HOMEWORK 1_801333188

hw1-1a

February 20, 2023

[17]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing  # import scikit-learn library (source: https:
 ↪//scikit-learn.org/stable/index.html)
```

[18]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[18]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	13300000	7420	4	2	3	yes	no	no	no	yes	2	yes	furnished
1	12250000	8960	4	4	4	yes	no	no	no	yes	3	no	furnished
2	12250000	9960	3	2	2	yes	no	yes	no	no	2	yes	semi-furnished
3	12215000	7500	4	2	2	yes	no	yes	no	yes	3	yes	furnished
4	11410000	7420	4	1	2	yes	yes	yes	no	yes	2	no	furnished
..	…	…	…	…	…	…	…	…	…	…	…	…	…
540	1820000	3000	2	1	1	yes	no	yes	no	no	2	no	unfurnished
541	1767150	2400	3	1	1	no	no	no	no	no	0	no	semi-furnished
542	1750000	3620	2	1	1	yes	no	no	no	no	0	no	unfurnished
543	1750000	2910	3	1	1	no	no	no	no	no	0	no	furnished
544	1750000	3850	3	1	2	yes	no	no	no	no	0	no	unfurnished

```
[545 rows x 13 columns]
```

[19]:
```
df = df.replace(to_replace=['yes', 'no'], value=[1, 0])
df
```

[19]:
```
        price  area  bedrooms  bathrooms  stories  mainroad  guestroom  \
0    13300000  7420         4          2        3         1          0
1    12250000  8960         4          4        4         1          0
2    12250000  9960         3          2        2         1          0
3    12215000  7500         4          2        2         1          0
4    11410000  7420         4          1        2         1          1
..        ...   ...       ...        ...      ...       ...        ...
540   1820000  3000         2          1        1         1          0
541   1767150  2400         3          1        1         0          0
542   1750000  3620         2          1        1         1          0
543   1750000  2910         3          1        1         0          0
544   1750000  3850         3          1        2         1          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
0           0                0                1        2         1
1           0                0                1        3         0
2           1                0                0        2         1
3           1                0                1        3         1
4           1                0                1        2         0
..        ...              ...              ...      ...       ...
540         1                0                0        2         0
541         0                0                0        0         0
542         0                0                0        0         0
543         0                0                0        0         0
544         0                0                0        0         0

    furnishingstatus
0          furnished
1          furnished
2     semi-furnished
3          furnished
4          furnished
..               ...
540      unfurnished
541   semi-furnished
542      unfurnished
543        furnished
544      unfurnished

[545 rows x 13 columns]
```

[20]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)

# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[21]:
```
test
```

[21]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom
2	12250000	9960	3	2	2	1	0
3	12215000	7500	4	2	2	1	0
7	10150000	16200	5	3	2	1	0
15	9100000	6000	4	1	2	1	0
22	8645000	8050	3	1	1	1	1
..	…	…	…	…	…	…	…
508	2590000	4400	2	1	1	1	0
513	2485000	4400	3	1	2	1	0
520	2450000	7700	2	1	1	1	0
537	1890000	1700	3	1	2	1	0
539	1855000	2990	2	1	1	0	0

	basement	hotwaterheating	airconditioning	parking	prefarea
2	1	0	0	2	1
3	1	0	1	3	1
7	0	0	0	0	0
15	1	0	0	2	0
22	1	0	1	1	0
..	…	…	…	…	…
508	0	0	0	0	0
513	0	0	0	0	0
520	0	0	0	0	0
537	0	0	0	0	0
539	0	0	0	1	0

	furnishingstatus
2	semi-furnished
3	furnished
7	unfurnished
15	semi-furnished
22	furnished
..	…
508	unfurnished
513	unfurnished
520	unfurnished
537	unfurnished
539	unfurnished

[109 rows x 13 columns]

[22]:
```
# select specific columns for the training set
train = train[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]

# select specific columns for the test set
test = test[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]
```

[23]:
```
import sklearn.preprocessing # import scikit-learn library for data preprocessing

# create an instance of the MinMaxScaler class for scaling features to a range of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()
```

[24]:
```
train
```

[24]:

	price	area	bedrooms	bathrooms	stories	parking
62	7070000	6240	4	2	2	1
247	4550000	8400	4	1	4	3
142	5600000	10500	4	2	2	1
107	6125000	6420	3	1	3	0
483	2940000	6615	3	1	2	0
..	…	…	…	…	…	…
359	3710000	3600	3	1	1	1
36	8043000	7482	3	2	3	1
30	8400000	7475	3	2	4	2
20	8750000	4320	3	1	2	2
527	2275000	1836	2	1	1	0

[436 rows x 6 columns]

[25]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

[25]:
```
array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
        5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
        6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
        8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
        4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
        4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
        6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
        4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
        5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
```

```
 3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
 4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
 4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
 6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
 5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
 3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
 7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
 2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
 6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
 8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
 3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
 6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
 5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
 6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
 5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
 4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
 6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
 4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
 7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
 4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
 4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
 2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
 2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
 5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
 6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
 7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
 7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
 8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
 3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
 7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
 6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
 6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
 6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
 6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
 8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
 5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
 7482,  7475,  4320,  1836])
```

[26]:
```
X1_t = np.array(train['area']) # 'area' column
X2_t = np.array(train['bedrooms']) # 'bedrooms' column
X3_t = np.array(train['bathrooms']) # 'bathrooms' column
X4_t = np.array(train['stories']) # 'stories' column
X5_t = np.array(train['parking']) # 'parking' column
```

```
# create a NumPy array of ones to represent the bias term
X0_t = np.ones(len(train))
```

[27]:
```
# stack the selected feature arrays vertically using np.vstack
X = np.vstack([X0_t, X1_t, X2_t, X3_t, X4_t, X5_t])

# transpose the stacked array to make it a 6 x 436 matrix
X = X.T

# convert the stacked array to a NumPy array
X = np.array(X)

# display the NumPy array
X
```

[27]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, 1.000e+00, 4.000e+00, 3.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, 2.000e+00, 4.000e+00, 2.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, 1.000e+00, 2.000e+00, 2.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, 1.000e+00, 1.000e+00, 0.000e+00]])
```

[28]:
```
# scale the feature matrix using the fit_transform() method of the scaler object
X_scaled = scaler.fit_transform(X)

# assign the scaled feature matrix to the original variable name 'X'
X = X_scaled
```

[29]:
```
# create a 1D NumPy array of zeros with length 6
theta = np.zeros(6)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (6,1))

# display the column vector
theta
```

[29]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[30]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
↪'train'
```

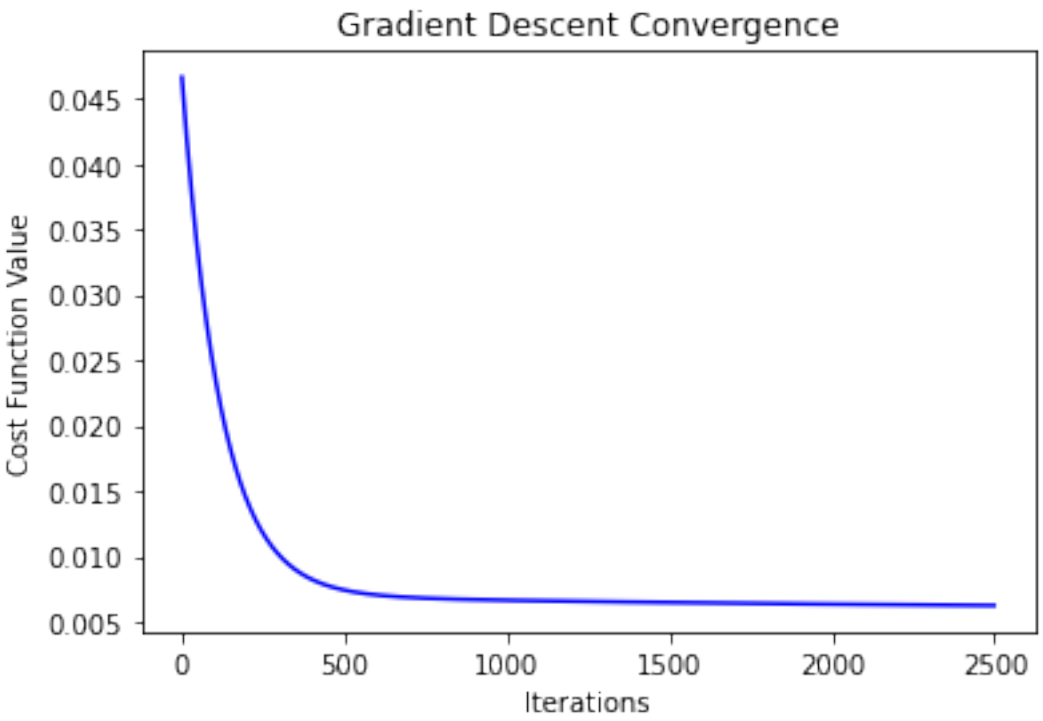
```
Y_t = np.array(train.price)

# create a copy of 'Y_t' to prevent changing the original data
Y = Y_t.copy()

# reshape 'Y' to a column vector using np.reshape
Y = np.reshape(Y, (436,1))

# create an instance of the MinMaxScaler class for scaling the target variable␣
 ↪to a range of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()

# scale the target variable using the fit_transform() method of the scaler␣
 ↪object
Y_scaled = scaler.fit_transform(Y)

# assign the scaled target variable to the original variable name 'Y'
Y = Y_scaled
```

[31]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 2500

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  6
Shape of 'j' array:  (2500, 1)
```

[32]:
```
# set the initial iteration count to zero
count = 0
```

```
# create a NumPy array 'j' to store the cost function values for each iteration
j = np.zeros(shape=(iterations, 1), dtype=float)

# perform gradient descent for the specified number of iterations
while count < iterations:

    # calculate the predicted values 'h' using the current parameters 'theta'
    h = X.dot(theta)

    # calculate the cost function value 'j' using the current parameters 'theta'
    j[count] = (1/(2*m)) * np.sum((h-Y)**2)

    # calculate the gradient of the cost function with respect to 'theta'
    grad = (1/m) * X_T.dot(h-Y)

    # update the parameters 'theta' using the learning rate 'alpha' and the␣
 ↪gradient 'grad'
    alpha = 0.01
    theta = theta - alpha * grad

    # increment the iteration count
    count += 1

# plot the cost function values over the iterations
plt.plot(j,'b-')
plt.xlabel('Iterations')
plt.ylabel('Cost Function Value')
plt.title('Gradient Descent Convergence')
plt.show()
```



[33]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
N_X0_t = np.ones(109)
```

[34]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[34]:
```
(109, 6)
```

[35]:
```
N_theta = np.array([0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(6,1)
N_theta
```

[35]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[36]:
```
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[36]:
```
(109, 1)
```

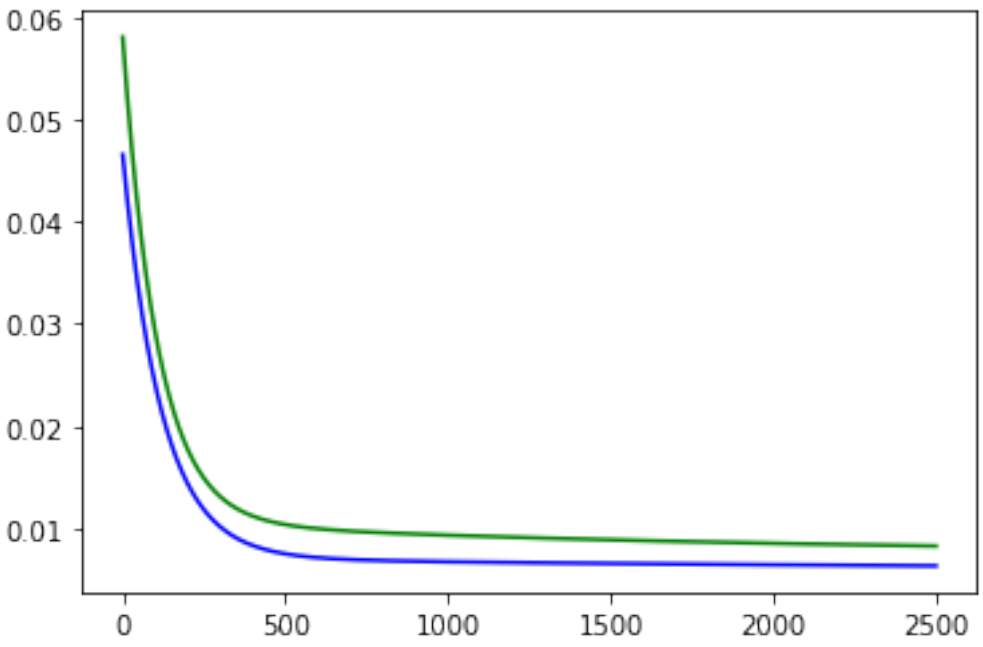
[37]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
m,n
```

[37]:
```
(109, 6)
```

[38]:
```
iterations = 2500
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)

while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)


    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)

    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta - 0.01*(grad)

    count += 1
```

[39]:
```
plt.plot(N_j,'g-')
plt.plot(j,'b-')
```

[39]:
```
[<matplotlib.lines.Line2D at 0x7f84302065e0>]
```


0.06
0.05
0.04
0.03
0.02
0.01
0
500
1000
1500
2000
2500

hw1-1b

February 20, 2023

[1]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing  # import scikit-learn library (source: https:
 ↪//scikit-learn.org/stable/index.html)
```

[2]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[2]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
0	13300000	7420	4	2	3	yes	no	no	
1	12250000	8960	4	4	4	yes	no	no	
2	12250000	9960	3	2	2	yes	no	yes	
3	12215000	7500	4	2	2	yes	no	yes	
4	11410000	7420	4	1	2	yes	yes	yes	
..	…	…	…	…	…	…	…	…	
540	1820000	3000	2	1	1	yes	no	yes	
541	1767150	2400	3	1	1	no	no	no	
542	1750000	3620	2	1	1	yes	no	no	
543	1750000	2910	3	1	1	no	no	no	
544	1750000	3850	3	1	2	yes	no	no	

	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	no	yes	2	yes	furnished
1	no	yes	3	no	furnished
2	no	no	2	yes	semi-furnished
3	no	yes	3	yes	furnished
4	no	yes	2	no	furnished
..	…	…	…	…	…
540	no	no	2	no	unfurnished
541	no	no	0	no	semi-furnished
542	no	no	0	no	unfurnished
543	no	no	0	no	furnished
544	no	no	0	no	unfurnished

```
[545 rows x 13 columns]
```

[3]:
```
df = df.replace(to_replace=['yes', 'no'], value=[1, 0])
df
```

[3]:
```
        price  area  bedrooms  bathrooms  stories  mainroad  guestroom  \
0    13300000  7420         4          2        3         1          0
1    12250000  8960         4          4        4         1          0
2    12250000  9960         3          2        2         1          0
3    12215000  7500         4          2        2         1          0
4    11410000  7420         4          1        2         1          1
..        ...   ...       ...        ...      ...       ...        ...
540   1820000  3000         2          1        1         1          0
541   1767150  2400         3          1        1         0          0
542   1750000  3620         2          1        1         1          0
543   1750000  2910         3          1        1         0          0
544   1750000  3850         3          1        2         1          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
0           0                0                1        2         1
1           0                0                1        3         0
2           1                0                0        2         1
3           1                0                1        3         1
4           1                0                1        2         0
..        ...              ...              ...      ...       ...
540         1                0                0        2         0
541         0                0                0        0         0
542         0                0                0        0         0
543         0                0                0        0         0
544         0                0                0        0         0

    furnishingstatus
0          furnished
1          furnished
2     semi-furnished
3          furnished
4          furnished
..               ...
540      unfurnished
541   semi-furnished
542      unfurnished
543        furnished
544      unfurnished

[545 rows x 13 columns]
```

[4]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)

# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[5]:
```
train
```

[5]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
62	7070000	6240	4	2	2	1	0	0	0	1	1	0	furnished
247	4550000	8400	4	1	4	1	0	0	0	0	3	0	unfurnished
142	5600000	10500	4	2	2	1	0	0	0	0	1	0	semi-furnished
107	6125000	6420	3	1	3	1	0	1	0	0	0	1	unfurnished
483	2940000	6615	3	1	2	1	0	0	0	0	0	0	semi-furnished
..	…	…	…	…	…	…	…	…	…	…	…	…	…
359	3710000	3600	3	1	1	1	0	0	0	0	1	0	unfurnished
36	8043000	7482	3	2	3	1	0	0	1	0	1	1	furnished
30	8400000	7475	3	2	4	1	0	0	0	1	2	0	unfurnished
20	8750000	4320	3	1	2	1	0	1	1	0	2	0	semi-furnished
527	2275000	1836	2	1	1	0	0	1	0	0	0	0	semi-furnished

[436 rows x 13 columns]

[6]:
```
test
```

[6]:
```
        price   area  bedrooms  bathrooms  stories  mainroad  guestroom  \
2    12250000   9960         3          2        2         1          0
3    12215000   7500         4          2        2         1          0
7    10150000  16200         5          3        2         1          0
15    9100000   6000         4          1        2         1          0
22    8645000   8050         3          1        1         1          1
..        …    …        …         …       …        …         …
508   2590000   4400         2          1        1         1          0
513   2485000   4400         3          1        2         1          0
520   2450000   7700         2          1        1         1          0
537   1890000   1700         3          1        2         1          0
539   1855000   2990         2          1        1         0          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
2           1                0                0        2         1
3           1                0                1        3         1
7           0                0                0        0         0
15          1                0                0        2         0
22          1                0                1        1         0
..        …                …                …       …        …
508         0                0                0        0         0
513         0                0                0        0         0
520         0                0                0        0         0
537         0                0                0        0         0
539         0                0                0        1         0

    furnishingstatus
2     semi-furnished
3          furnished
7        unfurnished
15    semi-furnished
22         furnished
..               …
508      unfurnished
513      unfurnished
520      unfurnished
537      unfurnished
539      unfurnished

[109 rows x 13 columns]
```

[7]:
```
# select specific columns for the training set
```

```
train = train[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
# select specific columns for the test set
test = test[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
```

[8]:
```
import sklearn.preprocessing # import scikit-learn library for data␣
 ↪preprocessing

# create an instance of the MinMaxScaler class for scaling features to a range␣
 ↪of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()
```

[9]:
```
train
```

[9]:

	price	area	bedrooms	bathrooms	stories	parking	mainroad	guestroom	basement	hotwaterheating	airconditioning	prefarea
62	7070000	6240	4	2	2	1	1	0	0	0	1	0
247	4550000	8400	4	1	4	3	1	0	0	0	0	0
142	5600000	10500	4	2	2	1	1	0	0	0	0	0
107	6125000	6420	3	1	3	0	1	0	1	0	0	1
483	2940000	6615	3	1	2	0	1	0	0	0	0	0
..	…	…	…	…	…	…	…	…	…	…	…	…
359	3710000	3600	3	1	1	1	1	0	0	0	0	0
36	8043000	7482	3	2	3	1	1	0	0	1	0	1
30	8400000	7475	3	2	4	2	1	0	0	0	1	0
20	8750000	4320	3	1	2	2	1	0	1	1	0	0
527	2275000	1836	2	1	1	0	0	0	1	0	0	0

[436 rows x 12 columns]

[10]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

[10]:
```
array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
        5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
        6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
        8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
        4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
        4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
        6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
        4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
        5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
        3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
        4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
        4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
        6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
        5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
       10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
        3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
        7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
        2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
        6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
        8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
        3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
        6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
        5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
        6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
        5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
        4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
        6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
        4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
        7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
       10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
        4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
        4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
        2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
        2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
        5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
        6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
        7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
        7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
        8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
        3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
        7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
```

```
 6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
 6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
 6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
 6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
 8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
 5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
 7482,  7475,  4320,  1836])
```

[11]:
```
X2_t = np.array(train.bedrooms)
X3_t = np.array(train.bathrooms)
X4_t = np.array(train.stories)
X5_t = np.array(train.parking)
X6_t = np.array(train.mainroad)
X7_t = np.array(train.guestroom)
X8_t = np.array(train.hotwaterheating)
X9_t = np.array(train.airconditioning)
X10_t = np.array(train.prefarea)
X11_t = np.array(train.basement)
X0_t= np.ones(436)
```

[12]:
```
# stack the selected feature arrays vertically using np.vstack
X = np.vstack([X0_t, X1_t, X2_t, X3_t, X4_t, X5_t,␣
 ↪X6_t,X7_t,X8_t,X9_t,X10_t,X11_t])

# transpose the stacked array to make it a 6 x 436 matrix
X = X.T

# convert the stacked array to a NumPy array
X = np.array(X)

# display the NumPy array
X
```

[12]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00]])
```

[13]:
```
# scale the feature matrix using the fit_transform() method of the scaler object
X_scaled = scaler.fit_transform(X)

# assign the scaled feature matrix to the original variable name 'X'
X = X_scaled
```

[14]:
```
# create a 1D NumPy array of zeros with length 12
theta = np.zeros(12)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (12,1))

# display the column vector
theta
```

[14]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[15]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
 ↪'train'
Y_t = np.array(train.price)

# create a copy of 'Y_t' to prevent changing the original data
Y = Y_t.copy()

# reshape 'Y' to a column vector using np.reshape
Y = np.reshape(Y, (436,1))

# create an instance of the MinMaxScaler class for scaling the target variable␣
 ↪to a range of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()

# scale the target variable using the fit_transform() method of the scaler␣
 ↪object
Y_scaled = scaler.fit_transform(Y)

# assign the scaled target variable to the original variable name 'Y'
```

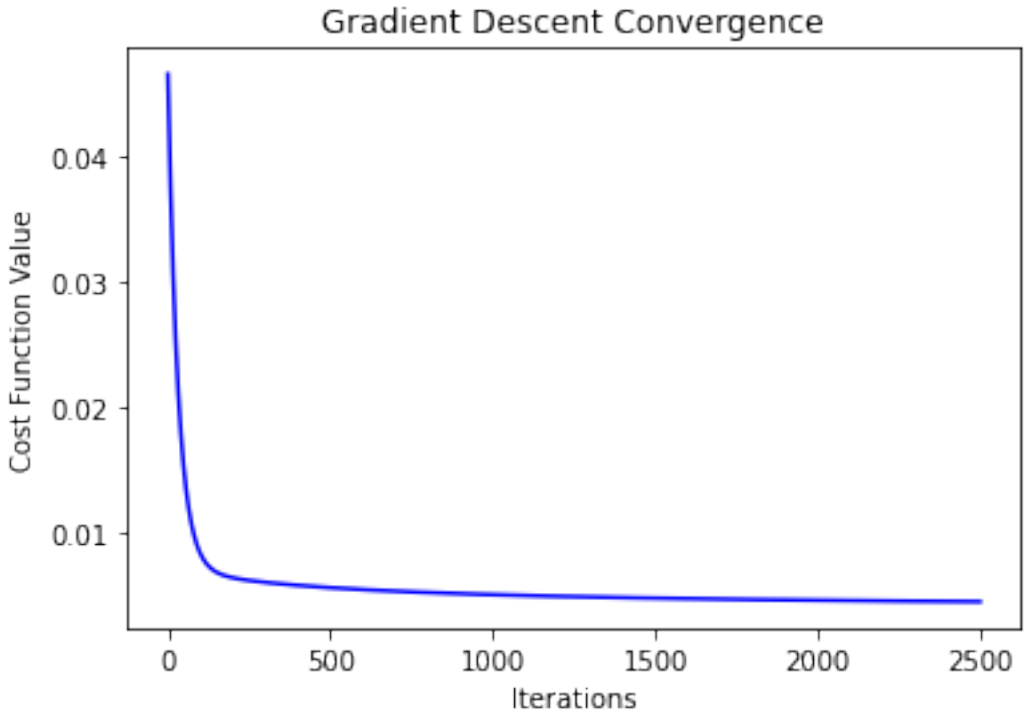
```
Y = Y_scaled
```

[16]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 2500

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  12
Shape of 'j' array:  (2500, 1)
```

[17]:
```
# set the initial iteration count to zero
count = 0

# create a NumPy array 'j' to store the cost function values for each iteration
j = np.zeros(shape=(iterations, 1), dtype=float)

# perform gradient descent for the specified number of iterations
while count < iterations:

    # calculate the predicted values 'h' using the current parameters 'theta'
    h = X.dot(theta)

    # calculate the cost function value 'j' using the current parameters 'theta'
    j[count] = (1/(2*m)) * np.sum((h-Y)**2)

    # calculate the gradient of the cost function with respect to 'theta'
    grad = (1/m) * X_T.dot(h-Y)
```

```
    # update the parameters 'theta' using the learning rate 'alpha' and the␣
 ↪gradient 'grad'
    alpha = 0.01
    theta = theta - alpha * grad

    # increment the iteration count
    count += 1

# plot the cost function values over the iterations
plt.plot(j,'b-')
plt.xlabel('Iterations')
plt.ylabel('Cost Function Value')
plt.title('Gradient Descent Convergence')
plt.show()
```



[18]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
N_X6_t = np.array(test.mainroad)
N_X7_t = np.array(test.guestroom)
```

```
N_X8_t = np.array(test.hotwaterheating)
N_X9_t = np.array(test.airconditioning)
N_X10_t = np.array(test.prefarea)
N_X11_t = np.array(test.basement)
N_X0_t = np.ones(109)
```

[19]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t,␣
 ↪N_X6_t,N_X7_t,N_X8_t,N_X9_t,N_X10_t,N_X11_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[19]:
```
(109, 12)
```

[20]:
```
# initialize the parameters for the test set
N_theta = np.array([0.,0.,0.,0.,0,0.,0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(12,1)
N_theta
```

[20]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[21]:
```
# initialize the target variable for the test set
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[21]:
```
(109, 1)
```

[22]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
```

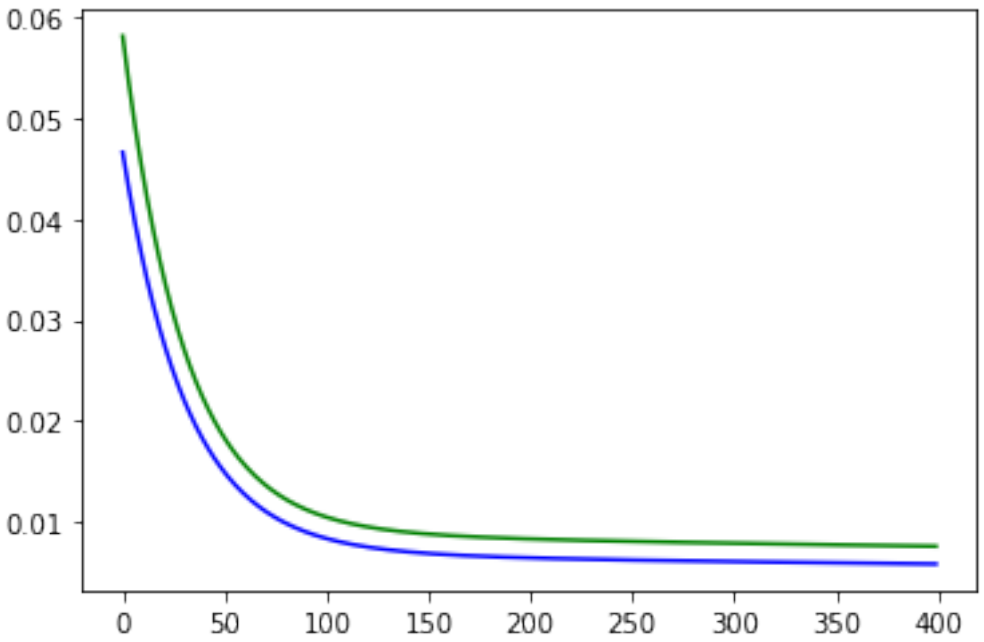
```
m,n
```

[22]: (109, 12)

[23]:
```
iterations = 2500
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)

while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)


    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)

    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta - 0.01*(grad)

    count += 1
```

[25]:
```
plt.plot(N_j[:400],'g-')
plt.plot(j[:400],'b-')
```

[25]: [<matplotlib.lines.Line2D at 0x7ffb24cc4f70>]



hw1-2a-normalization

February 20, 2023

[ ]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing  # import scikit-learn library (source: https:
↪//scikit-learn.org/stable/index.html)
```

[ ]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[ ]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
0	13300000	7420	4	2	3	yes	no	no	
1	12250000	8960	4	4	4	yes	no	no	
2	12250000	9960	3	2	2	yes	no	yes	
3	12215000	7500	4	2	2	yes	no	yes	
4	11410000	7420	4	1	2	yes	yes	yes	
..	…	…	…	…	…	…	…	…	
540	1820000	3000	2	1	1	yes	no	yes	
541	1767150	2400	3	1	1	no	no	no	
542	1750000	3620	2	1	1	yes	no	no	
543	1750000	2910	3	1	1	no	no	no	
544	1750000	3850	3	1	2	yes	no	no	

	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	no	yes	2	yes	furnished
1	no	yes	3	no	furnished
2	no	no	2	yes	semi-furnished
3	no	yes	3	yes	furnished
4	no	yes	2	no	furnished
..	…	…	…	…	…
540	no	no	2	no	unfurnished
541	no	no	0	no	semi-furnished
542	no	no	0	no	unfurnished
543	no	no	0	no	furnished
544	no	no	0	no	unfurnished

```
[545 rows x 13 columns]
```

[ ]:
```
df = df.replace(to_replace=['yes', 'no'], value=[1, 0])
df
```

[ ]:
```
        price  area  bedrooms  bathrooms  stories  mainroad  guestroom  \
0    13300000  7420         4          2        3         1          0
1    12250000  8960         4          4        4         1          0
2    12250000  9960         3          2        2         1          0
3    12215000  7500         4          2        2         1          0
4    11410000  7420         4          1        2         1          1
..        ...   ...       ...        ...      ...       ...        ...
540   1820000  3000         2          1        1         1          0
541   1767150  2400         3          1        1         0          0
542   1750000  3620         2          1        1         1          0
543   1750000  2910         3          1        1         0          0
544   1750000  3850         3          1        2         1          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
0           0                0                1        2         1
1           0                0                1        3         0
2           1                0                0        2         1
3           1                0                1        3         1
4           1                0                1        2         0
..        ...              ...              ...      ...       ...
540         1                0                0        2         0
541         0                0                0        0         0
542         0                0                0        0         0
543         0                0                0        0         0
544         0                0                0        0         0

    furnishingstatus
0          furnished
1          furnished
2     semi-furnished
3          furnished
4          furnished
..               ...
540      unfurnished
541   semi-furnished
542      unfurnished
543        furnished
544      unfurnished

[545 rows x 13 columns]
```

[ ]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)

# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[ ]:
```
test
```

[ ]:
```
        price   area  bedrooms  bathrooms  stories  mainroad  guestroom  \
2    12250000   9960         3          2        2         1          0
3    12215000   7500         4          2        2         1          0
7    10150000  16200         5          3        2         1          0
15    9100000   6000         4          1        2         1          0
22    8645000   8050         3          1        1         1          1
..        …    …        …         …       …        …         …
508   2590000   4400         2          1        1         1          0
513   2485000   4400         3          1        2         1          0
520   2450000   7700         2          1        1         1          0
537   1890000   1700         3          1        2         1          0
539   1855000   2990         2          1        1         0          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
2           1                0                0        2         1
3           1                0                1        3         1
7           0                0                0        0         0
15          1                0                0        2         0
22          1                0                1        1         0
..        …                …                …       …        …
508         0                0                0        0         0
513         0                0                0        0         0
520         0                0                0        0         0
537         0                0                0        0         0
539         0                0                0        1         0

    furnishingstatus
2     semi-furnished
3          furnished
7        unfurnished
15    semi-furnished
22         furnished
..               …
508      unfurnished
513      unfurnished
520      unfurnished
537      unfurnished
539      unfurnished
```

```
[109 rows x 13 columns]
```

[ ]:
```
# select specific columns for the training set
train = train[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]

# select specific columns for the test set
test = test[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]
```

[ ]:
```
import sklearn.preprocessing # import scikit-learn library for data preprocessing

# create an instance of the MinMaxScaler class for scaling features to a range of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()
```

[ ]:
```
train
```

[ ]:

	price	area	bedrooms	bathrooms	stories	parking
62	7070000	6240	4	2	2	1
247	4550000	8400	4	1	4	3
142	5600000	10500	4	2	2	1
107	6125000	6420	3	1	3	0
483	2940000	6615	3	1	2	0
..	…	…	…	…	…	…
359	3710000	3600	3	1	1	1
36	8043000	7482	3	2	3	1
30	8400000	7475	3	2	4	2
20	8750000	4320	3	1	2	2
527	2275000	1836	2	1	1	0

```
[436 rows x 6 columns]
```

[ ]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

[ ]:
```
array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
        5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
        6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
        8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
        4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
        4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
        6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
        4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
        5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
```

```
 3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
 4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
 4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
 6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
 5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
 3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
 7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
 2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
 6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
 8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
 3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
 6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
 5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
 6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
 5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
 4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
 6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
 4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
 7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
 4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
 4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
 2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
 2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
 5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
 6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
 7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
 7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
 8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
 3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
 7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
 6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
 6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
 6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
 6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
 8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
 5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
 7482,  7475,  4320,  1836])
```

[ ]:
```
X1_t = np.array(train['area']) # 'area' column
X2_t = np.array(train['bedrooms']) # 'bedrooms' column
X3_t = np.array(train['bathrooms']) # 'bathrooms' column
X4_t = np.array(train['stories']) # 'stories' column
X5_t = np.array(train['parking']) # 'parking' column
```

```
# create a NumPy array of ones to represent the bias term
X0_t = np.ones(len(train))
```

[ ]:
```
# stack the selected feature arrays vertically using np.vstack
X = np.vstack([X0_t, X1_t, X2_t, X3_t, X4_t, X5_t])

# transpose the stacked array to make it a 6 x 436 matrix
X = X.T

# convert the stacked array to a NumPy array
X = np.array(X)

# display the NumPy array
X
```

[ ]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, 1.000e+00, 4.000e+00, 3.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, 2.000e+00, 4.000e+00, 2.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, 1.000e+00, 2.000e+00, 2.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, 1.000e+00, 1.000e+00, 0.000e+00]])
```

[ ]:
```
# scale the feature matrix using the fit_transform() method of the scaler object
X_scaled = scaler.fit_transform(X)

# assign the scaled feature matrix to the original variable name 'X'
X = X_scaled
```

[ ]:
```
# create a 1D NumPy array of zeros with length 6
theta = np.zeros(6)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (6,1))

# display the column vector
theta
```

[ ]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[ ]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
↪'train'
```

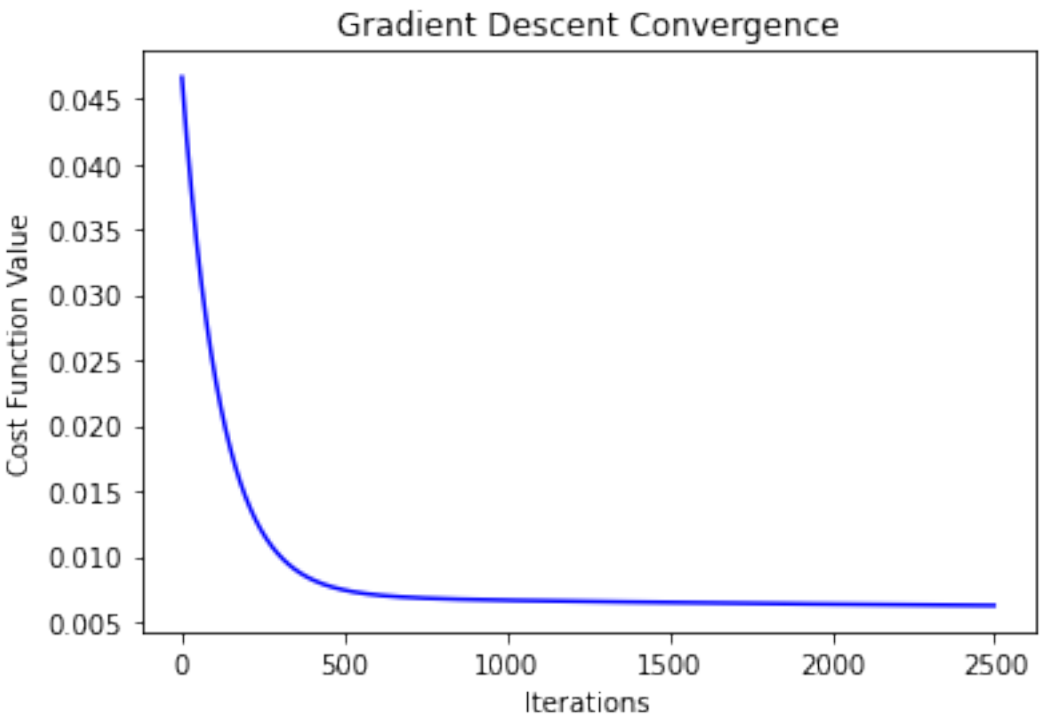
```
Y_t = np.array(train.price)

# create a copy of 'Y_t' to prevent changing the original data
Y = Y_t.copy()

# reshape 'Y' to a column vector using np.reshape
Y = np.reshape(Y, (436,1))

# create an instance of the MinMaxScaler class for scaling the target variable␣
 ↪to a range of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()

# scale the target variable using the fit_transform() method of the scaler␣
 ↪object
Y_scaled = scaler.fit_transform(Y)

# assign the scaled target variable to the original variable name 'Y'
Y = Y_scaled
```

[ ]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 2500

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  6
Shape of 'j' array:  (2500, 1)
```

[ ]:
```
# set the initial iteration count to zero
count = 0
```

```
# create a NumPy array 'j' to store the cost function values for each iteration
j = np.zeros(shape=(iterations, 1), dtype=float)

# perform gradient descent for the specified number of iterations
while count < iterations:

    # calculate the predicted values 'h' using the current parameters 'theta'
    h = X.dot(theta)

    # calculate the cost function value 'j' using the current parameters 'theta'
    j[count] = (1/(2*m)) * np.sum((h-Y)**2)

    # calculate the gradient of the cost function with respect to 'theta'
    grad = (1/m) * X_T.dot(h-Y)

    # update the parameters 'theta' using the learning rate 'alpha' and the␣
 ↪gradient 'grad'
    alpha = 0.01
    theta = theta - alpha * grad

    # increment the iteration count
    count += 1

# plot the cost function values over the iterations
plt.plot(j,'b-')
plt.xlabel('Iterations')
plt.ylabel('Cost Function Value')
plt.title('Gradient Descent Convergence')
plt.show()
```



[ ]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
N_X0_t = np.ones(109)
```

[ ]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[ ]:
```
(109, 6)
```

[ ]:
```
N_theta = np.array([0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(6,1)
N_theta
```

[ ]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[ ]:
```
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[ ]:
```
(109, 1)
```

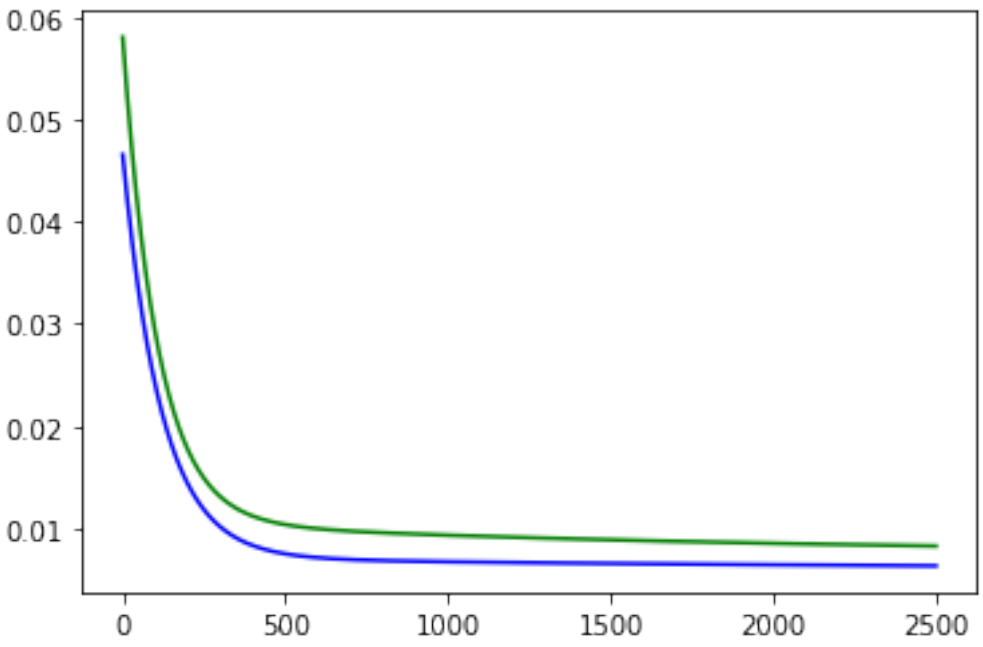
[ ]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
m,n
```

[ ]:
```
(109, 6)
```

[ ]:
```
iterations = 2500
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)

while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)


    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)

    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta - 0.01*(grad)

    count += 1
```

[ ]:
```
plt.plot(N_j,'g-')
plt.plot(j,'b-')
```

[ ]:
```
[<matplotlib.lines.Line2D at 0x7fdb15afbe20>]
```


0.06
0.05
0.04
0.03
0.02
0.01
0
500
1000
1500
2000
2500

hw1-2a-standardization

February 20, 2023

[ ]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing
```

[ ]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[ ]:
```
        price  area  bedrooms  bathrooms  stories mainroad guestroom basement  \
0    13300000  7420         4          2        3      yes        no       no
1    12250000  8960         4          4        4      yes        no       no
2    12250000  9960         3          2        2      yes        no      yes
3    12215000  7500         4          2        2      yes        no      yes
4    11410000  7420         4          1        2      yes       yes      yes
..        ...   ...       ...        ...      ...      ...       ...      ...
540   1820000  3000         2          1        1      yes        no      yes
541   1767150  2400         3          1        1       no        no       no
542   1750000  3620         2          1        1      yes        no       no
543   1750000  2910         3          1        1       no        no       no
544   1750000  3850         3          1        2      yes        no       no

    hotwaterheating airconditioning  parking prefarea furnishingstatus
0                no             yes        2      yes        furnished
1                no             yes        3       no        furnished
2                no              no        2      yes   semi-furnished
3                no             yes        3      yes        furnished
4                no             yes        2       no        furnished
..              ...             ...      ...      ...              ...
540              no              no        2       no      unfurnished
541              no              no        0       no   semi-furnished
542              no              no        0       no      unfurnished
543              no              no        0       no        furnished
544              no              no        0       no      unfurnished

[545 rows x 13 columns]
```

[ ]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)

# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[ ]:
```
train
```

[ ]:
```
       price   area  bedrooms  bathrooms  stories mainroad guestroom basement  \
62   7070000   6240         4          2        2      yes        no       no
247  4550000   8400         4          1        4      yes        no       no
142  5600000  10500         4          2        2      yes        no       no
107  6125000   6420         3          1        3      yes        no      yes
483  2940000   6615         3          1        2      yes        no       no
..       …     …       …         …       …      …       …       …
359  3710000   3600         3          1        1      yes        no       no
36   8043000   7482         3          2        3      yes        no       no
30   8400000   7475         3          2        4      yes        no       no
20   8750000   4320         3          1        2      yes        no      yes
527  2275000   1836         2          1        1       no        no      yes

    hotwaterheating airconditioning  parking prefarea furnishingstatus
62               no              yes        1       no        furnished
247              no               no        3       no      unfurnished
142              no               no        1       no   semi-furnished
107              no               no        0      yes      unfurnished
483              no               no        0       no   semi-furnished
..              …              …       …       …              …
359              no               no        1       no      unfurnished
36              yes               no        1      yes        furnished
30               no              yes        2       no      unfurnished
20              yes               no        2       no   semi-furnished
527              no               no        0       no   semi-furnished

[436 rows x 13 columns]
```

[ ]:
```
test
```

[ ]:
```
       price   area  bedrooms  bathrooms  stories mainroad guestroom  \
2   12250000   9960         3          2        2      yes        no
3   12215000   7500         4          2        2      yes        no
7   10150000  16200         5          3        2      yes        no
15   9100000   6000         4          1        2      yes        no
22   8645000   8050         3          1        1      yes       yes
..       …     …       …         …       …      …       …
508  2590000   4400         2          1        1      yes        no
513  2485000   4400         3          1        2      yes        no
```

```
520   2450000   7700         2          1        1      yes        no
537   1890000   1700         3          1        2      yes        no
539   1855000   2990         2          1        1       no        no

    basement hotwaterheating airconditioning  parking prefarea  \
2        yes              no              no        2      yes
3        yes              no             yes        3      yes
7         no              no              no        0       no
15       yes              no              no        2       no
22       yes              no             yes        1       no
..       …               …               …       …       …
508       no              no              no        0       no
513       no              no              no        0       no
520       no              no              no        0       no
537       no              no              no        0       no
539       no              no              no        1       no

    furnishingstatus
2     semi-furnished
3          furnished
7        unfurnished
15    semi-furnished
22         furnished
..               …
508      unfurnished
513      unfurnished
520      unfurnished
537      unfurnished
539      unfurnished

[109 rows x 13 columns]
```

[ ]:
```
# select specific columns for the training set
train = train[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]

# select specific columns for the test set
test = test[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]
```

[ ]:
```
scaler = preprocessing.StandardScaler()
```

[ ]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

```
[ ]: array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
             5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
             6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
             8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
             4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
             4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
             6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
             4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
             5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
             3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
             4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
             4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
             6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
             5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
            10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
             3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
             7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
             2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
             6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
             8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
             3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
             6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
             5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
             6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
             5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
             4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
             6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
             4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
             7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
            10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
             4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
             4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
             2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
             2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
             5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
             6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
             7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
             7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
             8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
             3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
             7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
             6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
             6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
             6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
             6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
             8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
             5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
```

```
        13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
         7482,  7475,  4320,  1836])
```

[ ]:
```
X1_t = np.array(train['area']) # 'area' column
X2_t = np.array(train['bedrooms']) # 'bedrooms' column
X3_t = np.array(train['bathrooms']) # 'bathrooms' column
X4_t = np.array(train['stories']) # 'stories' column
X5_t = np.array(train['parking']) # 'parking' column

# create a NumPy array of ones to represent the bias term
X0_t = np.ones(len(train))
```

[ ]:
```
# stack the selected feature arrays vertically using np.vstack
X = np.vstack([X0_t, X1_t, X2_t, X3_t, X4_t, X5_t])

# transpose the stacked array to make it a 6 x 436 matrix
X = X.T

# convert the stacked array to a NumPy array
X = np.array(X)

# display the NumPy array
X
```

[ ]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, 1.000e+00, 4.000e+00, 3.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, 2.000e+00, 4.000e+00, 2.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, 1.000e+00, 2.000e+00, 2.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, 1.000e+00, 1.000e+00, 0.000e+00]])
```

[ ]:
```
x = scaler.fit_transform(X)
X = x
```

[ ]:
```
# create a 1D NumPy array of zeros with length 6
theta = np.zeros(6)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (6,1))

# display the column vector
theta
```

[ ]:
```
array([[0.],
       [0.],
       [0.],
```

```
       [0.],
       [0.],
       [0.]])
```

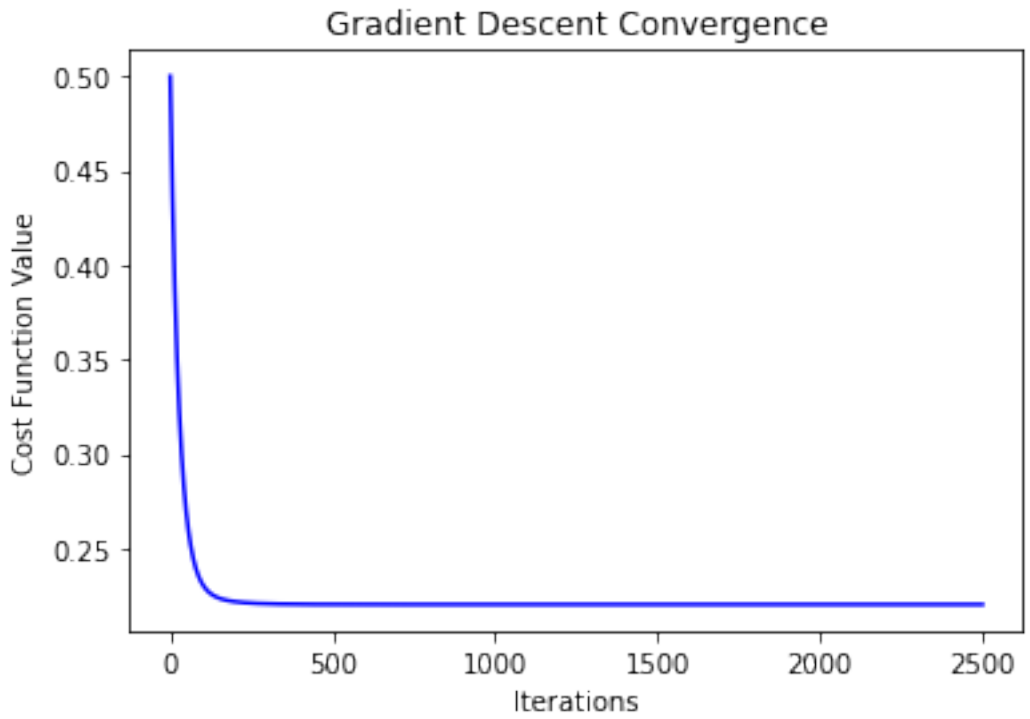
[ ]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
 ↪'train'
#h = np.matmul(X,theta)
Y_t = np.array(train.price)
Y = Y_t
Y = Y_t.reshape(436,1)
y = scaler.fit_transform(Y)
Y=y
```

[ ]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 2500

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  6
Shape of 'j' array:  (2500, 1)
```

[ ]:
```
# set the initial iteration count to zero
count = 0

# create a NumPy array 'j' to store the cost function values for each iteration
j = np.zeros(shape=(iterations, 1), dtype=float)

# perform gradient descent for the specified number of iterations
while count < iterations:
```

```
    # calculate the predicted values 'h' using the current parameters 'theta'
    h = X.dot(theta)

    # calculate the cost function value 'j' using the current parameters 'theta'
    j[count] = (1/(2*m)) * np.sum((h-Y)**2)

    # calculate the gradient of the cost function with respect to 'theta'
    grad = (1/m) * X_T.dot(h-Y)

    # update the parameters 'theta' using the learning rate 'alpha' and the␣
 ↪gradient 'grad'
    alpha = 0.01
    theta = theta - alpha * grad

    # increment the iteration count
    count += 1

# plot the cost function values over the iterations
plt.plot(j,'b-')
plt.xlabel('Iterations')
plt.ylabel('Cost Function Value')
plt.title('Gradient Descent Convergence')
plt.show()
```



[ ]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
N_X0_t = np.ones(109)
```

[ ]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[ ]:
```
(109, 6)
```

[ ]:
```
N_theta = np.array([0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(6,1)
N_theta
```

[ ]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[ ]:
```
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[ ]:
```
(109, 1)
```

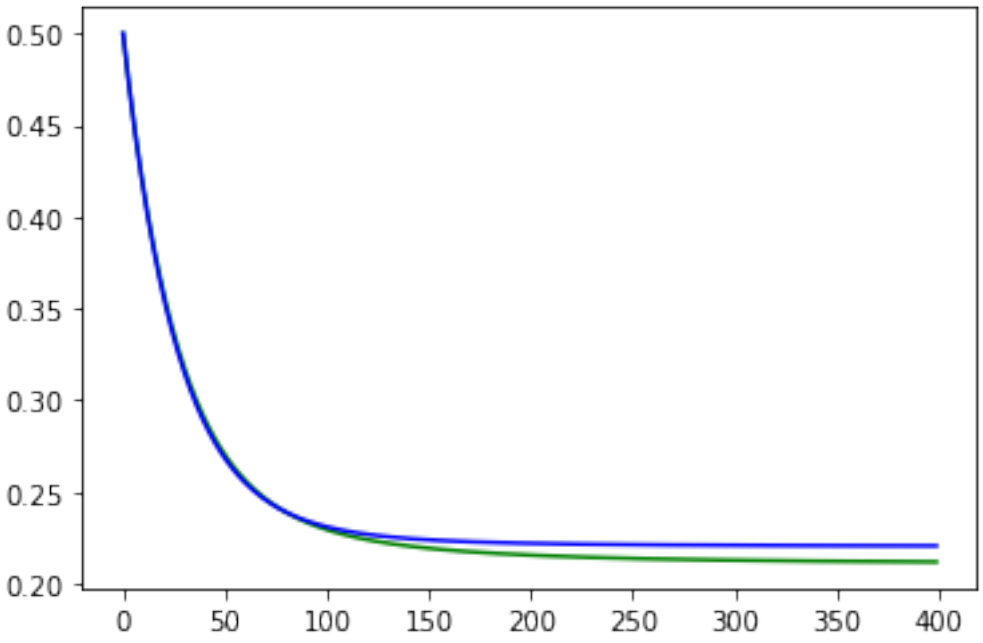
[ ]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
m,n
```

[ ]:
```
(109, 6)
```

[ ]:
```
iterations = 2500
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)

while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)


    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)

    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta - 0.01*(grad)

    count += 1
```

[ ]:
```
plt.plot(N_j[:400],'g-')
plt.plot(j[:400],'b-')
```

[ ]:
```
[<matplotlib.lines.Line2D at 0x7f9ecbdcf5b0>]
```



hw1-2b-normalization

February 20, 2023

[5]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing  # import scikit-learn library (source: https:
↪//scikit-learn.org/stable/index.html)
```

[ ]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[8]:
```
df = df.replace(to_replace=['yes', 'no'], value=[1, 0])
df
```

[8]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	\
0	13300000	7420	4	2	3	1	0	
1	12250000	8960	4	4	4	1	0	
2	12250000	9960	3	2	2	1	0	
3	12215000	7500	4	2	2	1	0	
4	11410000	7420	4	1	2	1	1	
..	…	…	…	…	…	…	…	
540	1820000	3000	2	1	1	1	0	
541	1767150	2400	3	1	1	0	0	
542	1750000	3620	2	1	1	1	0	
543	1750000	2910	3	1	1	0	0	
544	1750000	3850	3	1	2	1	0	

	basement	hotwaterheating	airconditioning	parking	prefarea	\
0	0	0	1	2	1	
1	0	0	1	3	0	
2	1	0	0	2	1	
3	1	0	1	3	1	
4	1	0	1	2	0	
..	…	…	…	…	…	
540	1	0	0	2	0	
541	0	0	0	0	0	
542	0	0	0	0	0	

```
543         0                 0                 0         0         0
544         0                 0                 0         0         0

    furnishingstatus
0          furnished
1          furnished
2     semi-furnished
3          furnished
4          furnished
..               …
540      unfurnished
541   semi-furnished
542      unfurnished
543        furnished
544      unfurnished

[545 rows x 13 columns]
```

[9]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)

# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[10]:
```
train
```

[10]:
```
       price   area  bedrooms  bathrooms  stories  mainroad  guestroom  \
62   7070000   6240         4          2        2         1          0
247  4550000   8400         4          1        4         1          0
142  5600000  10500         4          2        2         1          0
107  6125000   6420         3          1        3         1          0
483  2940000   6615         3          1        2         1          0
..       …    …       …         …       …         …        …
359  3710000   3600         3          1        1         1          0
36   8043000   7482         3          2        3         1          0
30   8400000   7475         3          2        4         1          0
20   8750000   4320         3          1        2         1          0
527  2275000   1836         2          1        1         0          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
62          0                0                1        1         0
247         0                0                0        3         0
142         0                0                0        1         0
107         1                0                0        0         1
483         0                0                0        0         0
..        …              …              …      …        …
359         0                0                0        1         0
```

```
36          0                 1                 0         1          1
30          0                 0                 1         2          0
20          1                 1                 0         2          0
527         1                 0                 0         0          0

    furnishingstatus
62         furnished
247      unfurnished
142   semi-furnished
107      unfurnished
483   semi-furnished
..                ...
359      unfurnished
36         furnished
30       unfurnished
20    semi-furnished
527   semi-furnished

[436 rows x 13 columns]
```

[11]:
```
test
```

[11]:
```
        price   area  bedrooms  bathrooms  stories  mainroad  guestroom  \
2    12250000   9960         3          2        2         1          0
3    12215000   7500         4          2        2         1          0
7    10150000  16200         5          3        2         1          0
15    9100000   6000         4          1        2         1          0
22    8645000   8050         3          1        1         1          1
..        ...    ...       ...        ...      ...       ...        ...
508   2590000   4400         2          1        1         1          0
513   2485000   4400         3          1        2         1          0
520   2450000   7700         2          1        1         1          0
537   1890000   1700         3          1        2         1          0
539   1855000   2990         2          1        1         0          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
2           1                0                0        2         1
3           1                0                1        3         1
7           0                0                0        0         0
15          1                0                0        2         0
22          1                0                1        1         0
..        ...              ...              ...      ...       ...
508         0                0                0        0         0
513         0                0                0        0         0
520         0                0                0        0         0
537         0                0                0        0         0
539         0                0                0        1         0
```

```
    furnishingstatus
2     semi-furnished
3          furnished
7        unfurnished
15    semi-furnished
22         furnished
..               …
508      unfurnished
513      unfurnished
520      unfurnished
537      unfurnished
539      unfurnished

[109 rows x 13 columns]
```

[12]:
```
# select specific columns for the training set
train = train[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
# select specific columns for the test set
test = test[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
```

[13]:
```
import sklearn.preprocessing # import scikit-learn library for data␣
 ↪preprocessing

# create an instance of the MinMaxScaler class for scaling features to a range␣
 ↪of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()
```

[14]:
```
train
```

[14]:
```
       price   area  bedrooms  bathrooms  stories  parking  mainroad  \
62   7070000   6240         4          2        2        1         1
247  4550000   8400         4          1        4        3         1
142  5600000  10500         4          2        2        1         1
107  6125000   6420         3          1        3        0         1
483  2940000   6615         3          1        2        0         1
..       …     …         …          …        …        …         …
359  3710000   3600         3          1        1        1         1
36   8043000   7482         3          2        3        1         1
30   8400000   7475         3          2        4        2         1
20   8750000   4320         3          1        2        2         1
527  2275000   1836         2          1        1        0         0
```

```
     guestroom  basement  hotwaterheating  airconditioning  prefarea
62           0         0                0                1         0
247          0         0                0                0         0
142          0         0                0                0         0
107          0         1                0                0         1
483          0         0                0                0         0
..         …         …              …                …       …
359          0         0                0                0         0
36           0         0                1                0         1
30           0         0                0                1         0
20           0         1                1                0         0
527          0         1                0                0         0

[436 rows x 12 columns]
```

[15]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

[15]:
```
array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
        5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
        6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
        8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
        4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
        4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
        6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
        4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
        5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
        3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
        4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
        4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
        6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
        5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
       10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
        3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
        7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
        2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
        6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
        8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
        3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
        6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
        5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
        6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
        5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
        4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
```

```
 6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
 4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
 7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
 4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
 4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
 2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
 2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
 5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
 6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
 7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
 7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
 8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
 3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
 7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
 6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
 6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
 6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
 6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
 8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
 5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
 7482,  7475,  4320,  1836])
```

[16]:
```
X2_t = np.array(train.bedrooms)
X3_t = np.array(train.bathrooms)
X4_t = np.array(train.stories)
X5_t = np.array(train.parking)
X6_t = np.array(train.mainroad)
X7_t = np.array(train.guestroom)
X8_t = np.array(train.hotwaterheating)
X9_t = np.array(train.airconditioning)
X10_t = np.array(train.prefarea)
X11_t = np.array(train.basement)
X0_t= np.ones(436)
```

[17]:
```
# stack the selected feature arrays vertically using np.vstack
X = np.vstack([X0_t, X1_t, X2_t, X3_t, X4_t, X5_t,␣
 ↪X6_t,X7_t,X8_t,X9_t,X10_t,X11_t])

# transpose the stacked array to make it a 6 x 436 matrix
X = X.T

# convert the stacked array to a NumPy array
X = np.array(X)

# display the NumPy array
```

```
X
```

[17]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00]])
```

[18]:
```
# scale the feature matrix using the fit_transform() method of the scaler object
X_scaled = scaler.fit_transform(X)

# assign the scaled feature matrix to the original variable name 'X'
X = X_scaled
```

[19]:
```
# create a 1D NumPy array of zeros with length 12
theta = np.zeros(12)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (12,1))

# display the column vector
theta
```

[19]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[20]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
↪'train'
Y_t = np.array(train.price)
```

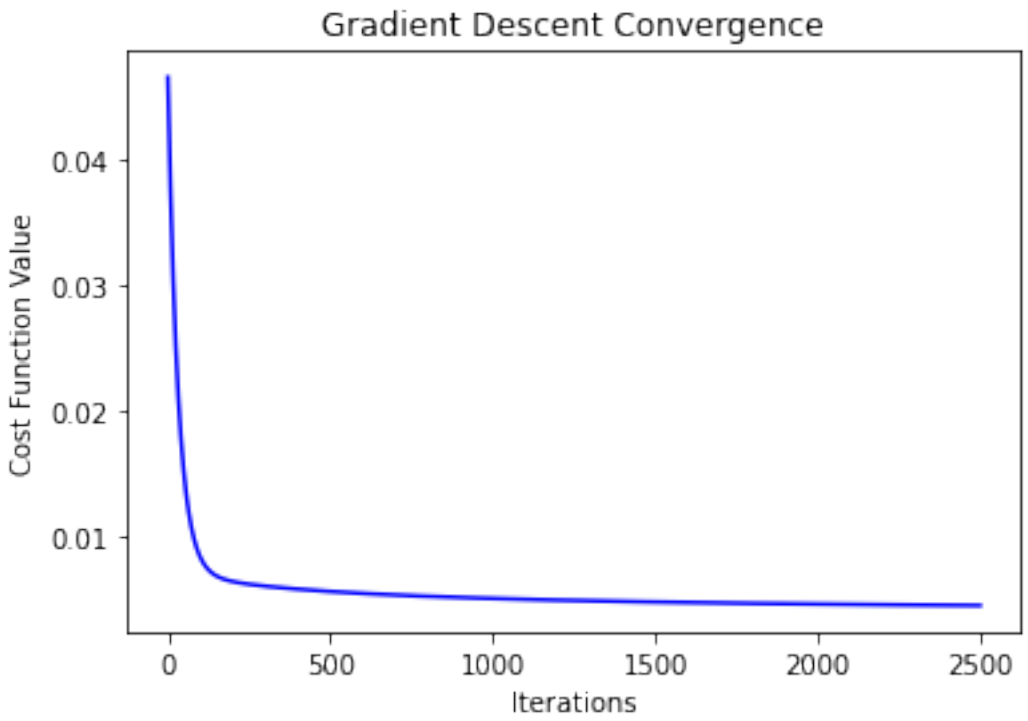
```
# create a copy of 'Y_t' to prevent changing the original data
Y = Y_t.copy()

# reshape 'Y' to a column vector using np.reshape
Y = np.reshape(Y, (436,1))

# create an instance of the MinMaxScaler class for scaling the target variable␣
 ↪to a range of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()

# scale the target variable using the fit_transform() method of the scaler␣
 ↪object
Y_scaled = scaler.fit_transform(Y)

# assign the scaled target variable to the original variable name 'Y'
Y = Y_scaled
```

[21]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 2500

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  12
Shape of 'j' array:  (2500, 1)
```

[22]:
```
# set the initial iteration count to zero
count = 0
```

```
# create a NumPy array 'j' to store the cost function values for each iteration
j = np.zeros(shape=(iterations, 1), dtype=float)

# perform gradient descent for the specified number of iterations
while count < iterations:

    # calculate the predicted values 'h' using the current parameters 'theta'
    h = X.dot(theta)

    # calculate the cost function value 'j' using the current parameters 'theta'
    j[count] = (1/(2*m)) * np.sum((h-Y)**2)

    # calculate the gradient of the cost function with respect to 'theta'
    grad = (1/m) * X_T.dot(h-Y)

    # update the parameters 'theta' using the learning rate 'alpha' and the␣
 ↪gradient 'grad'
    alpha = 0.01
    theta = theta - alpha * grad

    # increment the iteration count
    count += 1

# plot the cost function values over the iterations
plt.plot(j,'b-')
plt.xlabel('Iterations')
plt.ylabel('Cost Function Value')
plt.title('Gradient Descent Convergence')
plt.show()
```



[23]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
N_X6_t = np.array(test.mainroad)
N_X7_t = np.array(test.guestroom)
N_X8_t = np.array(test.hotwaterheating)
N_X9_t = np.array(test.airconditioning)
N_X10_t = np.array(test.prefarea)
N_X11_t = np.array(test.basement)
N_X0_t = np.ones(109)
```

[24]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t,␣
 ↪N_X6_t,N_X7_t,N_X8_t,N_X9_t,N_X10_t,N_X11_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[24]: (109, 12)

[25]:
```
# initialize the parameters for the test set
N_theta = np.array([0.,0.,0.,0.,0,0.,0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(12,1)
N_theta
```

[25]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[26]:
```
# initialize the target variable for the test set
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[26]: (109, 1)

[27]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
m,n
```

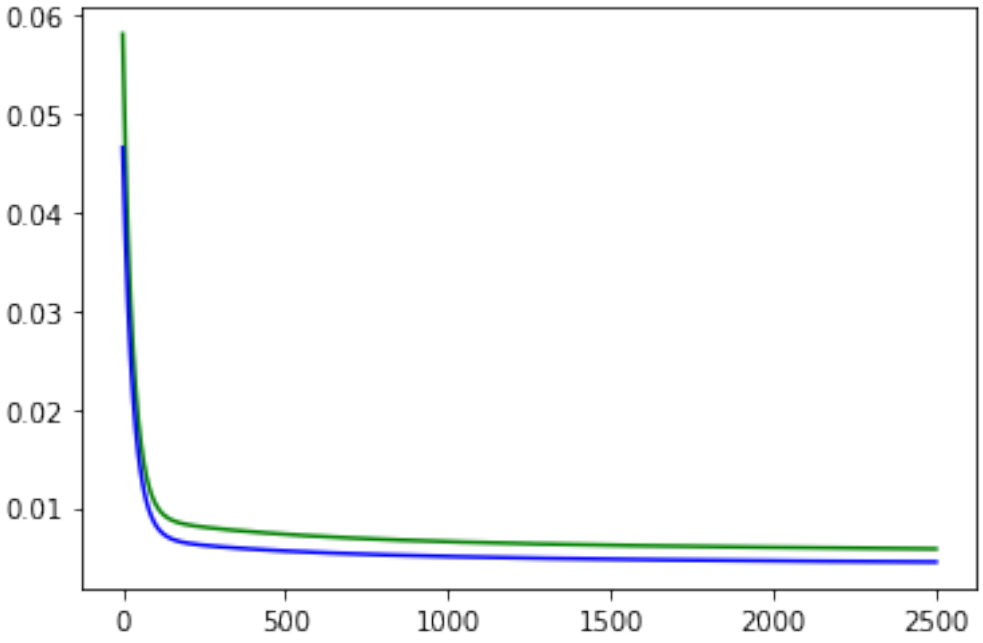
[27]: (109, 12)

[28]:
```
iterations = 2500
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)

while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)


    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)
```

```
    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta - 0.01*(grad)

    count += 1
```

[29]:
```
plt.plot(N_j,'g-')
plt.plot(j,'b-')
```

[29]: [<matplotlib.lines.Line2D at 0x7fb76ca055b0>]



hw1-2b-standardization

February 20, 2023

[ ]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing
```

[ ]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[ ]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	13300000	7420	4	2	3	yes	no	no	no	yes	2	yes	furnished
1	12250000	8960	4	4	4	yes	no	no	no	yes	3	no	furnished
2	12250000	9960	3	2	2	yes	no	yes	no	no	2	yes	semi-furnished
3	12215000	7500	4	2	2	yes	no	yes	no	yes	3	yes	furnished
4	11410000	7420	4	1	2	yes	yes	yes	no	yes	2	no	furnished
..	…	…	…	…	…	…	…	…	…	…	…	…	…
540	1820000	3000	2	1	1	yes	no	yes	no	no	2	no	unfurnished
541	1767150	2400	3	1	1	no	no	no	no	no	0	no	semi-furnished
542	1750000	3620	2	1	1	yes	no	no	no	no	0	no	unfurnished
543	1750000	2910	3	1	1	no	no	no	no	no	0	no	furnished
544	1750000	3850	3	1	2	yes	no	no	no	no	0	no	unfurnished

[545 rows x 13 columns]

[ ]:
```
df = df.replace(to_replace=['yes', 'no'], value=[1, 0])
df
```

[ ]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	\
0	13300000	7420	4	2	3	1	0	
1	12250000	8960	4	4	4	1	0	
2	12250000	9960	3	2	2	1	0	
3	12215000	7500	4	2	2	1	0	
4	11410000	7420	4	1	2	1	1	
..	…	…	…	…	…	…	…	
540	1820000	3000	2	1	1	1	0	
541	1767150	2400	3	1	1	0	0	
542	1750000	3620	2	1	1	1	0	
543	1750000	2910	3	1	1	0	0	
544	1750000	3850	3	1	2	1	0	

	basement	hotwaterheating	airconditioning	parking	prefarea	\
0	0	0	1	2	1	
1	0	0	1	3	0	
2	1	0	0	2	1	
3	1	0	1	3	1	
4	1	0	1	2	0	
..	…	…	…	…	…	
540	1	0	0	2	0	
541	0	0	0	0	0	
542	0	0	0	0	0	
543	0	0	0	0	0	
544	0	0	0	0	0	

	furnishingstatus
0	furnished
1	furnished
2	semi-furnished
3	furnished
4	furnished
..	…
540	unfurnished
541	semi-furnished
542	unfurnished
543	furnished
544	unfurnished

[545 rows x 13 columns]

[ ]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)
```

```
# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[ ]:
```
train
```

[ ]:
```
       price   area  bedrooms  bathrooms  stories  mainroad  guestroom  \
62   7070000   6240         4          2        2         1          0
247  4550000   8400         4          1        4         1          0
142  5600000  10500         4          2        2         1          0
107  6125000   6420         3          1        3         1          0
483  2940000   6615         3          1        2         1          0
..       …     …       …         …       …        …         …
359  3710000   3600         3          1        1         1          0
36   8043000   7482         3          2        3         1          0
30   8400000   7475         3          2        4         1          0
20   8750000   4320         3          1        2         1          0
527  2275000   1836         2          1        1         0          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
62          0                0                1        1         0
247         0                0                0        3         0
142         0                0                0        1         0
107         1                0                0        0         1
483         0                0                0        0         0
..        …              …              …       …        …
359         0                0                0        1         0
36          0                1                0        1         1
30          0                0                1        2         0
20          1                1                0        2         0
527         1                0                0        0         0

    furnishingstatus
62         furnished
247      unfurnished
142   semi-furnished
107      unfurnished
483   semi-furnished
..              …
359      unfurnished
36         furnished
30       unfurnished
20    semi-furnished
527   semi-furnished

[436 rows x 13 columns]
```

[ ]:
```
test
```

[ ]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
2	12250000	9960	3	2	2	1	0	1	0	0	2	1	semi-furnished
3	12215000	7500	4	2	2	1	0	1	0	1	3	1	furnished
7	10150000	16200	5	3	2	1	0	0	0	0	0	0	unfurnished
15	9100000	6000	4	1	2	1	0	1	0	0	2	0	semi-furnished
22	8645000	8050	3	1	1	1	1	1	0	1	1	0	furnished
..	…	…	…	…	…	…	…	…	…	…	…	…	…
508	2590000	4400	2	1	1	1	0	0	0	0	0	0	unfurnished
513	2485000	4400	3	1	2	1	0	0	0	0	0	0	unfurnished
520	2450000	7700	2	1	1	1	0	0	0	0	0	0	unfurnished
537	1890000	1700	3	1	2	1	0	0	0	0	0	0	unfurnished
539	1855000	2990	2	1	1	0	0	0	0	0	1	0	unfurnished

[109 rows x 13 columns]

[ ]:
```
# select specific columns for the training set
train = train[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
# select specific columns for the test set
```

```
test = test[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
```

[ ]:
```
scaler = preprocessing.StandardScaler()
```

[ ]:
```
train
```

[ ]:
```
       price   area  bedrooms  bathrooms  stories  parking  mainroad  \
62   7070000   6240         4          2        2        1         1
247  4550000   8400         4          1        4        3         1
142  5600000  10500         4          2        2        1         1
107  6125000   6420         3          1        3        0         1
483  2940000   6615         3          1        2        0         1
..       …     …       …         …       …      …       …
359  3710000   3600         3          1        1        1         1
36   8043000   7482         3          2        3        1         1
30   8400000   7475         3          2        4        2         1
20   8750000   4320         3          1        2        2         1
527  2275000   1836         2          1        1        0         0

     guestroom  basement  hotwaterheating  airconditioning  prefarea
62           0         0                0                1         0
247          0         0                0                0         0
142          0         0                0                0         0
107          0         1                0                0         1
483          0         0                0                0         0
..         …       …              …              …       …
359          0         0                0                0         0
36           0         0                1                0         1
30           0         0                0                1         0
20           0         1                1                0         0
527          0         1                0                0         0

[436 rows x 12 columns]
```

[ ]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

[ ]:
```
array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
        5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
        6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
        8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
        4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
```

```
 4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
 6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
 4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
 5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
 3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
 4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
 4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
 6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
 5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
 3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
 7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
 2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
 6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
 8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
 3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
 6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
 5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
 6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
 5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
 4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
 6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
 4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
 7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
 4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
 4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
 2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
 2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
 5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
 6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
 7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
 7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
 8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
 3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
 7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
 6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
 6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
 6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
 6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
 8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
 5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
 7482,  7475,  4320,  1836])
```

[ ]:
```
X2_t = np.array(train.bedrooms)
X3_t = np.array(train.bathrooms)
```

```
X4_t = np.array(train.stories)
X5_t = np.array(train.parking)
X6_t = np.array(train.mainroad)
X7_t = np.array(train.guestroom)
X8_t = np.array(train.hotwaterheating)
X9_t = np.array(train.airconditioning)
X10_t = np.array(train.prefarea)
X11_t = np.array(train.basement)
X0_t= np.ones(436)
```

[ ]:
```
X = np.vstack([X0_t,X1_t,X2_t,X3_t,X4_t,X5_t,X6_t,X7_t,X8_t,X9_t,X10_t,X11_t])
X = X.T
X = np.array(X)
X
```

[ ]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00]])
```

[ ]:
```
# scale the feature matrix using the fit_transform() method of the scaler object
X_scaled = scaler.fit_transform(X)

# assign the scaled feature matrix to the original variable name 'X'
X = X_scaled
```

[ ]:
```
# create a 1D NumPy array of zeros with length 12
theta = np.zeros(12)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (12,1))

# display the column vector
theta
```

[ ]:
```
array([[0.],
       [0.],
       [0.],
```

```
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

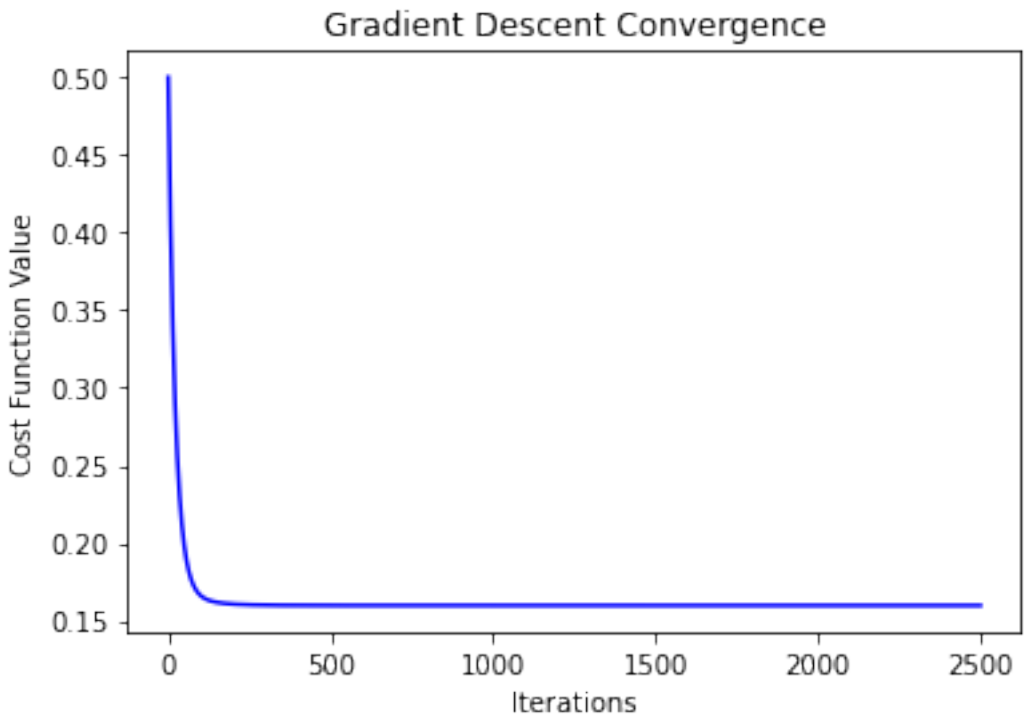
[ ]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
 ↪'train'
#h = np.matmul(X,theta)
Y_t = np.array(train.price)
Y = Y_t
Y = Y_t.reshape(436,1)
y = scaler.fit_transform(Y)
Y=y
```

[ ]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 2500

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  12
Shape of 'j' array:  (2500, 1)
```

[ ]:
```
# set the initial iteration count to zero
count = 0
```

```
# create a NumPy array 'j' to store the cost function values for each iteration
j = np.zeros(shape=(iterations, 1), dtype=float)

# perform gradient descent for the specified number of iterations
while count < iterations:

    # calculate the predicted values 'h' using the current parameters 'theta'
    h = X.dot(theta)

    # calculate the cost function value 'j' using the current parameters 'theta'
    j[count] = (1/(2*m)) * np.sum((h-Y)**2)

    # calculate the gradient of the cost function with respect to 'theta'
    grad = (1/m) * X_T.dot(h-Y)

    # update the parameters 'theta' using the learning rate 'alpha' and the␣
 ↪gradient 'grad'
    alpha = 0.01
    theta = theta - alpha * grad

    # increment the iteration count
    count += 1

# plot the cost function values over the iterations
plt.plot(j,'b-')
plt.xlabel('Iterations')
plt.ylabel('Cost Function Value')
plt.title('Gradient Descent Convergence')
plt.show()
```



[ ]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
N_X6_t = np.array(test.mainroad)
N_X7_t = np.array(test.guestroom)
N_X8_t = np.array(test.hotwaterheating)
N_X9_t = np.array(test.airconditioning)
N_X10_t = np.array(test.prefarea)
N_X11_t = np.array(test.basement)
N_X0_t = np.ones(109)
```

[ ]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t,␣
 ↪N_X6_t,N_X7_t,N_X8_t,N_X9_t,N_X10_t,N_X11_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[ ]: (109, 12)

[ ]:
```
# initialize the parameters for the test set
N_theta = np.array([0.,0.,0.,0.,0,0.,0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(12,1)
N_theta
```

[ ]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[ ]:
```
# initialize the target variable for the test set
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[ ]: (109, 1)

[ ]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
m,n
```

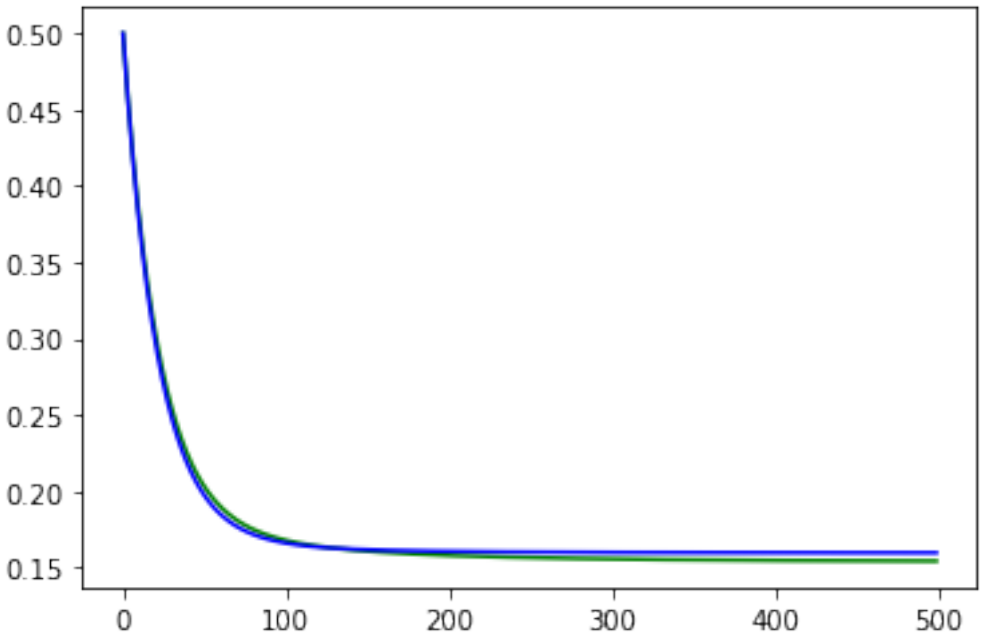
[ ]: (109, 12)

[ ]:
```
iterations = 2500
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)

while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)


    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)
```

```
    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta - 0.01*(grad)

    count += 1
```

[ ]:
```
plt.plot(N_j[:500],'g-')
plt.plot(j[:500],'b-')
```

[ ]:
```
[<matplotlib.lines.Line2D at 0x7f29d18e6df0>]
```



hw1-3a-normalization

February 20, 2023

[29]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing  # import scikit-learn library (source: https:
↪//scikit-learn.org/stable/index.html)
```

[30]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[30]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
0	13300000	7420	4	2	3	yes	no	no	
1	12250000	8960	4	4	4	yes	no	no	
2	12250000	9960	3	2	2	yes	no	yes	
3	12215000	7500	4	2	2	yes	no	yes	
4	11410000	7420	4	1	2	yes	yes	yes	
..	…	…	…	…	…	…	…	…	
540	1820000	3000	2	1	1	yes	no	yes	
541	1767150	2400	3	1	1	no	no	no	
542	1750000	3620	2	1	1	yes	no	no	
543	1750000	2910	3	1	1	no	no	no	
544	1750000	3850	3	1	2	yes	no	no	

	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	no	yes	2	yes	furnished
1	no	yes	3	no	furnished
2	no	no	2	yes	semi-furnished
3	no	yes	3	yes	furnished
4	no	yes	2	no	furnished
..	…	…	…	…	…
540	no	no	2	no	unfurnished
541	no	no	0	no	semi-furnished
542	no	no	0	no	unfurnished
543	no	no	0	no	furnished
544	no	no	0	no	unfurnished

```
[545 rows x 13 columns]
```

[31]:
```
df = df.replace(to_replace=['yes', 'no'], value=[1, 0])
df
```

[31]:
```
        price  area  bedrooms  bathrooms  stories  mainroad  guestroom  \
0    13300000  7420         4          2        3         1          0
1    12250000  8960         4          4        4         1          0
2    12250000  9960         3          2        2         1          0
3    12215000  7500         4          2        2         1          0
4    11410000  7420         4          1        2         1          1
..        ...   ...       ...        ...      ...       ...        ...
540   1820000  3000         2          1        1         1          0
541   1767150  2400         3          1        1         0          0
542   1750000  3620         2          1        1         1          0
543   1750000  2910         3          1        1         0          0
544   1750000  3850         3          1        2         1          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
0           0                0                1        2         1
1           0                0                1        3         0
2           1                0                0        2         1
3           1                0                1        3         1
4           1                0                1        2         0
..        ...              ...              ...      ...       ...
540         1                0                0        2         0
541         0                0                0        0         0
542         0                0                0        0         0
543         0                0                0        0         0
544         0                0                0        0         0

    furnishingstatus
0          furnished
1          furnished
2     semi-furnished
3          furnished
4          furnished
..               ...
540      unfurnished
541   semi-furnished
542      unfurnished
543        furnished
544      unfurnished

[545 rows x 13 columns]
```

[32]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)

# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[33]:
```
test
```

[33]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom
2	12250000	9960	3	2	2	1	0
3	12215000	7500	4	2	2	1	0
7	10150000	16200	5	3	2	1	0
15	9100000	6000	4	1	2	1	0
22	8645000	8050	3	1	1	1	1
..	…	…	…	…	…	…	…
508	2590000	4400	2	1	1	1	0
513	2485000	4400	3	1	2	1	0
520	2450000	7700	2	1	1	1	0
537	1890000	1700	3	1	2	1	0
539	1855000	2990	2	1	1	0	0

	basement	hotwaterheating	airconditioning	parking	prefarea
2	1	0	0	2	1
3	1	0	1	3	1
7	0	0	0	0	0
15	1	0	0	2	0
22	1	0	1	1	0
..	…	…	…	…	…
508	0	0	0	0	0
513	0	0	0	0	0
520	0	0	0	0	0
537	0	0	0	0	0
539	0	0	0	1	0

	furnishingstatus
2	semi-furnished
3	furnished
7	unfurnished
15	semi-furnished
22	furnished
..	…
508	unfurnished
513	unfurnished
520	unfurnished
537	unfurnished
539	unfurnished

[109 rows x 13 columns]

[34]:
```
# select specific columns for the training set
train = train[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]

# select specific columns for the test set
test = test[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]
```

[35]:
```
import sklearn.preprocessing # import scikit-learn library for data preprocessing

# create an instance of the MinMaxScaler class for scaling features to a range of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()
```

[36]:
```
train
```

[36]:

	price	area	bedrooms	bathrooms	stories	parking
62	7070000	6240	4	2	2	1
247	4550000	8400	4	1	4	3
142	5600000	10500	4	2	2	1
107	6125000	6420	3	1	3	0
483	2940000	6615	3	1	2	0
..	…	…	…	…	…	…
359	3710000	3600	3	1	1	1
36	8043000	7482	3	2	3	1
30	8400000	7475	3	2	4	2
20	8750000	4320	3	1	2	2
527	2275000	1836	2	1	1	0

[436 rows x 6 columns]

[37]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

[37]:
```
array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
        5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
        6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
        8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
        4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
        4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
        6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
        4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
        5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
```

```
 3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
 4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
 4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
 6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
 5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
 3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
 7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
 2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
 6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
 8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
 3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
 6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
 5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
 6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
 5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
 4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
 6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
 4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
 7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
 4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
 4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
 2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
 2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
 5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
 6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
 7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
 7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
 8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
 3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
 7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
 6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
 6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
 6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
 6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
 8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
 5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
 7482,  7475,  4320,  1836])
```

[38]:
```
X1_t = np.array(train['area']) # 'area' column
X2_t = np.array(train['bedrooms']) # 'bedrooms' column
X3_t = np.array(train['bathrooms']) # 'bathrooms' column
X4_t = np.array(train['stories']) # 'stories' column
X5_t = np.array(train['parking']) # 'parking' column
```

```
# create a NumPy array of ones to represent the bias term
X0_t = np.ones(len(train))
```

[39]:
```
# stack the selected feature arrays vertically using np.vstack
X = np.vstack([X0_t, X1_t, X2_t, X3_t, X4_t, X5_t])

# transpose the stacked array to make it a 6 x 436 matrix
X = X.T

# convert the stacked array to a NumPy array
X = np.array(X)

# display the NumPy array
X
```

[39]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, 1.000e+00, 4.000e+00, 3.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, 2.000e+00, 4.000e+00, 2.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, 1.000e+00, 2.000e+00, 2.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, 1.000e+00, 1.000e+00, 0.000e+00]])
```

[40]:
```
# scale the feature matrix using the fit_transform() method of the scaler object
X_scaled = scaler.fit_transform(X)

# assign the scaled feature matrix to the original variable name 'X'
X = X_scaled
```

[41]:
```
# create a 1D NumPy array of zeros with length 6
theta = np.zeros(6)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (6,1))

# display the column vector
theta
```

[41]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[42]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
↪'train'
```

```
Y_t = np.array(train.price)

# create a copy of 'Y_t' to prevent changing the original data
Y = Y_t.copy()

# reshape 'Y' to a column vector using np.reshape
Y = np.reshape(Y, (436,1))

# create an instance of the MinMaxScaler class for scaling the target variable␣
 ↪to a range of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()

# scale the target variable using the fit_transform() method of the scaler␣
 ↪object
Y_scaled = scaler.fit_transform(Y)

# assign the scaled target variable to the original variable name 'Y'
Y = Y_scaled
```

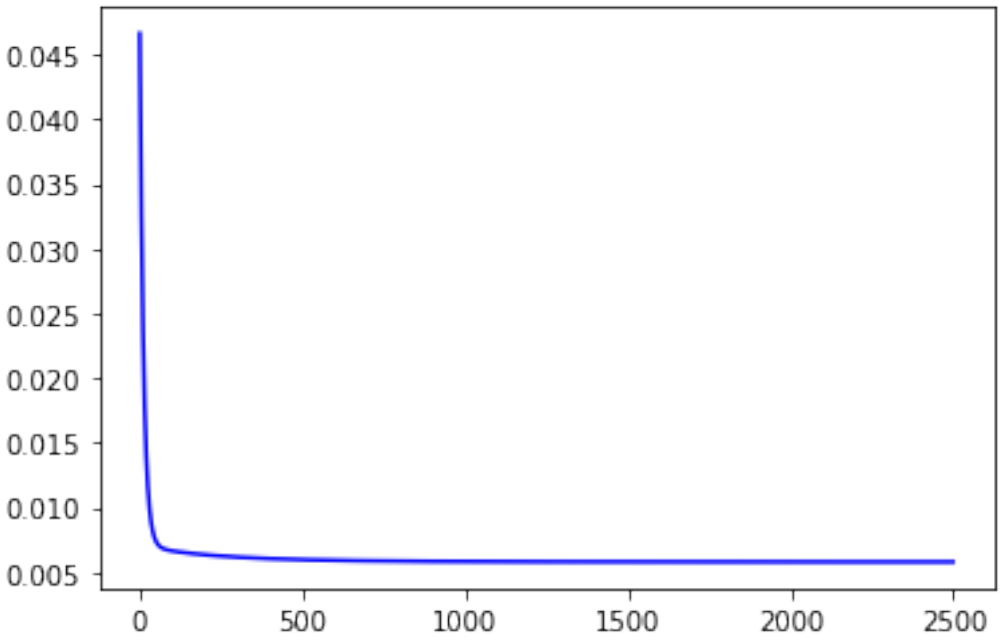
[43]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 2500

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  6
Shape of 'j' array:  (2500, 1)
```

[44]:
```
# Initialize variables
iterations = 2500
```

```
count = 0
alpha = 0.1
lambda_ = 0.001
j = np.zeros(iterations)

# Perform gradient descent
while count < iterations:
    h_t = X.dot(theta)
    h = np.array(h_t, float)

    j[count] = (1/(2*m)) * np.sum((h - Y)**2) + (lambda_ / (2*m)) * np.
 ↪sum(theta**2)

    grad_t = X_T.dot(h - Y)
    grad = grad_t * (1/m)

    theta = theta * (1 - alpha * (lambda_ / m)) - alpha * grad

    count += 1

# Plot the cost function
plt.plot(j, 'b-')
```

[44]: [<matplotlib.lines.Line2D at 0x7f6e719f9700>]



[45]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
N_X0_t = np.ones(109)
```

[46]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[46]:
```
(109, 6)
```

[47]:
```
N_theta = np.array([0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(6,1)
N_theta
```

[47]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[48]:
```
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[48]:
```
(109, 1)
```

[49]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
m,n
```

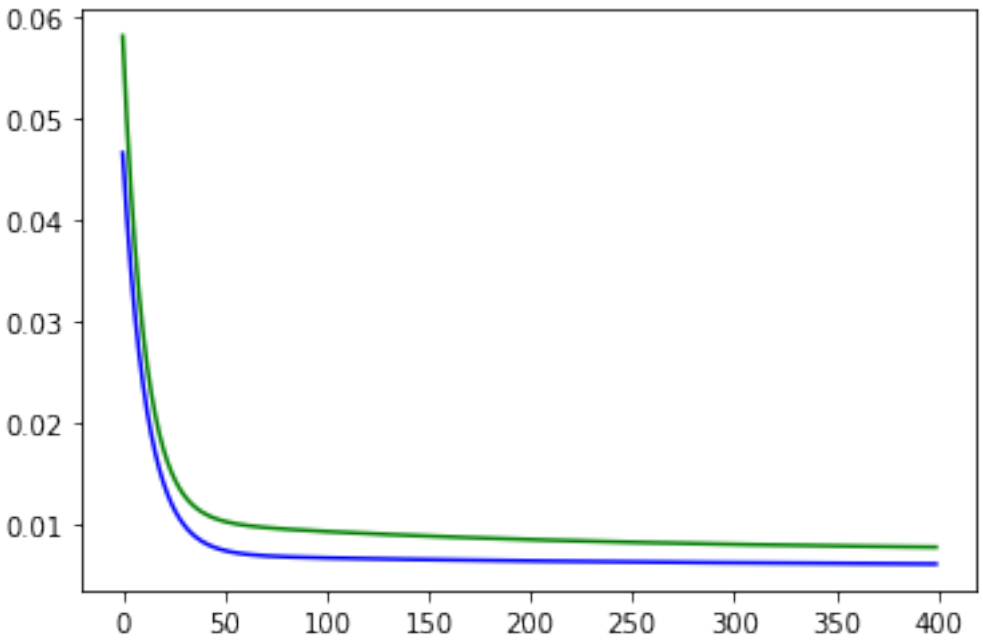
[49]:
```
(109, 6)
```

[50]:
```
iterations = 2500
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)
```

```
while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)


    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)+(0.001)*(1/(2*m))*np.
 ↪sum((N_theta)**2)

    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta*(1-0.1*(0.001/m)) - 0.1*(grad)

    count += 1
```

[51]:
```
plt.plot(N_j[:400],'g-')
plt.plot(j[:400],'b-')
```

[51]: [<matplotlib.lines.Line2D at 0x7f6e719beac0>]



hw1-3a-standardization

February 20, 2023

[1]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing
```

[2]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[2]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	13300000	7420	4	2	3	yes	no	no	no	yes	2	yes	furnished
1	12250000	8960	4	4	4	yes	no	no	no	yes	3	no	furnished
2	12250000	9960	3	2	2	yes	no	yes	no	no	2	yes	semi-furnished
3	12215000	7500	4	2	2	yes	no	yes	no	yes	3	yes	furnished
4	11410000	7420	4	1	2	yes	yes	yes	no	yes	2	no	furnished
..	…	…	…	…	…	…	…	…	…	…	…	…	…
540	1820000	3000	2	1	1	yes	no	yes	no	no	2	no	unfurnished
541	1767150	2400	3	1	1	no	no	no	no	no	0	no	semi-furnished
542	1750000	3620	2	1	1	yes	no	no	no	no	0	no	unfurnished
543	1750000	2910	3	1	1	no	no	no	no	no	0	no	furnished
544	1750000	3850	3	1	2	yes	no	no	no	no	0	no	unfurnished

[545 rows x 13 columns]

[3]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)

# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[4]:
```
train
```

[4]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
62	7070000	6240	4	2	2	yes	no	no	
247	4550000	8400	4	1	4	yes	no	no	
142	5600000	10500	4	2	2	yes	no	no	
107	6125000	6420	3	1	3	yes	no	yes	
483	2940000	6615	3	1	2	yes	no	no	
..	…	…	…	…	…	…	…	…	
359	3710000	3600	3	1	1	yes	no	no	
36	8043000	7482	3	2	3	yes	no	no	
30	8400000	7475	3	2	4	yes	no	no	
20	8750000	4320	3	1	2	yes	no	yes	
527	2275000	1836	2	1	1	no	no	yes	

	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
62	no	yes	1	no	furnished
247	no	no	3	no	unfurnished
142	no	no	1	no	semi-furnished
107	no	no	0	yes	unfurnished
483	no	no	0	no	semi-furnished
..	…	…	…	…	…
359	no	no	1	no	unfurnished
36	yes	no	1	yes	furnished
30	no	yes	2	no	unfurnished
20	yes	no	2	no	semi-furnished
527	no	no	0	no	semi-furnished

[436 rows x 13 columns]

[5]:
```
test
```

[5]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	\
2	12250000	9960	3	2	2	yes	no	
3	12215000	7500	4	2	2	yes	no	
7	10150000	16200	5	3	2	yes	no	
15	9100000	6000	4	1	2	yes	no	
22	8645000	8050	3	1	1	yes	yes	
..	…	…	…	…	…	…	…	
508	2590000	4400	2	1	1	yes	no	
513	2485000	4400	3	1	2	yes	no	

```
520   2450000   7700         2          1        1      yes        no
537   1890000   1700         3          1        2      yes        no
539   1855000   2990         2          1        1       no        no

    basement hotwaterheating airconditioning  parking prefarea  \
2        yes              no              no        2      yes
3        yes              no             yes        3      yes
7         no              no              no        0       no
15       yes              no              no        2       no
22       yes              no             yes        1       no
..       …              …               …       …       …
508       no              no              no        0       no
513       no              no              no        0       no
520       no              no              no        0       no
537       no              no              no        0       no
539       no              no              no        1       no

    furnishingstatus
2     semi-furnished
3          furnished
7        unfurnished
15    semi-furnished
22         furnished
..               …
508      unfurnished
513      unfurnished
520      unfurnished
537      unfurnished
539      unfurnished

[109 rows x 13 columns]
```

[6]:
```
# select specific columns for the training set
train = train[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]

# select specific columns for the test set
test = test[['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']]
```

[7]:
```
scaler = preprocessing.StandardScaler()
```

[8]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

```
[8]: array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
             5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
             6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
             8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
             4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
             4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
             6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
             4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
             5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
             3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
             4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
             4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
             6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
             5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
            10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
             3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
             7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
             2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
             6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
             8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
             3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
             6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
             5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
             6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
             5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
             4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
             6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
             4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
             7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
            10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
             4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
             4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
             2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
             2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
             5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
             6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
             7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
             7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
             8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
             3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
             7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
             6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
             6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
             6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
             6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
             8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
             5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
```

```
        13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
         7482,  7475,  4320,  1836])
```

[9]:
```
X1_t = np.array(train['area']) # 'area' column
X2_t = np.array(train['bedrooms']) # 'bedrooms' column
X3_t = np.array(train['bathrooms']) # 'bathrooms' column
X4_t = np.array(train['stories']) # 'stories' column
X5_t = np.array(train['parking']) # 'parking' column

# create a NumPy array of ones to represent the bias term
X0_t = np.ones(len(train))
```

[10]:
```
# stack the selected feature arrays vertically using np.vstack
X = np.vstack([X0_t, X1_t, X2_t, X3_t, X4_t, X5_t])

# transpose the stacked array to make it a 6 x 436 matrix
X = X.T

# convert the stacked array to a NumPy array
X = np.array(X)

# display the NumPy array
X
```

[10]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, 1.000e+00, 4.000e+00, 3.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, 2.000e+00, 2.000e+00, 1.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, 2.000e+00, 4.000e+00, 2.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, 1.000e+00, 2.000e+00, 2.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, 1.000e+00, 1.000e+00, 0.000e+00]])
```

[11]:
```
x = scaler.fit_transform(X)
X = x
```

[12]:
```
# create a 1D NumPy array of zeros with length 6
theta = np.zeros(6)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (6,1))

# display the column vector
theta
```

[12]:
```
array([[0.],
       [0.],
       [0.],
```

```
       [0.],
       [0.],
       [0.]])
```

[13]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
 ↪'train'
#h = np.matmul(X,theta)
Y_t = np.array(train.price)
Y = Y_t
Y = Y_t.reshape(436,1)
y = scaler.fit_transform(Y)
Y=y
```

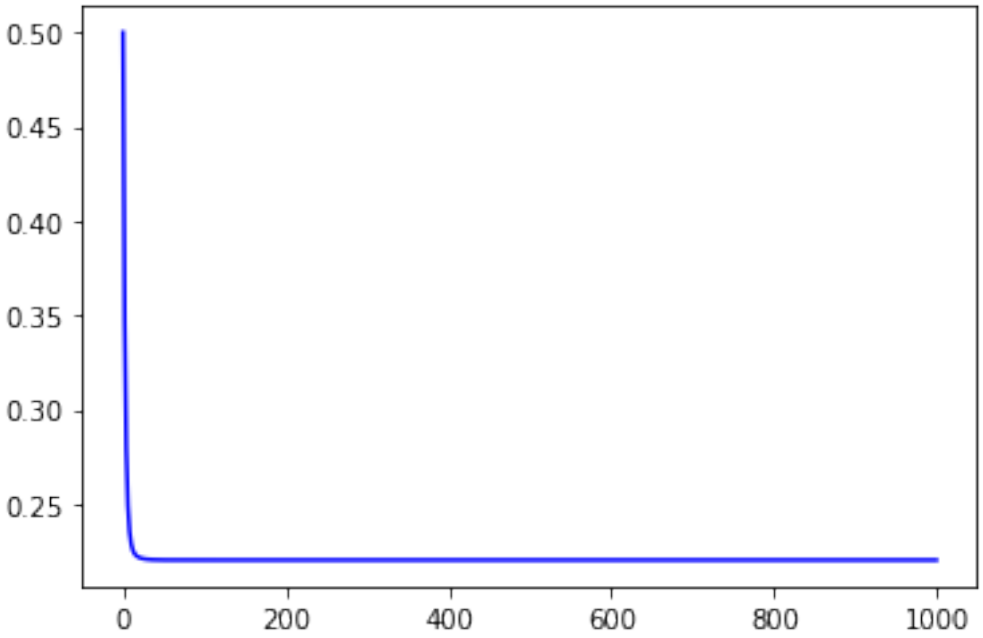
[14]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 1000

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  6
Shape of 'j' array:  (1000, 1)
```

[15]:
```
# Initialize variables
iterations = 1000
count = 0
alpha = 0.1
lambda_ = 0.001
j = np.zeros(iterations)

# Perform gradient descent
```

```
while count < iterations:
    h_t = X.dot(theta)
    h = np.array(h_t, float)

    j[count] = (1/(2*m)) * np.sum((h - Y)**2) + (lambda_ / (2*m)) * np.
 ↪sum(theta**2)

    grad_t = X_T.dot(h - Y)
    grad = grad_t * (1/m)

    theta = theta * (1 - alpha * (lambda_ / m)) - alpha * grad

    count += 1

# Plot the cost function
plt.plot(j, 'b-')
```

[15]: [<matplotlib.lines.Line2D at 0x7f153537e520>]



[16]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
```

```
N_X0_t = np.ones(109)
```

[17]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[17]:
```
(109, 6)
```

[18]:
```
N_theta = np.array([0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(6,1)
N_theta
```

[18]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[19]:
```
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[19]:
```
(109, 1)
```

[20]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
m,n
```

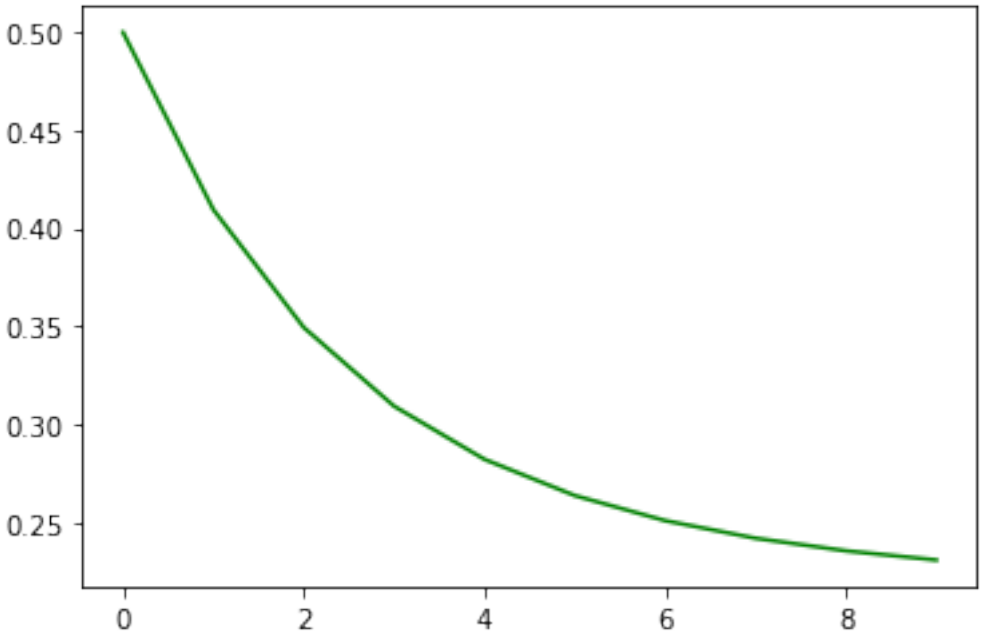
[20]:
```
(109, 6)
```

[22]:
```
iterations = 2500
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)

while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)
```

```
    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)+(0.001)*(1/(2*m))*np.
↪sum((N_theta)**2)

    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta*(1-0.1*(0.001/m)) - 0.1*(grad)

    count += 1
```

[25]:
```
plt.plot(N_j[:10],'g-')
#plt.plot(j[:10],'b-')
```

[25]:
```
[<matplotlib.lines.Line2D at 0x7f1534d59070>]
```



hw1-3b-normalization

February 20, 2023

[100]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing  # import scikit-learn library (source: https:
↪//scikit-learn.org/stable/index.html)
```

[101]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[101]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
0	13300000	7420	4	2	3	yes	no	no	
1	12250000	8960	4	4	4	yes	no	no	
2	12250000	9960	3	2	2	yes	no	yes	
3	12215000	7500	4	2	2	yes	no	yes	
4	11410000	7420	4	1	2	yes	yes	yes	
..	…	…	…	…	…	…	…	…	
540	1820000	3000	2	1	1	yes	no	yes	
541	1767150	2400	3	1	1	no	no	no	
542	1750000	3620	2	1	1	yes	no	no	
543	1750000	2910	3	1	1	no	no	no	
544	1750000	3850	3	1	2	yes	no	no	

	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	no	yes	2	yes	furnished
1	no	yes	3	no	furnished
2	no	no	2	yes	semi-furnished
3	no	yes	3	yes	furnished
4	no	yes	2	no	furnished
..	…	…	…	…	…
540	no	no	2	no	unfurnished
541	no	no	0	no	semi-furnished
542	no	no	0	no	unfurnished
543	no	no	0	no	furnished
544	no	no	0	no	unfurnished

```
[545 rows x 13 columns]
```

[102]:
```
df = df.replace(to_replace=['yes', 'no'], value=[1, 0])
df
```

[102]:
```
        price  area  bedrooms  bathrooms  stories  mainroad  guestroom  \
0    13300000  7420         4          2        3         1          0
1    12250000  8960         4          4        4         1          0
2    12250000  9960         3          2        2         1          0
3    12215000  7500         4          2        2         1          0
4    11410000  7420         4          1        2         1          1
..        ...   ...       ...        ...      ...       ...        ...
540   1820000  3000         2          1        1         1          0
541   1767150  2400         3          1        1         0          0
542   1750000  3620         2          1        1         1          0
543   1750000  2910         3          1        1         0          0
544   1750000  3850         3          1        2         1          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
0           0                0                1        2         1
1           0                0                1        3         0
2           1                0                0        2         1
3           1                0                1        3         1
4           1                0                1        2         0
..        ...              ...              ...      ...       ...
540         1                0                0        2         0
541         0                0                0        0         0
542         0                0                0        0         0
543         0                0                0        0         0
544         0                0                0        0         0

    furnishingstatus
0          furnished
1          furnished
2     semi-furnished
3          furnished
4          furnished
..               ...
540      unfurnished
541   semi-furnished
542      unfurnished
543        furnished
544      unfurnished

[545 rows x 13 columns]
```

[103]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)

# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[104]:
```
train
```

[104]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom
62	7070000	6240	4	2	2	1	0
247	4550000	8400	4	1	4	1	0
142	5600000	10500	4	2	2	1	0
107	6125000	6420	3	1	3	1	0
483	2940000	6615	3	1	2	1	0
..	…	…	…	…	…	…	…
359	3710000	3600	3	1	1	1	0
36	8043000	7482	3	2	3	1	0
30	8400000	7475	3	2	4	1	0
20	8750000	4320	3	1	2	1	0
527	2275000	1836	2	1	1	0	0

	basement	hotwaterheating	airconditioning	parking	prefarea
62	0	0	1	1	0
247	0	0	0	3	0
142	0	0	0	1	0
107	1	0	0	0	1
483	0	0	0	0	0
..	…	…	…	…	…
359	0	0	0	1	0
36	0	1	0	1	1
30	0	0	1	2	0
20	1	1	0	2	0
527	1	0	0	0	0

	furnishingstatus
62	furnished
247	unfurnished
142	semi-furnished
107	unfurnished
483	semi-furnished
..	…
359	unfurnished
36	furnished
30	unfurnished
20	semi-furnished
527	semi-furnished

```
[436 rows x 13 columns]
```

[105]:
```
test
```

[105]:
```
        price   area  bedrooms  bathrooms  stories  mainroad  guestroom  \
2    12250000   9960         3          2        2         1          0
3    12215000   7500         4          2        2         1          0
7    10150000  16200         5          3        2         1          0
15    9100000   6000         4          1        2         1          0
22    8645000   8050         3          1        1         1          1
..        …    …        …         …       …        …         …
508   2590000   4400         2          1        1         1          0
513   2485000   4400         3          1        2         1          0
520   2450000   7700         2          1        1         1          0
537   1890000   1700         3          1        2         1          0
539   1855000   2990         2          1        1         0          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
2           1                0                0        2         1
3           1                0                1        3         1
7           0                0                0        0         0
15          1                0                0        2         0
22          1                0                1        1         0
..        …                …                …       …        …
508         0                0                0        0         0
513         0                0                0        0         0
520         0                0                0        0         0
537         0                0                0        0         0
539         0                0                0        1         0

    furnishingstatus
2     semi-furnished
3          furnished
7        unfurnished
15    semi-furnished
22         furnished
..               …
508      unfurnished
513      unfurnished
520      unfurnished
537      unfurnished
539      unfurnished

[109 rows x 13 columns]
```

[106]:
```
# select specific columns for the training set
```

```
train = train[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
# select specific columns for the test set
test = test[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
```

[107]:
```
import sklearn.preprocessing # import scikit-learn library for data␣
 ↪preprocessing

# create an instance of the MinMaxScaler class for scaling features to a range␣
 ↪of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()
```

[108]:
```
train
```

[108]:

	price	area	bedrooms	bathrooms	stories	parking	mainroad	guestroom	basement	hotwaterheating	airconditioning	prefarea
62	7070000	6240	4	2	2	1	1	0	0	0	1	0
247	4550000	8400	4	1	4	3	1	0	0	0	0	0
142	5600000	10500	4	2	2	1	1	0	0	0	0	0
107	6125000	6420	3	1	3	0	1	0	1	0	0	1
483	2940000	6615	3	1	2	0	1	0	0	0	0	0
..	…	…	…	…	…	…	…	…	…	…	…	…
359	3710000	3600	3	1	1	1	1	0	0	0	0	0
36	8043000	7482	3	2	3	1	1	0	0	1	0	1
30	8400000	7475	3	2	4	2	1	0	0	0	1	0
20	8750000	4320	3	1	2	2	1	0	1	1	0	0
527	2275000	1836	2	1	1	0	0	0	1	0	0	0

[436 rows x 12 columns]

[109]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

[109]:
```
array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
        5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
        6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
        8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
        4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
        4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
        6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
        4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
        5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
        3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
        4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
        4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
        6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
        5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
       10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
        3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
        7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
        2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
        6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
        8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
        3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
        6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
        5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
        6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
        5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
        4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
        6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
        4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
        7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
       10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
        4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
        4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
        2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
        2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
        5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
        6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
        7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
        7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
        8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
        3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
        7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
```

```
 6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
 6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
 6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
 6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
 8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
 5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
 7482,  7475,  4320,  1836])
```

[110]:
```
X2_t = np.array(train.bedrooms)
X3_t = np.array(train.bathrooms)
X4_t = np.array(train.stories)
X5_t = np.array(train.parking)
X6_t = np.array(train.mainroad)
X7_t = np.array(train.guestroom)
X8_t = np.array(train.hotwaterheating)
X9_t = np.array(train.airconditioning)
X10_t = np.array(train.prefarea)
X11_t = np.array(train.basement)
X0_t= np.ones(436)
```

[111]:
```
# stack the selected feature arrays vertically using np.vstack
X = np.vstack([X0_t, X1_t, X2_t, X3_t, X4_t, X5_t,␣
 ↪X6_t,X7_t,X8_t,X9_t,X10_t,X11_t])

# transpose the stacked array to make it a 6 x 436 matrix
X = X.T

# convert the stacked array to a NumPy array
X = np.array(X)

# display the NumPy array
X
```

[111]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00]])
```

[112]:
```
# scale the feature matrix using the fit_transform() method of the scaler object
X_scaled = scaler.fit_transform(X)

# assign the scaled feature matrix to the original variable name 'X'
X = X_scaled
```

[113]:
```
# create a 1D NumPy array of zeros with length 12
theta = np.zeros(12)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (12,1))

# display the column vector
theta
```

[113]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[114]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
 ↪'train'
Y_t = np.array(train.price)

# create a copy of 'Y_t' to prevent changing the original data
Y = Y_t.copy()

# reshape 'Y' to a column vector using np.reshape
Y = np.reshape(Y, (436,1))

# create an instance of the MinMaxScaler class for scaling the target variable␣
 ↪to a range of [0, 1]
scaler = sklearn.preprocessing.MinMaxScaler()

# scale the target variable using the fit_transform() method of the scaler␣
 ↪object
Y_scaled = scaler.fit_transform(Y)

# assign the scaled target variable to the original variable name 'Y'
```

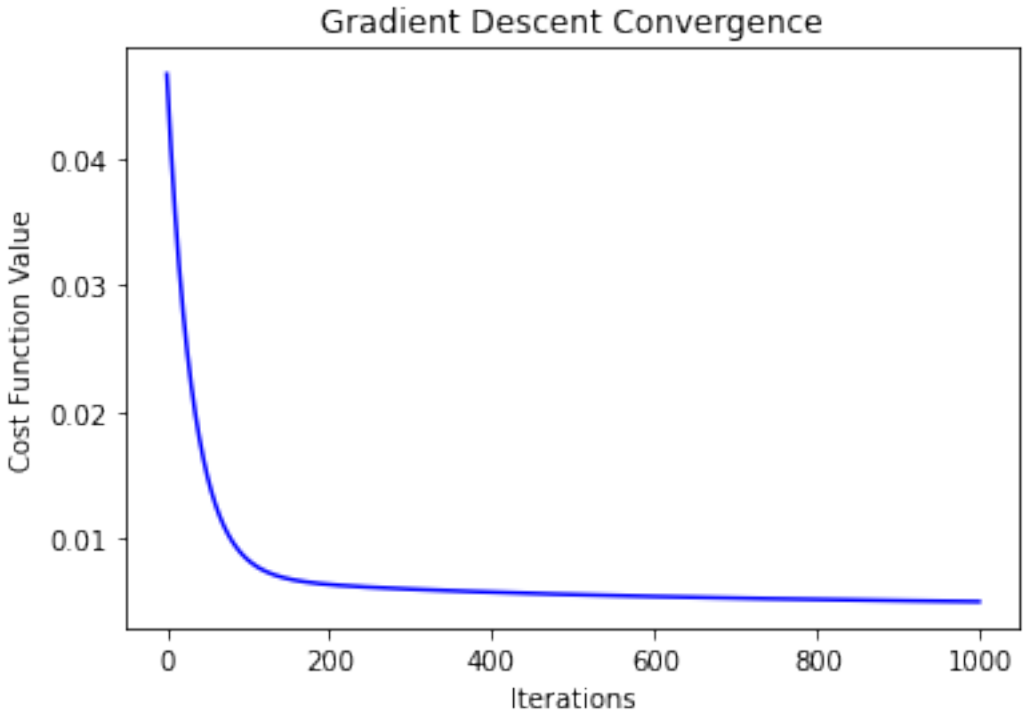
```
Y = Y_scaled
```

[115]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 1000

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  12
Shape of 'j' array:  (1000, 1)
```

[116]:
```
# set the initial iteration count to zero
count = 0

# create a NumPy array 'j' to store the cost function values for each iteration
j = np.zeros(shape=(iterations, 1), dtype=float)

# perform gradient descent for the specified number of iterations
while count < iterations:

    # calculate the predicted values 'h' using the current parameters 'theta'
    h = X.dot(theta)

    # calculate the cost function value 'j' using the current parameters 'theta'
    j[count] = (1/(2*m)) * np.sum((h-Y)**2)

    # calculate the gradient of the cost function with respect to 'theta'
    grad = (1/m) * X_T.dot(h-Y)
```

```
    # update the parameters 'theta' using the learning rate 'alpha' and the␣
 ↪gradient 'grad'
    alpha = 0.01
    theta = theta - alpha * grad

    # increment the iteration count
    count += 1

# plot the cost function values over the iterations
plt.plot(j,'b-')
plt.xlabel('Iterations')
plt.ylabel('Cost Function Value')
plt.title('Gradient Descent Convergence')
plt.show()
```



[117]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
N_X6_t = np.array(test.mainroad)
N_X7_t = np.array(test.guestroom)
```

```
N_X8_t = np.array(test.hotwaterheating)
N_X9_t = np.array(test.airconditioning)
N_X10_t = np.array(test.prefarea)
N_X11_t = np.array(test.basement)
N_X0_t = np.ones(109)
```

[118]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t,␣
 ↪N_X6_t,N_X7_t,N_X8_t,N_X9_t,N_X10_t,N_X11_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[118]:
```
(109, 12)
```

[119]:
```
# initialize the parameters for the test set
N_theta = np.array([0.,0.,0.,0.,0,0.,0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(12,1)
N_theta
```

[119]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[120]:
```
# initialize the target variable for the test set
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[120]:
```
(109, 1)
```

[121]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
```

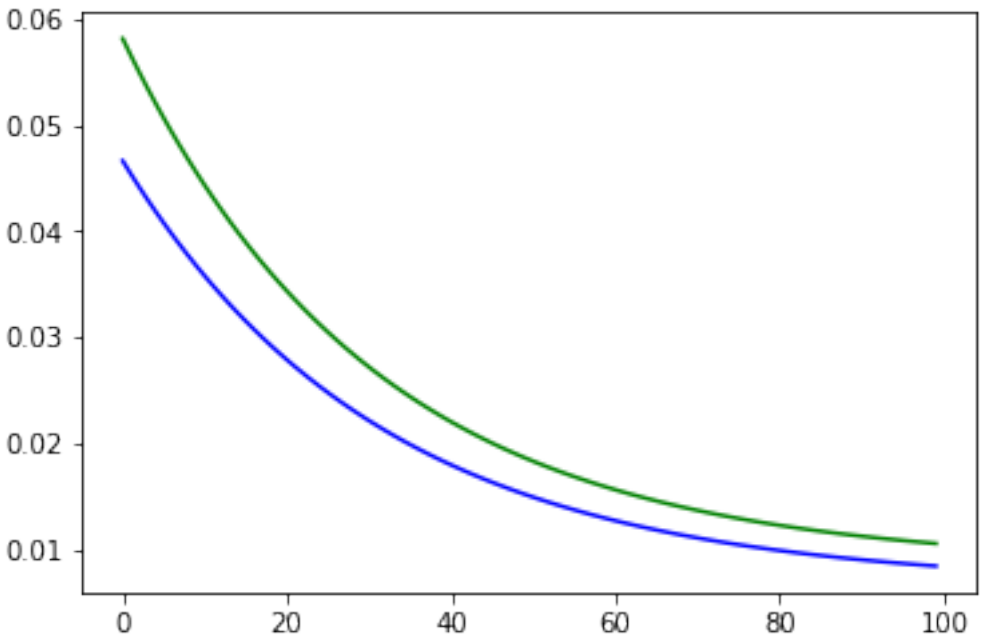
```
m,n
```

[121]: (109, 12)

[122]:
```
iterations = 1000
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)

while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)


    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)

    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta - 0.01*(grad)

    count += 1
```

[127]:
```
plt.plot(N_j[:100],'g-')
plt.plot(j[:100],'b-')
```

[127]: [<matplotlib.lines.Line2D at 0x7faee47df520>]



hw1-3b-standardization

February 20, 2023

[1]:
```
import numpy as np # import numpy library
import pandas as pd # import pandas library
import matplotlib.pyplot as plt # import matplotlib library
from sklearn import preprocessing
```

[2]:
```
df = pd.read_csv("/content/sample_data/Housing.csv")

# display DataFrame
df
```

[2]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	13300000	7420	4	2	3	yes	no	no	no	yes	2	yes	furnished
1	12250000	8960	4	4	4	yes	no	no	no	yes	3	no	furnished
2	12250000	9960	3	2	2	yes	no	yes	no	no	2	yes	semi-furnished
3	12215000	7500	4	2	2	yes	no	yes	no	yes	3	yes	furnished
4	11410000	7420	4	1	2	yes	yes	yes	no	yes	2	no	furnished
..	…	…	…	…	…	…	…	…	…	…	…	…	…
540	1820000	3000	2	1	1	yes	no	yes	no	no	2	no	unfurnished
541	1767150	2400	3	1	1	no	no	no	no	no	0	no	semi-furnished
542	1750000	3620	2	1	1	yes	no	no	no	no	0	no	unfurnished
543	1750000	2910	3	1	1	no	no	no	no	no	0	no	furnished
544	1750000	3850	3	1	2	yes	no	no	no	no	0	no	unfurnished

[545 rows x 13 columns]

[3]:
```
df = df.replace(to_replace=['yes', 'no'], value=[1, 0])
df
```

[3]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	13300000	7420	4	2	3	1	0	0	0	1	2	1	furnished
1	12250000	8960	4	4	4	1	0	0	0	1	3	0	furnished
2	12250000	9960	3	2	2	1	0	1	0	0	2	1	semi-furnished
3	12215000	7500	4	2	2	1	0	1	0	1	3	1	furnished
4	11410000	7420	4	1	2	1	1	1	0	1	2	0	furnished
..	…	…	…	…	…	…	…	…	…	…	…	…	…
540	1820000	3000	2	1	1	1	0	1	0	0	2	0	unfurnished
541	1767150	2400	3	1	1	0	0	0	0	0	0	0	semi-furnished
542	1750000	3620	2	1	1	1	0	0	0	0	0	0	unfurnished
543	1750000	2910	3	1	1	0	0	0	0	0	0	0	furnished
544	1750000	3850	3	1	2	1	0	0	0	0	0	0	unfurnished

[545 rows x 13 columns]

[4]:
```
# create a training set by randomly selecting 80% of the rows from the DataFrame
train = df.sample(frac=0.8, random_state=1)
```

```
# create a test set by dropping the rows in the training set from the DataFrame
test = df.drop(train.index)
```

[5]:
```
train
```

[5]:
```
       price   area  bedrooms  bathrooms  stories  mainroad  guestroom  \
62   7070000   6240         4          2        2         1          0
247  4550000   8400         4          1        4         1          0
142  5600000  10500         4          2        2         1          0
107  6125000   6420         3          1        3         1          0
483  2940000   6615         3          1        2         1          0
..       …     …       …         …       …        …         …
359  3710000   3600         3          1        1         1          0
36   8043000   7482         3          2        3         1          0
30   8400000   7475         3          2        4         1          0
20   8750000   4320         3          1        2         1          0
527  2275000   1836         2          1        1         0          0

     basement  hotwaterheating  airconditioning  parking  prefarea  \
62          0                0                1        1         0
247         0                0                0        3         0
142         0                0                0        1         0
107         1                0                0        0         1
483         0                0                0        0         0
..        …              …              …       …        …
359         0                0                0        1         0
36          0                1                0        1         1
30          0                0                1        2         0
20          1                1                0        2         0
527         1                0                0        0         0

    furnishingstatus
62         furnished
247      unfurnished
142   semi-furnished
107      unfurnished
483   semi-furnished
..              …
359      unfurnished
36         furnished
30       unfurnished
20    semi-furnished
527   semi-furnished

[436 rows x 13 columns]
```

[6]:
```
test
```

[6]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom
2	12250000	9960	3	2	2	1	0
3	12215000	7500	4	2	2	1	0
7	10150000	16200	5	3	2	1	0
15	9100000	6000	4	1	2	1	0
22	8645000	8050	3	1	1	1	1
..	…	…	…	…	…	…	…
508	2590000	4400	2	1	1	1	0
513	2485000	4400	3	1	2	1	0
520	2450000	7700	2	1	1	1	0
537	1890000	1700	3	1	2	1	0
539	1855000	2990	2	1	1	0	0

	basement	hotwaterheating	airconditioning	parking	prefarea
2	1	0	0	2	1
3	1	0	1	3	1
7	0	0	0	0	0
15	1	0	0	2	0
22	1	0	1	1	0
..	…	…	…	…	…
508	0	0	0	0	0
513	0	0	0	0	0
520	0	0	0	0	0
537	0	0	0	0	0
539	0	0	0	1	0

	furnishingstatus
2	semi-furnished
3	furnished
7	unfurnished
15	semi-furnished
22	furnished
..	…
508	unfurnished
513	unfurnished
520	unfurnished
537	unfurnished
539	unfurnished

[109 rows x 13 columns]

[7]:
```
# select specific columns for the training set
train = train[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
# select specific columns for the test set
```

```
test = test[['price','area','bedrooms','bathrooms','stories','parking',␣
 ↪'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',␣
 ↪'prefarea']]
```

[8]:
```
scaler = preprocessing.StandardScaler()
```

[9]:
```
train
```

[9]:
```
       price   area  bedrooms  bathrooms  stories  parking  mainroad  \
62   7070000   6240         4          2        2        1         1
247  4550000   8400         4          1        4        3         1
142  5600000  10500         4          2        2        1         1
107  6125000   6420         3          1        3        0         1
483  2940000   6615         3          1        2        0         1
..       …     …       …         …       …      …       …
359  3710000   3600         3          1        1        1         1
36   8043000   7482         3          2        3        1         1
30   8400000   7475         3          2        4        2         1
20   8750000   4320         3          1        2        2         1
527  2275000   1836         2          1        1        0         0

     guestroom  basement  hotwaterheating  airconditioning  prefarea
62           0         0                0                1         0
247          0         0                0                0         0
142          0         0                0                0         0
107          0         1                0                0         1
483          0         0                0                0         0
..         …       …              …              …       …
359          0         0                0                0         0
36           0         0                1                0         1
30           0         0                0                1         0
20           0         1                1                0         0
527          0         1                0                0         0

[436 rows x 12 columns]
```

[10]:
```
# create a NumPy array of the 'area' column from the training set
X1_t = np.array(train['area'])

# display the NumPy array
X1_t
```

[10]:
```
array([ 6240,  8400, 10500,  6420,  6615,  3600,  3240,  6600,  2700,
        5000,  2650,  4775,  4800,  3700,  7700,  7420,  4280,  6000,
        6600,  3649,  3420,  5500,  3630,  3180,  3600,  8400,  3000,
        8880,  5750,  2145,  6360,  6525,  1950,  5850,  8372,  2870,
        4990,  2684,  5200,  6321,  4960,  3480,  3210,  4950,  6840,
```

```
 4350,  5850,  4410,  2500,  3850,  3180,  3162,  3500,  4340,
 6440,  5010,  3000,  4920,  3760,  3816,  6000,  7000,  3640,
 4080,  4160,  2910,  6060,  3000,  2787,  4815,  4785,  6600,
 5300,  3600,  6000,  2176,  3000,  7420,  7020,  3480,  5960,
 3510,  6420,  6450,  6210,  4500,  3000,  3180,  5700,  3520,
 4040,  5800,  2800,  6480,  4960,  4260,  7500,  5880, 10500,
 4500,  3850,  8500,  3120,  3990,  4095,  4800, 13200,  7770,
 6100,  4075,  6550,  4100,  4370,  3180,  7350,  3510,  3640,
 5500,  8250,  6600,  8250,  2475,  3850,  4500,  3720,  4360,
10240,  5500,  3970,  3450,  3850,  5500,  3520,  2145,  6600,
 3640,  3986,  2953,  8250,  4130,  8580,  6000,  3500,  5885,
 7680,  2430,  3150,  6450,  8100,  5500,  1650,  3040,  4079,
 2747,  4600,  2325,  7231,  3520,  2145,  3450,  3620,  4000,
 6000,  6000,  4500,  3540,  7200,  3120,  4000,  2015,  4040,
 8000,  2787,  3512,  3420,  6060,  4500,  6360,  5450,  8250,
 3960,  7410, 10360,  3630,  6020,  4100,  6254,  4500,  4560,
 6710,  3500,  8880,  3600,  7152,  6000,  4040,  4000,  4040,
 5360,  6600,  3800,  3960,  4900,  3480,  3584,  2275,  4000,
 6500, 10500,  8960,  3290,  8875,  8580,  3450,  6600,  2800,
 5640,  3745, 10269,  6100, 12090,  5880,  6750,  6000,  5320,
 4000,  4040, 15600,  3090,  3970,  5450,  4770,  4095,  6000,
 6540,  6550,  4320,  3100,  4050,  3650,  3850,  5600,  2817,
 4510,  3000,  4995, 11410,  3000,  4840,  3600,  4000,  3500,
 7800,  5300,  4840,  3000,  3480,  2970,  5828,  3800,  4040,
10700,  7320,  5000,  6325,  2880,  4300,  3150,  4000,  9500,
 4500,  3420,  3180,  2145,  5400,  3630,  6750,  4820,  5136,
 4120,  6825,  4600,  6650,  5800,  5720,  5000,  4352,  3300,
 2160,  5900,  3000,  4500,  3350,  5400,  4600,  9800,  3630,
 2610,  9667,  3635,  4000,  3180,  3630,  6600,  2610,  4960,
 5150,  6000,  3640,  2910,  3650,  3450,  4032,  7980,  1905,
 6000,  3360,  9620,  1950, 12900,  3240,  4320,  6540,  6000,
 7440,  3760,  8100,  4880,  6000,  2000,  5200,  4050,  9166,
 7950,  5500,  2700,  6000,  6900,  3500,  5076,  5985,  4300,
 8050,  5320,  5960,  7000,  7260,  6360,  3000,  3460, 12944,
 3880,  2400,  4080,  6000,  4500,  6050,  7000,  3930,  4600,
 7155,  4100,  2400,  3460,  4632,  4200,  4640,  8800,  3000,
 6300,  7000,  7000,  6900,  3420,  3264,  2640,  3150,  4320,
 6862, 11440,  4992,  3069,  3185,  3750,  5300,  7200,  6400,
 6800,  3400,  6420,  3792,  5500,  4600,  6800,  6000,  8520,
 6480,  8150,  5948,  3185,  5830,  3410,  3000,  8400,  6350,
 8100,  4800,  2856,  3185,  3780,  3640,  6000,  6000,  4800,
 5800,  6360,  4120,  5400,  2850,  5400,  2145,  4500,  3240,
13200,  3900,  9000,  4646,  3840,  9000,  3520,  3640,  3600,
 7482,  7475,  4320,  1836])
```

[11]:
```
X2_t = np.array(train.bedrooms)
X3_t = np.array(train.bathrooms)
```

```
X4_t = np.array(train.stories)
X5_t = np.array(train.parking)
X6_t = np.array(train.mainroad)
X7_t = np.array(train.guestroom)
X8_t = np.array(train.hotwaterheating)
X9_t = np.array(train.airconditioning)
X10_t = np.array(train.prefarea)
X11_t = np.array(train.basement)
X0_t= np.ones(436)
```

[12]:
```
X = np.vstack([X0_t,X1_t,X2_t,X3_t,X4_t,X5_t,X6_t,X7_t,X8_t,X9_t,X10_t,X11_t])
X = X.T
X = np.array(X)
X
```

[12]:
```
array([[1.000e+00, 6.240e+03, 4.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 8.400e+03, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 1.050e+04, 4.000e+00, …, 0.000e+00, 0.000e+00,
        0.000e+00],
       …,
       [1.000e+00, 7.475e+03, 3.000e+00, …, 1.000e+00, 0.000e+00,
        0.000e+00],
       [1.000e+00, 4.320e+03, 3.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00],
       [1.000e+00, 1.836e+03, 2.000e+00, …, 0.000e+00, 0.000e+00,
        1.000e+00]])
```

[13]:
```
# scale the feature matrix using the fit_transform() method of the scaler object
X_scaled = scaler.fit_transform(X)

# assign the scaled feature matrix to the original variable name 'X'
X = X_scaled
```

[14]:
```
# create a 1D NumPy array of zeros with length 12
theta = np.zeros(12)

# reshape the 1D array to a column vector using np.reshape
theta = np.reshape(theta, (12,1))

# display the column vector
theta
```

[14]:
```
array([[0.],
       [0.],
       [0.],
```

```
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[15]:
```
# create a 1D NumPy array 'Y_t' from the 'price' column of the training set␣
 ↪'train'
#h = np.matmul(X,theta)
Y_t = np.array(train.price)
Y = Y_t
Y = Y_t.reshape(436,1)
y = scaler.fit_transform(Y)
Y=y
```

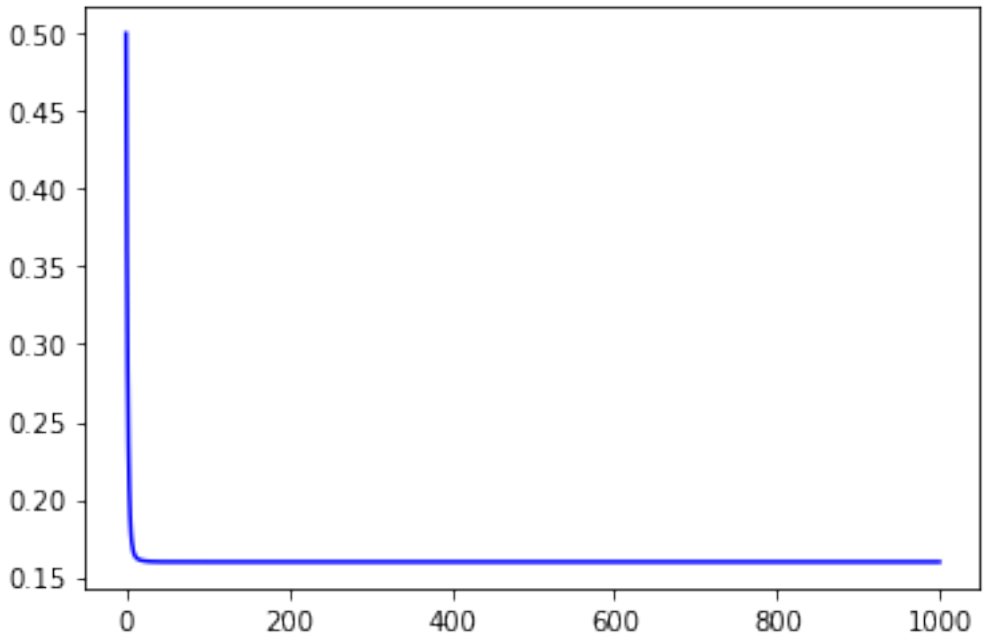
[16]:
```
# create a NumPy array 'X_T' containing the transpose of the feature matrix 'X'
X_T = np.array(X.T)

# retrieve the number of rows 'm' and the number of columns 'n' from the␣
 ↪feature matrix 'X'
m, n = X.shape

# display the values of 'm' and 'n'
print("Number of training examples (m): ", m)
print("Number of features (n): ", n)

# set the number of iterations for gradient descent
iterations = 1000

# create a counter variable 'count' and a NumPy array 'j' to store the cost␣
 ↪function values for each iteration
count = 0
j = np.zeros(shape=(iterations, 1), dtype=float)

# display the shape of the 'j' array
print("Shape of 'j' array: ", j.shape)
```

```
Number of training examples (m):  436
Number of features (n):  12
Shape of 'j' array:  (1000, 1)
```

[17]:
```
# Initialize variables
iterations = 1000
```

```
count = 0
alpha = 0.1
lambda_ = 0.001
j = np.zeros(iterations)

# Perform gradient descent
while count < iterations:
    h_t = X.dot(theta)
    h = np.array(h_t, float)

    j[count] = (1/(2*m)) * np.sum((h - Y)**2) + (lambda_ / (2*m)) * np.
 ↪sum(theta**2)

    grad_t = X_T.dot(h - Y)
    grad = grad_t * (1/m)

    theta = theta * (1 - alpha * (lambda_ / m)) - alpha * grad

    count += 1

# Plot the cost function
plt.plot(j, 'b-')
```

[17]: [<matplotlib.lines.Line2D at 0x7f03efeffd60>]



[18]:
```
# extract the test set features into NumPy arrays
N_X1_t = np.array(test.area)
N_X2_t = np.array(test.bedrooms)
N_X3_t = np.array(test.bathrooms)
N_X4_t = np.array(test.stories)
N_X5_t = np.array(test.parking)
N_X6_t = np.array(test.mainroad)
N_X7_t = np.array(test.guestroom)
N_X8_t = np.array(test.hotwaterheating)
N_X9_t = np.array(test.airconditioning)
N_X10_t = np.array(test.prefarea)
N_X11_t = np.array(test.basement)
N_X0_t = np.ones(109)
```

[19]:
```
# stack the test set features into a design matrix
N_X = np.vstack([N_X0_t,N_X1_t,N_X2_t,N_X3_t,N_X4_t,N_X5_t,␣
 ↪N_X6_t,N_X7_t,N_X8_t,N_X9_t,N_X10_t,N_X11_t])
N_X_T = N_X.T
N_X = np.array(N_X_T)
N_x = scaler.fit_transform(N_X)
N_X = N_x
N_X.shape
```

[19]:
```
(109, 12)
```

[20]:
```
# initialize the parameters for the test set
N_theta = np.array([0.,0.,0.,0.,0,0.,0.,0.,0.,0.,0,0.])
N_theta = N_theta.reshape(12,1)
N_theta
```

[20]:
```
array([[0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.],
       [0.]])
```

[21]:
```
# initialize the target variable for the test set
N_Y_t = np.array(test.price)
N_Y = N_Y_t
N_Y = N_Y_t.reshape(109,1)
```

```
N_y = scaler.fit_transform(N_Y)
N_Y=N_y
N_Y.shape
```

[21]: (109, 1)

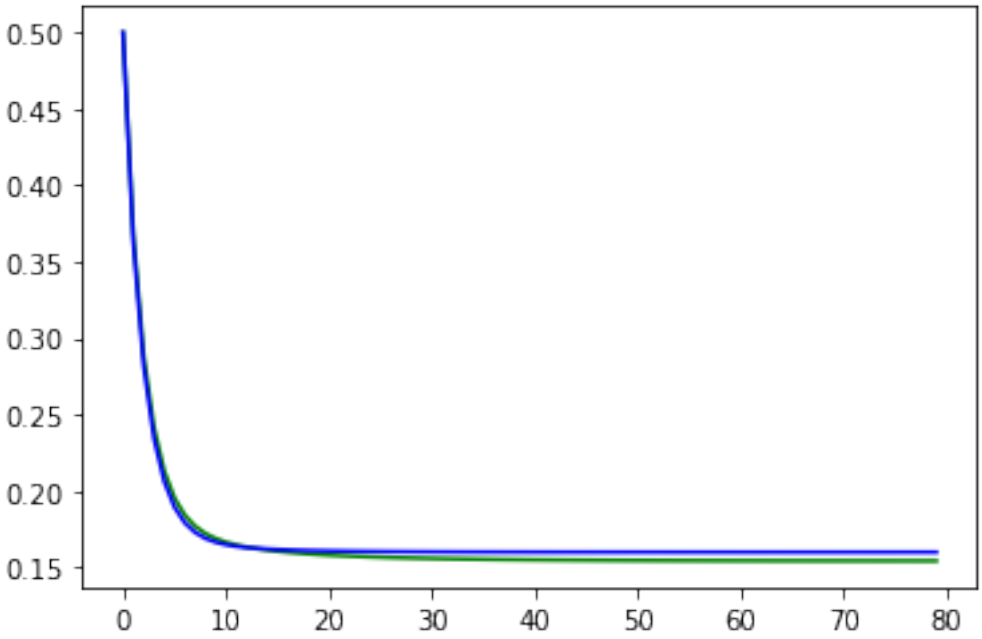
[22]:
```
N_X_T = np.array(N_X.T)
m,n = N_X.shape
m,n
```

[22]: (109, 12)

[24]:
```
iterations = 2500
count=0
N_j = np.zeros(shape=(iterations, 1), dtype=float)

while(count < iterations):

    N_h_t = N_X.dot(N_theta)
    N_h = np.array(N_h_t)


    N_j[count]= (1/(2*m))*np.sum((N_h - N_Y)**2)+(0.001)*(1/(2*m))*np.
 ↪sum((N_theta)**2)

    grad_t = N_X_T.dot(N_h-N_Y)
    grad = grad_t*(1/m)

    N_theta = N_theta*(1-0.1*(0.001/m)) - 0.1*(grad)

    count += 1
```

[26]:
```
plt.plot(N_j[:80],'g-')
plt.plot(j[:80],'b-')
```

[26]: [<matplotlib.lines.Line2D at 0x7f03ef903eb0>]


0.50
0.45
0.40
0.35
0.30
0.25
0.20
0.15
0
10
20
30
40
50
60
70
80