main_transcripts

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/transcripts/diffExpr_maleVfemale_full.txt',
         'dlpfc': '../../dlpfc/_m/transcripts/diffExpr_maleVfemale_full.txt',
         'hippo': '../../hippocampus/ m/transcripts/diffExpr_maleVfemale full.
     }
[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)
```

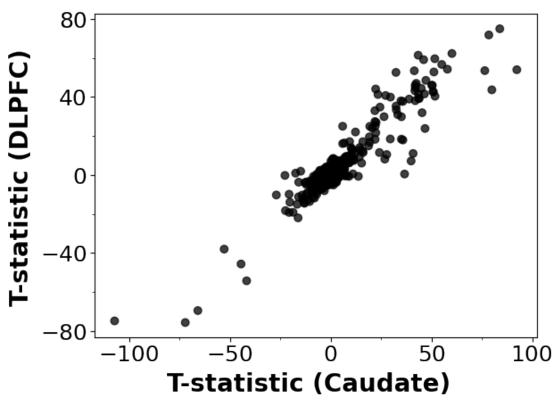
```
y: array of PC
             Outputs:
              1. r2
               2. p-value, two-sided test
                 - whose null hypothesis is that two sets of data are uncorrelated
              3. slope (beta): directory of correlations
         slope, intercept, r_value, p_value, std_err = linregress(xx, 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 svg, 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

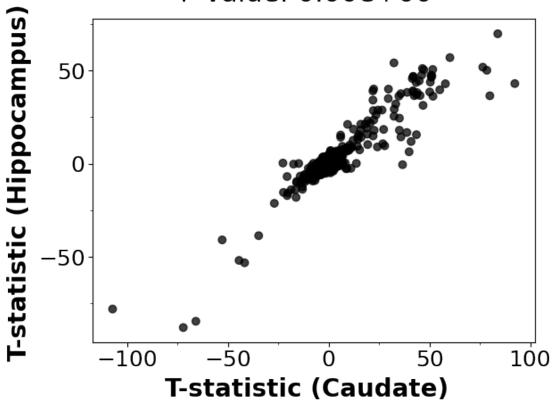
```
[6]: pheno_file = '/ceph/projects/v3_phase3_paper/inputs/phenotypes/merged/_m/
      →merged_phenotypes.csv'
      pheno = pd.read_csv(pheno_file, index_col=0)
      pheno = pheno[(pheno['Age'] > 17) & (pheno['Dx'].isin(['Schizo', 'Control']))]
      pheno.head(2)
 [6]:
               BrNum
                        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
              51647
       1.0
              55569
      dtype: int64
[10]: caudate[(caudate['adj.P.Val'] < 0.05)].shape
[10]: (462, 6)
[11]: dlpfc = get_deg(config['dlpfc'])
      dlpfc.groupby('Dir').size()
```

```
[11]: Dir
     -1.0
              38553
              43774
       1.0
      dtype: int64
[12]: dlpfc[(dlpfc['adj.P.Val'] < 0.05)].shape</pre>
[12]: (422, 6)
[13]: hippo = get_deg(config['hippo'])
      hippo.groupby('Dir').size()
[13]: Dir
     -1.0
              42370
       1.0
              39061
      dtype: int64
[14]: hippo[(hippo['adj.P.Val'] < 0.05)].shape
[14]: (252, 6)
     1.2.1 Enrichment of DEG
[15]: cal_fishers('caudate', 'dlpfc')
     [[184, 158], [216, 78168]]
[15]: (421.44022503516175, 0.0)
[16]: cal_fishers('caudate', 'hippo')
     [[183, 160], [60, 77793]]
[16]: (1482.9290625, 0.0)
[17]: cal_fishers('dlpfc', 'hippo')
     [[179, 224], [54, 77583]]
[17]: (1148.094990079365, 0.0)
     1.2.2 Correlation
[18]: pp = plot_corr('caudate', 'dlpfc', merge_dataframes)
```

R2: 0.38 P-value: 0.00e+00

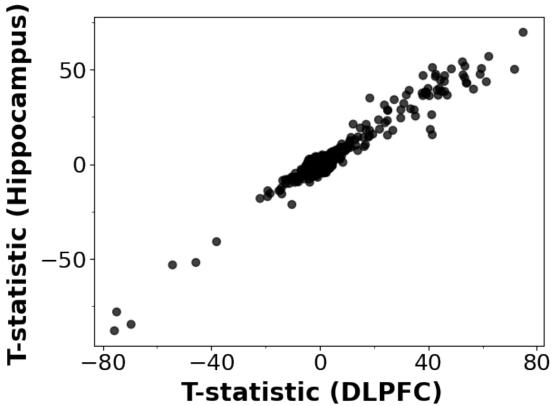


R2: 0.46 P-value: 0.00e+00



```
[19]: <ggplot: (8767858379578)>
[20]: ww = plot_corr('dlpfc', 'hippo', merge_dataframes)
ww
```

R2: 0.38 P-value: 0.00e+00

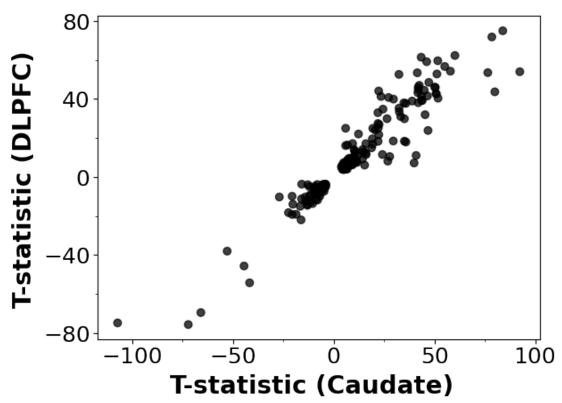


```
[20]: <ggplot: (8767856901834)>
```

1.2.3 Significant correlation, FDR < 0.05

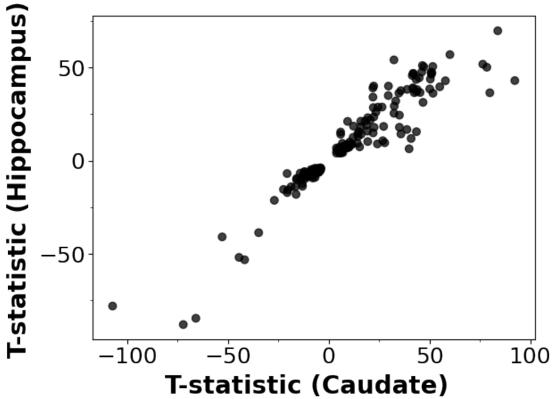
```
[21]: pp = plot_corr('caudate', 'dlpfc', merge_dataframes_sig)
pp
```

R2: 0.90 P-value: 9.82e-92

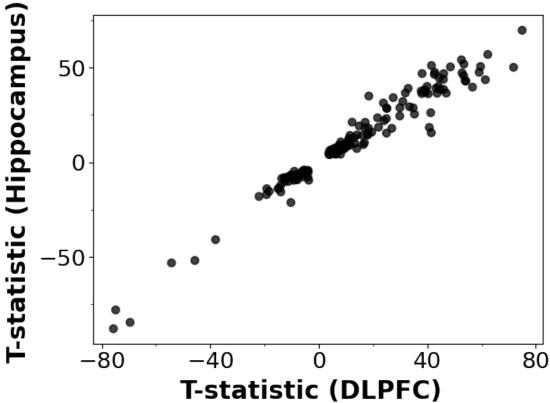


```
[21]: <ggplot: (8767854509135)>
[22]: qq = plot_corr('caudate', 'hippo', merge_dataframes_sig)
    qq
```

R2: 0.88 P-value: 3.17e-86



R2: 0.95 P-value: 2.60e-119



```
[25]: 1e-323
[26]: enrichment_binom('dlpfc', 'hippo', merge_dataframes)
        agree
     0 -1.0 35300
          1.0 42740
     1
[26]: 1.8250523208853138e-156
     Significant DEG (FDR < 0.05)
[27]: enrichment_binom('caudate', 'dlpfc', merge_dataframes_sig)
        agree
                0
     0
         1.0 184
     All directions agree!
[28]: enrichment_binom('caudate', 'hippo', merge_dataframes_sig)
        agree
                 0
          1.0 183
     All directions agree!
[29]: enrichment_binom('dlpfc', 'hippo', merge_dataframes_sig)
        agree
         1.0 179
     All directions agree!
 []:
```