Heat Maps using Matplotlib and Seaborn

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```
In [1]: %pylab inline
        import matplotlib.pyplot as plt
        import pandas as pd
        import numpy as np
Populating the interactive namespace from numpy and matplotlib
In [2]: helix = pd.read_csv('Data/helix_parameters.csv')
        helix.head() # just seeing that data was imported properly by outputing first 5 cells
                   Energy n_helices r0_A r0_B r0_C omega0 delta_omega0_A \
Out[2]:
           job_n
        0 36019 -387.167
                                    3
                                        6.0
                                              6.0
                                                    6.0
                                                               0
                                                                               0
        1 36022 -402.606
                                    3
                                        6.0
                                              6.0
                                                     6.0
                                                               0
                                                                               0
        2 36020 -395.944
                                    3
                                        6.0
                                              6.0
                                                     6.0
                                                               0
                                                                               0
        3 36002 -389.788
                                    3
                                        6.0
                                              6.0
                                                    6.0
                                                               0
                                                                               0
        4 36005 -388.016
                                        6.0
                                              6.0
                                                     6.0
           delta_omega0_B delta_omega0_C ...
                                                 invert_B invert_C z1_offset_A \
        0
                      120
                                       240 ...
                                                          1
                                                                    0
                                                                                  0
        1
                      120
                                       240
                                                                                  0
                                           . . .
        2
                      120
                                       240
                                                                    0
                                                                                  0
                                                          1
        3
                      120
                                       240
                                                          1
                                                                    0
                                                                                  0
                      120
                                       240
                                                          1
                                                                                  0
           \verb|z1_offset_B| | \verb|z1_offset_C| | | delta_t\_A| | delta_t\_B| | delta_t\_C|
                                                                         omega1
        0
                                -3.0
                                              0
                                                          0
                                                                        99.999268 1.51
                   0.6
                                              0
                                                          0
        1
                   0.6
                                 0.6
                                                                     0 99.999268 1.51
        2
                   0.6
                                -1.8
                                              0
                                                         0
                                                                     0 99.999268 1.51
```

0

0

0

0 99.999268 1.51

0 99.999268 1.51

[5 rows x 27 columns]

-3.0

-3.0

-1.8

1.8

In [3]: # shape of the dataframe
 helix.shape

Out[3]: (47475, 27)

3

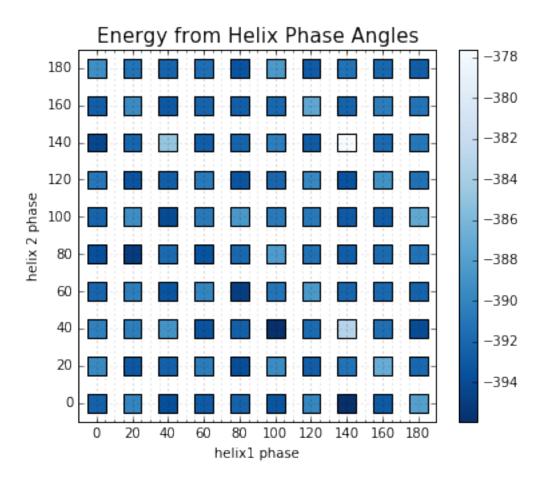
In [4]: # checking what the columns are
 helix.columns

```
u'helix3 phase', u'invert_A', u'invert_B', u'invert_C', u'z1_offset_A',
               u'z1_offset_B', u'z1_offset_C', u'delta_t_A', u'delta_t_B',
               u'delta_t_C', u'omega1', u'z1'],
              dtype='object')
  Selecting Columns (by different methods)
In [5]: # selecting a couple columns
        couple_columns = helix[['Energy','helix 2 phase', 'helix1 phase']]
        couple_columns.head()
Out[5]:
            Energy helix 2 phase helix1 phase
        0 -387.167
                                0
        1 -402.606
                                0
                                               0
        2 -395.944
                                0
                                               0
        3 -389.788
                                0
                                               0
        4 -388.016
                                0
In [6]: # selecting same columns a different way
        helix.ix[:,['Energy','helix 2 phase', 'helix1 phase']].head()
Out[6]:
            Energy helix 2 phase helix1 phase
        0 -387.167
                                0
        1 - 402.606
                                0
                                               0
        2 -395.944
                                0
                                               0
        3 -389.788
                                               0
                                0
        4 -388.016
                                0
                                               0
  Heat Map
In [7]: # this is essentially would be taking the average of each unique combination.
        # one important mention is notice how little the data varies from eachother.
        phase_1_2 = couple_columns.groupby(['helix1 phase', 'helix 2 phase']).mean()
        print phase_1_2.shape
        phase_1_2.head(10)
(100, 1)
Out[7]:
                                         Energy
        helix1 phase helix 2 phase
                                    -392.419841
        0
                     0
                     20
                                    -389.622691
                     40
                                    -390.318620
                     60
                                    -392.198537
                     80
                                    -393.661624
                     100
                                    -392.226253
                     120
                                    -390.955112
                     140
                                    -394.319969
                     160
                                    -392.594862
                     180
                                    -389.254009
In [8]: phase_1_2 = phase_1_2.reset_index()
```

u'z0_offset_B', u'z0_offset_C', u'helix1 phase', u'helix 2 phase',

phase_1_2.head()

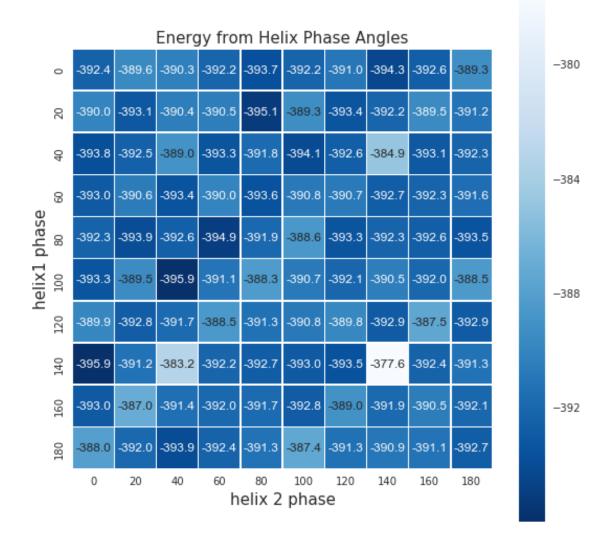
```
helix1 phase helix 2 phase
Out[8]:
                                            Energy
                                     0 -392.419841
                      0
                      0
                                    20 -389.622691
        1
        2
                      0
                                    40 -390.318620
        3
                      0
                                    60 -392.198537
        4
                      0
                                    80 -393.661624
  Heat Map using Matplotlib
In [9]: major_ticks = np.arange(0, 200, 20)
        minor_ticks = np.arange(0, 180, 5)
       fig = plt.figure(figsize = (6,5))
        ax = fig.add_subplot(1,1,1)
        s = ax.scatter('helix1 phase', 'helix 2 phase', c = 'Energy',data = phase_1_2, cmap = 'Blues_r'
        ax.axis([phase_1_2['helix1 phase'].min()-10, phase_1_2['helix1 phase'].max()+10, phase_1_2['hel
        ax.set_xticks(major_ticks)
        ax.set_xticks(minor_ticks, minor=True)
        ax.set_yticks(major_ticks)
        ax.grid(which='both', alpha = 0.3)
        ax.grid(which='major', alpha=0.3)
        ax.set_xlabel('helix1 phase', fontsize=10);
        ax.set_ylabel('helix 2 phase', fontsize=10);
        ax.set_title('Energy from Helix Phase Angles', size = 15)
        {\it \# http://stackoverflow.com/questions/13943217/how-to-add-colorbars-to-scatterplots-created-like}
        cbar = plt.colorbar(mappable = s,ax = ax)
       plt.show()
```



Heat Map using Seaborn

```
In [10]: import numpy as np;
         import seaborn as sns;
         # To translate into Excel Terms for those familiar with Excel
         # string 1 is row labels 'helix1 phase'
         # string 2 is column labels 'helix 2 phase'
         # string 3 is values 'Energy'
         # Official pivot documentation
         {\it \# http://pandas.pydata.org/pandas-docs/stable/generated/pandas.DataFrame.pivot.html}
         phase_1_2.pivot('helix1 phase', 'helix 2 phase', 'Energy').head()
Out[10]: helix 2 phase
                                0
                                            20
                                                         40
                                                                     60
                                                                                  80
         helix1 phase
                        -392.419841 -389.622691 -390.318620 -392.198537 -393.661624
