## Part One and Two-ML

## October 13, 2018

```
In [58]: import pandas as pd
         import numpy as np
         import seaborn as sns
         import matplotlib.pyplot as plt
In [59]: weight1 = pd.read_csv('weight_0.0001.csv')
         weight2 = pd.read_csv('weight_1e-05.csv')
         weight3 = pd.read_csv('weight_1e-06.csv')
         weight4 = pd.read_csv('weight_1e-07.csv')
In [60]: weight1
Out [60]:
               weight feature
          -6.314734
         1 -0.016919
                             2
         2
           7.585033
                             3
         3 -0.756257
                             4
         4 4.918300
                             5
         5
             2.960866
                             6
            7.681923
                             7
            8.606310
                             8
         8
             2.481608
                             9
         9
            4.239114
                            10
         10 0.166684
                            11
         11 -3.195887
                            12
In [61]: diff_wt ={
             0.0001:weight1.weight,
             0.00001:weight2.weight,
             0.000001:weight3.weight,
             0.0000001:weight4.weight
         }
In [62]: features = pd.read_csv('features_used.csv')
         x_names = np.array(features['names'])
In [63]: x_names
```

```
Out[63]: array(['bedrooms', 'bathrooms', 'sqft_living', 'floors', 'waterfront',
                 'view', 'grade', 'sqft_above', 'sqft_basement', 'lat',
                 'sqft_living15', 'dummy'], dtype=object)
In [64]: main_df_val = pd.read_csv('PA1_dev.csv.xls')
In [65]: df val = main df val[x names]
         df_val.head()
Out [65]:
            bedrooms
                      bathrooms
                                  sqft_living floors
                                                        waterfront
                                                                     view
                   3
                            2.25
                                          2570
                                                   2.0
                                                                        0
                                                                               7
                                                                  0
                   3
                            2.00
                                          1680
                                                   1.0
                                                                  0
                                                                        0
                                                                               8
         1
                   3
         2
                            2.50
                                         3560
                                                   1.0
                                                                  0
                                                                        0
                                                                               8
         3
                   2
                                                   1.0
                                                                  0
                                                                        0
                                                                               7
                            1.00
                                         1160
         4
                   3
                                                   2.0
                            2.00
                                         1890
                                                                        0
                                                                               7
            sqft_above sqft_basement
                                             lat
                                                  sqft_living15
         0
                  2170
                                   400
                                        47.7210
                                                           1690
                                                                      1
         1
                  1680
                                     0 47.6168
                                                           1800
                                                                      1
         2
                  1860
                                  1700 47.6007
                                                           2210
                                                                      1
         3
                   860
                                   300 47.6900
                                                           1330
                                                                      1
                                                                      1
                  1890
                                     0 47.7277
                                                           1890
In [66]: df_val = df_val.drop(['dummy'],1)
In [67]: for feature in df_val.columns:
             mn = df_val[feature].min()
             rnge = df_val[feature].max() - df_val[feature].min()
             df_val[feature] = df_val[feature].apply(lambda x : (x-mn)/rnge)
In [68]: df_val['dummy']=main_df_val['dummy']
In [69]: df_val.head()
Out [69]:
            bedrooms bathrooms
                                  sqft_living floors waterfront view
                                                                           grade
                                     0.165779
                                                               0.0
               0.250
                        0.206897
                                                   0.4
                                                                      0.0
                                                                             0.4
         0
         1
               0.250
                        0.172414
                                     0.098099
                                                   0.0
                                                               0.0
                                                                      0.0
                                                                             0.5
         2
                                                               0.0
               0.250
                       0.241379
                                     0.241065
                                                   0.0
                                                                      0.0
                                                                             0.5
         3
               0.125
                        0.034483
                                     0.058555
                                                   0.0
                                                               0.0
                                                                      0.0
                                                                             0.4
               0.250
                        0.172414
                                     0.114068
                                                   0.4
                                                               0.0
                                                                      0.0
                                                                             0.4
            sqft_above sqft_basement
                                                   sqft_living15 dummy
                                              lat
              0.197339
                              0.082988
                                        0.908027
                                                        0.184116
         0
                                                                       1
         1
              0.143016
                              0.000000
                                        0.738707
                                                        0.203971
                                                                       1
         2
              0.162971
                              0.352697
                                        0.712545
                                                        0.277978
         3
              0.052106
                              0.062241
                                        0.857654
                                                        0.119134
                                                                       1
              0.166297
                              0.000000 0.918915
                                                        0.220217
                                                                       1
In [70]: def calculate_sse(weight):
             print(df_val.shape, weight.shape)
```

```
p = df_val.dot(np.array(weight))
             y = main_df_val['price']
             diff = p - y
             sse = (np.transpose(diff)).dot(diff)
             return sse
In [71]: learning_rate = [0.0001,0.00001,0.000001,0.000001]
        for lr in learning rate:
             sse = calculate_sse(diff_wt[lr])
             print("lr = ",lr," sse = ",sse)
(5597, 12) (12,)
lr = 0.0001 sse = 42092.233530221536
(5597, 12) (12,)
lr = 1e-05 sse = 42392.654374075384
(5597, 12) (12,)
lr = 1e-06 sse = 31865.8394104541
(5597, 12) (12,)
lr = 1e-07 sse = 26098.742670731357
In [53]: ############### thus choosing learn rate = 10**(-7)
In [97]: main_df_train = pd.read_csv('PA1_train.csv.xls')
        df_train = main_df_train.drop(['id','dummy'],1)
In [98]: df_train.columns
Out[98]: Index(['date', 'bedrooms', 'bathrooms', 'sqft_living', 'sqft_lot', 'floors',
                'waterfront', 'view', 'condition', 'grade', 'sqft_above',
                'sqft_basement', 'yr_built', 'yr_renovated', 'zipcode', 'lat', 'long',
                'sqft_living15', 'sqft_lot15', 'price'],
               dtype='object')
In [99]: df_train['date'] = pd.to_datetime(main_df_train['date'])
        df train['month'] = df train['date'].apply(lambda x:x.month)
        df_train['year'] = df_train['date'].apply(lambda x:x.year)
        df_train['day'] = df_train['date'].apply(lambda x:x.day)
In [100]: df_train.columns.shape
Out[100]: (23,)
In [101]: df_train = df_train.drop(['date'],1)
          df_train.shape
Out[101]: (10000, 22)
In [102]: df_train.columns
```

```
Out[102]: Index(['bedrooms', 'bathrooms', 'sqft_living', 'sqft_lot', 'floors',
                 'waterfront', 'view', 'condition', 'grade', 'sqft_above',
                 'sqft_basement', 'yr_built', 'yr_renovated', 'zipcode', 'lat', 'long',
                 'sqft_living15', 'sqft_lot15', 'price', 'month', 'year', 'day'],
                dtype='object')
In [103]: y_train = df_train['price']
          df_train = df_train.drop(['price'],1)
In [104]: for feature in df_train.columns:
              mn = df_train[feature].min()
              rnge = df_train[feature].max() - df_train[feature].min()
              df_train[feature] = df_train[feature].apply(lambda x : (x-mn)/rnge)
In [105]: df_train.head()
Out[105]:
             bedrooms
                      bathrooms sqft_living sqft_lot
                                                         floors
                                                                 waterfront view \
          0
              0.09375
                        0.275862
                                     0.246849
                                               0.005715
                                                            0.4
                                                                        0.0
                                                                              0.0
              0.03125
                                                            0.0
                                                                        0.0
                                                                              0.0
          1
                        0.275862
                                     0.234244
                                               0.002682
             0.12500
                        0.172414
                                     0.134454
                                               0.001471
                                                            0.2
                                                                        0.0
                                                                              0.0
          3
              0.09375
                        0.172414
                                     0.141807
                                               0.005469
                                                            0.0
                                                                        0.0
                                                                              0.0
             0.15625
                        0.241379
                                     0.363445
                                                            0.0
                                                                        0.0
                                                                              0.5
                                               0.006426
                                                        yr_built yr_renovated \
             condition
                           grade sqft_above
                                                . . .
          0
                  0.50 0.555556
                                    0.276796
                                                . . .
                                                        0.773913
                                                                           0.0
          1
                  1.00 0.444444
                                    0.109541
                                                        0.226087
                                                                           0.0
                                                . . .
          2
                                                                           0.0
                  0.50 0.444444
                                    0.150766
                                                        0.017391
          3
                  0.75 0.444444
                                    0.159011
                                                        0.600000
                                                                           0.0
          4
                                                                            0.0
                  1.00 0.555556
                                    0.243816
                                                        0.539130
              zipcode
                            lat
                                     long
                                           sqft_living15 sqft_lot15
                                                                         month
                                                                                year \
                                 0.394979
                                                                                  0.0
          0 0.368687
                       0.761782
                                                0.400000
                                                            0.011601
                                                                      0.545455
          1
            0.631313
                      0.683127
                                 0.112971
                                                0.318584
                                                            0.004985
                                                                      0.545455
                                                                                  0.0
          2 0.722222 0.707093
                                 0.174059
                                                0.226549
                                                            0.003837
                                                                      0.545455
                                                                                  0.0
          3 0.292929
                      0.515522
                                                0.212389
                                                            0.011556
                                                                                  1.0
                                 0.300418
                                                                      0.272727
                      0.690043 0.335565
          4 0.035354
                                                0.361062
                                                            0.011188 0.727273
                                                                                  0.0
                  day
          0 0.266667
            0.566667
          2 0.200000
          3 0.900000
          4 0.100000
          [5 rows x 21 columns]
In [106]: def check_explode(norm):
              does_explode = True
              for i in range(1,len(norm)-1):
```

## if norm[i]<norm[i-1]: does\_explode = False return does\_explode</pre>

In [107]: df\_train.describe()

	_						
Out[107]:		bedrooms	bathrooms	sqft_living	sqft_lot	floors	\
	count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	
	mean	0.074225	0.223293	0.179645	0.008794	0.201480	
	std	0.029476	0.105535	0.095728	0.024960	0.217059	
	min	0.000000	0.000000	0.000000	0.000000	0.000000	
	25%	0.062500	0.172414	0.111345	0.002704	0.000000	
	50%	0.062500	0.241379	0.162815	0.004269	0.200000	
	75%	0.093750	0.275862	0.228992	0.006166	0.400000	
	max	1.000000	1.000000	1.000000	1.000000	1.000000	
		waterfront	view	condition	grade	sqft_above	\
	count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	
	mean	0.007000	0.057350	0.602275	0.408133	0.167621	
	std	0.083377	0.188983	0.163398	0.131118	0.097864	
	min	0.000000	0.000000	0.000000	0.000000	0.000000	
	25%	0.000000	0.000000	0.500000	0.333333	0.096584	
	50%	0.000000	0.000000	0.500000	0.333333	0.140165	
	75%	0.000000	0.000000	0.750000	0.444444	0.219081	
	max	1.000000	1.000000	1.000000	1.000000	1.000000	
			yr_built	yr_renovated	zipcode	lat	\
	count	• • •	10000.000000	10000.000000	10000.000000	10000.000000	`
	mean	• • •	0.618477	0.040311	0.390369	0.649693	
	std		0.256353	0.195722	0.270295	0.223018	
	min		0.000000	0.000000	0.000000	0.000000	
	25%		0.443478	0.000000	0.161616	0.505549	
	50%		0.660870	0.000000	0.323232	0.668168	
	75%	• • •	0.843478	0.000000	0.587121	0.838829	
	max		1.000000	1.000000	1.000000	1.000000	
		long	sqft_living15	sqft_lot15	month	year	\
	count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	
	mean	0.251643	0.271562	0.013884	0.508400	0.318500	
	std	0.118330	0.122460	0.032441	0.282858	0.465918	
	min	0.000000	0.000000	0.000000	0.000000	0.000000	
	25%	0.155649	0.182301	0.005091	0.272727	0.000000	
	50%	0.237657	0.246018	0.007995	0.454545	0.000000	
	75%	0.325523	0.339823	0.010839	0.727273	1.000000	
	max	1.000000	1.000000	1.000000	1.000000	1.000000	

day count 10000.000000

```
0.493403
          mean
                     0.287392
          std
                     0.000000
          min
          25%
                     0.233333
          50%
                     0.500000
          75%
                     0.733333
          max
                     1.000000
          [8 rows x 21 columns]
In [142]: all_info={}
In [143]: import math
          def calculate_weights_and_sse(df,max_iteration,check_iter_for_explode,lmbda,learning
              weight = np.array(np.random.randn(1,21)[0])
              number_of_iteration = 0
              sse_for_each_iter = []
              norm_for_this_lr = []
              while True:
                  diff_in_y = df.dot(weight)-y_train
                  gradient = (np.transpose(df)).dot(diff_in_y) + 2*lmbda*weight
                  weight = weight - learning_rate*gradient
                  norm_for_gradient = math.sqrt(gradient.dot(gradient))
                  sse = (np.transpose(diff_in_y)).dot(diff_in_y) + lmbda*(norm_for_gradient**2
                  #print('number of iteration',number_of_iteration,norm_for_gradient,'
                                                                                            sse =
                  sse_for_each_iter.append(sse)
                  norm_for_this_lr.append(norm_for_gradient)
                  {\tt number\_of\_iteration = number\_of\_iteration + 1}
                  if(norm_for_gradient<=0.5):</pre>
```

```
"weights":weight,
                                         "sse_for_each_iteration":sse_for_each_iter,
                                         "number_of_iteration_used":number_of_iteration
                      print("##############")
                     print("number of itertion used for lmbda =",lmbda , " is ",number_of_ite
                     print("############")
                     break
                  if(number_of_iteration == check_iter_for_explode and check_explode(norm_for_
                     print('Gradient explodes for learning_rate =',learning_rate)
                      all_info[lmbda] = {"whether converged":False,
                                        "weights":np.array(np.zeros(12)),
                                         "sse_for_each_iteration":np.array([-1]*12),
                                         "number_of_iteration_used":number_of_iteration
                                       }
                     break
                  if(number_of_iteration==max_iteration):
                      all_info[lmbda] = {"whether converged":True,
                                         "weights":weight,
                                         "sse_for_each_iteration":sse_for_each_iter,
                                         "number_of_iteration_used":number_of_iteration
                     print("converges but takes too many iterations for lambda =",lmbda)
                     break
                  if (number_of_iteration%5000)==0:
                     print('number of iteration',number_of_iteration,norm_for_gradient,sse)
             print('checked for lambda = ',lmbda," now returning")
             return
In [150]: lmbda_values = [0.001,0.01,0.1,0,1,10,100]
         for lmb in lmbda_values[:1]:
             print('calculating everything for lmbda = ',lmb,'\n\n')
              calculate_weights_and_sse(df_train,100000,100,lmb)
calculating everything for lmbda = 0.001
```

all\_info[lmbda] = {"whether converged":True,

```
number of iteration 5000 3953.391108623146 94866.21992209915
number of iteration 10000 3070.158008505448 76476.81486105984
number of iteration 15000 2454.127470784738 65502.98799012292
number of iteration 20000 1999.0736869370403 58549.807360480416
number of iteration 25000 1654.389053324791 53971.69073226203
number of iteration 30000 1389.0689212703576 50859.72213456328
number of iteration 35000 1182.2606822533098 48682.16510989132
number of iteration 40000 1019.2534905389226 47116.326075368575
number of iteration 45000 889.3757042781946 45960.85953753021
number of iteration 50000 784.7590907108071 45087.17129534546
number of iteration 55000 699.5483828567017 44411.37665208678
number of iteration 60000 629.3622849932688 43877.64091572249
number of iteration 65000 570.9094212334195 43448.05710898071
number of iteration 70000 521.707427465312 43096.38004649695
number of iteration 75000 479.87490297088675 42804.08295862966
number of iteration 80000 443.976802669359 42557.836993896286
number of iteration 85000 412.9097423048035 42347.875573415324
number of iteration 90000 385.81724058388437 42166.916681490955
number of iteration 95000 362.02733307132956 42009.44169314308
converges but takes too many iterations for lambda = 0.001
checked for lambda = 0.001 now returning
In [151]: lmbda_values = [0.001,0.01,0.1,0,1,10,100]
          for lmb in lmbda values[1:3]:
              print('calculating everything for lmbda = ',lmb,'\n\n')
              calculate_weights_and_sse(df_train, 100000, 100, lmb)
calculating everything for lmbda = 0.01
number of iteration 5000 4666.816624672163 302907.1552322915
number of iteration 10000 3426.5732150585563 186458.46143740794
number of iteration 15000 2628.709836422897 129076.48769014035
number of iteration 20000 2066.88629598309 97228.56298687053
number of iteration 25000 1657.3890157795715 78534.58094440239
number of iteration 30000 1353.6742526554626 67137.20988470822
number of iteration 35000 1126.186968927287 59967.25990349583
number of iteration 40000 954.6891315745524 55321.75431663274
number of iteration 45000 824.6386988783155 52219.927325916506
number of iteration 50000 725.2923069694199 50081.995348747485
number of iteration 55000 648.6223294662576 48558.32128415069
number of iteration 60000 588.6353046119526 47434.49701307612
```

number of iteration 65000 540.8935645351668 46576.99111484759 number of iteration 70000 502.14839919464487 45901.34834804996 number of iteration 75000 470.0469760139227 45353.27938374704 number of iteration 80000 442.8986522207564 44897.26746076563 number of iteration 85000 419.4932019929069 44509.62549533512

```
number of iteration 90000 398.9637388607468 44174.22035965555 number of iteration 95000 380.68624745335524 43879.81226060978 converges but takes too many iterations for lambda = 0.01 checked for lambda = 0.01 now returning calculating everything for lmbda = 0.1
```

```
number of iteration 5000 4819.962211891929 2416435.8844847404
number of iteration 10000 3677.527484664394 1427847.5761126648
number of iteration 15000 2887.92878627315 898746.3328010812
number of iteration 20000 2306.5284064761768 590038.1757720907
number of iteration 25000 1872.4626102876457 404301.5658121514
number of iteration 30000 1546.111319252982 289829.16599484574
number of iteration 35000 1299.1301247261022 217542.05334511524
number of iteration 40000 1110.7460705667131 170697.17023809694
number of iteration 45000 965.599246902963 139484.86124315692
number of iteration 50000 852.3207722623415 118067.13731614628
number of iteration 55000 762.5345624508221 102916.77439354235
number of iteration 60000 690.1169680086808 91869.69829302837
number of iteration 65000 630.6271211690956 83576.42940240884
number of iteration 70000 580.8641106570394 77180.47772169925
number of iteration 75000 538.5265661270003 72127.60772486268
number of iteration 80000 501.95667787456256 68051.47737207776
number of iteration 85000 469.9523419072625 64704.324381360646
number of iteration 90000 441.6322720690797 61914.48329580018
number of iteration 95000 416.3407937673375 59560.03368649325
converges but takes too many iterations for lambda = 0.1
checked for lambda = 0.1 now returning
```

calculating everything for lmbda = 1

```
number of iteration 5000 3850.0948119418767 14900658.282675557 number of iteration 10000 2929.785709809465 8649825.87373344 number of iteration 15000 2357.623423891707 5617614.645058472 number of iteration 20000 1936.2612107421692 3803727.598937871 number of iteration 25000 1614.3752205595774 2657669.3012909056 number of iteration 30000 1364.8304516164796 1911994.9456908188 number of iteration 35000 1169.3240879934826 1414934.4627051463 number of iteration 40000 1014.6646274562754 1075955.603118469 number of iteration 45000 891.0994211311307 839549.9467794424 number of iteration 50000 791.3332777435535 670981.5493019668
```

```
number of iteration 55000 709.880361361556 548129.6929425572 number of iteration 60000 642.6058031992367 456674.9885331249 number of iteration 65000 586.3882911419032 387197.59479101864 number of iteration 70000 538.8668600341632 333399.1538465037 number of iteration 75000 498.24979546798636 290998.04162663524 number of iteration 80000 463.170727669272 257034.1689033956 number of iteration 85000 432.5808694926658 229426.07158857497 number of iteration 90000 405.66881761992477 206685.21008836775 number of iteration 95000 381.8011665517486 187729.31558226433 converges but takes too many iterations for lambda = 1 checked for lambda = 1 now returning calculating everything for lmbda = 10
```

number of iteration 5000 4706.11459155437 221568151.62205747 number of iteration 10000 3617.2731396423883 130922315.29330994 number of iteration 15000 2821.136115553441 79653227.49370821 number of iteration 20000 2231.9176089916105 49873105.39102261 number of iteration 25000 1793.3421725449907 32215038.708569307 number of iteration 30000 1465.426970913255 21526182.977578547 number of iteration 35000 1219.0434288071124 14910107.20140356 number of iteration 40000 1032.6910107204496 10712518.66867503 number of iteration 45000 890.4110438462532 7975265.852932706 number of iteration 50000 780.3780105253586 6136026.568537295 number of iteration 55000 693.8955975651231 4860389.485776216 number of iteration 60000 624.6381952829643 3946679.0926117967 number of iteration 65000 568.0554085057954 3271381.3954565357 number of iteration 70000 520.9016412357752 2757526.8178267977 number of iteration 75000 480.87138519071686 2356197.2416046592 number of iteration 80000 446.32480501625344 2035607.620173191 number of iteration 85000 416.08753792594536 1774596.9110438835 number of iteration 90000 389.3084507027722 1558706.684593951 number of iteration 95000 365.360593003546 1377790.7039406803 converges but takes too many iterations for lambda = 10 checked for lambda = 10 now returning calculating everything for lmbda = 100

```
number of iteration 5000 3556.5356250240084 1264967018.966031 number of iteration 10000 2367.9016706171165 560759046.5179408 number of iteration 15000 1669.1996611934858 278681110.6023904 number of iteration 20000 1226.7234916541383 150540604.98710194 number of iteration 25000 937.0898324520397 87867547.91651021 number of iteration 30000 740.7940385281794 54930253.79704245 number of iteration 35000 602.2691317674758 36324704.698476784 number of iteration 40000 500.2971313296554 25081065.022267584 number of iteration 45000 422.2312982442847 17878869.718305625 number of iteration 50000 360.45846568985843 13043675.99505407
```

```
number of iteration 55000 310.29020958481914 9678421.733222196
number of iteration 60000 268.7399794063006 7272364.66670678
number of iteration 65000 233.82562333374997 5517553.832706962
number of iteration 70000 204.1729225142143 4218662.659391614
number of iteration 75000 178.7883001531508 3246444.1780800414
number of iteration 80000 156.92609251202197 2512428.854025964
number of iteration 85000 138.00896041040545 1954439.4540890653
number of iteration 90000 121.57857772907617 1527880.3041882827
number of iteration 95000 107.2639280964741 1200261.316914767
converges but takes too many iterations for lambda = 100
checked for lambda = 100 now returning
In [153]: ###Training SSE
          all_info.keys()
          calculate_weights_and_sse(df_train,100000,100,0.00001)
          # for l in lmbda_values:
                data_l = all_info[l]
               sse_value_train=data_l['sse_for_each_iteration']
               length = len(sse_value_train)
                print("lambda = ",l," and sse = ",sse_value_train[length-1] ," for 200000 iter
number of iteration 5000 4243.150229605829 84858.54839396563
number of iteration 10000 3293.642007258502 70784.62495467352
number of iteration 15000 2631.103319351139 62028.71924793841
number of iteration 20000 2130.6351654692376 56369.27308698399
number of iteration 25000 1748.722925470233 52612.962923711064
number of iteration 30000 1456.4553165527186 50048.846539998325
number of iteration 35000 1231.9752565446545 48244.775401440595
number of iteration 40000 1058.5183482363175 46935.06922287955
number of iteration 45000 923.2757514109747 45954.396166208535
number of iteration 50000 816.5541112091834 45198.37864474703
number of iteration 55000 731.1069937546567 44599.98557276937
number of iteration 60000 661.5840318115103 44115.30657367803
number of iteration 65000 604.075168076209 43714.9254191745
number of iteration 70000 555.7409535476118 43378.64922669948
number of iteration 75000 514.5220707955848 43092.25839368556
number of iteration 80000 478.91877674832836 42845.48018066739
number of iteration 85000 447.82851240627804 42630.708283544205
number of iteration 90000 420.42928891341296 42442.18093625701
number of iteration 95000 396.0974911247715 42275.443672688685
converges but takes too many iterations for lambda = 1e-05
checked for lambda = 1e-05 now returning
In [155]: calculate_weights_and_sse(df_train,100000,100,0.0001)
number of iteration 5000 4827.788623297959 94592.61469582534
```

