

# INTRODUCTION TO SINGLE-CELL RNA-SEQ

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BIOC8145

2020/04/27 to 2020/05/01

# OUTLINE

## **2020/04/27: Lecture 1**

- scRNA-seq introduction
- Available Technologies and understanding limitations
- STUDY DESIGN!!!
- 10x FASTQ alignment

## **2020/04/29: Lecture 2**

- Quality control / gene and cell filtering
- Dimensional Reduction and plotting genes and cells

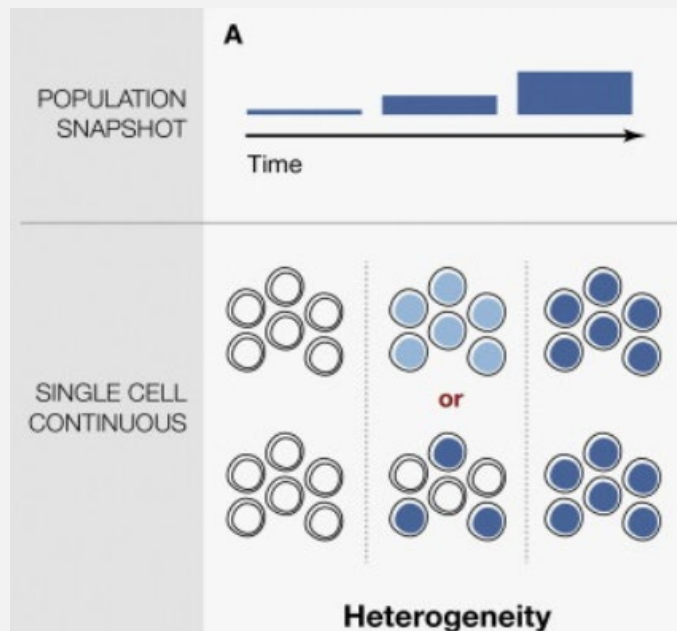
## **2020/05/01: Lecture 3**

- Differential gene expression
- Downstream analysis

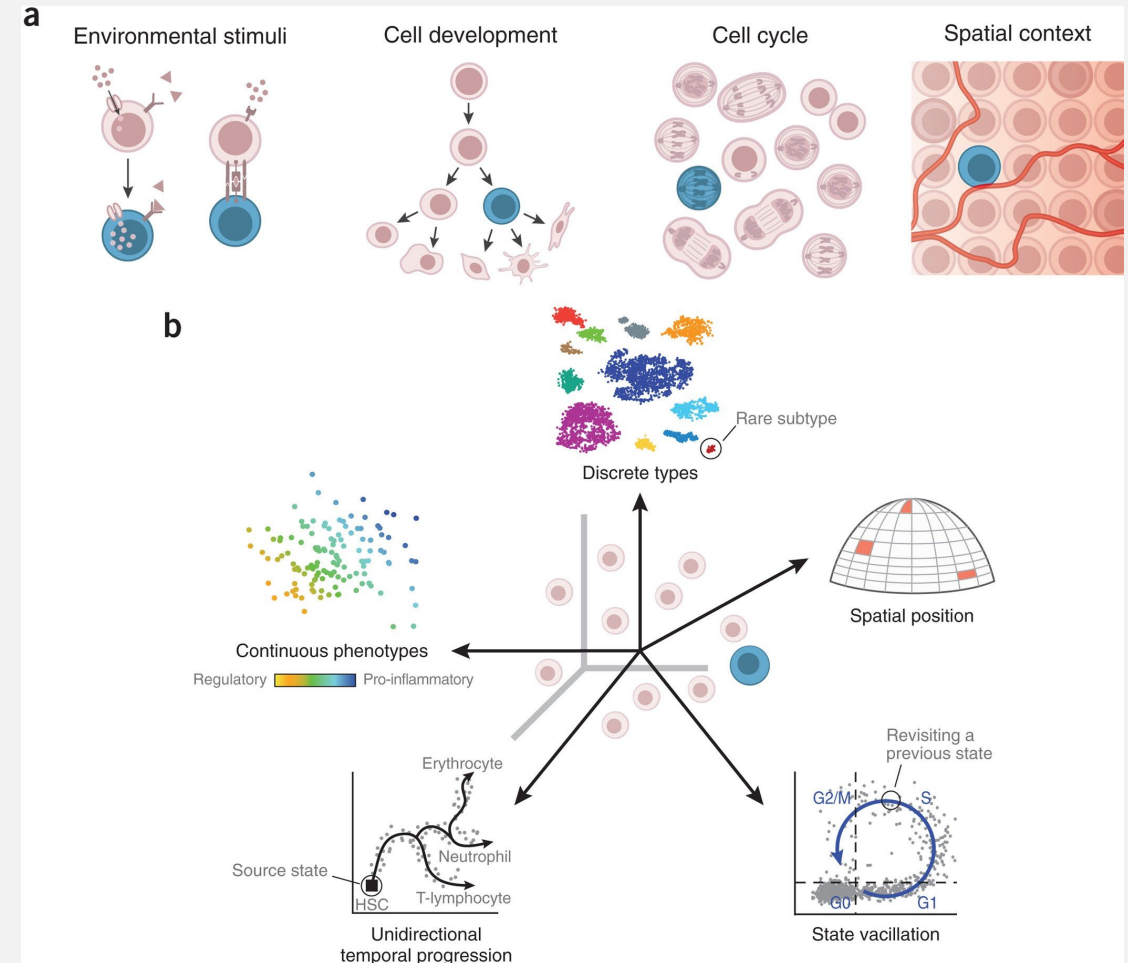
# Why do single cell RNA-seq?

scRNA-seq will help you understand what are the “identities” of cells in your specific model.

- Identify expression profiles of individual cells (that may be missed with bulk RNA-seq)



Etzrodt M., Ende M & Schroeder T. **Quantitative Single-Cell Approaches to Stem Cell Research.** *Cell Stem Cell* 2014

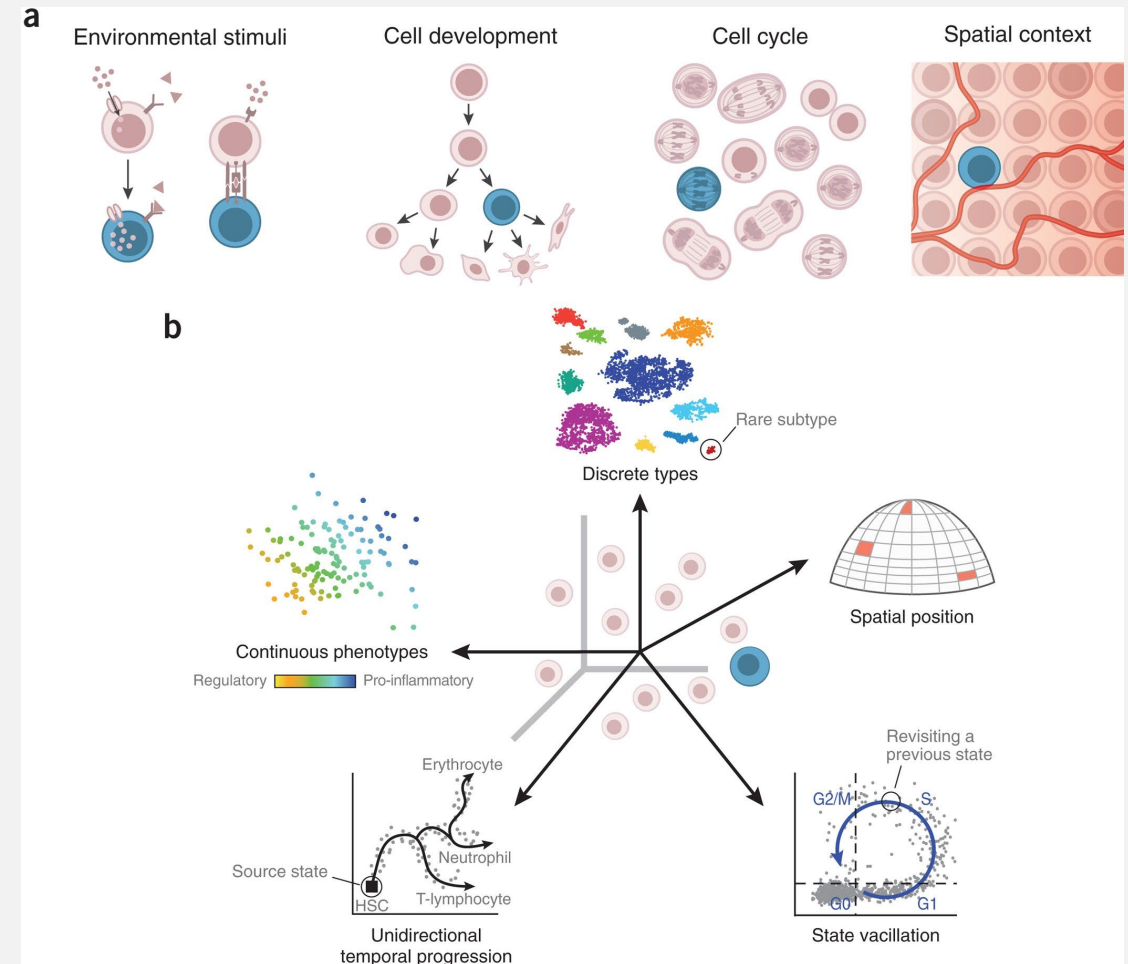


Wagner A., Regev A & Yosef N. **Revealing the vectors of cellular identity with single-cell genomics.** *Nat biotechnology* 2016

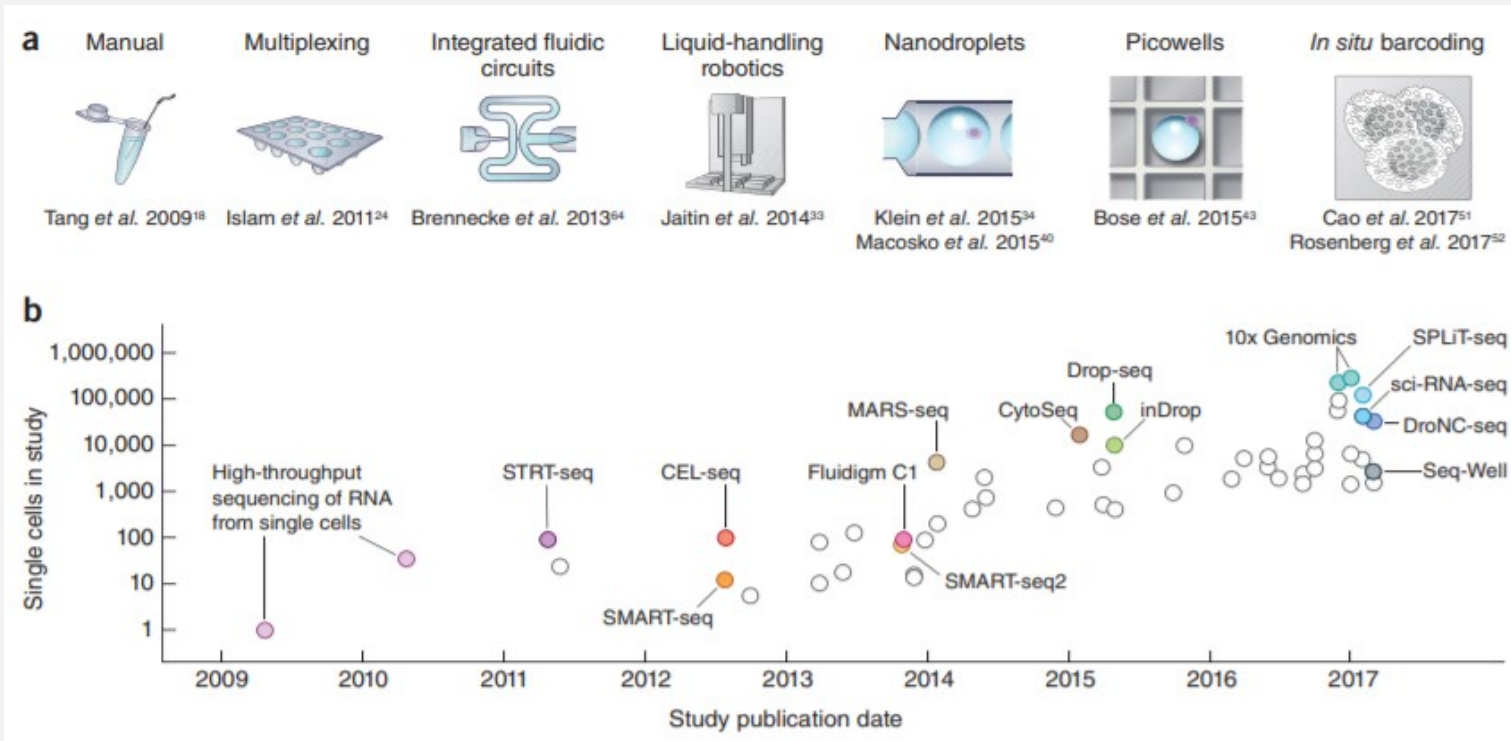
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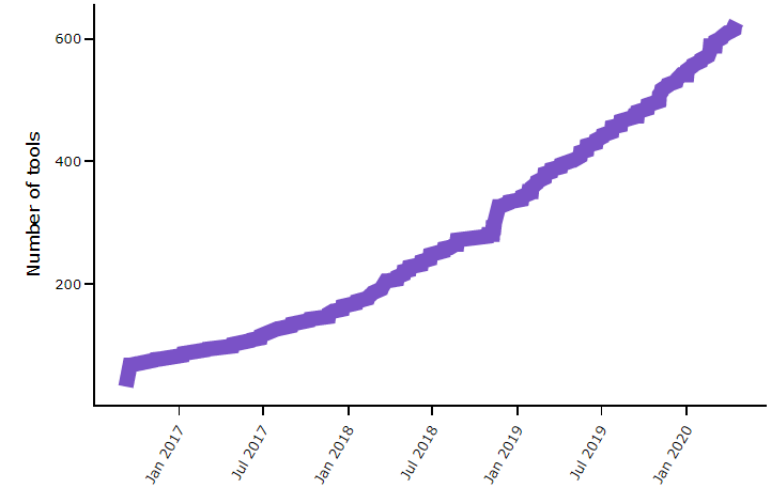
- Identify expression profiles of individual cells (that may be missed with bulk RNA-seq)
- Discover new rare populations and/or cell states
- Response to the environment and development trajectory



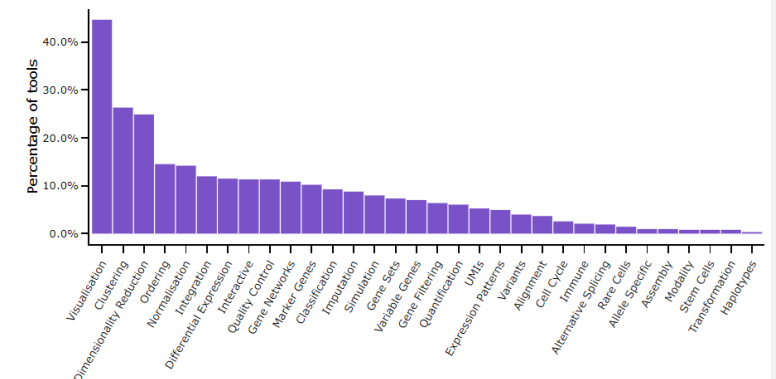
# Advances in scRNA-seq assays led to an exponential growth in publications and methods in the last 5 to 10 years



Svensson V., Vento-Tormo R & Teichmann S. **Exponential scaling of single-cell RNA-seq in the past decade.** *Nat protocols* 2018



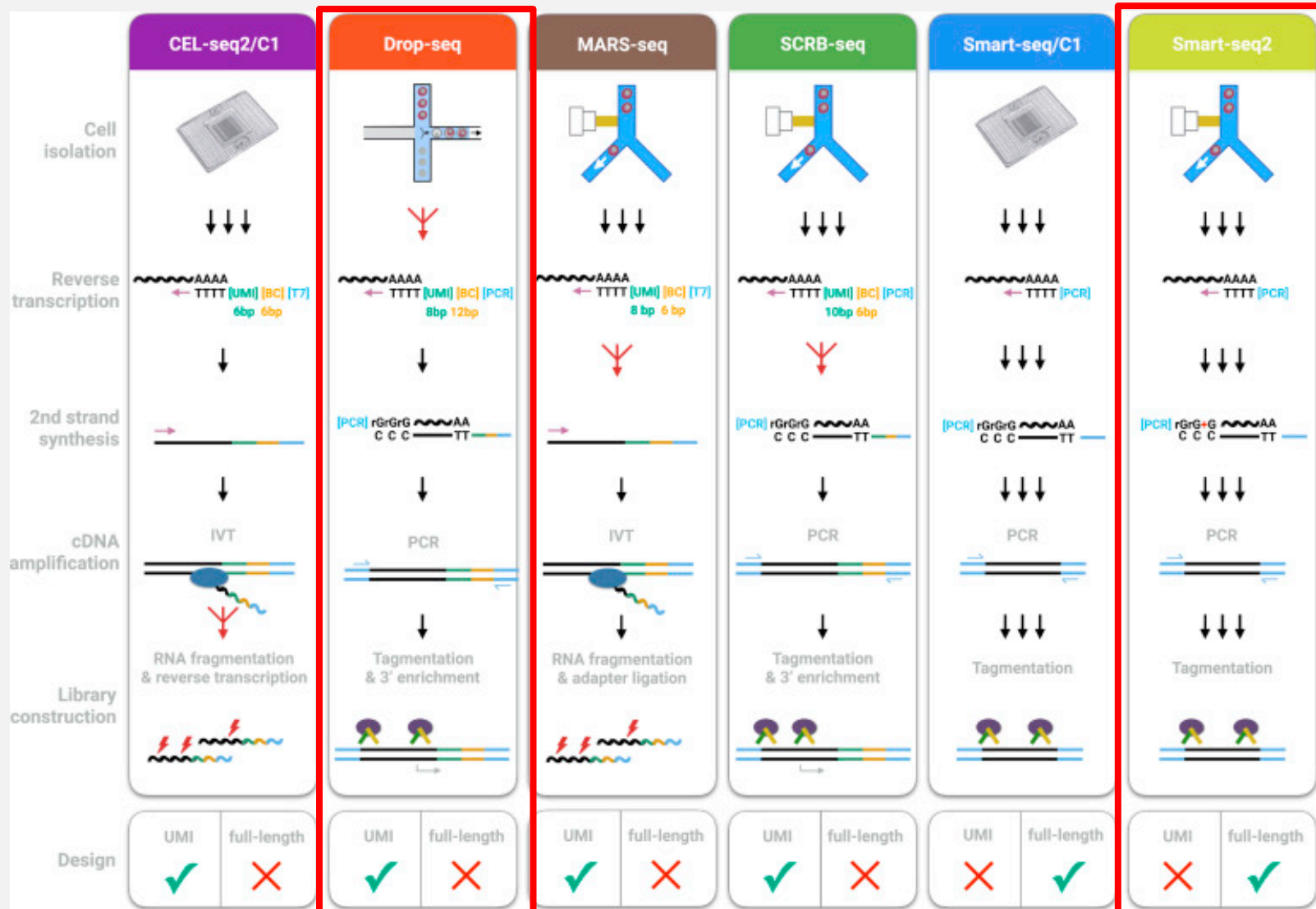
We currently track **627** tools...



...in over **30** categories

[www.scrna-seq.org](http://www.scrna-seq.org)

scRNA-seq assays differs in several areas, including library prep, UMI or not, transcript information, COST



Ziegenhain C. **Comparative Analysis of Single-Cell RNA Sequencing Methods**. *Molecular Cell* 2017

# Smart-seq2

## nature methods

Brief Communication | Published: 22 September 2013

### Smart-seq2 for sensitive full-length transcriptome profiling in single cells

Simone Picelli, Åsa K Björklund, Omid R Faridani, Sven Sagasser, Gösta Winberg & Rickard Sandberg 


*Nature Methods* **10**, 1096–1098(2013) | [Cite this article](#)

**11k** Accesses | **666** Citations | **113** Altmetric | [Metrics](#)

## nature protocols

Protocol | Published: 02 January 2014

### Full-length RNA-seq from single cells using Smart-seq2

Simone Picelli, Omid R Faridani, Åsa K Björklund, Gösta Winberg, Sven Sagasser & Rickard Sandberg 

*Nature Protocols* **9**, 171–181(2014) | [Cite this article](#)

**28k** Accesses | **857** Citations | **53** Altmetric | [Metrics](#)

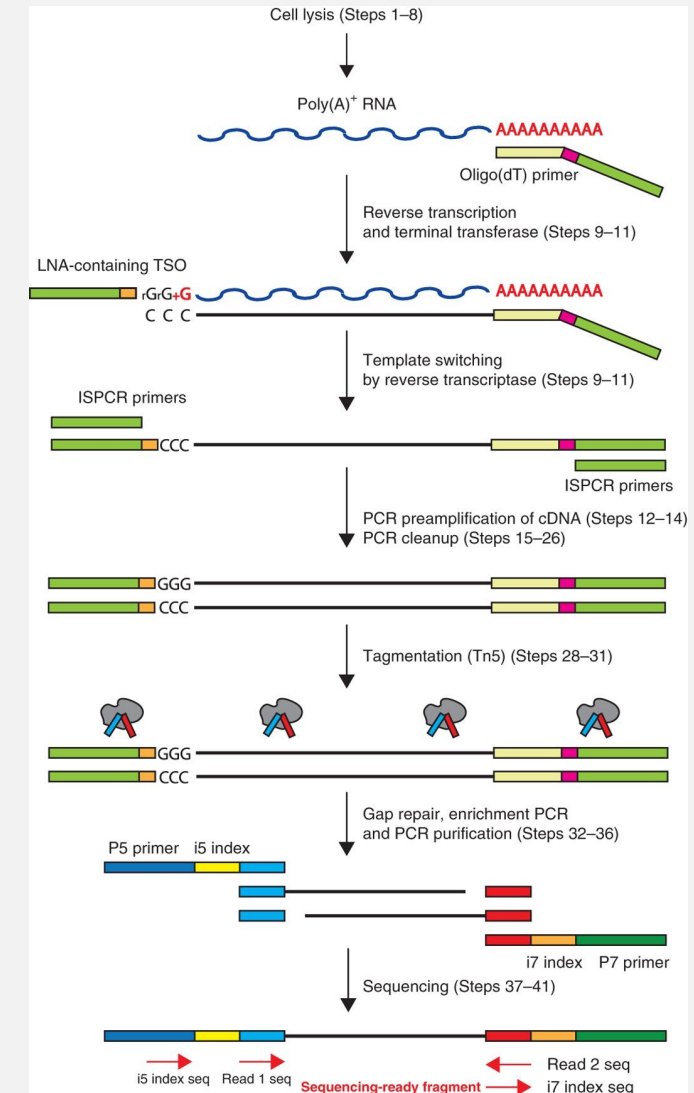


Also developed NASC-seq in 2019



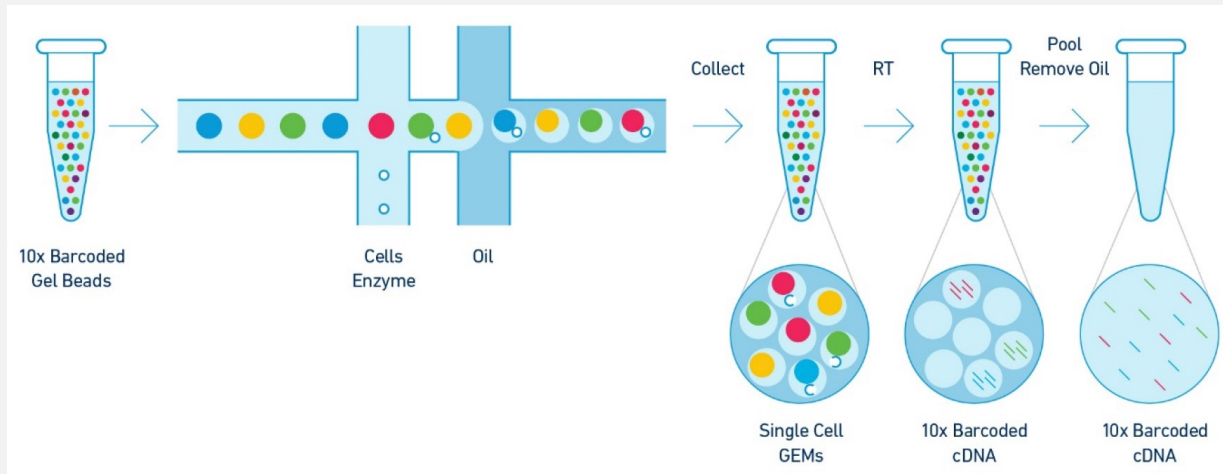
# Smart-seq2: Description and technology overview

- Developed for single cell, but has been adapted and used for total RNA (excellent for low input RNA)
- Selects for poly-A tail
- **Full transcript scRNA-seq**
- Off-the-shelf products
- Hundreds of samples
- Most often used without UMIs, but with spike-ins





# Drop-seq: Description and technology overview (also, if you are using UVA cores, this is what they offer)



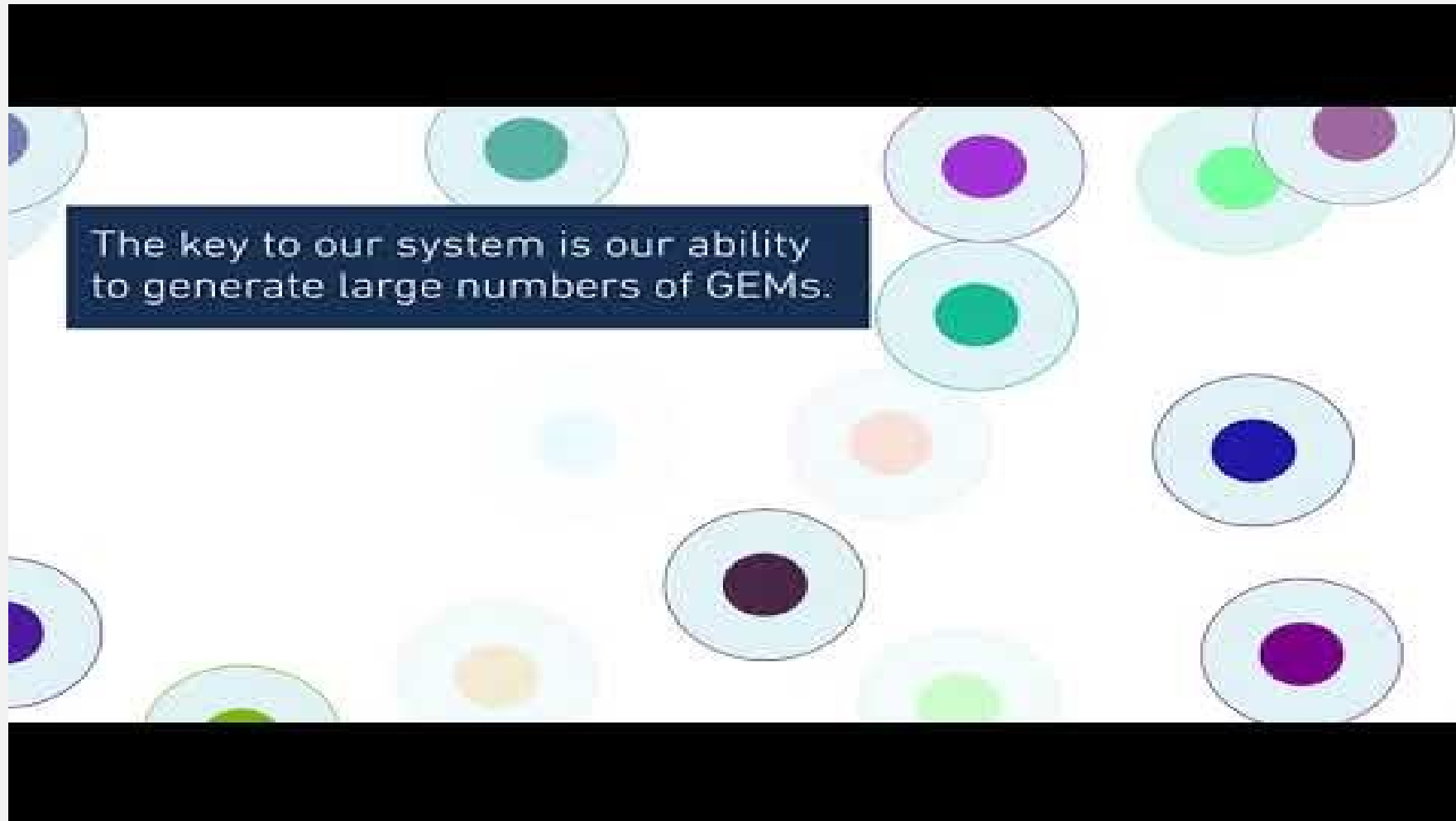
From [www.10xgenomics.com](http://www.10xgenomics.com)

- Throughput from hundreds to thousands.
- Droplet-based processing using microfluidics
- Nanoliter scale aqueous drops in oil.
- 3' End
- Uses UMI (Unique Molecular Identifier).

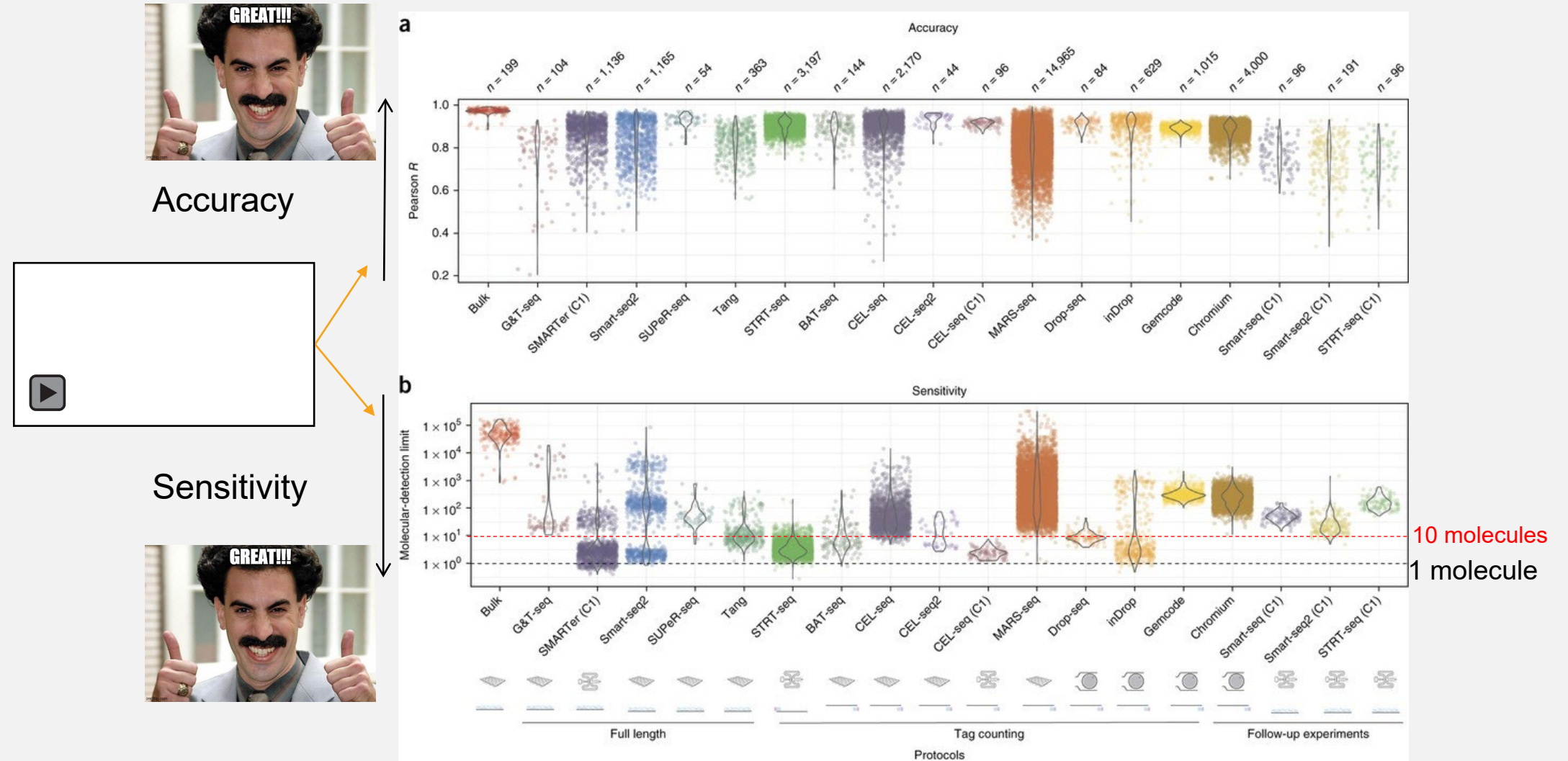
## More specifically to 10x:

- GEM (Gel Bead in Emulsion)
- Standardized instrumentation and reagents.
- Scalable to tens of thousands.
- Less processing time.
- Software is available and very supported by company

# Drop-seq: Description and technology overview (also, if you are using UVA cores, this is what they offer)



# Comparison between different assays



# Final thoughts on different technologies/assays

- Different assays have different throughput.
  - Smart-seq2 < Drop-seq < 10X
- Transcript length
  - 10x (only partial) < Smart-seq2 (full)
- Plate-based methods get lysed in wells and so do not leak.
  - Droplet-based can have leaky RNA.
- In Drop-seq assays RT happens outside the droplets

Understanding (or at least knowing the basics) of the assay/technology you selected is crucial for pipeline and downstream analysis

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- Plate-based methods get lysed in wells and so do not leak.
  - Droplet-based can have leaky RNA.
- In Drop-seq assays RT happens outside the droplets
  - Can use harsher lysis buffers.
- 10X is more standardized and comes with a pipeline.
- Drop-seq is more customizable.
- **Cost per library varies greatly.**

# Different technologies have different sequencing FASTQ files

ACGTACGTACGT

**SmartSeq2**

Left (R1)

Paired Transcript Sequence

**Dropseq**

Left (R1)

Cell Barcode

UMI

PolyT

**10X**

Cell Barcode

ACGTACGTACGT

Right (R2)

Paired Transcript Sequence

Right (R2)

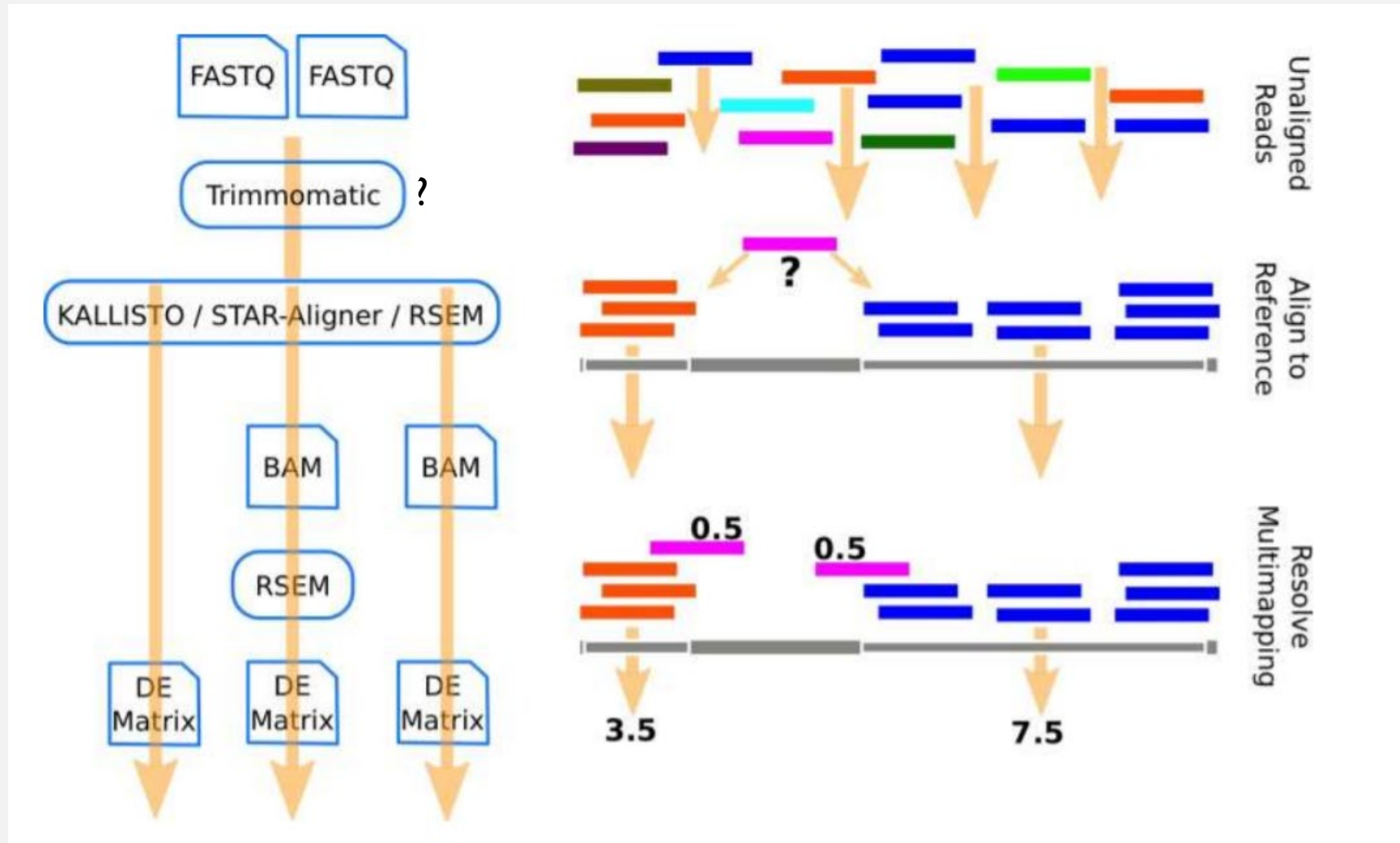
Transcript Sequence

I1 RA

Transcript Sequence

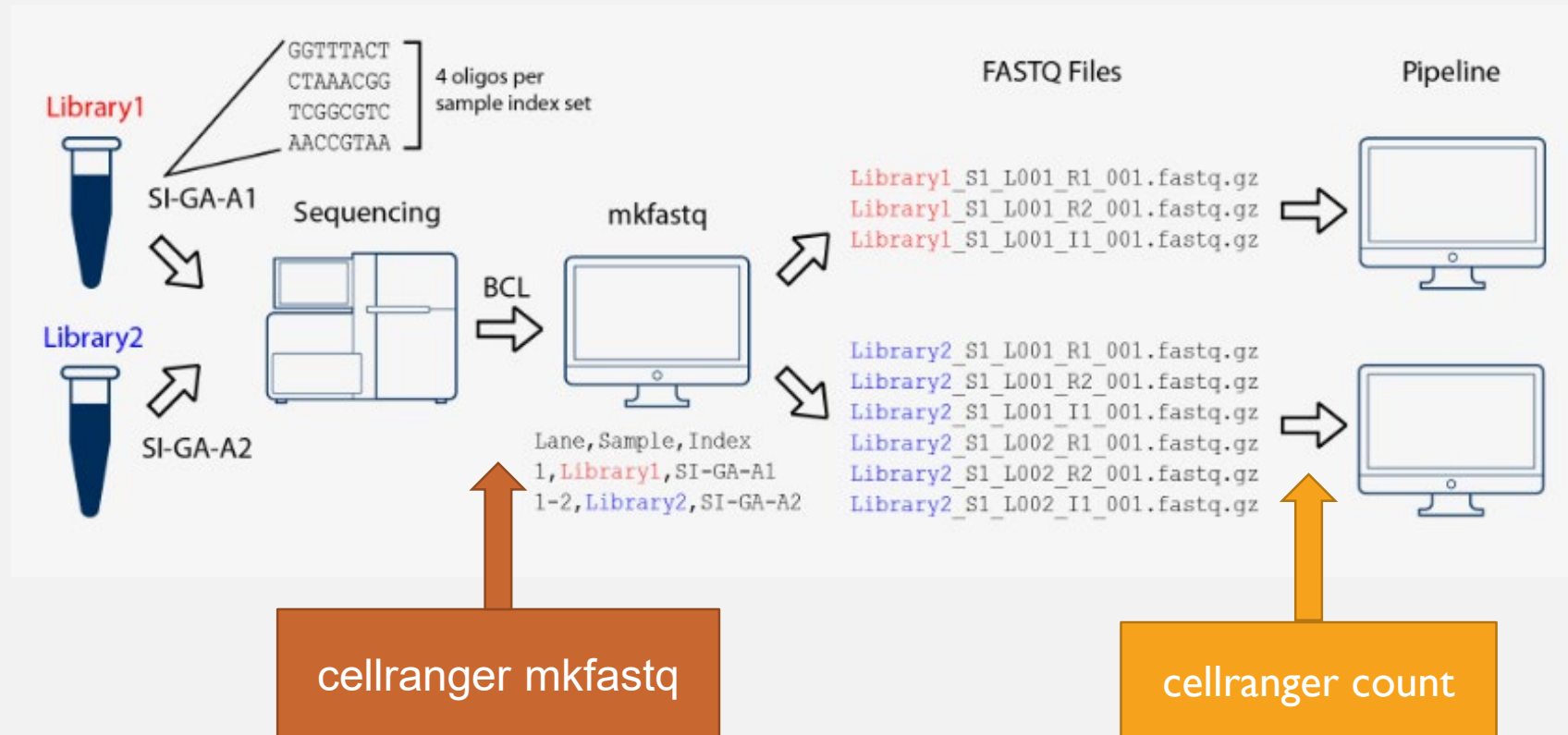
UMI

# Smart-seq2 Pipeline





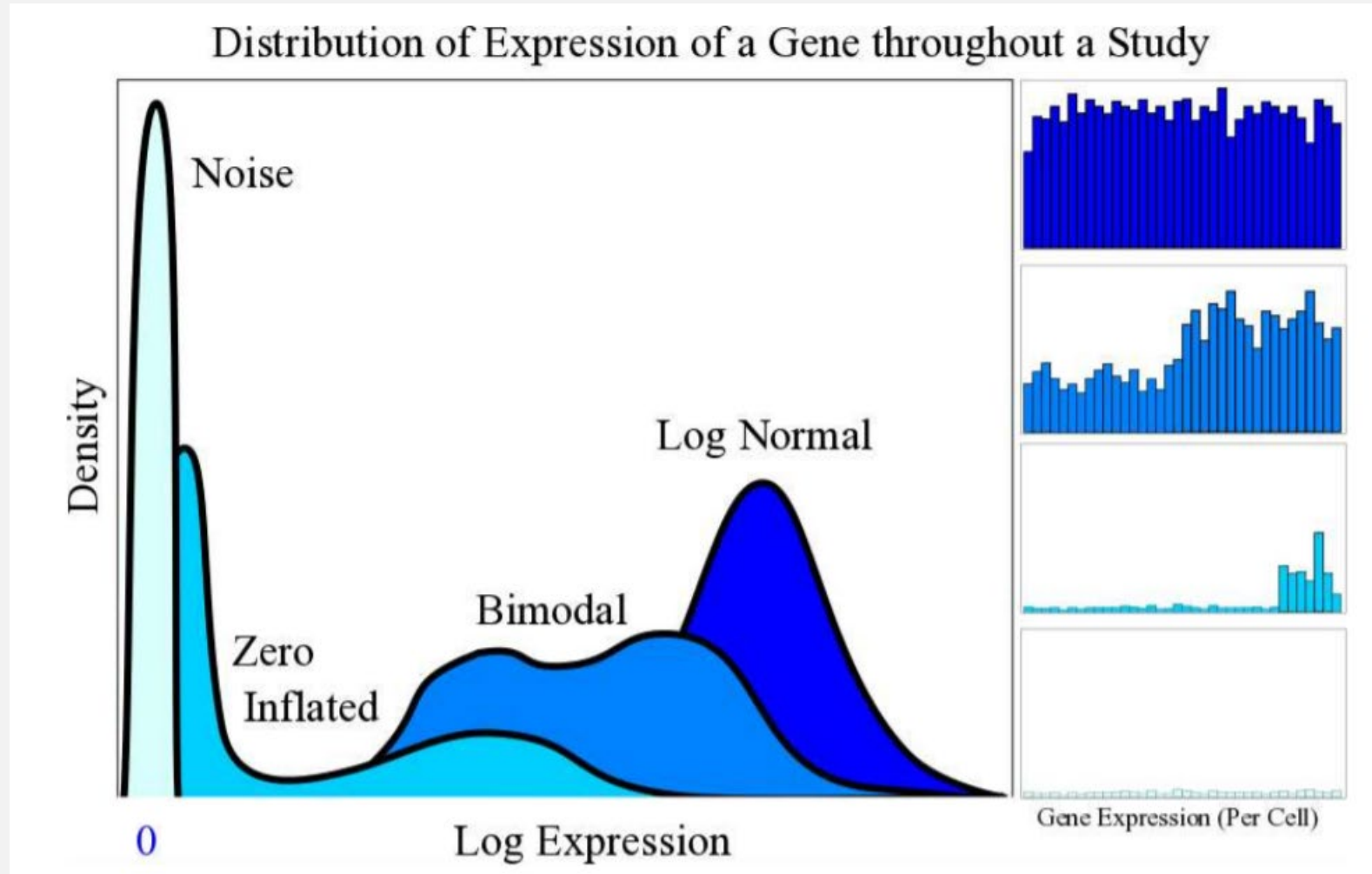
# 10X Genomics Pipeline



# scRNA-seq Study Design

- How many cells should you sequence?
  - HIGHLY dependent on: technology available and/or what type of transcript you are interested in looking at
  - What is your biology and what is the expectation of finding rare populations?
- scRNA-seq is VERY prone to technical batch affects
- Use UMIs or spike-ins
- Satija lab online tool:
  - [How many cells?](#)

# Genes can have different distributions in a scRNA-seq experiment



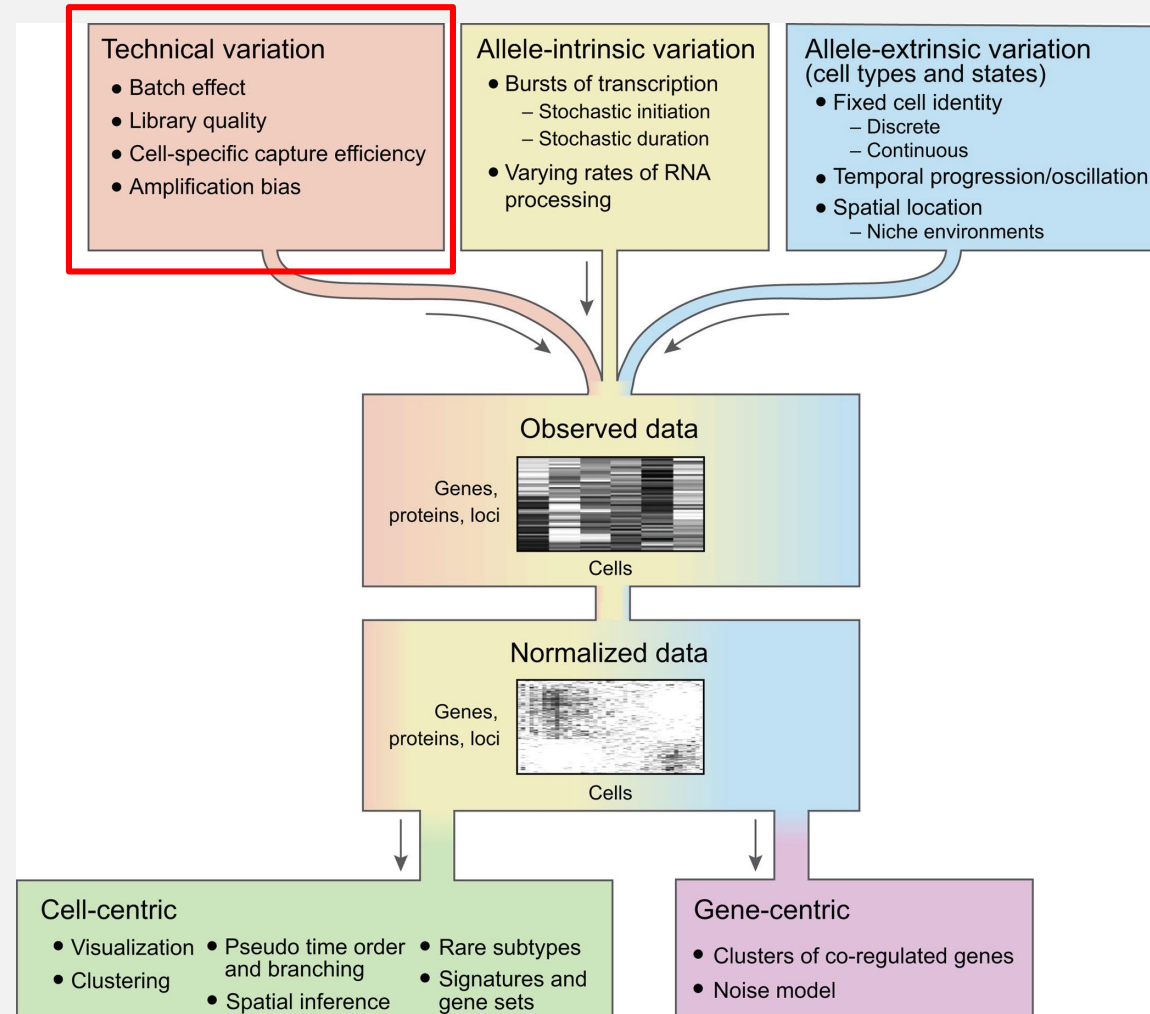
House keeping  
genes

Bimodal (different  
populations)

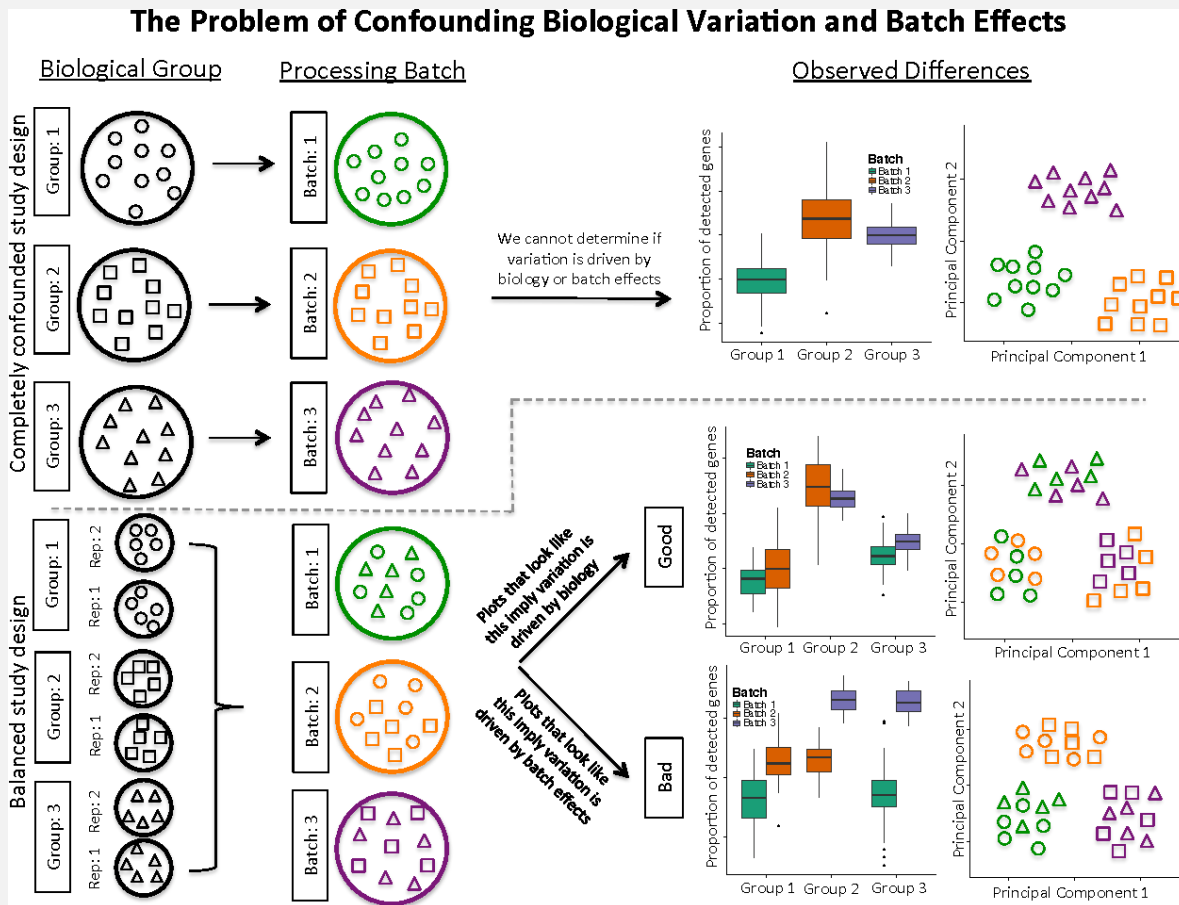
Rare populations

Noise

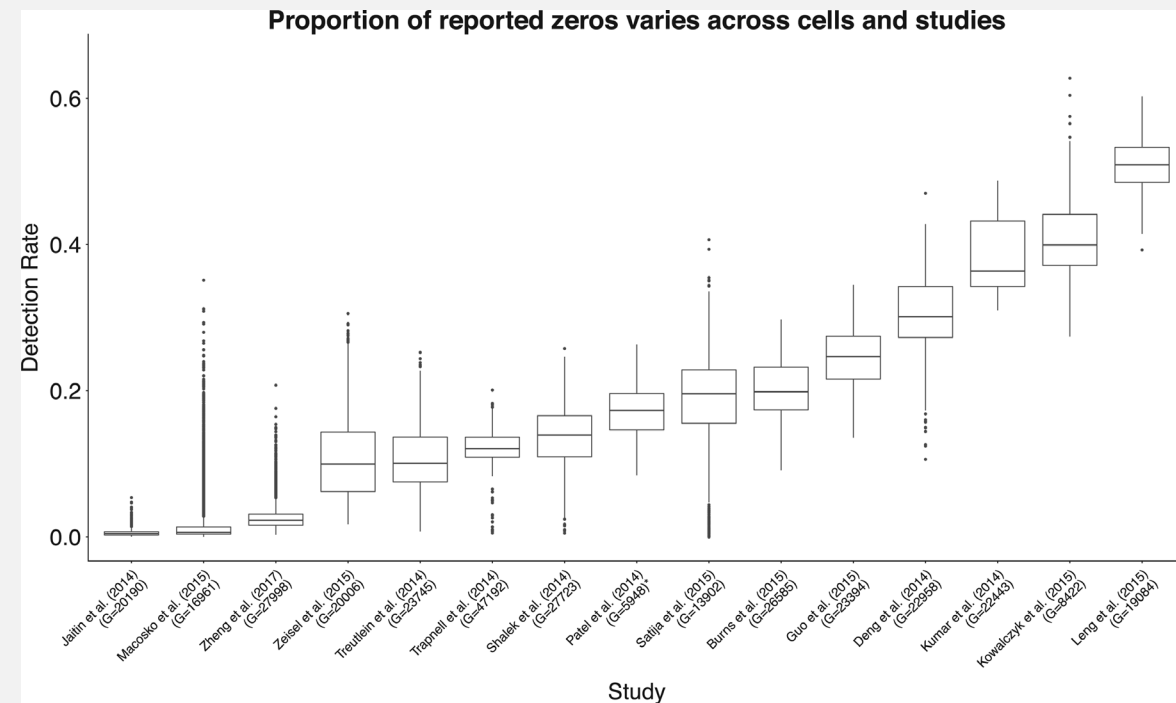
# Expression can have several sources of variation that will affect single-cell



# Batch Effect and “Missing” data



Hicks S., Townes F., Teng M., Irizarry R. **On the widespread and critical impact of systematic bias and batch effects in single-cell RNA-seq data.** *bioRxiv* 2015



Hicks S., Townes F., Teng M., Irizarry R. **Missing data and technical variability in single-cell RNA-sequencing experiments.** *Biostatistics* 2018

# Summary of the data

We are still early in the process in understanding scData and how to apply it:

- Data can be NOT normal
- Data can be Zero-inflated
- Data can be very noisy
- Cells vary in library complexity
- And many other factors

# Analysis pipeline

