main

September 1, 2021

1 Plotting eQTLs, increase font sizes

1.0.1 Kynon Jade Benjamin and Apuã Paquola

```
[1]: import re
  import functools
  import subprocess
  import numpy as np
  import pandas as pd
  from plotnine import *
  from pandas_plink import read_plink
  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.*')
```

1.1 Configuration

1.2 Functions

1.2.1 Expression functions

```
[3]: Ofunctools.lru cache()
    def tissue_map(tissue):
        return {"caudate": "Caudate", "dlpfc": "DLPFC",
                 "hippocampus": "Hippocampus"}[tissue]
    @functools.lru_cache()
    def feature_map(feature):
        return {"genes": "Gene", "transcripts": "Transcript",
                 "exons": "Exon", "junctions": "Junction"}[feature]
    @functools.lru_cache()
    def get biomart df():
        biomart = pd.read_csv(config['biomart_file'], index_col=0)
        biomart['description'] = biomart['description'].str.replace('\[Source.])
     →*$','', regex=True)
        return biomart
    @functools.lru_cache()
    def get_residual_expression_df():
        return pd.read_csv(config['residual_expression_file'], index_col=0).
     →transpose()
    @functools.lru_cache()
    def get_pheno_df():
        return pd.read_csv(config['phenotype_file']).set_index("BrNum").loc[:,__
     @functools.lru_cache()
    def get_expression_and_pheno_df():
        return pd.merge(get_pheno_df(), get_residual_expression_df(),
                        left_index=True, right_index=True)
    @functools.lru_cache()
    def get_gene_id_df():
        return pd.DataFrame({'gene_id': get_residual_expression_df().columns,
                              'ensembl_gene_id': get_residual_expression_df().
     →columns.str.replace('\..+$','', regex=True)})
```

1.2.2 Genotype and eQTL functions

```
[4]: def letter_snp(number, a0, a1):
         Example:
         letter_snp(0, 'A', 'G') is 'AA'
         letter_snp(1, 'A', 'G') is 'AG'
         letter_snp(2, 'A', 'G') is 'GG'
         if np.isnan(number):
             return np.nan
         if len(a0) == 1 and len(a1) == 1:
             sep = ''
         else:
             sep = ' '
         return sep.join(sorted([a0]*int(number) + [a1]*(2-int(number))))
     @functools.lru_cache()
     def get_plink_tuple():
         Usage: (bim, fam, bed) = get_plink_tuple()
         return read_plink(config['plink_file_prefix'])
     @functools.lru_cache()
     def get_eFeature_df():
         eqtl_df = pd.read_csv(config["eqtl_output_file"], sep='\t')
         return eqtl_df[(eqtl_df["Type"] == feature_map(feature)) &
                        (eqtl df["Tissue"] == tissue map(tissue))]
```

```
@functools.lru_cache()
def get_gwas_snps():
    return pd.read_csv(config['gwas_snp_file'], sep='\t', index_col=0)
@functools.lru_cache()
def get_risk_allele(snp_id):
    gwas_snp = get_gwas_snp(snp_id)
    if gwas snp['OR'].iloc[0] > 1:
        ra = gwas_snp['A1'].iloc[0]
    else:
        ra = gwas_snp['A2'].iloc[0]
    return ra
@functools.lru_cache()
def get_snp_df(snp_id):
    Returns a dataframe containing the genotype on snp snp_id.
    The allele count is the same as in the plink files.
    Example:
    get_snp_df('rs653953').head(5)
            rs653953\_num\ rs653953\_letter\ rs653953
    Br5168
                       0
                                      GG
                                             O \setminus nGG
                                       AG
                                             1 \setminus nAG
    Br2582
                       1
    Br2378
                       1
                                       AG \qquad 1 \setminus nAG
    Br5155
                       2
                                       AA
                                             2 \backslash nAA
    Br5182
                        2
                                             2 \ nAA
                                       AA
    (bim, fam, bed) = get_plink_tuple()
    brain_ids = list(set(get_expression_and_pheno_df().index).
→intersection(set(fam['fid'])))
    snp info = bim[bim['snp']==snp id]
    snp_pos = snp_info.iloc[0]['i']
    fam_pos = list(fam.drop_duplicates(subset="fid").set_index('fid').
 →loc[brain_ids]['i'])
    dfsnp = (pd.DataFrame(bed[[snp_pos]].compute()[:,fam_pos],
                           columns=brain_ids, index=[snp_id + '_num'])
             .transpose().dropna())
    my_letter_snp = functools.partial(letter_snp, a0=snp_info.iloc[0]['a0'],_
\rightarrowa1=snp_info.iloc[0]['a1'])
    dfsnp[[snp id + ' num']] = 2 - dfsnp[[snp id + ' num']].astype('int')
    dfsnp[snp_id + '_letter'] = dfsnp[snp_id + '_num'].apply(my_letter_snp)
    dfsnp[snp_id] = (dfsnp[snp_id + '_num'].astype('str') + '\n' +
                     dfsnp[snp_id + '_letter'].astype('str')).astype('category')
```

```
return dfsnp
@functools.lru_cache()
def get_gwas_ordered_snp_df(snp_id):
    Returns a dataframe containing the genotype on snp snp_id.
    The allele count is the number of risk alleles according to GWAS.
    Example:
    get_gwas_ordered_snp_df('rs653953').head(5)
            rs653953 num rs653953 letter rs653953
    Br5168
                        2
                                        GG
                                               2 \backslash nGG
    Br2582
                                              1 \setminus nAG
                        1
                                        AG
    Br2378
                        1
                                        AG
                                              1 \setminus nAG
                        0
                                        AA \quad O \setminus nAA
    Br5155
    Br5182
                                        AA
                                              0 \backslash nAA
    pgc = get_gwas_snps()
    dfsnp = get_snp_df(snp_id).copy()
    gwas_snp = get_gwas_snp(snp_id)
    if gwas_snp['pgc2_a1_same_as_our_counted'].iloc[0]:
        if gwas snp['OR'].iloc[0] > 1:
            pass
        else:
            dfsnp[[snp_id + '_num']] = 2 - dfsnp[[snp_id + '_num']]
    else:
        if gwas_snp['OR'].iloc[0] > 1:
            dfsnp[[snp_id + '_num']] = 2 - dfsnp[[snp_id + '_num']]
        else:
    dfsnp[snp_id] = (dfsnp[snp_id + '_num'].astype('str') + '\n' +
                      dfsnp[snp_id + '_letter'].astype('str')).astype('category')
    return dfsnp
```

1.2.3 Plotting functions

```
def get_snp_gene_pheno_df(snp_id, gene_id, snp_df_func):
    pheno_columns = list(get_pheno_df().columns)
    expr_df = get_expression_and_pheno_df()[pheno_columns + [gene_id]]
    snp_df = snp_df_func(snp_id)
    return expr_df.merge(snp_df, left_index=True, right_index=True)

def simple_snp_expression_plot_impl(snp_id, gene_id, snp_df_func):
    df = get_snp_gene_pheno_df(snp_id, gene_id, snp_df_func)
```

```
y0 = df[gene_id].quantile(.01) - 0.26
   y1 = df[gene_id].quantile(.99) + 0.26
   p = ggplot(df, aes(x=snp_id, y=gene_id, fill='Sex')) \
   + geom_boxplot(alpha=0.4, outlier_alpha=0) \
   + geom_jitter(position=position_jitterdodge(jitter_width=0.25),
                  stroke=0, alpha=0.6) \
   + ylim(y0, y1) \
   + theme_bw(base_size=15) \
   + theme(panel_grid=element_blank(),
            axis_title=element_text(face="bold"))
   return p
def simple_snp_expression_plot(snp_id, gene_id):
   return simple_snp_expression_plot_impl(snp_id, gene_id, get_snp_df)
def simple_gwas_ordered_snp_expression_plot(snp_id, gene_id):
   return simple_snp_expression_plot_impl(snp_id, gene_id,__
→get_gwas_ordered_snp_df)
def get_gene_symbol(gene_id, biomart=get_biomart_df()):
   ensge = re.sub('\..+$','', gene_id)
   ggg = biomart[biomart['ensembl_gene_id']==ensge]
    if ggg.shape[0]==0:
       return '', ''
   gs = ggg['external_gene_name'].values[0]
   de = ggg['description'].values[0]
   if type(de)!=str:
       de = ''
   de = re.sub('\[Source:.*$','',de)
   return gs, de
def get_gwas_snp(snp_id):
   gwas = get_gwas_snps()
   r = gwas[gwas['our_snp_id']==snp_id]
   assert len(r) == 1
   return r
def gwas_annotation(snp_id):
   return 'SZ GWAS pvalue: %.1e' % get_gwas_snp(snp_id).iloc[0]['P']
def eqtl_annotation(snp_id, gene_id):
```

```
eqtl_df = get_eFeature_df()
   r = eqtl_df[(eqtl_df['variant_id']==snp_id) & (eqtl_df['gene_id']==gene_id)]
   assert len(r)==1
   return 'eQTL adjusted p-value: %.1e' % r.iloc[0]['BF']
def risk_allele_annotation(snp_id):
   return 'SZ risk allele: %s' % get_risk_allele(snp_id)
def annotated eqtl plot(snp id, gene id):
   p = simple_snp_expression_plot(snp_id, gene_id)
   gene_symbol, gene_description = get_gene_symbol(gene_id)
   title ="\n".join([gene_symbol,
                     eqtl_annotation(snp_id, gene_id)
   p += ggtitle(title) + ylab('Residualized Expression')
   return p
def gwas_annotated_eqtl_plot(snp_id, gene_id):
   p = simple_gwas_ordered_snp_expression_plot(snp_id, gene_id)
   gene_symbol, gene_description = get_gene_symbol(gene_id)
   title ="\n".join([gene_symbol,
                     eqtl_annotation(snp_id, gene_id),
                     gwas annotation(snp id),
                     risk_allele_annotation(snp_id)
                     1)
   p += ggtitle(title) + ylab('Residualized Expression')
   return p
def save_plot(p, fn):
   for ext in ['png', 'pdf', 'svg']:
       p.save(fn + '.' + ext)
```

1.3 Plot eQTLs

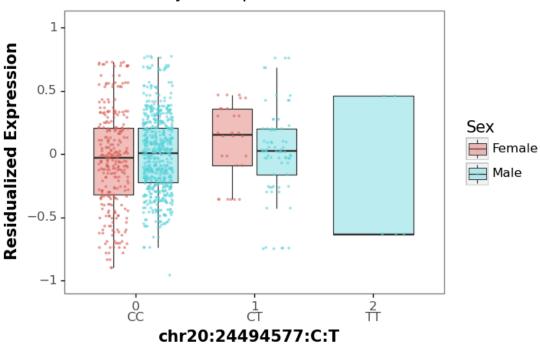
1.3.1 DRD2

```
[6]: get_eFeature_df()[(get_eFeature_df()["gene_id"] == gene_id_from_symbol('DRD2'))]
[6]: Empty DataFrame
    Columns: [variant_id, gene_id, gencodeID, slope, statistic, pval_nominal, BF, eigenMT_BH, TESTS, Type, Tissue]
    Index: []
```

1.3.2 Top 5 eQTLs

```
[7]: eqtl_df = get_eFeature_df()
    eqtl_df.head()
[7]:
                           variant_id
                                                  gene_id
                                                                    gencodeID \
    7125
                    chr7:42936690:A:C
                                       ENSG00000002746.14
                                                           ENSG00000002746.14
    7126
                                       ENSG00000002919.14
                                                           ENSG00000002919.14
                   chr17:48075934:C:T
    7127
                     chr7:7934340:A:C
                                       ENSG0000003147.17
                                                           ENSG00000003147.17
    7128 chr16:89536451:CGGTGAGGCG:C
                                       ENSG00000003249.13
                                                           ENSG00000003249.13
    7129
                    chr8:17754947:C:T
                                       ENSG0000003989.16
                                                           ENSG0000003989.16
             slope statistic pval_nominal
                                                   BF eigenMT_BH
                                                                          Type \
                                                                   TESTS
    7125 0.436502
                    7.820971
                                                         0.445692
                                                                          Gene
                                   0.000023 0.010399
                                                                     445
    7126 -0.536903 -7.959488
                                   0.000127
                                             0.037424
                                                         0.549367
                                                                     294 Gene
    7127 0.213896 11.357107
                                   0.000022 0.022195
                                                         0.503766
                                                                    1002
                                                                          Gene
    7128 -0.524657 -9.921271
                                   0.000028 0.014577
                                                         0.470811
                                                                     529
                                                                          Gene
    7129 -0.497351 -11.806077
                                   0.000007 0.007416
                                                         0.417015
                                                                    1040 Gene
           Tissue
    7125 Caudate
    7126 Caudate
    7127 Caudate
    7128 Caudate
    7129 Caudate
[8]: top_5 = eqtl_df.sort_values('pval_nominal').reset_index(drop=True).head(5)
    for x in top_5.itertuples():
        filename = "top_%d_eqtl_%s" % (x.Index, tissue)
        p = annotated_eqtl_plot(x.variant_id, x.gene_id)
        print(filename, x.Index, x.variant_id, x.gene_id)
        print(p)
        save_plot(p, filename)
    Mapping files: 100%|
                             | 3/3 [00:26<00:00, 8.79s/it]
    top_0_eqtl_caudate 0 chr20:24494577:C:T ENSG00000101474.11
```

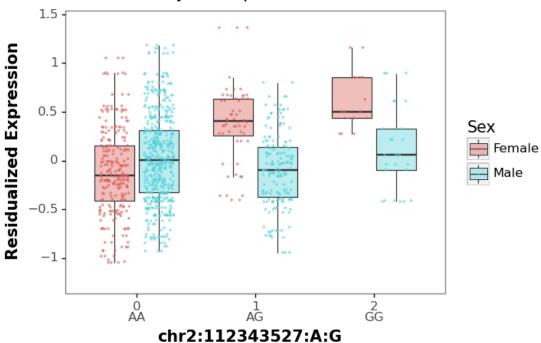
APMAP eQTL adjusted p-value: 4.2e-06



<ggplot: (8728063515606)>

top_1_eqtl_caudate 1 chr2:112343527:A:G ENSG00000153214.9

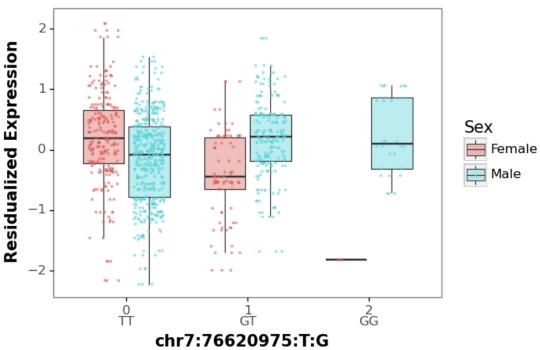
TMEM87B eQTL adjusted p-value: 1.1e-05



<ggplot: (8728063342174)>

top_2_eqtl_caudate 2 chr7:76620975:T:G ENSG00000205482.9

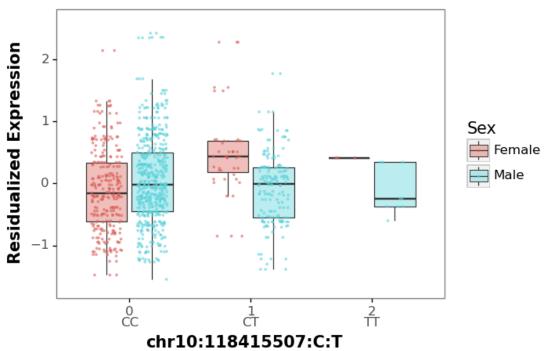
SPDYE18 eQTL adjusted p-value: 2.0e-05



<ggplot: (8728063254492)>

top_3_eqtl_caudate 3 chr10:118415507:C:T ENSG00000119973.5

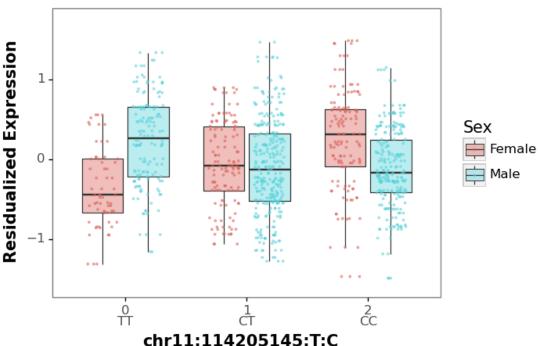
PRLHR eQTL adjusted p-value: 3.4e-05



<ggplot: (8728062620182)>

top_4_eqtl_caudate 4 chr11:114205145:T:C ENSG00000166736.11

HTR3A eQTL adjusted p-value: 3.7e-05



<ggplot: (8728062803366)>

1.3.3 Top 5 X-linked genes

```
[9]: top_5_x = eqtl_df[eqtl_df['variant_id'].str.contains("chrX")].

→sort_values("pval_nominal").reset_index(drop=True).head(5)

for x in top_5_x.itertuples():
    filename = "top_%d_eqtl_xlinked_%s" % (x.Index, tissue)
    p = annotated_eqtl_plot(x.variant_id, x.gene_id)
    print(filename, x.Index, x.variant_id, x.gene_id)
    print(p)
    save_plot(p, filename)
```

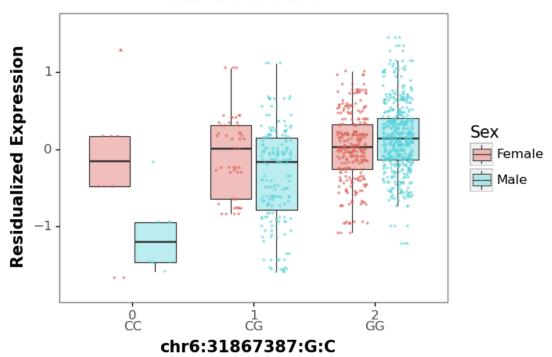
1.3.4 Top 5 eQTL with GWAS significant index SNP

```
[10]:
                                                          gencodeID
                                                                        slope \
                variant_id
                                        gene_id
      0 chr22:42119633:G:C
                                                  ENSG00000183172.8 0.215764
                              ENSG00000183172.8
      1
         chr6:32140074:G:A ENSG00000204371.11 ENSG00000204371.11 -0.509024
         chr6:31867387:G:C
                             ENSG00000244731.7
                                                  ENSG00000244731.7 -0.632753
        statistic pval nominal
                                           eigenMT BH
                                                        TESTS
                                       \mathsf{BF}
                                                              Type
                                                                    ... A2
                                                                               OR \
      0 10.770408 5.788120e-05
                                 0.024426
                                              0.516399
                                                          422
                                                               Gene
                                                                    ... C
                                                                          1.0629
      1 -8.596720 5.852160e-05
                                                          503
                                                               Gene
                                 0.029436
                                              0.536823
                                                                    ... A 1.1331
      2 -13.250431 1.564220e-07 0.000079
                                              0.123254
                                                          502 Gene ... C 1.1744
              SE
                                hg19chrc hg38chrc
                                                    hg38pos \
                                    chr22
      0 0.010193 2.130000e-09
                                             chr22 42119633
      1 0.014200 1.370000e-18
                                     chr6
                                              chr6 32140074
      2 0.014203 1.050000e-29
                                     chr6
                                              chr6 31867387
       pgc2_a1_same_as_our_counted
                                         rsid is_index_snp
                              False rs8143153
                                                       False
                              False rs3134962
                                                       False
      1
      2
                              False
                                     rs693906
                                                       False
      [3 rows x 33 columns]
[11]: top gwas_eqtl_df = gwas_eqtl_df [(gwas_eqtl_df ['is index_snp'])].

→sort_values(['BF', 'P'])
      print(top_gwas_eqtl_df.shape)
      top_gwas_eqtl_df.head()
     (0, 33)
[11]: Empty DataFrame
      Columns: [variant_id, gene_id, gencodeID, slope, statistic, pval_nominal, BF,
      eigenMT_BH, TESTS, Type, Tissue, chrN, our_snp_id, cm, pos, our_counted,
      our_alt, chrom, SNP, Freq.A1, CHR, BP, A1, A2, OR, SE, P, hg19chrc, hg38chrc,
     hg38pos, pgc2_a1_same_as_our_counted, rsid, is_index_snp]
      Index: []
      [0 rows x 33 columns]
[12]: top_gwas_eqtl_df = gwas_eqtl_df.sort_values(['BF', 'P']).reset_index(drop=True)
      print(top_gwas_eqtl_df.shape)
      top gwas eqtl df.head(10)
     (3, 33)
[12]:
                variant_id
                                        gene_id
                                                          gencodeID
                                                                        slope \
         chr6:31867387:G:C
                              ENSG00000244731.7
                                                  ENSG00000244731.7 -0.632753
      1 chr22:42119633:G:C
                              ENSG00000183172.8
                                                  ENSG00000183172.8 0.215764
         chr6:32140074:G:A ENSG00000204371.11 ENSG00000204371.11 -0.509024
```

```
statistic pval_nominal
                                          eigenMT_BH TESTS Type
                                       BF
                                                                   ... A2
                                                                             OR \
     0 -13.250431 1.564220e-07 0.000079
                                                        502 Gene
                                            0.123254
                                                                   ... C 1.1744
                                                             Gene
                                                                   ... C 1.0629
     1 10.770408 5.788120e-05
                                 0.024426
                                             0.516399
                                                        422
                                                        503 Gene ... A 1.1331
     2 -8.596720 5.852160e-05 0.029436
                                            0.536823
              SE
                             P hg19chrc hg38chrc
                                                   hg38pos \
     0 0.014203 1.050000e-29
                                    chr6
                                            chr6 31867387
                                   chr22
     1 0.010193 2.130000e-09
                                            chr22 42119633
     2 0.014200 1.370000e-18
                                    chr6
                                             chr6 32140074
       pgc2_a1_same_as_our_counted
                                         rsid is_index_snp
                             False
                                     rs693906
                                                     False
     1
                             False rs8143153
                                                     False
     2
                             False rs3134962
                                                     False
     [3 rows x 33 columns]
[13]: top_5_gwas = top_gwas_eqtl_df.head(5)
     for x in top_5_gwas.itertuples():
         filename = "top_%d_eqtl_in_gwas_significant_snps_%s" % (x.Index, tissue)
         p = gwas_annotated_eqtl_plot(x.variant_id, x.gene_id)
         print(filename, x.Index, x.variant_id, x.gene_id)
         print(p)
         save_plot(p, filename)
     top_0_eqtl_in_gwas_significant_snps_caudate 0 chr6:31867387:G:C
```

C4A eQTL adjusted p-value: 7.9e-05 SZ GWAS pvalue: 1.1e-29 SZ risk allele: G

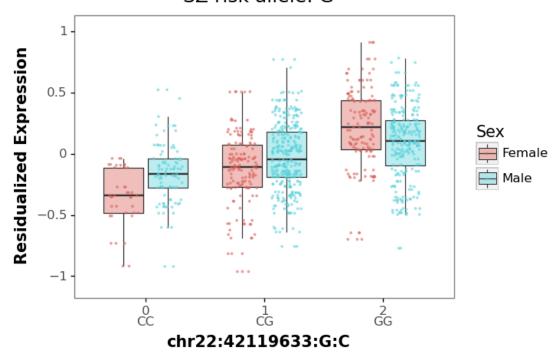


<ggplot: (8728062453709)>

 $\verb"top_1_eqtl_in_gwas_significant_snps_caudate 1 chr22:42119633:G:C"$

ENSG00000183172.8

SMDT1 eQTL adjusted p-value: 2.4e-02 SZ GWAS pvalue: 2.1e-09 SZ risk allele: G

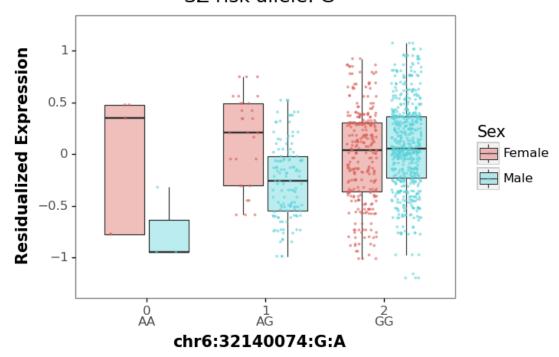


<ggplot: (8728184152691)>

 ${\tt top_2_eqtl_in_gwas_significant_snps_caudate~2~chr6:32140074:G:A}$

ENSG00000204371.11

EHMT2 eQTL adjusted p-value: 2.9e-02 SZ GWAS pvalue: 1.4e-18 SZ risk allele: G



<ggplot: (8728062942880)>

[]: