

*UiO IN-BIOS5000/9000*

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# RNA-seq

## differential expression analysis

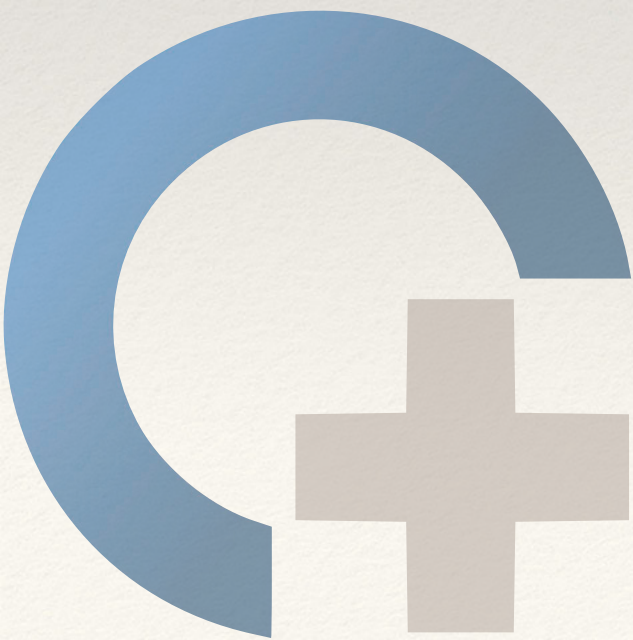
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Nov 07, 2024

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OUS, Ullevål, Oslo

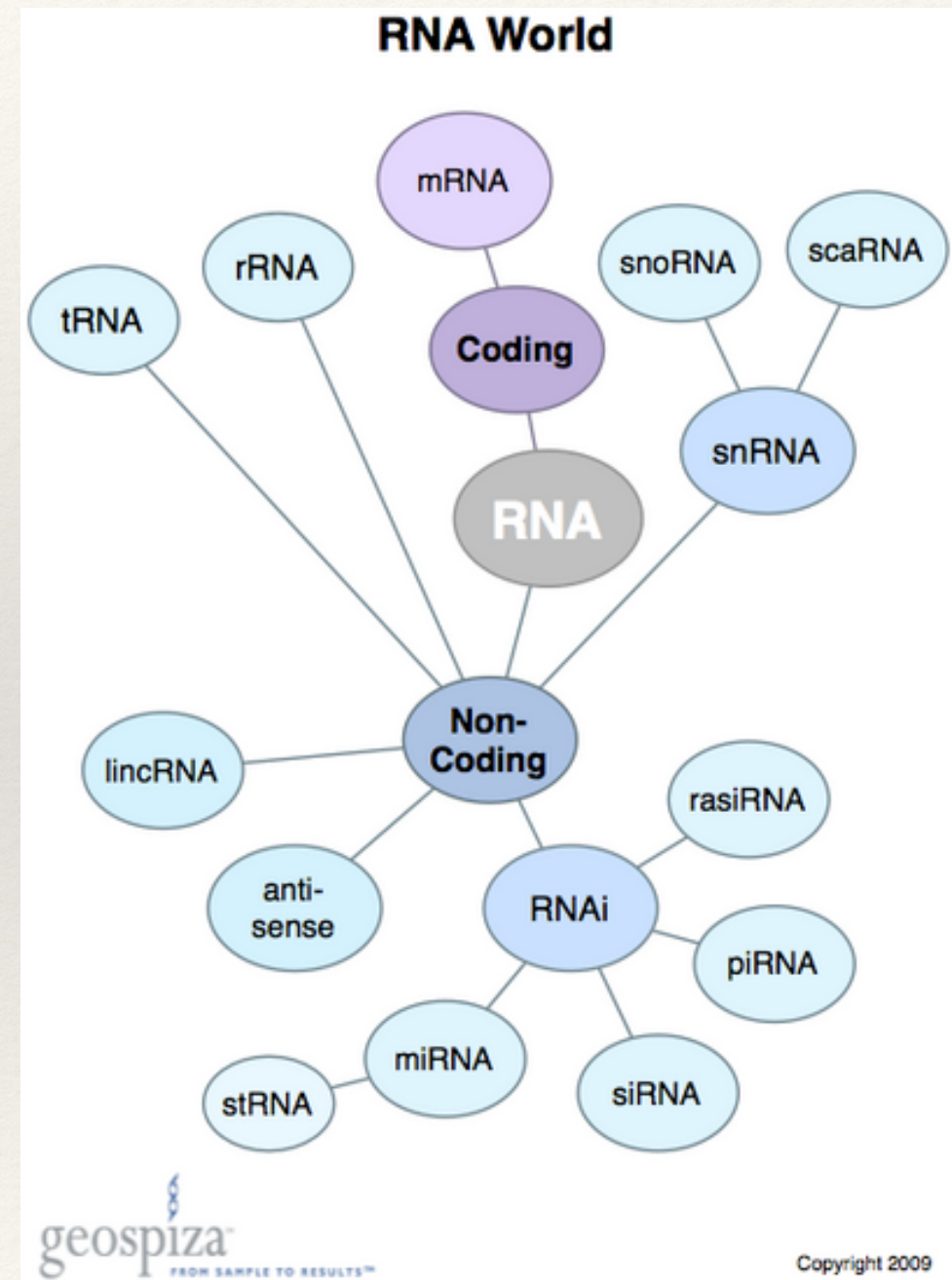
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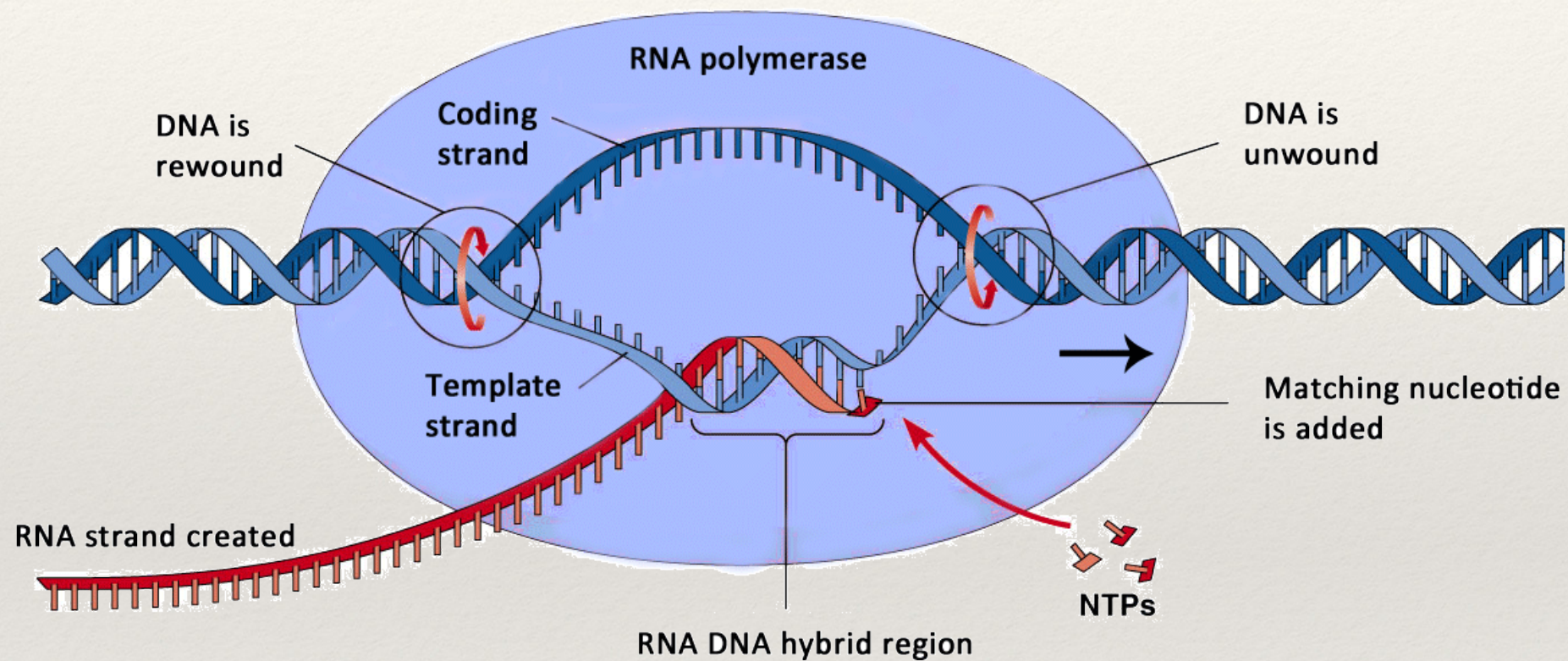
# Transcriptome

A transcriptome is a snapshot in time of all RNAs present in a sample isolated from a given cell, tissue or organism





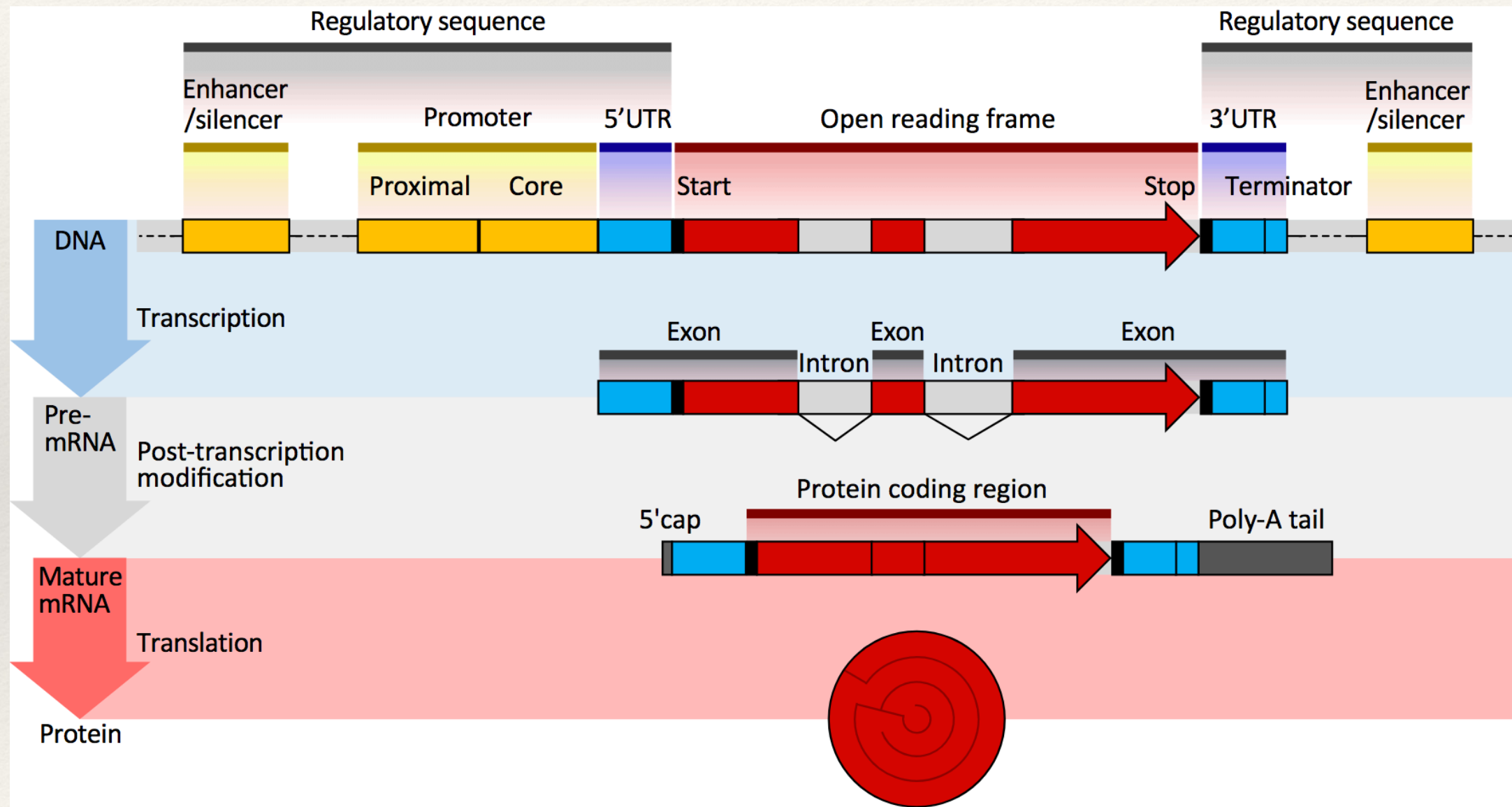
# Transcription



Copying information from DNA to a RNA molecule for regulation or translation to protein

