main

July 10, 2021

1 Calculate RXE for public dataset, females only based on Apua's script

```
[1]: import re
    import functools
    import numpy as np
    import pandas as pd
    from os import environ
    from plotnine import *
    from scipy.stats import variation
    from warnings import filterwarnings
    from scipy.stats import mannwhitneyu
    from matplotlib.cbook import mplDeprecation
[2]: filterwarnings("ignore", category=mplDeprecation)
    filterwarnings('ignore', category=UserWarning, module='plotnine.*')
    filterwarnings('ignore', category=DeprecationWarning, module='plotnine.*')
[3]: environ['NUMEXPR MAX THREADS'] = '10'
[4]: config = {
         'pheno_file': 'cmc_phenotypes.csv',
         'cpm_file': 'cmc_log2tpm.tsv',
         'tpm_file': '/ceph/users/jbenja13/projects/sex_sz_ria/input/commonMind/
     'annot_file': 'cmc_gene_annotation.tsv',
[5]: Ofunctools.lru_cache()
    def get_pheno():
        return pd.read_csv(config['pheno_file'], index_col=0)
    @functools.lru_cache()
    def get_annotation():
        return pd.read_csv(config['annot_file'], index_col=0, sep='\t')
```

```
[6]: def chrom_type(chrom):
          if chrom=='X':
              return 'X'
          elif re.search('^\d+', chrom):
              return 'autosome'
          else:
              return 'other'
      def save_ggplot(p, fn):
          for ext in ['.pdf', '.png', '.svg']:
              p.save(fn+ext, width=10, height=7)
 [7]: gene_annot = get_annotation()
      gene_annot.head(2)
 [7]:
                          ensembl_gene_id hgnc_symbol chromosome_name
     gene_id
     ENSG0000000003.14 ENSG0000000003
                                               TSPAN6
                                                                     Х
      ENSG0000000005.5
                          ENSG00000000005
                                                 TNMD
                                                                    Х
 [8]: gene_annot.shape
 [8]: (57765, 3)
 [9]: log2tpm = pd.read csv(config['tpm_file'], index_col=0, sep=',', engine='c')
      log2tpm.iloc[0:2, 0:5]
 [9]:
                          MSSM_RNA_BP_PFC_2 MSSM_RNA_BP_PFC_4 MSSM_RNA_BP_PFC_5 \
      gene_id
     ENSG0000000003.14
                                   1.927755
                                                      1.341998
                                                                          1.531728
      ENSG0000000005.5
                                   0.089456
                                                      0.036288
                                                                          0.030684
                          MSSM_RNA_BP_PFC_6 MSSM_RNA_BP_PFC_7
      gene_id
      ENSG0000000003.14
                                   1.565244
                                                      2.178456
      ENSG0000000005.5
                                   0.088621
                                                      0.235076
[10]: log2tpm.shape
[10]: (50689, 606)
[11]: df0 = gene_annot.merge(log2tpm, right_index=True, left_index=True).
       →drop_duplicates(subset=['ensembl_gene_id'])
      df0.shape
[11]: (50097, 609)
```

```
[12]: df0['chrom_type'] = df0['chromosome_name'].apply(chrom_type)
      df0.shape
[12]: (50097, 610)
[13]: df0.groupby('chrom_type').size()
[13]: chrom type
                   1970
      autosome
                  47618
      other
                   509
      dtype: int64
[14]: samples = np.array(log2tpm.columns[log2tpm.columns.str.contains("RNA|CMC", ____
      →regex=True)])
      df = df0[(df0[samples].sum(axis=1) >= 0.2 * len(samples)) & (df0['chrom_type'] !
      →= 'other')]
      df1 = df.groupby('chrom_type').mean().transpose()
      df1['sample'] = df1.index
      df1['RXE'] = df1['X'] - df1['autosome']
      df1.head(2)
                                                                    RXE
[14]: chrom_type
                               X autosome
                                                       sample
                                  2.124278 MSSM RNA BP PFC 2 0.076663
      MSSM_RNA_BP_PFC_2 2.200941
      MSSM_RNA_BP_PFC_4 1.666506 1.628365 MSSM_RNA_BP_PFC_4 0.038141
[15]: df2 = df1.merge(get_pheno(), left_index=True, right_index=True)
      df2.head(2)
[15]:
                            X autosome
                                                 sample
                                                              RXE Individual_ID \
      MSSM_RNA_PFC_1 2.594985 2.529523 MSSM_RNA_PFC_1 0.065462 CMC MSSM 025
      MSSM RNA_PFC_2 2.672613 2.594684 MSSM_RNA_PFC_2 0.077929 CMC_MSSM_136
                     Institution Reported_Gender
                                                      Dx Age_of_Death
      MSSM RNA PFC 1
                           MSSM
                                           Male
                                                     SCZ
                                                                  68.0
     MSSM_RNA_PFC_2
                           MSSM
                                           Male Control
                                                                  76.0
[16]: df2.loc[:, ['Reported Gender', 'Institution', 'RXE']].groupby(["Institution", u
       →'Reported_Gender']).describe()
Г16]:
                                    RXF.
                                  count
                                                        std
                                                                  min
                                                                            25%
                                             mean
      Institution Reported_Gender
     MSSM
                 Female
                                  126.0 0.065413 0.067790 -0.090430 0.012126
                 Male
                                  181.0 0.058109 0.061735 -0.122148 0.028566
      Penn
                 Female
                                   54.0 0.089080 0.050252 -0.061299 0.057125
                 Male
                                   40.0 0.087665 0.060816 -0.053713 0.051894
     Pitt
                 Female
                                   41.0 0.097466 0.031784 0.035957 0.074324
```

```
50%
                                                  75%
                                                            max
      Institution Reported_Gender
     MSSM
                 Female
                                   0.077704 0.109618 0.255605
                 Male
                                   0.071640 0.100297 0.192727
                 Female
     Penn
                                   0.088885 0.118559 0.238989
                 Male
                                   0.090865 0.115655 0.243992
     Pitt
                  Female
                                   0.096307 0.126792 0.157051
                 Male
                                   0.085425 0.115341 0.186669
[17]: for institution in df2.Institution.unique():
          for sex in ["Female", "Male"]:
              var = variation(df2[(df2['Institution']==institution) &
                                  (df2['Reported_Gender']==sex)].RXE)
              print("There is %.3f variation for %s in %s." % (var, sex, institution))
     There is 1.032 variation for Female in MSSM.
     There is 1.059 variation for Male in MSSM.
     There is 0.559 variation for Female in Penn.
     There is 0.685 variation for Male in Penn.
     There is 0.322 variation for Female in Pitt.
     There is 0.436 variation for Male in Pitt.
[18]: df2.loc[:, ['Reported_Gender', 'Institution', 'RXE']].

¬groupby(['Reported_Gender']).describe()
[18]:
                         RXE
                                                                                \
                                                                 25%
                                                                           50%
                       count
                                  mean
                                             std
                                                       min
      Reported_Gender
     Female
                       221.0 0.077142 0.059949 -0.090430 0.047601
                                                                      0.085898
      Male
                      327.0 0.071475 0.057008 -0.122148 0.044566
                            75%
                                      max
      Reported_Gender
      Female
                       0.113375 0.255605
      Male
                       0.106715 0.243992
[19]: for sex in ["Female", "Male"]:
          var = variation(df2[(df2['Reported_Gender']==sex)].RXE)
          print("There is %.3f variation for %s." % (var, sex)) # Not sure if this_
       →variation is real
     There is 0.775 variation for Female.
     There is 0.796 variation for Male.
```

1.0.1 Separate out tissues

```
[20]: for institution in df2.Institution.unique():
          dft = df2[(df2['Institution'] == institution)].copy()
          stat, pval = mannwhitneyu(dft[(dft['Reported_Gender']=="Female")].RXE,
                                    dft[(dft['Reported_Gender'] == "Male")].RXE)
         print("Mann-WhitneyU for female vs male (RXE) for %s: %.4f" % (institution,
       →pval))
     Mann-WhitneyU for female vs male (RXE) for MSSM: 0.1548
     Mann-WhitneyU for female vs male (RXE) for Penn: 0.4137
     Mann-WhitneyU for female vs male (RXE) for Pitt: 0.0719
[21]: stat, pval = mannwhitneyu(df2[(df2['Reported Gender']=="Female")].RXE,
                                df2[(df2['Reported_Gender']=="Male")].RXE)
      print("Mann-WhitneyU for female vs male (RXE) for %s: %.4f, %.1f" % ("DLPFC", )
       →pval, stat))
     Mann-WhitneyU for female vs male (RXE) for DLPFC: 0.0734, 33495.0
[22]: for institution in df2.Institution.unique():
         dft = df2[(df2['Institution'] == institution)].copy()
          stat, pval = mannwhitneyu(dft[(dft['Dx']=='Control')].RXE,
                                    dft[(dft['Dx']=='SCZ')].RXE)
         print("Mann-WhitneyU for ctl vs sz (RXE) for %s: \%.4e" \% (institution, \Box
       →pval))
     Mann-WhitneyU for ctl vs sz (RXE) for MSSM: 1.6429e-02
     Mann-WhitneyU for ctl vs sz (RXE) for Penn: 1.6475e-01
     Mann-WhitneyU for ctl vs sz (RXE) for Pitt: 1.7510e-01
[23]: stat, pval = mannwhitneyu(df2[(df2['Dx']=="Control")].RXE,
                                df2[(df2['Dx']=="SCZ")].RXE)
      print("Mann-WhitneyU for female vs male (RXE) for %s: %.4e, %.1f" % ("DLPFC", u
      →pval, stat))
     Mann-WhitneyU for female vs male (RXE) for DLPFC: 3.1883e-02, 33911.0
[24]: for institution in df2.Institution.unique():
         dft = df2[(df2['Institution'] == institution)].copy()
         female = dft[(dft['Reported_Gender']=='Female')].copy()
         male = dft[(dft['Reported_Gender']=='Male')].copy()
         stat_f, pval_f = mannwhitneyu(female[(female['Dx']=='Control')].RXE,
                                        female[(female['Dx']=='SCZ')].RXE)
         stat_m, pval_m = mannwhitneyu(male[(male['Dx']=='Control')].RXE,
                                       male[(male['Dx']=='SCZ')].RXE)
         print("Mann-WhitneyU of female, ctl vs sz (RXE) for %s: %.4f" %
```

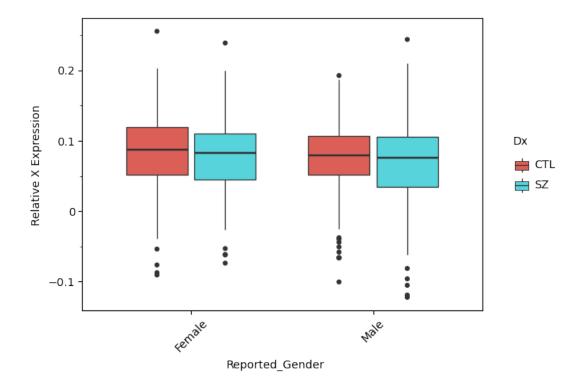
```
print("Mann-WhitneyU of male, ctl vs sz (RXE) for %s: %.4f" % (institution,
       →pval_m))
     Mann-WhitneyU of female, ctl vs sz (RXE) for MSSM: 0.1330
     Mann-WhitneyU of male, ctl vs sz (RXE) for MSSM: 0.0499
     Mann-WhitneyU of female, ctl vs sz (RXE) for Penn: 0.3124
     Mann-WhitneyU of male, ctl vs sz (RXE) for Penn: 0.0294
     Mann-WhitneyU of female, ctl vs sz (RXE) for Pitt: 0.3120
     Mann-WhitneyU of male, ctl vs sz (RXE) for Pitt: 0.1254
[25]: | female = df2[(df2['Reported_Gender']=='Female')].copy()
      male = df2[(df2['Reported_Gender']=='Male')].copy()
      stat_f, pval_f = mannwhitneyu(female[(female['Dx']=='Control')].RXE,
                                        female[(female['Dx']=='SCZ')].RXE)
      stat_m, pval_m = mannwhitneyu(male[(male['Dx']=='Control')].RXE,
                                        male[(male['Dx']=='SCZ')].RXE)
      print("Mann-WhitneyU of female, ctl vs sz (RXE) for %s: %.4f, %.2f" % ("DLPFC", _
      →pval_f, stat_f))
      print("Mann-WhitneyU of male, ctl vs sz (RXE) for %s: %.4f, %.2f" % ("DLPFC", L
       →pval_m, stat_m))
     Mann-WhitneyU of female, ctl vs sz (RXE) for DLPFC: 0.2171, 5601.00
     Mann-WhitneyU of male, ctl vs sz (RXE) for DLPFC: 0.0487, 11943.00
     1.0.2 Interaction model
[26]: df2.Reported_Gender = df2.Reported_Gender.astype("category").cat.
      →reorder_categories(["Female", "Male"], ordered=True)
      df2.Dx = df2.Dx.astype("category").cat.rename categories({"SCZ": "SZ", ...

¬"Control": "CTL"})
      df2.head(2)
[26]:
                             X autosome
                                                  sample
                                                               RXE Individual ID \
     MSSM RNA PFC 1 2.594985 2.529523 MSSM RNA PFC 1 0.065462 CMC MSSM 025
     MSSM_RNA_PFC_2 2.672613 2.594684 MSSM_RNA_PFC_2 0.077929 CMC_MSSM_136
                     Institution Reported_Gender
                                                   Dx Age_of_Death
      MSSM_RNA_PFC_1
                            MSSM
                                            Male
                                                   SZ
                                                               68.0
                                                               76.0
     MSSM_RNA_PFC_2
                            MSSM
                                            Male CTL
[27]: df3 = df2.loc[:, ['RXE', 'Reported_Gender', "Dx", 'Institution']]
      df3.loc[:, "Reported_Gender"] = df3.Reported_Gender.cat.codes
      df3.Dx = df3.Dx.cat.codes
      df3.loc[:, "Institution"] = df3.Institution.astype("category").cat.codes
      df3.head(2)
[27]:
                           RXE Reported_Gender Dx Institution
      MSSM_RNA_PFC_1 0.065462
                                                  1
```

	sum_sq	df	F	PR(>F)
C(Reported_Gender)	0.002992	1.0	0.938861	3.330063e-01
C(Dx)	0.008155	1.0	2.559338	1.102315e-01
C(Institution)	0.111778	2.0	17.540066	4.173517e-08
C(Reported_Gender):C(Dx)	0.000445	1.0	0.139511	7.089152e-01
<pre>C(Reported_Gender):C(Institution)</pre>	0.000732	2.0	0.114794	8.915716e-01
C(Dx):C(Institution)	0.013260	2.0	2.080725	1.258432e-01
Residual	1.714256	538.0	NaN	NaN

There is a massive difference between institutions, but no interaction between institution and sex or sex and diagnosis

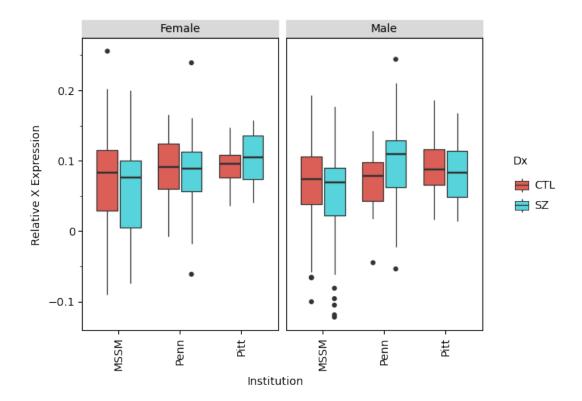
1.0.3 Plot RXE

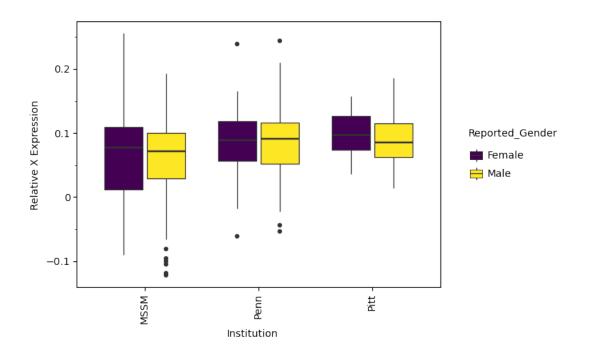


[29]: <ggplot: (8781457543801)>

/home/jbenja13/.local/lib/python3.9/sitepackages/plotnine/facets/facet_grid.py:136: FutureWarning: Index.__and__
operating as a set operation is deprecated, in the future this will be a logical
operation matching Series.__and__. Use index.intersection(other) instead
/home/jbenja13/.local/lib/python3.9/sitepackages/plotnine/facets/facet_grid.py:137: FutureWarning: Index.__and__
operating as a set operation is deprecated, in the future this will be a logical

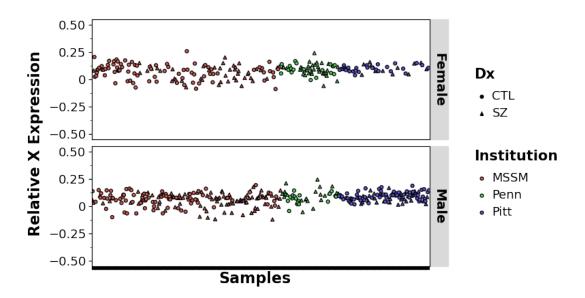
operation matching Series.__and__. Use index.intersection(other) instead





[31]: <ggplot: (8781458928014)>

/home/jbenja13/.local/lib/python3.9/sitepackages/plotnine/facets/facet_grid.py:136: FutureWarning: Index.__and__
operating as a set operation is deprecated, in the future this will be a logical
operation matching Series.__and__. Use index.intersection(other) instead
/home/jbenja13/.local/lib/python3.9/sitepackages/plotnine/facets/facet_grid.py:137: FutureWarning: Index.__and__
operating as a set operation is deprecated, in the future this will be a logical
operation matching Series.__and__. Use index.intersection(other) instead



[32]: <ggplot: (8781458553733)>

```
[33]: df2.to_csv('RXE_public.csv')
      save_ggplot(p, "RXE_public_bySex")
     /home/jbenja13/.local/lib/python3.9/site-
     packages/plotnine/facets/facet_grid.py:136: FutureWarning: Index.__and__
     operating as a set operation is deprecated, in the future this will be a logical
     operation matching Series.__and__. Use index.intersection(other) instead
     /home/jbenja13/.local/lib/python3.9/site-
     packages/plotnine/facets/facet_grid.py:137: FutureWarning: Index.__and__
     operating as a set operation is deprecated, in the future this will be a logical
     operation matching Series.__and__. Use index.intersection(other) instead
     /home/jbenja13/.local/lib/python3.9/site-
     packages/plotnine/facets/facet_grid.py:136: FutureWarning: Index.__and__
     operating as a set operation is deprecated, in the future this will be a logical
     operation matching Series.__and__. Use index.intersection(other) instead
     /home/jbenja13/.local/lib/python3.9/site-
     packages/plotnine/facets/facet grid.py:137: FutureWarning: Index. and
     operating as a set operation is deprecated, in the future this will be a logical
     operation matching Series.__and__. Use index.intersection(other) instead
     /home/jbenja13/.local/lib/python3.9/site-
     packages/plotnine/facets/facet_grid.py:136: FutureWarning: Index.__and__
     operating as a set operation is deprecated, in the future this will be a logical
     operation matching Series.__and__. Use index.intersection(other) instead
     /home/jbenja13/.local/lib/python3.9/site-
     packages/plotnine/facets/facet grid.py:137: FutureWarning: Index. and
     operating as a set operation is deprecated, in the future this will be a logical
     operation matching Series.__and__. Use index.intersection(other) instead
```

[]:[