

DTU



# A Short Introduction to Transcriptomics

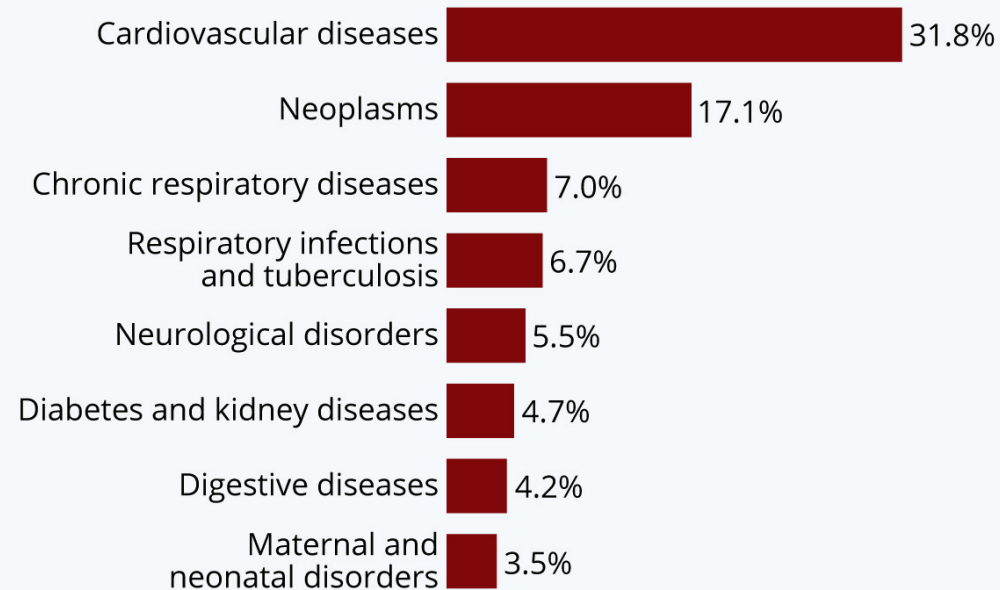
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Assistant Professor in Bioinformatics

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## Top Global Causes of Death

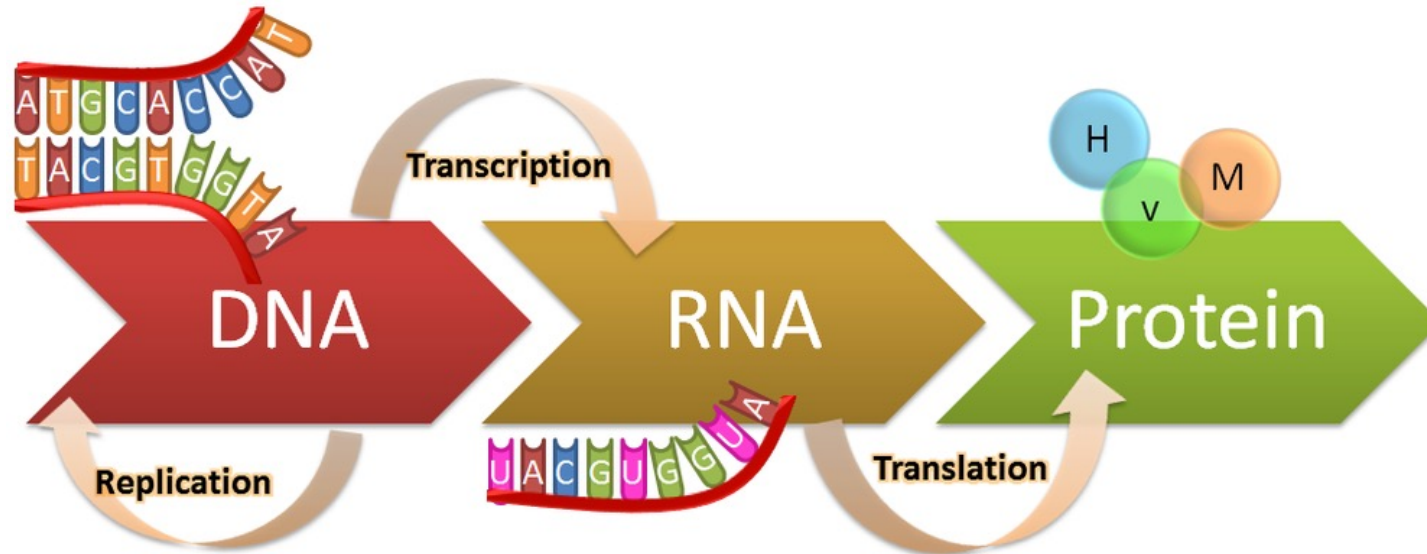
Share of all global deaths in 2017,  
by most common causes



Source: World Economic Forum / Institute for Health Metrics and Evaluation

To treat most of these we need to understand the molecular mechanisms and how they are change by disease

Aim: Profile difference between healthy and sick



Solution: Measure all DNA, RNA or all protein in healthy and sick

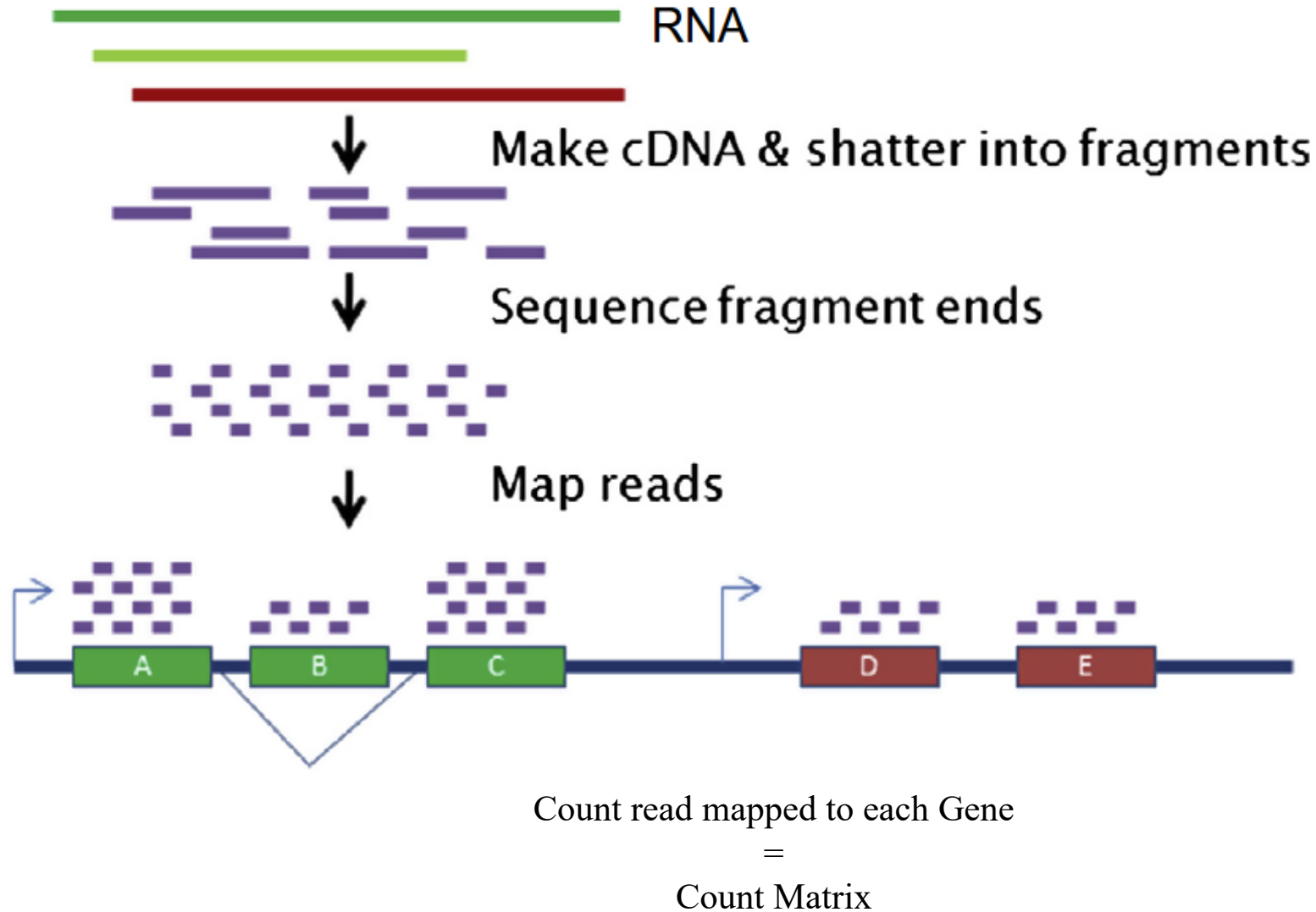
Genomics

**Transcriptomics**

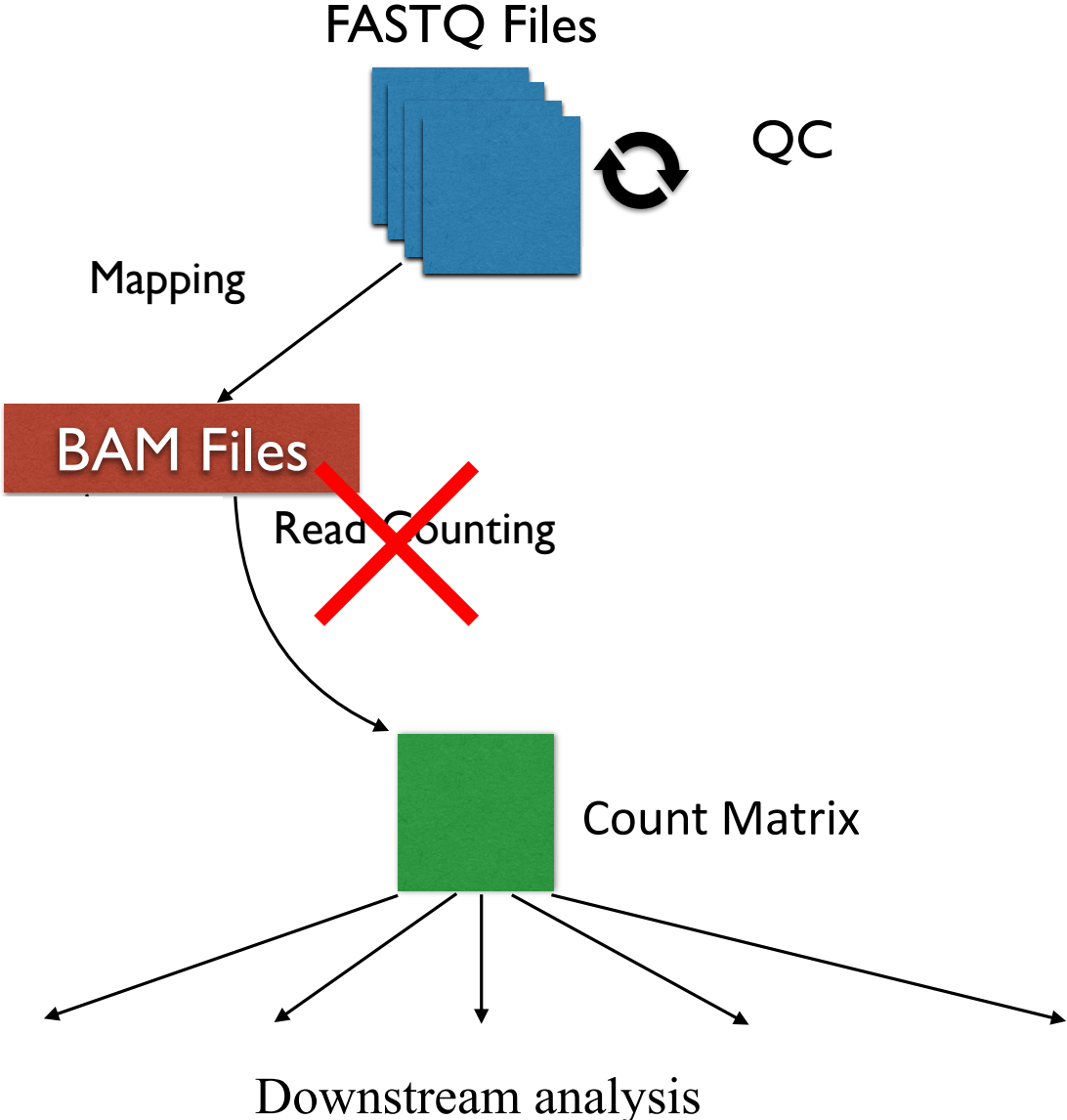
Proteomics

# RNA-sequencing 101

# RNA-sequencing

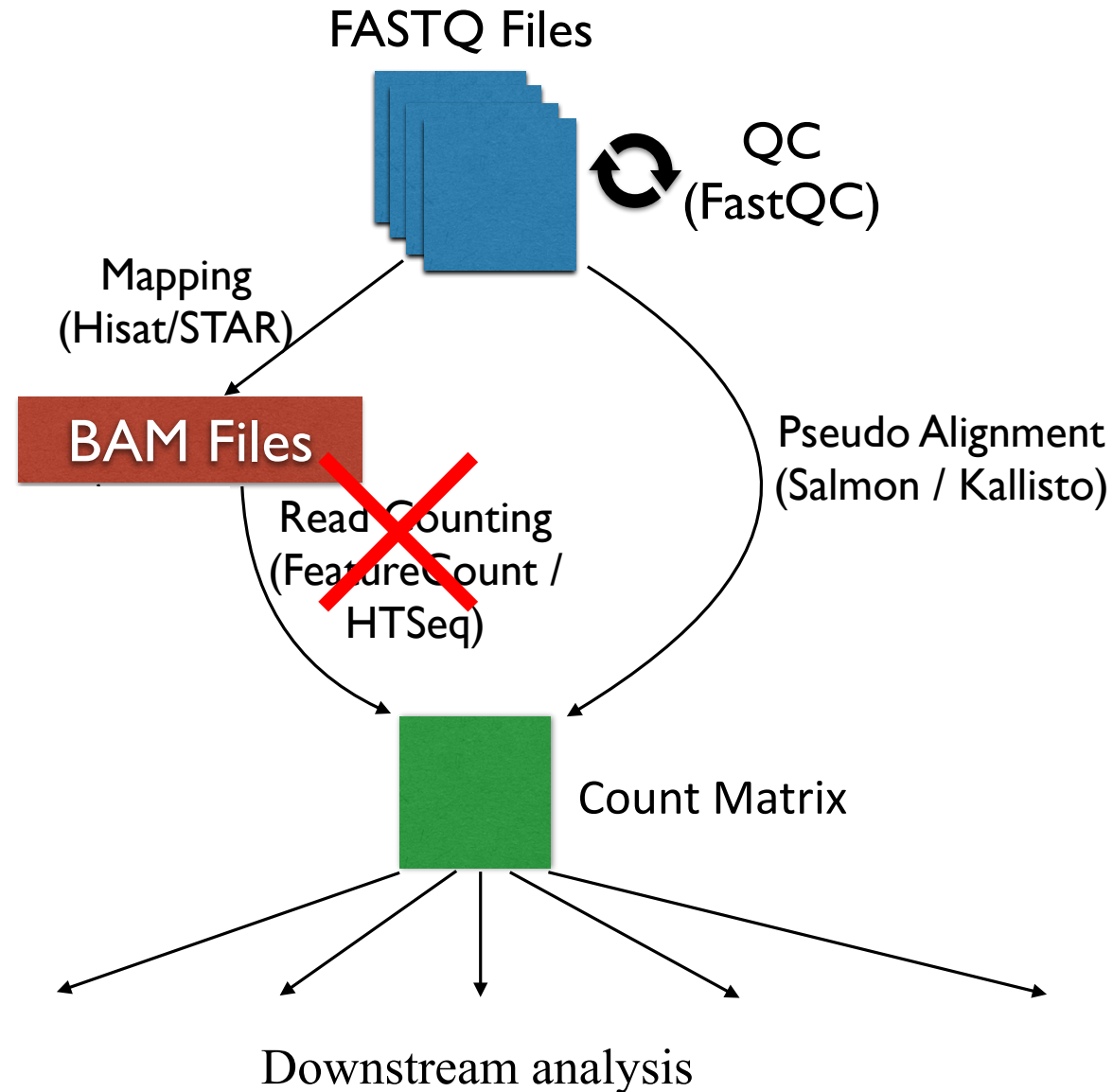


# More detailed workflow



# More detailed workflow

(Tool name)



# Gene Count Matrix

The results from a single sample RNA-seq datasets:

	sample1	
Identifier	DDR1	884
	RFC2	422
	HSPA6	621
	PAX8	658
	GUCA1A	426
	UBA7	524
	THRA	564
	PTPN21	909
	CCL5	771
	CYP2E1	315
	EPHB3	362
	ESRRA	911
	CYP2A6	409
	GAS6	368
	MMP14	3
	TRADD	102
	FNTB	368
	PLD1	661

← Read count of each “feature”

↓ Typically, 20,000 – 50,000 genes



# Gene Count Matrix

Samples →

Genes ↓

	sample1	sample2	sample3	sample4	sample5
DDR1	884	81	622	946	541
RFC2	422	858	305	830	28
HSPA6	621	130	401	221	110
PAX8	658	176	573	999	296
GUCA1A	426	171	723	659	64
UBA7	524	579	19	199	510
THRA	564	110	119	384	488
PTPN21	909	668	810	781	852
CCL5	771	838	217	837	816
CYP2E1	315	978	289	718	583
EPHB3	362	488	359	850	558
ESRRA	911	724	41	476	264
CYP2A6	409	683	286	998	690
GAS6	368	222	212	642	747
MMP14	3	833	878	552	511
TRADD	102	479	844	327	43
FNTB	368	284	591	729	747
PLD1	661	489	706	859	587

# RNA-sequencing Count Matrix

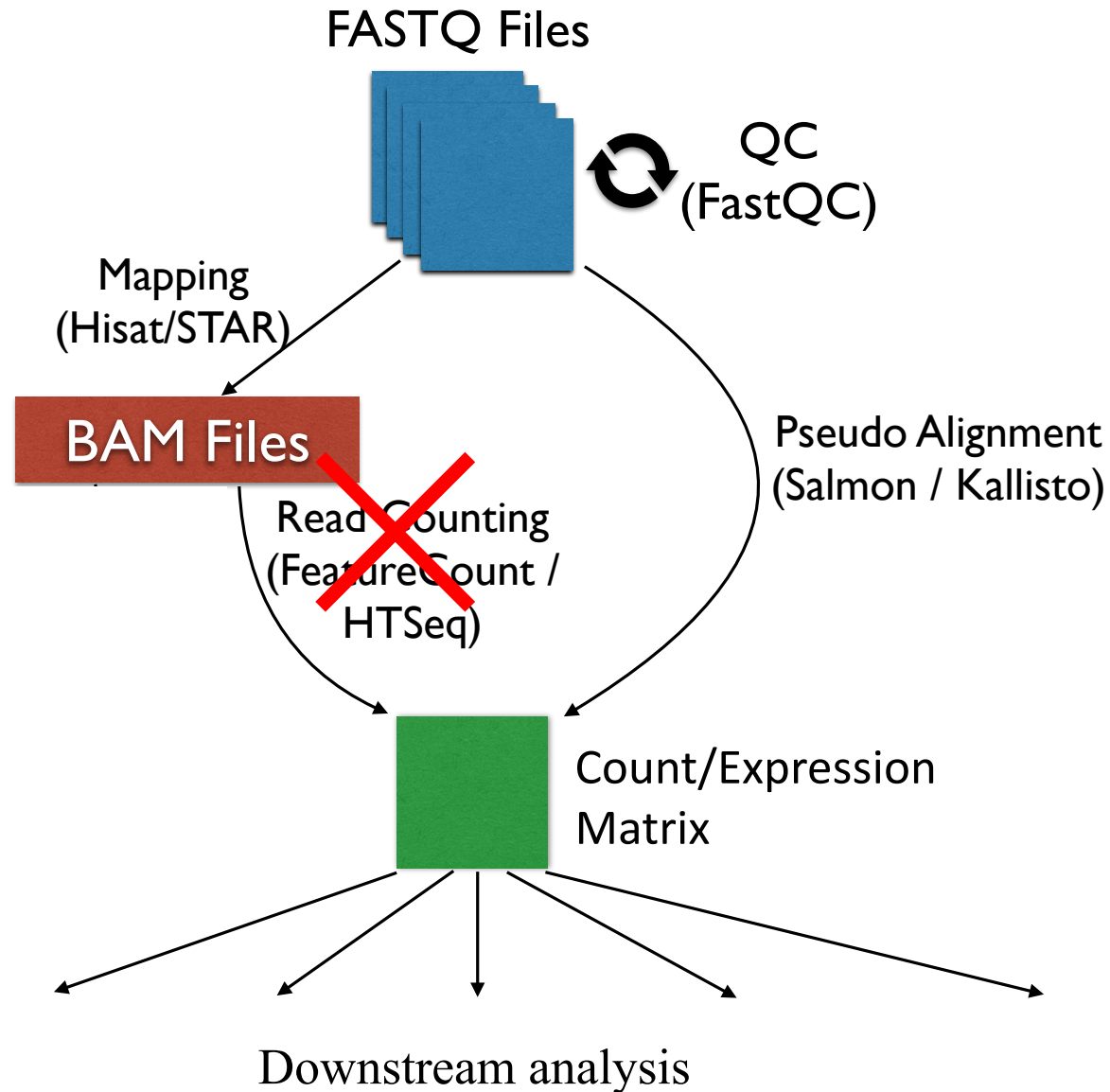
- 3 minutes with neighbour:
- You are analysing 2 genes (gene A and B) in two conditions (condition 1 and 2) on the basis of a RNA-seq experiment that resulted the following number of reads:

	Condition 1	Condition 2
Gene A	1000	3000
Gene B	2000	4000

- Question: Is the following statement correct:  
“Both gene A and B are more expressed in condition 2”  
Explain why/why not.

# More detailed workflow

(Tool name)



# Downstream Analysis

Two of the hundreds of possible uses

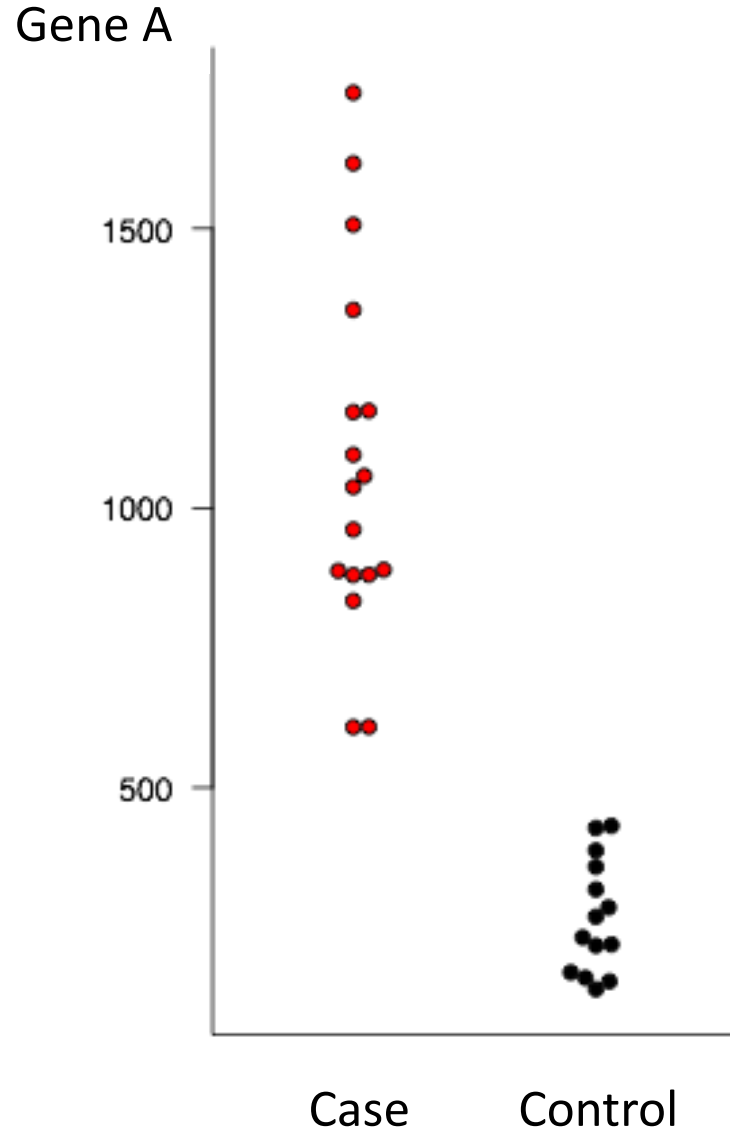
# What question do you want to answer?

Typically, we use transcriptomics to compare between two or more groups, generally referred to as a case/control study.

Examples include:

- Disease vs. normal
- Drug treatment vs. control
- Good prognosis vs. bad prognosis
- Timepoint 1 vs timepoint 2

# Case/Control Study



2 Min with neighbor:

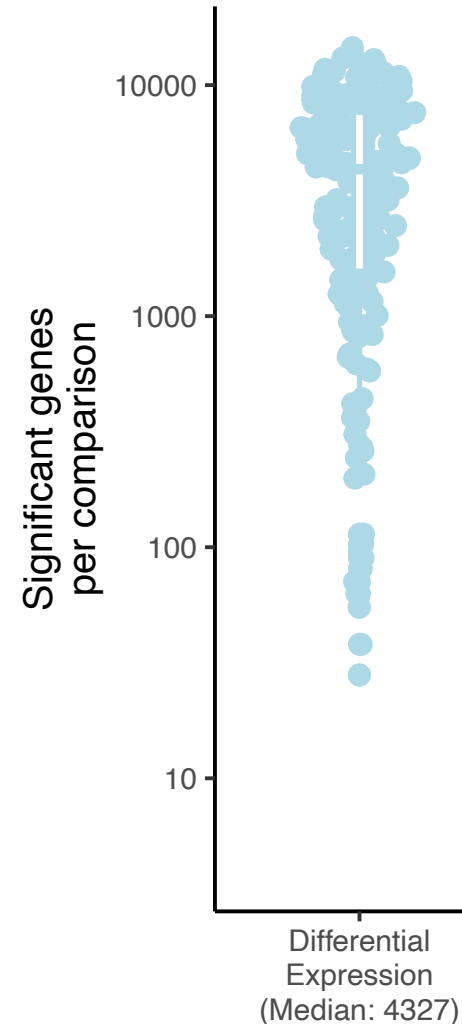
- What would we like to know/summarize?

# Differential Expression Analysis

- Done by advanced well-tested bioinformatics tools
  - DESeq2
  - EdgeR
  - Voom-limma
- Count matrix as input (do normalization internally)
- They use (generalized) linear models
  - Can take unwanted effects into account

# In the real world

- We recently did a systematic analysis of 100 RNA-seq datasets
- On average thousands of genes change significantly between conditions (!)
- How do you make sense of such a list?





# Gene-sets

- Collection of genes that have something in common
  - Participate in the same process (e.g. cell cycle)
  - Have the same molecular function (e.g. DNA binding)
  - Cellular location (e.g. nucleus)
  - Identified by Kristoffer (e.g. what I just found in my data)
- Many databases
  - Gene Oncology (GO-terms) (<http://geneontology.org/>)
  - MSigDB (<https://www.gsea-msigdb.org/gsea/msigdb/>)
  - Enrichr (<https://maayanlab.cloud/Enrichr/#libraries>)

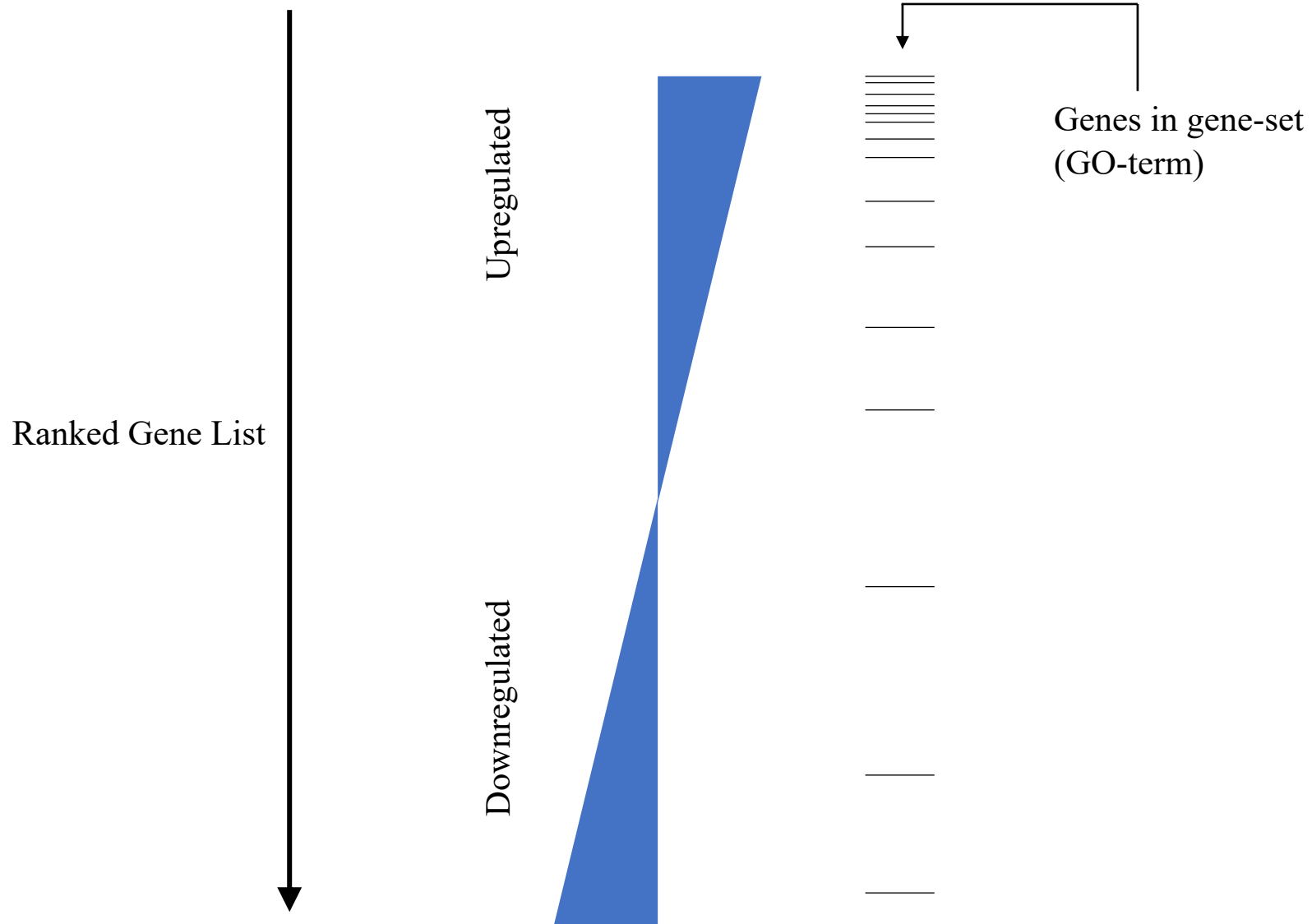
# Gene Set Enrichment Analysis

Gene Set Enrichment Analysis (GSEA) can be done in two ways:

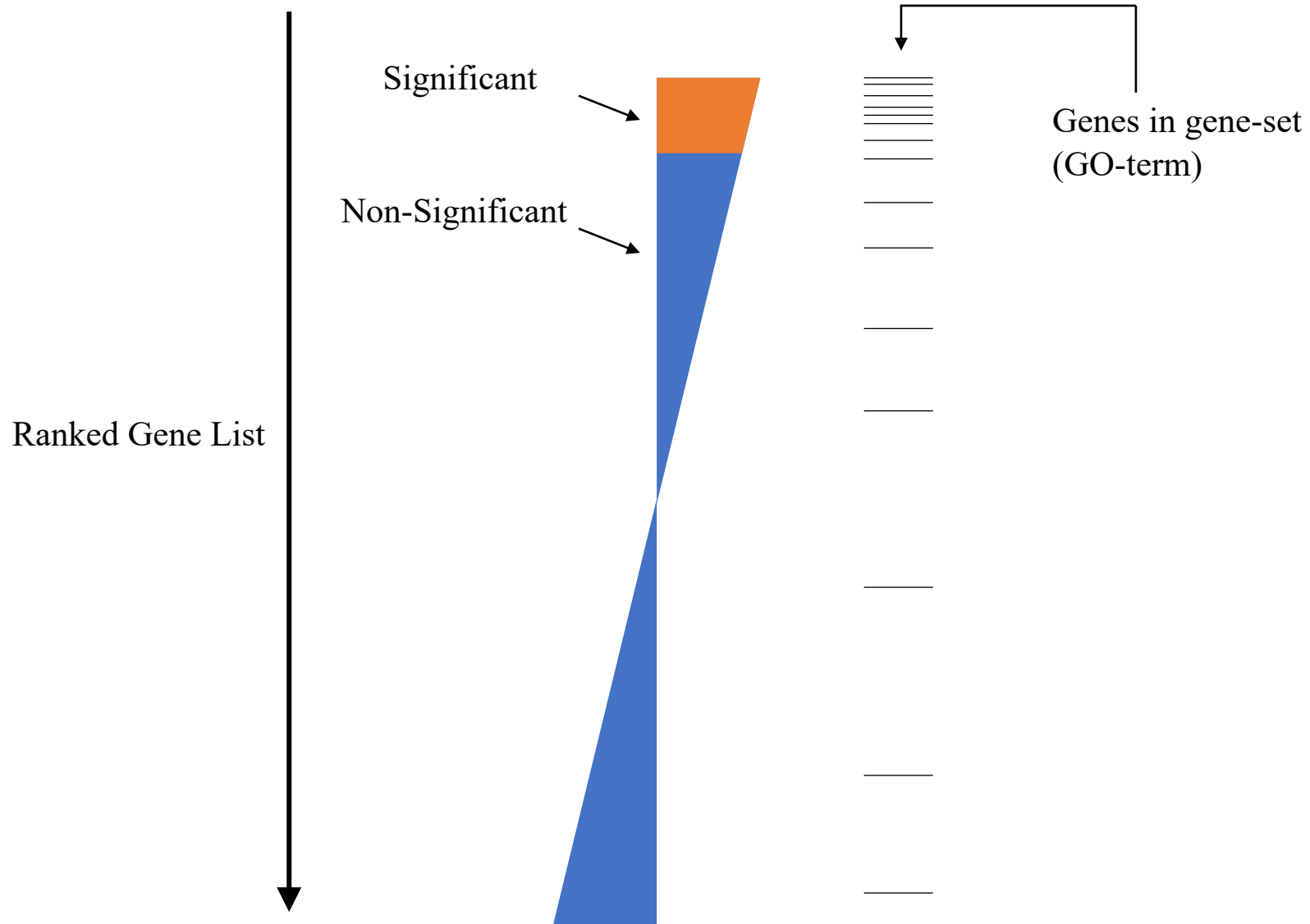
1. Overrepresentation Analysis (OA/OR<sub>A</sub>)
2. Functional Class Scoring (FSC)

