HN analysis part 2 - ANALYSIS 1-12-2016

December 1, 2016

In [1]: import pandas as pd

```
import numpy as np
        import matplotlib.pyplot as plt
        %matplotlib inline
        from scipy import stats
        from scipy.stats import wilcoxon # must import explicitly
        import seaborn as sns
        sns.set(style="whitegrid", palette="pastel", color_codes=True)
0.1 Helper functions
In [2]: def diff_percent(Dm, AAA):
            return 100 * ((Dm-AAA)/AAA)
        print (diff_percent (105.0, 100.0))
5.0
In [3]: def my_wilcox(data):
                                                # helper function to return p-value
            z_stat, p_val = wilcoxon(data) # if only 1 value passed
            #z_stat, p_val = wilcoxon(data.as_matrix()) # if only 1 value passed
            return p_val
In [4]: my_wilcox([1,2,3,4,6,7,2,352,3])
C:\Users\RCole02.ROYALSURREY\AppData\Local\Continuum\Anaconda3\lib\site-packages\so
 warnings.warn("Warning: sample size too small for normal approximation.")
Out [4]: 0.0075792819433897285
In [5]: def bland_altman_plot(data1, data2, structure, metric, *args, **kwargs):
           data1 = np.asarray(data1)
                     = np.asarray(data2)
            z_stat, p_val = wilcoxon(data1, data2)
           mean
                   = np.mean([data1, data2], axis=0)
            diff
                     = data1 - data2
                                                        # Difference between data1
```

```
sd
                     = np.std(diff, axis=0)
                                                     # Standard deviation of the
           plt.scatter(mean, diff, *args, **kwargs)
                                   color='red', linestyle='-')
           plt.axhline(md,
           plt.axhline(md + 1.96*sd, color='blue', linestyle='--')
           plt.axhline(md - 1.96*sd, color='blue', linestyle='--')
           plt.xlabel('Dose (Gy)')
           plt.ylabel('Difference Gy')
           plt.title( str(np.round(md, decimals = 3)) + ' Gy mean difference with
           plt.savefig('BA.png')
In [6]: def query_data(df, Col, structure, metric): # helper function to get data
           return df[Col][(df['structure'] == structure) & (df['metric'] == metric
1 Start analysis
In [7]: HN_df = pd.read_csv('HN_df_clean_28_11.csv') # read in the cleaned data
In [8]: HN_df['Dm-AAA'] = HN_df['Dm'] - HN_df['AAA'] # get abs diff
In [9]: HN_df['Dm-AAA%'] = diff_percent(HN_df['Dm'], HN_df['AAA'])
In [10]: HN df.head()
Out[10]:
           Case structure metric
                                        AAA
                                                    Dm
                                                          Dm-AAA
                                                                    Dm-AAA%
        0 Case1
                     PTV2 D2% 65.392496 64.671451 -0.721045 -1.102642
        1 Case1
                      PTV2
                             D5% 61.221910 60.574434 -0.647476 -1.057589
        2 Case1
                      PTV2 D50% 56.935482 56.306392 -0.629090 -1.104918
        3 Case1
                     PTV2 D95% 52.908380 53.783883 0.875503
                                                                 1.654753
        4 Case1
                     PTV2 D99% 38.261698 45.021250 6.759552 17.666628
In [11]: HN_sub_df = HN_df[['Case', 'structure', 'metric', 'Dm-AAA%']]
        HN_sub_df.head()
Out[11]:
            Case structure metric
                                   Dm-AAA%
        0 Case1
                      PTV2
                             D2% -1.102642
        1 Case1
                      PTV2
                             D5% -1.057589
        2 Case1
                     PTV2 D50% -1.104918
        3 Case1
                      PTV2 D95% 1.654753
        4 Case1
                      PTV2 D99% 17.666628
In [12]: HN_sub_df['structure'].unique()
Out[12]: array(['PTV2', 'PTV1', 'Body', 'R Parotid', 'L Parotid', 'Brain Stem',
               'Spinal Cord', 'CTV65'], dtype=object)
```

