# MLExpResso: differential expression and methylation analysis

# Case study using RTCGA data

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# 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" "LumA" "other"
```

# **2.1.2** Testing

```
res_met <- calculate_test(BRCA_methylation_gen, condition_met, test="ttest")
head(res_met)</pre>
```

```
##
         id
              log2.fold
                                pval mean_LumA mean_other
## 1
      ICAM2 -0.15151320 3.754116e-17 0.2547275 0.4062407 0.3330801
## 2
       RILP -0.05073691 2.575168e-13 0.3079069 0.3586438 0.3341447
      PIPOX 0.11505558 5.360053e-12 0.4242804 0.3092248 0.3647812
## 3
## 4 TNFSF12 -0.13412855 5.867083e-12 0.1791401 0.3132686 0.2485025
## 5
        CD7 0.09822690 1.641919e-11 0.8635077 0.7652808 0.8127112
## 6
       KSR1 0.19973400 2.054467e-11 0.658270
                                                0.458536 0.5549808
```

#### 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
                                    LumA
                                              2
                                                  1846 5656
                                                             312
## TCGA-A1-A0SE-01A-11R-A084-07
                                                  3391 9522
                                                             736
                                    T.11m A
                                             11
## 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)</pre>
```

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

## 2.2.2 Testing

```
res_exp <- calculate_test(BRCA_mRNAseq_chr17[,-1], condition_exp, test="lrt")
head(res_exp)
##
        id log2.fold
                            pval mean_LumA mean_other
                                                          mean
## 1 AURKB 2.339920 3.191000e-32
                                 539.0426
                                            2323.8868
                                                       1485.01
                                                       2574.48
## 2 CBX2 2.895062 2.834335e-26
                                            4296.6038
                                  632.5106
## 3 KPNA2 1.447288 8.551812e-24 11547.36
                                              26427.38 19433.77
## 4 PRR11 3.822148 2.286874e-22
                                   396.383
                                              3479.981
                                                       2030.69
## 5 BIRC5 1.988998 1.953941e-21 1957.085
                                                       4448.76
                                              6658.358
     GSG2 1.405039 3.527773e-21 278.2128
## 6
                                              629.3396
                                                        464.31
```

# 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")
## Warning in sqrt(result[, 2] * result[, 4]): wyprodukowano wartości NaN
## Warning: Column `id` joining character vector and factor, coercing into
## character vector
head(genes_comparison)
##
             id nbinom2.log2.fold nbinom2.pval ttest.log2.fold ttest.pval
## 354
          LSM12
                      0.056616564
                                     0.4954925
                                                  9.234253e-05 0.87094671
## 579
       SMARCE1
                     -0.023092119
                                     0.8318721
                                                 -2.377235e-04 0.60251281
```

```
## 77 C17orf61
                     -0.006548806
                                      0.9585559
                                                  -1.027549e-03 0.10515489
## 367
           MFD9
                     -0.003427851
                                      0.9677837
                                                  -2.615948e-03 0.03653572
## 482
        PRKAR1A
                     -0.003404099
                                      0.9791404
                                                  -4.099641e-03 0.08891401
           GRB2
                     -0.068323804
                                      0.5412281
                                                  -2.148793e-04 0.85842993
## 273
       geom.mean.rank No.probes
##
          0.002286507
## 354
## 579
          0.002342976
                               2
          0.002594074
                               2
## 77
## 367
          0.002994508
                               1
                               2
## 482
          0.003735717
## 273
          0.003831628
                               2
```

# 2.4 Choosing most different genes

```
Sorting \dots
```

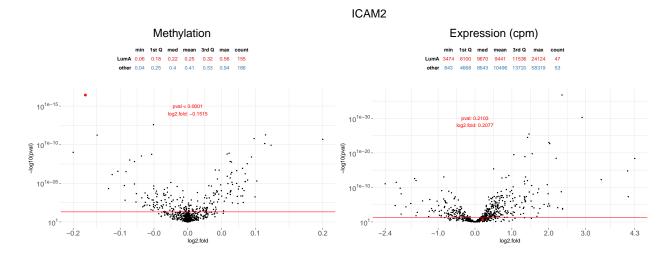
```
genes_comparison_sorted <- genes_comparison[order(genes_comparison$ttest.pval), ]
head(genes_comparison_sorted)</pre>
```

```
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
## 466
         PIPOX
                       0.3589979 4.371416e-02
                                                    0.11505558 5.360053e-12
## 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
##
       geom.mean.rank No.probes
## 302
                  NaN
## 519
                  NaN
                               2
## 466
            0.2032356
                               2
## 652
            0.2417546
                               2
## 133
            0.3672149
                               1
## 341
            0.2917644
                               1
```

ICAM2!

## 2.5 Visualization

Visualizing chosen gene - IGFALS.



Note that plot\_gene methylation require data frame with cpg islands, not genes. plot\_gene(BRCA\_methylation\_chr17,BRCA\_mRNAseq\_chr17,condition\_met, condition\_exp, "ICAM2")

