main

March 7, 2022

1 Comparison with public TWAS associations for SZ

```
[1]: import urllib
import numpy as np
import pandas as pd
from html.parser import HTMLParser
from scipy.stats import fisher_exact
from html.entities import name2codepoint
[2]: class MyHTMLParser(HTMLParser):
    def __init__(self):
        super().__init__()
```

```
[2]: class MyHTMLParser(HTMLParser):
    def __init__(self):
        super().__init__()
        self.genes = []

    def handle_starttag(self, tag, attrs):
        for attr in attrs:
        if attr[0] == 'href' and attr[1].startswith('/genes/'):
            gene = [attr[1][len('/genes/'):]]
        self.genes += gene
```

1.1 MHC genes

[3]: 105

1.2 Caudate specific of BrainSeq brain regions

```
[4]: fn = "../../libd_twas_comparison/_m/caudate_only_twasList_genes.txt"
caudate = pd.read_csv(fn, sep='\t')
caudate.shape
```

```
[4]: (309, 47)
```

```
[5]: len(set(caudate.ID) - set(mhc_genes))
 [5]: 241
 [6]: caudate_noMHC = caudate[(caudate["ID"].isin(list(set(caudate.ID) -__
      ⇒set(mhc_genes))))].copy()
      caudate_noMHC.iloc[0:2, 0:5]
 [6]:
                    FILE
                                ID
                                    CHR_TWAS
                                                    P0
                                                                P1
      0 ENSG00000242866
                              STRC
                                          15 43599398 43618800.0
      1 ENSG00000183379
                         SYNDIG1L
                                          14 74405893 74426102.0
     1.3 TWAS Hub comparison
     1.3.1 Schizophrenia 2014
 [7]: parser = MyHTMLParser()
      html_str = urllib.request.urlopen('http://twas-hub.org/traits/Schizophrenia/').
      →read().decode()
      parser.feed(html_str)
      scz_2014 = np.unique(parser.genes)
 [8]: scz_2014.shape
 [8]: (49,)
     1.3.2 Schizophrenia 2018
 [9]: parser = MyHTMLParser()
      html_str = urllib.request.urlopen('http://twas-hub.org/traits/SCZ_2018/').
      →read().decode()
      parser.feed(html str)
      scz_2018 = np.unique(parser.genes)
     1.3.3 TWAS hub comparison
[10]: twas_hub = np.unique(np.append(scz_2014, scz_2018))
      len(twas_hub)
[10]: 67
[11]: twas_hub
[11]: array(['AC011816.1', 'AC103965.1', 'ACTR5', 'AKT3', 'ALMS1P', 'ANKRD44',
             'AS3MT', 'ATG13', 'BAI1', 'C12orf65', 'C17orf39', 'C2orf47',
             'CACNA2D4', 'CEP170', 'CHRNA5', 'CLCN3', 'CNTN4', 'CPEB1', 'CPNE7',
             'CRELD2', 'CUL3', 'CYSTM1', 'DCP1B', 'ELAC2', 'ERCC8', 'FAM53C',
             'FAM83H', 'FES', 'FLJ10661', 'GIGYF1', 'GMIP', 'HSPD1', 'IK',
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```
'IMMP2L', 'ITIH4-AS1', 'KIAA0319', 'KLC1', 'LGSN', 'LRRC48',
             'MAD1L1', 'MAP7D1', 'MAPK3', 'MGAT3', 'NAGA', 'NGEF', 'PCCB',
             'PCNX', 'PITPNM2', 'PLEKHO1', 'PPP1R13B', 'PPP1R14B', 'RERE',
             'RP11-981G7.6', 'SDCCAG8', 'SF3B1', 'SLCO4C1', 'SMG6', 'SNAP91',
             'SUGP2', 'THOC7', 'TRIM38', 'TTC14', 'U91328.21', 'VPS29',
             'VPS37A', 'XPNPEP3', 'ZNF318'], dtype='<U12')
[12]: print("There are %d caudate only genes present in the TWAS hub" %
            len(set(caudate.ID) & set(twas_hub)))
      set(caudate.ID) & set(twas_hub)
     There are 6 caudate only genes present in the TWAS hub
[12]: {'ELAC2', 'MAD1L1', 'PPP1R13B', 'TRIM38', 'VPS29', 'VPS37A'}
[13]: print("There are %d caudate only genes (no MHC) present in the TWAS hub" %
            len(set(caudate_noMHC.ID) & set(twas_hub)))
      set(caudate.ID) & set(twas_hub)
     There are 6 caudate only genes (no MHC) present in the TWAS hub
[13]: {'ELAC2', 'MAD1L1', 'PPP1R13B', 'TRIM38', 'VPS29', 'VPS37A'}
     1.4 Gandal comparison
[14]: gandal = pd.read excel("../h/aat8127_Table_S4.xlsx", sheet_name="SCZ.TWAS")
      gandal_twas = gandal[(gandal['TWAS.Bonferroni'] <= 0.05)].copy()</pre>
      np.sum(gandal.loc[:, 'TWAS.Bonferroni'] < 0.05)</pre>
[14]: 193
[15]: caudate_bonferroni = caudate[(caudate['Bonferroni'] <= 0.05)].copy()</pre>
      np.sum(caudate.Bonferroni <= 0.05)</pre>
[15]: 85
[16]: caudate noMHC bonferroni = caudate noMHC[(caudate noMHC['Bonferroni'] <= 0.05)].
       →copy()
      np.sum(caudate noMHC.Bonferroni <= 0.05)
[16]: 34
[17]: print("There are %d caudate only genes present in the Gandal at Bonferroni < 0.

→05." %

            len(set(caudate_bonferroni.FILE) & set(gandal_twas.GeneID)))
      #list(set(caudate bonferroni.ID) & set(qandal twas.gene name))
      print("There are %d caudate only genes present in the Gandal." %
            len(set(caudate.FILE) & set(gandal_twas.GeneID)))
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np.array(list(set(caudate.FILE) & set(gandal_twas.GeneID)))
     There are 21 caudate only genes present in the Gandal at Bonferroni < 0.05.
     There are 29 caudate only genes present in the Gandal.
[17]: array(['ENSG00000175662', 'ENSG00000219891', 'ENSG00000186522',
             'ENSG00000197279', 'ENSG00000205702', 'ENSG00000100162',
             'ENSG00000088808', 'ENSG0000006744', 'ENSG00000186470',
             'ENSG00000189298', 'ENSG00000204264', 'ENSG00000161896',
             'ENSG00000163938', 'ENSG00000231389', 'ENSG00000187987',
             'ENSG00000259404', 'ENSG00000158406', 'ENSG00000226314',
             'ENSG00000124613', 'ENSG00000219392', 'ENSG00000262074',
             'ENSG00000111237', 'ENSG00000272009', 'ENSG00000110492',
             'ENSG00000198315', 'ENSG00000231925', 'ENSG00000204963',
             'ENSG00000216901', 'ENSG00000168405'], dtype='<U15')
[18]: print("There are %d caudate only genes (no MHC) present in the Gandal at ⊔
      →Bonferroni < 0.05." %
            len(set(caudate noMHC_bonferroni.FILE) & set(gandal_twas.GeneID)))
      #list(set(caudate_bonferroni.ID) & set(qandal_twas.qene_name))
      print("There are %d caudate only genes (no MHC) present in the Gandal." %
            len(set(caudate_noMHC.FILE) & set(gandal_twas.GeneID)))
      np.array(list(set(caudate_noMHC.FILE) & set(gandal_twas.GeneID)))
     There are 9 caudate only genes (no MHC) present in the Gandal at Bonferroni <
     0.05.
     There are 13 caudate only genes (no MHC) present in the Gandal.
[18]: array(['ENSG00000175662', 'ENSG00000262074', 'ENSG0000006744',
             'ENSG00000111237', 'ENSG00000163938', 'ENSG00000186522',
             'ENSG00000110492', 'ENSG00000259404', 'ENSG00000205702',
             'ENSG00000100162', 'ENSG00000088808', 'ENSG00000204963',
             'ENSG00000168405'], dtype='<U15')
     MHC is present within Gandal analysis
[19]: gandal[(gandal["TWAS.Bonferroni"] < 0.05)].shape
[19]: (193, 31)
[20]: ## Not MHC
      gandal[~(gandal['gene_name'].isin(list(set(mhc_genes)))) &
             (gandal["TWAS.Bonferroni"] < 0.05)].shape
[20]: (177, 31)
[21]: print("There are {} MHC genes within Gandal significant TWAS (Bonferroni < 0.

→05)!"\
```

```
.format(len(gandal[(gandal['gene_name'].isin(list(set(mhc_genes)))) &
             (gandal["TWAS.Bonferroni"] < 0.05)].gene_name)))</pre>
      gandal[(gandal['gene_name'].isin(list(set(mhc_genes)))) &
             (gandal["TWAS.Bonferroni"] < 0.05)].gene_name
     There are 16 MHC genes within Gandal significant TWAS (Bonferroni < 0.05)!
[21]: 2
              HIST1H4J
      3
              HIST1H3C
      5
             ZSCAN12P1
      10
               ZSCAN23
      15
                BTN3A2
      22
               ZKSCAN3
      29
               ZKSCAN8
      38
                ZNF165
      39
              HIST1H4H
      59
              ZNF192P1
      67
                 PSMB8
      70
                 IP6K3
      71
                BTN2A2
                ZNF391
      124
      146
              HLA-DPA1
      166
                 TAPBP
      Name: gene_name, dtype: object
     1.4.1 Calculated enrichment with Gandal
[22]: caudate.shape
[22]: (309, 47)
[23]: dft = caudate.loc[:, ['FILE', 'ID', 'Bonferroni']]\
                    .merge(gandal.loc[:, ['GeneID', 'gene_name', 'TWAS.Bonferroni']],
                           left_on='FILE', right_on='GeneID',
                           suffixes=[" Benjamin", " Gandal"])
      dft.shape
[23]: (226, 6)
               [[np.sum((dft['Bonferroni']<0.05) & ((dft['TWAS.Bonferroni']<.05))),
[24]: table =
                 np.sum((dft['Bonferroni']<0.05) & ((dft['TWAS.Bonferroni']>=.05)))],
                [np.sum((dft['Bonferroni']>=0.05) & ((dft['TWAS.Bonferroni']<.05))),</pre>
                 np.sum((dft['Bonferroni']>=0.05) & ((dft['TWAS.Bonferroni']>=.05)))]]
      print(table)
      fisher_exact(table)
     [[21, 30], [8, 167]]
[24]: (14.6125, 7.051905747948124e-10)
```

1.4.2 Extract and save overlapping genes

```
Bonferroni < 0.05
[25]: overlapping_twas = np.append(np.
      →array(caudate_bonferroni[(caudate_bonferroni['FILE'].
       ⇒isin(list(set(caudate_bonferroni.FILE) &
      ⇒set(gandal twas.GeneID))))].ID),
                                  np.array(list(set(caudate_bonferroni.ID) &∟

set(twas_hub))))
     len(overlapping_twas)
[25]: 23
[26]: caudate_bonferroni[~(caudate_bonferroni['ID'].isin(overlapping_twas))].
      →to_csv("caudate_only_twasList_genes_bonferroni.txt",
                                                            sep='\t', index=False)
     caudate_bonferroni[~(caudate_bonferroni['ID'].isin(overlapping_twas))].shape
[26]: (64, 47)
[27]: drop_caudate = caudate_bonferroni[~(caudate_bonferroni['ID'].
      →isin(overlapping_twas))].copy()
     drop_caudate[(drop_caudate['P'] > 5e-8)].shape
[27]: (5, 47)
[28]: drop_caudate[(drop_caudate['P'] > 5e-8)].sort_values('FDR')\
                                             .loc[:, ['ID', 'our_snp_id', _
      [28]:
                            our_snp_id CHR_TWAS
               ID
                                                      FDR
     270
            CROCC
                     chr1:16497972:T:C
                                              1 0.000042 1.200000e-04
     85
          ANKRD45
                    chr1:173743105:T:C
                                              1 0.000054 5.050000e-07
     0
             STRC
                    chr15:43782086:A:G
                                             15 0.000183
                                                           2.750000e-06
     70
           ADRA2A chr10:111513602:C:T
                                             10 0.000244 1.650000e-05
     56
           CCDC92 chr12:123431550:A:G
                                             12 0.000334 7.100000e-07
                     FILE
     270 ENSG00000058453
     85
          ENSG00000183831
     0
          ENSG00000242866
     70
          ENSG00000150594
     56
          ENSG00000119242
     FDR < 0.05
[29]: overlapping_twas = np.append(np.array(caudate[(caudate['FILE'].
      →isin(list(set(caudate.FILE) &
```

```
⇒set(gandal_twas.GeneID))))].ID),
                                  np.array(list(set(caudate.ID) & set(twas_hub))))
     len(overlapping twas)
[29]: 35
[30]: caudate[~(caudate['ID'].isin(overlapping twas))].
      →to_csv("caudate_only_twasList_genes.txt",
                                                             sep='\t', index=False)
     caudate[~(caudate['ID'].isin(overlapping_twas))].shape
[30]: (277, 47)
[31]: drop_caudate = caudate[~(caudate['ID'].isin(overlapping_twas))].copy()
     drop_caudate[(drop_caudate['P'] > 5e-8)].shape
[31]: (173, 47)
[32]: drop_caudate[(drop_caudate['P'] > 5e-8)].sort_values('FDR')\
                                             .loc[:, ['ID', 'our_snp_id', _
       [32]:
                                    our_snp_id CHR_TWAS
                       ID
                                                               FDR
                                                                               Ρ
                                                                                  \
     270
                    CROCC
                             chr1:16497972:T:C
                                                          0.000042 1.200000e-04
                                                       1
     85
                  ANKRD45
                            chr1:173743105:T:C
                                                       1
                                                          0.000054
                                                                    5.050000e-07
     0
                     STRC
                            chr15:43782086:A:G
                                                      15 0.000183
                                                                    2.750000e-06
     70
                                                          0.000244 1.650000e-05
                   ADRA2A
                           chr10:111513602:C:T
                                                      10
     56
                   CCDC92
                           chr12:123431550:A:G
                                                         0.000334
                                                                    7.100000e-07
     183
                                                         0.000449 5.770000e-08
                   TBC1D5
                             chr3:16834975:C:T
                                                       3
     71
                    STIM2
                             chr4:26781831:A:G
                                                       4
                                                          0.000558 4.210000e-07
     205
                                                      19 0.000645 1.100000e-06
                     NTN5
                            chr19:48746940:T:C
     182
                 SLC39A10
                            chr2:195559315:A:C
                                                       2 0.000664 3.000000e-05
     21
          ENSG00000263715
                            chr17:46055092:G:A
                                                      17 0.000707 1.140000e-05
     212
                            chr12:50258497:G:A
                                                      12 0.000725 2.390000e-05
                    ASIC1
     215
                   UBQLN4
                            chr1:155906822:G:A
                                                       1 0.000816 9.600000e-06
     232
                    PPM1E
                            chr17:58681018:A:G
                                                      17
                                                         0.001272 2.390000e-06
     222
                     SNF8
                            chr17:48963251:G:A
                                                      17 0.001370 5.910000e-06
     40
                   PCDHB7
                            chr5:140841554:G:A
                                                       5 0.001405 8.660000e-07
     18
                    TRPS1
                            chr8:115455975:A:G
                                                       8
                                                         0.002413 4.280000e-07
     92
                     SSR2
                                                       1 0.002756 9.600000e-06
                            chr1:155906822:G:A
     25
                  SNORD13
                             chr8:33920637:G:A
                                                         0.002756 1.330000e-05
                                                       8
                                                         0.002925 1.200000e-03
     68
                   ANAPC1
                            chr2:111884332:C:T
     134
               LRRC37A17P
                            chr17:46784796:G:A
                                                         0.003155 3.050000e-05
                                                      17
     86
                                                      17 0.003230 4.400000e-06
                      CPD
                            chr17:30556723:C:T
                                                       7 0.003444 1.740000e-04
     12
                     EGFR
                             chr7:54797031:G:A
     193
                  TUBGCP4
                            chr15:43782086:A:G
                                                      15 0.004538
                                                                    2.750000e-06
```

```
147
                  PABPC1L
                           chr20:45053910:T:G
                                                     20 0.004817 3.320000e-06
     63
                                                     17 0.004940 1.140000e-05
                  RPS26P8
                            chr17:46055092:G:A
                     FILE
     270 ENSG00000058453
     85
          ENSG00000183831
     0
          ENSG00000242866
     70
          ENSG00000150594
     56
          ENSG00000119242
     183 ENSG00000131374
     71
          ENSG00000109689
     205 ENSG00000142233
     182 ENSG00000196950
     21
          ENSG00000263715
     212 ENSG00000110881
     215 ENSG00000160803
     232 ENSG00000175175
     222 ENSG00000159210
     40
          ENSG00000113212
          ENSG00000104447
          ENSG00000163479
     25
          ENSG00000239039
     68
          ENSG00000153107
     134 ENSG00000263142
     86
          ENSG00000108582
          ENSG00000146648
     193 ENSG00000137822
     147 ENSG00000101104
          ENSG00000204652
[33]: | drop_caudate[(drop_caudate["ID"] == "MIAT")].loc[:, ['ID', 'our_snp_id', __
      [33]: Empty DataFrame
     Columns: [ID, our_snp_id, CHR_TWAS, FDR, P, FILE]
     Index: []
     1.5 TWAS tissue summary
[34]: brainseq = pd.read_csv("../../libd_twas_comparison/_m/TWAS_gene_tissue_summary.
      ⇔csv")
     brainseq.shape
[34]: (8348, 11)
[35]: brainseq.head(2)
```

```
[35]:
                             Symbol Caudate_TWAS.Z
                                                      Caudate_FDR Caudate_GWAS.SNP \
                  Geneid
     0 ENSG00000219891 ZSCAN12P1
                                           12.375993 1.871099e-31
                                                                           Risk SNP
      1 ENSG00000204338
                           CYP21A1P
                                           12.163700 1.287468e-30
                                                                           Risk SNP
         DLPFC_TWAS.Z DLPFC_FDR DLPFC_GWAS.SNP HIPPO_TWAS.Z HIPPO_FDR \
      0
                  {\tt NaN}
                             {\tt NaN}
                                             {\tt NaN}
                                                           {\tt NaN}
                                                                      {\tt NaN}
                  NaN
                             NaN
                                                                      NaN
      1
                                             NaN
                                                           NaN
       HIPPO_GWAS.SNP
                   {\tt NaN}
      0
      1
                   {\tt NaN}
[36]: bb = brainseq.merge(pd.DataFrame({'Symbol': twas_hub, 'inTWAS_HUB': 1}), ___
       .merge(pd.DataFrame({'Geneid': gandal_twas.GeneID, 'inGandal': 1}),
                          on="Geneid", how='left')
      bb.to_csv('TWAS_gene_tissue_summary.csv', index=False)
```