= np.mean(diff)

md

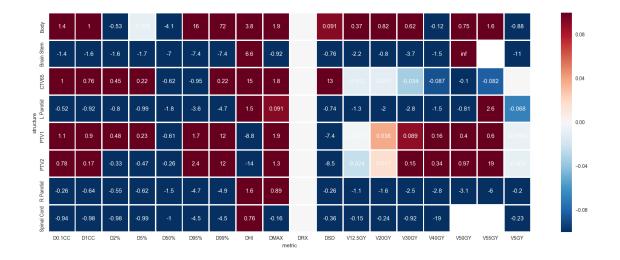
Mean of the difference

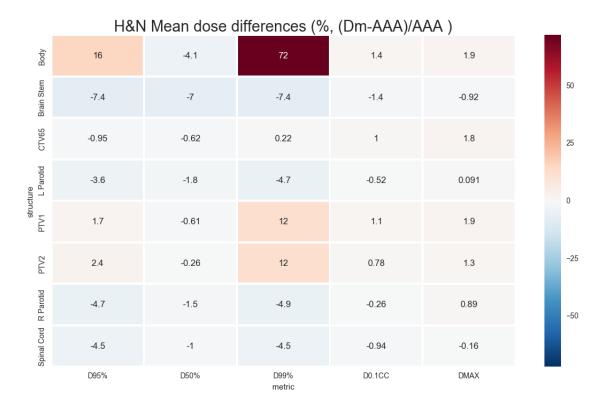
Everything imported OK

1.1 Get mean dose differences

C:\Users\RCole02.ROYALSURREY\AppData\Local\Continuum\Anaconda3\lib\site-packages\ma
resdat /= (vmax - vmin)

C:\Users\RCole02.ROYALSURREY\AppData\Local\Continuum\Anaconda3\lib\site-packages\ma
cbook._putmask(xa, xa < 0.0, -1)</pre>

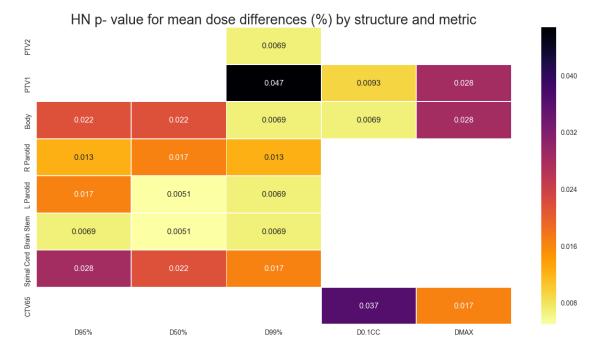




1.2 Get wilcoxon p value of differences

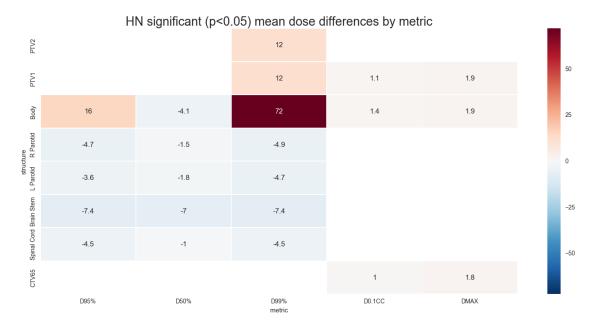
```
In [16]: structures_of_interest = HN_sub_df['structure'].unique()
In [17]: structures_of_interest
Out[17]: array(['PTV2', 'PTV1', 'Body', 'R Parotid', 'L Parotid', 'Brain Stem',
                'Spinal Cord', 'CTV65'], dtype=object)
In [18]: i = len(structures_of_interest)
         j = len(metrics_of_interest)
         wilcox_data = np.zeros((i,j)) # init an empty array
In [19]: d = HN_df # for convenience just copy
         j = 0
         for structure in structures_of_interest:
             i = 0
             for metric in metrics_of_interest:
                 A = d[(d['structure'] == structure) & (d['metric'] == metric)]
                 D = A['Dm-AAA%']
                 wilcox_data[j][i] = my_wilcox(D.values)
                 i = i + 1
             j = j+ 1
```

```
In [20]: wilcox_data_df = pd.DataFrame(data=wilcox_data, # values
                         index=structures_of_interest, # 1st column as index
                         columns=metrics_of_interest) # 1st row as the column nam
         . . .
In [21]: wilcox_data_df
Out [21]:
                         D95%
                                   D50%
                                            D99%
                                                    D0.1CC
                                                                DMAX
        PTV2
                     0.059336 0.139414 0.006910 0.332880 0.059336
        PTV1
                     0.332880 0.074462 0.046853 0.009344 0.028417
        Body
                     0.021824 0.021824 0.006910 0.006910 0.028417
        R Parotid
                     0.012515 0.016605 0.012515 0.114128 0.646462
                    0.016605 0.005062 0.006910 0.092601 0.386271
        L Parotid
        Brain Stem
                     0.006910 0.005062 0.006910 0.059336 0.139414
        Spinal Cord 0.028417 0.021824 0.016605 0.074462 0.646462
        CTV65
                     0.092601
                              0.074462 0.646462 0.036658 0.016605
In [22]: confidence = 0.05
        plt.figure(figsize=(16, 8))
        ax2 = sns.heatmap(wilcox_data_df[wilcox_data_df<confidence], annot=True, 1
        ax2.set_title('HN p- value for mean dose differences (%) by structure and
        plt.savefig('HN p value for mean dose differences.png', dpi=500)
```



```
mask = np.zeros_like(data, dtype=np.bool) # create a mask
mask[np.triu_indices_from(mask)] = True
```

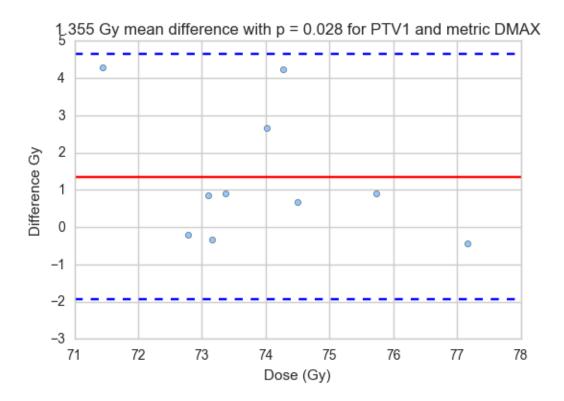
ax3 = sns.heatmap(data, annot=**True**, linewidths=.5, center=0) # mask=mask, ax3.set_title('HN significant (p<0.05) mean dose differences by metric', splt.savefig('HN significant mean dose differences by metric.png', dpi=500)



2 Investigate differences

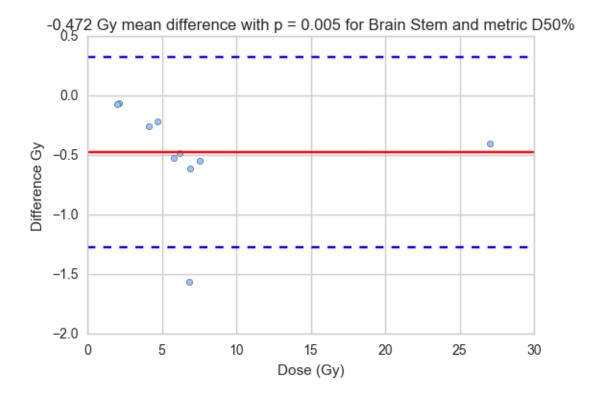
2.0.1 PTV1 DMAX

```
In [25]: structure = 'PTV1'
    metric = 'DMAX'
    plt.figure(figsize=(6, 4))
    bland_altman_plot(query_data(HN_df, 'Dm', structure, metric), query_data(House)
    plt.show()
    HN_df[(HN_df['structure'] == structure) & (HN_df['metric'] == metric)]
```



```
Case structure metric
Out [25]:
                                               AAA
                                                          Dm
                                                               Dm-AAA
                                                                         Dm-AAA%
         25
                Case1
                                         73.32000
                                                    72.99500 -0.32500 -0.443262
                            PTV1
                                   DMAX
                                         75.28158
         169
                Case2
                            PTV1
                                   DMAX
                                                    76.19172
                                                             0.91014
                                                                        1.208981
         313
                Case3
                            PTV1
                                   DMAX
                                         74.16500
                                                    74.83000 0.66500
                                                                        0.896649
         457
                Case4
                            PTV1
                                         72.92000
                                                    73.83500 0.91500
                                                                        1.254800
                                   DMAX
         583
                Case5
                            PTV1
                                   DMAX
                                         77.39100 76.94400 -0.44700 -0.577587
         745
                Case6
                            PTV1
                                   DMAX
                                         72.16110 76.38675 4.22565
                                                                        5.855856
         979
                Case7
                                         69.30066
                                                    73.59132
                                                              4.29066
                                                                        6.191370
                            PTV1
                                   DMAX
         1033
                Case8
                            PTV1
                                   DMAX
                                         72.68118
                                                    73.52631
                                                              0.84513
                                                                        1.162791
         1267
                                         72.68118
                                                    75.34659
                Case9
                            PTV1
                                   DMAX
                                                              2.66541
                                                                        3.667263
         1411
               Case10
                            PTV1
                                   DMAX
                                         72.87621
                                                    72.68118 -0.19503 -0.267618
```

2.0.2 Brain stem **D50%**

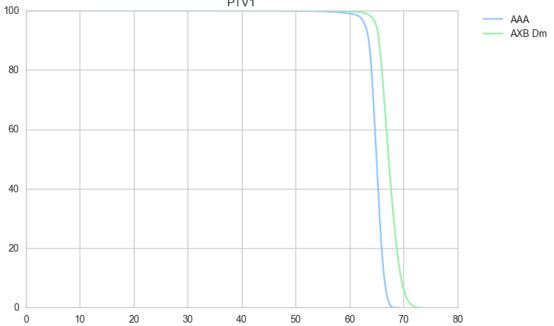


Out[26]:	Case	structure	metric	AAA	Dm	Dm-AAA	Dm-AAA
92	Case1	Brain Stem	D50%	27.252817	26.854172	-0.398645	-1.462765
236	Case2	Brain Stem	D50%	2.143964	2.082395	-0.061569	-2.871744
380	Case3	Brain Stem	D50%	4.236256	3.983629	-0.252627	-5.963459
506	Case4	Brain Stem	D50%	6.073990	5.556459	-0.517532	-8.520461
686	Case5	Brain Stem	D50%	2.046338	1.974176	-0.072162	-3.526414
830	Case6	Brain Stem	D50%	4.792651	4.578888	-0.213763	-4.460219
920	Case7	Brain Stem	D50%	7.637237	6.071817	-1.565420	-20.497204
1064	Case8	Brain Stem	D50%	6.425576	5.942164	-0.483412	-7.523243
1172	Case9	Brain Stem	D50%	7.804884	7.261682	-0.543202	-6.959772
1370	Case10	Brain Stem	D50%	7.165912	6.556099	-0.609812	-8.509905

3 Case 7

```
In [31]: HN_df[(HN_df['Case'] == 'Case7') & (HN_df['metric'] == 'DMAX')]
Out[31]:
               Case
                       structure metric
                                               AAA
                                                          Dm
                                                               Dm-AAA
                                                                         Dm-AAA%
         871
              Case7
                           CTV65
                                    DMAX
                                          69.30066
                                                    73.59132
                                                               4.29066
                                                                        6.191370
         889
              Case7
                       L Parotid
                                    DMAX
                                          57.79389
                                                    60.52431
                                                               2.73042
                                                                        4.724409
                                                    73.26627
         907
              Case7
                       R Parotid
                                   DMAX
                                          68.45553
                                                               4.81074
                                                                        7.027540
         925
              Case7
                      Brain Stem
                                    DMAX
                                          48.95253
                                                    50.64279
                                                               1.69026
                                                                        3.452855
         943
              Case7
                     Spinal Cord
                                    DMAX
                                          46.09209
                                                    47.39229
                                                               1.30020
                                                                        2.820874
         961
              Case7
                            Body
                                    DMAX 69.30066
                                                    73.59132
                                                               4.29066
                                                                       6.191370
```

```
979 Case7
                            PTV1
                                   DMAX 69.30066 73.59132 4.29066 6.191370
         997 Case7
                            PTV2
                                   DMAX 68.13048 72.68118 4.55070 6.679389
In [36]: from pyEclipseDVH import eclipse_DVH
In [40]: Case7_AAA = eclipse_DVH('Case7_AAA.txt')
         Case7_Dm = eclipse_DVH('Case7_Dm.txt')
Case7 AAA.txt loaded
                             patID = Case7 AAA
                                                        Prescription [Gy] = 65.010
Case7_Dm.txt loaded
                                                      Prescription [Gy] = 65.010
                            patID = Case7\_Dm
In [41]: Case7_AAA.DVH_df.columns
Out[41]: Index(['Artifact', 'GTV', 'GTVn', 'GTV + 15mm', 'GTVn + 10mm', 'CTV65',
                'CTV54', 'L Parotid', 'R Parotid', 'L Lens', 'R Lens', 'L Orbit',
                'R Orbit', 'Brain Stem', 'Spinal Cord', 'BS + 3mm', 'BS + 5mm', 'BG
                'Lt Par edit', 'PTV1', 'PTV1 edit', 'PTV2', 'PTV2 edit',
                'Pseudo Midline', 'Pseudo Oral Cav', 'Pseudo PostBrain', 'Pseudo Po
                'Rt Par edit', 'SC + 3mm', 'SC + 5mm', 'Pseudo PTV1', 'Pseudo PTV2
                'Left Couch Bar', 'Right Couch Bar', 'Annulus1', 'Dose 6175[cGy]',
                'Dose 5130[cGy]', 'missing1', 'missing2'],
               dtype='object')
In [42]: structure = 'PTV1'
         plt.plot(Case7_AAA.DVH_df[structure], label="AAA")
         plt.plot(Case7_Dm.DVH_df[structure], label="AXB Dm")
         plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)
         plt.title(structure)
Out[42]: <matplotlib.text.Text at 0xbd2e9d0>
                              PTV1
    100
                                                               AAA
                                                               AXB Dm
     80
```



In []: