MLExpResso: differential expression and methylation analysis

Case study using RTCGA data

Aleksandra Dabrowska, Alicja Gosiewska

Contents

1	Pac	ekage (Abstract?)	1
2	Star	ndard Workflow	1
	2.1	Methylation	1
		2.1.1 BRCA_methylation_chr17 data set	-
		2.1.2 Testing	2
	2.2	Expression	
		2.2.1 BRCA_mRNAseq_chr17 data set	
		2.2.2 Testing	
	2.3	Comparing test results	
	2.4	Choosing most different genes	4
	2.5	Visualization	4

1 Package (Abstract?)

It is considered that the result of increased methylation is decreased gene expression. While, recent studies suggest that the relationship between methylation and expression is more complex than was previously thought. The package MLExpResso provides methods to test for differential expression and methylation by use of the negative binonial distribution and t-test. Additionaly 'MExpResso allows to visualize results in a simple way.

2 Standard Workflow

In this vignette we will work with the data sets containing information about gene expression and methylation for patients with breast cancer. We will analyze differences in methylation and expression for patients with different subtypes of BRCA cancer.

2.1 Methylation

2.1.1 BRCA_methylation_chr17 data set

In this section, we will work with the methylation level data from TCGA database. Package contains BRCA_methylation_chr17 dataset. BRCA_methylation_chr17 contains information about methylation of CpG islands for patients with breast cancer. Rows of this data set correspond to patients, more precisely, to samples taken from patients. First column SUBTYPEcorresponds to a subtype of BRCA cancer, next columns correspond to CpG islands. Values inside the table indicate the methylation level of CpG island for specified sample.

```
library(MLExpResso)
```

library(MLExpRessodata)

```
head(BRCA_methylation_chr17)[1:5,1:4]
```

In this analysis we would like to find genes with different methylation and expression. At first we need to use function aggregate_probes, which generates new data frame with CpG islands mapped to genes.

```
BRCA_methylation_gen <- aggregate_probes(BRCA_methylation_chr17[,-1])
head(BRCA_methylation_gen)[1:5,1:4]</pre>
```

```
## TCGA-A2-A04N-01A-11D-A112-05 0.7148533 0.8625816 0.24294092 0.7835302 ## TCGA-A2-A04N-01A-11D-A112-05 0.5850106 0.8355825 0.21367129 0.8466190 ## TCGA-A2-A04P-01A-31D-A032-05 0.4495537 0.8786166 0.03277413 0.3417919 ## TCGA-A2-A04Q-01A-21D-A032-05 0.7120650 0.8819490 0.03460160 0.7264985 ## TCGA-A2-A04T-01A-21D-A032-05 0.6010397 0.7739978 0.02501599 0.6276399
```

In this case we have two conditions, connected with subtypes of breast cancer.

Before we go to the testing, we need to define condition values for each sample. We would like to test for differences between LumA subtype and other subtypes of breast cancer, so we create a vector, which each element corresponds to a sample. Our division into this two groups relies on numbers of occurences of each subtype. The LumA subtype is the most common, in case of breast cancer.

```
condition_met <- ifelse(BRCA_methylation_chr17$SUBTYPE=="LumA","LumA", "other")
head(condition_met, 8)</pre>
```

```
## [1] "LumA" "LumA" "other" "other" "other" "other" "LumA" "other"
```

2.1.2 Testing

```
res_met <- calculate_test(BRCA_methylation_chr17[,-1], condition_met, test="ttest")
head(res_met)</pre>
```

```
## cg10275770 cg10275770 0.3330801 -0.1515132 3.795188e-17
## cg06144905 cg06144905 0.4111616 0.1738623 9.908386e-14
## cg04049033 cg04049033 0.6462602 -0.1027159 1.513031e-13
## cg12045829 cg12045829 0.2485025 -0.1341285 5.910592e-12
## cg02473123 cg02473123 0.8127112 0.0982269 1.653082e-11
## cg05246522 cg05246522 0.5549808 0.1997340 2.069380e-11
```

2.2 Expression

2.2.1 BRCA_mRNAseq_chr17 data set

Data set BRCA_mRNAseq_chr17 contains information about gene expression. This data set contains per-gene read counts computed for genes for 736 patients with breast cancer. Rows of this data set correspond to samples taken from patients. First column SUBTYPEcorresponds to a subtype of BRCA cancer, next columns correspond to genes.

```
BRCA_mRNAseq_chr17[1:5,1:5]
```

```
SUBTYPE AANAT AARSD1 AATF AATK
##
## TCGA-A1-A0SB-01A-11R-A144-07
                                  Normal
                                              9
                                                  2354 2870
                                                             317
## TCGA-A1-AOSD-O1A-11R-A115-07
                                              2
                                                  1846 5656
                                    LumA
                                                             312
## TCGA-A1-A0SE-01A-11R-A084-07
                                    LumA
                                             11
                                                  3391 9522
                                                             736
## TCGA-A1-A0SF-01A-11R-A144-07
                                    LumA
                                              0
                                                  2169 4625
                                                             169
## TCGA-A1-A0SG-01A-11R-A144-07
                                    LumA
                                              1
                                                  2273 3473
                                                               92
```

In our example we will test for differential expression between groups with LumA breast cancer subtype and other subtypes of that cancer. Again we will use vector conditions, which consist of two values corresponds to subtype of breast cancer: LumA and other.

```
condition_exp <- ifelse(BRCA_mRNAseq_chr17$SUBTYPE=="LumA","LumA","other")
head(condition_exp, 8)
## [1] "other" "LumA" "LumA" "LumA" "LumA" "LumA" "other" "LumA"</pre>
```

2.2.2 Testing

```
res_exp <- calculate_test(BRCA_mRNAseq_chr17[,-1], condition_exp, test="lrt")
head(res_exp)
##
         id
                      log2.fold
               mean
                                      pval
                     0.35015880 0.20167059
      AANAT
               3.64
## 2 AARSD1 2857.44 -0.09262390 0.35921511
## 3
       AATF 6973.70
                     0.11885989 0.24882591
## 4
       AATK 376.04 0.01594810 0.94744134
## 5
      ABCA5 1997.68 -0.09521489 0.64228765
      ABCA6 674.18 -0.46348803 0.06796017
## 6
```

2.3 Comparing test results

```
genes_comparison <- calculate_comparison_table(BRCA_mRNAseq_chr17[ ,-1], BRCA_methylation_gen,
                                   condition_exp, condition_met, test1="nbinom2", test2="ttest")
head(genes_comparison)
##
         id nbinom2.log2.fold nbinom2.pval ttest.log2.fold
                                                               ttest.pval
                                 0.22109578
                                                 0.07237890 1.026904e-05
## 1
     AANAT
                   0.36342624
## 2 AARSD1
                  -0.07435954
                                 0.45027777
                                                 0.01114861 5.052981e-01
## 3
       AATF
                   0.13308688
                                 0.18597163
                                                 0.01360521 2.136847e-01
## 4
       AATK
                   0.02308606
                                 0.92313513
                                                 0.06619897 1.427292e-05
## 5
      ABCA5
                  -0.07034249
                                 0.73314806
                                                 0.00510806 6.353608e-01
## 6
      ABCA6
                  -0.44685708
                                 0.08266088
                                                -0.03402781 4.393248e-02
##
      genes_rank
```

```
## 1 0.001506798
## 2 0.476995295
## 3 0.199347171
## 4 0.003629853
## 5 0.682505353
## 6 0.060261904
```

2.4 Choosing most different genes

```
Sorting \dots
```

```
genes_comparison_sorted <- genes_comparison[order(genes_comparison$ttest.pval), ]</pre>
head(genes_comparison_sorted)
            id nbinom2.log2.fold nbinom2.pval ttest.log2.fold
##
                                                                 ttest.pval
## 302
         ICAM2
                       0.2077055 2.102511e-01
                                                   -0.15151320 3.754116e-17
## 519
          RILP
                       0.3139026 5.252448e-02
                                                   -0.05073691 2.575168e-13
                                                    0.11505558 5.360053e-12
         PIPOX
## 466
                       0.3589979 4.371416e-02
## 652 TNFSF12
                      -0.4357410 3.486764e-04
                                                   -0.13412855 5.867083e-12
## 133
           CD7
                       1.3728093 9.099866e-07
                                                    0.09822690 1.641919e-11
## 341
          KSR1
                       0.4261993 1.658465e-02
                                                    0.19973400 2.054467e-11
##
         genes_rank
## 302 2.809461e-09
```

133 3.865391e-09

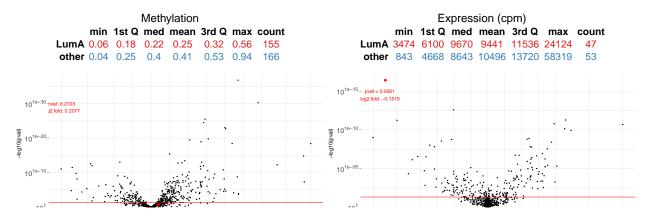
341 5.837174e-07

519 1.163011e-07 ## 466 4.840560e-07 ## 652 4.522956e-08

ICAM2!

2.5 Visualization

Visualizing chosen gene - IGFALS.



Note that plot_gene methylation require data frame with cpg islands, not genes.

plot_gene(condition_exp, condition_met, BRCA_mRNAseq_chr17, BRCA_methylation_chr17, "ICAM2")

