

# HN analysis part 2 - ANALYSIS 15-12-2016

December 15, 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\scipy\stats\wilcoxon.py:191:
warnings.warn("Warning: sample size too small for normal approximation.")
```

```
Out[4]: 0.0075792819433897285
```

```
In [5]: 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 [6]: HN_df = pd.read_csv('HN_df_clean_28_11.csv') # read in the cleaned data
HN_df['Dm-AAA'] = HN_df['Dm'] - HN_df['AAA'] # get abs diff
HN_df['Dm-AAA%'] = diff_percent(HN_df['Dm'], HN_df['AAA'])
```

Exclude Case5 which has 54 Gy prescription

```
In [7]: to_exclude = ['Case5', 'Case7'] # has prescription of 54 Gy
to_include = list(set(HN_df['Case'].unique()) - set(to_exclude))
HN_df = HN_df[HN_df['Case'].isin(to_include)]
HN_df['Case'].unique()
```

```
Out[7]: array(['Case1', 'Case2', 'Case3', 'Case4', 'Case6', 'Case8', 'Case9',
              'Case10'], dtype=object)
```

```
In [8]: HN_df.head()
```

```
Out[8]:
```

	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

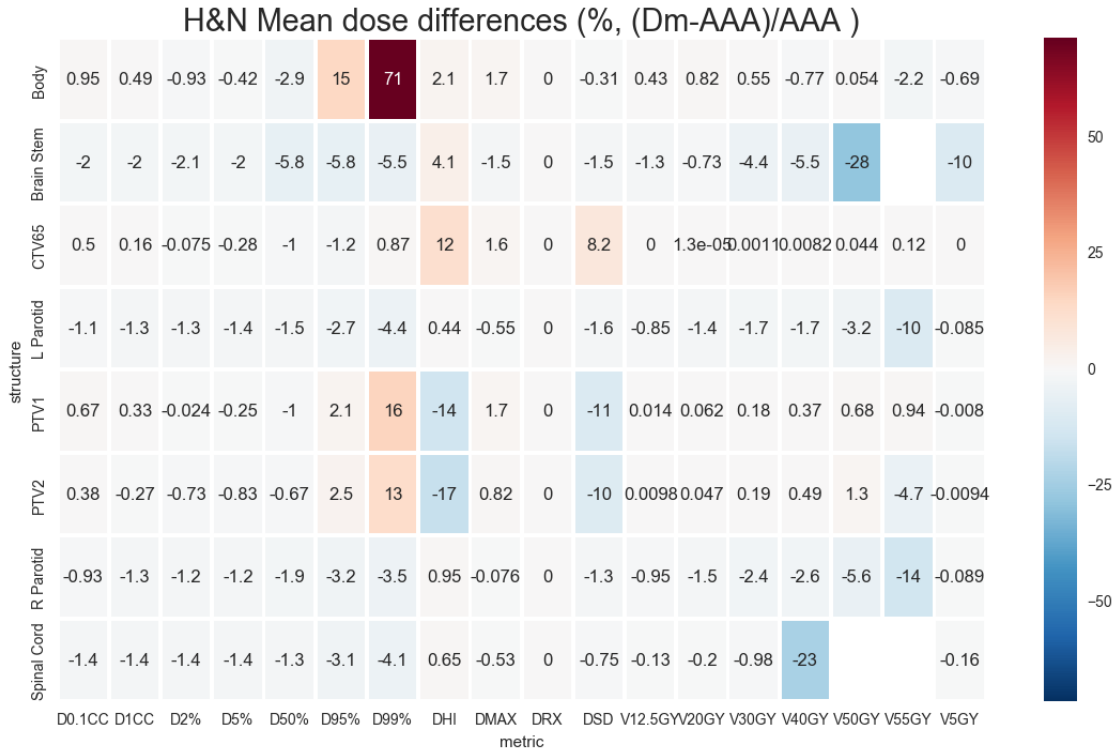
Everything imported OK

## 1.1 Get mean dose differences

```
In [9]: mean_diff_table = HN_df.groupby(['structure', 'metric'], as_index=False).mean()
```

```
In [10]: #metrics_of_interest = [ 'D95%', 'D50%', 'D99%', 'D0.1CC', 'DMAX'] # use
metrics_of_interest = mean_diff_table.columns # use a
```

```
In [11]: plt.figure(figsize=(14, 8))
ax1 = sns.heatmap(mean_diff_table[metrics_of_interest], annot=True, linewidths=1)
ax1.set_title('H&N Mean dose differences (% (Dm-AAA)/AAA )', size='xx-large')
plt.savefig('H&N Mean dose difference.png', dpi=200)
```



## 1.2 Get wilcoxon p value of differences

```
In [12]: metrics_of_interest = [ 'D95%', 'D50%', 'D0.1CC', 'DMAX' ] # use limited s
        #HN_df['structure'].unique() # all structures
        structures_of_interest = ['PTV1', 'PTV2', 'CTV65', 'L Parotid', 'R Parotid']

In [13]: i = len(structures_of_interest)
        j = len(metrics_of_interest)
        wilcox_data = np.zeros((i,j)) # init an empty array

In [14]: 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
```

C:\Users\RCole02.ROYALSURREY\AppData\Local\Continuum\Anaconda3\lib\site-packages\scipy\stats\wilcoxon.py:112: RuntimeWarning: sample size too small for normal approximation.

```

In [15]: 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 name

In [16]: # Keep for plotting all p-vals

#confidence = 1.0
#plt.figure(figsize=(16, 8))
#ax2 = sns.heatmap(wilcox_data_df[wilcox_data_df<confidence], annot=True,
#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)

In [17]: sub_diff_table = mean_diff_table[metrics_of_interest].loc[structures_of_interest]

In [18]: plt.figure(figsize=(18, 8))
data = sub_diff_table[wilcox_data_df<0.05]
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', s

#plt.savefig('HN significant mean dose differences by metric.png', dpi=500)
print('Mean differences')
print(sub_diff_table)
print('*****')
print('p-vals of differences')
print(wilcox_data_df)

```

Mean differences

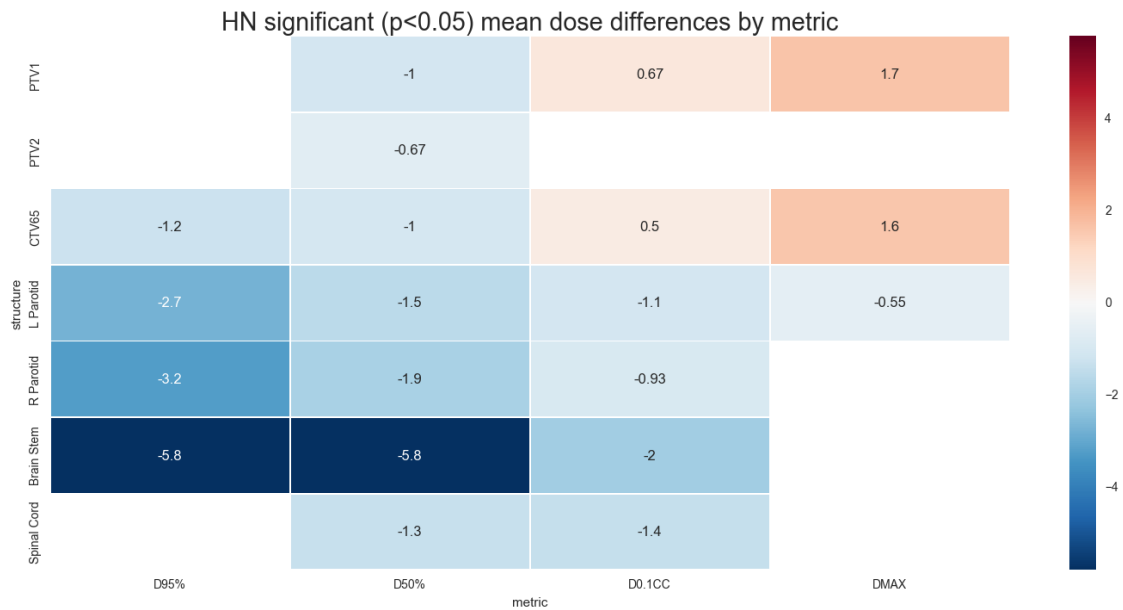
metric	D95%	D50%	D0.1CC	DMAX
structure				
PTV1	2.104196	-1.041277	0.674281	1.666932
PTV2	2.514504	-0.665242	0.377605	0.819839
CTV65	-1.239658	-1.039090	0.496371	1.584108
L Parotid	-2.725370	-1.549682	-1.059832	-0.545207
R Parotid	-3.243811	-1.857180	-0.934542	-0.076388
Brain Stem	-5.783298	-5.783946	-2.001884	-1.487810
Spinal Cord	-3.117142	-1.316931	-1.364837	-0.534718

\*\*\*\*\*

p-vals of differences

	D95%	D50%	D0.1CC	DMAX
PTV1	0.400814	0.011719	0.011719	0.035692
PTV2	0.092892	0.035692	0.483840	0.123485
CTV65	0.017290	0.011719	0.035692	0.017290
L Parotid	0.035692	0.011719	0.011719	0.035692
R Parotid	0.035692	0.017290	0.017290	0.483840
Brain Stem	0.017290	0.011719	0.011719	0.068704

Spinal Cord 0.092892 0.017290 0.011719 0.326989



## 2 Investigate differences

```
In [22]: def AAA_diff_plot(Dm, AAA, structure, metric, *args, **kwargs):
    Dm      = np.asarray(Dm)      # convert to arrays
    AAA      = np.asarray(AAA)
    z_stat, p_val = wilcoxon(Dm, AAA)

    diff_pct = diff_percent(Dm, AAA)
    AAA_norm_pct = diff_percent(AAA, AAA.mean()) # how far is a measurement from the mean

    plt.scatter(AAA_norm_pct, diff_pct, linewidths=3, *args, **kwargs)

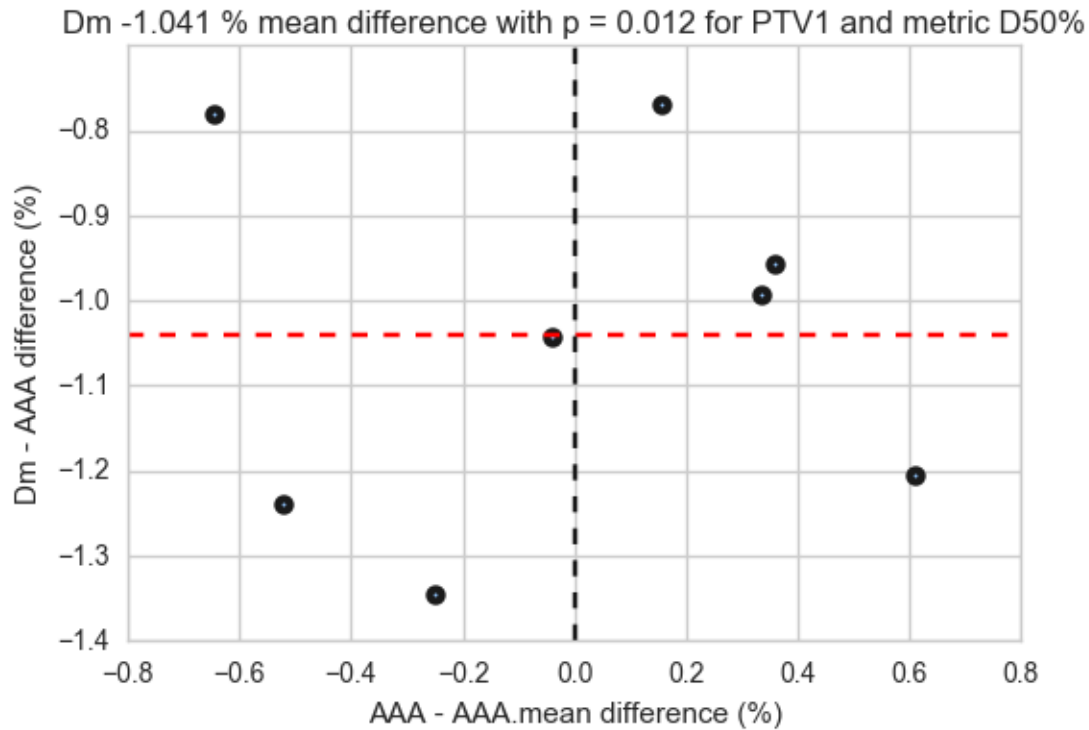
    plt.axhline(diff_pct.mean(), color='red', linestyle='--') # horitz line
    plt.axvline(0, color='k', linestyle='--') # v line

    plt.title('Dm ' + str(np.round(diff_pct.mean(), decimals = 3)) + ' % mean')
    plt.ylabel('Dm - AAA difference (%)')
    plt.xlabel('AAA - AAA.mean difference (%)')

In [23]: structure = 'PTV1'
metric = 'D50%'
plt.figure(figsize=(6, 4))
#bland_altman_plot(query_data(HN_df, 'Dm', structure, metric), query_data(HN_df, 'AAA', structure, metric))
AAA_diff_plot(query_data(HN_df, 'Dm', structure, metric), query_data(HN_df, 'AAA', structure, metric))
```

```
plt.show()
HN_df[(HN_df['structure'] == structure) & (HN_df['metric'] == metric)]
```

C:\Users\RCole02.ROYALSURREY\AppData\Local\Continuum\Anaconda3\lib\site-packages\sc  
warnings.warn("Warning: sample size too small for normal approximation.")

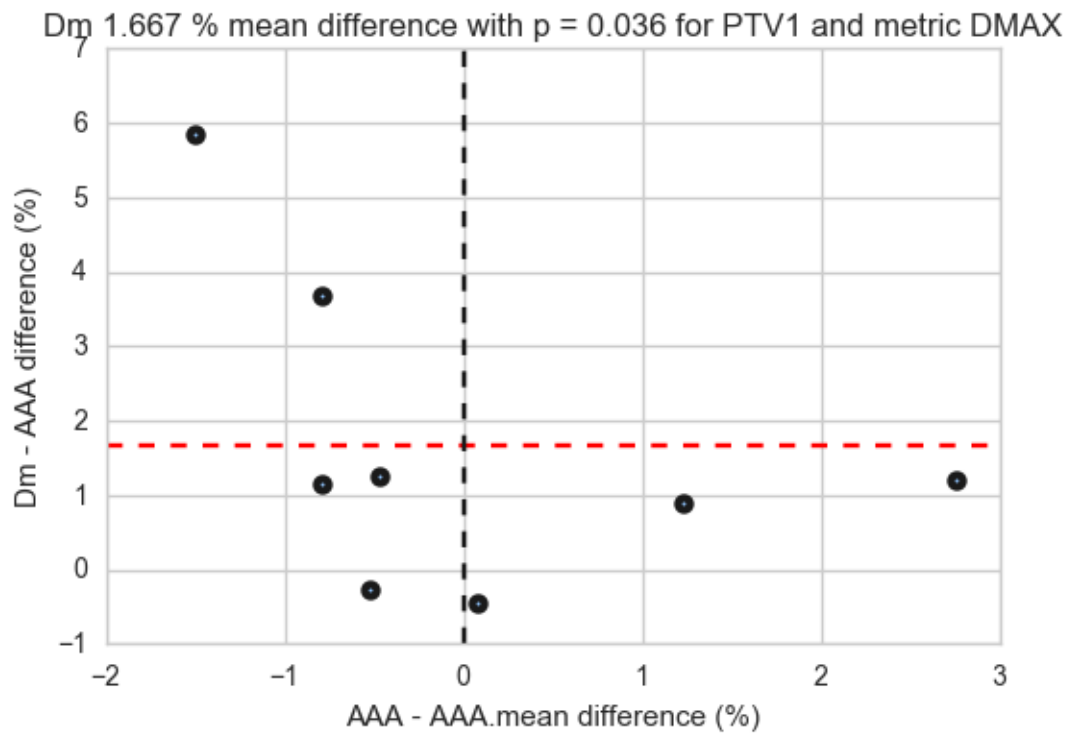


```
Out [23]:
```

	Case	structure	metric	AAA	Dm	Dm-AAA	Dm-AAA%
20	Case1	PTV1	D50%	68.559571	67.733731	-0.825840	-1.204558
164	Case2	PTV1	D50%	68.250510	67.725632	-0.524878	-0.769047
308	Case3	PTV1	D50%	68.372452	67.693337	-0.679114	-0.993257
452	Case4	PTV1	D50%	68.388653	67.734436	-0.654217	-0.956616
740	Case6	PTV1	D50%	67.787586	66.948429	-0.839157	-1.237922
1028	Case8	PTV1	D50%	68.116240	67.405729	-0.710511	-1.043086
1262	Case9	PTV1	D50%	67.705087	67.176668	-0.528418	-0.780471
1406	Case10	PTV1	D50%	67.973479	67.059061	-0.914418	-1.345257

```
In [24]: structure = 'PTV1'
metric = 'D50%'
plt.figure(figsize=(6, 4))
AAA_diff_plot(query_data(HN_df, 'Dm', structure, metric), query_data(HN_df, 'AAA', structure, metric))
#bland_altman_plot(query_data(HN_df, 'Dm', structure, metric), query_data(HN_df, 'AAA', structure, metric))
plt.show()
HN_df[(HN_df['structure'] == structure) & (HN_df['metric'] == metric)]
```

```
C:\Users\RCole02.ROYALSURREY\AppData\Local\Continuum\Anaconda3\lib\site-packages\sc
warnings.warn("Warning: sample size too small for normal approximation.")
```



```
Out[24]:
```

	Case	structure	metric	AAA	Dm	Dm-AAA	Dm-AAA%
25	Case1	PTV1	DMAX	73.32000	72.99500	-0.32500	-0.443262
169	Case2	PTV1	DMAX	75.28158	76.19172	0.91014	1.208981
313	Case3	PTV1	DMAX	74.16500	74.83000	0.66500	0.896649
457	Case4	PTV1	DMAX	72.92000	73.83500	0.91500	1.254800
745	Case6	PTV1	DMAX	72.16110	76.38675	4.22565	5.855856
1033	Case8	PTV1	DMAX	72.68118	73.52631	0.84513	1.162791
1267	Case9	PTV1	DMAX	72.68118	75.34659	2.66541	3.667263
1411	Case10	PTV1	DMAX	72.87621	72.68118	-0.19503	-0.267618

```
In [ ]:
```