main_junctions

July 11, 2021

1 Tissue comparison for differential expression analysis

```
[1]: import functools
     import numpy as np
     import pandas as pd
     from plotnine import *
     from scipy.stats import binom_test, fisher_exact, linregress
     from warnings import filterwarnings
     from matplotlib.cbook import mplDeprecation
     filterwarnings('ignore', category=mplDeprecation)
     filterwarnings('ignore', category=UserWarning, module='plotnine.*')
     filterwarnings('ignore', category=DeprecationWarning, module='plotnine.*')
[2]: config = {
         'caudate': '../../caudate/_m/junctions/diffExpr_maleVfemale_full.txt',
         'dlpfc': '../../dlpfc/_m/junctions/diffExpr_maleVfemale_full.txt',
         'hippo': '../../hippocampus/ m/junctions/diffExpr maleVfemale_full.txt',
     }
[3]: Ofunctools.lru_cache()
     def get_deg(filename):
         dft = pd.read_csv(filename, sep='\t', index_col=0)
         dft['Feature'] = dft.index
         dft['Dir'] = np.sign(dft['t'])
         if 'gene_id' in dft.columns:
             dft['ensemblID'] = dft.gene_id.str.replace('\\..*', '', regex=True)
         return dft[['Feature', 'ensemblID', 'adj.P.Val', 'logFC', 't', 'Dir']]
     @functools.lru_cache()
     def get_deg_sig(filename, fdr):
         dft = get_deg(filename)
         return dft[(dft['adj.P.Val'] < fdr)]</pre>
     @functools.lru_cache()
     def merge_dataframes(tissue1, tissue2):
```

```
return get_deg(config[tissue1]).merge(get_deg(config[tissue2]),
                                               on='Feature', suffixes=['_%s' %__
     →tissue1, '_%s' % tissue2])
     @functools.lru cache()
     def merge_dataframes_sig(tissue1, tissue2):
         fdr1 = 0.05 if tissue1 != 'dlpfc' else 0.05
         fdr2 = 0.05 if tissue2 != 'dlpfc' else 0.05
         return get_deg_sig(config[tissue1], fdr1).
     →merge(get_deg_sig(config[tissue2], fdr2),
                                                          on='Feature', ...

→suffixes=['_%s' % tissue1, '_%s' % tissue2])
[4]: def enrichment_binom(tissue1, tissue2, merge_fnc):
         df = merge_fnc(tissue1, tissue2)
         df['agree'] = df['Dir_%s' % tissue1] * df['Dir_%s' % tissue2]
         dft = df.groupby('agree').size().reset_index()
         print(dft)
         return binom_test(dft[0].iloc[1], dft[0].sum()) if dft.shape[0] != 1 else__
     →print("All directions agree!")
     def cal_fishers(tissue1, tissue2):
         df = merge dataframes(tissue1, tissue2)
         fdr1 = 0.05 if tissue1 != 'dlpfc' else 0.05
         fdr2 = 0.05 if tissue2 != 'dlpfc' else 0.05
         table = [[np.sum((df['adj.P.Val_%s' % tissue1]<fdr1) &
                          ((df['adj.P.Val_%s' % tissue2]<fdr2))),
                   np.sum((df['adj.P.Val_%s' % tissue1]<fdr1) &
                          ((df['adj.P.Val_%s' % tissue2]>=fdr2)))],
                  [np.sum((df['adj.P.Val_%s' % tissue1]>=fdr1) &
                          ((df['adj.P.Val_%s' % tissue2]<fdr2))),
                   np.sum((df['adj.P.Val_%s' % tissue1]>=fdr1) &
                          ((df['adj.P.Val_%s' % tissue2]>=fdr2)))]]
         print(table)
         return fisher_exact(table)
     def calculate corr(xx, yy):
         '''This calculates R^2 correlation via linear regression:
              - used to calculate relationship between 2 arrays
              - the arrays are principal components 1 or 2 (PC1, PC2) AND gender
              - calculated on a scale of 0 to 1 (with 0 being no correlation)
             Inputs:
               x: array of Gender (converted to binary output)
               y: array of PC
```

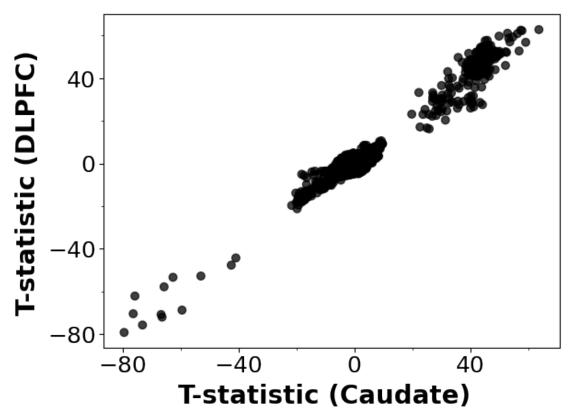
```
Outputs:
               1. r2
               2. p-value, two-sided test
                 - whose null hypothesis is that two sets of data are uncorrelated
               3. slope (beta): directory of correlations
        slope, intercept, r_value, p_value, std_err = linregress(xx, yy)
        return r_value, p_value
    def corr_annotation(tissue1, tissue2, merge_fnc):
        dft = merge_fnc(tissue1, tissue2)
        xx = dft['t_%s' % tissue1]
        yy = dft['t_%s' % tissue2]
        r_value1, p_value1 = calculate_corr(xx, yy)
        return 'R2: %.2f\nP-value: %.2e' % (r_value1**2, p_value1)
    def tissue_annotation(tissue):
        return {'dlpfc': 'DLPFC', 'hippo': 'Hippocampus', 'caudate':
     [5]: def plot_corr_impl(tissue1, tissue2, merge_fnc):
        dft = merge fnc(tissue1, tissue2)
        title = '\n'.join([corr_annotation(tissue1, tissue2, merge_fnc)])
        xlab = 'T-statistic (%s)' % tissue annotation(tissue1)
        ylab = 'T-statistic (%s)' % tissue_annotation(tissue2)
        pp = ggplot(dft, aes(x='t_%s'%tissue1, y='t_%s' % tissue2))\
        + geom_point(alpha=0.75, size=3)\
        + theme_matplotlib()\
        + theme(axis_text=element_text(size=18),
                 axis_title=element_text(size=20, face='bold'),
                 plot_title=element_text(size=22))
        pp += labs(x=xlab, y=ylab, title=title)
        return pp
    def plot_corr(tissue1, tissue2, merge_fnc):
        return plot_corr_impl(tissue1, tissue2, merge_fnc)
    def save_plot(p, fn, width=7, height=7):
         '''Save plot as sug, png, and pdf with specific label and dimension.'''
        for ext in ['.svg', '.png', '.pdf']:
            p.save(fn+ext, width=width, height=height)
```

1.1 Sample summary

```
[6]: pheno_file = '/ceph/projects/v3_phase3_paper/inputs/phenotypes/merged/_m/
      →merged_phenotypes.csv'
      pheno = pd.read_csv(pheno_file, index_col=0)
      pheno = pheno[(pheno['Age'] > 17) & (pheno['Dx'].isin(['Schizo', 'Control']))]
      pheno.head(2)
 [6]:
               BrNum
                        RNum
                               Region RIN
                                              Age Sex Race
                                                                 Dx
      R12864 Br1303 R12864
                              Caudate
                                       9.6
                                            42.98
                                                    F
                                                         AA
                                                            Schizo
      R12865 Br1320 R12865
                              Caudate 9.5 53.12
                                                    М
                                                        AA
                                                            Schizo
 [7]: pheno.groupby(['Region']).size()
 [7]: Region
      Caudate
                 394
      DLPFC
                 379
      HIPPO
                 376
      dtype: int64
 [8]: pheno.groupby(['Region', 'Sex']).size()
 [8]: Region
               Sex
      Caudate
              F
                      121
                      273
               Μ
      DLPFC
               F
                      117
                      262
     HIPPO
               F
                      121
               M
                      255
      dtype: int64
     1.2 BrainSeq Tissue Comparison
 [9]: caudate = get_deg(config['caudate'])
      caudate.groupby('Dir').size()
 [9]: Dir
      -1.0
              71339
       1.0
              87311
      dtype: int64
[10]: caudate[(caudate['adj.P.Val'] < 0.05)].shape
[10]: (772, 6)
[11]: dlpfc = get_deg(config['dlpfc'])
      dlpfc.groupby('Dir').size()
```

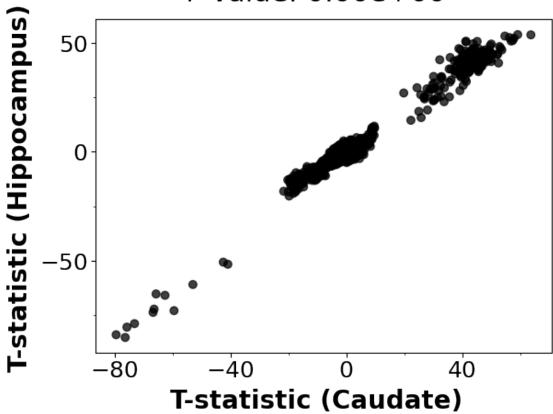
```
[11]: Dir
     -1.0
              68843
       1.0
              76371
      dtype: int64
[12]: dlpfc[(dlpfc['adj.P.Val'] < 0.05)].shape</pre>
[12]: (1809, 6)
[13]: hippo = get_deg(config['hippo'])
      hippo.groupby('Dir').size()
[13]: Dir
     -1.0
              69998
              72267
       1.0
      dtype: int64
[14]: hippo[(hippo['adj.P.Val'] < 0.05)].shape
[14]: (687, 6)
     1.2.1 Enrichment of DEG
[15]: cal_fishers('caudate', 'dlpfc')
     [[517, 187], [1226, 137962]]
[15]: (311.11284905479323, 0.0)
[16]: cal_fishers('caudate', 'hippo')
     [[491, 193], [175, 137620]]
[16]: (2000.6341968911918, 0.0)
[17]: cal_fishers('dlpfc', 'hippo')
     [[497, 1245], [150, 135279]]
[17]: (360.0196144578313, 0.0)
     1.2.2 Correlation
[18]: pp = plot_corr('caudate', 'dlpfc', merge_dataframes)
```

R2: 0.49 P-value: 0.00e+00

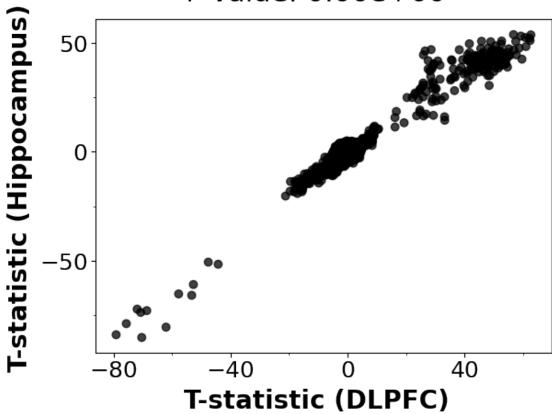


```
[18]: <ggplot: (8760097938044)>
[19]: qq = plot_corr('caudate', 'hippo', merge_dataframes)
    qq
```

R2: 0.58 P-value: 0.00e+00



R2: 0.48 P-value: 0.00e+00

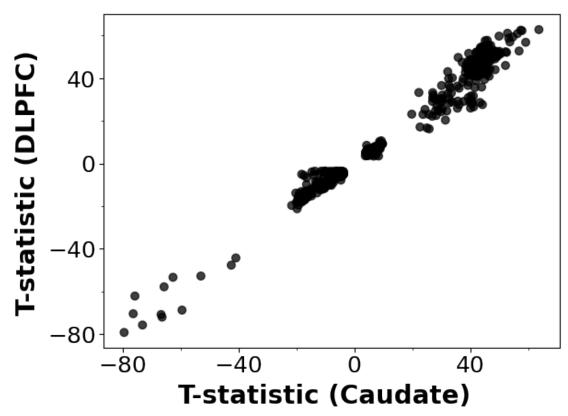


```
[20]: <ggplot: (8760095352227)>
```

1.2.3 Significant correlation, FDR < 0.05

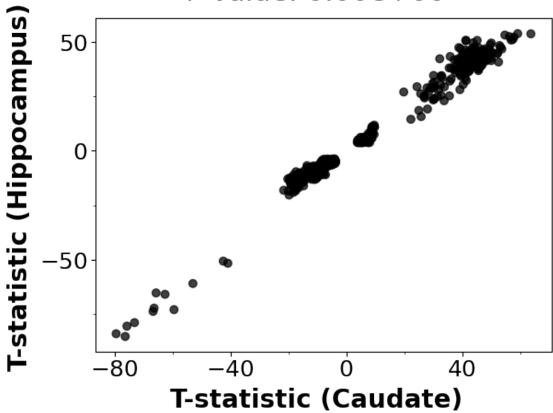
```
[21]: pp = plot_corr('caudate', 'dlpfc', merge_dataframes_sig)
pp
```

R2: 0.98 P-value: 0.00e+00

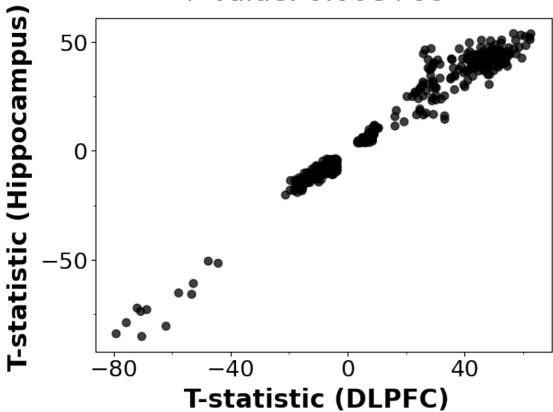


```
[21]: <ggplot: (8760091745515)>
[22]: qq = plot_corr('caudate', 'hippo', merge_dataframes_sig)
    qq
```

R2: 0.98 P-value: 0.00e+00



R2: 0.97 P-value: 0.00e+00



```
[23]: <ggplot: (8760091741284)>
     1.2.4 Directionality test
     All genes
[24]: enrichment_binom('caudate', 'dlpfc', merge_dataframes)
        agree
         -1.0
               61293
     1
          1.0
               78599
[24]: 1e-323
[25]: enrichment_binom('caudate', 'hippo', merge_dataframes)
        agree
                   0
         -1.0 57734
     0
          1.0 80745
```

```
[25]: 1e-323
[26]: enrichment_binom('dlpfc', 'hippo', merge_dataframes)
        agree
     0 -1.0 60570
          1.0 76601
     1
[26]: 1e-323
     Significant DEG (FDR < 0.05)
[27]: enrichment_binom('caudate', 'dlpfc', merge_dataframes_sig)
        agree
         1.0 517
     0
     All directions agree!
[28]: enrichment_binom('caudate', 'hippo', merge_dataframes_sig)
        agree
                 0
          1.0 491
     All directions agree!
[29]: enrichment_binom('dlpfc', 'hippo', merge_dataframes_sig)
        agree
         1.0 497
     All directions agree!
 []:
```