```
Assigment 3
         This assignment focuses on getting comfortable with working with multidimensional data and linear regression. Key items
         include:
           · Creating random n-dimensional data
           · Creating a Model that can handle the data

    Plot a subset of the data along with the prediction

           • Using a Dataset to read in and choose certain columns to produce a model
           · Create several models from various combinations of columns
           · Plot a few of the results
           . BONUS: Perform all the plots in 3D instead of 2D
 In [4]: import numpy as np
          import matplotlib.pylab as plt
          %matplotlib inline
         1. Create a 4 dimensional data set with 64 elements and show 2D plots of the
         data x_1 \rightarrow y, x_2 \rightarrow y, etc.
 In [8]: #make the dataset
          n = 64
         x = np.linspace(0, 1, n) + np.random.rand(4, n)
         x = np.vstack([x, np.ones(len(x.T))]).T
         y = np.linspace(0, 1, n) + np.random.rand(n) - 1
 In [9]: #plot y,x1
         plt.scatter(x.T[0],y)
 Out[9]: <matplotlib.collections.PathCollection at 0x2a967d95668>
           1.00
           0.75
           0.50
           0.25
           0.00
           -0.25
           -0.50
          -0.75
                                         1.25
In [10]: #plot y,x2
         plt.scatter(x.T[1],y)
Out[10]: <matplotlib.collections.PathCollection at 0x2a967e20ba8>
           1.00
           0.75
           0.50
           0.25
           0.00
           -0.25
           -0.50
          -0.75
                              0.75
                                   1.00
                                         1.25
                                              1.50 1.75
In [11]: #plot y,x3
          plt.scatter(x.T[2],y)
Out[11]: <matplotlib.collections.PathCollection at 0x2a96809d7b8>
           1.00
           0.75
           0.50
           0.25
           0.00
           -0.25
          -0.50
           -0.75
                               0.75
                                    1.00
                                         1.25 1.50
                    0.25
                         0.50
In [15]: #plot y,x4
         plt.scatter(x.T[3],y)
Out[15]: <matplotlib.collections.PathCollection at 0x2a9681ca470>
           1.00
           0.75
           0.50
           0.25
           0.00
           -0.25
          -0.50
           -0.75
                                            1.50 1.75
                                       1.25
                             0.75
                                  1.00
         2. Create a model to fit the data. Hint: follow the example from Lesson 3
In [16]: left = np.linalg.inv(np.dot(x.T,x))
In [18]: right = np.dot(y.T, x)
In [19]: np.dot(left, right)
Out[19]: array([ 3.20241053e-01, 2.75656281e-01, 2.85218671e-01, -8.06996665e-04,
                 -8.20575701e-01])
In [20]: beta = np.linalg.lstsq(x, y) [0]
         D:\Python\lib\site-packages\ipykernel_launcher.py:1: FutureWarning: `rcond` parameter will chang
         e to the default of machine precision times ``max(M, N)`` where M and N are the input matrix \dim
         To use the future default and silence this warning we advise to pass `rcond=None`, to keep using
         the old, explicitly pass `rcond=-1`.
           """Entry point for launching an IPython kernel.
Out[20]: array([ 3.20241053e-01, 2.75656281e-01, 2.85218671e-01, -8.06996665e-04,
                -8.20575701e-01])
In [50]: pred = np.dot(x, beta)
         3. Plot the model's prediction in 2D for 2 of the dimensions (x_1 \rightarrow y_p, x_2 \rightarrow y_p)
         along with the original points
In [22]: #plot dimension 1
         plt.scatter(x.T[0], pred, c='red')
         plt.scatter(x.T[0], y, c='blue')
Out[22]: <matplotlib.collections.PathCollection at 0x2a968a536a0>
           0.75
           0.50
           0.25
           0.00
           -0.25
          -0.50
           -0.75
          -1.00
                              0.75
                                   1.00
                                         1.25
                                              1.50
                        0.50
In [23]: #plot dimension 2
         plt.scatter(x.T[1], pred, c='red')
         plt.scatter(x.T[1], y, c='blue')
Out[23]: <matplotlib.collections.PathCollection at 0x2a968ab59b0>
           0.75
           0.50
           0.25
           0.00
           -0.25
          -0.50
           -0.75
                   0.25
                         0.50
                              0.75 1.00
                                         1.25 1.50
         4. Read in mlnn/data/Credit.csv with Pandas and create a model to
         predict Credit Rating (Rating). Use only the numeric columns in your model,
         but feel free to experiment which which columns you believe are better
         predicters of Credit Rating
 In [5]: import pandas as pd
          credit = pd.read_csv('../data/Credit.csv')
         credit.head()
 Out[5]:
            Unnamed: 0 | Income | Limit | Rating | Cards | Age | Education | Gender | Student | Married
                                                                                          Ethnicity | Balance
          0 1
                                                                                          Caucasian 333
                        14.891
                                3606 283
                                                   34
                                                        11
                                                                  Male
                                                                         No
                                                                                  Yes
          1 2
                        106.025 6645
                                     483
                                                   82
                                                       15
                                                                                          Asian
                                                                                                   903
                                                                  Female Yes
                                                                                  Yes
          2 3
                                                   71
                                                        11
                        104.593 7075
                                                                  Male
                                                                                                   580
                                     514
                                                                                  No
                                                                                          Asian
          3 4
                                                        11
                                                   36
                                                                                  No
                        148.924
                                9504
                                     681
                                                                  Female
                                                                                          Asian
                                                                                                    964
          4 5
                        55.882
                                4897
                                     357
                                                   68
                                                       16
                                                                  Male
                                                                         No
                                                                                  Yes
                                                                                                   331
                                                                                          Caucasian
```

3 4 148.924 9504 681 3 36 11 Female No No Asian 964
4 5 55.882 4897 357 2 68 16 Male No Yes Caucasian 331

In [6]: X = credit[['Income', 'Limit']].as_matrix()

X = np.vstack([X.T, np.ones(len(X))]).T

D:\Python\lib\site-packages\ipykernel_launcher.py:1: FutureWarning: Method .as_matrix will be removed in a future version. Use .values instead.

"""Entry point for launching an IPython kernel.

```
"""Entry point for launching an IPython kernel.

In [7]: Y = credit['Rating']

In [8]: left = np.linalg.inv(np.dot(X.T, X))
    right = np.dot(Y.T, X)
    np.dot(left, right)
    beta = np.linalg.lstsq(X, Y)[0]
```

e to the default of machine precision times ``max(M, N)`` where M and N are the input matrix dim ensions.

To use the future default and silence this warning we advise to pass `rcond=None`, to keep using the old, explicitly pass `rcond=-1`.

after removing the cwd from sys.path.

5. Plot your results (Bonus if you use 3D plots). Show as many of your columns vs. credit rating that you can.

D:\Python\lib\site-packages\ipykernel launcher.py:4: FutureWarning: `rcond` parameter will chang

In [9]: import matplotlib.pyplot as plt from mpl_toolkits.mplot3d import Axes3D

```
ax.view_init(16,20)
ax.scatter(X.T[0], X.T[1], pred, zdir='z')
ax.scatter(X.T[0], X.T[1], Y, zdir='z', c='r')

Out[9]: <mpl_toolkits.mplot3d.art3d.Path3DCollection at 0x2469ff35668>
```

```
800
600
200
200040006000800010000q200q4000
```

```
In [10]: fig = plt.figure()
    ax = fig.add_subplot(111, projection='3d')
    ax.view_init(16,20)
    ax.scatter(X.T[0], X.T[2], pred, zdir='z')
    ax.scatter(X.T[0], X.T[2], Y, zdir='z', c='r')
```

pred = np.dot(X, beta)

fig = plt.figure()

ax = fig.add subplot(111, projection='3d')

Out[10]: <mpl_toolkits.mplot3d.art3d.Path3DCollection at 0x246a04a07f0>