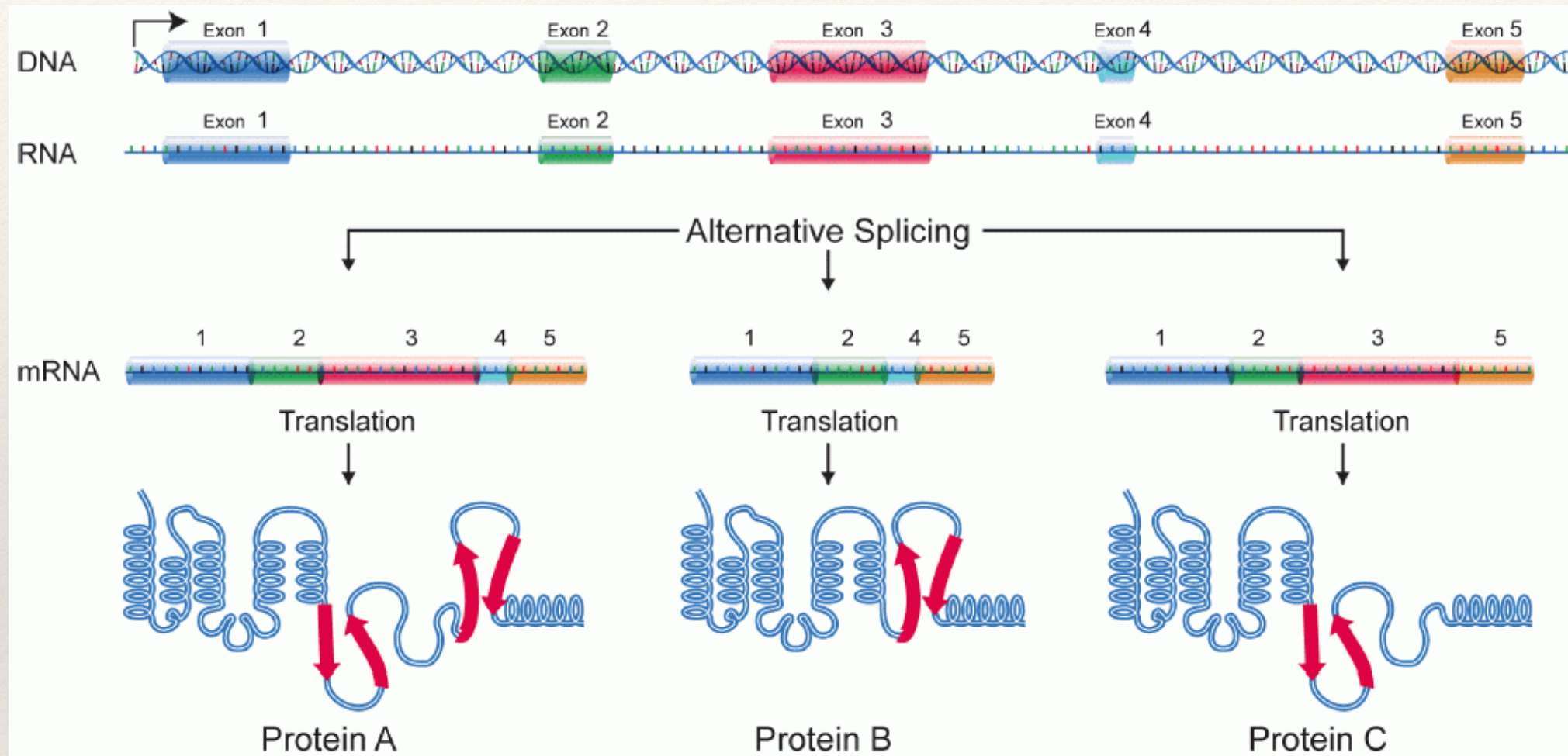


# Eukaryotic mRNA processing





# Eukaryotic mRNA processing



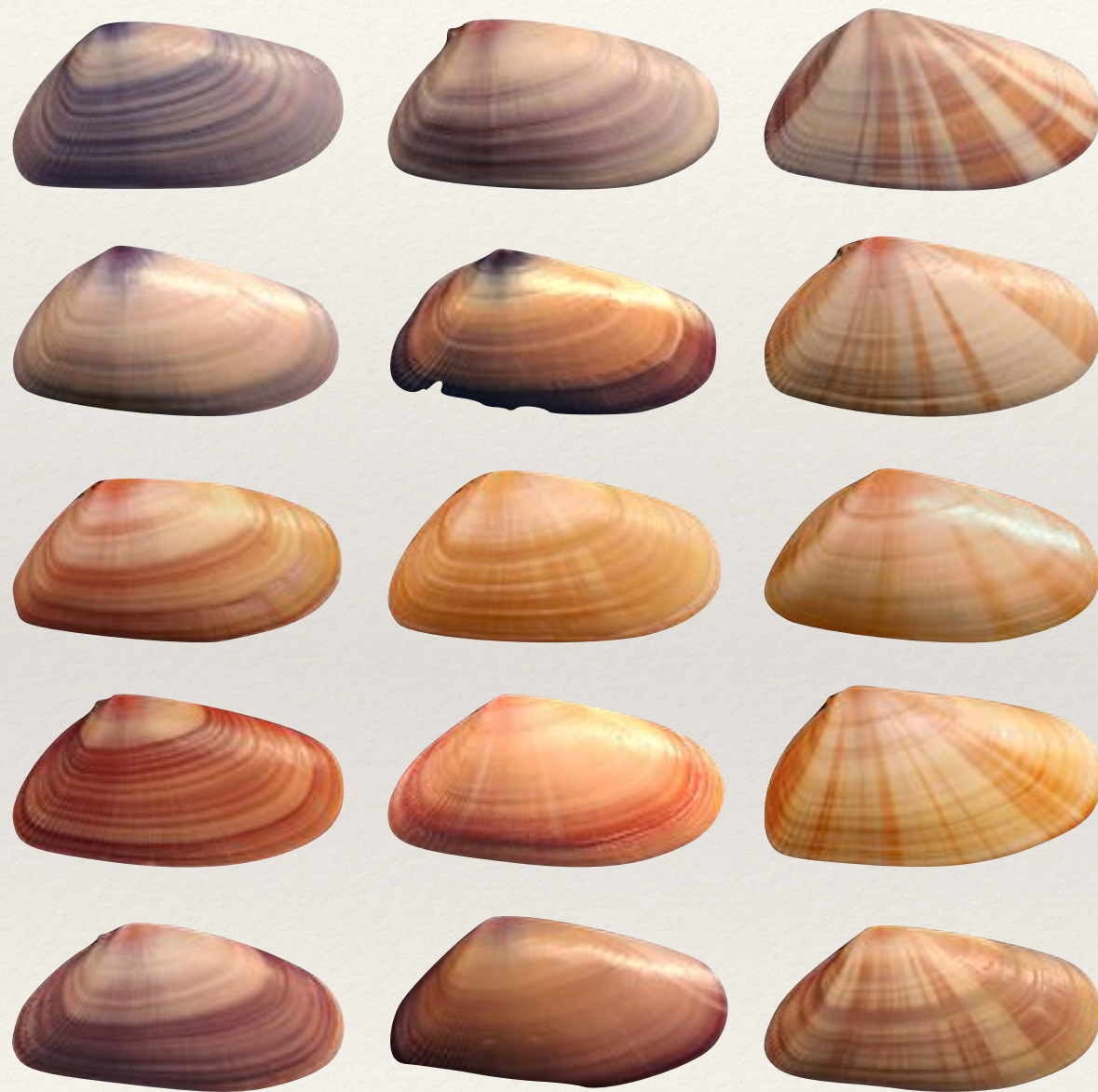
Splicing



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# Transcriptome

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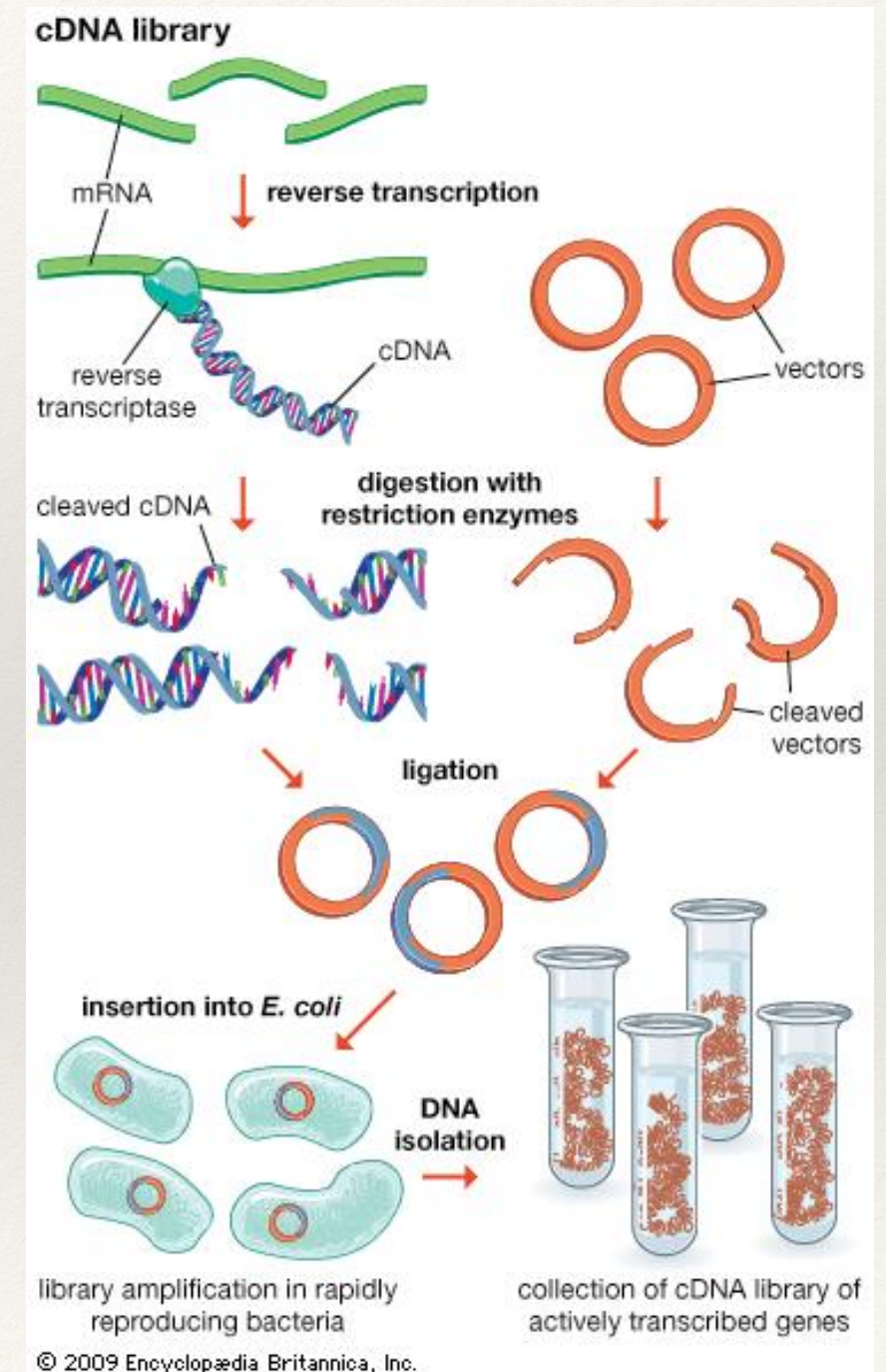
Study individual variation



# Obtaining transcriptome

## ❖ Sanger sequencing

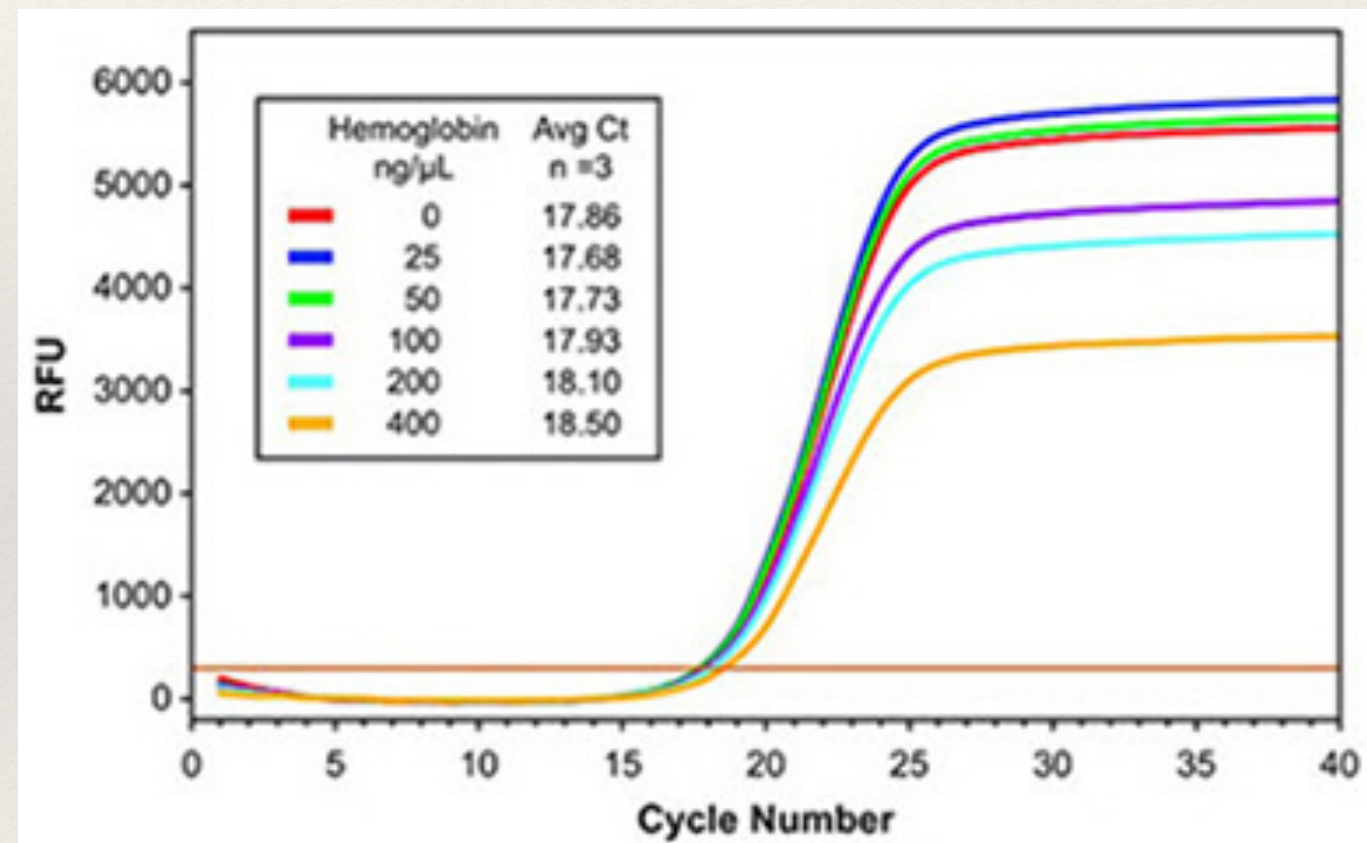
- ❖ mRNA converted to the more stable cDNA
- ❖ cDNA cleaved and ligated into vectors
- ❖ Vectors amplified (cloned) in *E. coli*
- ❖ DNA isolated = cDNA library
- ❖ Sequenced on Sanger
- ❖ Low throughput
- ❖ High accuracy





# Quantifying expression

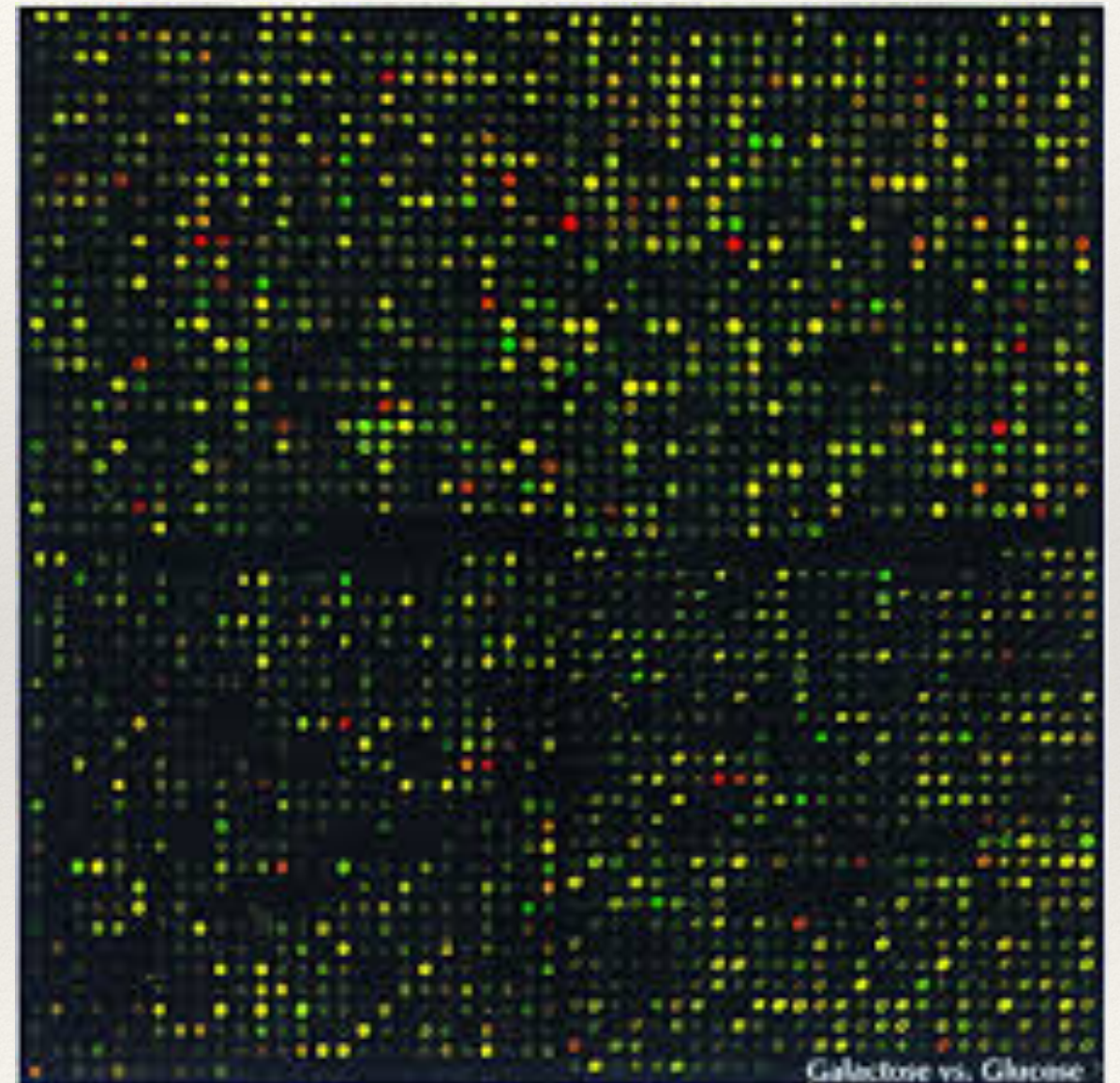
- ❖ Quantitative RT-PCR
  - ❖ qRT-PCR requires knowledge of gene sequence
  - ❖ Hard manual work
  - ❖ Low throughput
  - ❖ Expression level relative to control (house-keeping gene)





# Quantifying expression

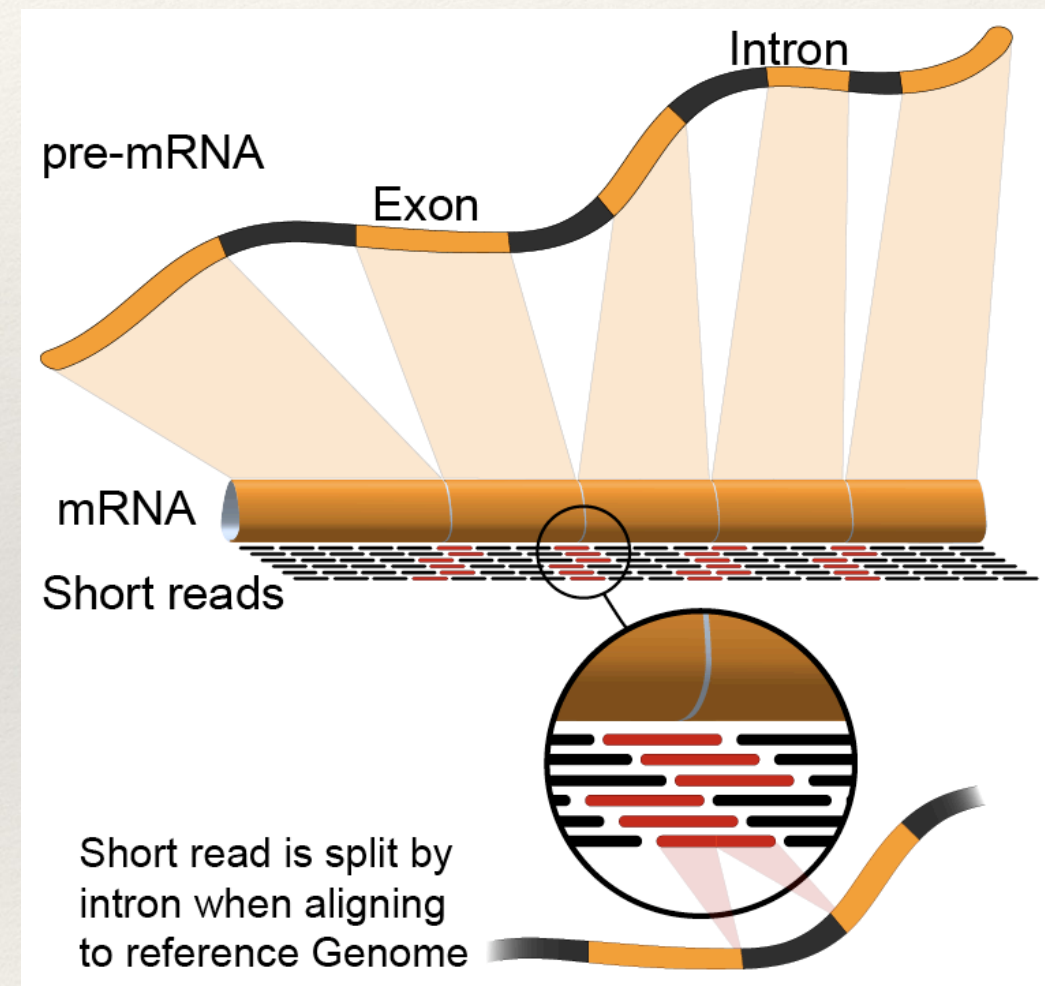
- ❖ Microarray
  - ❖ Requires gene sequences for probe design
  - ❖ High throughput compared to qRT-PCR
  - ❖ Possibility of outsourcing
  - ❖ Expression results relative to all probes





# Quantifying expression

- ❖ RNA-seq
  - ❖ Transcriptome and expression in one go
  - ❖ No need for gene sequence information
  - ❖ High throughput
  - ❖ Can be outsourced
  - ❖ Costly, but effective
  - ❖ Expression results relative to all transcripts





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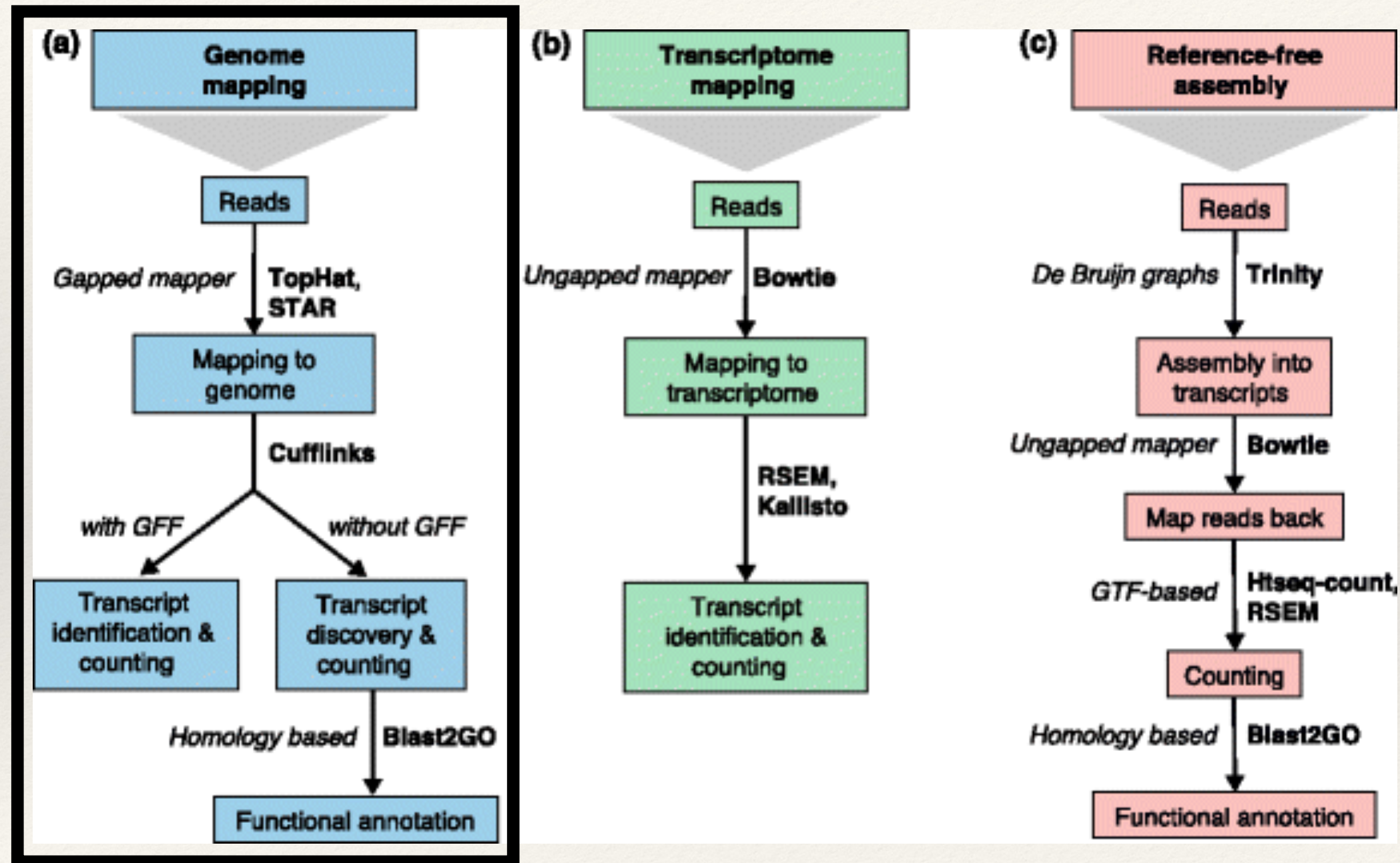
# Sequence data analysis

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- ❖ Is genome available?
- ❖ Well annotated?
- ❖ *De novo* approach
- ❖ Reference based approach
- ❖ Transcriptome
- ❖ Genome+Transcriptome
- ❖ Mixed approach??
- ❖ Short reads (Illumina) + Long reads (PacBio, ONT)



# Mapping sequence data





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# Library prep (Illumina)

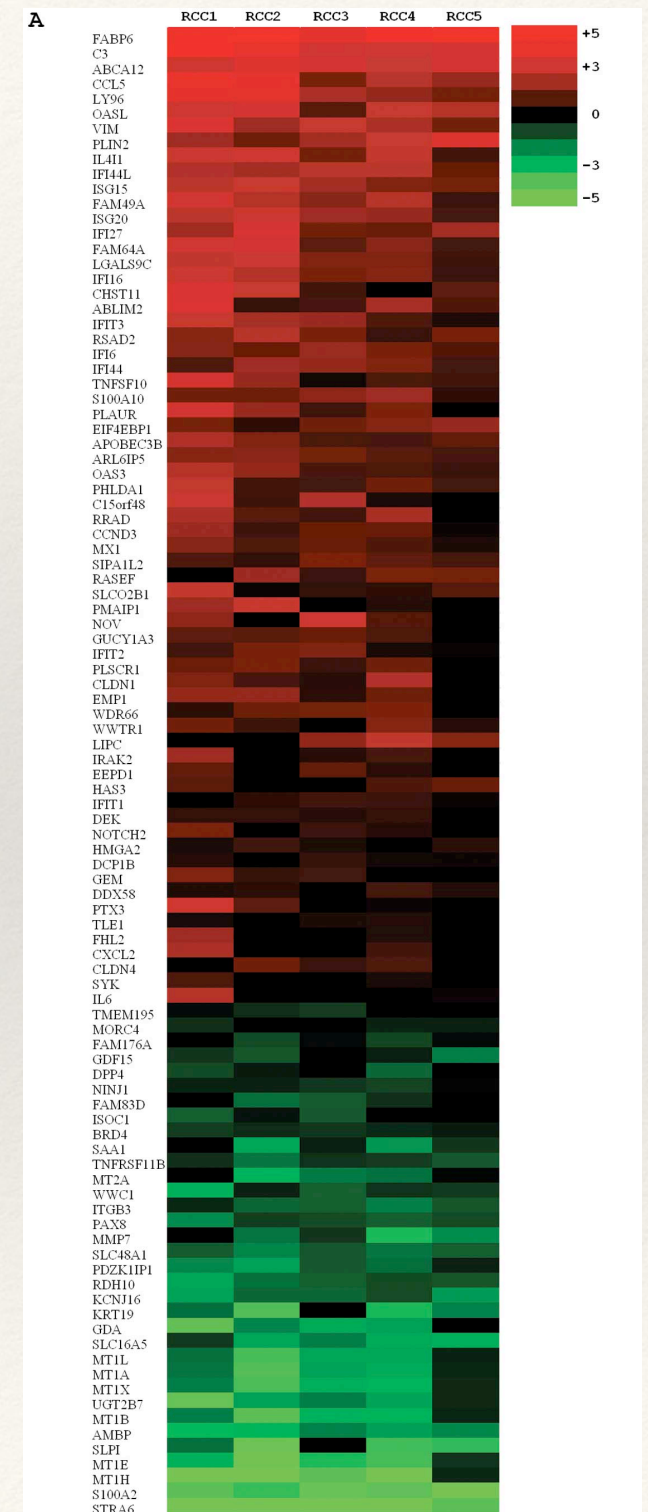
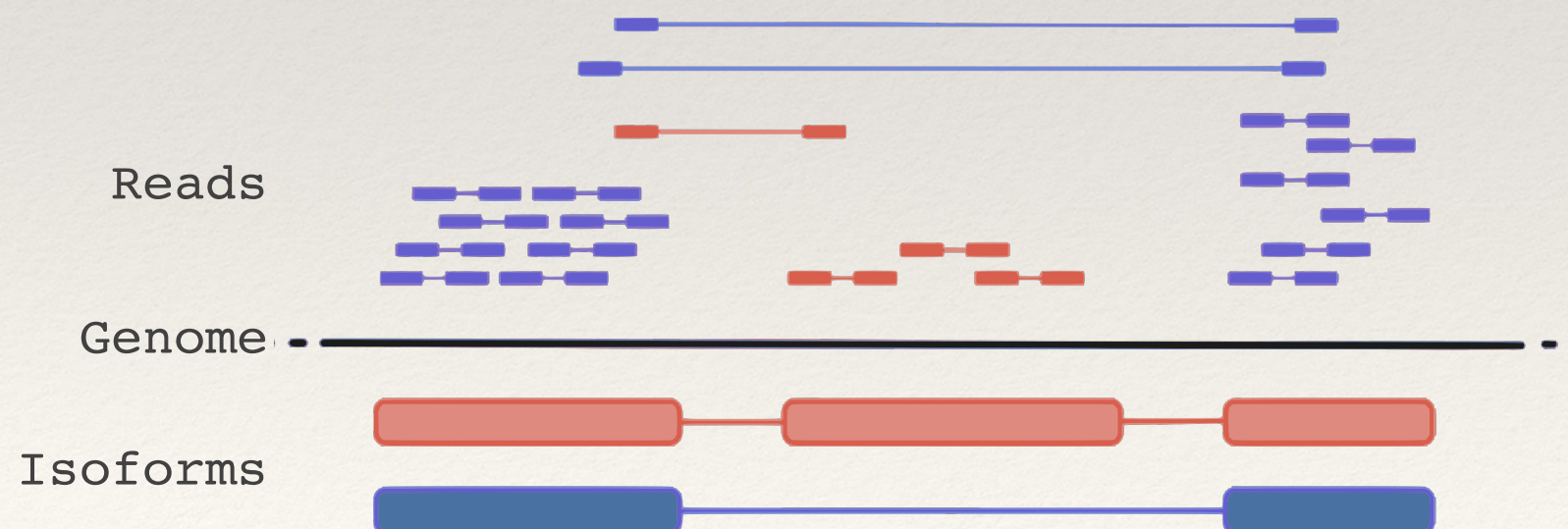
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- ❖ RNA sequencing
  - ❖ Total RNA
    - ❖ TruSeq **Stranded** Total RNA kit
  - ❖ mRNA
    - ❖ TruSeq **Stranded** mRNA kit
  - ❖ small RNA
    - ❖ TruSeq small RNA kit
  - ❖ Ribosome profiling
    - ❖ High quality and quantity of RNA
    - ❖ Do you want to sequence rRNA??



# Depth

- ❖ RNA sequencing
  - ❖ Highly expressed known transcripts
  - ❖ Novel isoforms
  - ❖ Low expressed / rare transcripts





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# Counting

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- ❖ 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.



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# Differential expression

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



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# Normalisation

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- ❖ 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>



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# DESeq2

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- ❖ 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



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# Multiple hypothesis testing and FDR

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## 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  $P$ -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
    - ❖ FDR = 0.05 common for exploratory / broad scope studies
    - ❖ FDR  $< 0.05$  common for medical applications and hunts for candidate genes



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# DESeq2

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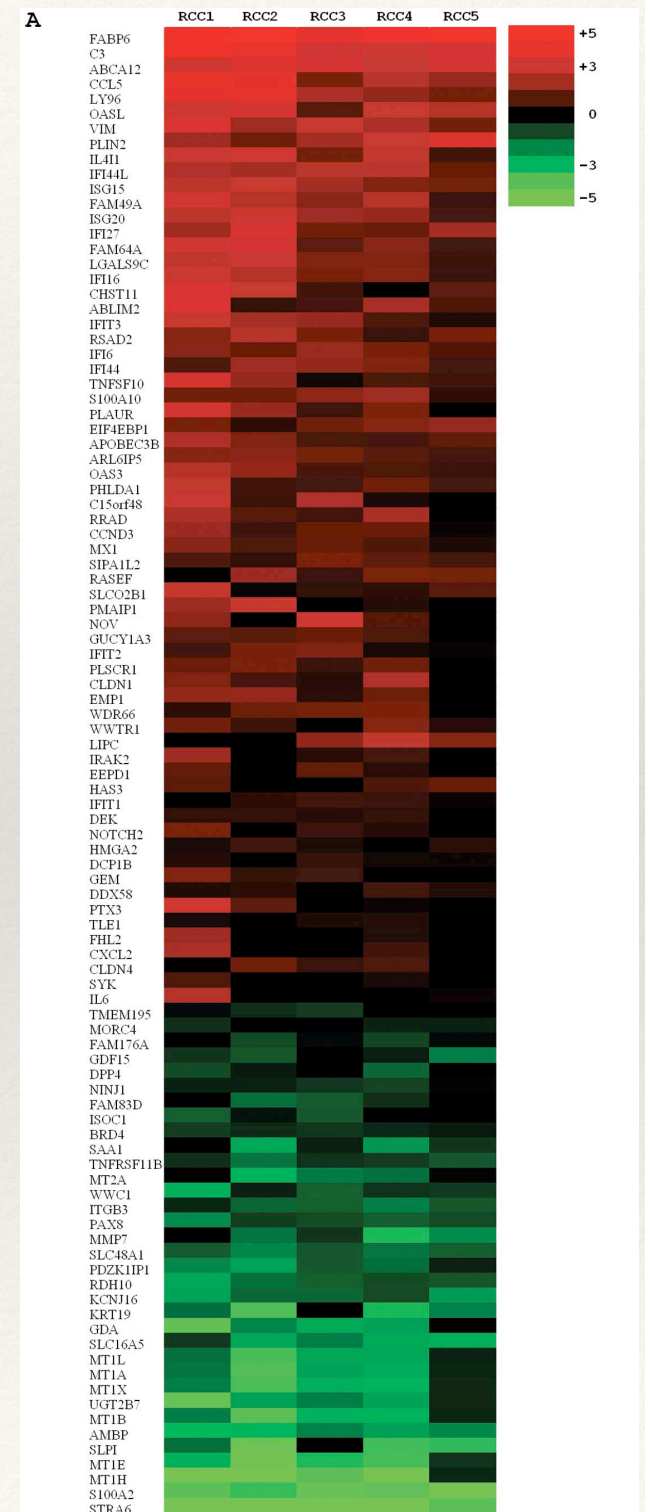
- ❖ `plotDispEsts()`: To look at the dispersion plots
- ❖ `plotPCA()`: To find outliers
- ❖ `plotMA()`: Exploring DE results



# DE list. What next?

## ❖ Heatmap

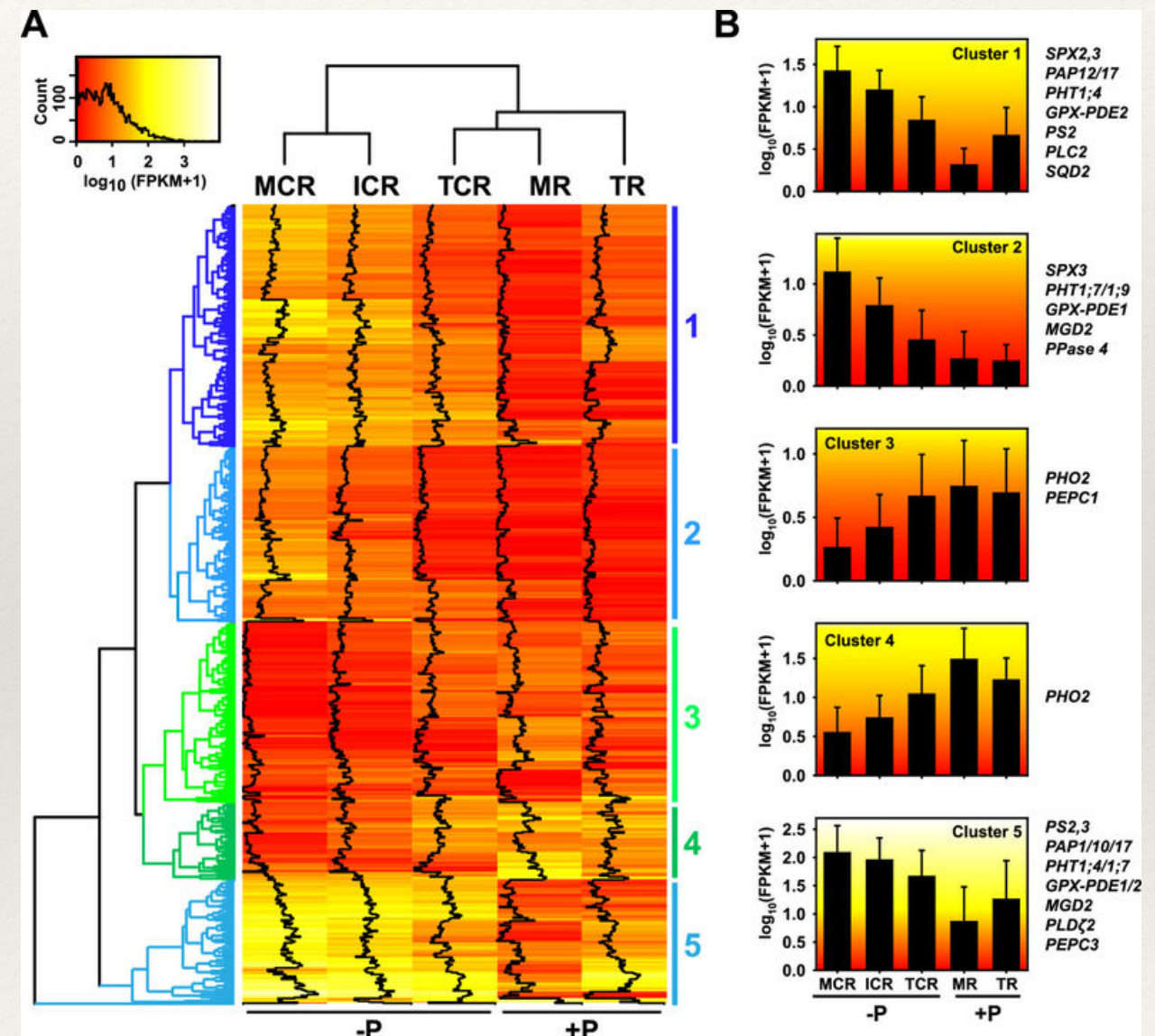
- ❖ Possible to make heatmaps using cummeRbund, DESeq2 and edgeR or in R
- ❖ Tool: MeV TM4
- ❖ Using normalised count information





# DE list. What next?

- ❖ Clustering
  - ❖ Gene (feature level)
  - ❖ Sample level
  - ❖ Hierarchical
  - ❖ CAST: Clustering Affinity Search Technique
  - ❖ Personal favorite - MeV TM4
  - ❖ Possible in R



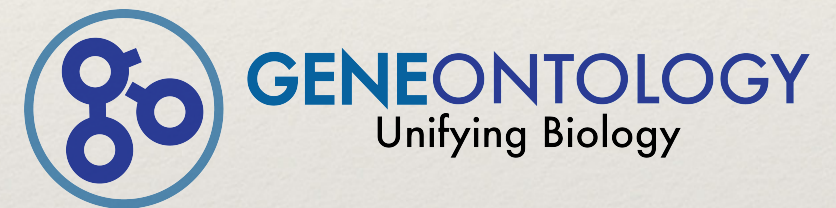


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# DE list. What next?

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- ❖ Functional profiling of gene lists
  - ❖ Gene Ontology (GO) enrichment analysis
    - ❖ Biological process
    - ❖ Cellular components
    - ❖ Molecular function
  - ❖ KEGG pathway enrichment analysis



- ❖ Tools
  - ❖ GOrilla (<http://cbl-gorilla.cs.technion.ac.il/>)
  - ❖ Comprehensive tool - g:Profiler (<http://biit.cs.ut.ee/gprofiler/>)