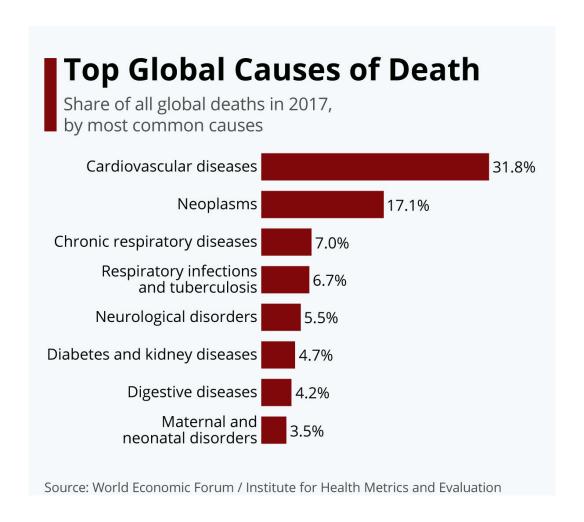




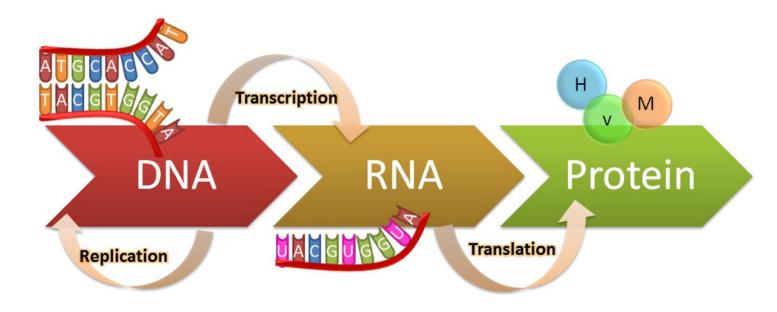
## A Short Introduction to Transcriptomics

Kristoffer Vitting-Seerup, PhD Assistant Professor in Bioinformatics krivi@dtu.dk

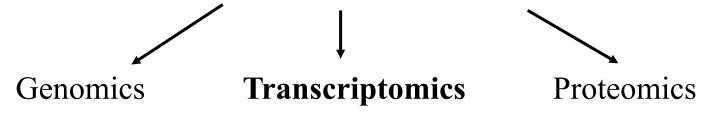


To treat most of these we need to understand the molecular mechanisms and how they are change by disease

#### Aim: Profile difference between healthy and sick

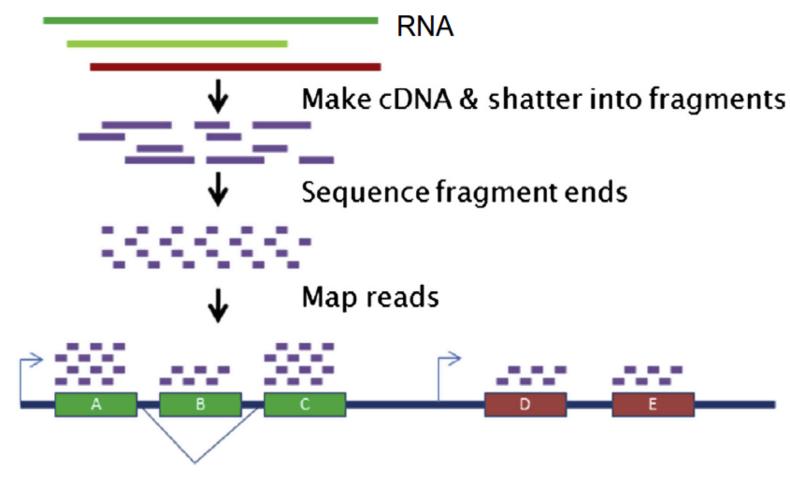


Solution: Measure all <u>DNA</u>, <u>RNA</u> or all <u>protein</u> in healthy and sick



# RNA-sequencing 101

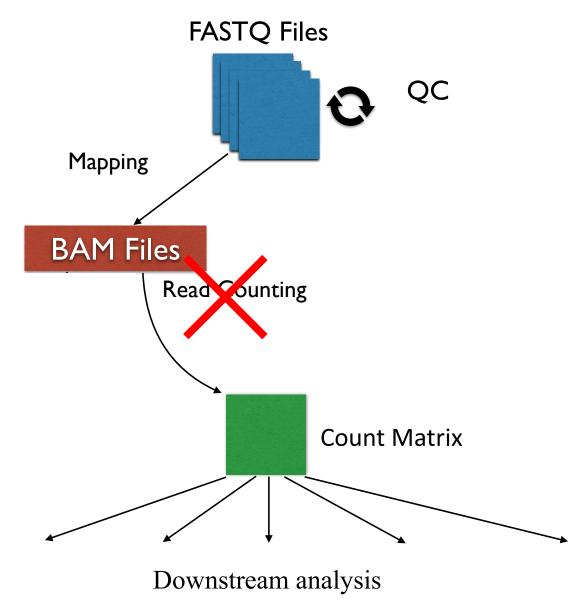
## RNA-sequencing



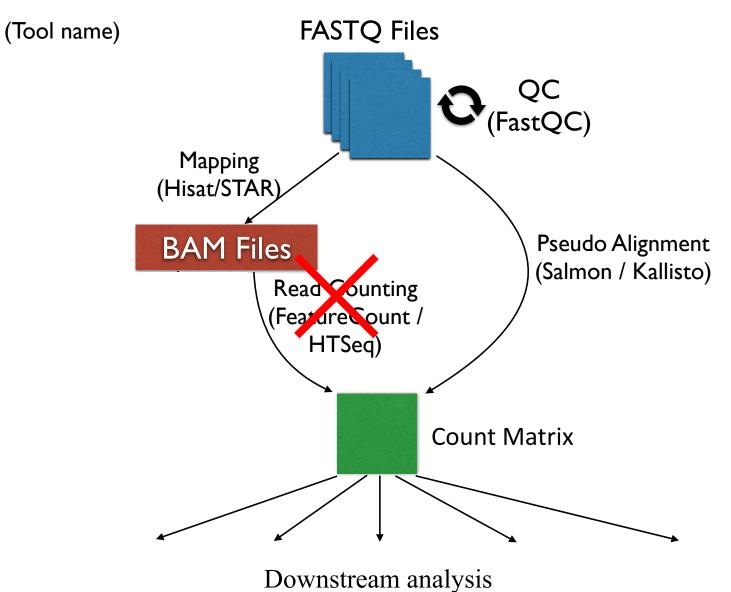
Count read mapped to each Gene

Count Matrix

#### More detailed workflow

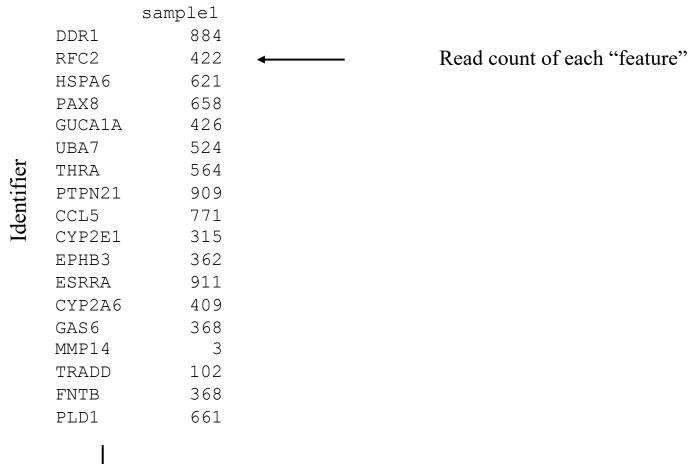


#### More detailed workflow



#### Gene Count Matrix

The results from a single sample RNA-seq datasets:



Typically, 20,000 – 50,000 genes

#### Gene Count Matrix

Samples ----

		sample1	sample2	sample3	sample4	sample5
Genes	DDR1	884	81	622	946	541
	RFC2	422	858	305	830	28
	HSPA6	621	130	401	221	110
<b>↓</b>	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

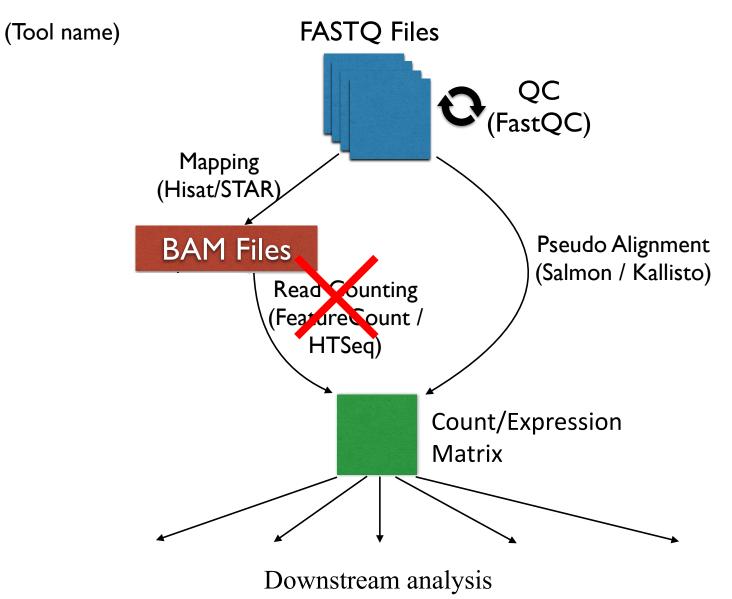
#### RNA-sequencing Count Matrix

- 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



## Downstream Analysis

Two of the hundreds of possible uses

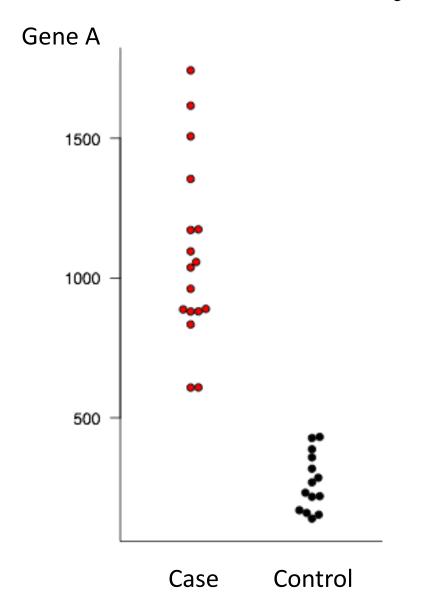
#### What question do you want to answer?

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

• A nice, normalized expression matrix:

	Healthy 1	Healthy 2	Healthy 3	Disease 1	Disease 2	Disease 3
Gene A	10	12	9	9	8	13
Gene B	0	5	1	25	26	22
Gene C	35	45	55	0	8	10

Modern bioinformatic tool jointly estimates both log2FCs and P-values using (generalized) linear modeling

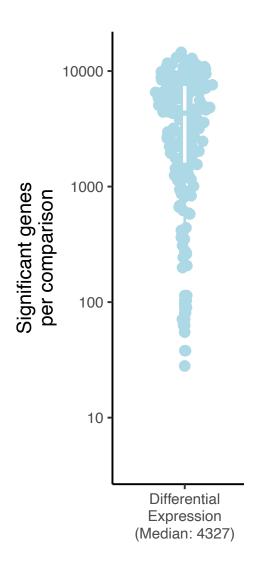
## Differential Expression Analysis

- 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) (<a href="http://geneontology.org/">http://geneontology.org/</a>)
  - MSigDB (<a href="https://www.gsea-msigdb.org/gsea/msigdb/">https://www.gsea-msigdb.org/gsea/msigdb/</a>)
  - Enrichr (<a href="https://maayanlab.cloud/Enrichr/#libraries">https://maayanlab.cloud/Enrichr/#libraries</a>)

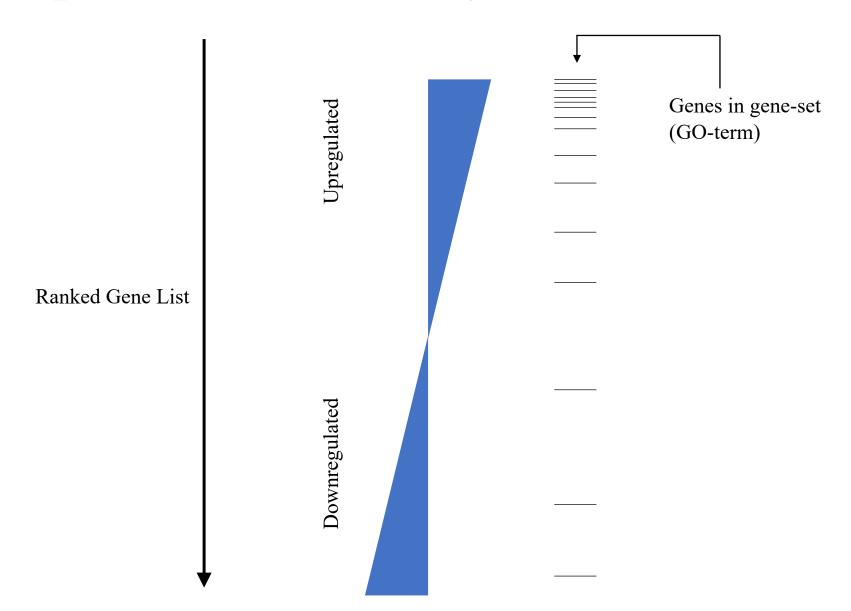
#### Gene Set Enrichment Analysis

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

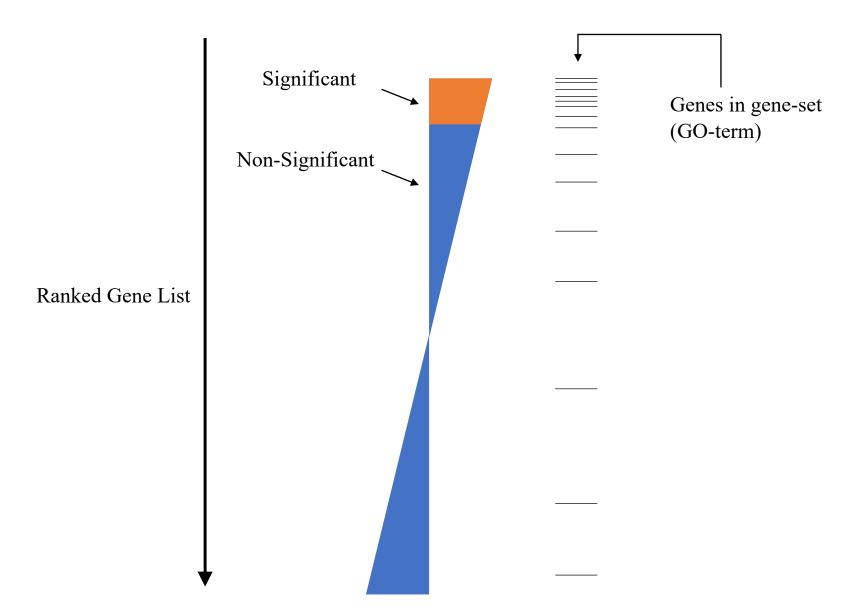
- 1. Overrepresentation Analysis (OA/ORA)
- 2. Functional Class Scoring (FSC)

Confusingly these are both referred to as GSEA

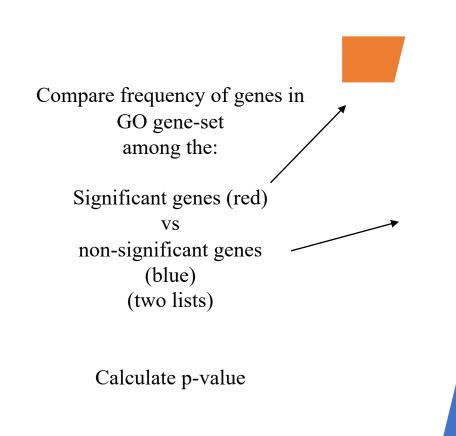
## Overrepresentation Analysis

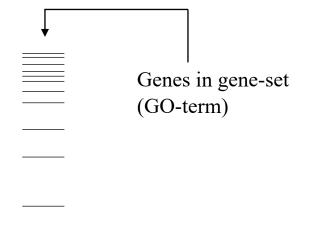


## Overrepresentation Analysis



## Overrepresentation Analysis





	Not in gene-set	In gene-set
Not significant	1000	100
Significant	10	10

#### 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**

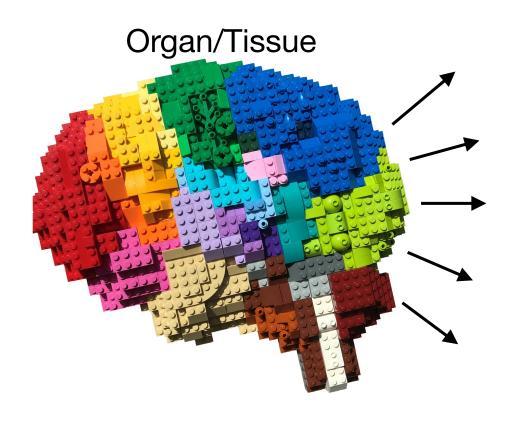
- Literally hundreds of tools for doing it!
- R packages
  - <u>fgsea</u>
  - clusterProfiler
  - limma
  - gProfiler
  - pairedGSEA
- Web tools
  - <a href="http://geneontology.org/">http://geneontology.org/</a>
  - <a href="https://david.ncifcrf.gov/">https://david.ncifcrf.gov/</a>
  - <a href="https://biit.cs.ut.ee/gprofiler/gost">https://biit.cs.ut.ee/gprofiler/gost</a>
- Pay attention to your background!

# RNA-sequencing 101 – Done!

## Extention #1

Bulk vs Single cell vs Spatial

## Transcriptomics



# Bulk RNA-seq

Single Cell RNA-seq



#### Bulk RNA-seq

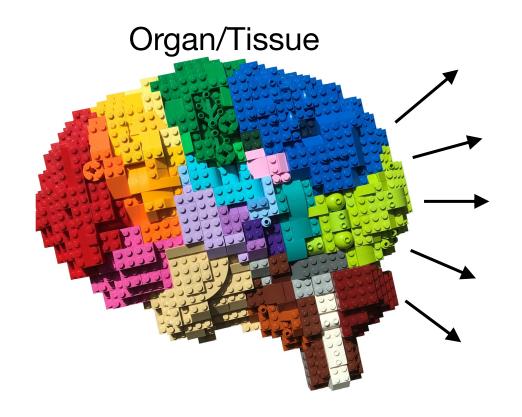
• 2 Min with neighbors: What potential problems are there with just measuring the average signal (compared to single cell)?



Single Cell RNA-seq



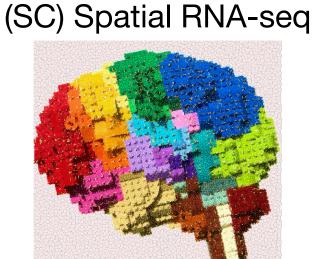
## Transcriptomics



Bulk RNA-seq

Spatial RNA-seq

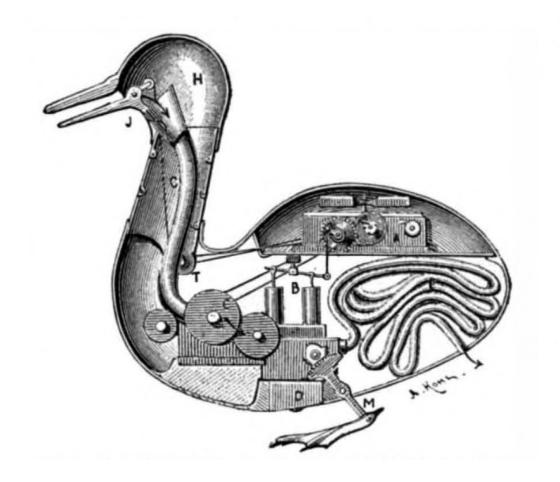
Single Cell RNA-seq



Modified from @BoXia7, https://twitter.com/adj\_23/status/1261928476811997184

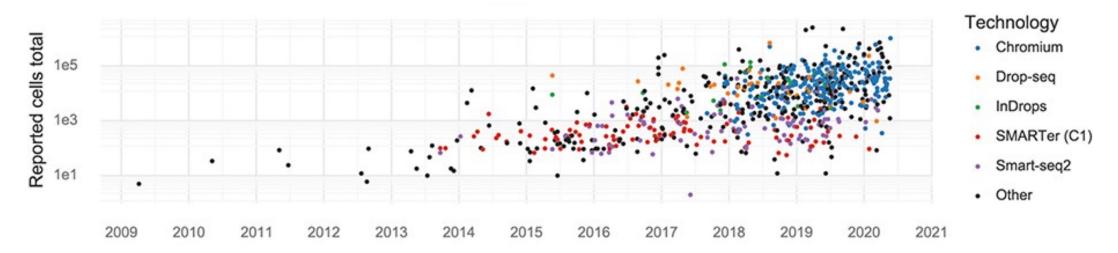
#### Bulk RNA-seq

- 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 the advanced master 22102 I'm running



## Single Cell Multi Omics

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

#### nature methods

Explore content > About the journal > Publish with us >

nature > nature methods > editorials > article

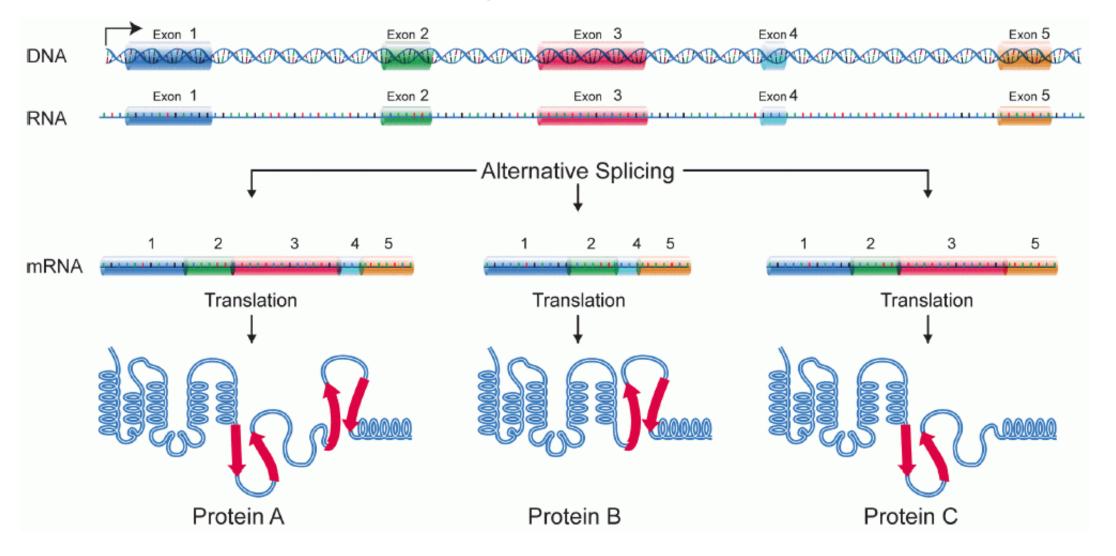
Editorial Published: 06 January 2020

Method of the Year 2019: Single-cell multimodal omics

## Extention #2

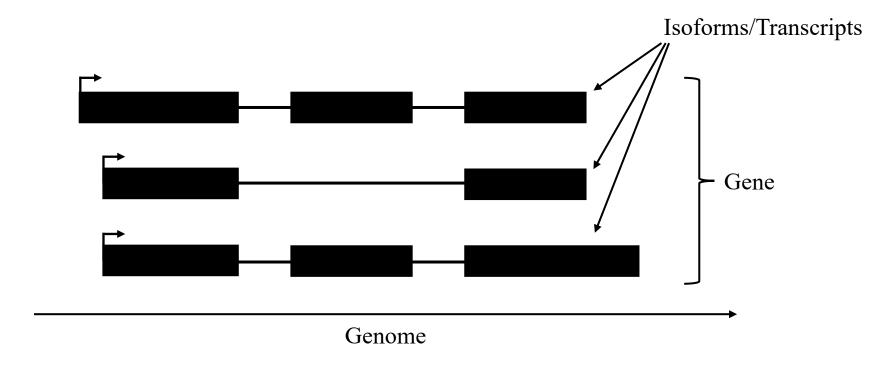
There is no such thing as a "gene"

## 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

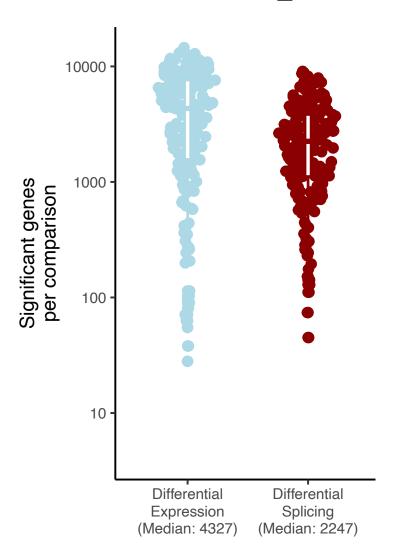
• Opposite effects in apoptosis

• Interact with different proteins

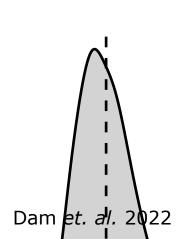
• Located different places in the cell

• P53

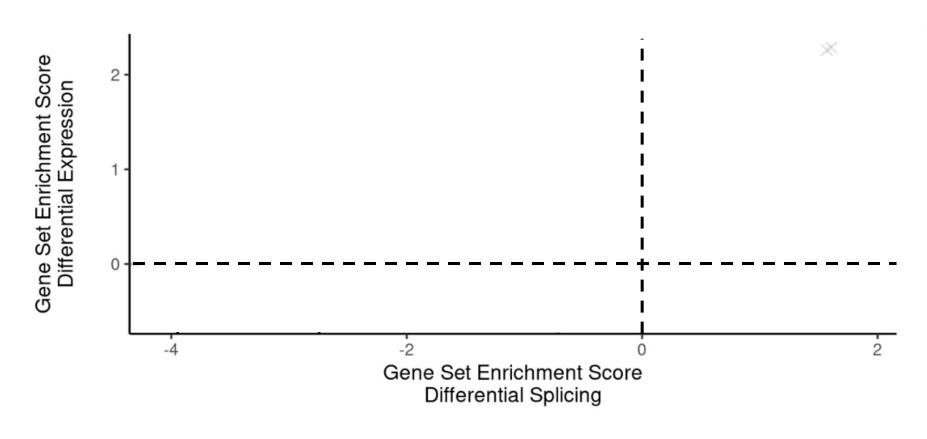
## Differential Splicing is Omnipresent



93% of all multi-isoform genes

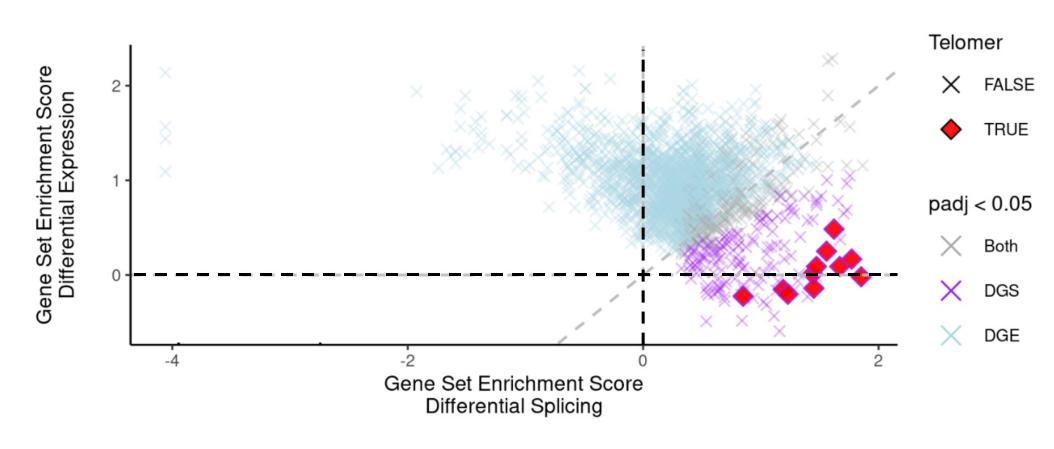


## Splicing Mediate Distinct Biological Signals



pairedGSEAR package

## Splicing Mediate Distinct Biological Signals



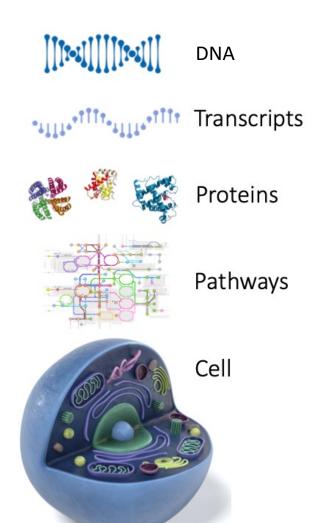
pairedGSEAR package

## 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

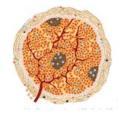
Exceptionally understudied!

#### Transcriptomics Enable high-level Integrative Analysis





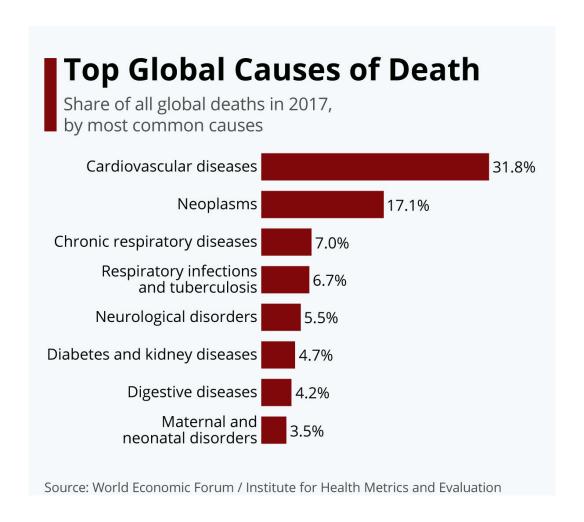
Cellular interactions



Tissue/Cancer organization



Functional phenotype



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

#### Come work with me

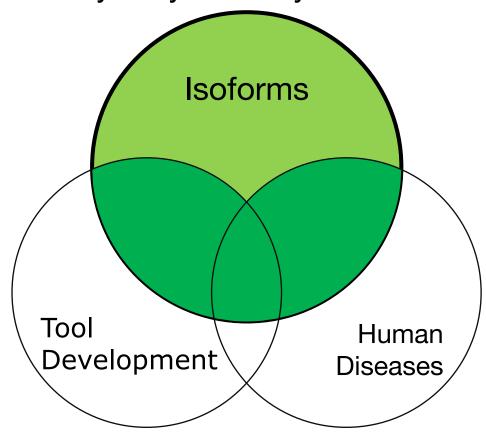
#### **Projects**

- BSc
- MSc (30-60 ETCS)
- Special Courses

#### **Technologies**

- RNA-seq (bulk/SC)
- Proteomics

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



#### 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

# Assignment Time!