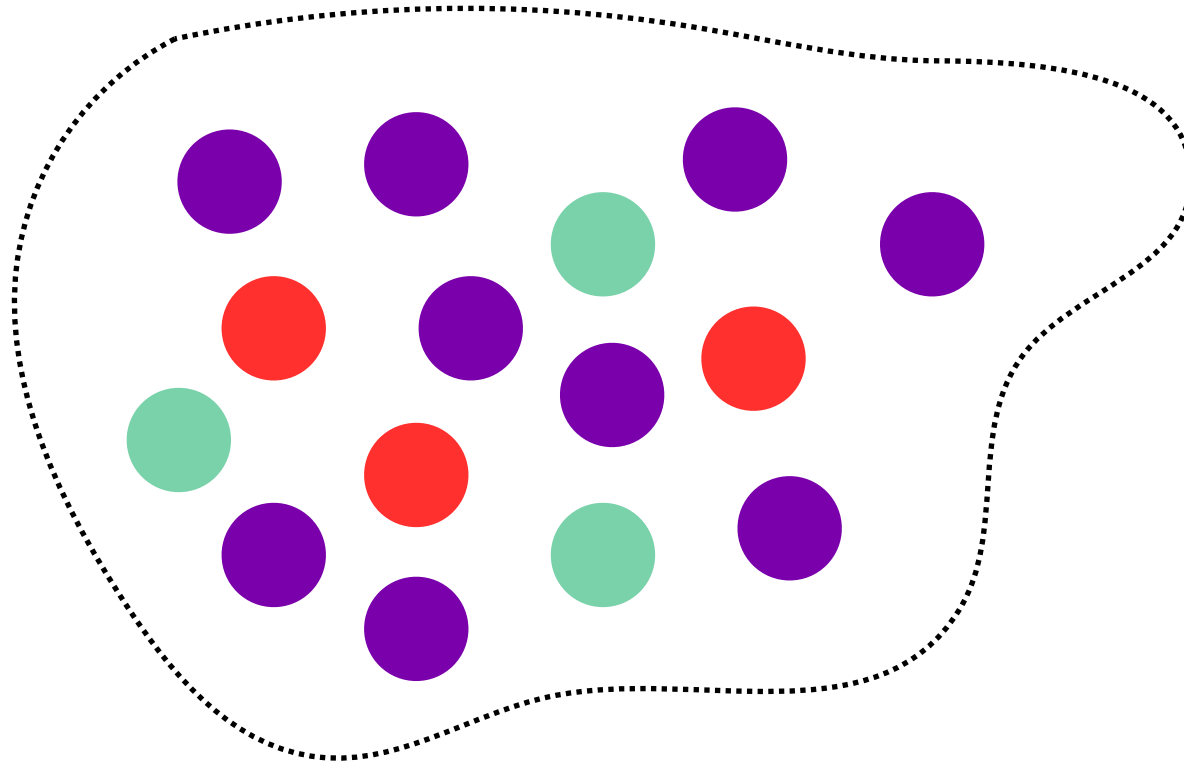


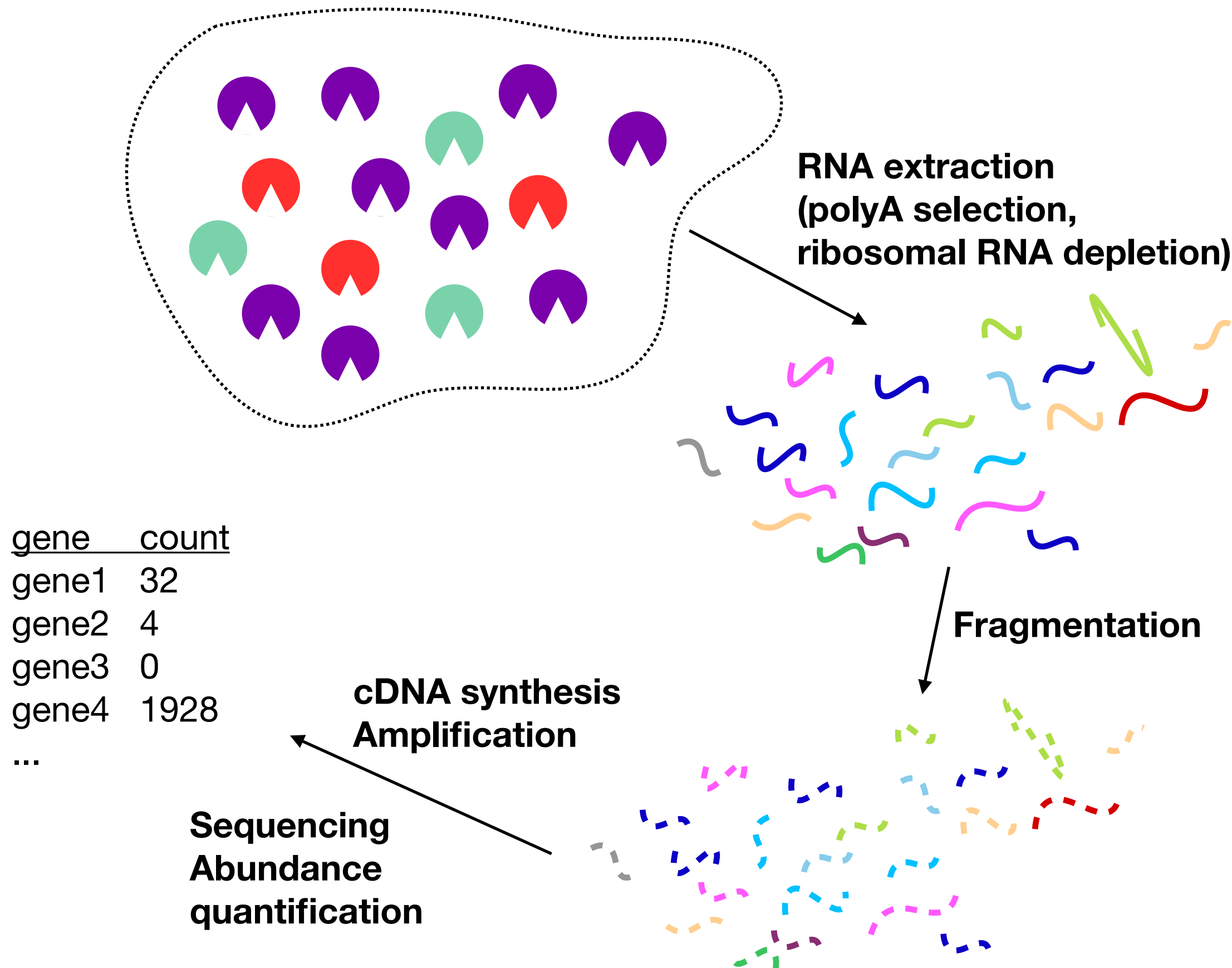
Single-cell RNA-seq (intro)

Charlotte Soneson
CSAMA, Brixen, July 10 2018

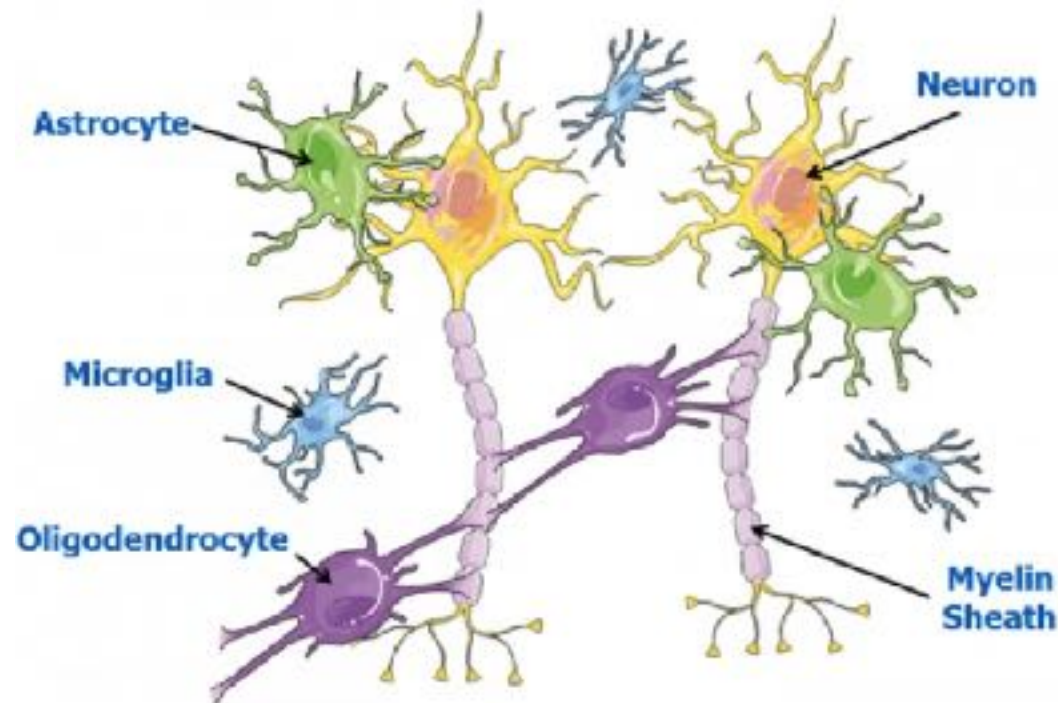
Bulk RNA-seq



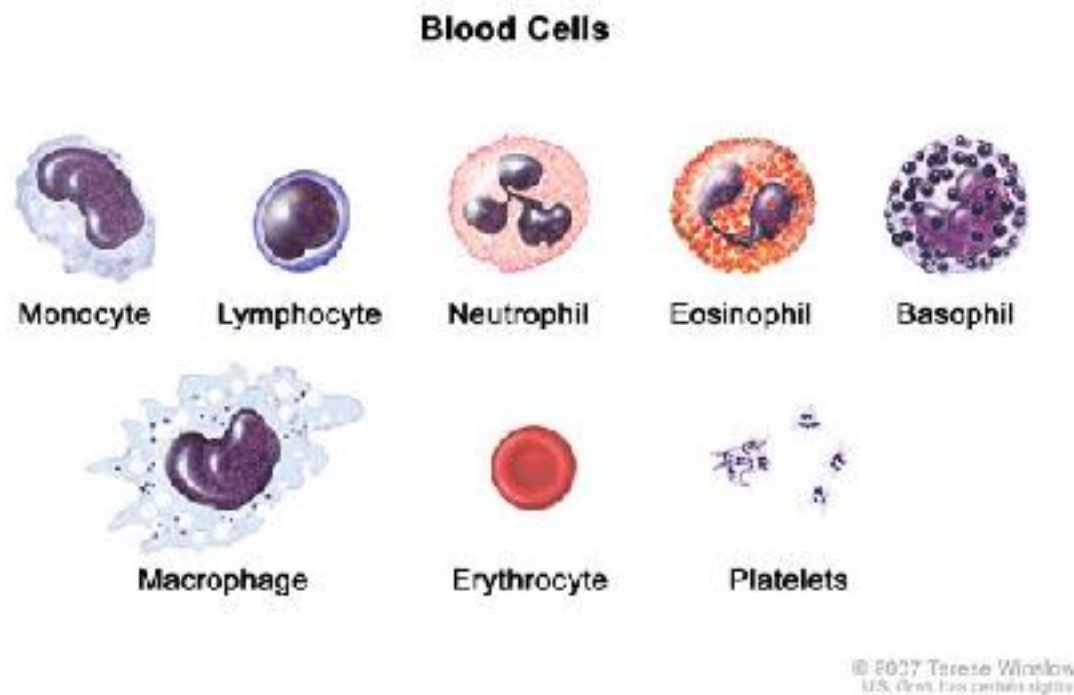
Bulk RNA-seq



There are lots of different sorts of cells



- Wikipedia lists >200 cell types present in the human body
- Cell *type* often refers to more “permanent” aspects of a cell’s identity
- Cell *state* instead reflects aspects arising in more transient processes (differentiation, cell cycle, circadian rhythm)
- Samples are often heterogeneous in terms of cell types/states

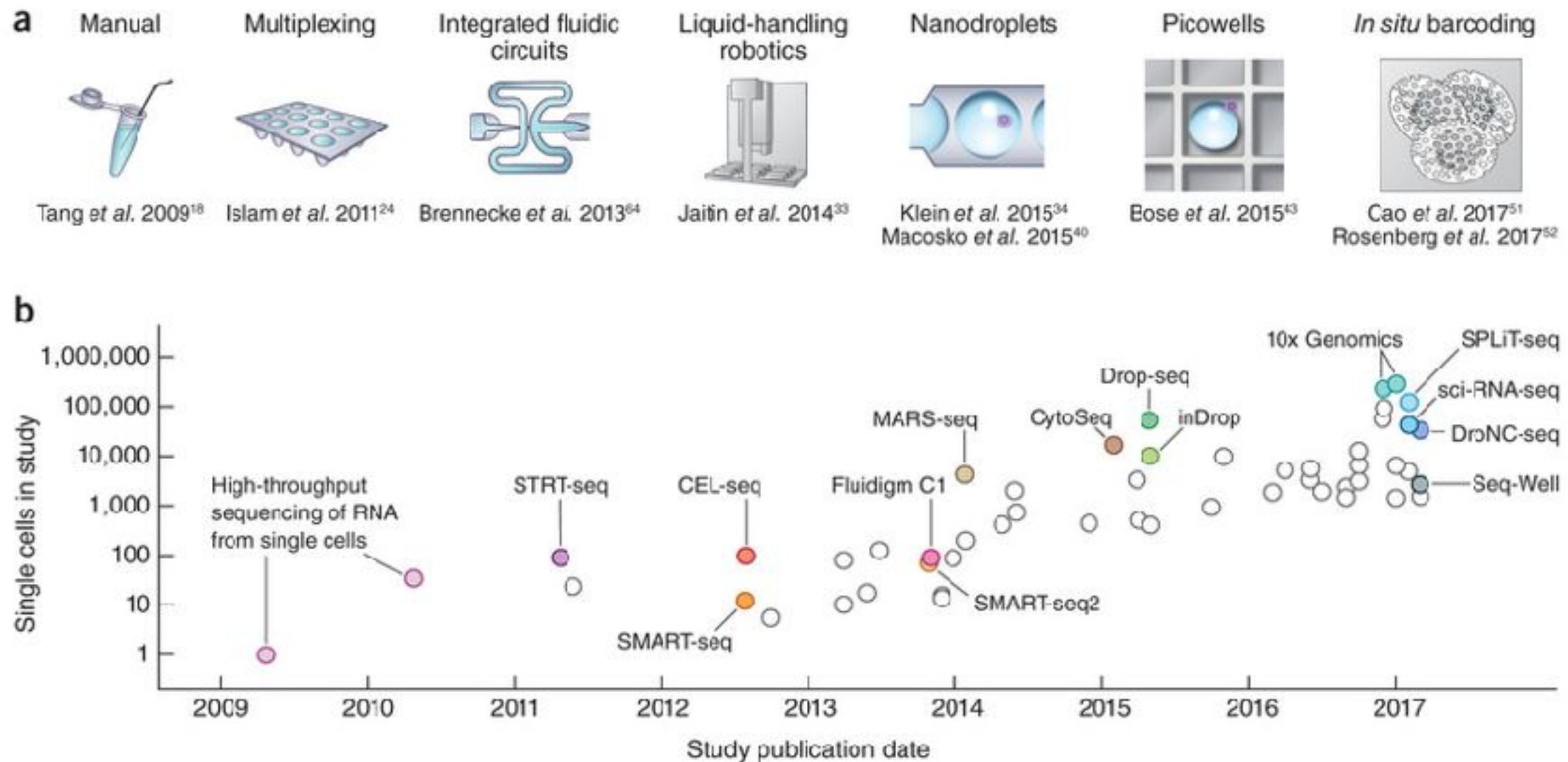


Opportunities with single-cell (RNA-seq) assays

- Quantify abundance of individual cell types, and detect new ones
- Discriminate between differences in gene abundance due to changes in cell type composition, and differences due to changes in the expression levels within a given cell type
- Study the variability (more generally, the distribution) of a gene's expression among cells
- ...

scRNA-seq is not a single assay

- SmartSeq(2)
- CEL-Seq
- 10x Genomics (GemCode, Chromium)
- DropSeq
- STRT-Seq
- MARSseq
- QuartzSeq
- inDrop
- ...



Differences between single-cell protocols

- Throughput (many shallowly sequenced cells or few deeply sequenced ones)
- Full-length coverage vs 3' (or 5') end tag sequencing
- Use of UMIs (unique molecular identifiers) to account for PCR amplification artifacts or not
- Possibilities to incorporate external spike-in molecules for normalization and technical variance estimation

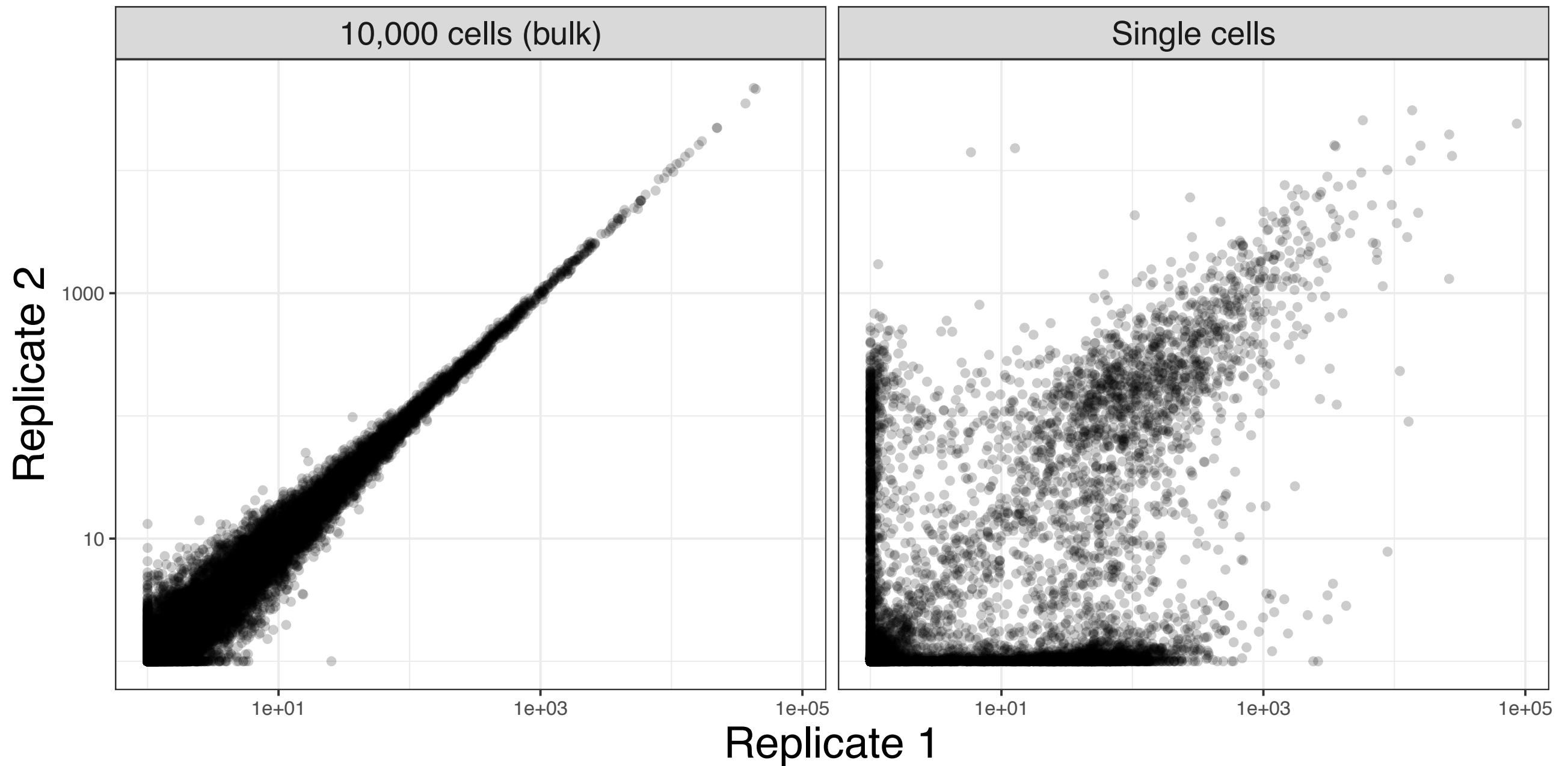
UMIs - accounting for amplification biases

- Low input -> strong amplification needed
- UMIs are short random “barcodes” that are added to transcripts during reverse transcription
- All reads arising from the same transcript molecule will have the same barcode
- Instead of counting reads aligning to a gene, count unique UMIs (“molecules”)

Challenges with scRNA-seq data

- Doublets (can be difficult to identify, depends on number of loaded cells) - single-cell isolation may not be straightforward
- Not all cells are captured for sequencing
- Not all transcripts (only 10-20%) in a cell are captured (reverse transcribed) -> dropouts (zero observed reads for a gene that is actually expressed)
- The same cell can not be sequenced repeatedly to quantify technical variability
- Variability between cells is often high
- Proper experimental design can be challenging

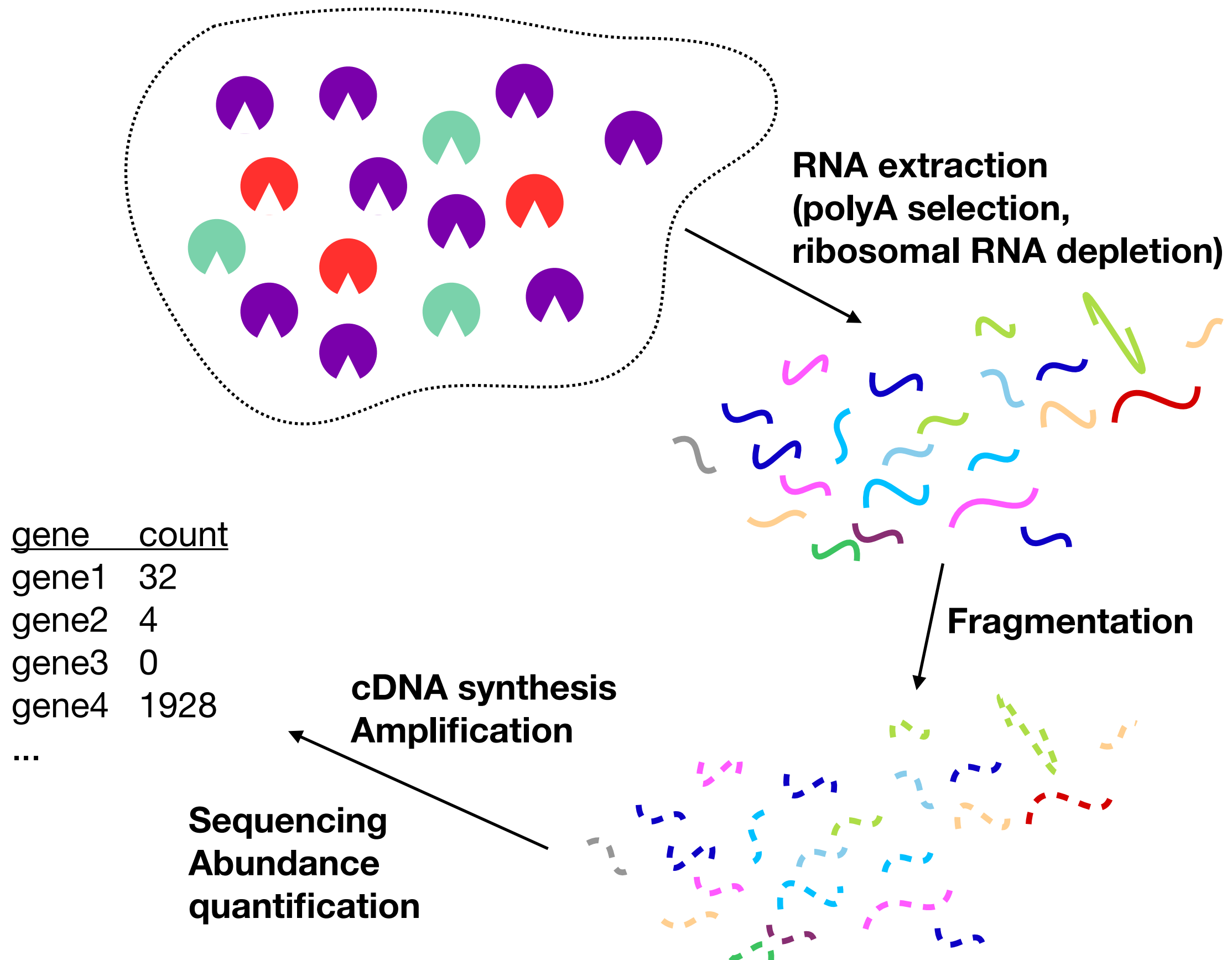
Higher noise level in scRNA-seq than in bulk



Long-read RNA-seq (intro)

Charlotte Soneson
CSAMA, Brixen, July 10 2018

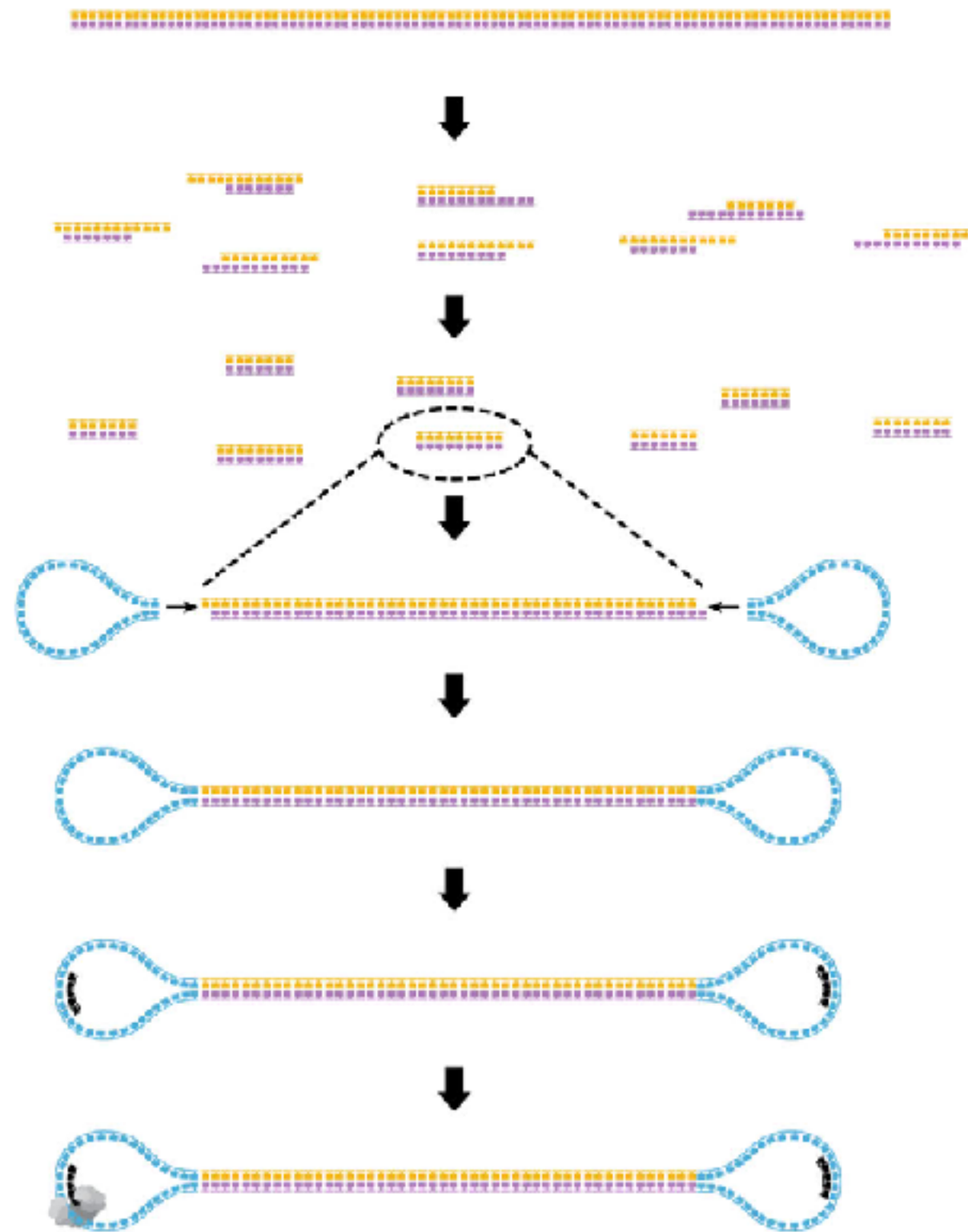
“Bulk” RNA-seq



Why long reads?

- With Illumina sequencing, we need to fragment and amplify the (c)DNA before sequencing
- Application areas for long-read technologies:
 - genome assembly
 - metagenomics
 - transcriptomics
 - DNA/RNA modification detection
- Advantages for RNA-seq: identification of splice isoforms, transcriptome annotation

SMRT Sequencing (PacBio)

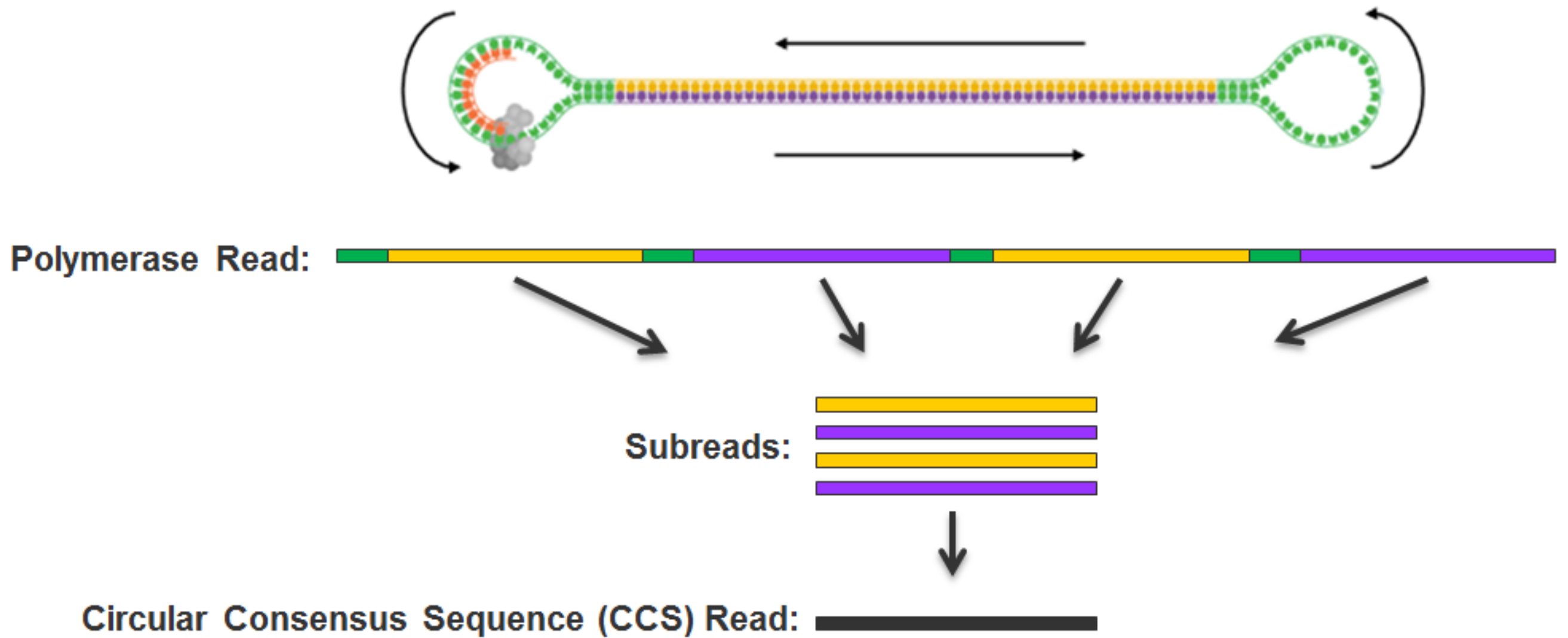


SMRT Sequencing (PacBio)

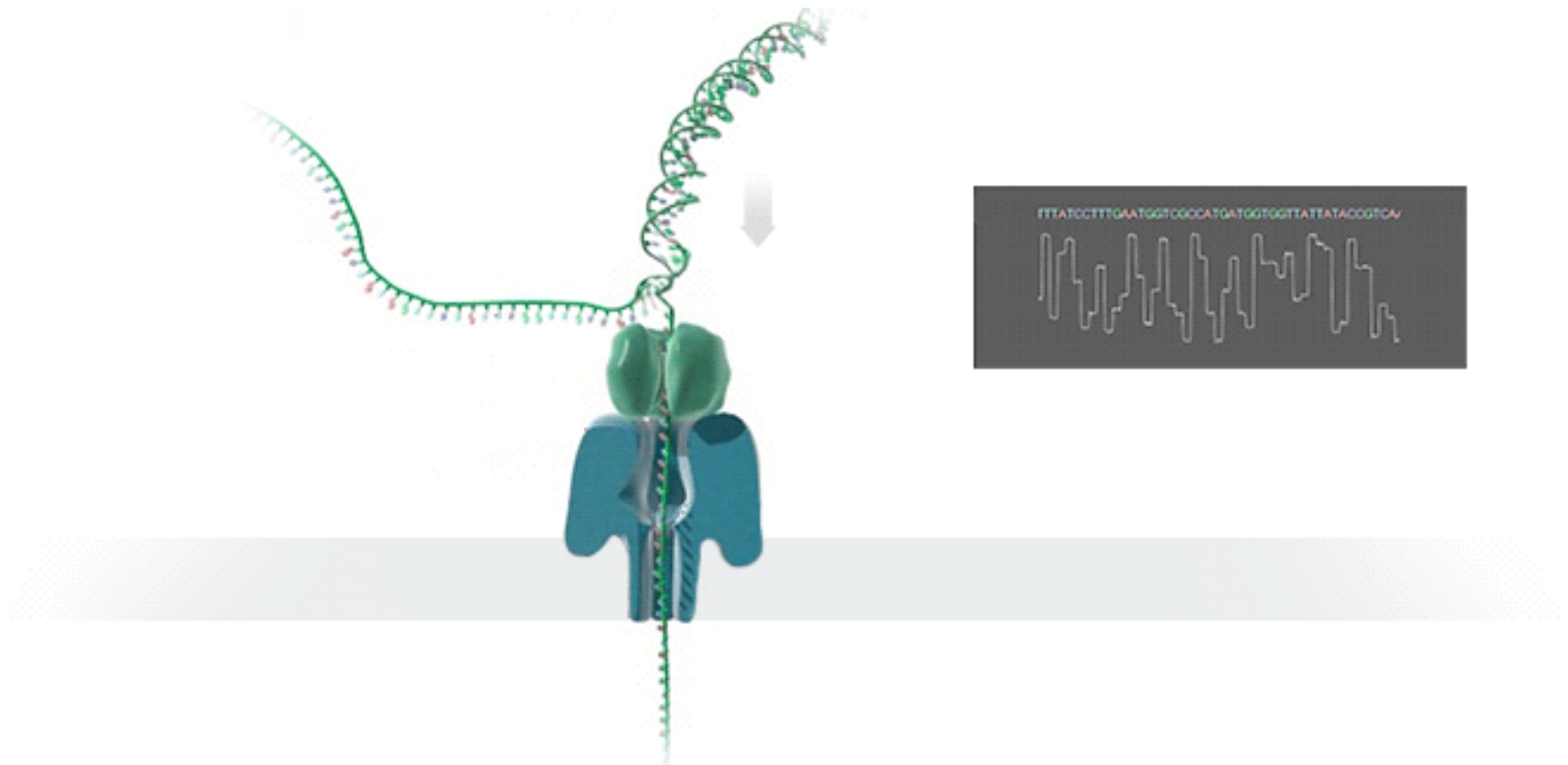


- SMRT cell containing tens of thousands of zero-mode wave guides (ZMWs) - small detection volumes
- ZMWs are illuminated from below
- A DNA template-polymerase complex is immobilized at the bottom of each ZMW
- Phospho-linked nucleotides are introduced in the ZMW
- As a base is held in the detection volume, a light pulse is generated

SMRT Sequencing (PacBio)



Oxford Nanopore sequencing



- Pass DNA or RNA through a nanopore
- The current is changed as the bases pass through the pore

Challenges with long-read cDNA/RNA seq data

- Individual reads have high error rate (nanopore: 2-20%, PacBio subreads: 13-15%, PacBio CCS: 2%)
- Relatively low throughput
- Unequal read length
- Not all reads represent full-length transcript molecules
- How quantitative?