



Single cell experimental design and hypothesis testing

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Key Topics



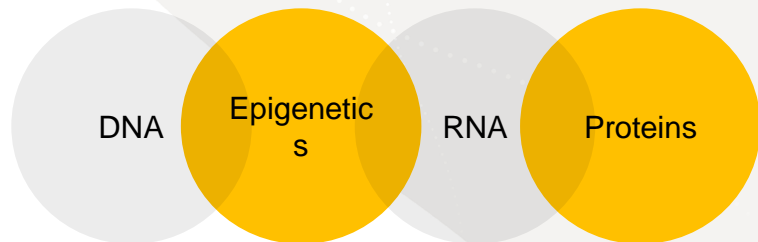
- Introduction to the range of techniques for generating scRNA-seq data
- Key challenges and considerations
- What analysis questions can I address?

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Cellular genomics



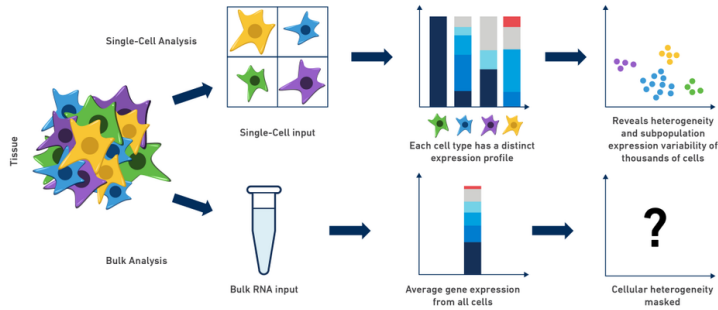
Science
Breakthrough technology
of 2018



Cellular genomics technology allows you to measure the genomic contents for an individual cell. But at 10,000s of cells at a time.

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Why single cells?

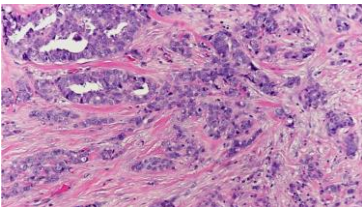


Do we expect cells to have the same transcriptional signatures?

Bulk methods give an average RNA signal

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A typical challenge that we might face

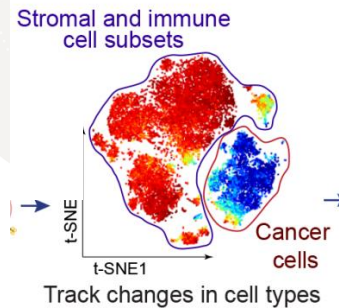


A tumor biopsy contains many cell types

- Tumor cells (with different mutations)
- Healthy non-tumor cells
- Immune cells
- Nerves
- Blood vessels
-

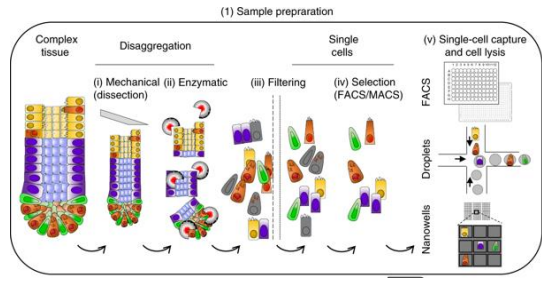
Traditional genomic assays average the signals across all cells

Single cell sequencing enables complete resolution of cell types and their genomic functions



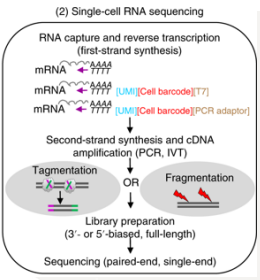
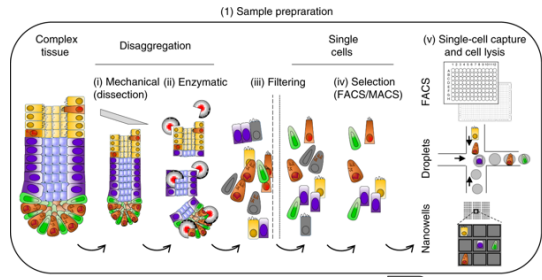
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Typical workflow



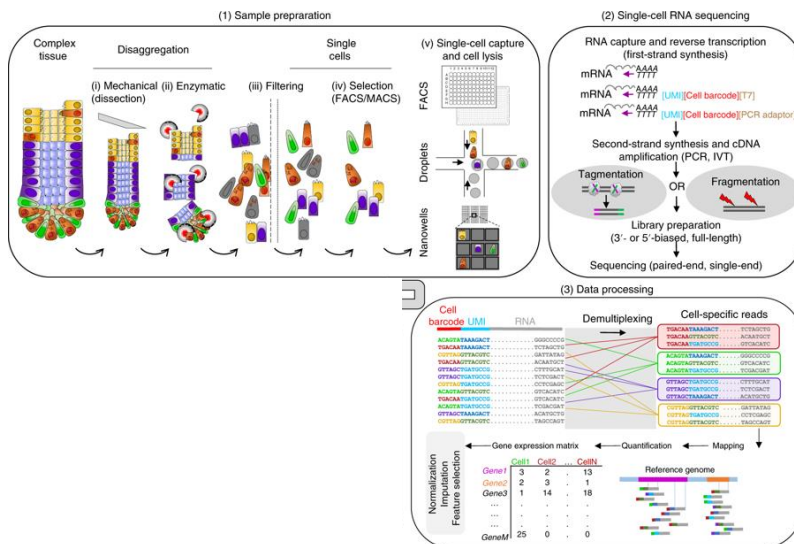
7 Lafzi et al. Nature Protocols 2018

Typical workflow



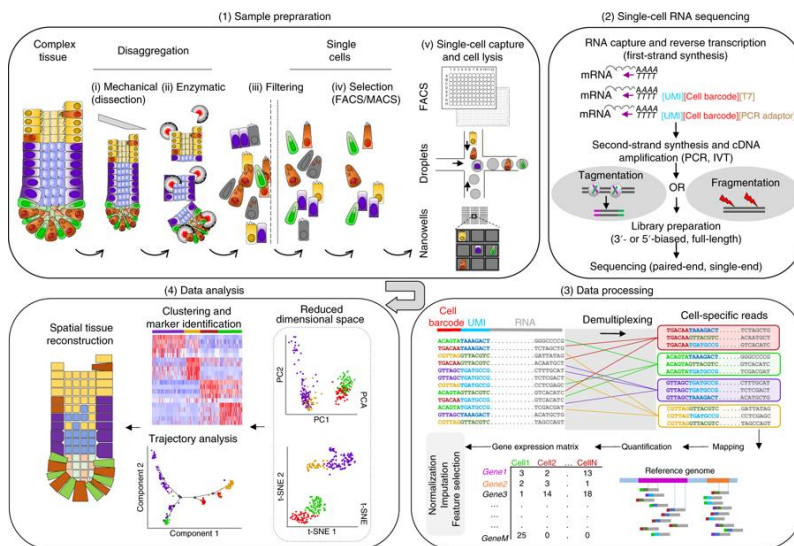
8 Lafzi et al. Nature Protocols 2018

Typical workflow



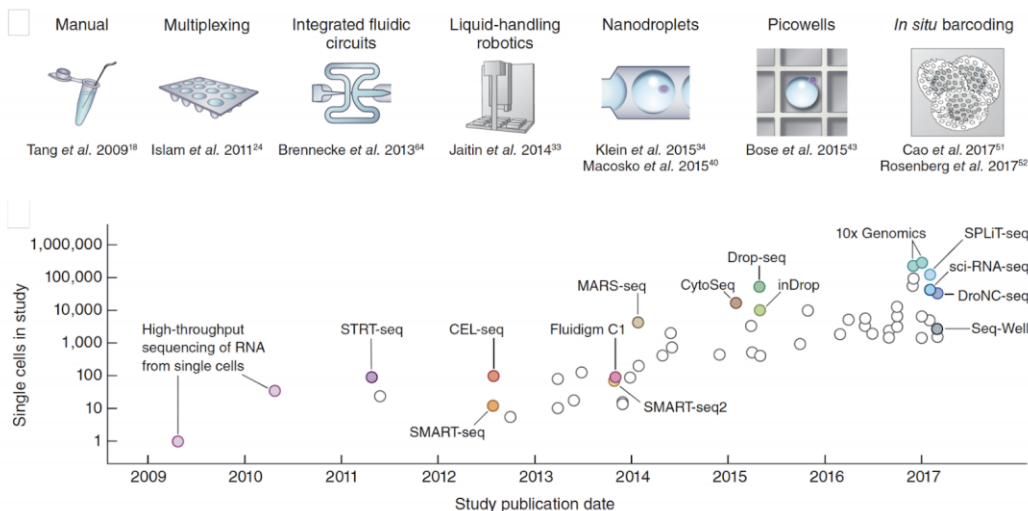
9 Lafzi et al. Nature Protocols 2018

Typical workflow



10 Lafzi et al. Nature Protocols 2018

Technology has progressed rapidly



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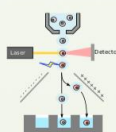
Cell capture defines methods

Microfluidic device



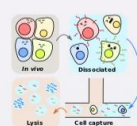
- 96 or 800 well format
- Physically check presence of cells
- High capture efficiency
- Doublet issues
- Expensive
- Full-length cDNA (SMART-seq{2})
- Spike-in control RNA
- **High gene coverage**

Plate-based



- 96 or 384 well format
- Sort specific population(s) of cells
- High capture efficiency
- Experimental design considerations
- Full-length cDNA (SMART-seq{2}) or end-tagging; UMIs
- Spike-in control RNA
- **High gene coverage**

Droplet-based



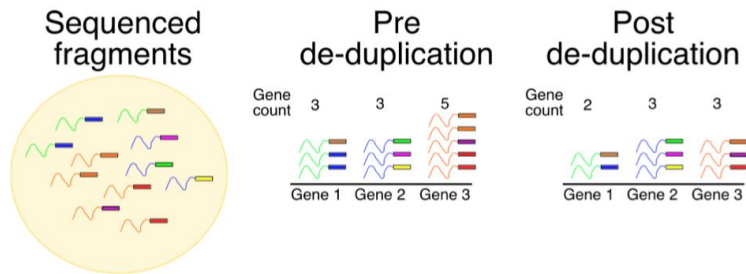
- 100-1000's of cells
- Doublet issues
- Variable capture efficiency
- Low per-cell cost
- 3' end tag; UMIs
- No spike-in control RNA
- **High cell coverage**

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What are UMIs?

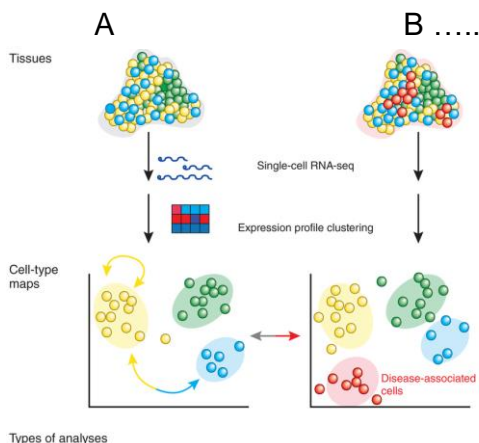
Unique molecular identifiers give (almost) exact molecule counts in sequencing experiments.

They reduce the amplification noise by allowing (almost) complete de-duplication of sequenced fragments.



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What questions need to be considered in the experimental design?



How common / rare are the cells in am most interested in?

How different are they likely to be from other cell types?


Can the 'phenotype' be detected By RNA differences?


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- Within cell type**
 - Stochasticity, variability of transcription
 - Regulatory network inference
 - Allelic expression patterns
 - Scaling laws of transcription
- Between cell types**
 - Identify biomarkers
 - (Post)-transcriptional differences
- Between tissues**
 - Cell-type compositions
 - Altered transcription in matched cell types

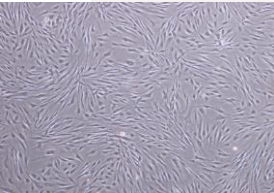
Considerations for single cell experiments

Cell Prep






Liquid



cell cultures



solid tissue

How much starting material is there?
Human/non-human

The critical point is **getting single cells and keeping them that way**


Complete dissociation is required

- Cell Strainers
- Sorting – FACS
- Magnetics beads

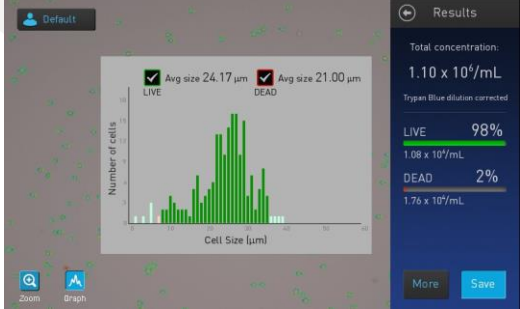
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Considerations for single cell experiments

Cell Prep



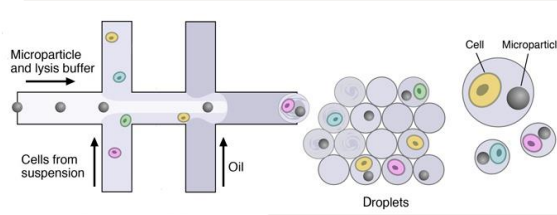
- How big are they?
- How sensitive are they?
 - Handling (e.g. pipetting force)
 - Enzymatic dissociation (timing and harshness)
 - FACS pressure/nozzle gauge
 - Post-dissociation viability?
- Will they lyse in the reaction buffer?
- Are they sticky/liable to clump?



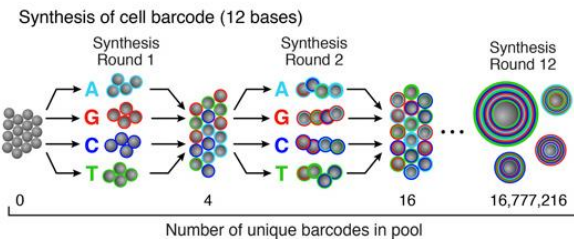
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Barcoding

Microfluidic

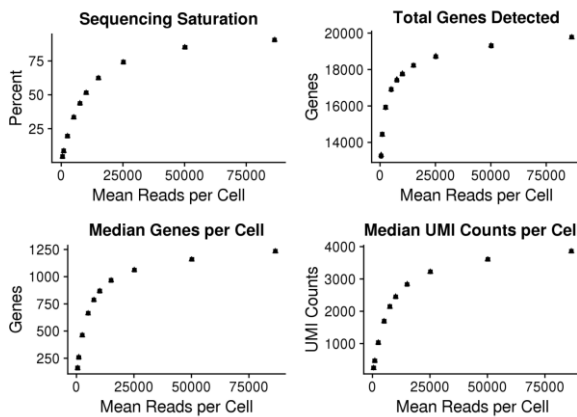


Barcodes



17 Makosco et al Cell 2015

How deep should I sequence?



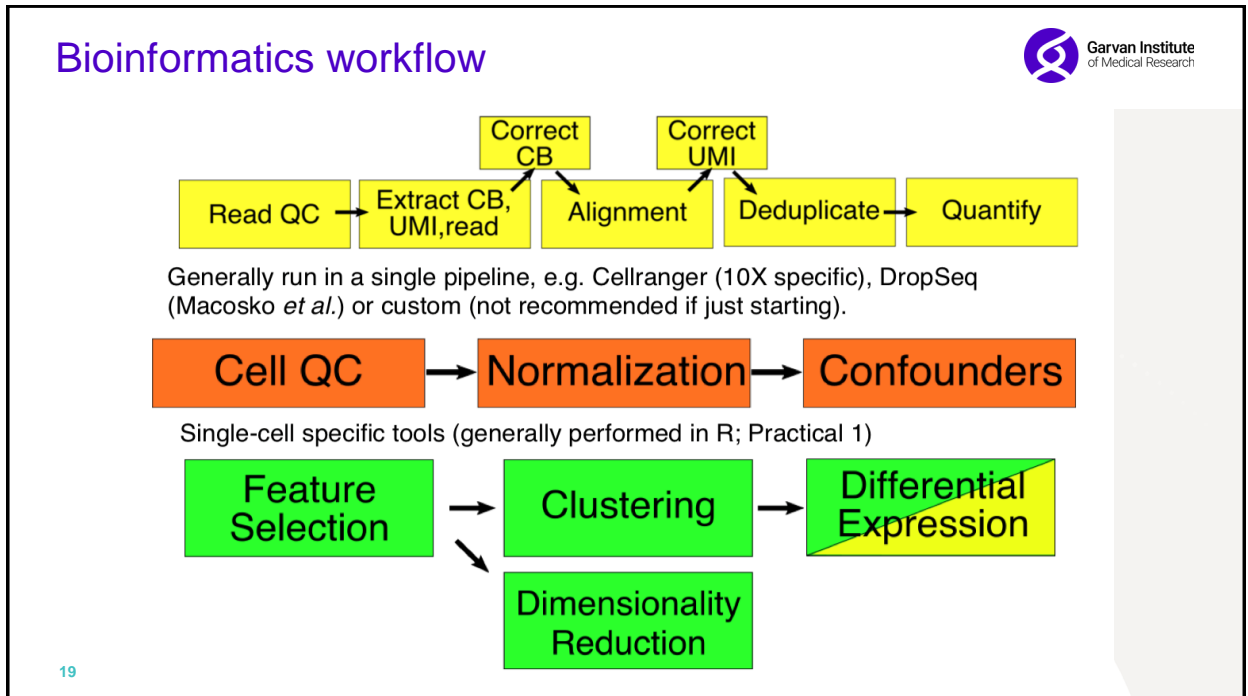
Costs

Per 10,000 reads* per cell

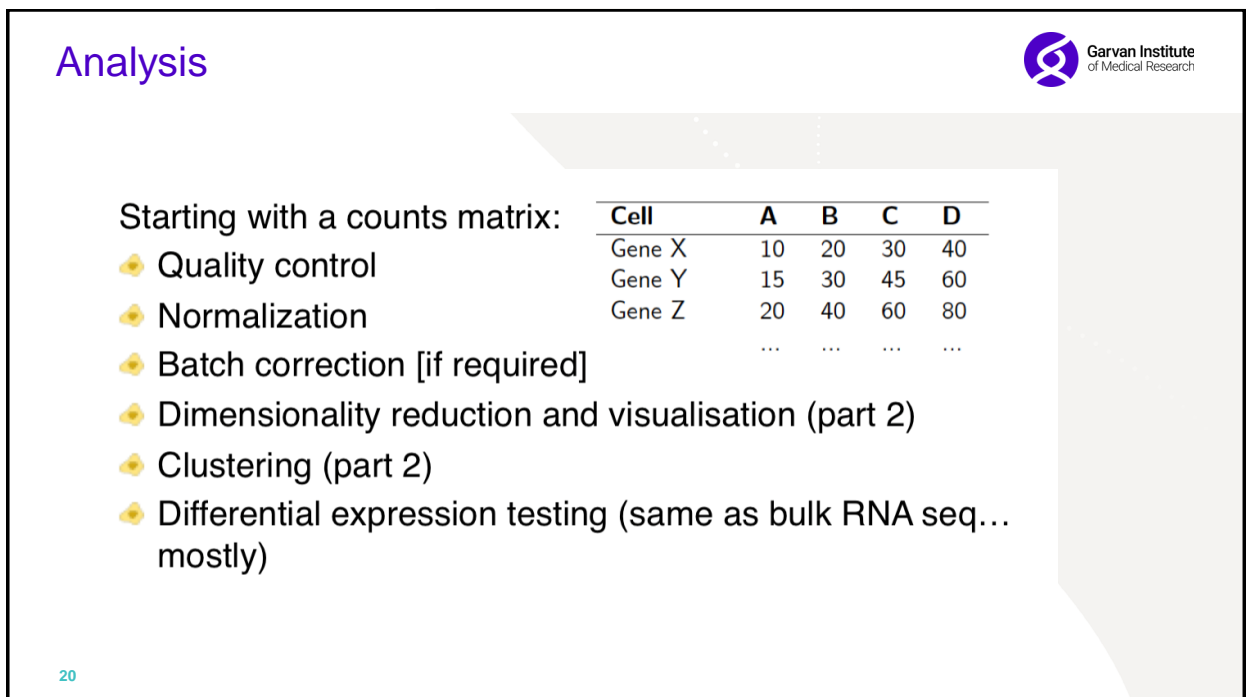
NextSeq	5-10 cents
HighSeq2500	10-15 cents
NovaSeq S2	3-5 cents
NovaSeq S4	2-4 cents

* With 3' scRNA-seq

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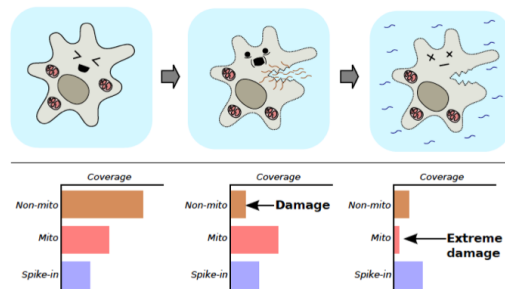
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Dealing with low quality cells

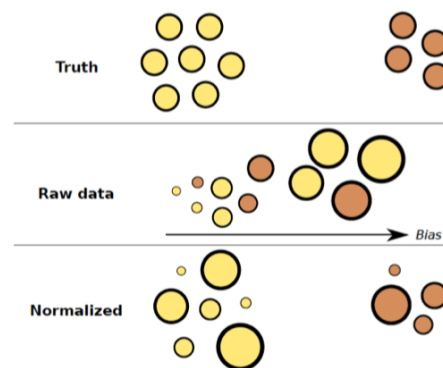
- ✦ Low sequencing depth
- ✦ Low numbers of expressed genes (i.e. any non-zero count)
- ✦ High spike-in (if present) or mitochondrial content



21 From Aaron Lun

Normalization

- ✦ The aim is bring all cells onto the same *distribution* to remove biases between them
- ✦ We want to preserve biological variability, not introduce new technical variation
- ✦ Primary source of bias is sequencing depth – scale down counts accordingly
- ✦ Need a method that is robust to sparsity and composition bias
 - ✦ TMM & DESeq size factors are not!



22 Image from Aaron Lun

Confounding and batch effects

- ✦ A segue into proper experimental design
- ✦ Some batch effects cannot be avoided
- ✦ Some can, make sure you know which is which

A simple batch correction is
To use linear regression

Providing your biological groups
and processing groups are not
confounded.

But these methods assume most
cells are transcriptionally the same.

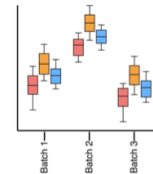
We will discuss more complex
normalization later

Not confounded
design

Biological
groups



Processing
groups



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Resources

Single Cell Resources:

Single cell course (Hemberg Lab; Wellcome Sanger Institute):
<http://hemberg-lab.github.io/scRNA.seq.course/index.html>

Aaron Lun's single cell workflow (very detailed):
<https://www.bioconductor.org/packages/release/workflows/html/simpleSingleCell.html>

Cellranger pipeline:
<https://support.10xgenomics.com/single-cell-gene-expression/software/pipelines/latest/what-is-cell-ranger>

Thank You!

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Post Doc positions



Interested in a single cell postdoc?

My lab is recruiting 3 positions:

- Population genetics at single-cell resolution
- Clinical translation of single-cell seq methods
- Modeling common disease with stem cells at single-cell resolution

Advert on Garvan Jobs

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<https://www.garvan.org.au/research/cancer/computational-genomics>