Final_Plots-Copy1

September 16, 2021

```
[2]: import numpy as np
  import pandas as pd
  import matplotlib.pyplot as plt
  import seaborn as sns
  import os
[50]: input_path = 'Z:/V11/vM1_Ref_Frame/FINAL/'
  cm = 1/2.54
```

1 Clustering

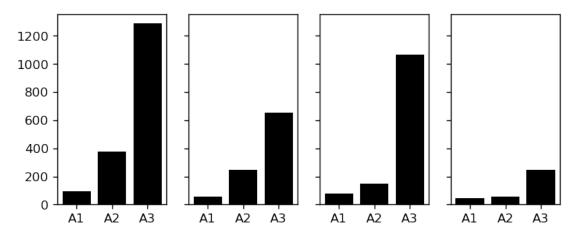
1.1 Absolute rabies counts and cov

```
[71]: # the total number of rabies labelled neurons varies a lot across 3 animals
      df_rabies_counts = pd.read_csv(input_path + 'Fig_1/Rabies_Distribution/
       →Clustering/'+'landmarks_counts.csv',index_col=0)
      #mq48 count = df rabies counts['MG48 lhs vM1']
      mg_48_count_total = df_rabies_counts[df_rabies_counts.
       →index==0][['MG48_lhs_vM1','MG48_rhs_vM1','MG48_lhs_vS1','MG48_rhs_vS1']].
       \rightarrow values [0] . sum()
      mg_48_count_total_lhs = df_rabies_counts[df_rabies_counts.
       \rightarrowindex==0][['MG48_lhs_vM1','MG48_lhs_vS1']].values[0].sum()
      mg_48_count_total_rhs = df_rabies_counts[df_rabies_counts.
       \rightarrowindex==0][['MG48_rhs_vM1','MG48_rhs_vS1']].values[0].sum()
      mg_49_count_total = df_rabies_counts[df_rabies_counts.
       →index==0][['MG49_lhs_vM1','MG49_rhs_vM1','MG49_lhs_vS1','MG49_rhs_vS1']].
       \rightarrow values [0].sum()
      mg_49_count_total_lhs = df_rabies_counts[df_rabies_counts.
       \rightarrowindex==0][['MG49_lhs_vM1','MG49_lhs_vS1']].values[0].sum()
      mg_49_count_total_rhs = df_rabies_counts[df_rabies_counts.

→index==0] [['MG49_rhs_vM1', 'MG49_rhs_vS1']].values[0].sum()
      mg_50_count_total = df_rabies_counts[df_rabies_counts.
       \rightarrow \texttt{index==0} \texttt{[['MG50\_lhs\_vM1','MG50\_rhs\_vM1','MG50\_lhs\_vS1','MG50\_rhs\_vS1']]}.
       \rightarrow values [0] . sum()
```

[71]: (1452.0, 1295.5155987739656, 89.2228373811271)

```
[72]: fig, axes = plt.subplots(nrows=1,ncols=4,__
       ⇒sharex=True, sharey=True, figsize=(18*cm, 7*cm), dpi=120)
      axes[0].bar(['A1','A2','A3'],df_rabies_counts[df_rabies_counts.
       \rightarrowindex==0] [['MG48_lhs_vM1','MG49_lhs_vM1','MG50_lhs_vM1']].
       \rightarrow values [0], color=(0,0,0))
      axes[1].bar(['A1','A2','A3'],df_rabies_counts[df_rabies_counts.
       →index==0][['MG48_rhs_vM1','MG49_rhs_vM1','MG50_rhs_vM1']].
       \rightarrow values [0], color=(0,0,0))
      axes[2].bar(['A1','A2','A3'],df_rabies_counts[df_rabies_counts.
       \rightarrowindex==0] [['MG48_lhs_vS1','MG49_lhs_vS1','MG50_lhs_vS1']].
       \rightarrow values[0],color=(0,0,0))
      axes[3].bar(['A1','A2','A3'],df_rabies_counts[df_rabies_counts.
       \rightarrowindex==0][['MG48_rhs_vS1','MG49_rhs_vS1','MG50_rhs_vS1']].
       \rightarrow values[0],color=(0,0,0))
      fig.savefig(input path + 'Fig 1/Rabies Distribution/Clustering/total counts.
        ⇔eps')
```



```
[73]: df_rabies_counts[df_rabies_counts.
       →index==0][['MG48_lhs_vM1','MG49_lhs_vM1','MG50_lhs_vM1']]
[73]:
              MG48_lhs_vM1 MG49_lhs_vM1 MG50_lhs_vM1
      Cutoff
      0
                        96
                                      376
                                                    1292
[74]: vm1_lhs = df_rabies_counts[df_rabies_counts.
       →index==0][['MG48_lhs_vM1','MG49_lhs_vM1','MG50_lhs_vM1']]
      vm1 lhs.mean(axis=1),vm1 lhs.std(axis=1),vm1 lhs.std(axis=1)/vm1 lhs.
       \rightarrowmean(axis=1),
[74]: (Cutoff
            588.0
       dtype: float64,
       Cutoff
            625.549359
       dtype: float64,
       Cutoff
            1.063859
       dtype: float64)
[75]: df_rabies_counts[df_rabies_counts.

→index==0][['MG48_rhs_vM1','MG49_rhs_vM1','MG50_rhs_vM1']]
[75]:
              MG48_rhs_vM1 MG49_rhs_vM1 MG50_rhs_vM1
      Cutoff
      0
                        58
                                      245
                                                     653
[76]: vm1_lhs = df_rabies_counts[df_rabies_counts.
       →index==0][['MG48_rhs_vM1','MG49_rhs_vM1','MG50_rhs_vM1']]
      vm1_lhs.mean(axis=1),vm1_lhs.std(axis=1),vm1_lhs.std(axis=1)/vm1_lhs.
       \rightarrowmean(axis=1),
[76]: (Cutoff
            318.666667
       dtype: float64,
       Cutoff
            304.263592
       dtype: float64,
       Cutoff
            0.954802
       dtype: float64)
[77]: df_rabies_counts[df_rabies_counts.

→index==0][['MG48_lhs_vS1','MG49_lhs_vS1','MG50_lhs_vS1']]
```

```
[77]:
              MG48_lhs_vS1 MG49_lhs_vS1 MG50_lhs_vS1
      Cutoff
                         76
      0
                                      149
                                                    1066
[78]: vm1_lhs = df_rabies_counts[df_rabies_counts.
       \rightarrowindex==0][['MG48_lhs_vS1','MG49_lhs_vS1','MG50_lhs_vS1']]
      vm1 lhs.mean(axis=1),vm1 lhs.std(axis=1),vm1 lhs.std(axis=1)/vm1 lhs.
       \rightarrowmean(axis=1),
[78]: (Cutoff
            430.333333
       dtype: float64,
       Cutoff
            551.712183
       dtype: float64,
       Cutoff
            1.282058
       dtype: float64)
[79]: df_rabies_counts[df_rabies_counts.
       →index==0][['MG48_rhs_vS1','MG49_rhs_vS1','MG50_rhs_vS1']]
[79]:
              MG48_rhs_vS1 MG49_rhs_vS1 MG50_rhs_vS1
      Cutoff
      0
                         43
                                       57
                                                     245
[80]: vm1_lhs = df_rabies_counts[df_rabies_counts.
       →index==0][['MG48_rhs_vS1','MG49_rhs_vS1','MG50_rhs_vS1']]
      vm1_lhs.mean(axis=1),vm1_lhs.std(axis=1),vm1_lhs.std(axis=1)/vm1_lhs.
       \rightarrowmean(axis=1),
[80]: (Cutoff
            115.0
       dtype: float64,
       Cutoff
            112.800709
       dtype: float64,
       Cutoff
            0.980876
       dtype: float64)
[81]: np.array([1.063859,0.954802,1.282058,0.980876]).mean()
[81]: 1.07039875
```

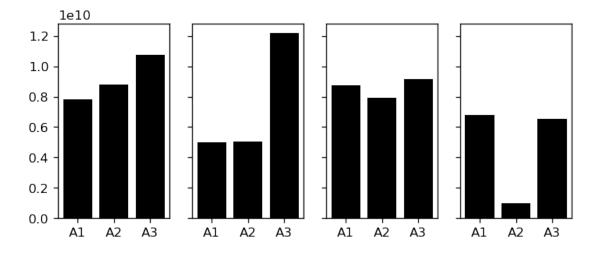
1.2 Clustered rabies volumes and cov

```
[82]: df_surf_vols_amira = pd.read_csv(input_path + 'Fig_1/Rabies_Distribution/
       →Clustering/surf_volumes_from_amira.csv',\
                                          sep=';',thousands='.',index_col=0)
      mg_48_volume_total = df_surf_vols_amira[df_surf_vols_amira.
       →index==10][['MG48_lhs_vM1','MG48_rhs_vM1','MG48_lhs_vS1','MG48_rhs_vS1']].
       \rightarrow values [0].sum()
      mg_48_volume_total_lhs = df_surf_vols_amira[df_surf_vols_amira.
       \rightarrowindex==10][['MG48_lhs_vM1','MG48_lhs_vS1']].values[0].sum()
      mg_48_volume_total_rhs = df_surf_vols_amira[df_surf_vols_amira.
       \rightarrowindex==10][['MG48_rhs_vM1','MG48_rhs_vS1']].values[0].sum()
      mg_49_volume_total = df_surf_vols_amira[df_surf_vols_amira.
       →index==10][['MG49_lhs_vM1','MG49_rhs_vM1','MG49_lhs_vS1','MG49_rhs_vS1']].
       \rightarrow values [0] . sum()
      mg_49_volume_total_lhs = df_surf_vols_amira[df_surf_vols_amira.
       \rightarrowindex==10][['MG49_lhs_vM1','MG49_lhs_vS1']].values[0].sum()
      mg 49 volume total rhs = df_surf_vols_amira[df_surf_vols_amira.
       \rightarrowindex==10][['MG49_rhs_vM1','MG49_rhs_vS1']].values[0].sum()
      mg_50_volume_total = df_surf_vols_amira[df_surf_vols_amira.
       \rightarrow \texttt{index} == 10] \texttt{[['MG50\_lhs\_vM1', 'MG50\_rhs\_vM1', 'MG50\_lhs\_vS1', 'MG50\_rhs\_vS1']]}.
       \rightarrow values [0] . sum()
      mg_50_volume_total_lhs = df_surf_vols_amira[df_surf_vols_amira.
       \rightarrowindex==10][['MG50_lhs_vM1','MG50_lhs_vS1']].values[0].sum()
      mg_50_volume_total_rhs = df_surf_vols_amira[df_surf_vols_amira.
       →index==10][['MG50_rhs_vM1','MG50_rhs_vS1']].values[0].sum()
      totals = np.array([mg_48_volume_total,mg_49_volume_total,mg_50_volume_total])
      totals_lhs = np.
       array([mg 48 volume total lhs,mg 49 volume total lhs,mg 50 volume total lhs])
      totals rhs = np.
       -array([mg_48_volume_total_rhs,mg_49_volume_total_rhs,mg_50_volume_total_rhs])
      totals.mean(),totals.std(),totals.std()/totals.mean()*100
[82]: (29918853888.0, 6588226373.641311, 22.02031668159504)
[83]: totals_lhs.mean(),totals_lhs.std(),totals_lhs.std()/totals_lhs.mean()*100
[83]: (17738112000.0, 1535953136.6981075, 8.659056480746695)
[84]: totals_rhs.mean(),totals_rhs.std(),totals_rhs.std()/totals_rhs.mean()*100
[84]: (12180741888.0, 5207223830.138087, 42.749644299318454)
```

```
[85]: fig, axes = plt.subplots(nrows=1,ncols=4,__
       ⇒sharex=True, sharey=True, figsize=(18*cm, 7*cm), dpi=120)
      axes[0].bar(['A1','A2','A3'],df surf vols amira[df surf vols amira.
       \rightarrow \texttt{index==}10\texttt{][['MG48\_lhs\_vM1','MG49\_lhs\_vM1','MG50\_lhs\_vM1']]}.
       \rightarrow values[0],color=(0,0,0))
      axes[1].bar(['A1','A2','A3'],df_surf_vols_amira[df_surf_vols_amira.

→index==10][['MG48_rhs_vM1','MG49_rhs_vM1','MG50_rhs_vM1']].
       \rightarrow values[0],color=(0,0,0))
      axes[2].bar(['A1','A2','A3'],df_surf_vols_amira[df_surf_vols_amira.

→index==10][['MG48 lhs vS1','MG49 lhs vS1','MG50 lhs vS1']].
       \rightarrow values[0],color=(0,0,0))
      axes[3].bar(['A1','A2','A3'],df_surf_vols_amira[df_surf_vols_amira.
       \rightarrowindex==10][['MG48_rhs_vS1','MG49_rhs_vS1','MG50_rhs_vS1']].
       \rightarrow values[0],color=(0,0,0))
      fig.savefig(input_path + 'Fig_1/Rabies_Distribution/Clustering/total_volumes.
        ⇔eps')
```



[87]: (Cutoff 10 9.118986e+09

```
dtype: float64,
       Cutoff
       10
             1.489716e+09
       dtype: float64,
       Cutoff
       10
             0.163364
       dtype: float64)
[88]: df_surf_vols_amira[df_surf_vols_amira.
       \rightarrowindex==10][['MG48_rhs_vM1','MG49_rhs_vM1','MG50_rhs_vM1']]
[88]:
              MG48_rhs_vM1 MG49_rhs_vM1 MG50_rhs_vM1
      Cutoff
      10
                              5035372032
                5000535552
                                            12209556480
[89]: bla = df_surf_vols_amira[df_surf_vols_amira.
       →index==10][['MG48_rhs_vM1','MG49_rhs_vM1','MG50_rhs_vM1']]
      bla.mean(axis=1),bla.std(axis=1),bla.std(axis=1)/bla.mean(axis=1),
[89]: (Cutoff
       10
             7.415155e+09
       dtype: float64,
       Cutoff
       10
             4.152110e+09
       dtype: float64,
       Cutoff
             0.559949
       10
       dtype: float64)
[90]: df_surf_vols_amira[df_surf_vols_amira.

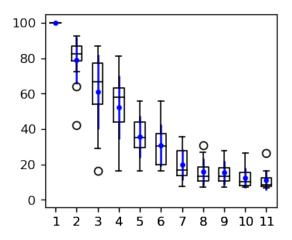
→index==10][['MG48_lhs_vS1','MG49_lhs_vS1','MG50_lhs_vS1']]
[90]:
              MG48_lhs_vS1 MG49_lhs_vS1 MG50_lhs_vS1
      Cutoff
      10
                8765255680
                              7920558080
                                             9171564544
[91]: bla = df_surf_vols_amira[df_surf_vols_amira.
       →index==10][['MG48_lhs_vS1','MG49_lhs_vS1','MG50_lhs_vS1']]
      bla.mean(axis=1),bla.std(axis=1),bla.std(axis=1)/bla.mean(axis=1),
[91]: (Cutoff
       10
             8.619126e+09
       dtype: float64,
       Cutoff
       10
             6.381768e+08
       dtype: float64,
       Cutoff
```

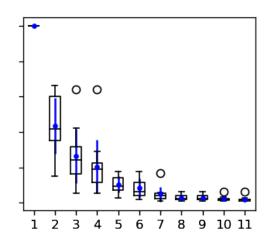
```
10
            0.074042
      dtype: float64)
[92]: df_surf_vols_amira[df_surf_vols_amira.
      →index==10][['MG48 rhs vS1','MG49 rhs vS1','MG50 rhs vS1']]
[92]:
            MG48_rhs_vS1 MG49_rhs_vS1 MG50_rhs_vS1
     Cutoff
     10
               6777475072
                            978758912
                                        6540527616
[93]: bla = df_surf_vols_amira[df_surf_vols_amira.

→index==10][['MG48_rhs_vS1','MG49_rhs_vS1','MG50_rhs_vS1']]
     bla.mean(axis=1),bla.std(axis=1),bla.std(axis=1)/bla.mean(axis=1),
[93]: (Cutoff
      10
            4.765587e+09
      dtype: float64,
      Cutoff
            3.281629e+09
      10
      dtype: float64,
      Cutoff
      10
            0.68861
      dtype: float64)
     1.3 Effect of density cutoff thersholds on clustering
[94]: df_rabies_counts = pd.read_csv(input_path + 'Fig_1/Rabies_Distribution/
      →Clustering/'+'landmarks_counts.csv',index_col=0)
     df_surf_areas = pd.read_csv(input_path + 'Fig_1/Rabies_Distribution/Clustering/
      df_surf_volumes_per = pd.read_csv(input_path + 'Fig_1/Rabies_Distribution/
      [95]: df_counts = df_rabies_counts.filter(regex='per').transpose()
     df surfs = df surf areas.filter(regex='per').transpose()
     fig, axes = plt.subplots(nrows=1,ncols=2,__
      ⇒sharex=True, sharey=True, figsize=(18*cm, 7*cm), dpi=120)
     b1 = axes[0].boxplot(df counts[df counts.columns[::-1]])
     eb2 = axes[0].errorbar(x=[1,2,3,4,5,6,7,8,9,10,11],y=np.
      →array(df_counts[df_counts.columns[::-1]].transpose()).mean(axis=1),yerr=np.
      →array(df_counts[df_counts.columns[::-1]].transpose()).std(axis=1),\
                 color='blue',marker='.',fmt='.')
     b2 = axes[1].boxplot(df_surfs[df_surfs.columns[::-1]])
     eb2 = axes[1].errorbar(x=[1,2,3,4,5,6,7,8,9,10,11],y=np.array(df_surfs[df_surfs.
      →columns[::-1]].transpose()).mean(axis=1),yerr=np.array(df_surfs[df_surfs.
      color='blue',marker='.',fmt='.')
```

```
plt.setp(b1['boxes'], color='black')
plt.setp(b1['whiskers'], color='black')
plt.setp(b1['medians'], color='black')
plt.setp(b1['caps'], color='black')
plt.setp(b2['boxes'], color='black')
plt.setp(b2['whiskers'], color='black')
plt.setp(b2['medians'], color='black')
plt.setp(b2['caps'], color='black')
fig.savefig(input_path + 'Fig_1/Rabies_Distribution/Clustering/density_cutoffs.

oeps')
```





```
[96]: (Cutoff
       0
              100.000000
       10
               78.972736
       20
               61.044467
       30
               52.317847
       40
               35.613992
       50
               30.848289
       60
               20.047840
               15.788484
       70
       80
               15.332909
       90
               12.629794
       100
               11.061912
       dtype: float64,
       Cutoff
               0.000000
```

```
10
              13.959851
       20
              21.806309
       30
               18.581325
       40
              12.173787
       50
               12.421827
       60
                9.206613
       70
                7.951789
       80
                7.135830
       90
                5.902816
       100
                5.772488
       dtype: float64,
       Cutoff
              0.000000
       10
              0.176768
       20
              0.357220
       30
              0.355162
       40
              0.341826
       50
              0.402675
       60
              0.459232
       70
              0.503645
       80
              0.465393
       90
              0.467372
       100
              0.521835
       dtype: float64)
[97]: df_surfs[df_surfs.columns[::-1]].mean(),df_surfs[df_surfs.columns[::-1]].

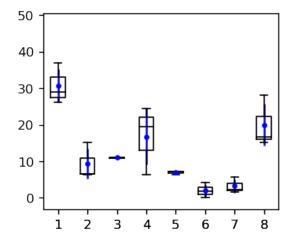
→std(),df_surfs[df_surfs.columns[::-1]].std()/df_surfs[df_surfs.columns[::
       \rightarrow-1]].mean()
[97]: (Cutoff
       0
               100.000000
       10
                43.652848
       20
                26.505614
       30
                20.498617
       40
                10.416689
       50
                 8.327781
       60
                 4.837567
       70
                 3.024383
       80
                 2.952427
       90
                 2.244578
       100
                 2.007696
       dtype: float64,
       Cutoff
       0
                0.000000
       10
               16.398099
       20
               16.109189
       30
               15.614610
```

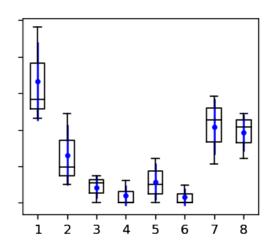
```
40
        4.843768
50
        5.562739
60
        4.200960
70
        1.617775
80
        1.555020
90
        1.361272
100
        1.413926
dtype: float64,
Cutoff
0
       0.000000
10
       0.375648
20
       0.607765
30
       0.761740
40
       0.465001
50
       0.667974
60
       0.868403
70
       0.534911
80
       0.526692
90
       0.606471
100
       0.704253
dtype: float64)
```

2 Rabies Distribution in Atlas

```
[98]: df_atlas = pd.read_csv(input_path + 'Fig_1/Rabies_Distribution/Atlas/
       df_atlas_clustered = pd.read_csv(input_path + 'Fig_1/Rabies_Distribution/
       →Atlas_Clustered/'+'Atlas_Counts_per.csv',index_col=0)
      df = df_atlas_clustered
[99]: df_atlas_clustered.index
[99]: Index(['M1', 'M2', 'S1_ULp', 'SBF', 'S2', 'S1_fl', 'Insular', 'Others'],
      dtype='object', name='Atlas_Region')
[100]: df lhs = df[['lhs MG48','lhs MG49','lhs MG50']]
      df_rhs = df[['rhs_MG48','rhs_MG49','rhs_MG50']]
      \#df [lhs.mean(axis=1), df [lhs.std(axis=1), df [rhs.mean(axis=1), df [rhs.std(axis=1)]
      #df = pd.DataFrame()
      #df = df.append(df_lhs)
      #df = df.append(df_rhs)
      #df_lhs.transpose().plot(kind='box', stacked=True)
      \#df\_rhs.transpose().plot(kind='box', stacked=True)
      #axes.boxplot(df_lhs.transpose(),)
      #axes.boxplot(df_rhs.transpose())
```

```
#axes.set_edgecolor('red')
fig, axes = plt.subplots(nrows=1,ncols=2,__
  ⇒sharex=True, sharey=True, figsize=(18*cm, 7*cm), dpi=120)
\#axes.errorbar(df\_lhs.index.values, df\_lhs.transpose().mean(axis=0), yerr=df\_lhs.
  \rightarrow transpose().std(axis=0))
\#axes.errorbar(df\_rhs.index.values, df\_rhs.transpose().mean(axis=0), yerr=df\_rhs.transpose().mean(axis=0), yerr=df\_rhs.transpose().mean(axis=0),
  \hookrightarrow transpose().std(axis=0))
b1 = axes[0].boxplot(df_lhs.transpose())
eb1 = axes[0].errorbar(x=[1,2,3,4,5,6,7,8],y=np.array(df_lhs).
  →mean(axis=1),yerr=np.array(df_lhs).std(axis=1),\
                                     color='blue',marker='.',fmt='.')
plt.setp(b1['boxes'], color='black')
plt.setp(b1['whiskers'], color='black')
plt.setp(b1['medians'], color='black')
plt.setp(b1['caps'], color='black')
b2 = axes[1].boxplot(df_rhs.transpose())
eb2 = axes[1].errorbar(x=[1,2,3,4,5,6,7,8],y=np.array(df_rhs).
  →mean(axis=1),yerr=np.array(df_rhs).std(axis=1),\
                                     color='blue',marker='.',fmt='.')
plt.setp(b2['boxes'], color='black')
plt.setp(b2['whiskers'], color='black')
plt.setp(b2['medians'], color='black')
plt.setp(b2['caps'], color='black')
fig.savefig(input_path + 'Fig_1/Rabies Distribution/Atlas_Clustered/
   →atlas_counts.eps')
```





 $[101]: df_{lhs.mean(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.mean(axis=1)})}, df_{lhs.std(axis=1),df_{lhs.mean(axis=1)}/df_{lhs.mean(axis=1)}, df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.mean(axis=1)}, df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.mean(axis=1)}, df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.std(axis=1)}, df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.std(axis=1)}, df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.std(axis=1)}/df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1)}/df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.std(axis=1),df_{lhs.st$

```
[101]: (Atlas_Region
                   30.680840
        M2
                    9.421677
        S1_ULp
                   10.981062
        SBF
                   16.776272
        S2
                    6.984500
        S1 fl
                    1.936321
        Insular
                    3.236422
        Others
                   19.982906
        dtype: float64,
        Atlas_Region
        M1
                   5.633135
        M2
                   5.020520
        S1_ULp
                   0.169554
        SBF
                   9.290977
        S2
                   0.469766
        S1_fl
                   2.059179
        Insular
                   2.252403
        Others
                   6.977257
        dtype: float64,
        Atlas_Region
        M1
                   0.183604
        M2
                   0.532869
        S1_ULp
                   0.015441
        SBF
                   0.553817
        S2
                   0.067258
        S1_fl
                   1.063449
        Insular
                   0.695955
        Others
                   0.349161
        dtype: float64)
[102]: df_rhs.mean(axis=1),df_rhs.std(axis=1),df_rhs.std(axis=1)/df_rhs.mean(axis=1)
[102]: (Atlas_Region
                   33.224430
        M1
        М2
                   13.061121
                    4.231139
        S1_ULp
        SBF
                    2.077323
        S2
                    5.767550
        S1_f1
                    1.612903
        Insular
                   20.864665
        Others
                   19.160868
        dtype: float64,
        Atlas_Region
        M1
                   13.159230
        M2
                   10.152341
        S1_ULp
                    3.790582
```

```
SBF
             3.482277
S2
             6.124294
S1_fl
             2.793630
Insular
             9.460580
Others
             6.298316
dtype: float64,
Atlas_Region
M1
            0.396071
M2
            0.777295
S1_ULp
            0.895877
SBF
            1.676329
S2
            1.061854
S1 f1
            1.732051
Insular
            0.453426
Others
            0.328707
dtype: float64)
```

3 Variability

```
[103]: df = pd.read_csv(input_path + 'Fig_1/Rabies_Distribution/Variability/
        →'+'variability.csv',index_col=0)
[104]: df
[104]:
                                             surf_area rabies_surf_vol
                                                                               surf_vol \
             Hem
                   BR
                      rabies_surf_area
       Exp
       MG48
             lhs
                  vM1
                            3.864979e+07
                                          5.949542e+07
                                                            9.998786e+09
                                                                          2.933011e+10
       MG49
             lhs
                  vM1
                            3.100946e+07
                                          5.446138e+07
                                                            8.817117e+09
                                                                          2.770998e+10
      MG50
             lhs
                  vM1
                            4.487086e+07
                                          6.644812e+07
                                                            1.073652e+10
                                                                          3.424272e+10
       MG48
             rhs
                  vM1
                            4.027432e+07
                                          5.095105e+07
                                                            8.319941e+09
                                                                          2.491589e+10
       MG49
             rhs
                  vM1
                            2.457457e+07
                                          4.002189e+07
                                                            8.473321e+09
                                                                          1.525190e+10
       MG50
             rhs
                  vM1
                            4.538989e+07
                                          7.190225e+07
                                                            1.220956e+10
                                                                          3.769797e+10
      MG48
             lhs
                  vS1
                            3.517157e+07
                                          7.991888e+07
                                                            8.765256e+09
                                                                          2.227071e+10
      MG49
             lhs
                  vS1
                            3.440360e+07
                                          5.075986e+07
                                                            7.920558e+09
                                                                          2.250900e+10
       MG50
                  vS1
             lhs
                            3.654979e+07
                                          6.439866e+07
                                                            1.155127e+10
                                                                          3.034522e+10
       MG48
             rhs
                  vS1
                            3.642832e+07
                                          5.875139e+07
                                                            5.389144e+09
                                                                          2.025440e+10
       MG49
             rhs
                  vS1
                                                                          1.421633e+09
                            7.360863e+06
                                          8.591115e+06
                                                            9.787589e+08
       MG50
             rhs
                  vS1
                            2.953847e+07
                                          6.121407e+07
                                                            6.540527e+09
                                                                          2.753162e+10
             pia_surf_area wm_surf_area pca0_extent
                                                        pca1_extent
       Exp
      MG48
              3.405203e+07
                            7.648197e+06
                                           5495.218316
                                                         2438.709082
      MG49
              3.153497e+07
                             6.590995e+06
                                           4363.242266
                                                         2817.266806
      MG50
              3.756157e+07
                             1.037292e+07
                                           5190.418578
                                                         3059.023762
       MG48
              2.965448e+07
                             4.680838e+06
                                           4566.534440
                                                         2312.278507
       MG49
              1.924605e+07
                             4.796520e+06
                                           3853.662749
                                                         1645.224640
```

```
MG50
       4.467729e+07
                      1.128978e+07
                                     4991.162955
                                                  2963.189153
MG48
       3.345650e+07
                      3.569468e+07
                                     4670.086117
                                                   3047.370760
MG49
       1.796299e+07
                      9.505442e+06
                                     3733.429637
                                                   3224.832867
MG50
       2.672434e+07
                      1.409410e+07
                                     4993.667943
                                                   3630.501278
MG48
       3.147497e+07
                      1.746175e+07
                                                   2663.134437
                                     3855.213838
MG49
       1.872575e+06
                      6.059152e+05
                                     1631.573866
                                                   1442.449528
                                     5237.160116
                                                  3096.969710 ...
MG50
       2.593566e+07
                      1.260094e+07
      pca0_angle_Z_axis pca1_angle_X_axis pca1_angle_Y_axis \
Exp
MG48
              130.450097
                                   13.544005
                                                       76.455995
MG49
              124.009155
                                    2.757566
                                                       87.242434
MG50
              125.523610
                                    8.952160
                                                       81.047840
MG48
              117.988712
                                  170.214887
                                                       80.214887
MG49
              94.926912
                                  175.890868
                                                       85.890868
MG50
              109.686205
                                  166.053442
                                                       76.053442
MG48
              125.021209
                                   59.618737
                                                       30.381263
MG49
              117.787791
                                   56.662826
                                                       33.337174
MG50
              130.446944
                                   52.095473
                                                       37.904527
MG48
              110.233981
                                   86.120410
                                                        3.879590
MG49
              121.481232
                                  101.552473
                                                       11.552473
MG50
              74.921986
                                  110.241371
                                                       20.241371
      pca1 angle Z axis pca2 angle X axis pca2 angle Y axis
Exp
MG48
              170.874873
                                   95.607085
                                                        5.607085
MG49
              168.582760
                                   88.529888
                                                        1.470112
MG50
                                  102.546421
                                                       12.546421
                6.103461
MG48
                8.778636
                                   85.023098
                                                        4.976902
MG49
               30.609510
                                   87.485591
                                                        2.514409
MG50
                                                        0.253644
               21.252799
                                   90.253644
MG48
               32.448471
                                  160.362035
                                                       70.362035
MG49
                                                       70.358336
               17.305465
                                  160.358336
MG50
               34.382812
                                  166.764675
                                                       76.764675
MG48
               18.424177
                                    5.416217
                                                       84.583783
MG49
                7.722050
                                    0.068894
                                                       89.931106
MG50
              166.212065
                                   20.395848
                                                       69.604152
      pca2 angle Z axis Dist From Bregma X Dist From Bregma Y \
Exp
MG48
               41.621342
                                  2108.854392
                                                       3243.377530
MG49
               34.622195
                                  2231.130414
                                                       3088.905952
MG50
               35.552018
                                  2210.779601
                                                       2989.982653
MG48
               28.937604
                                  2382.954455
                                                       3379.329850
MG49
                2.115601
                                  2285.709061
                                                       3883.483915
MG50
              21.421980
                                  2627.218475
                                                       2788.760230
MG48
              160.158370
                                  4810.509570
                                                      -1019.393555
```

```
MG50
                    153.548737
                                        4615.292784
                                                            -860.982242
      MG48
                    153.978686
                                        5369.017905
                                                           -1163.679269
      MG49
                    136.819901
                                        5290.379033
                                                           -1274.961181
       MG50
                      1.236915
                                        5196.967151
                                                            -685.370806
             Dist_From_Bregma_Z
      Exp
      MG48
                    2762.964212
      MG49
                    2745.502601
      MG50
                    2677.345192
      MG48
                    2839.165783
      MG49
                    3213.812551
      MG50
                    2462.813891
      MG48
                    3584.832360
      MG49
                    3648.417585
      MG50
                    3333.408306
      MG48
                    5010.493602
      MG49
                    6123.180491
      MG50
                    5292.105781
       [12 rows x 23 columns]
[105]: # surf areas and volumes
       df_lhs_vm1 = df[((df['Hem']=='lhs') & (df['BR']=='vM1'))]
       df_rhs_vm1 = df[((df['Hem']=='rhs') & (df['BR']=='vM1'))]
       df_lhs_vs1 = df[((df['Hem']=='lhs') & (df['BR']=='vS1'))]
       df_rhs_vs1 = df[((df['Hem']=='rhs') & (df['BR']=='vS1'))]
[106]: | surf_areas = [df_lhs_vm1['surf_area'].values,df_rhs_vm1['surf_area'].
        -values,df_lhs_vs1['surf_area'].values,df_rhs_vs1['surf_area'].values]
       pia_areas = [df_lhs_vm1['pia_surf_area'].values,df_rhs_vm1['pia_surf_area'].
        -values, df_lhs_vs1['pia_surf_area'].values, df_rhs_vs1['pia_surf_area'].values]
       wm_areas = [df_lhs_vm1['wm_surf_area'].values,df_rhs_vm1['wm_surf_area'].
        →values,df_lhs_vs1['wm_surf_area'].values,df_rhs_vs1['wm_surf_area'].values]
       fig, axes = plt.subplots(nrows=1,ncols=2,__
       →sharex=True, sharey=True, figsize=(12*cm, 7*cm), dpi=120)
       b1 = axes[0].boxplot(surf_areas)
       eb1 = axes[0].errorbar(x=[1,2,3,4],y=np.array(surf_areas).mean(axis=1),yerr=np.
       →array(surf_areas).std(axis=1),\
                    color='blue',marker='.',fmt='.')
       b2 = axes[1].boxplot(pia_areas)
       eb2 = axes[1].errorbar(x=[1,2,3,4],y=np.array(pia_areas).mean(axis=1),yerr=np.
        →array(pia_areas).std(axis=1),\
                    color='blue',marker='.',fmt='.')
```

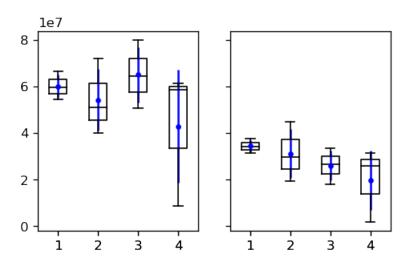
4607.222303

-290.536024

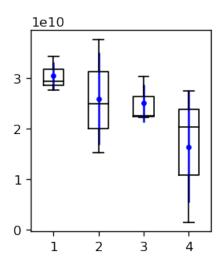
MG49

138.652532

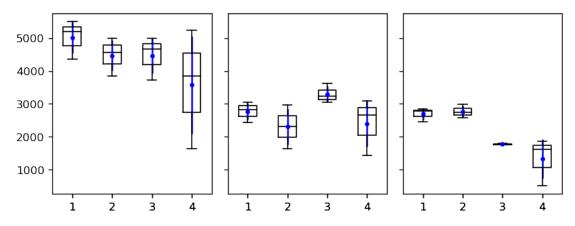
```
\#b3 = axes[2].boxplot(wm_areas)
\#eb3 = axes[2].errorbar(x=[1,2,3,4],y=np.array(wm areas).mean(axis=1),yerr=np.
\rightarrow array(wm\_areas).std(axis=1),
              color='blue',marker='.',fmt='.')
#b4 = axes[3].boxplot(surf_vols)
plt.setp(b1['boxes'], color='black')
plt.setp(b1['whiskers'], color='black')
plt.setp(b1['medians'], color='black')
plt.setp(b1['caps'], color='black')
plt.setp(b2['boxes'], color='black')
plt.setp(b2['whiskers'], color='black')
plt.setp(b2['medians'], color='black')
plt.setp(b2['caps'], color='black')
#plt.setp(b3['boxes'], color='black')
#plt.setp(b3['whiskers'], color='black')
#plt.setp(b3['medians'], color='black')
#plt.setp(b3['caps'], color='black')
fig.savefig(input_path + 'Fig_1/Rabies_Distribution/variability/surf_areas.eps')
```



```
[107]: (array([60134974.32372921, 54291731.92702731, 65025796.55837191,
              42852192.23701403]),
       array([ 4914419.41713892, 13227734.75623022, 11912375.74921831,
               24247092.5934162 ]),
        array([0.08172315, 0.24364179, 0.18319461, 0.56583086]))
[108]: np.array(pia_areas).mean(axis=1),np.array(pia_areas).std(axis=1),np.
        →array(pia_areas).std(axis=1)/np.array(pia_areas).mean(axis=1)
[108]: (array([34382857.70773227, 31192606.80273371, 26047941.52964634,
               19761069.92644515]),
        array([ 2471445.42786537, 10439070.18533514, 6343256.15517651,
               12849635.24565877]),
        array([0.07188016, 0.33466489, 0.24352236, 0.65024998]))
[109]: np.array(wm areas).mean(axis=1),np.array(wm areas).std(axis=1),np.
       →array(wm areas).std(axis=1)/np.array(wm areas).mean(axis=1)
[109]: (array([ 8204036.0609121 , 6922380.77612065, 19764739.79579503,
               10222870.92297374]),
        array([ 1593203.8704809 , 3088582.06154907, 11418876.31084508,
                7083843.11825806]),
        array([0.19419757, 0.44617338, 0.57773977, 0.69294068]))
[110]: surf vols = [df lhs vm1['surf vol'].values,df rhs vm1['surf vol'].
       →values,df_lhs_vs1['surf_vol'].values,df_rhs_vs1['surf_vol'].values]
       fig, axes = plt.subplots(nrows=1,ncols=1,_u
       ⇒sharex=True, sharey=True, figsize=(6*cm,7*cm), dpi=120)
       b1 = axes.boxplot(surf vols)
       #print(b1)
       eb = axes.errorbar(x=[1,2,3,4],y=np.array(surf_vols).mean(axis=1),yerr=np.
       →array(surf_vols).std(axis=1),\
                    color='blue',marker='.',fmt='.')
       plt.setp(b1['boxes'], color='black')
       plt.setp(b1['whiskers'], color='black')
       plt.setp(b1['medians'], color='black')
       plt.setp(b1['caps'], color='black')
       plt.tight layout()
       fig.savefig(input_path + 'Fig_1/Rabies_Distribution/variability/surf_vols.eps')
```



```
[111]: np.array(surf_vols).mean(axis=1),np.array(surf_vols).std(axis=1),np.
        →array(surf_vols).std(axis=1)/np.array(surf_vols).mean(axis=1)
[111]: (array([3.04276071e+10, 2.59552539e+10, 2.50416434e+10, 1.64025540e+10]),
       array([2.77759451e+09, 9.19299494e+09, 3.75145557e+09, 1.10018323e+10]),
        array([0.09128534, 0.35418628, 0.14980868, 0.67073898]))
[112]: pca0 extent = [df_lhs_vm1['pca0_extent'].values,df_rhs_vm1['pca0_extent'].
       →values,df_lhs_vs1['pca0_extent'].values,df_rhs_vs1['pca0_extent'].values]
       pca1 extent = [df lhs vm1['pca1 extent'].values,df rhs vm1['pca1 extent'].
       -values,df_lhs_vs1['pca1_extent'].values,df_rhs_vs1['pca1_extent'].values]
       pca2 extent = [df lhs vm1['pca2 extent'].values.df rhs vm1['pca2 extent'].
       -values,df_lhs_vs1['pca2 extent'].values,df_rhs_vs1['pca2 extent'].values]
       fig, axes = plt.subplots(nrows=1,ncols=3,__
       ⇒sharex=True, sharey=True, figsize=(18*cm, 7*cm), dpi=120)
       b1 = axes[0].boxplot(pca0_extent)
       eb1 = axes[0].errorbar(x=[1,2,3,4],y=np.array(pca0_extent).mean(axis=1),yerr=np.
       →array(pca0_extent).std(axis=1),\
                    color='blue',marker='.',fmt='.')
       b2 = axes[1].boxplot(pca1_extent)
       eb2 = axes[1].errorbar(x=[1,2,3,4],y=np.array(pca1_extent).mean(axis=1),yerr=np.
       →array(pca1_extent).std(axis=1),\
                    color='blue',marker='.',fmt='.')
       b3 = axes[2].boxplot(pca2_extent)
       eb3 = axes[2].errorbar(x=[1,2,3,4],y=np.array(pca2_extent).mean(axis=1),yerr=np.
       →array(pca2_extent).std(axis=1),\
                    color='blue',marker='.',fmt='.')
```

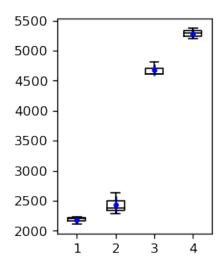


```
[115]: (array([2692.52605639, 2772.80077258, 1771.99340483, 1329.11796089]),
        array([172.5738613 , 169.59636646, 26.36089975, 588.59135003]),
        array([0.06409366, 0.06116428, 0.01487641, 0.44284358]))
[116]: pca0_angle_X_axis = [df_lhs_vm1['pca0_angle_X_axis'].
        →values,df_rhs_vm1['pca0_angle_X_axis'].
        →values,df_lhs_vs1['pca0_angle_X_axis'].
        ⇔values,df_rhs_vs1['pca0_angle_X_axis'].values]
       #pcaO angle Y axis = [df lhs vm1['pcaO angle Y axis'].
        ⇒values, df rhs vm1['pca0 angle Y axis'].
        \rightarrow values, df_lhs_vs1['pca0_angle_Y_axis'].
        \rightarrow values, df_rhs_vs1['pca0_angle_Y_axis'].values]
       pca0 angle Z axis = [df lhs vm1['pca0 angle Z axis'].
        →values,df_rhs_vm1['pca0_angle_Z_axis'].
        →values,df_lhs_vs1['pca0_angle_Z_axis'].
        →values,df_rhs_vs1['pca0_angle_Z_axis'].values]
       pca1 angle X axis = [df lhs vm1['pca1 angle X axis'].
        ⇒values, df rhs vm1['pca1 angle X axis'].
        →values,df_lhs_vs1['pca1_angle_X_axis'].
        →values,df_rhs_vs1['pca0_angle_X_axis'].values]
       \#pca0 angle Y axis = [df lhs vm1['pca0 angle Y axis'].
        ⇒values, df rhs vm1['pca0 angle Y axis'].
        \rightarrow values, df_lhs_vs1['pca0_angle_Y_axis'].
       \rightarrow values, df_rhs_vs1['pca0_angle_Y_axis'].values]
       pca1 angle Z axis = [df lhs vm1['pca1 angle Z axis'].
        →values,df_rhs_vm1['pca1_angle_Z_axis'].
        →values,df_lhs_vs1['pca1_angle_Z_axis'].
        →values,df_rhs_vs1['pca0_angle_Z_axis'].values]
       pca2 angle X axis = [df lhs vm1['pca2 angle X axis'].
        ⇒values, df rhs vm1['pca2 angle X axis'].
        →values,df lhs vs1['pca2 angle X axis'].
        →values,df_rhs_vs1['pca0_angle_X_axis'].values]
       \#pcaO\_angle\_Y\_axis = [df\_lhs\_vm1['pcaO\_angle\_Y\_axis'].
        \rightarrow values, df_rhs_vm1['pca0_angle_Y_axis'].
        →values, df lhs vs1['pca0 angle Y axis'].
        \rightarrow values, df_rhs_vs1['pca0_angle_Y_axis'].values]
       pca2 angle Z axis = [df lhs vm1['pca2 angle Z axis'].
        →values,df_rhs_vm1['pca2_angle_Z_axis'].
        ⇒values, df lhs vs1['pca2 angle Z axis'].
        →values,df_rhs_vs1['pca0_angle_Z_axis'].values]
       fig, axes = plt.subplots(nrows=1,ncols=6,__
       ⇒sharex=True, sharey=True, figsize=(18*cm, 7*cm), dpi=120)
       b1 = axes[0].boxplot(pca0 angle X axis)
       eb1 =axes[0].errorbar(x=[1,2,3,4],y=np.array(pca0_angle_X_axis).
        →mean(axis=1),yerr=np.array(pca0_angle_X_axis).std(axis=1),\
```

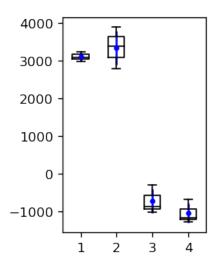
```
color='blue',marker='.',fmt='.')
b2 = axes[1].boxplot(pca0 angle Z axis)
eb2 =axes[1].errorbar(x=[1,2,3,4],y=np.array(pca0_angle_Z_axis).
→mean(axis=1),yerr=np.array(pca0_angle_Z_axis).std(axis=1),\
             color='blue',marker='.',fmt='.')
b3 = axes[2].boxplot(pca1 angle X axis)
eb3 = axes[2].errorbar(x=[1,2,3,4],y=np.array(pca1 angle X axis).
→mean(axis=1),yerr=np.array(pca1_angle_X_axis).std(axis=1),\
             color='blue',marker='.',fmt='.')
b4 = axes[3].boxplot(pca1_angle_Z_axis)
eb4 = axes[3].errorbar(x=[1,2,3,4],y=np.array(pca1_angle_Z_axis).
→mean(axis=1), yerr=np.array(pca1 angle Z axis).std(axis=1),\
             color='blue',marker='.',fmt='.')
b5 = axes[4].boxplot(pca2_angle_X_axis)
eb5 = axes[4].errorbar(x=[1,2,3,4],y=np.array(pca2_angle_X_axis).
→mean(axis=1),yerr=np.array(pca2_angle_X_axis).std(axis=1),\
             color='blue',marker='.',fmt='.')
b6 = axes[5].boxplot(pca2_angle_Z_axis)
eb6 = axes[5].errorbar(x=[1,2,3,4],y=np.array(pca2_angle_Z_axis).
→mean(axis=1),yerr=np.array(pca2_angle_Z_axis).std(axis=1),\
             color='blue',marker='.',fmt='.')
plt.setp(b1['boxes'], color='black')
plt.setp(b1['whiskers'], color='black')
plt.setp(b1['medians'], color='black')
plt.setp(b1['caps'], color='black')
plt.setp(b2['boxes'], color='black')
plt.setp(b2['whiskers'], color='black')
plt.setp(b2['medians'], color='black')
plt.setp(b2['caps'], color='black')
plt.setp(b3['boxes'], color='black')
plt.setp(b3['whiskers'], color='black')
plt.setp(b3['medians'], color='black')
plt.setp(b3['caps'], color='black')
plt.setp(b4['boxes'], color='black')
plt.setp(b4['whiskers'], color='black')
plt.setp(b4['medians'], color='black')
plt.setp(b4['caps'], color='black')
plt.setp(b5['boxes'], color='black')
plt.setp(b5['whiskers'], color='black')
plt.setp(b5['medians'], color='black')
plt.setp(b5['caps'], color='black')
plt.setp(b6['boxes'], color='black')
plt.setp(b6['whiskers'], color='black')
plt.setp(b6['medians'], color='black')
plt.setp(b6['caps'], color='black')
plt.tight_layout()
```

```
[117]: np.array(pca0_angle_X_axis).mean(axis=1),np.array(pca0_angle_X_axis).
       ⇒std(axis=1),np.array(pca0_angle_X_axis).std(axis=1)/np.
        →array(pca0 angle X axis).mean(axis=1)
[117]: (array([102.29198451, 40.35601563, 108.01072233, 111.84864211]),
       array([7.84102708, 28.60588025, 8.03867268, 68.73721322]),
        array([0.07665339, 0.70883807, 0.07442477, 0.61455563]))
[118]: np.array(pca0_angle_Z_axis).mean(axis=1),np.array(pca0_angle_Z_axis).
        ⇒std(axis=1),np.array(pca0_angle_Z_axis).std(axis=1)/np.
        →array(pca0_angle_Z_axis).mean(axis=1)
[118]: (array([126.66095436, 107.53394327, 124.41864802, 102.21239963]),
       array([ 2.74973905, 9.53714962, 5.18561173, 19.83599734]),
        array([0.02170945, 0.08868967, 0.04167873, 0.19406645]))
[119]: np.array(pca1_angle_X_axis).mean(axis=1),np.array(pca1_angle_X_axis).
        \rightarrowstd(axis=1),np.array(pca1_angle_X_axis).std(axis=1)/np.
        →array(pca1_angle_X_axis).mean(axis=1)
[119]: (array([ 8.41791036, 170.71973246, 56.12567879, 111.84864211]),
       array([ 4.41971974, 4.03194657, 3.09475575, 68.73721322]),
        array([0.52503763, 0.02361734, 0.05513975, 0.61455563]))
[120]: np.array(pca1_angle_Z_axis).mean(axis=1),np.array(pca1_angle_Z_axis).
       ⇒std(axis=1),np.array(pca1_angle_Z_axis).std(axis=1)/np.
        →array(pca1_angle_Z_axis).mean(axis=1)
```

```
[120]: (array([115.18703146, 20.21364853, 28.04558306, 102.21239963]),
        array([77.13940821, 8.94265571, 7.63535702, 19.83599734]),
        array([0.66968831, 0.44240681, 0.27224811, 0.19406645]))
[121]: np.array(pca2_angle_X_axis).mean(axis=1),np.array(pca2_angle_X_axis).
       ⇒std(axis=1),np.array(pca2_angle_X_axis).std(axis=1)/np.
        →array(pca2_angle_X_axis).mean(axis=1)
[121]: (array([ 95.56113108, 87.58744438, 162.49501569, 111.84864211]),
       array([5.72231793, 2.13657571, 3.01910558, 68.73721322]),
        array([0.05988123, 0.02439363, 0.01857968, 0.61455563]))
[122]: np.array(pca2 angle_Z_axis).mean(axis=1),np.array(pca2 angle_Z_axis).
       ⇒std(axis=1),np.array(pca2_angle_Z_axis).std(axis=1)/np.
        →array(pca2 angle Z axis).mean(axis=1)
[122]: (array([ 37.26518465, 17.49172831, 150.78654658, 102.21239963]),
       array([ 3.10357 , 11.29720117, 8.99435146, 19.83599734]),
        array([0.08328337, 0.64585963, 0.05964956, 0.19406645]))
[130]: bregma_dist_X = [df_lhs_vm1['Dist_From_Bregma_X'].
       →values,df rhs vm1['Dist From Bregma X'].
       ⇒values, df lhs vs1['Dist From Bregma X'].
       →values,df_rhs_vs1['Dist_From_Bregma_X'].values]
      fig, axes = plt.subplots(nrows=1,ncols=1,__
       ⇒sharex=True, sharey=True, figsize=(6*cm, 7*cm), dpi=120)
      b1 = axes.boxplot(bregma dist X)
      eb1 =axes.errorbar(x=[1,2,3,4],y=np.array(bregma_dist_X).mean(axis=1),yerr=np.
       →array(bregma dist X).std(axis=1),\
                    color='blue',marker='.',fmt='.')
      plt.setp(b1['boxes'], color='black')
      plt.setp(b1['whiskers'], color='black')
      plt.setp(b1['medians'], color='black')
      plt.setp(b1['caps'], color='black')
      plt.tight layout()
      fig.savefig(input_path + 'Fig_1/Rabies_Distribution/variability/bregma_dist_X.
       →eps')
```

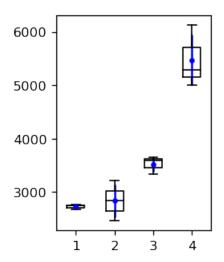


```
[131]: np.array(bregma_dist_X).mean(axis=1),np.array(bregma_dist_X).std(axis=1),np.
        →array(bregma_dist_X).std(axis=1)/np.array(bregma_dist_X).mean(axis=1)
[131]: (array([2183.58813574, 2431.96066365, 4677.67488569, 5285.45469656]),
        array([ 53.49385142, 143.66251302, 93.98607431, 70.32568205]),
        array([0.02449814, 0.05907271, 0.02009248, 0.01330551]))
[132]: bregma_dist_Y = [df_lhs_vm1['Dist_From_Bregma_Y'].
        →values,df_rhs_vm1['Dist_From_Bregma_Y'].
        →values,df_lhs_vs1['Dist_From_Bregma_Y'].
        →values,df_rhs_vs1['Dist_From_Bregma_Y'].values]
       fig, axes = plt.subplots(nrows=1,ncols=1,__
       ⇒sharex=True, sharey=True, figsize=(6*cm, 7*cm), dpi=120)
       b1 = axes.boxplot(bregma dist Y)
       eb1 =axes.errorbar(x=[1,2,3,4],y=np.array(bregma_dist_Y).mean(axis=1),yerr=np.
       →array(bregma_dist_Y).std(axis=1),\
                    color='blue',marker='.',fmt='.')
       plt.setp(b1['boxes'], color='black')
       plt.setp(b1['whiskers'], color='black')
       plt.setp(b1['medians'], color='black')
       plt.setp(b1['caps'], color='black')
       plt.tight layout()
       fig.savefig(input_path + 'Fig_1/Rabies_Distribution/variability/bregma_dist_Y.
        →eps')
```



```
[133]: np.array(bregma_dist_Y).mean(axis=1),np.array(bregma_dist_Y).std(axis=1),np.
        →array(bregma_dist_Y).std(axis=1)/np.array(bregma_dist_Y).mean(axis=1)
[133]: (array([ 3107.42204459, 3350.52466476, -723.63727363, -1041.33708555]),
       array([104.27327944, 447.38297608, 313.00272124, 255.77321977]),
        array([ 0.0335562 , 0.13352624, -0.43254091, -0.24562
[134]: bregma_dist_Z = [df_lhs_vm1['Dist_From_Bregma_Z'].

→values,df_rhs_vm1['Dist_From_Bregma_Z'].
        ⇔values,df lhs vs1['Dist From Bregma Z'].
       →values,df_rhs_vs1['Dist_From_Bregma_Z'].values]
       fig, axes = plt.subplots(nrows=1,ncols=1,__
       ⇒sharex=True, sharey=True, figsize=(6*cm,7*cm), dpi=120)
       b1 = axes.boxplot(bregma_dist_Z)
       eb1 =axes.errorbar(x=[1,2,3,4],y=np.array(bregma_dist_Z).mean(axis=1),yerr=np.
       →array(bregma_dist_Z).std(axis=1),\
                    color='blue',marker='.',fmt='.')
       plt.setp(b1['boxes'], color='black')
       plt.setp(b1['whiskers'], color='black')
       plt.setp(b1['medians'], color='black')
       plt.setp(b1['caps'], color='black')
       plt.tight_layout()
       fig.savefig(input_path + 'Fig_1/Rabies_Distribution/variability/bregma_dist_Z.
        →eps')
```



4 Layer Borders

4.1 Comparison of layer borders with others

4.1.1 vS1

```
[18]: df_mine = pd.read_csv(input_path+'Final_Data/Layer_Borders_V2/vS1_Borders.
       ⇔csv',index_col=0)
      df mine.mean()
[18]: L1_Border
                          -75.000000
     L2_3_Border
                         -470.833333
     L4 Border
                         -879.166667
     L5_Border
                        -1520.833333
     L1_Border_Per
                            0.034884
     L2_3_Border_Per
                            0.218992
     L4 Border Per
                            0.408915
     L5 Border Per
                            0.707364
     dtype: float64
[34]: N=5
      nd = np.arange(N)
      width = 0.35
      fig = plt.figure(figsize=(6*cm,7*cm),dpi=120)
      ax = fig.add_subplot(111)
      rects1 = ax.bar(ind, [df_mine.mean()['L1_Border_Per']*100,df_mine.

--mean()['L2_3_Border_Per']*100,df_mine.mean()['L4_Border_Per']*100,\

                           df_mine.mean()['L5_Border_Per']*100,100], width,

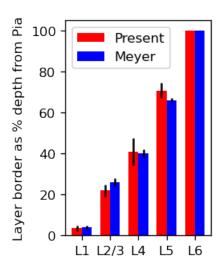
color='red', \

                      yerr=[df mine.std()['L1 Border Per']*100,df mine.
       →std()['L2_3_Border_Per']*100,df_mine.std()['L4_Border_Per']*100,\
                           df mine.std()['L5 Border Per']*100,0])
      rects2 = ax.bar(ind+width, df_meyer['Mean'].values, width, color='blue',_

    yerr=df_meyer['STD'].values)
      # add some
      ax.set ylabel('Layer border as % depth from Pia')
      #ax.set_title('BF Layer border comparison')
      ax.set xticks(ind + width / 2)
      ax.set_xticklabels(('L1', 'L2/3', 'L4', 'L5', 'L6'))
      ax.legend( (rects1[0], rects2[0]), ('Present', 'Meyer') )
      plt.tight_layout()
      fig.savefig(input_path + 'Fig_2/vS1_Border_Comparison.eps')
```

The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.

The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.



4.1.2 vM1

```
[35]: # vS1
      df = pd.DataFrame()
      df['Layers'] = ['L1','L2/3','LA','LB','L6']
      df['Mean'] = [10,16,34,67,100]
      df['STD'] = [1,1,2,0,0]
      df.to_csv(input_path+'Fig_2/vM1_Layer_Borders_Shepherd.csv')
[39]: df_hooks = pd.read_csv(input_path+'Fig_2/vM1_Layer_Borders_Shepherd.
       ⇔csv',index_col=0)
[40]: df_hooks
[40]:
        Layers
                Mean
                      STD
      0
            L1
                  10
                        1
      1
          L2/3
                  16
                        1
      2
            LA
                  34
                        2
                        0
      3
            LB
                  67
      4
            L6
                        0
                 100
[41]: df_mine = pd.read_csv(input_path+'Final_Data/Layer_Borders/vM1_Borders.
      ⇔csv',index_col=0)
      df_mine.mean()
[41]: L1_Border
                         -222.093023
      L2_3_Border
                         -544.186047
     L5A_Border
                        -1017.571059
```

```
L5B_Border
     L1_Border_Per
                           0.085420
     L2_3_Border_Per
                           0.209302
     L5A_Border_Per
                           0.391373
     L5B_Border_Per
                           0.779219
     dtype: float64
[42]: N=5
     nd = np.arange(N)
     width = 0.35
     fig = plt.figure(figsize=(6*cm,7*cm),dpi=120)
     ax = fig.add_subplot(111)
     rects1 = ax.bar(ind, [df_mine.mean()['L1_Border_Per']*100,df_mine.

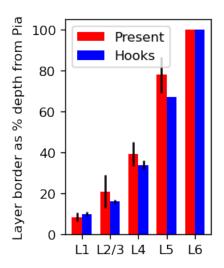
-mean()['L2_3_Border_Per']*100,df_mine.mean()['L5A_Border_Per']*100,\

                          df_mine.mean()['L5B_Border_Per']*100,100], width,
      yerr=[df_mine.std()['L1_Border_Per']*100,df_mine.
      →std()['L2_3_Border_Per']*100,df_mine.std()['L5A_Border_Per']*100,\
                          df mine.std()['L5B Border Per']*100,0])
     rects2 = ax.bar(ind+width, df_hooks['Mean'].values, width, color='blue',u
      →yerr=df_hooks['STD'].values)
      # add some
     ax.set_ylabel('Layer border as % depth from Pia')
     #ax.set_title('BF Layer border comparison')
     ax.set xticks(ind + width / 2)
     ax.set_xticklabels(('L1', 'L2/3', 'L4', 'L5', 'L6'))
     ax.legend( (rects1[0], rects2[0]), ('Present', 'Hooks') )
     plt.tight_layout()
     fig.savefig(input_path + 'Fig_2/vM1_Border_Comparison.eps')
```

-2025.968992

The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.

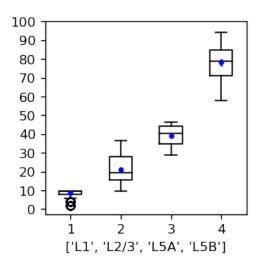
The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.



5 vM1 Laminar organization

```
[48]: # show layer border variability
       df = pd.read_csv(input_path+'Final_Data/Layer_Borders/vM1_Borders.
        ⇔csv',index_col=0)
       df.columns
[48]: Index(['Exp_Name', 'L1_Border', 'L2_3_Border', 'L5A_Border', 'L5B_Border',
              'L1_Border_Per', 'L2_3_Border_Per', 'L5A_Border_Per', 'L5B_Border_Per'],
             dtype='object')
[121]: mg48_lhs_df = df[df['Exp_Name']=='MG48_lhs']
       mg48_rhs_df = df[df['Exp_Name']=='MG48_rhs']
       mg49_lhs_df = df[df['Exp_Name'] == 'MG49_lhs']
       mg49_rhs_df = df[df['Exp_Name']=='MG49_rhs']
       mg50_lhs_df = df[df['Exp_Name'] == 'MG50_lhs']
       mg50_rhs_df = df[df['Exp_Name']=='MG50_rhs']
[122]: 11_means = [mg48_lhs_df['L1_Border_Per'].mean(),mg48_rhs_df['L1_Border_Per'].
        →mean(),mg49_lhs_df['L1_Border_Per'].mean(),\
                   mg49_rhs_df['L1_Border_Per'].mean(),mg50_lhs_df['L1_Border_Per'].
        →mean(),mg50_rhs_df['L1_Border_Per'].mean()]
       12_3_means = [mg48_lhs_df['L2_3_Border_Per'].
        →mean(),mg48_rhs_df['L2_3_Border_Per'].mean(),mg49_lhs_df['L2_3_Border_Per'].
        \rightarrowmean(),\
```

```
mg49_rhs_df['L2_3_Border_Per'].
        →mean(),mg50_lhs_df['L2_3_Border_Per'].mean(),mg50_rhs_df['L2_3_Border_Per'].
        \rightarrowmean()]
       15A_means = [mg48_lhs_df['L5A_Border_Per'].mean(),mg48_rhs_df['L5A_Border_Per'].
        →mean(),mg49 lhs df['L5A Border Per'].mean(),\
                   mg49_rhs_df['L5A_Border_Per'].mean(),mg50_lhs_df['L5A_Border_Per'].
        →mean(),mg50_rhs_df['L5A_Border_Per'].mean()]
       15B_means = [mg48_lhs_df['L5B_Border_Per'].mean(),mg48_rhs_df['L5B_Border_Per'].
        →mean(),mg49_lhs_df['L5B_Border_Per'].mean(),\
                   mg49 rhs df['L5B Border Per'].mean(),mg50 lhs df['L5B Border Per'].
        →mean(),mg50_rhs_df['L5B_Border_Per'].mean()]
[133]: per_layer_borders_df = df[['L1_Border_Per', 'L2_3_Border_Per', u
       → 'L5A Border Per', 'L5B Border Per']]*100
       fig, axes = plt.subplots(nrows=1,ncols=1,__
        ⇒sharex=True, sharey=True, figsize=(7*cm,7*cm), dpi=120)
       b1 = axes.boxplot(per_layer_borders df)
       eb1 = axes.errorbar(x=[1,2,3,4], y=[np.array(11_means).mean()*100,np.
        \rightarrowarray(12_3_means).mean()*100,np.array(15A_means).mean()*100,np.
        \rightarrowarray(15B_means).mean()*100],\
                          yerr=[np.array(11\_means).std()*100,np.array(12\_3\_means).
        \rightarrowstd()*100,np.array(15A_means).std()*100,np.array(15B_means).std()*100],
                    color='blue',marker='.',fmt='.')
       plt.yticks(np.arange(0,110,10))
       plt.xlabel(['L1','L2/3','L5A','L5B'])
       plt.setp(b1['boxes'], color='black')
       plt.setp(b1['whiskers'], color='black')
       plt.setp(b1['medians'], color='black')
       plt.setp(b1['caps'], color='black')
       plt.tight_layout()
       fig.savefig(input_path + 'Fig_2/laminar_variability.eps')
```



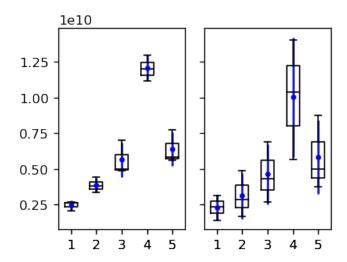
```
[126]: np.array(l1_means).mean(),np.array(l1_means).std()
```

[126]: (0.08623863020839255, 0.011960957728402168)

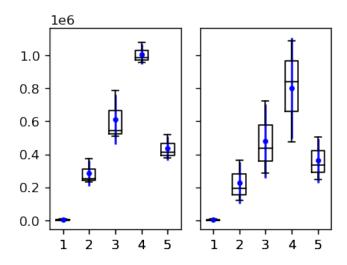
5.1 3d counts and densities

```
[47]: df_2d_counts = pd.read_csv(input_path+'Fig_2/Layerwise_2D_Cell_Counts.csv')
    df_3d_counts = pd.read_csv(input_path+'Fig_2/Layerwise 3D_Cell_Counts.csv')
    df_3d_densities = pd.read_csv(input_path+'Fig_2/Layerwise_3D_Densities.csv')
    df layerwise volumes = pd.read csv(input path+'Fig 2/Layer Volumes.csv')
    df_3d_counts_lhs = df_3d_counts[df_3d_counts['Hem']=='lhs']
    df_3d_counts_rhs = df_3d_counts[df_3d_counts['Hem']=='rhs']
    df_3d_densities_lhs = df_3d_densities[df_3d_densities['Hem']=='lhs']
    df_3d_densities_rhs = df_3d_densities[df_3d_densities['Hem']=='rhs']
    df_layerwise_volumes_lhs =_
     →df_layerwise_volumes[(df_layerwise_volumes['Exp']=='MG48_lhs') |
     df_layerwise_volumes_rhs =__
     ⇒df_layerwise_volumes[(df_layerwise_volumes['Exp']=='MG48_rhs') |
```

```
eb1 = axes[0].
 \rightarrowerrorbar(x=[1,2,3,4,5],y=df_layerwise_volumes_lhs[['L1','L2_3','L5A','L5B','L6']].
 \rightarrowmean(),\
 color='blue',marker='.',fmt='.')
b2 = axes[1].boxplot(df_layerwise_volumes_rhs[['L1','L2_3','L5A','L5B','L6']])
eb2 = axes[1].
\rightarrowerrorbar(x=[1,2,3,4,5],y=df_layerwise_volumes_rhs[['L1','L2_3','L5A','L5B','L6']].
\rightarrowmean(),\
color='blue',marker='.',fmt='.')
#plt.yticks(np.arange(0,110,10))
#plt.xlabel(['L1','L2/3','L5A','L5B'])
plt.setp(b1['boxes'], color='black')
plt.setp(b1['whiskers'], color='black')
plt.setp(b1['medians'], color='black')
plt.setp(b1['caps'], color='black')
plt.tight_layout()
plt.setp(b2['boxes'], color='black')
plt.setp(b2['whiskers'], color='black')
plt.setp(b2['medians'], color='black')
plt.setp(b2['caps'], color='black')
plt.tight_layout()
fig.savefig(input_path + 'Fig_2/layerwise_volumes.eps')
```



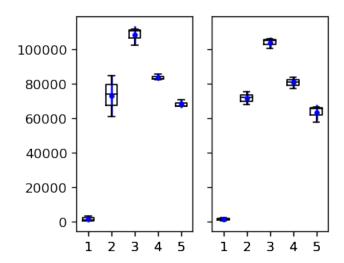
```
[49]: fig, axes = plt.subplots(nrows=1,ncols=2,__
       →sharex=True,sharey=True,figsize=(9*cm,7*cm),dpi=120)
      b1 = axes[0].boxplot(df_3d_counts_lhs[['L1','L2_3','L5A','L5B','L6']])
      eb1 = axes[0].
       \rightarrowerrorbar(x=[1,2,3,4,5],y=df_3d_counts_lhs[['L1','L2_3','L5A','L5B','L6']].
       \rightarrowmean(),\
                          yerr=df_3d_counts_lhs[['L1','L2_3','L5A','L5B','L6']].std(),\
                   color='blue',marker='.',fmt='.')
      b2 = axes[1].boxplot(df_3d_counts_rhs[['L1','L2_3','L5A','L5B','L6']])
      eb2 = axes[1].
       →errorbar(x=[1,2,3,4,5],y=df_3d_counts_rhs[['L1','L2_3','L5A','L5B','L6']].
       \rightarrowmean(),\
                          yerr=df_3d_counts_rhs[['L1','L2_3','L5A','L5B','L6']].std(),\
                   color='blue',marker='.',fmt='.')
      #plt.yticks(np.arange(0,110,10))
      #plt.xlabel(['L1','L2/3','L5A','L5B'])
      plt.setp(b1['boxes'], color='black')
      plt.setp(b1['whiskers'], color='black')
      plt.setp(b1['medians'], color='black')
      plt.setp(b1['caps'], color='black')
      plt.tight_layout()
      plt.setp(b2['boxes'], color='black')
      plt.setp(b2['whiskers'], color='black')
      plt.setp(b2['medians'], color='black')
      plt.setp(b2['caps'], color='black')
      plt.tight_layout()
      fig.savefig(input_path + 'Fig_2/layerwise_3d_cellcounts.eps')
```



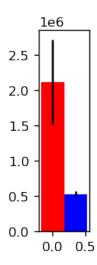
```
[50]: fig, axes = plt.subplots(nrows=1,ncols=2,__
       →sharex=True,sharey=True,figsize=(9*cm,7*cm),dpi=120)
      b1 = axes[0].boxplot(df_3d_densities_lhs[['L1','L2_3','L5A','L5B','L6']])
      eb1 = axes[0].
       \rightarrowerrorbar(x=[1,2,3,4,5],y=df_3d_densities_lhs[['L1','L2_3','L5A','L5B','L6']].
       \rightarrowmean(),\
                          yerr=df_3d_densities_lhs[['L1','L2_3','L5A','L5B','L6']].
       →std(),\
                    color='blue',marker='.',fmt='.')
      b2 = axes[1].boxplot(df_3d_densities_rhs[['L1','L2_3','L5A','L5B','L6']])
      eb2 = axes[1].
       \rightarrowerrorbar(x=[1,2,3,4,5],y=df_3d_densities_rhs[['L1','L2_3','L5A','L5B','L6']].
       \rightarrowmean(),\
                          yerr=df_3d_densities_rhs[['L1','L2_3','L5A','L5B','L6']].

std(),\
                    color='blue',marker='.',fmt='.')
      #plt.yticks(np.arange(0,110,10))
      #plt.xlabel(['L1', 'L2/3', 'L5A', 'L5B'])
      plt.setp(b1['boxes'], color='black')
      plt.setp(b1['whiskers'], color='black')
      plt.setp(b1['medians'], color='black')
      plt.setp(b1['caps'], color='black')
      plt.tight_layout()
      plt.setp(b2['boxes'], color='black')
      plt.setp(b2['whiskers'], color='black')
```

```
plt.setp(b2['medians'], color='black')
plt.setp(b2['caps'], color='black')
plt.tight_layout()
fig.savefig(input_path + 'Fig_2/layerwise_3d_densities.eps')
```



5.2 Comapare counts with vS1 counts



```
[61]: volumes = df_volumes[df_volumes['BR']=='vM1']['surf_vol'].values

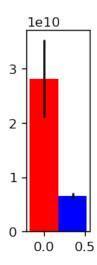
N=1
  ind = np.arange(N)
  width = 0.35

fig = plt.figure(figsize=(3*cm,7*cm),dpi=120)
  ax = fig.add_subplot(111)

rects1 = ax.bar(ind, volumes.mean(), width, color='red', yerr=volumes.std())

rects2 = ax.bar(ind+width, 6.6*1e9, width, color='blue', yerr=0.58*1e9)

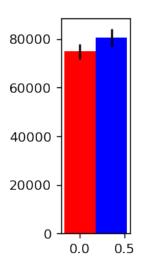
# add some
  plt.tight_layout()
  fig.savefig(input_path + 'Fig_2/Volume_Comparison.eps')
```



```
[64]: densities = []
     df_volumes = pd.read_csv(input_path+'Fig_1/Rabies_Distribution/Variability/
      ⇔variability.csv')
     for exp in ['MG48','MG49','MG50']:
         for hem in ['lhs','rhs']:
             vol = df_volumes[(df_volumes['Exp'] == exp) & (df_volumes['Hem'] == hem) &_

    df_volumes['BR'] == 'vM1')]['surf_vol'].values[0]

             counts = df_3d_counts[(df_3d_counts['Exp']==exp) &__
      densities.append(counts*1e9/vol)
     N=1
     ind = np.arange(N)
     width = 0.35
     fig = plt.figure(figsize=(4*cm,7*cm),dpi=120)
     ax = fig.add_subplot(111)
     rects1 = ax.bar(ind, np.array(densities).mean(), width, color='red', yerr=np.
      →array(densities).std())
     rects2 = ax.bar(ind+width, 80419, width, color='blue', yerr=3688)
     # add some
     plt.tight_layout()
     fig.savefig(input_path + 'Fig_2/Densities_Comparison.eps')
```



6 Global v/s Local Ref Frame Comparison

```
[523]: df_global =pd.read_csv('Z:/V11/vM1_Ref_Frame/Outputs/Alignment_Analysis/
        →Global Alignment stats.csv')
       df_vm1 =pd.read_csv('Z:/V11/vM1_Ref_Frame/Outputs/Alignment_Analysis/vM1_stats.
        ⇔csv')
       df_s1bf =pd.read_csv('Z:/V11/vM1_Ref_Frame/Outputs/Alignment_Analysis/

¬S1BF_stats.csv')
[524]: df_global.columns
[524]: Index(['Unnamed: 0', 'Angular_Error_SEM', 'Angular_Precision_Mean',
              'Angular_Precision_Std', 'BR', 'Barrel_Cortex_Center_Ellipsoid',
              'Barrel_Cortex_Center_Ellipsoid_Norm',
              'Barrel_Cortex_Center_Error_Dists_SEM',
              'Barrel_Cortex_Center_Precision_Mean',
              'Barrel_Cortex_Center_Precision_Std', 'Barrels_Angular_Error_SEM',
              'Barrels_Angular_Precision_Mean', 'Barrels_Angular_Precision_Std',
              'Barrels_Barrel_Center_Precision_Mean',
              'Barrels_Barrel_Center_Precision_Std',
              'Barrels_Barrel_Centers_Error_Ellipsoid_Mean',
              'Barrels_Barrel_Centers_Error_Ellipsoid_Mean_Norm',
              'Barrels_Barrel_Centers_Error_Ellipsoid_Std',
              'Barrels_Barrel_Centers_Error_Ellipsoid_Std_Norm',
              'Barrels_Column_Center_Precision_Mean',
              'Barrels_Column_Center_Precision_Std',
              'Barrels_Column_Error_Ellipsoid_Mean',
              'Barrels_Column_Error_Ellipsoid_Mean_Norm',
              'Barrels_Column_Error_Ellipsoid_Std',
```

```
'Barrels_Column_Error_Ellipsoid_Std_Norm',
 'Barrels_Pia_Error_Ellipsoid_Mean',
 'Barrels_Pia_Error_Ellipsoid_Mean_Norm',
 'Barrels_Pia_Error_Ellipsoid_Std',
 'Barrels_Pia_Error_Ellipsoid_Std_Norm', 'Barrels_Pia_Precision_Mean',
 'Barrels_Pia_Precision_std', 'Barrels_WM_Error_Ellipsoid_Mean',
 'Barrels WM Error Ellipsoid Mean Norm',
 'Barrels_WM_Error_Ellipsoid_Std', 'Barrels_WM_Error_Ellipsoid_Std_Norm',
 'Barrels_WM_Precision_Mean', 'Barrels_WM_Precision_Std', 'Hem',
 'Inner_Alignment_Type', 'Number_Of_Axis_Used', 'Outer_Alignment_Type',
 'Overall_Rabies_landmarks', 'Overall_Volume',
 'Overlap_Rabies_landmarks', 'Overlap_Volume',
 'PCA_O_Angle_Between_Axes_Mean', 'PCA_O_Angle_Between_Axes_Std',
 'PCA_1_Angle_Between_Axes_Mean', 'PCA_1_Angle_Between_Axes_Std',
 'PCA_2_Angle_Between_Axes_Mean', 'PCA_2_Angle_Between_Axes_Std',
 'PCA_Angular_Precision_Mean', 'PCA_Angular_Precision_SEM',
 'PCA_Angular_Precision_Std', 'Pia_Error_Dist_SEM',
 'Pia_Error_Ellipsoid_Mean', 'Pia_Error_Ellipsoid_Mean_Norm',
 'Pia_Error_Ellipsoid_Std', 'Pia_Error_Ellipsoid_Std_Norm',
 'Pia_Precision_Mean', 'Pia_Precision_Std', 'Rabies_Center_Ellipsoid',
 'Rabies_Center_Ellipsoid_Norm', 'Rabies_Center_Error_Dists_SEM',
 'Rabies Center Precision Mean', 'Rabies Center Precision Std',
 'Rabies_Landmarks_Overlap_Percentage', 'Rabies_Surf_Center_Dists_Mean',
 'Rabies Surf Center Dists Std', 'Rabies Surf Center Ellipsoid',
 'Rabies_Surf_Center_Ellipsoid_Norm',
 'Rabies Surf Center Error Dists SEM', 'Surf Center Ellipsoid',
 'Surf_Center_Ellipsoid_Norm', 'Surf_Center_Error_Dists_SEM',
 'Surf_Center_Precision_Mean', 'Surf_Center_Precision_Std',
 'Volume_Overlap_Percentage', 'WM_Error_Dist_SEM',
 'WM_Error_Ellipsoid_Mean', 'WM_Error_Ellipsoid_Mean_Norm',
 'WM_Precision_Mean', 'WM_Precision_Std'],
dtype='object')
```

6.1 Center Precisions

```
→df_global[(df_global['Inner Alignment Type']=='Using_Local_Centers') &
 → (df_global['BR']=='WB')][['Surf_Center_Precision_Mean',]].mean().values[0],\
-df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_
 →& (df global['BR']=='WB')][['Surf Center Precision Mean',]].mean().
 \rightarrow values [0],
                     df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Center') &__

→ (df_vm1['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0],\
\hookrightarrow df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use_Barrels_Alone') &
→ (df_s1bf['BR']=='WB')][['Surf_Center_Precision_Mean',]].mean().values[0],\
                     width, color='red', \
 →yerr=[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') & Using_Surface')
→ (df_global['BR']=='WB')][['Surf_Center_Precision_Std',]].mean().values[0],\
\hookrightarrow df_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Centers') & \sqcup
→ (df_global['BR']=='WB')][['Surf_Center_Precision_Std',]].mean().values[0],\
→df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')__
→& (df_global['BR']=='WB')][['Surf_Center_Precision_Std',]].mean().values[0],\
                     df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Center') &_
 → (df_vm1['BR']=='WB')][['Surf_Center_Precision_Std',]].mean().values[0],\
→df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use_Barrels_Alone') & U
→ (df_s1bf['BR']=='WB')][['Surf_Center_Precision_Std',]].mean().values[0],\
                     ],\
                     )
rects2 = ax.
→bar(ind+width,[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface')
 \rightarrow values [0],\
→df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & ∪
 → (df_global['BR']=='vM1')][['Rabies_Surf_Center_Dists_Mean',]].mean().
\rightarrow values [0],
→df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_
→& (df_global['BR']=='vM1')][['Rabies_Surf_Center_Dists_Mean',]].mean().
\rightarrow values [0],\
                     df_vm1[(df_vm1['Inner_Alignment_Type'] == 'Using_Center') &_
 → (df_vm1['BR']=='vM1')][['Rabies Surf_Center_Dists_Mean',]].mean().values[0],\
```

```
→df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use_Barrels_Alone') &__
 → (df_s1bf['BR']=='vM1')][['Rabies_Surf_Center_Dists_Mean',]].mean().
 →values[0],\
                     ],\
                      width, color='green', \
 -yerr=[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') & ∪
 → (df_global['BR']=='vM1')][['Rabies_Surf_Center_Dists_Std',]].mean().
 →values[0],\
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & U
 → (df_global['BR']=='vM1')][['Rabies_Surf_Center_Dists_Std',]].mean().
\rightarrow values [0],
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_
 →& (df_global['BR']=='vM1')][['Rabies_Surf_Center_Dists_Std',]].mean().
 →values[0],\
                      df_vm1[(df_vm1['Inner_Alignment_Type'] == 'Using_Center') &__

→ (df_vm1['BR'] == 'vM1')][['Rabies_Surf_Center_Dists_Std',]].mean().values[0],\
 →df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use_Barrels_Alone') &__

→ (df_s1bf['BR']=='vM1')][['Rabies_Surf_Center_Dists_Std',]].mean().values[0],\
                      ],\
                      )
rects3 = ax.
 →bar(ind+2*width,[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface')
→& (df_global['BR']=='vM1')][['Barrels_Barrel_Center_Precision_Mean',]].
\rightarrowmean().values[0],\
-df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & ∪
→ (df_global['BR']=='vM1')][['Barrels_Barrel_Center_Precision_Mean',]].mean().
 \rightarrow values [0],

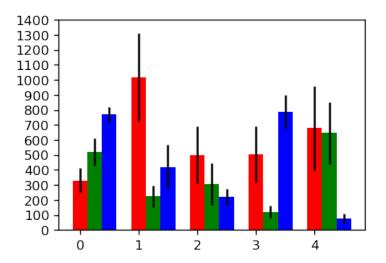
¬df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')
□
 →& (df_global['BR']=='vM1')][['Barrels_Barrel_Center_Precision_Mean',]].
 \rightarrowmean().values[0],\
                      df_vm1[(df_vm1['Inner_Alignment_Type'] == 'Using_Center') &__
 → (df_vm1['BR']=='vM1')][['Barrels_Barrel_Center_Precision_Mean',]].mean().
 \rightarrow values [0],
\hookrightarrow df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use_Barrels_Alone') &
 → (df_s1bf['BR']=='vM1')][['Barrels_Barrel_Center_Precision_Mean',]].mean().
 →values[0],\
                    ],\
```

```
width, color='blue', \
 →yerr=[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') & Using_Surface')
 \rightarrow values [0],
→df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & ∪
 → (df global['BR']=='vS1')][['Barrels Barrel Center Precision Std',]].mean().
 \rightarrow values [0],\
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_
 →& (df global['BR']=='vS1')][['Barrels Barrel Center Precision Std',]].mean().
\rightarrow values [0],
                      df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Center') &__
 → (df_vm1['BR']=='vS1')][['Barrels_Barrel_Center_Precision_Std',]].mean().
 \rightarrow values [0],\
- df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use_Barrels_Alone') كان المالية
→ (df_s1bf['BR']=='vS1')][['Barrels_Barrel_Center_Precision_Std',]].mean().
\hookrightarrow values [0],\
                      ],\
                      )
ticks = plt.yticks(np.arange(0,1500,100))
\#rects2 = ax.bar(ind+width, \bot
→ df qlobal[(df qlobal['Inner Alignment Type']=='Using Local Centers') &
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0], 
⇒width, color='blue'. \
yerr=df qlobal[(df qlobal['Inner Alignment Type']=='Using Local Centers') &
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Std',]].mean().values[0])
\#rects1 = ax.bar(ind, \dots)
→df qlobal[(df qlobal['Inner Alignment Type']=='Using Surface') &
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0], 
\rightarrow width, color='red', \
→yerr=df_qlobal[(df_qlobal['Inner Alignment Type']=='Using Surface') &
→ (df qlobal['BR']=='WB')][['Surf Center Precision Std',]].mean().values[0])
\#rects2 = ax.bar(ind+width, \square
\rightarrow df_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Centers') 
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0],
 \rightarrow width, color='blue', \
```

```
#

-yerr=df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') &
-(df_global['BR']=='WB')][['Surf_Center_Precision_Std',]].mean().values[0])

plt.tight_layout()
fig.savefig(input_path + 'Fig_3/Centers_Precision.eps')
```



6.2 Pia Precisions

```
→df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use Projections Alone') & U
 \rightarrow (df_s1bf['BR']=='WB')][['Pia_Precision_Mean',]].mean().values[0],\
                   ],\
                    width, color='red', \
 -yerr=[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') & ∪
 → (df global['BR']=='WB')][['Pia Precision Std',]].mean().values[0],\
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & ∪
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')⊔
 →& (df_global['BR']=='WB')][['Pia_Precision_Std',]].mean().values[0],\
                    df_vm1[(df_vm1['Inner_Alignment_Type'] == 'Using_Surface') &__
 → (df_vm1['BR']=='WB')][['Pia Precision Std',]].mean().values[0],\
→df s1bf[(df s1bf['Inner Alignment Type'] == 'Use Projections Alone') & ...

    → (df_s1bf['BR']=='WB')][['Pia_Precision_Std',]].mean().values[0],\

                    ],\
                    )
rects2 = ax.
→bar(ind+width, [df_global[(df_global['Inner_Alignment_Type']=='Using_Center')

    d (df_global['BR']=='vM1')][['Pia_Precision_Mean',]].values[0][0],\

 \hookrightarrow df_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Centers') &
 → (df_global['BR']=='vM1')][['Pia_Precision_Mean',]].mean().values[0],\
→df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')⊔
 →& (df_global['BR']=='vM1')][['Pia_Precision_Mean',]].mean().values[0],\
                    df_vm1[(df_vm1['Inner_Alignment_Type'] == 'Using_Surface') &__

    → (df_vm1['BR']=='vM1')][['Pia_Precision_Mean',]].values[0][0],\

→df s1bf[(df s1bf['Inner Alignment Type']=='Use Projections Alone') & ...
 → (df_s1bf['BR']=='vM1')][['Pia_Precision_Mean',]].mean().values[0],\
                    width, color='green', \
 -yerr=[df_global[(df_global['Inner_Alignment_Type']=='Using_Center') & ∪
 →df global[(df global['Inner Alignment Type']=='Using Local Centers') & ...
 → (df_global['BR']=='vM1')][['Pia_Precision_Std',]].mean().values[0],\
```

```
→df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_
  →& (df_global['BR']=='vM1')][['Pia_Precision_Std',]].mean().values[0],\
                                         df vm1[(df vm1['Inner Alignment Type']=='Using Surface') & ...

df_vm1['BR']=='vM1')][['Pia_Precision_Std',]].values[0][0],\

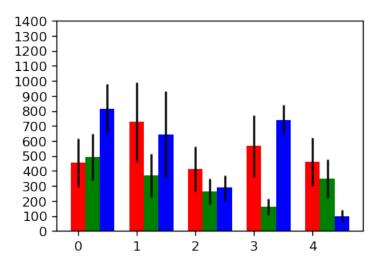
[ of vm1['BR']=='vM1'] [['Pia_Precision_Std',]].values[0][0],\

[ of vm1['BR']=='vM1'] [['Pia_Precision_Std',]].values[0][0][].

[ of vm1['BR']=='vM1'] [['Pia_Precision_Std',]].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values[[vm1['BR']=='vM1']].values
 →df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use_Projections_Alone') &_
  → (df_s1bf['BR']=='vM1')][['Pia Precision_Std',]].mean().values[0],\
                                         ],\
                                         )
rects3 = ax.
  →bar(ind+2*width,[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface')
  →& (df_global['BR']=='vM1')][['Barrels_Pia_Precision_Mean',]].mean().
 \rightarrow values [0],
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & U
 →df_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Rabies_Surfaces')
  →& (df_global['BR']=='vM1')][['Barrels_Pia_Precision_Mean',]].mean().
  \rightarrow values [0],\
                                         df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Center') &_
  → (df_vm1['BR']=='vM1')][['Barrels_Pia_Precision_Mean',]].mean().values[0],\
 →df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use Projections Alone') & ...

→ (df_s1bf['BR']=='vM1')][['Barrels_Pia_Precision_Mean',]].mean().values[0],\
                                      ],\
                                         width, color='blue', \
  → (df_global['BR']=='vS1')][['Barrels_Pia_Precision_std',]].mean().values[0],\
 \rightarrowdf_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Centers') &
  → (df_global['BR']=='vS1')][['Barrels_Pia_Precision_std',]].mean().values[0],\
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')⊔
  →& (df_global['BR']=='vS1')][['Barrels_Pia_Precision_std',]].mean().
 \rightarrow values [0],
                                         df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Center') &__
 → (df_vm1['BR']=='vS1')][['Barrels_Pia_Precision_std',]].mean().values[0],\
 →df s1bf[(df s1bf['Inner Alignment Type']=='Use Projections Alone') & ...
  → (df_s1bf['BR']=='vS1')][['Barrels_Pia_Precision_std',]].mean().values[0],\
                                         ],\
                                         )
```

```
ticks = plt.yticks(np.arange(0,1500,100))
\#rects2 = ax.bar(ind+width, \dots
 → df_qlobal[(df_qlobal['Inner Alignment Type']=='Using_Local Centers') &
 \rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0], 
 ⇒width, color='blue', \
  \neg yerr = df\_global[(df\_global['Inner\_Alignment\_Type'] == 'Using\_Local\_Centers') \  \, \mathfrak{E}_{\square} 
 \rightarrow (df qlobal['BR']=='WB')][['Surf Center Precision Std',]].mean().values[0])
\#rects1 = ax.bar(ind, \square)
 \rightarrow df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') 
 → (df qlobal['BR']=='WB')][['Surf Center Precision Mean',]].mean().values[0],
\rightarrow width, color='red', \
 →yerr=df_qlobal[(df_qlobal['Inner_Alignment_Type']=='Using_Surface') &
 → (df qlobal['BR']=='WB')][['Surf Center Precision Std',]].mean().values[0])
\#rects2 = ax.bar(ind+width, \dots)
 → df qlobal[(df qlobal['Inner Alignment Type'] == 'Using Local Centers') &
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0], 
\rightarrow width, color='blue', \
#
                 ш
 yerr=df qlobal[(df qlobal['Inner Alignment Type']=='Using Local Centers') &
 → (df qlobal['BR']=='WB')][['Surf Center Precision Std',]].mean().values[0])
plt.tight_layout()
fig.savefig(input_path + 'Fig_3/Pia_Precision.eps')
```



6.3 WM Precisions

```
[527]: N=5
      ind = np.arange(N)
      width = 0.25
      fig = plt.figure(figsize=(10*cm,7*cm),dpi=120)
      ax = fig.add_subplot(111)
      rects1 = ax.
       →bar(ind,[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') &
       → (df_global['BR']=='WB')][['WM_Precision_Mean',]].mean().values[0],\
       →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & ∪
       → (df_global['BR']=='WB')][['WM_Precision_Mean',]].mean().values[0],\
       →df_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Rabies_Surfaces')
       →& (df global['BR']=='WB')][['WM_Precision_Mean',]].mean().values[0],\
                           df vm1[(df vm1['Inner Alignment Type']=='Using Surface') & ...

    → (df_vm1['BR']=='WB')][['WM_Precision_Mean',]].mean().values[0],\

       →df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use Projections Alone') & ...
       → (df_s1bf['BR']=='WB')][['WM_Precision_Mean',]].mean().values[0],\
                          ],\
                           width, color='red', \
       → (df_global['BR']=='WB')][['WM_Precision_Std',]].mean().values[0],\
       →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & ∪
       → (df_global['BR']=='WB')][['WM_Precision_Std',]].mean().values[0],\
       →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')⊔
       →& (df_global['BR']=='WB')][['WM_Precision_Std',]].mean().values[0],\
                           df_vm1[(df_vm1['Inner_Alignment_Type'] == 'Using_Surface') &__

    → (df_vm1['BR']=='WB')][['WM_Precision_Std',]].mean().values[0],\
       →df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use_Projections_Alone') & U

    → (df_s1bf['BR']=='WB')][['WM_Precision_Std',]].mean().values[0],\
                           ],\
                           )
      rects2 = ax.
       →bar(ind+width, [df_global[(df_global['Inner_Alignment_Type']=='Using_Center')
       →& (df global['BR']=='vM1')][['WM Precision Mean',]].mean().values[0],\
```

```
→df_global[(df_global['Inner Alignment Type']=='Using_Local_Centers') &
 → (df_global['BR']=='vM1')][['WM Precision Mean',]].mean().values[0],\
→df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')⊔
 →& (df global['BR']=='vM1')][['WM Precision Mean',]].mean().values[0],\
                      df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Surface') & ∪

df_vm1['BR']=='vM1')][['WM_Precision_Mean',]].values[0][0],\

(df_vm1['BR']=='vM1')][['WM_Precision_Mean',]].values[0][0],\

(df_vm1['BR']=='vM1')][['WM_Precision_Mean',]].values[0][0],\

(df_vm1['BR']=='vM1')][['WM_Precision_Mean',]].values[0][0],\
 →df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use_Projections_Alone') & U

→ (df_s1bf['BR']=='vM1')][['WM_Precision_Mean',]].mean().values[0],\
                     ],\
                      width, color='green', \
 → (df_global['BR']=='vM1')][['WM_Precision_Std',]].mean().values[0],\
-df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & ∪

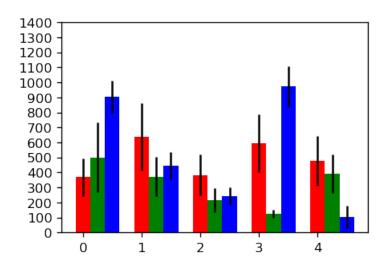
    →(df_global['BR']=='vM1')][['WM_Precision_Std',]].mean().values[0],\

¬df_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Rabies_Surfaces')
□
 →& (df global['BR']=='vM1')][['WM Precision Std',]].mean().values[0],\
                      df vm1[(df vm1['Inner Alignment Type']=='Using Surface') &___
→df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use_Projections_Alone') & U
 \hookrightarrow (df_s1bf['BR']=='vM1')][['WM_Precision_Std',]].mean().values[0],\
                      ],\
                      )
rects3 = ax.
 →bar(ind+2*width,[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface')
 →& (df global['BR']=='vM1')][['Barrels_WM_Precision_Mean',]].mean().
→values[0],\
-df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') &∟
→ (df_global['BR']=='vM1')][['Barrels_WM_Precision_Mean',]].mean().values[0],\
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_
 →& (df_global['BR']=='vM1')][['Barrels_WM_Precision_Mean',]].mean().
 \rightarrow values [0],\
                      df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Center') &_

    →(df_vm1['BR']=='vM1')][['Barrels_WM_Precision_Mean',]].mean().values[0],\
{}_{\hookrightarrow} df\_s1bf [(df\_s1bf['Inner\_Alignment\_Type'] == 'Use\_Projections\_Alone') \ \&_{\sqcup}

→ (df_s1bf['BR'] == 'vM1')][['Barrels_WM_Precision_Mean',]].mean().values[0],\
```

```
width, color='blue', \
 →yerr=[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') & Using_Surface'
 → (df_global['BR']=='vS1')][['Barrels_WM_Precision_Std',]].mean().values[0],\
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & U
 → (df global['BR'] == 'vS1')][['Barrels WM Precision Std',]].mean().values[0],\
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')__
 →& (df_global['BR']=='vS1')][['Barrels_WM_Precision_Std',]].mean().values[0],\
                      df vm1[(df vm1['Inner Alignment Type'] == 'Using Center') &
 → (df_vm1['BR']=='vS1')][['Barrels_WM_Precision_Std',]].mean().values[0],\
→df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use_Projections_Alone') &_
 → (df_s1bf['BR']=='vS1')][['Barrels_WM_Precision_Std',]].mean().values[0],\
                      ],\
ticks = plt.yticks(np.arange(0,1500,100))
\#rects2 = ax.bar(ind+width, \square)
→ df qlobal[(df qlobal['Inner Alignment Type']=='Using Local Centers') &
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0], <math>\sqcup
⇒width, color='blue', \
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Std',]].mean().values[0])
\#rects1 = ax.bar(ind, \square)
→df_qlobal[(df_qlobal['Inner_Alignment_Type']=='Using_Surface') &
\hookrightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0], <math>\sqcup
→width, color='red', \
\rightarrow yerr=df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') 
\rightarrow (df\_global['BR'] == 'WB')][['Surf\_Center\_Precision\_Std',]].mean().values[0])
\#rects2 = ax.bar(ind+width, \square
\rightarrow df_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Centers') 
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0], 
→width, color='blue', \
→yerr=df qlobal[(df qlobal['Inner Alignment Type']=='Using Local Centers') &
\hookrightarrow (df_global['BR']=='WB')][['Surf_Center_Precision_Std',]].mean().values[0])
plt.tight_layout()
fig.savefig(input_path + 'Fig_3/WM_Precision.eps')
```



6.4 Angular Precisions

```
[541]: N=5
      ind = np.arange(N)
      width = 0.25
      fig = plt.figure(figsize=(10*cm,7*cm),dpi=120)
      ax = fig.add_subplot(111)
      rects1 = ax.
       →bar(ind,[df_global['Inner_Alignment_Type']=='Using_Surface') &
       → (df global['BR'] == 'WB')][['PCA Angular Precision Mean',]].mean().values[0],\
       \hookrightarrow df_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Centers') & \sqcup

    → (df_global['BR']=='WB')][['PCA_Angular_Precision_Mean',]].mean().values[0],\

¬df_global[(df_global['Inner_Alignment_Type'] == 'Using_Local_Rabies_Surfaces')
□
       →values[0],\
       →df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Rabies_Surface') &<sub>□</sub>

    →(df_vm1['BR']=='WB')][['PCA_Angular_Precision_Mean',]].mean().values[0],\
       → (df_s1bf['BR']=='WB')][['PCA_Angular_Precision_Mean',]].mean().values[0],\
                        ],\
                         width, color='red', \
```

```
→yerr=[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') & Using_Surface'
 → (df_global['BR']=='WB')][['PCA_Angular_Precision_Std',]].mean().values[0],\
 df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') كال
 → (df global['BR']=='WB')][['PCA Angular Precision Std',]].mean().values[0],\
→df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_
 →& (df_global['BR']=='WB')][['PCA_Angular_Precision_Std',]].mean().values[0],\
→df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Rabies_Surface') &_
 → (df_vm1['BR']=='WB')][['PCA_Angular_Precision_Std',]].mean().values[0],\
→df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use Projections Alone') & □

→ (df_s1bf['BR'] == 'WB')][['PCA_Angular_Precision_Std',]].mean().values[0],\
                      ],\
                      )
rects2 = ax.
 →bar(ind+width, [df_global [(df_global ['Inner_Alignment_Type'] == 'Using_Surface')
 \rightarrow values [0],\
-df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Centers') & ∪

df_global['BR']=='vM1')][['PCA_Angular_Precision_Mean',]].mean().values[0],\

df_global['BR']=='vM1')][['PCA_Angular_Precision_Mean',]].mean().values[0],\
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_
 →& (df_global['BR']=='vM1')][['PCA_Angular_Precision_Mean',]].mean().
 →values[0],\
                      df_vm1[(df_vm1['Inner_Alignment_Type'] == 'Using_Surface') &__

→ (df_vm1['BR']=='vM1')][['PCA_Angular_Precision_Mean',]].values[0][0],\
→df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use Projections Alone') & ...
 → (df_s1bf['BR']=='vM1')][['PCA_Angular_Precision_Mean',]].mean().values[0],\
                     1.\
                      width, color='green', \
-yerr=[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') & ∪

→ (df_global['BR']=='vM1')][['PCA_Angular_Precision_Std',]].mean().values[0],\
{}_{\hookrightarrow} df\_global[(df\_global['Inner\_Alignment\_Type'] == 'Using\_Local\_Centers') \ \&_{\sqcup}

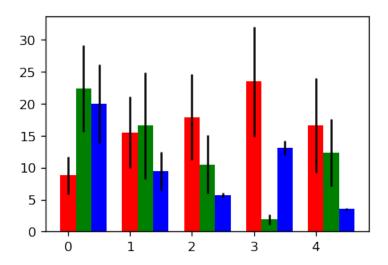
→ (df_global['BR']=='vM1')][['PCA_Angular_Precision_Std',]].mean().values[0],\
-df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_
→& (df_global['BR']=='vM1')][['PCA Angular Precision Std',]].mean().
 \rightarrow values [0],\
```

```
df_vm1[(df_vm1['Inner_Alignment_Type'] == 'Using_Surface') &__
 → (df vm1['BR']=='vM1')][['PCA Angular Precision Std',]].values[0][0],\
 →df s1bf[(df s1bf['Inner Alignment Type'] == 'Use Projections Alone') & ...
 → (df_s1bf['BR']=='vM1')][['PCA_Angular_Precision_Std',]].mean().values[0],\
                      ],\
                      )
rects3 = ax.
 →bar(ind+2*width,[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface')
 →& (df_global['BR']=='vM1')][['Barrels_Angular_Precision_Mean',]].mean().
 \rightarrow values [0],
→df_global[(df_global['Inner Alignment_Type']=='Using_Local_Centers') &
 → (df_global['BR']=='vM1')][['Barrels_Angular_Precision_Mean',]].mean().
 →values[0],\
 →df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_⊔
 →& (df_global['BR']=='vM1')][['Barrels_Angular_Precision_Mean',]].mean().
\rightarrow values [0],\
 →df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Rabies_Surface') & Using_Rabies_Surface'
 → (df_vm1['BR']=='vM1')][['Barrels_Angular_Precision_Mean',]].mean().
 →values[0],\
→df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use_Projections_Alone') & U
 → (df_s1bf['BR']=='vM1')][['Barrels_Angular_Precision_Mean',]].mean().
 \rightarrowvalues[0],\
                     ],\
                      width, color='blue', \
 -yerr=[df_global[(df_global['Inner_Alignment_Type']=='Using_Surface') & ∪
 → (df_global['BR']=='vS1')][['Barrels_Angular_Precision_Std',]].mean().
 \rightarrow values [0],
 →df_global[(df_global['Inner Alignment Type']=='Using_Local_Centers') &

→ (df_global['BR']=='vS1')][['Barrels_Angular_Precision_Std',]].mean().
 →values[0],\
→df_global[(df_global['Inner_Alignment_Type']=='Using_Local_Rabies_Surfaces')_⊔
 →& (df_global['BR']=='vS1')][['Barrels_Angular_Precision_Std',]].mean().
 \rightarrow values [0],
 →df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Rabies_Surface') & U

→ (df_vm1['BR']=='vS1')][['Barrels_Angular_Precision_Std',]].mean().values[0],\
```

```
⇒df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use_Projections_Alone') & 
 → (df_s1bf['BR']=='vS1')][['Barrels_Angular_Precision_Std',]].mean().
 \rightarrow values [0],\
                    ],\
                    )
\#ticks = plt.yticks(np.arange(0, 1300, 100))
\#rects2 = ax.bar(ind+width, \square
→ df_qlobal[(df_qlobal['Inner Alignment Type']=='Using_Local Centers') &
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0], <math>\sqcup
⇒width, color='blue', \
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Std',]].mean().values[0])
\#rects1 = ax.bar(ind, \dots)
→df qlobal[(df qlobal['Inner Alignment Type']=='Using Surface') &
→ (df qlobal['BR'] == 'WB')][['Surf Center Precision Mean',]].mean().values[0],
⇒width, color='red', \
\rightarrow (df\_global['BR'] == 'WB')][['Surf\_Center\_Precision\_Std',]].mean().values[0])
\#rects2 = ax.bar(ind+width, \square
→df_qlobal[(df_qlobal['Inner_Alignment_Type']=='Using_Local_Centers') &
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Mean',]].mean().values[0], 
\rightarrow width, color='blue'.
yerr=df qlobal[(df qlobal['Inner Alignment Type']=='Using Local Centers') &
\rightarrow (df_global['BR'] == 'WB')][['Surf_Center_Precision_Std',]].mean().values[0])
plt.tight_layout()
fig.savefig(input_path + 'Fig_3/Angular_Precision.eps')
```



```
[577]: df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use_Projections_Alone') &__

→ (df_s1bf['BR']=='vS1')][['Barrels_Angular_Precision_Mean',]]
[577]:
          Barrels_Angular_Precision_Mean
       6
                                3.249892
       7
                                3.805283
[571]: (1/55)*38
[571]: 0.6909090909090909
  []: 55
[576]: df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Using_Surface') &__
        → (df_vm1['BR']=='vM1')][['Barrels_Angular_Precision_Mean',]].mean()
[576]: PCA_Angular_Precision_Mean
                                     5.334361
       dtype: float64
[531]: df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Center') &__

→ (df_vm1['BR']=='vM1')][['PCA_Angular_Precision_Mean',]].mean().values[0]
[531]: 22.36244632510002
[532]: df_vm1.columns
[532]: Index(['Unnamed: 0', 'Angular_Error_SEM', 'Angular_Precision_Mean',
              'Angular_Precision_Std', 'BR', 'Barrel_Cortex_Center_Ellipsoid',
              'Barrel_Cortex_Center_Ellipsoid_Norm',
              'Barrel_Cortex_Center_Error_Dists_SEM',
```

```
'Barrel_Cortex_Center_Precision_Mean',
 'Barrel_Cortex_Center_Precision_Std', 'Barrels_Angular_Error_SEM',
 'Barrels Angular Precision Mean', 'Barrels Angular Precision Std',
 'Barrels_Barrel_Center_Precision_Mean',
 'Barrels_Barrel_Center_Precision_Std',
 'Barrels_Barrel_Centers_Error_Ellipsoid_Mean',
 'Barrels Barrel Centers Error Ellipsoid Mean Norm',
 'Barrels_Barrel_Centers_Error_Ellipsoid_Std',
 'Barrels Barrel Centers Error Ellipsoid Std Norm',
 'Barrels Column Center Precision Mean',
 'Barrels Column Center Precision Std',
 'Barrels_Column_Error_Ellipsoid_Mean',
 'Barrels_Column_Error_Ellipsoid_Mean_Norm',
 'Barrels_Column_Error_Ellipsoid_Std',
 'Barrels_Column_Error_Ellipsoid_Std_Norm',
 'Barrels_Pia_Error_Ellipsoid_Mean',
 'Barrels_Pia_Error_Ellipsoid_Mean_Norm',
 'Barrels_Pia_Error_Ellipsoid_Std',
 'Barrels_Pia_Error_Ellipsoid_Std_Norm', 'Barrels_Pia_Precision_Mean',
 'Barrels_Pia_Precision_std', 'Barrels_WM_Error_Ellipsoid_Mean',
 'Barrels_WM_Error_Ellipsoid_Mean_Norm',
 'Barrels_WM_Error_Ellipsoid_Std', 'Barrels_WM_Error_Ellipsoid_Std_Norm',
 'Barrels_WM_Precision_Mean', 'Barrels_WM_Precision_Std', 'Hem',
 'Inner Alignment Type', 'Number Of Axis Used', 'Outer Alignment Type',
 'Overall_Rabies_landmarks', 'Overall_Volume',
 'Overlap_Rabies_landmarks', 'Overlap_Volume',
 'PCA_O_Angle_Between_Axes_Mean', 'PCA_O_Angle_Between_Axes_Std',
 'PCA_1_Angle_Between_Axes_Mean', 'PCA_1_Angle_Between_Axes_Std',
 'PCA_2_Angle_Between_Axes_Mean', 'PCA_2_Angle_Between_Axes_Std',
 'PCA_Angular_Precision_Mean', 'PCA_Angular_Precision_SEM',
 'PCA_Angular_Precision_Std', 'Pia_Error_Dist_SEM',
'Pia_Error_Ellipsoid_Mean', 'Pia_Error_Ellipsoid_Mean_Norm',
 'Pia_Error_Ellipsoid_Std', 'Pia_Error_Ellipsoid_Std_Norm',
 'Pia_Precision_Mean', 'Pia_Precision_Std', 'Rabies_Center_Ellipsoid',
 'Rabies_Center_Ellipsoid_Norm', 'Rabies_Center_Error_Dists_SEM',
 'Rabies_Center_Precision_Mean', 'Rabies_Center_Precision_Std',
 'Rabies_Landmarks_Overlap_Percentage', 'Rabies_Surf_Center_Dists_Mean',
 'Rabies_Surf_Center_Dists_Std', 'Rabies_Surf_Center_Ellipsoid',
 'Rabies Surf Center Ellipsoid Norm',
 'Rabies_Surf_Center_Error_Dists_SEM', 'Surf_Center_Ellipsoid',
 'Surf_Center_Ellipsoid_Norm', 'Surf_Center_Error_Dists_SEM',
 'Surf_Center_Precision_Mean', 'Surf_Center_Precision_Std',
 'Volume_Overlap_Percentage', 'WM_Error_Dist_SEM',
 'WM_Error_Ellipsoid_Mean', 'WM_Error_Ellipsoid_Mean_Norm',
 'WM_Precision_Mean', 'WM_Precision_Std'],
dtype='object')
```

```
[546]: | df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Surface') &__
                  #[(30.8+59.1)/2, (54.5+106.7)/2, (61.9+51.3)/2]
[546]:
                                                                WM_Error_Ellipsoid_Mean
                                [49.1533137 77.38980311 78.4623976 ]
                16
                17
                         [ 84.96801149 108.21712732 179.83579373]
[557]:
[557]: 1.1636363636363636
[551]: df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Center') &__

    → (df_vm1['BR'] == 'vM1')][['Rabies_Surf_Center_Ellipsoid',]]
[551]:
                                                   Rabies_Surf_Center_Ellipsoid
                     [ 42.27441856 110.01097201 84.38809675]
                              [58.50609879 14.00859704 95.14052509]
                5
[535]: df_vm1[(df_vm1['Inner_Alignment_Type']=='Using_Center') &__
                  → (df_vm1['BR']=='vM1')][['Surf_Center_Precision_Mean',]].values[0][0]
[535]: 289.59023782377807
[536]: | df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use Projections Alone') &__
                  [536]:
                         Barrels_Pia_Precision_Mean
                10
                                                              112.001154
                                                                87.909474
                11
[537]: df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use_Projections_Alone') &__

    →(df_s1bf['BR']=='vM1')][['Barrels_WM_Precision_Mean']]

                         Barrels_WM_Precision_Mean
[537]:
                10
                                                            118.938618
                11
                                                              85.799177
[538]: df_s1bf[(df_s1bf['Inner_Alignment_Type']=='Use_Barrels_Alone') &_

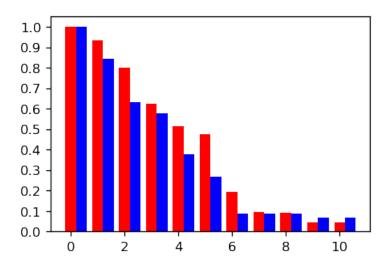
df_s1bf['BR']=='vM1')][['Barrels_Angular_Precision_Mean',]]

display="block" | Graph | Comparison | Compar
                #[(30.8+59.1)/2, (54.5+106.7)/2, (61.9+51.3)/2]
[538]:
                      Barrels_Angular_Precision_Mean
                4
                                                                          6.085805
                5
                                                                          3.800505
[539]: | df_s1bf[(df_s1bf['Inner_Alignment_Type'] == 'Use Projections Alone') &__

→ (df_s1bf['BR']=='vM1')][['Barrels_Angular_Precision_Mean',]]
```

```
#[(
           Barrels_Angular_Precision_Mean
[539]:
       10
                                   3.249892
       11
                                   3.805283
          Density Cluster Analysis for Registration
[578]: df = pd.read_csv('Z:/V11/Registration_Final_Coronal_Ref_Frame/Input_Ref_Frame/
        →Local_Ref_Frame_View/density_clusters.csv')
[580]: df
[580]:
           Unnamed: 0
                        dens_cutoff
                                     morph_count
                                                   morph_count_per
                                                                     rabies_count \
                     0
                                0.0
                                                           1.000000
       0
                                               90
                                                                              1567
                     0
                                0.1
                                               76
       1
                                                           0.844444
                                                                              1462
       2
                                0.2
                     0
                                               57
                                                                              1252
                                                           0.633333
       3
                     0
                                0.3
                                               52
                                                           0.577778
                                                                               976
                                0.4
                                               34
       4
                     0
                                                           0.377778
                                                                               804
                                0.5
       5
                     0
                                               24
                                                           0.266667
                                                                               743
       6
                     0
                                0.6
                                                8
                                                           0.088889
                                                                               303
       7
                     0
                                0.7
                                                8
                                                           0.088889
                                                                               148
       8
                     0
                                0.8
                                                8
                                                           0.088889
                                                                               143
       9
                     0
                                0.9
                                                6
                                                                                72
                                                           0.066667
       10
                     0
                                1.0
                                                6
                                                           0.066667
                                                                                72
           rabies_count_per
       0
                    1.000000
                    0.932993
       1
       2
                    0.798979
       3
                    0.622846
       4
                    0.513082
       5
                    0.474154
       6
                    0.193363
       7
                    0.094448
       8
                    0.091257
       9
                    0.045948
       10
                    0.045948
[592]: N=11
       ind = np.arange(N)
       width = 0.4
       fig = plt.figure(figsize=(10*cm,7*cm),dpi=120)
       ax = fig.add_subplot(111)
```

```
rects1 = ax.bar(ind,df['rabies_count_per'].values,width, color='red',)
rects2 = ax.bar(ind+width,df['morph_count_per'].values,width, color='blue',)
plt.yticks(np.arange(0,1.1,0.1))
plt.tight_layout()
fig.savefig(input_path + 'Fig_4/density_clustering.eps')
```



8 SBF old vs new

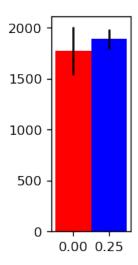
[31]: (152.35663949166354, 197.38083968394773, 347.75, 17.40377270444044)

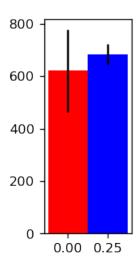
→std(),df_old['Barrel Height'].mean(),df_old['Barrel Height'].std()

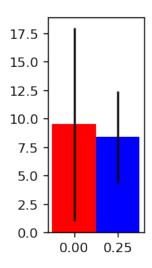
[31]: df_new['Barrel_Height'].mean(),df_new['Barrel_Height'].

```
[34]: df_new['Column_Height'].mean(),df_new['Column_Height'].

→std(),df_old['Column_Height'].mean(),df_old['Column_Height'].std()
[34]: (1772.2760047070042, 234.6472462002318, 1887.833333333333, 155.70531438437234)
[38]: df_new['Barrel_Center'].mean(),df_new['Barrel_Center'].std(),\
      (df_new['Barrel_Top'].mean()+df_new['Barrel_Bottom'].mean())/
       →2, (df_new['Barrel_Top'].std()+df_new['Barrel_Bottom'].std())/2,
[38]: (621.0285499617579, 157.62421794701226, 629.1231474012748, 167.91383888268817)
[42]: df_old['Column_Orientation'].mean(),df_old['Column_Orientation_Std'].mean()
[42]: (8.375, 4.05416666666666)
[48]: df_new_orientations['Orientation'].mean(),df_new_orientations['Orientation'].
       →std()
[48]: (9.50977985178146, 8.469195617154067)
[62]: N=1
      ind = np.arange(N)
      width = 0.25
      fig = plt.figure(figsize=(4*cm,7*cm),dpi=120)
      ax = fig.add_subplot(111)
      rects1 = ax.bar(ind,[df_new['Column_Height'].mean(),],\
               width, color='red', \
               yerr=[df_new['Column_Height'].std()],\
      rects2 = ax.bar(ind+width,[df_old['Column_Height'].mean(),],\
               width, color='blue', \
               yerr=[df_old['Column_Height_Std'].mean()],\
      plt.tight_layout()
      fig.savefig('D:/vM1/Fig_2/column_ht_comparison.eps')
```







[]: