

Preparation



• Create the DESeq2 object

```
library(DESeq2)
mr$Group <- factor(mr$Group)
d <- DESeqDataSetFromMatrix(countData=cf,colData=mr,design=~Group)
d</pre>
```

```
## class: DESeqDataSet
## dim: 17515 6
## metadata(1): version
## assays(1): counts
## rownames(17515): mt-Cytb mt-Td ... 4930447M23Rik Gm6518
## rowData names(0):
## colnames(6): DSSd00_1 DSSd00_2 ... DSSd07_2 DSSd07_3
## colData names(7): SampleName SampleID ... Group Replicate
```

- Categorical variables must be factors
- Building GLM models: ~var , ~covar+var

Size factors



• Normalisation factors are computed

```
d <- DESeq2::estimateSizeFactors(d,type="ratio")
sizeFactors(d)

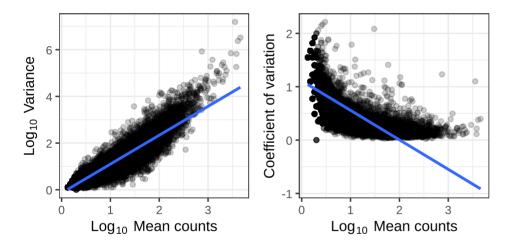
## DSSd00_1 DSSd00_2 DSSd00_3 DSSd07_1 DSSd07_2 DSSd07_3
## 1.0153287 0.9597101 0.9984645 1.0358161 1.0787996 0.9988740</pre>
```

Dispersion



• We need to measure the variability of gene counts

```
dm <- apply(cf,1,mean)
dv <- apply(cf,1,var)
cva <- function(x) sd(x)/mean(x)
dc <- apply(cf,1,cva)</pre>
```



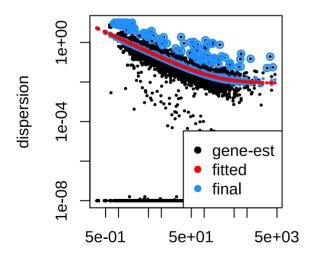
• Dispersion is a measure of variability in gene expression for a given mean

Dispersion



- Dispersion is unreliable for low mean counts
- Genes with similar mean values must have similar dispersion
- Estimate likely (ML) dispersion for each gene based on counts
- Fit a curve through the gene-wise estimates
- Shrink dispersion towards the curve

```
d <- DESeq2::estimateDispersions(d)
{par(mar=c(4,4,1,1))
plotDispEsts(d)}</pre>
```



mean of normalized counts

Testing



Log2 fold changes changes are computed after GLM fitting

```
dg <- nbinomWaldTest(d)
resultsNames(dg)

## [1] "Intercept" "Group_day07_vs_day00"</pre>
```

- Use results() to customise/return results
 - Set coefficients using contrast or name
 - Filtering results by fold change using lfcThreshold
 - cooksCutoff removes outliers
 - independentFiltering removes low count genes
 - pAdjustMethod sets method for multiple testing correction
 - o alpha set the significance threshold

Testing



```
res1 <- results(dg,name="Group_day07_vs_day00",alpha=0.05)
summary(res1)

##

## out of 17515 with nonzero total read count

## adjusted p-value < 0.05

## LFC > 0 (up) : 194, 1.1%

## LFC < 0 (down) : 217, 1.2%

## outliers [1] : 1, 0.0057%

## low counts [2] : 9169, 52%

## (mean count < 10)

## [1] see 'cooksCutoff' argument of ?results

## [2] see 'independentFiltering' argument of ?results
```

Testing



head(res1)

```
## log2 fold change (MLE): Group day07 vs day00
## Wald test p-value: Group day07 vs day00
## DataFrame with 6 rows and 6 columns
          baseMean log2FoldChange lfcSE
                                              stat pvalue
                                                               padi
##
         <numeric>
                       <numeric> <numeric> <numeric> <numeric> <numeric>
## mt-Cvtb 1.304697
                       0.658760 1.658716 0.397150 0.691257
                                                                 NA
## mt-Td 1.492515
                       -1.080724 1.521562 -0.710273 0.477535
                                                                 NA
## mt-Co1 0.327758 1.813649 3.064543 0.591817 0.553973
                                                                 NA
## mt-Tw 18.839105 0.209559 0.425694
                                          0.492277 0.622524 0.930584
## mt-Ti 2.649343
                       -1.200646 1.135727 -1.057160 0.290438
                                                                 NA
## mt-Nd1 23.325014
                        0.657781 0.401785 1.637148 0.101599 0.529489
```

• Use lfcShrink() to correct fold changes for high dispersion genes