Confusingly these are both referred to as GSEA

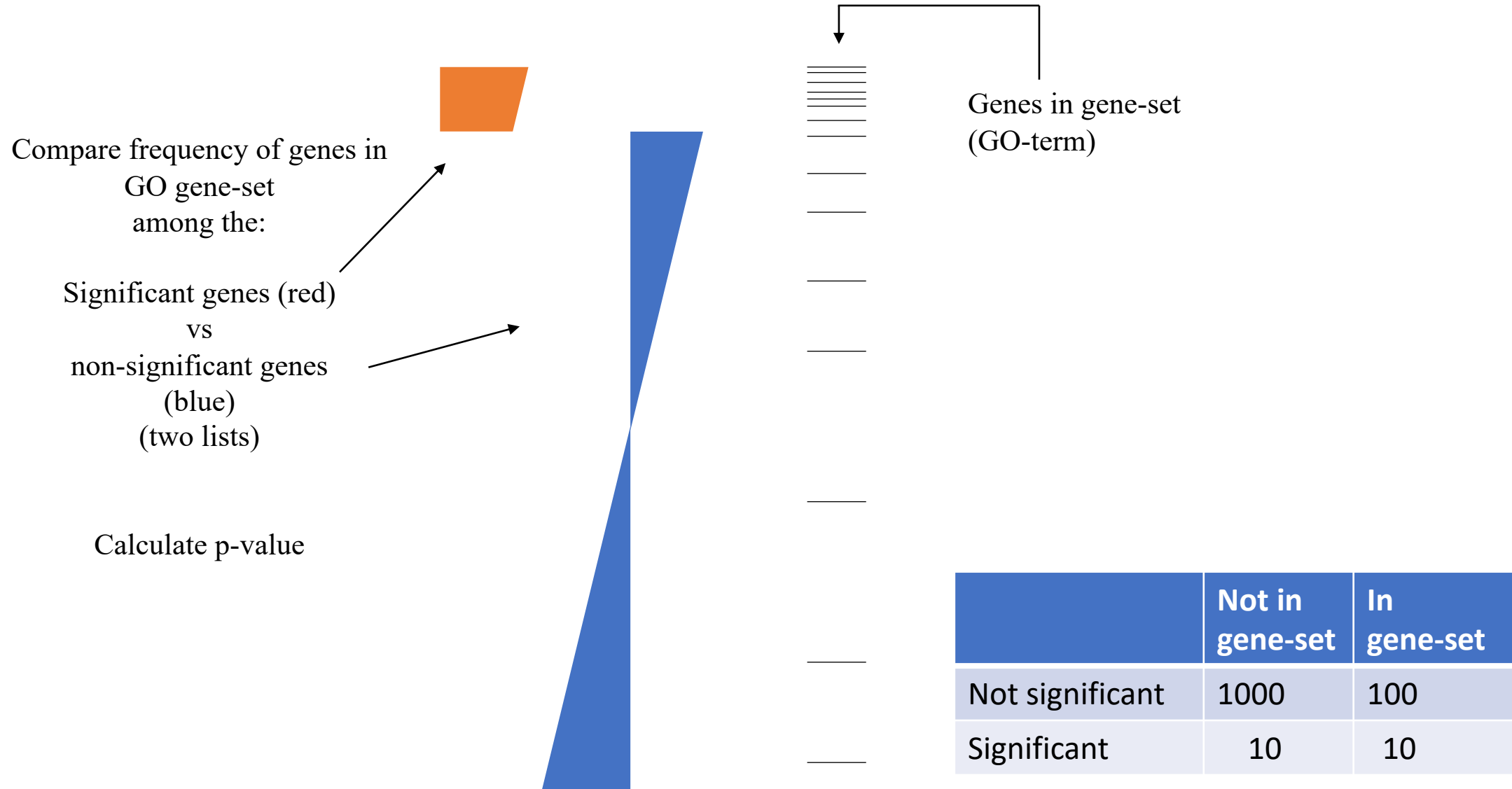
# Overrepresentation Analysis



# Overrepresentation Analysis



# Overrepresentation Analysis



# Note on Overrepresentation Analysis

Can be use for anything where you can divide observations into 4 groups based on two binary categories

- Are significant genes enriched for genes in a gene-set?
- Are people with horn-rimmed glasses also typically taller than 2m?
- Are people with biking helmets enriched amongst students at DTU?
- Etc...

# GSEA

- Literally hundreds of tools for doing it!
- R packages
  - [fgsea](#)
  - [clusterProfiler](#)
  - [limma](#)
  - [gProfiler](#)
  - [pairedGSEA](#)
- Web tools
  - <http://geneontology.org/>
  - <http://cbl-gorilla.cs.technion.ac.il/>
  - <https://david.ncifcrf.gov/>
  - <https://biit.cs.ut.ee/gprofiler/gost>
- Pay attention to your background!

RNA-sequencing 101 – Done!

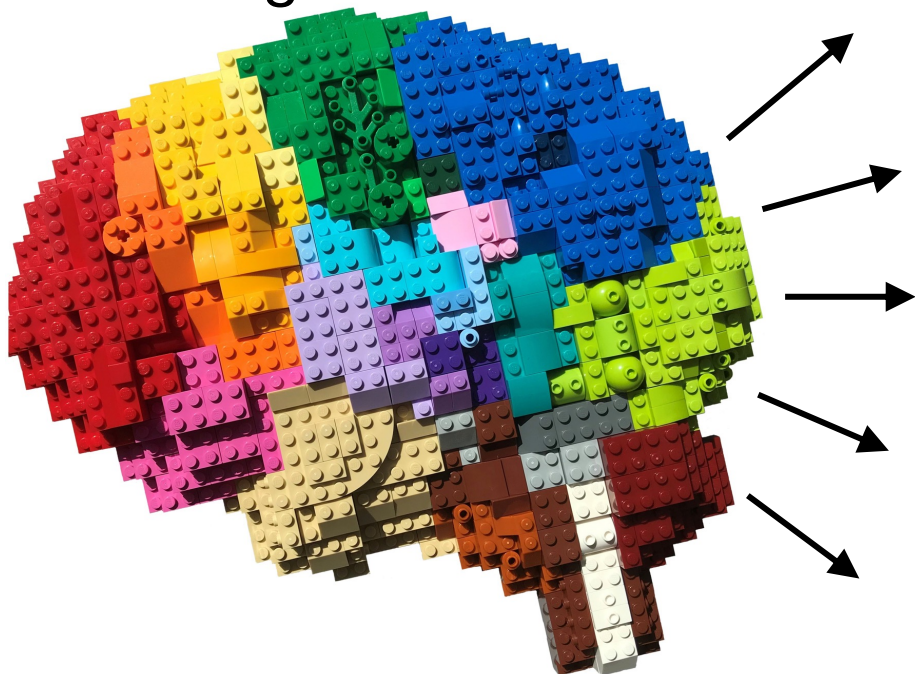


# Extention #1

Bulk vs Single cell vs Spatial

# Transcriptomics

Organ/Tissue



Bulk RNA-seq



Single Cell RNA-seq



# Bulk RNA-seq

- 2 Min with neighbors:  
What potential problems are there with just measuring the average signal (compared to single cell)?

Bulk RNA-seq

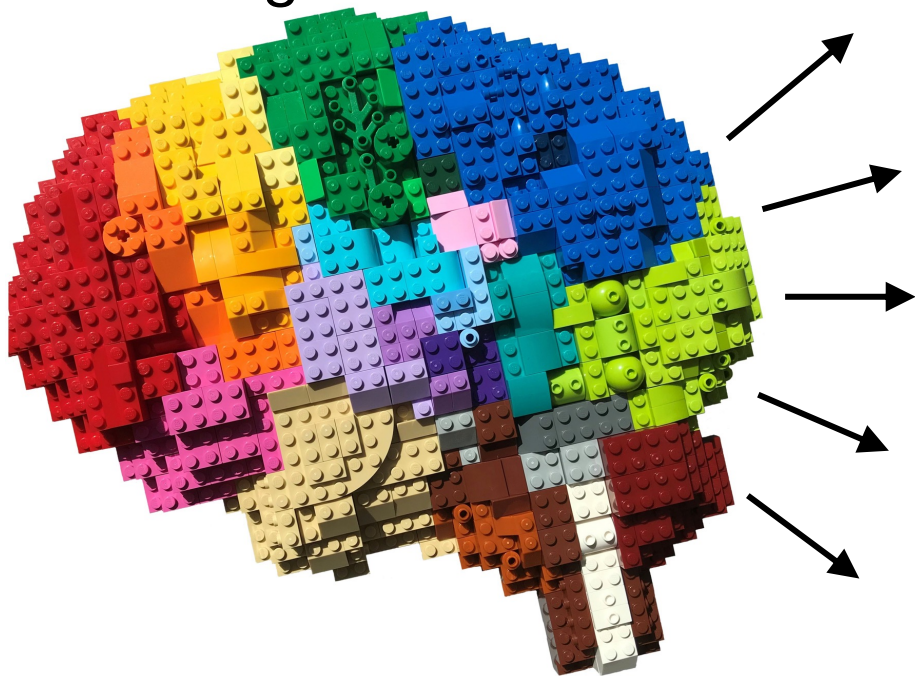


Single Cell RNA-seq



# Transcriptomics

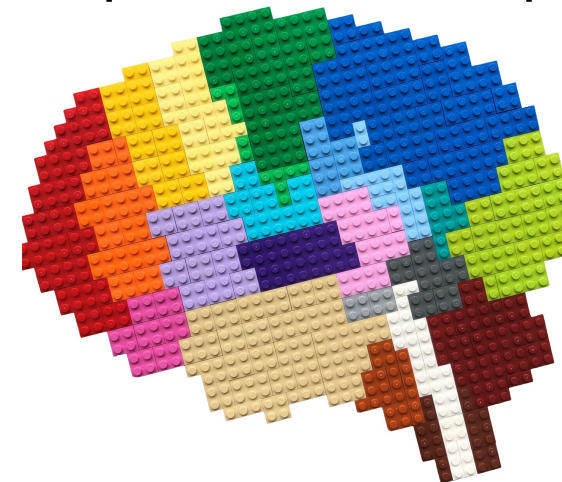
Organ/Tissue



Bulk RNA-seq



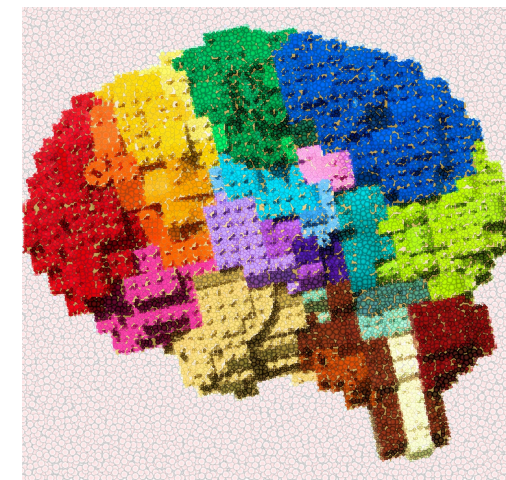
Spatial RNA-seq



Single Cell RNA-seq

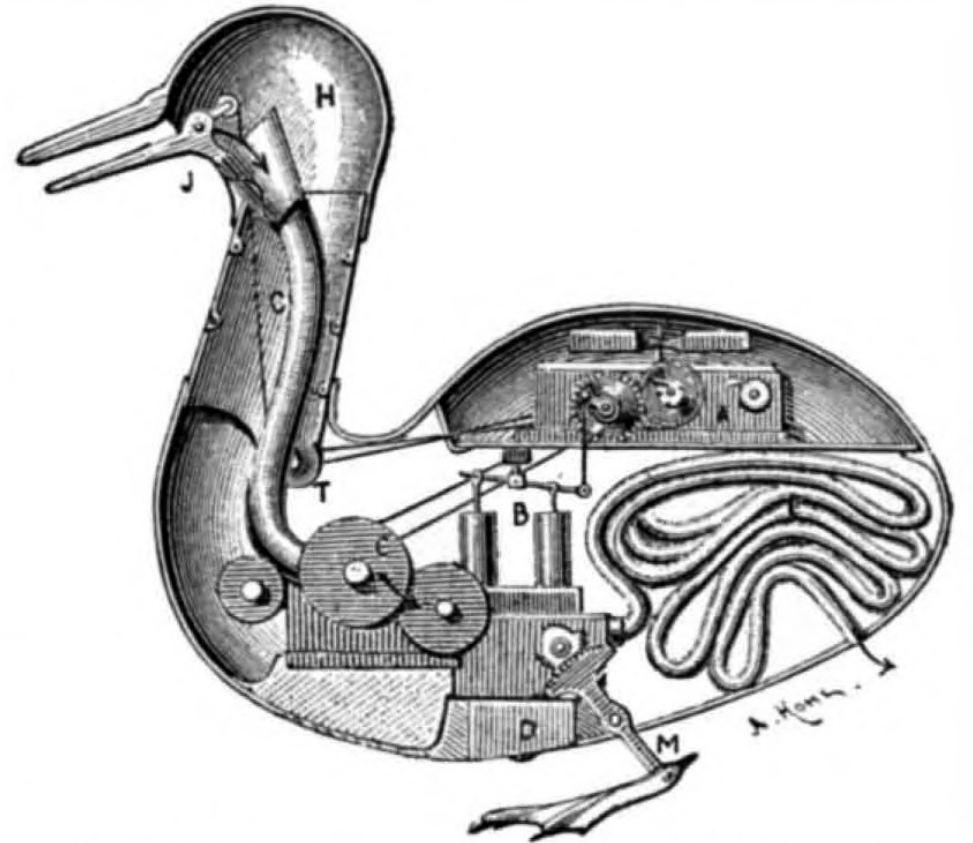


(SC) Spatial RNA-seq



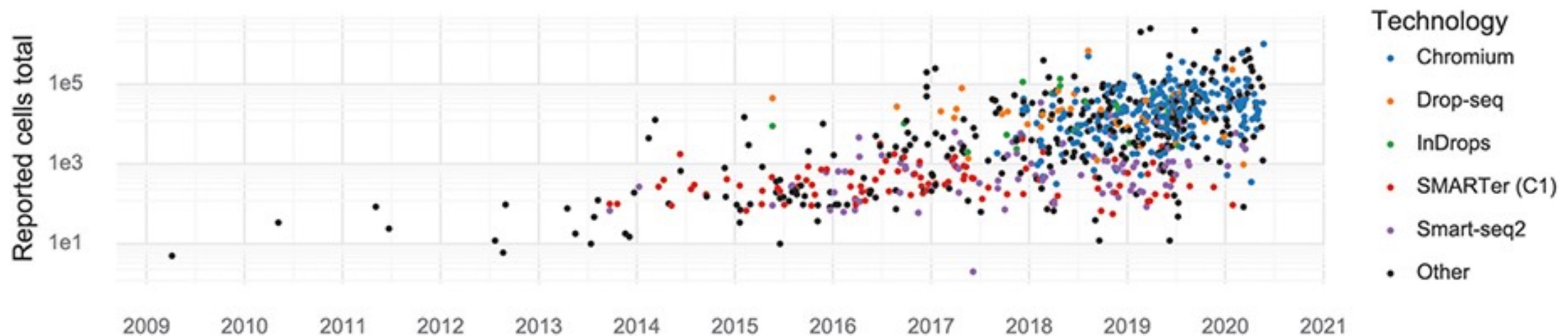
# Bulk RNA-seq

- The workhorse for the last almost 20 years
- Cannot be understated how important it has been!
- Have pushed the reductionistic → integrative (holistic?) research paradigm
- Will continue to be relevant due to limitations of single-cell / spatial



# (Spatial) Single Cell RNA-sequencing

- What everybody wants to do!
- Enables exciting insights into both normal and disease states
- Many limitations including much harder to analyze(!)
  - Only the 300-5000 highest expressed genes (fewer for spatial)
  - Very labor intensive and expensive!
- Is the subject of a new advanced master course I am making



# Single Cell Multi Omics

Joint analysis of multiple modalities (DNA, RNA, etc)  
pushes us towards a holistic research paradigm

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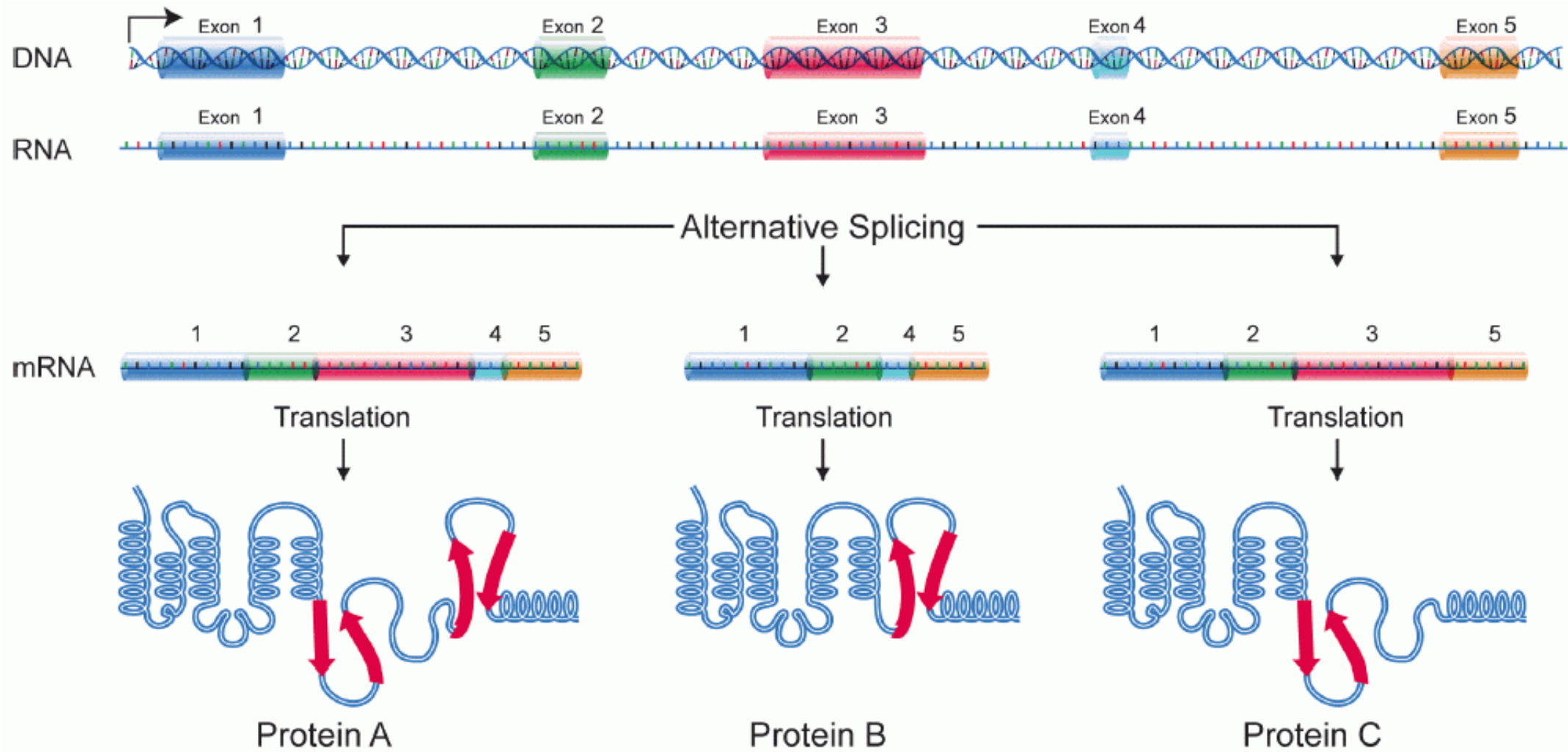
## Method of the Year 2019: Single-cell multimodal omics

# Extention #2

There is no such thing as a “gene”

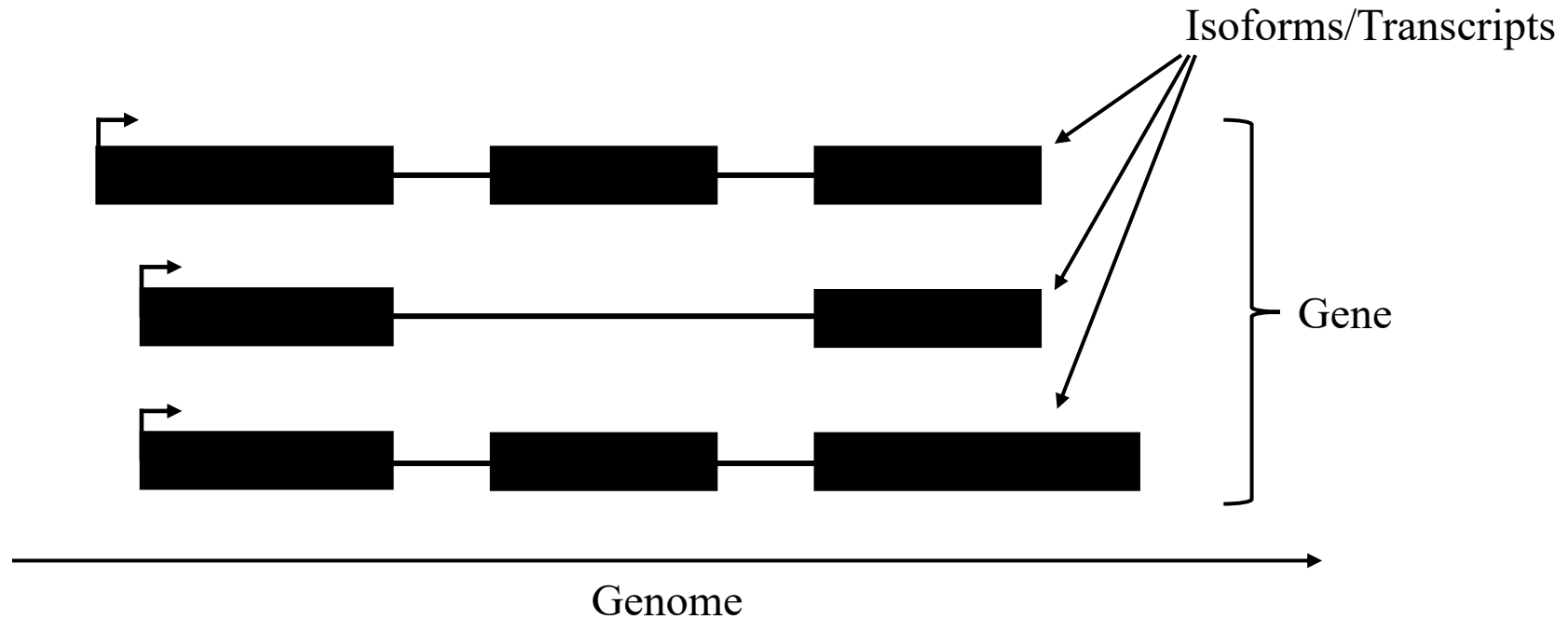


# Alternative Splicing



# Genes vs Isoforms

The terms isoforms and transcripts are (unfortunately) used interchangeably



Isoforms/Transcripts can easily be quantified from RNA-seq data via pseudo-alignment

# Genes vs Isoforms

Take 2 min with your neighbor and discuss:

What would you gain by profiling the transcriptome with isoform resolution (instead of gene resolution)?

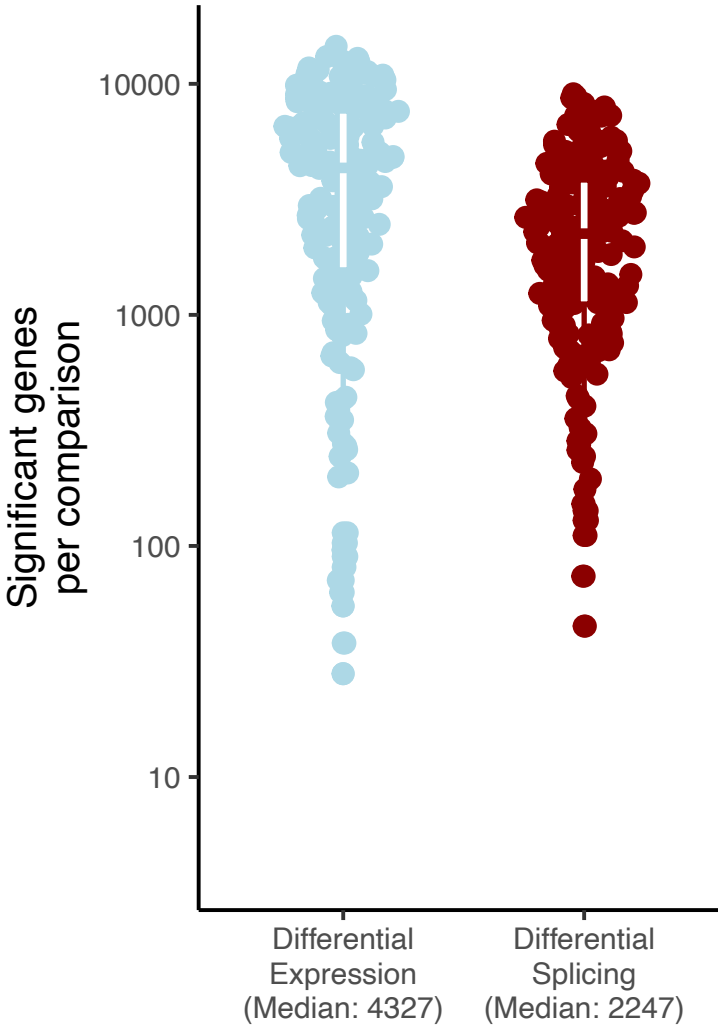
# Analysis of isoforms

- Mostly called differential transcript usage (DTU)
- Good tools:
  - DEXSeq (in family with DESeq2)
  - limma
  - satuRn
- Can be done at two levels
  - Gene level: This gene have changes in isoform usage (unknown which)
  - Transcript level: This isoform has changed usage
- Long read RNA-seq is really useful (both PacBio and ONT)

# Isoforms have different function

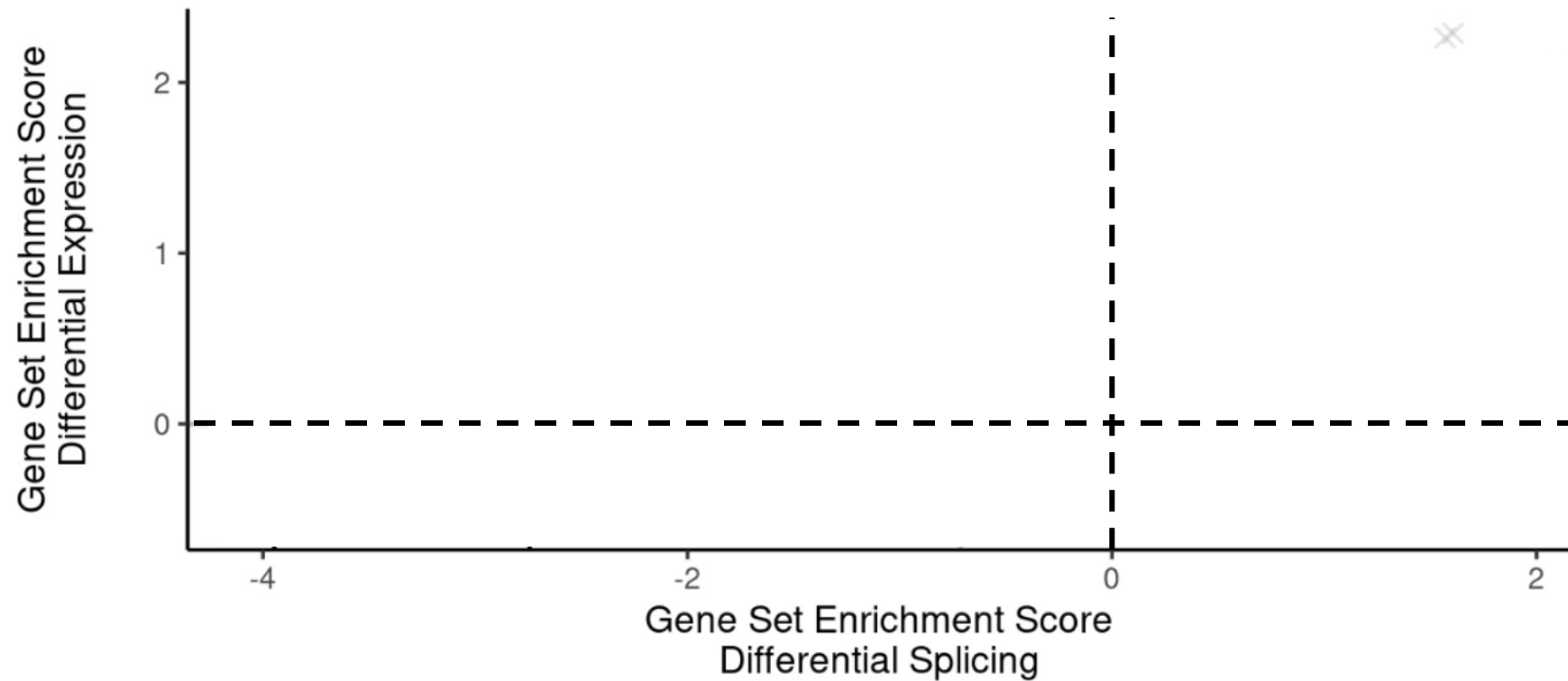
- Opposite effects in apoptosis
- Interact with different proteins
- Located different places in the cell
- P53

# Differential Splicing is Omnipresent



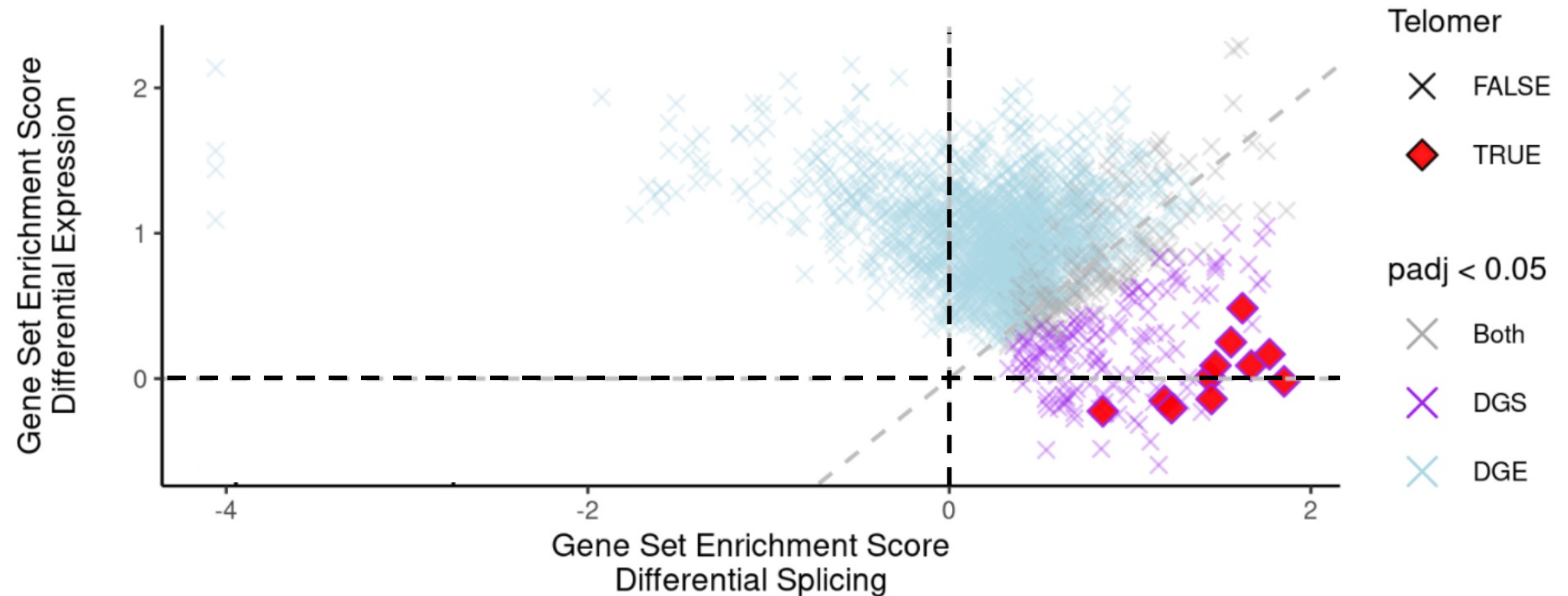
93% of all multi-isoform genes

# Splicing Mediate Distinct Biological Signals



pairedGSEA R package

# Splicing Mediate Distinct Biological Signals



pairedGSEA R package



# Isoforms are important

- For analysis of high-throughput data – including single cell analysis
- In clinical settings
  - Diagnosis
  - Treatment
- In genetics
- In most diseases – especially cancer

Exceptionally understudied!

# Main take-aways

- Transcriptomics profile the RNA content of a cell
  - Bulk
  - Single Cell
  - Spatial
- The resulting count/expression matrix enables downstream analysis
  - Differential Expression
  - Gene set enrichment analysis
- Isoforms are important and overlooked
- Enable high-level Integrative Analysis

# Transcriptomics Enable high-level Integrative Analysis



DNA



Transcripts



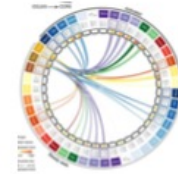
Proteins



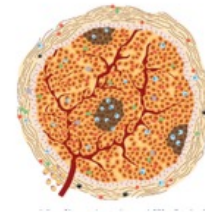
Pathways



Cell



Cellular interactions



Tissue/Cancer organization



Functional phenotype

# Come work with me

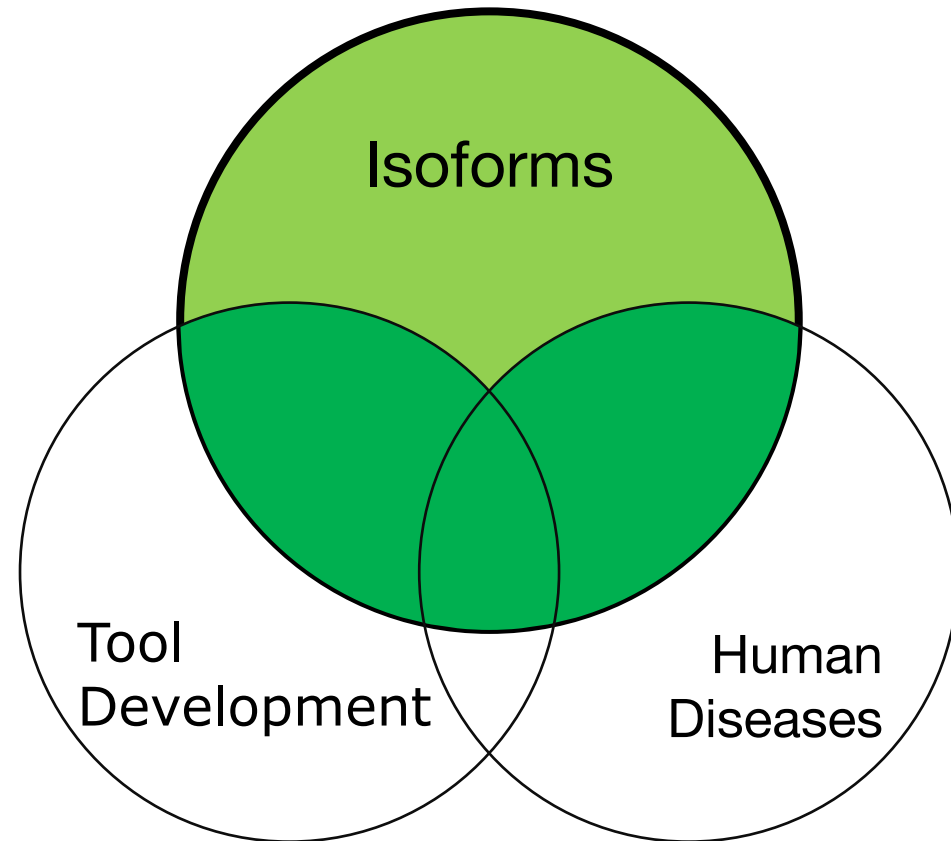
## Projects

- BSc
- MSc
- Special Courses

## Technologies

- RNA-seq (bulk/SC)
- Proteomics

I aim to *inspire* and *enable* everybody to study *isoforms*



# Assignment Time!

[shorturl.at/nzBPQ](https://shorturl.at/nzBPQ)

Remember the -X when you log in