number of iteration 10000 3681.867468079248 75676.19162278641

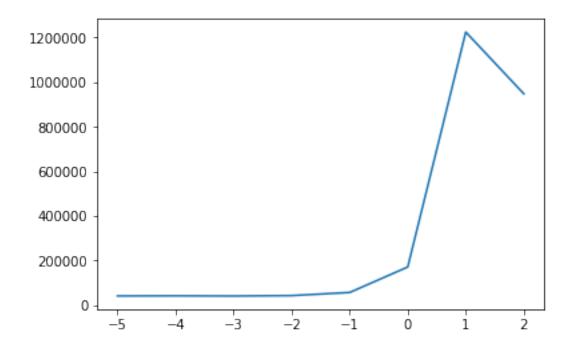
```
number of iteration 15000 2855.9649389844135 64536.47628575211
number of iteration 20000 2252.776152337358 57752.08757090865
number of iteration 25000 1808.1237828847595 53477.324249819896
number of iteration 30000 1477.8643862616598 50686.72012986799
number of iteration 35000 1230.9507983774254 48796.27844690202
number of iteration 40000 1045.1280853954513 47465.8115758571
number of iteration 45000 904.1942100306545 46492.91339779663
number of iteration 50000 796.2378644152137 45754.68452867197
number of iteration 55000 712.4771273145049 45174.988216534475
number of iteration 60000 646.458405822714 44705.681934085915
number of iteration 65000 593.4722658215547 44315.6657789068
number of iteration 70000 550.1090177999284 43984.38805779545
number of iteration 75000 513.9154781902178 43697.936242040574
number of iteration 80000 483.1320836148171 43446.65052412926
number of iteration 85000 456.49564933159473 43223.6464395616
number of iteration 90000 433.0947935865609 43023.88715424604
number of iteration 95000 412.2661449840231 42843.592153797785
converges but takes too many iterations for lambda = 0.0001
checked for lambda = 0.0001 now returning
In [154]: all_info.keys()
Out[154]: dict_keys([0.01, 0.1, 0.001, 1, 10, 100, 1e-05])
In [131]: for lmbda in lmbda values:
              sse_values = all_info[lmbda]['sse_for_each_iteration']
              1 = len(sse values)
              print(1)
              iteration = np.array([x for x in range(1,1+1)])
              print(len(iteration))
              output = pd.DataFrame({'sse': all_info[lmbda]['sse_for_each_iteration'],'iteration']
              name = 'sse_regular_'+ str(lmbda) + '.csv'
              output.to_csv(name, index=False)
              weight_values = all_info[lmbda]['weights']
              print(len(weight_values))
              itera = np.array([x for x in range(1,13)])
              print(len(itera))
              output = pd.DataFrame({'weight': all_info[lmbda]['weights'],'feature':itera})
              name = 'weight_regular_'+ str(lmbda) + '.csv'
              output.to csv(name, index=False)
In [165]: df_validation = main_df_val
          y_validation = df_validation['price']
          df_validation = df_validation.drop(['id','dummy','price'],1)
