## Differential Expression Analysis using limma

COMBINE RNA-seq Workshop



## Many plotting options available...



## Linear models for differential expression



## limma package:

## Linear Models for Microarrays \& RNA-seq

## Analysis of differential expression studies

- arbitrarily complex experiments: linear models, contrasts
- empirical Bayes methods for differential expression: $t$-tests, $F$-tests, posterior odds
- analyse log-ratios, log-intensities, log-CPM values
- accommodate quality weights in analysis
- control of FDR across genes and contrasts
- many plotting functions to help visualize raw data and final results from statistical analysis
- gene set testing at various levels
- fast, numerically efficient methods


## RNA-seq of Mouse mammary gland



Fu et al. (2015) 'EGF-mediated induction of Mcl-1 at the switch to lactation is essential for alveolar cell survival' Nat Cell Biol

## (some) questions we can ask

- Which genes are differentially expressed between basal and luminal cells?
- ... between basal and luminal in virgin mice?
- ... between pregnant and lactating mice?
- ... between pregnant and lactating mice in basal cells?


## What do we need to perform a statistical test?

- Measure of average expression
- Measure of variability



## One of the most useful statistics: $\boldsymbol{t}$-test

- We want to test the null hypothesis:

H0: mean(GroupA) = mean(GroupB)
against the alternative hypothesis:
H1: mean(GroupA) $\neq$ mean(GroupB)

- An important assumption of the $t$-test is that the data is roughly normally distributed
- A statistician's best trick is to transform data that isn't normally distributed into something that looks more normally distributed


## Log-counts vs counts for one gene

GeneID: 58175


GeneID: 58175

*A quick check to see how normal your data is: compare the mean and the median

## We can perform t-tests on log-counts

- Take into account different sequencing depths
- Take into account normalisation factors
- Take into account we can't log a zero
- The cpm(y, log=TRUE) function does this for you


## Now we have log-counts

- Calculate means and variances on the logcounts
$T=\log F C / S t d D e v / \sqrt{ } n$
- logFC is the difference in means between the two groups for the log-counts
- The t-statistic is t-distributed on n -1 degrees of freedom
- P-values!



## RNA-seq data is more complicated

- Mean-variance relationship. Use voom



# Although we test one gene at a time, we can share information about all the genes to help with testing 

Before sharing


After sharing


## Multiple testing burden

- Problem: We are performing tens of thousands of tests, which increases our chances of getting false discoveries
- Solution: Calculate false discovery rates ("adjusted $p$-values" in limma)
- Interpretation: If there are 100 genes significant at FDR<5\%, we are willing to accept that 5 will be false discoveries


# Linear modelling analysis pipeline for RNA-seq data 

- model.matrix / makeContrasts
- voom
- lmFit
- contrasts.fit
- treat
- eBayes
- topTable / topTreat
- decideTests

