

# Data Preprocessing

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Workshop on RNA-Seq

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NBIS, SciLifeLab

# Raw data

- Raw count table

##	<i>DSSd00_1</i>	<i>DSSd00_2</i>	<i>DSSd00_3</i>	<i>DSSd07_1</i>	<i>DSSd07_2</i>	<i>DSSd07_3</i>
## <i>ENSMUSG000000102693</i>	0	0	0	0	0	0
## <i>ENSMUSG000000064842</i>	0	0	0	0	0	0
## <i>ENSMUSG000000051951</i>	0	1	2	0	3	2
## <i>ENSMUSG000000102851</i>	0	0	0	0	0	0
## <i>ENSMUSG000000103377</i>	0	0	0	0	0	0
## <i>ENSMUSG000000104017</i>	0	0	0	0	0	0

- Metadata

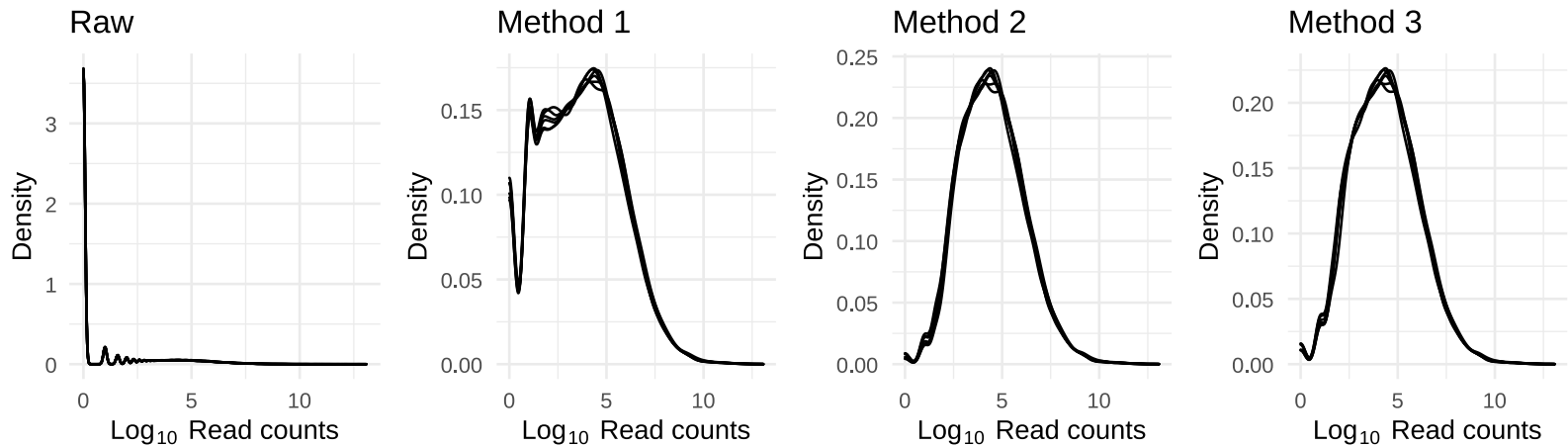
##	<i>SampleName</i>	<i>SampleID</i>	<i>No</i>	<i>Model</i>	<i>Day</i>	<i>Group</i>	<i>Replicate</i>
## <i>DSSd00_1</i>	<i>DSSd00_1</i>	<i>KI_PC1606_01</i>	1	<i>DSS</i>	0	<i>day00</i>	1
## <i>DSSd00_2</i>	<i>DSSd00_2</i>	<i>KI_PC1606_02</i>	2	<i>DSS</i>	0	<i>day00</i>	2
## <i>DSSd00_3</i>	<i>DSSd00_3</i>	<i>KI_PC1606_03</i>	3	<i>DSS</i>	0	<i>day00</i>	3
## <i>DSSd07_1</i>	<i>DSSd07_1</i>	<i>KI_PC1606_13</i>	13	<i>DSS</i>	7	<i>day07</i>	1
## <i>DSSd07_2</i>	<i>DSSd07_2</i>	<i>KI_PC1606_14</i>	14	<i>DSS</i>	7	<i>day07</i>	2
## <i>DSSd07_3</i>	<i>DSSd07_3</i>	<i>KI_PC1606_15</i>	15	<i>DSS</i>	7	<i>day07</i>	3

# Filtering

- Remove genes and samples with low counts

```
cf1 <- cr[rowSums(cr>0) >= 3, ]  
cf2 <- cr[rowSums(cr>3) >= 3, ]  
cf3 <- cr[rowSums(edgeR::cpm(cr)>5) >= 3, ]
```

- Inspect distribution



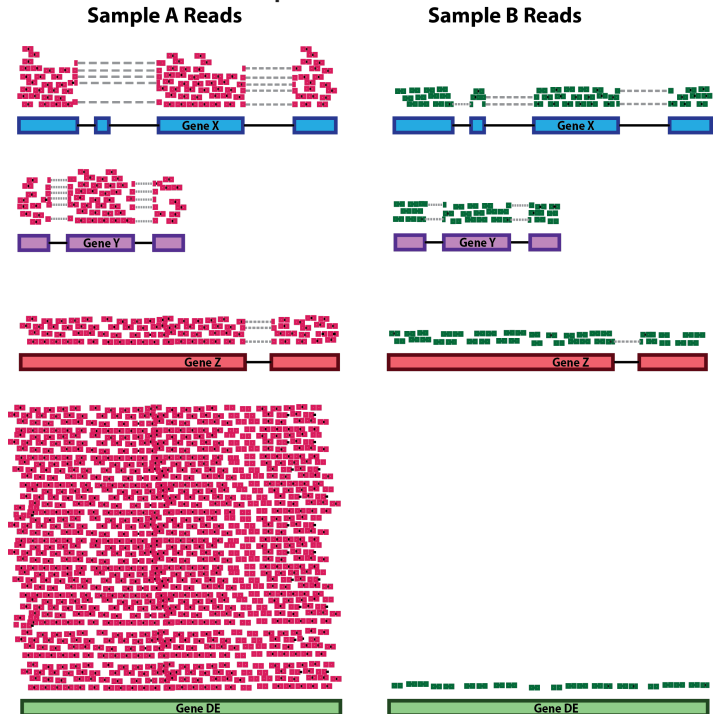
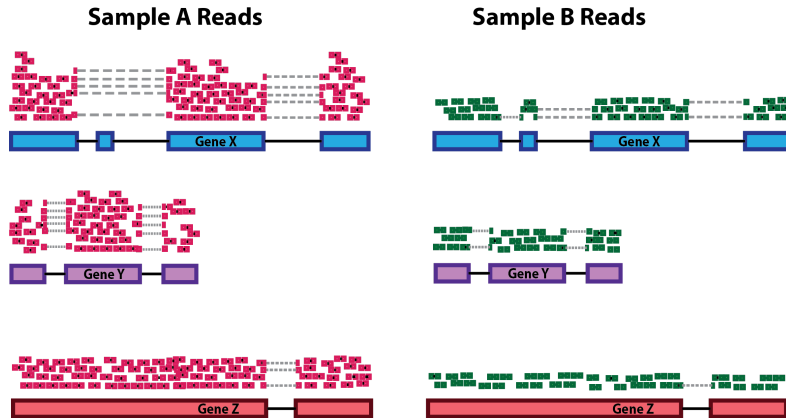
- Inspect the number of rows (genes) available after filtering

```
## Raw: 55487, Method 1: 16099, Method 2: 11783, Method 3: 12496
```

# Normalisation

- Make counts comparable across samples
- Control for sequencing depth

- Control for compositional bias



```
##      A B A_tc B_tc
## x    20 6 0.33 0.38
## y    25 6 0.42 0.38
## z    15 4 0.25 0.25
```

```
##      A B A_tc B_tc
## x      0 20 0.00 0.39
## y     25 25 0.18 0.49
## z     15  4 0.11 0.08
## de  100  2 0.71 0.04
```

# Normalisation

- Make counts comparable across features (genes)

## Sample A Reads



##	counts	gene_length	norm_counts
## x	50	10	5
## y	25	5	5

- Bring counts to a human-friendly scale

# Normalisation

## Normalisation by library size

- Assumes total expression is the same under different experimental conditions
- Methods include TC, RPKM, FPKM, TPM
- RPKM, FPKM and TPM control for sequencing depth and gene length
- Total number of RPKM/FPKM normalized counts for each sample will be different, therefore, you cannot compare the normalized counts for each gene equally between samples.
- TPM enables better comparison between samples because total per sample sums to equal value

##		A	B	len	A_rpm	B_rpm	A_rpk	B_rpk	A_tpm	B_tpm		
##	x	20	6	20000	408163	222222	20.41	11.11	0.001000	0.00030	493827	153846
##	y	25	6	40000	510204	222222	12.76	5.56	0.000625	0.00015	308642	76923
##	z	4	15	10000	81633	555556	8.16	55.56	0.000400	0.00150	197531	769231
##	sum	49	27	70000	1000000	1000000	41.33	72.23	0.002025	0.00195	1000000	1000000

## Normalisation by distribution

- Assumes technical effects are same for DE and non-DE genes
- Assumes number of over and under-expressed genes are roughly same across conditions
- Corrects for compositional bias
- Methods include Q, UQ, M, RLE, TMM, MRN
- `edgeR::calcNormFactors()` implements TMM, TMMwzp, RLE & UQ
- `DESeq2::estimateSizeFactors()` implements median ratio method (RLE)
- Does not correct for gene length
- `geTMM` is gene length corrected TMM

```
##      A  B   len   ref A_ratio B_ratio   A_mrn   B_mrn
## x 20  6 20000 10.95   1.83   0.55 10.928962 10.90909
## y 25  6 40000 12.25   2.04   0.49 13.661202 10.90909
## z  4 15 10000  7.75   0.52   1.94  2.185792 27.27273
```

## Normalisation by testing

- A more robust version of normalisation by distribution.
- A set of non-DE genes are detected through hypothesis testing
- Tolerates a larger difference in number of over and under expressed genes between conditions
- Methods include PoissonSeq, DEGES

## Normalisation using Controls

- Assumes controls are not affected by experimental condition and technical effects are similar to all other genes
- Useful in conditions with global shift in expression
- Controls could be house-keeping genes or spike-ins
- Methods include RUV, CLS

## Stabilizing variance

- Variance is stabilised across the range of mean values
- Methods include VST, RLOG, VROOM
- For use in exploratory analyses. Not for DE.
- `vst()` and `rlog()` functions from *DESeq2*
- `voom()` function from *Limma* converts data to normal distribution



## Recommendations

- Most tools use a mix of many different normalisations
- For DGE using DGE R packages (DESeq2, edgeR, Limma etc), use raw counts
- For visualisation (PCA, clustering, heatmaps etc), use VST or RLOG
- For own analysis with gene length correction, use TPM (maybe geTMM?)
- Custom solutions: spike-ins/house-keeping genes



# Thank you. Questions?

R version 4.0.3 (2020-10-10)

Platform: x86\_64-pc-linux-gnu (64-bit)

OS: Ubuntu 18.04.5 LTS

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Built on: 📅 01-Dec-2020 at 🕒 20:57:09

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