main

July 10, 2021

1 Cell type enrichment analysis

```
[1]: import os, errno
import functools
import numpy as np
import pandas as pd
from matplotlib_venn import venn3
from matplotlib import pyplot as plt
from scipy.stats import fisher_exact
from statsmodels.stats.multitest import multipletests
```

1.1 Functions

1.1.1 Cached functions

```
[2]: @functools.lru_cache()
def get_deg(tissue):
    fn = '../../../differential_expression/%s/_m/genes/
    diffExpr_maleVfemale_full.txt' % tissue
    return pd.read_csv(fn, sep='\t', index_col=0)

@functools.lru_cache()
def get_xci():
    return pd.read_csv('../_h/xci_status_hg19.txt', sep='\t')
```

1.1.2 Simple functions

```
raise
def cal_fishers(status, tissue):
   deg = get_deg(tissue)
   xci = get_xci()
   xci['ensemblID'] = xci['Gene ID'].str.replace("\\..*", "", regex=True)
   df = deg.merge(xci, on='ensemblID', how='left')
   table = [[np.sum((df['adj.P.Val']<0.05) & (df['Combined XCI status'] ==_
 ⇔status)),
              np.sum((df['adj.P.Val']<0.05) & (df['Combined XCI status'] !=_

status))],
             [np.sum((df['adj.P.Val']>0.05) & (df['Combined XCI status'] ==_
⇒status)),
              np.sum((df['adj.P.Val']>0.05) & (df['Combined XCI status'] !=__
 →status))]]
    #print(table)
   return fisher exact(table)
def cal_fishers_direction(status, direction, tissue):
   deg = get deg(tissue)
   if direction == 'Up':
        deg = deg[(deg['t'] > 0)].copy()
   else:
        deg = deg[(deg['t'] < 0)].copy()
   xci = get_xci()
   xci['ensemblID'] = xci['Gene ID'].str.replace("\\..*", "", regex=True)
   df = deg.merge(xci, on='ensemblID', how='left')
   table = [[np.sum((df['adj.P.Val']<0.05) & (df['Combined XCI status'] ==__
⇒status)),
              np.sum((df['adj.P.Val']<0.05) & (df['Combined XCI status'] !=_
 ⇒status))],
             [np.sum((df['adj.P.Val']>0.05) & (df['Combined XCI status'] ==__
 ⇒status)),
              np.sum((df['adj.P.Val']>0.05) & (df['Combined XCI status'] !=|

status))]]
    #print(table)
   return fisher_exact(table)
def cal_fisher_by_xci_status(tissue):
   xci_status = xci = get_xci().loc[:, 'Combined XCI status'].unique()
   xci_lt = []; pval_lt = []; oddratio_lt = []; dir_lt = [];
   for status in xci_status:
        odd_ratio, pval = cal_fishers(status, tissue)
```

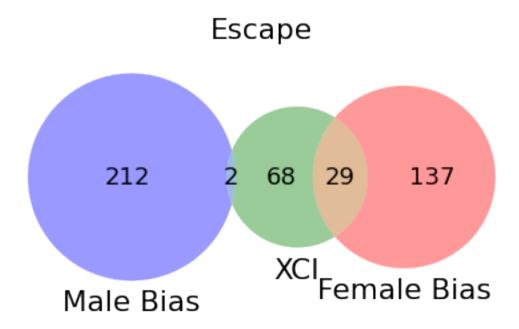
```
xci_lt.append(status); pval_lt.append(pval);
       oddratio_lt.append(odd_ratio); dir_lt.append('All')
       if pval < 0.05:</pre>
           print("There is a significant enrichment (p-value < %.1e) of %s!" %
                  (pval, status))
   for status in xci_status:
       odd_ratio, pval = cal_fishers_direction(status, "Up", tissue)
       xci lt.append(status); pval lt.append(pval);
       oddratio_lt.append(odd_ratio); dir_lt.append('Male Bias')
       if pval < 0.05:</pre>
           print("There is a significant enrichment of male bias genes⊔
\hookrightarrow (p-value < %.1e) of %s!" %
                  (pval, status))
   for status in xci_status:
       odd_ratio, pval = cal_fishers_direction(status, "Down", tissue)
       xci_lt.append(status); pval_lt.append(pval);
       oddratio_lt.append(odd_ratio); dir_lt.append('Female Bias')
       if pval < 0.05:
           print("There is a significant enrichment of female bias genes_
\rightarrow (p-value < %.1e) of %s!" %
                  (pval, status))
   return pd.DataFrame({"Tissue": tissue_annotation(tissue),
                         'XCI status': xci_lt, 'OR': oddratio_lt,
                         'PValue': pval_lt, 'Direction': dir_lt})
```

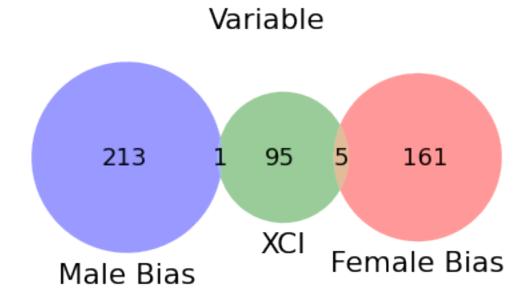
1.2 Gene set overlaps

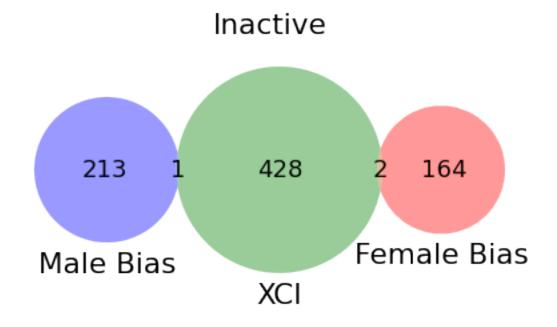
```
[4]: for tissue in ["caudate", "dlpfc", "hippocampus"]:
         mkdir p(tissue)
         print(tissue_annotation(tissue))
         deg = get_deg(tissue)
         xci = get_xci()
         xci['ensemblID'] = xci['Gene ID'].str.replace("\\..*", "", regex=True)
         xci_status = get_xci().loc[:, 'Combined XCI status'].unique()
         ## Plot venn diagram
         for status in xci status:
             setA = set(deg[(deg["adj.P.Val"] < 0.05) & (deg['t'] < 0)].ensemblID)
             setB = set(xci[(xci["Combined XCI status"] == status)].ensemblID)
             setC = set(deg[(deg["adj.P.Val"] < 0.05) & (deg['t'] > 0)].ensemblID)
             plt.rcParams.update({'font.size': 18})
             plt.figure(figsize=(7,7))
             v = venn3([setA, setB, setC], ("Female Bias", "XCI", "Male Bias"))
             plt.title(status.title())
             fn = '%s/venn_DEGs_%s' % (tissue, status)
             for ext in ['png', 'pdf', 'svg']:
```

```
plt.savefig(fn + '.' + ext)
plt.show()
```

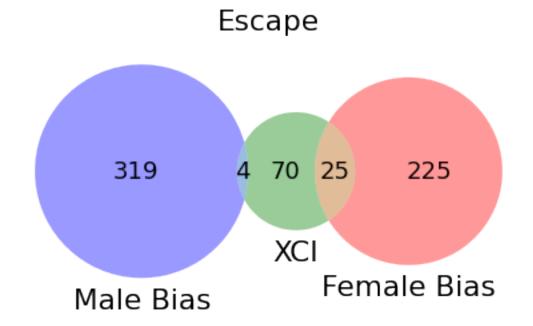
Caudate

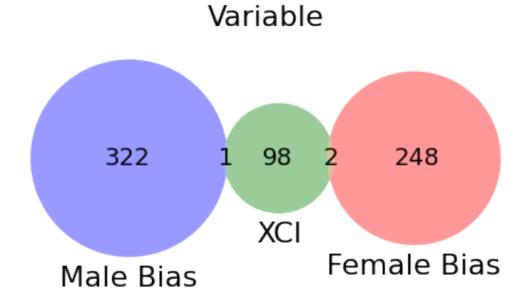


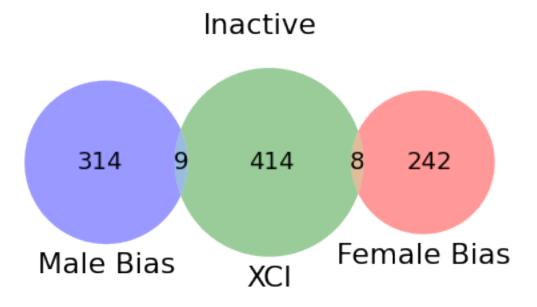




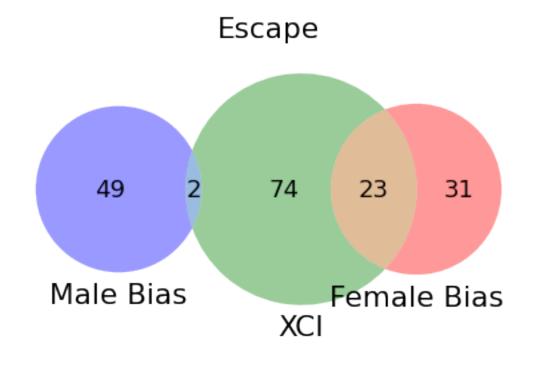
DLPFC

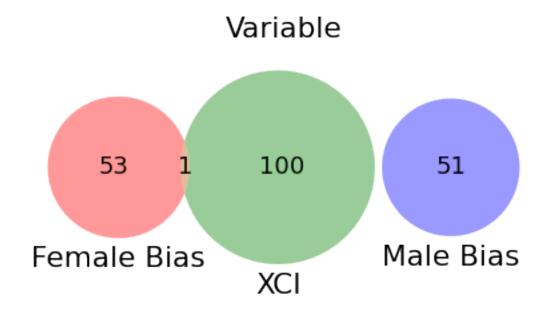




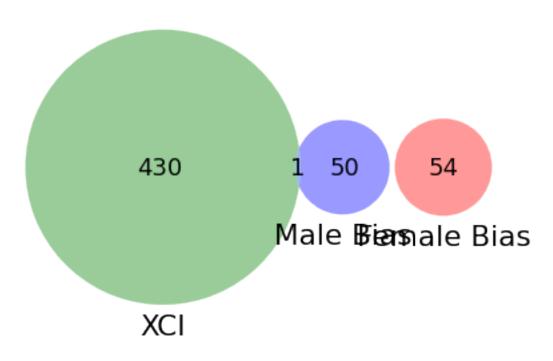


Hippocampus





Inactive



1.3 Calculated fisher's exact for each XCI status

```
[5]: d = dict()
for tissue in ["caudate", "dlpfc", "hippocampus"]:
    print(tissue_annotation(tissue))
    df = cal_fisher_by_xci_status(tissue)
    _, fdr, _, _ = multipletests(df.PValue, method='bonferroni')
    df['Bonferroni'] = fdr
    d[tissue] = df
```

Caudate

```
There is a significant enrichment (p-value < 3.5e-37) of escape!

There is a significant enrichment (p-value < 1.0e-03) of variable!

There is a significant enrichment of female bias genes (p-value < 1.2e-42) of escape!

There is a significant enrichment of female bias genes (p-value < 8.6e-05) of variable!

DLPFC

There is a significant enrichment (p-value < 2.3e-28) of escape!

There is a significant enrichment (p-value < 2.0e-02) of inactive!

There is a significant enrichment of male bias genes (p-value < 3.8e-03) of escape!

There is a significant enrichment of female bias genes (p-value < 6.5e-30) of
```

escape!

Hippocampus

There is a significant enrichment (p-value < 2.8e-41) of escape!

There is a significant enrichment of male bias genes (p-value < 2.4e-03) of escape!

There is a significant enrichment of female bias genes (p-value < 2.9e-41) of escape!

```
[6]: dft = pd.concat(d.values(), axis=0)
dft[(dft.Bonferroni <= 0.05)]</pre>
```

[6]:	Tissue	XCI status	OR	PValue	Direction	Bonferroni
0	Caudate	escape	52.541253	3.525156e-37	All	3.172640e-36
1	Caudate	variable	5.687289	1.018592e-03	All	9.167329e-03
6	Caudate	escape	136.346071	1.182377e-42	Female Bias	1.064139e-41
7	Caudate	variable	13.068562	8.613979e-05	Female Bias	7.752581e-04
0	DLPFC	escape	29.157326	2.319487e-28	All	2.087538e-27
3	DLPFC	escape	7.182445	3.797823e-03	Male Bias	3.418041e-02
6	DLPFC	escape	58.037037	6.541881e-30	Female Bias	5.887693e-29
0	Hippocampus	escape	148.678385	2.768297e-41	All	2.491467e-40
3	Hippocampus	escape	32.317784	2.362670e-03	Male Bias	2.126403e-02
6	Hippocampus	escape	256.447818	2.873171e-41	Female Bias	2.585853e-40

[7]: dft.to_csv('xci_enrichment_analysis_3brainRegions.txt', sep='\t', index=False)

[]: