

Solutions for chapter Differential Expression

Exercise 1

- There is a weak relationship but it is not dominant. We may safely proceed with the nonspecific filtering based on variability.
- It switches between plotting the x -axis (means) on the original scale (**FALSE**) or on the rank scale (**TRUE**). The latter distributes the data more evenly along the x -axis and allows a better visual assessment of the standard deviation as a function of the mean.

Exercise 2

The number of probe sets with p -value less than 0.05 and mean \log_2 fold-change larger than 0.5 is

```
> sum(tt$p.value<0.05 & abs(tt$dm)>0.5)
[1] 224
```

This choice of thresholds is of course arbitrary.

Exercise 3

```
> mtyp = ALLset1$mol.biol
> sel = rep(1:2, each=rev(table(mtyp)))
> plot(exprs(ALLset1)[j, order(mtyp)], pch=c(1,15)[sel],
       col=c("black", "red")[sel],
       main=featureNames(ALLset1)[j],
       ylab=expression(log[2]~expression~level))
> legend("bottomleft", col=c("black", "red"),
       pch=c(1,15), levels(mtyp), bty="n")
```

Exercise 4

The curve for a bad discriminator would be close to the diagonal because the classification would be almost random. The curve for a perfect discriminator shows both high sensitivity and high specificity over the whole plot, that is, a rectangle from $[0,1]$ to $[1,1]$.

Exercise 5

The identification of differentially expressed genes by area under the ROC curve is not so much affected by the sample size as the t -statistic is. For the t -test the number of differentially expressed genes increases constantly with the sample size. For the ROC curves this number stabilizes with a sufficient sample size.