         20
                        -390.013375 -393.121438 -390.435441 -390.489424 -395.112815
         40
                        -393.751152 -392.486692 -388.967952 -393.297131 -391.750444
         60
                       -392.960539 \ -390.578310 \ -393.441091 \ -390.007804 \ -393.577453
                        -392.347250 -393.919370 -392.575410 -394.918433 -391.945480
         80
         helix 2 phase
                                100
                                            120
                                                         140
                                                                     160
                                                                                  180
```

```
helix1 phase
         0
                       -392.226253 -390.955112 -394.319969 -392.594862 -389.254009
                       -389.307695 -393.365839 -392.171938 -389.534864 -391.171958
         20
                       -394.077600 -392.610059 -384.873483 -393.059775 -392.283178
         40
         60
                       -390.840711 -390.734280 -392.706732 -392.278805 -391.558888
         80
                       -388.600299 -393.314092 -392.258363 -392.636642 -393.471684
In [11]: # To translate into Excel Terms for those familiar with Excel
         # string 1 is row labels 'helix1 phase'
         # string 2 is column labels 'helix 2 phase'
         # ['Energy'] is values
         phase_1_2.pivot('helix1 phase', 'helix 2 phase')['Energy'].head()
                                                        40
Out[11]: helix 2 phase
                                           20
                                                                                80
         helix1 phase
                       -392.419841 -389.622691 -390.318620 -392.198537 -393.661624
         0
                       -390.013375 -393.121438 -390.435441 -390.489424 -395.112815
         20
         40
                       -393.751152 -392.486692 -388.967952 -393.297131 -391.750444
         60
                       -392.960539 -390.578310 -393.441091 -390.007804 -393.577453
         80
                       -392.347250 -393.919370 -392.575410 -394.918433 -391.945480
         helix 2 phase
                               100
                                           120
                                                        140
                                                                    160
                                                                                180
         helix1 phase
                       -392.226253 -390.955112 -394.319969 -392.594862 -389.254009
         20
                       -389.307695 -393.365839 -392.171938 -389.534864 -391.171958
                       -394.077600 -392.610059 -384.873483 -393.059775 -392.283178
         40
         60
                       -390.840711 -390.734280 -392.706732 -392.278805 -391.558888
                       -388.600299 -393.314092 -392.258363 -392.636642 -393.471684
         80
In [12]: # seaborn heatmap documentation
         # https://stanford.edu/~mwaskom/software/seaborn/qenerated/seaborn.heatmap.html
         # cmap choices: http://matplotlib.org/users/colormaps.html
         plt.figure(figsize=(9,9))
         pivot_table = phase_1_2.pivot('helix1 phase', 'helix 2 phase', 'Energy')
         plt.xlabel('helix 2 phase', size = 15)
         plt.ylabel('helix1 phase', size = 15)
         plt.title('Energy from Helix Phase Angles', size = 15)
         sns.heatmap(pivot_table, annot=True, fmt=".1f", linewidths=.5, square = True, cmap = 'Blues_r'
```



In []: