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quasispectral

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Using quasispectral

To download the library, first install the "devtools" package, which will allow you to install packages from github; this only needs to be done once:

```
install.packages("devtools")
```

Then install the quasispectral package from github:

```
devtools::install_github("mooredf22/quasispectral")
```

This will make the functions and data in the package available to you in R.

Next, load the libraries; the libraries quasiseq and fdrtool must have been previously installed from CRAN.

```
library(QuasiSeq)
library(fdrtool)
library(quasispectral)
```

The first few lines of the data may be examined:

```
head(spectralData_CLN1_late)
    geneName CLN1.1 CLN1.2 CLN1.3 CLN1.4 CLN1.5 CLN1.6 wt.1 wt.2 wt.3 wt.4
                          67
                                135
#> 1
              149
                    142
                                      150
                                            142 135 155 159
                                                              202
       Aga
#> 2
              577 542
                                601
       Arsa
                          314
                                      569
                                            590 602 639
                                                         617
                                                              599
#> 3
       Arsb
              466
                    482
                          258
                                409
                                      465
                                            465 624 599
                                                         645 688
#> 4
       Arsq
              68
                    65
                           41
                                55 69
                                             63
                                                 87
                                                      84
                                                          46
                                                              98
#> 5
       Arsk
               59
                     60
                           22
                                 39
                                       45
                                             54
                                                 75
                                                      75
                                                          80 107
#> 6
      Asah1
              806
                    905
                          533
                                729
                                      748
                                            827 865 812 887 861
   wt.5 wt.6
#> 1 195 181
#> 2 613 631
#> 3 662 667
#> 4 107
          99
          79
#> 5
     77
#> 6
    918 904
```

The first column is the gene name, the next 6 columns are spectral counts for mutant animals, and the final 6 columns are counts for wildtype animals.

Finally, call the package as follows, and look at the first few lines of output:

```
result.QLfit <- quasiSpectral(spectralDataAll=spectralData_CLN1_late, n.mut=6, n.wt=6)
#> Currently, bias reduction has not yet been implemented for Poisson model.
#> Spline scaling factor: 0.983018553199358
#> Warning in fdrtool(p.values, statistic = "pvalue"): There may be too few
```

5/11/2019 quasispectral #> input test statistics for reliable FDR calculations!

```
#> Step 1... determine cutoff point
#> Step 2... estimate parameters of null distribution and eta0
#> Warning: Censored sample for null model estimation has only size 8 !
#> Step 3... compute p-values and estimate empirical PDF/CDF
#> Step 4... compute q-values and local fdr
#> Step 5... prepare for plotting
QLfit.out <- result.QLfit$QLfit
head(QLfit.out)
    geneName CLN1.1 CLN1.2 CLN1.3 CLN1.4 CLN1.5 CLN1.6 wt.1 wt.2 wt.3 wt.4
#> 1
                149
                    142
                             67
                                   135
                                          150
                                                 142 135 155 159
                                                                    202
                    542
#> 2
        Arsa
               577
                             314
                                    601
                                          569
                                                 590 602 639 617
                                                                    599
#> 3
        Arsb
               466
                      482
                             258
                                   409
                                          465
                                                 465 624
                                                          599
                                                               645
                                                                    688
#> 4
        Arsq
                68
                      65
                              41
                                    55
                                           69
                                                  63
                                                      87
                                                           84
                                                                46
                                                                     98
        Arsk
                59
                                           45
                                                      75
#> 5
                       60
                              22
                                    39
                                                  54
                                                           75
                                                                80
                                                                   107
#> 6
       Asah1
                806
                      905
                             533
                                    729
                                          748
                                                 827 865 812 887 861
    wt.5 wt.6 coef.main
                           p.values p.values.bonf p.values.holm
#> 1 195 181 0.3129084 1.254920e-03 1.016486e-01 4.141237e-02
    613 631 0.1382393 5.303151e-03 4.295553e-01 1.198925e-01
#> 3 662 667 0.5354857 4.922626e-11 3.987327e-09 3.790422e-09
           99 0.4545213 3.430625e-03 2.778806e-01 8.919292e-02
#> 4 107
#> 5
     77
          79 0.7465593 6.204245e-06 5.025439e-04 2.978038e-04
#> 6 918 904 0.1314973 1.497866e-02 1.000000e+00 2.845946e-01
     q.values.BH q.values.strimmer
#> 1 2.074460e-03
                     2.914893e-04
#> 2 7.159254e-03
                    1.029129e-03
#> 3 7.974655e-10
                     1.120724e-10
#> 4 4.875099e-03
                    7.007858e-04
#> 5 1.478070e-05
                   2.104575e-06
#> 6 1.925828e-02
                     2.698284e-03
```

To see the estimated proportion of null samples in the population, from fdrtool, look at eta0:

```
result.QLfit$eta0
                      # estimated proportion of null samples in population
#> [1] 0.143748
```