

INF-BIOx121 2017

RNA-seq

differential expression analysis

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RNA-seq analysis

Differential expression (DE)

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Differential expression

- ❖ Genes
- ❖ Transcripts (Isoforms)
- ❖ Allele specific expression
- ❖ Exon level expression

Counting

- ❖ Feature - genes, transcript or exon
- ❖ How many reads aligned to each feature of interest?
- ❖ What is the length of the feature?

- ❖ Raw count calculated from BAM files using featureCounts, HTSeq, etc
- ❖ Most (all) DE tools would require raw count file and not (pre) scaled data.

Normalisation

- ❖ Normalisation within and across samples
- ❖ Count gets converted to RPKM, FPKM or TPM
- ❖ RPKM (Reads Per Kilobase Million)

Scaling factor = Total number reads / 1,000,000

RPM = Read count per feature / scaling factor

RPKM = RPM / Feature length in kilo bases

- ❖ FPKM (Fragments Per Kilobase Million)

FPM = Fragment count per feature / scaling factor

FPKM = FPM / Feature length in kilo bases

- ❖ TPM (Transcripts Per Kilobase Million)

RPK = Read count per feature / Feature length in kilo bases

Scaling factor = sum of RPK / 1,000,000

TPM = RPK / Scaling factor

DESeq2 (or edgeR) is different!!

<https://www.youtube.com/watch?v=UFB993xufUU>

<https://www.youtube.com/watch?v=TTUrtCY2k-w>

DESeq2

- ❖ Generalised linear model fit
 - ❖ Using negative binomial distribution (aka gamma-Poisson distribution)
- ❖ Empirical Bayes shrinkage
 - ❖ for within-group variability, i.e., variability between replicates
- ❖ Fold change estimation
- ❖ Not just pair-wise comparison. Allows for complicated nested designs to be compared

DESeq2

- ❖ `plotDispEsts()`: To look at the dispersion plots
- ❖ `plotPCA()`: To find outliers
- ❖ `plotMA()`: Exploring DE results

Multiple hypothesis testing and FDR

Multiple hypothesis testing

- ❖ Thousands of genes = thousands of hypothesis tests (simultaneously)
- ❖ Increased chance of false positives! (Type I error)
 - ❖ e.g. you test for differential expression in 1000 genes that are not differentially expressed
 - ❖ You would expect $1000 \times 0.05 = 50$ of them to have a $P\text{-value} < 0.05$
- ❖ Individual P -values not useful : Need multiple testing statistic instead

False Discovery rate (Benjamini & Hochberg 1995)

- ❖ The expected proportion of Type I errors among the rejected hypotheses
 - ❖ i.e. the proportion of false positives
 - ❖ Tends to be conservative if many genes are DE
 - ❖ $\text{FDR} = 0.05$ common for exploratory / broad scope studies
 - ❖ $\text{FDR} < 0.05$ common for medical applications and hunts for candidate genes