.,
           0.,    0.,    2.,    2.,    0.,    0.,    0.,    0.,    0.,
           0.,    2.,    0.,    0.,    0.,    0.,    0.,    0.,    0.,
           0.,    0.,    0.,    2.]),
 array([0.     , 0.06895, 0.1379 , 0.20685, 0.2758 , 0.34475, 0.4137 ,
        0.48265, 0.5516 , 0.62055, 0.6895 , 0.75845, 0.8274 , 0.89635,
        0.9653 , 1.03425, 1.1032 , 1.17215, 1.2411 , 1.31005, 1.379  ,
        1.44795, 1.5169 , 1.58585, 1.6548 , 1.72375, 1.7927 , 1.86165,
        1.9306 , 1.99955, 2.0685 , 2.13745, 2.2064 , 2.27535, 2.3443 ,
        2.41325, 2.4822 , 2.55115, 2.6201 , 2.68905, 2.758  ]),
 <a list of 40 Patch objects>)
```



In [501]:

```
propertiesAndTransactions['year_month'] = pd.to_datetime(propertiesAndTra
nsactions.transactiondate)
propertiesAndTransactions.to_sql('propertiesAndTransactions', conn, if_ex
ists='replace', index=False)
```

In [502]:
```
prop_and_trans_groupby_month = propertiesAndTransactions[['year_month','abs_logerror']].groupby(['year_month']).mean()

prop_and_trans_groupby_month
```

Out[502]:

	abs_logerror
year_month	
2016-01-04	0.011100
2016-01-05	0.071750
2016-01-06	0.023700
2016-01-07	0.065100
2016-01-08	0.064150
...	...
2017-09-13	0.065672
2017-09-14	0.012129
2017-09-15	0.049597
2017-09-18	0.069977
2017-09-19	0.016698

378 rows × 1 columns

In [503]:
```
print(prop_and_trans_groupby_month.groupby(pd.Grouper(freq='D')).mean())
```

```
            abs_logerror
year_month              
2016-01-04      0.011100
2016-01-05      0.071750
2016-01-06      0.023700
2016-01-07      0.065100
2016-01-08      0.064150
...                  ...
2017-09-15      0.049597
2017-09-16           NaN
2017-09-17           NaN
2017-09-18      0.069977
2017-09-19      0.016698

[625 rows x 1 columns]
```

In [504]:
```
prop_and_trans_groupby_month = prop_and_trans_groupby_month.reset_index()
```

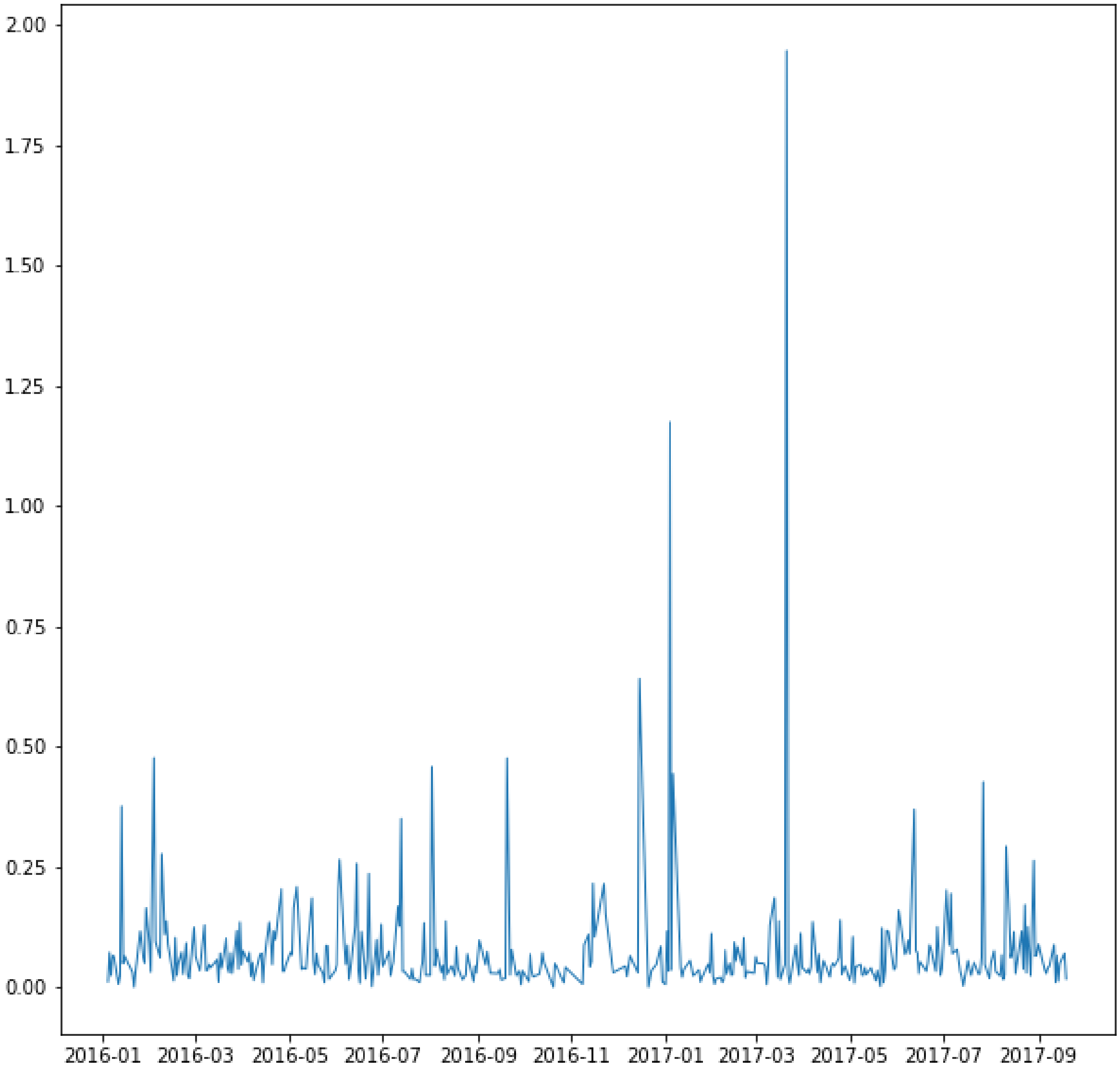
This graph shows the mean error between estimated value and actual sale value per month. It shows spikes in January and March of 2017.

In [505]:

```
plt.figure(figsize=(10, 10))
plt.plot(prop_and_trans_groupby_month['year_month'],prop_and_trans_groupby_month.abs_logerror,linewidth=1.0)
```

Out[505]:

```
[<matplotlib.lines.Line2D at 0x1de45e90>]
```



In [278]:

```
conn.close()
```

Milestone Conclusion

In this milestone, we stored the three datasets in the sqlite database. Using some queries and also usingdataframe's groupby function, we were able to produce some graphs and tables. We also used gmaps package from google to locate some properties on the map using longitude and latitude. The heat maps showed the intensoty of absoloute log erros and also the price.

Project Conclusion

This project involved collecting data related to the properties that were listed and sold in souther california in 2016 and 2017. We also has a corresponding dataset that stored the sale transaction date and error between estimated price and actual sale proce. The intent is to miimize this error. We took a sampling of these two data sets and used longitude and latitude of the properties to find compariable properities in the same zip code. We achived this by web scraping the web site https://www.trulia.com/ (https://www.trulia.com/). We also used obtained Geographic Location Coordinates of 20 properties(due to response time limitation cosnstraint) using google maps API. Given the longitue and latitude, this API privided a host of information, but we decided to collect location type along with longitude and latitude and the view from the property(not so useful!).