          df_validation['date'] = pd.to_datetime(main_df_val['date'])
          df_validation['month'] = df_validation['date'].apply(lambda x:x.month)
          df_validation['year'] = df_validation['date'].apply(lambda x:x.year)
```

```
df_validation['day'] = df_validation['date'].apply(lambda x:x.day)
          df_validation = df_validation.drop(['date'],1)
In [166]: df_validation.shape
Out[166]: (5597, 21)
In [167]: df_validation.columns
Out[167]: Index(['bedrooms', 'bathrooms', 'sqft_living', 'sqft_lot', 'floors',
                 'waterfront', 'view', 'condition', 'grade', 'sqft_above',
                 'sqft_basement', 'yr_built', 'yr_renovated', 'zipcode', 'lat', 'long',
                 'sqft_living15', 'sqft_lot15', 'month', 'year', 'day'],
                dtype='object')
In [175]: training_x =[]
          testing_x = []
          for feature in df_validation.columns:
              mn = df_validation[feature].min()
              mx = df_validation[feature].max()
              df_validation[feature] = df_validation[feature].apply(lambda x:((x-mn)/(mx-mn)))
          def calculate_sse_for_diff_lambda(weight):
              diff_in_y = df_validation.dot(weight)-y_validation
              error = (np.transpose(diff_in_y)).dot(diff_in_y)
              return error
          lambda_values = [0.00001,0.0001,0.001,0.01,0.1,1,10,100]
          for lmbda in lambda_values:
              info = all_info[lmbda]
              weights = np.array(info['weights'])
              sse_all = info['sse_for_each_iteration']
              training_sse = sse_all[len(sse_all)-1] - weights.dot(weights)
              training_x.append(training_sse)
              c = calculate_sse_for_diff_lambda(weights)
              testing_x.append(c)
              print('For lambda = ',lmbda , ' the training sse is = ',training_sse,' whereas to
                   ,c)
For lambda = 1e-05 the training sse is = 41965.42201510622 whereas the validation sse is
```

In [176]: #Log scale variation with respect to lambda

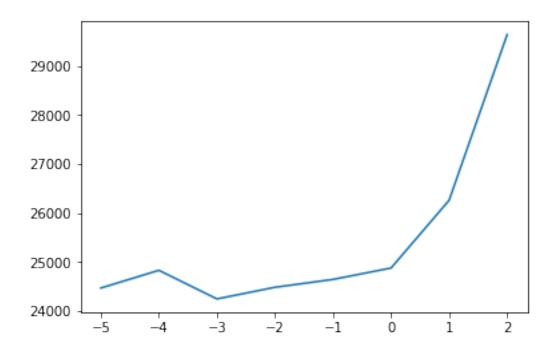
In [177]: sns.lineplot(y=training\_x,x=[-5,-4,-3,-2,-1,0,1,2])

Out[177]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1974f288438>



In [178]:  $sns.lineplot(y=testing_x,x=[-5,-4,-3,-2,-1,0,1,2])$ 

Out[178]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1974f30f128>



#Thus for lambda = 0.001 minimum SSE is observed

In []:

In []: