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In [80]: '''
Patrick Ballou
ID: 801130521
ECGR 4105
Homework 1
Problem 1
'''
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Out[80]: '\nPatrick Ballou\nID: 801130521\nECGR 4105\nHomework 1\nProblem 1\n'
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```
In [81]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
```

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In [82]: #create pandas dataframe and print first 5 rows
df = pd.read_csv("Housing.csv")
df_copy = df.copy() #copy will help later on
df.head()
```

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Out[82]:
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|   | price    | area | bedrooms | bathrooms | stories | mainroad | guestroom | basement | hotwaterheating |
|---|----------|------|----------|-----------|---------|----------|-----------|----------|-----------------|
| 0 | 13300000 | 7420 | 4        | 2         | 3       | yes      | no        | no       | no              |
| 1 | 12250000 | 8960 | 4        | 4         | 4       | yes      | no        | no       | no              |
| 2 | 12250000 | 9960 | 3        | 2         | 2       | yes      | no        | yes      | no              |
| 3 | 12215000 | 7500 | 4        | 2         | 2       | yes      | no        | yes      | no              |
| 4 | 11410000 | 7420 | 4        | 1         | 2       | yes      | yes       | yes      | no              |

```
In [83]: #categorical inputs that need to be mapped to numbers
non_num_varlist = ["mainroad", "guestroom", "basement", "hotwaterheating", "airconditioning"]
```

```
In [84]: #mapping function
def to_num(x):
    return x.map({"yes": 1, "no": 0})
```

```
In [85]: #map inputs and output new dataframe
df[non_num_varlist] = df_copy[non_num_varlist].apply(to_num) #copy df is to avoid problem
df.head()
```

```
Out[85]:
```

|   | price    | area | bedrooms | bathrooms | stories | mainroad | guestroom | basement | hotwaterheating |
|---|----------|------|----------|-----------|---------|----------|-----------|----------|-----------------|
| 0 | 13300000 | 7420 | 4        | 2         | 3       | 1        | 0         | 0        | 0               |
| 1 | 12250000 | 8960 | 4        | 4         | 4       | 1        | 0         | 0        | 0               |
| 2 | 12250000 | 9960 | 3        | 2         | 2       | 1        | 0         | 1        | 0               |
| 3 | 12215000 | 7500 | 4        | 2         | 2       | 1        | 0         | 1        | 0               |
| 4 | 11410000 | 7420 | 4        | 1         | 2       | 1        | 1         | 1        | 0               |

```
In [86]: #train/test split, random_state functions as seed
df_train, df_test = train_test_split(df, train_size=.8, test_size=.2, random_state=7)
```

```
In [87]: #don't need all variables for problem 1
vars = ['area', 'bedrooms', 'bathrooms', 'stories', 'parking', 'price']
train_set = df_train[vars]
test_set = df_test[vars]
```

```
In [88]: #create variables for test and train set since they are different sizes
m_train = len(train_set)
x_train = train_set[['area', 'bedrooms', 'bathrooms', 'stories', 'parking']]
x_0_train = np.ones((m_train,1))
X_train = np.hstack((x_0_train, x_train))
Y_train = train_set['price']

m_test = len(test_set)
x_test = test_set[['area', 'bedrooms', 'bathrooms', 'stories', 'parking']]
x_0_test = np.ones((m_test,1))
X_test = np.hstack((x_0_test, x_test))
Y_test = test_set['price']
```

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In [89]: #Loss function
def compute_cost(X, y, theta, m):
    predictions = X.dot(theta)
    errors = np.subtract(predictions, y)
    sqrErrors = np.square(errors)
    J = (1/(2*m))*np.sum(sqrErrors)

    return J
```

```
In [90]: #gradient descent function
def gradient_descent(X, y, theta, alpha, iterations):
    train_cost_history = np.zeros(iterations)
    test_cost_history = np.zeros(iterations)

    for i in range(iterations):
        predictions = X.dot(theta)
        errors = np.subtract(predictions, y)
        sum_delta = (alpha/m_train) * X.transpose().dot(errors)
        theta -= sum_delta
        train_cost_history[i] = compute_cost(X_train, Y_train, theta, m_train)
        test_cost_history[i] = compute_cost(X_test, Y_test, theta, m_test)

    return theta, train_cost_history, test_cost_history
```

```
In [91]: #initialize theta, # of iterations, and Learning rate
theta = np.zeros(6)
iterations = 20000
#very small theta because input 'area' is messing up the model since it is not scaled
alpha = .00000000001
```

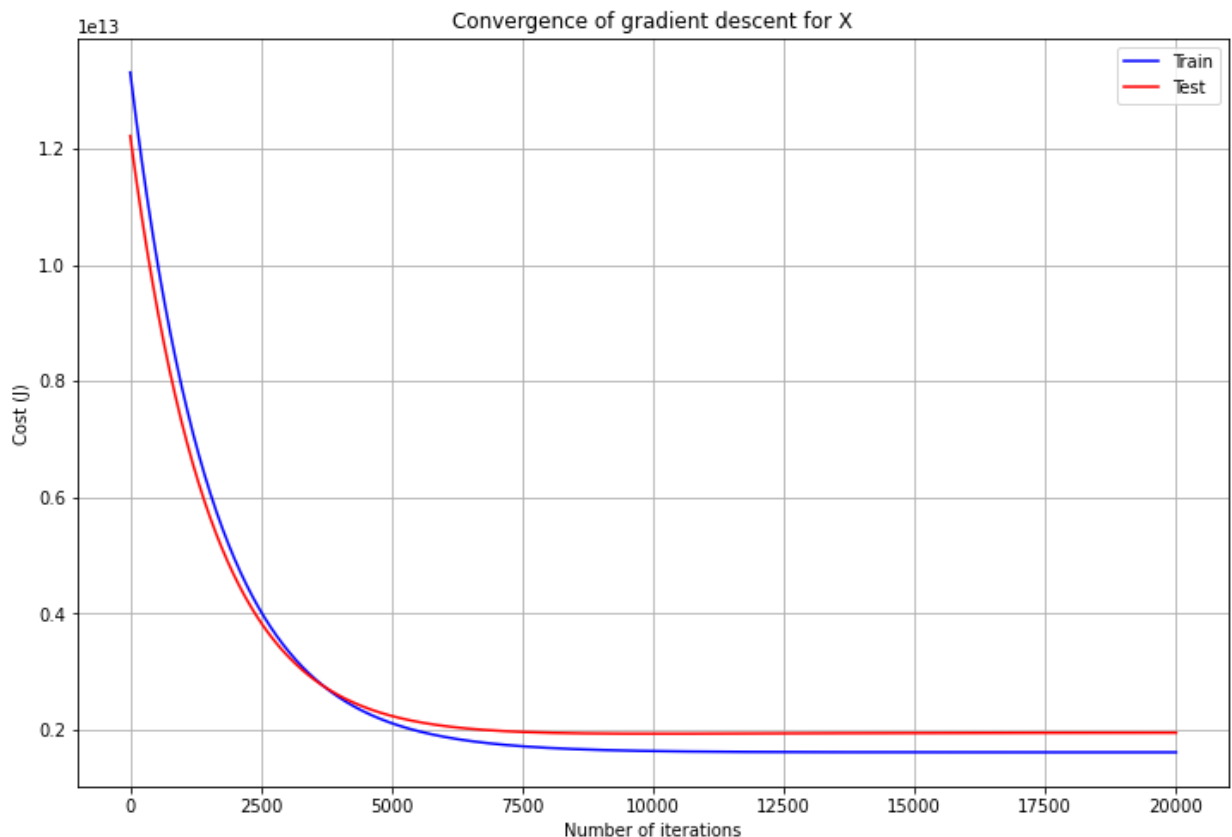
```
In [92]: #calculate test and train costs, and theta
theta, train_cost_history, test_cost_history = gradient_descent(X_train, Y_train, theta, alpha, iterations)
print("Final theta values for a:", theta)
```

Final theta values for a: [2.11132525e-01 8.61083612e+02 6.94719483e-01 3.44382851e-01  
1  
4.97239587e-01 1.72230062e-01]

```
In [93]: #plot loss vs iterations
plt.rcParams["figure.figsize"] = (12,8)
plt.plot(range(1, iterations + 1), train_cost_history, color='blue', label='Train')
plt.plot(range(1, iterations + 1), test_cost_history, color='red', label='Test')
plt.grid()
plt.legend()
plt.xlabel('Number of iterations')
plt.ylabel('Cost (J)')
plt.title('Convergence of gradient descent for X')
print("Training cost:", train_cost_history[-1])
print("Testing cost:", test_cost_history[-1])
```

Training cost: 1606259143218.0923

Testing cost: 1944119106864.634



```
In [94]: #Problem 1, part b
#same process as part a except more input variables

vars_b = ['area', 'bedrooms', 'bathrooms', 'stories', 'mainroad', 'guestroom', 'basement']
train_set_b = df_train[vars_b]
test_set_b = df_test[vars_b]

#train and test set variables
m_train_b = len(train_set_b)
x_train_b = train_set_b[['area', 'bedrooms', 'bathrooms', 'stories', 'mainroad', 'guestroom', 'basement']]
x_0_train_b = np.ones((m_train_b,1))
X_train_b = np.hstack((x_0_train_b, x_train_b))
Y_train_b = train_set_b['price']
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m_test_b = len(test_set_b)
x_test_b = test_set_b[['area', 'bedrooms', 'bathrooms', 'stories', 'mainroad', 'guestrooms']]
x_0_test_b = np.ones((m_test_b,1))
X_test_b = np.hstack((x_0_test_b, x_test_b))
Y_test_b = test_set_b['price']

#small theta is needed for same reason as part a
theta_b = np.zeros(12)
iterations_b = 1600
alpha_b = .000000001

```

```

In [95]: #new gradient descent for part b to avoid problems, same structure as part a
def gradient_descent_b(X, y, theta, alpha, iterations):
    train_cost_history_b = np.zeros(iterations)
    test_cost_history_b = np.zeros(iterations)

    for i in range(iterations):
        predictions = X.dot(theta)
        errors = np.subtract(predictions, y)
        sum_delta = (alpha/m_train_b) * X.transpose().dot(errors)
        theta -= sum_delta
        train_cost_history_b[i] = compute_cost(X_train_b, Y_train_b, theta, m_train_b)
        test_cost_history_b[i] = compute_cost(X_test_b, Y_test_b, theta, m_test_b)

    return theta, train_cost_history_b, test_cost_history_b

```

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In [96]: theta_b, train_cost_history_b, test_cost_history_b = gradient_descent_b(X_train_b, Y_train_b, theta_b, alpha_b, iterations_b)
print("Final theta values for b:", theta_b)

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Final theta values for b: [1.96526674e-01 8.57123320e+02 6.39843637e-01 3.12821014e-01
1.49239267e-01 1.74607418e-01 5.08318396e-02 9.38560300e-02
1.43031168e-02 1.02501195e-01 1.61114151e-01 6.11611234e-02]

```

```

In [97]: plt.plot(range(1, iterations_b + 1), train_cost_history_b, color='blue', label='Train')
plt.plot(range(1, iterations_b + 1), test_cost_history_b, color='red', label='Test')
plt.grid()
plt.legend()
plt.xlabel('Number of iterations')
plt.ylabel('Cost (J)')
plt.title('Convergence of gradient descent for X')
print("Training cost:", train_cost_history_b[-1])
print("Testing cost:", test_cost_history_b[-1])

```

```

Training cost: 1606703724154.2678
Testing cost: 1940242093479.1543

```

