



## Single Cell RNA-Seq at CCR-SF: Sample prep and best practices using 10X Genomics

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# CCR Sequencing Facility



Core facility dedicated to providing Next-Generation Sequencing (NGS) services to Center for Cancer Research investigators, operating one Illumina HiSeq 4000, two HiSeq 2500, three NextSeq 500, one MiSeq, one PacBio RSII / Sequel, 10X Chromium and BioNano Irys system.

DNA/RNA-Protein Interaction



**Illumina**

Transcript Profiling and Discovery

Genome-wide Methylation study

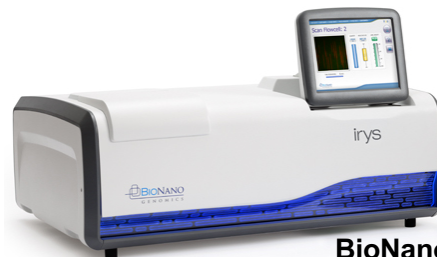
Targeted Exome sequencing

Whole Genome Structural Variants

Single Cell Gene Expression Profiles



**10X Chromium**



**BioNano Irys**

Next-Generation Mapping

De Novo Whole Genome Sequencing

Targeted Sequencing

Base Modifications



**PacBio RS II**

DNA Structure Variation analysis

Transcript Isoform Detection



**PacBio Sequel**



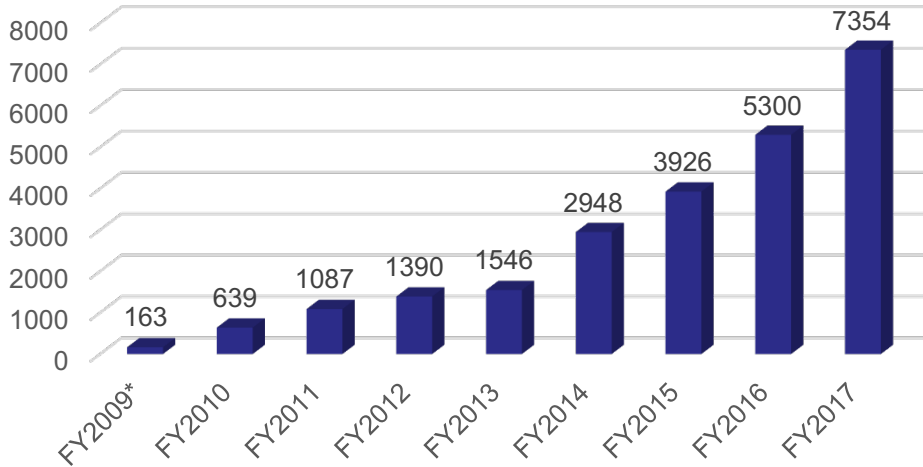
# CCR Sequencing Facility



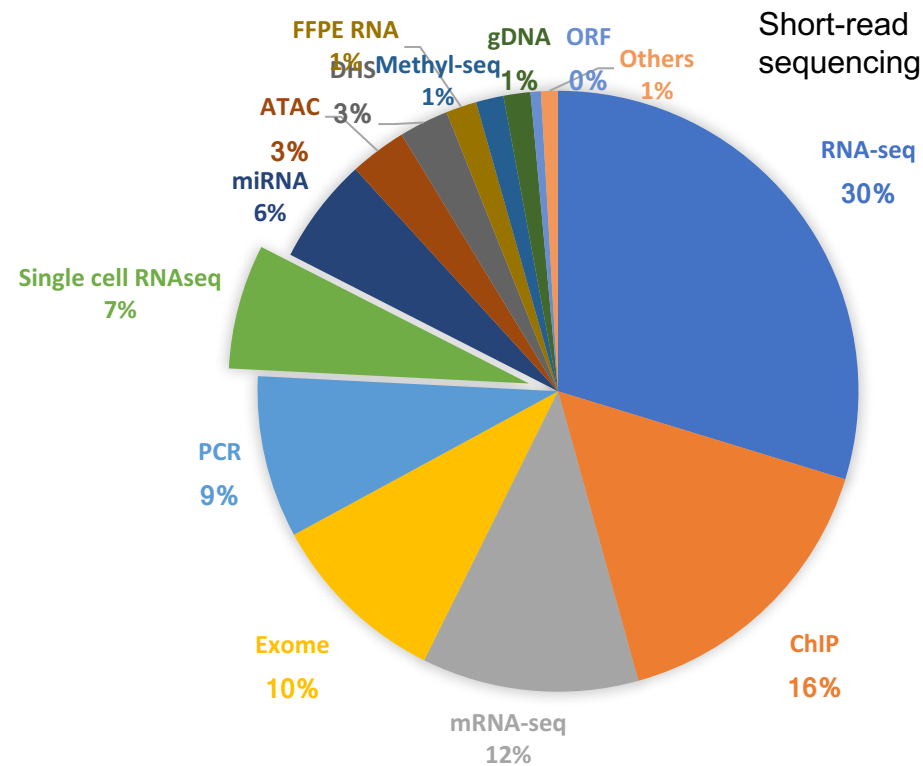
## FY2017 Statistics:

- 120 NCI investigators
- More than 7300 samples.
- Delivered over 45 trillion bases

Number of samples



## Sample types – FY2017



### Single cell RNA-Seq:

Fluidigm C1  
10X Chromium  
DropSeq

### Others

FFPE Exome MNase  
RIP-Seq 10X gDNA  
RiboSeq shRNA  
Drip sgRNA  
HiC VirScan



# Single-cell mRNA-seq



- Cell-to-cell variation in tissues and populations – normal, and important for biological functions, under normal and diseased states.
- Frequently manifested through changes in gene expression.
- Single-cell studies: Help to uncover the variation masked by the average of heterogeneous subpopulations, identification of rare populations.



- Crucial in understanding cancer cells, stem cells, immune cells, cell lineages etc.

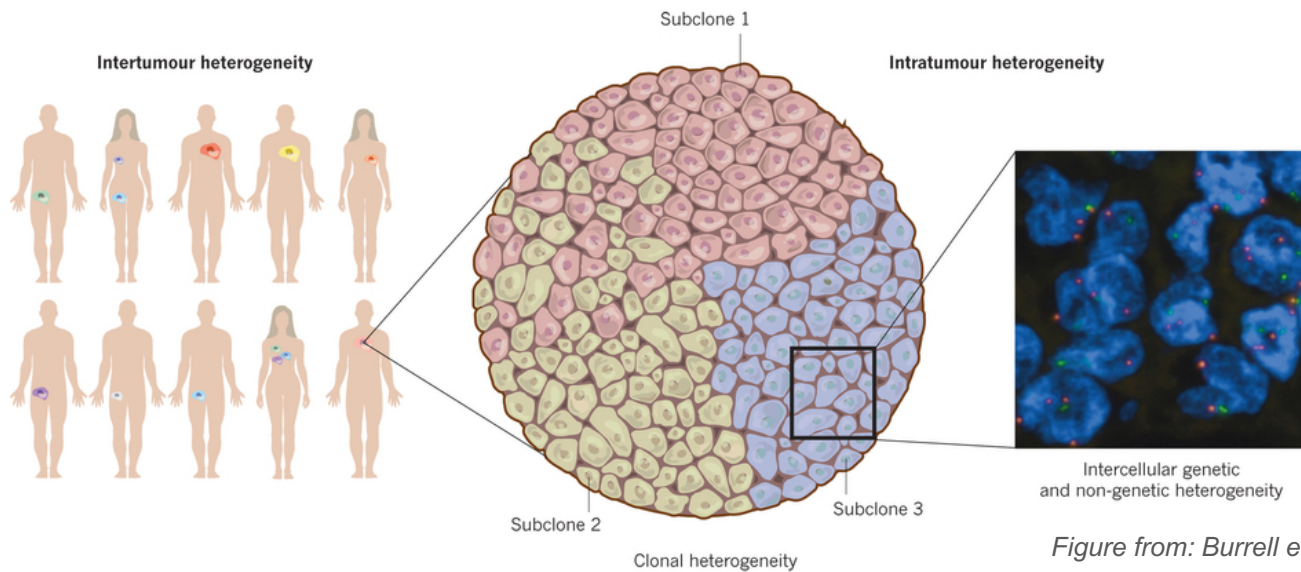


Figure from: Burrell et al., 2013. Nature, 501: 338

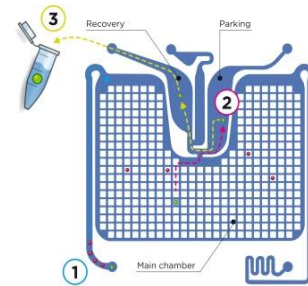
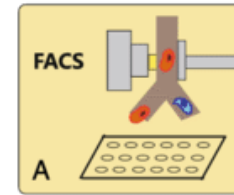


# Single-cell mRNA-seq: Gene expression profiling at single cell level



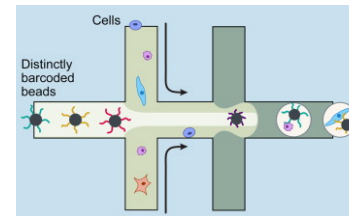
A. Single cell-per-well protocols (low-throughput, full-length transcript analysis possible)

- A. FACS-based methods
- B. Fluidigm C1
- C. DEPArray



B. Droplet-based massively parallel protocols (high-throughput, expression analysis only)

- A. DropSeq
- B. 10X Genomics Chromium
- C. ddSeq
- D. BD Rhapsody

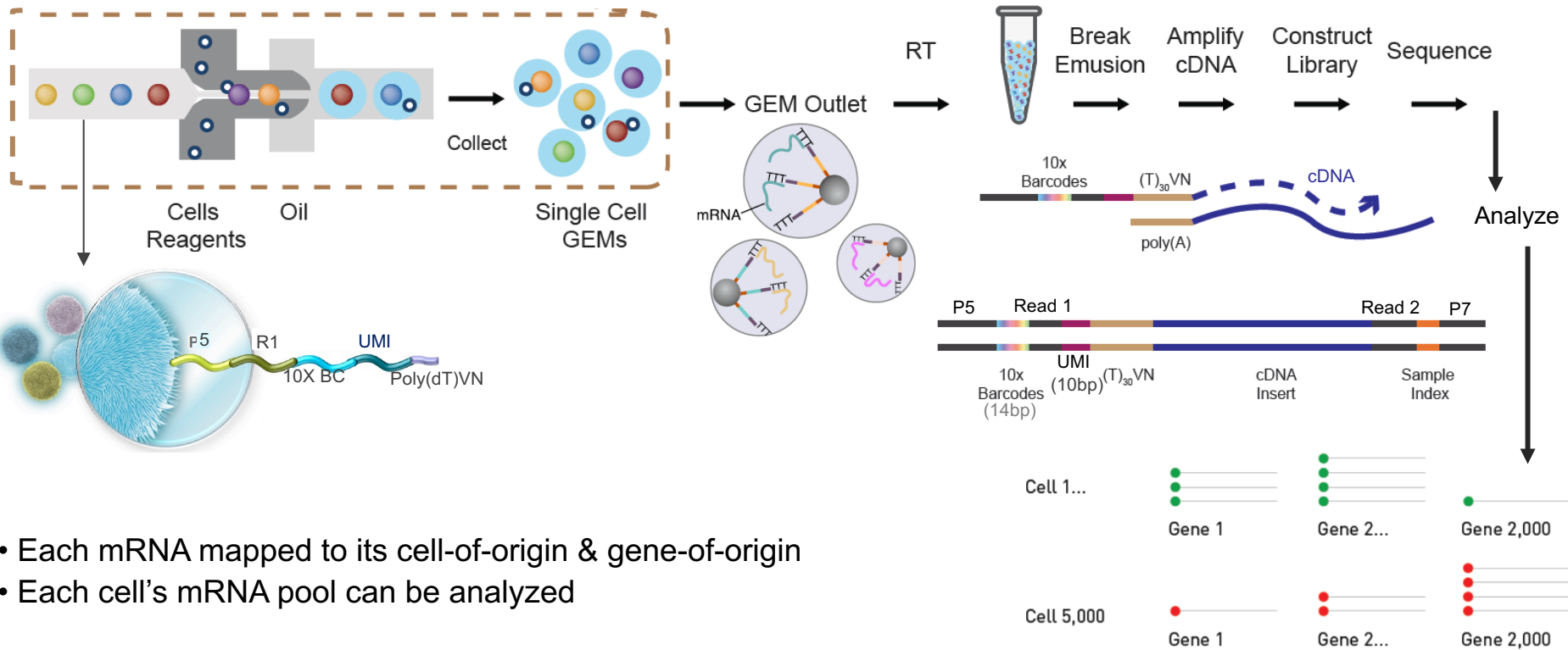


SF: lib prep/sequencing services for all approaches, end-to-end (sample to sequence) service provided for **10XChromium Single-cell RNAseq**.

# Single cell 3' mRNA-Seq using 10X Chromium

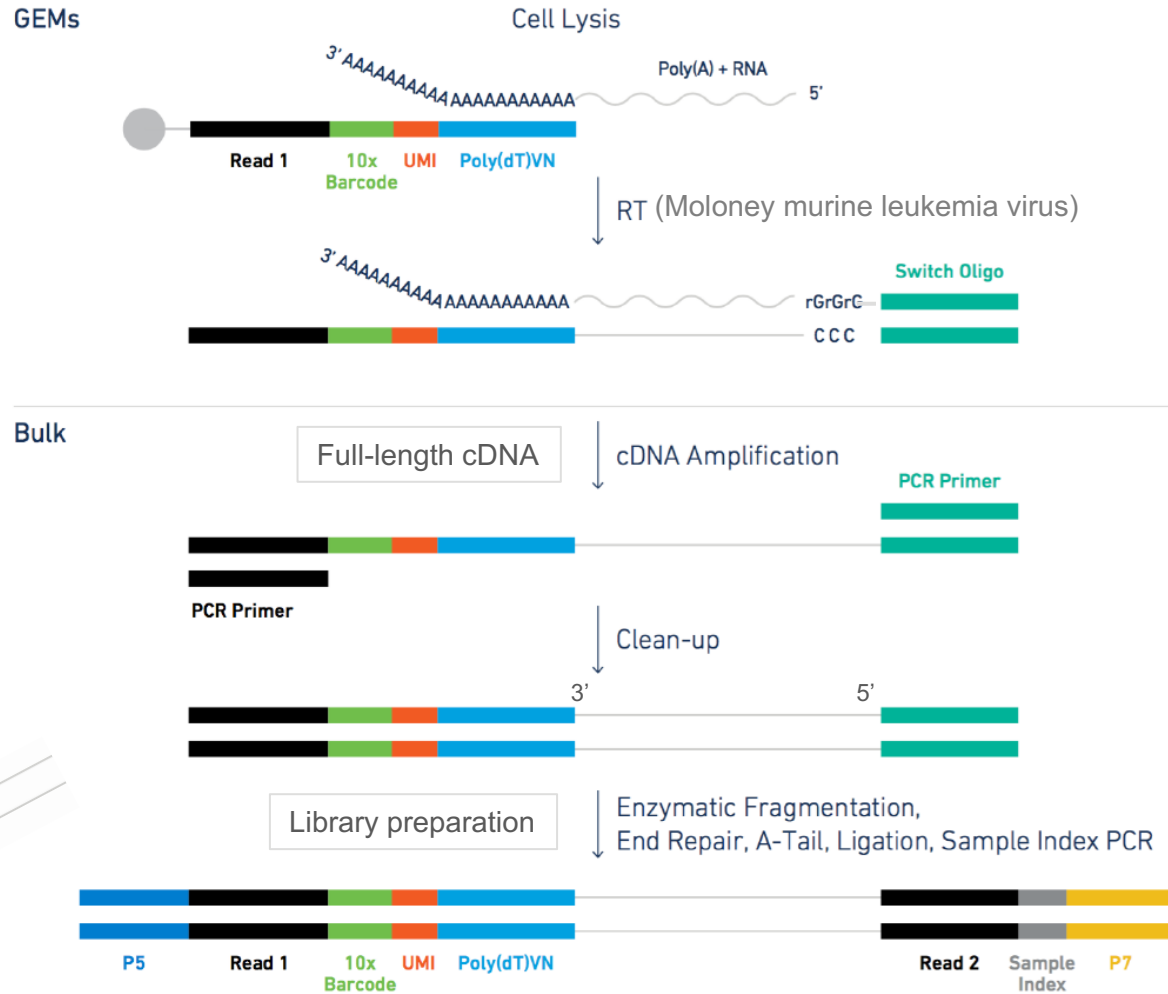


- Based on droplet microfluidics technologies such as Drop-Seq and inDrop
- Combines microfluidics with molecular barcoding to enable high-throughput single-cell RNA sequencing
- Transcriptional profiling of thousands of individual cells by partitioning into nanoliter-scale Gel Bead-In-Emulsions (GEMs)
- Uncovers the complexity of heterogeneous populations.



- Each mRNA mapped to its cell-of-origin & gene-of-origin
- Each cell's mRNA pool can be analyzed

# Generation of full-length cDNA using template switching and 3' sequencing



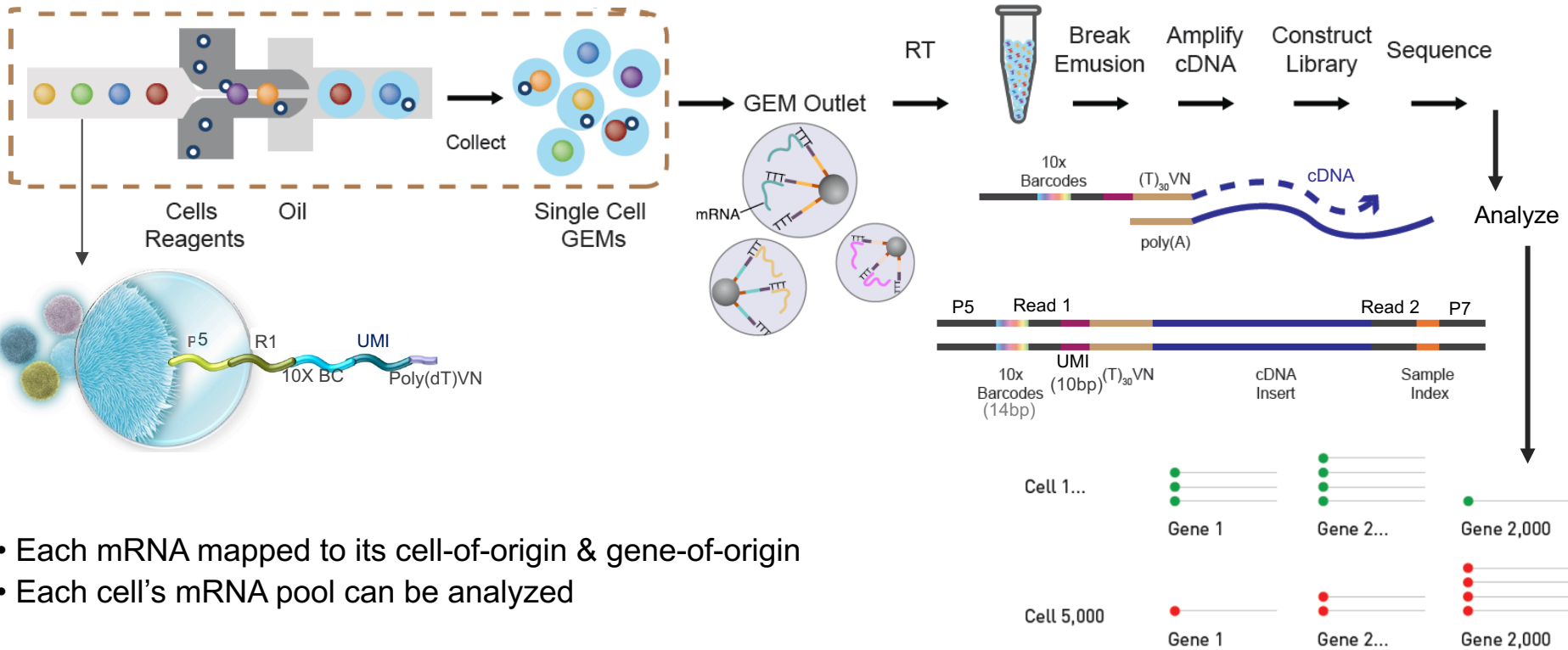
Sequencing adaptors, Read 1 and Read 2, Barcodes and Sample Index



# Single cell 3' mRNA-Seq using 10X Chromium



- Based on droplet microfluidics technologies such as Drop-Seq and inDrop
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- Transcriptional profiling of thousands of individual cells by partitioning into nanoliter-scale Gel Bead-In-Emulsons (GEMs)
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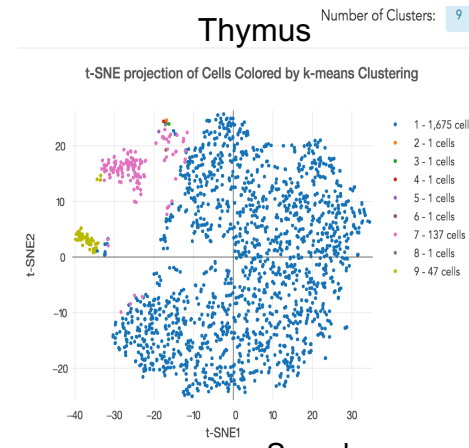
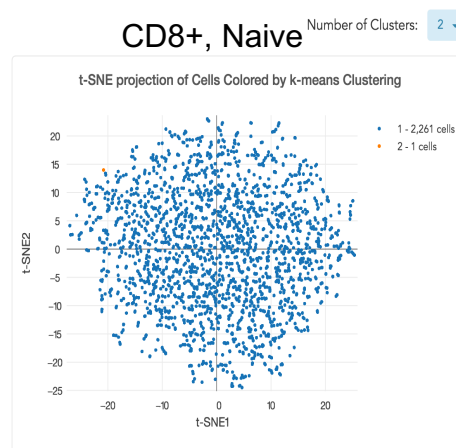
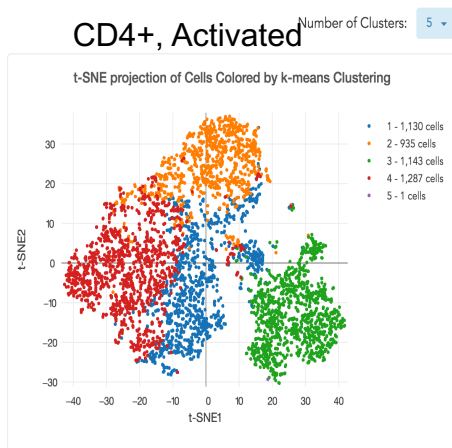
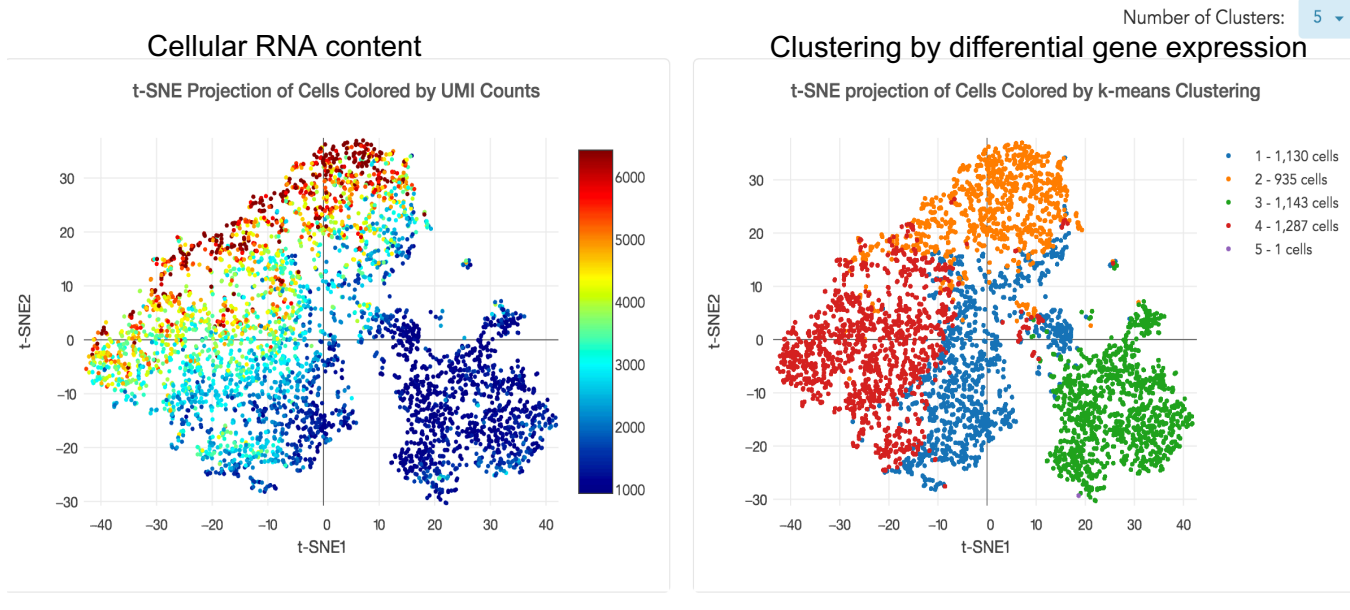


- Each mRNA mapped to its cell-of-origin & gene-of-origin
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# Single cell 3' mRNA-Seq using 10X Chromium



## Examples of clusters generated by 10X Chromium

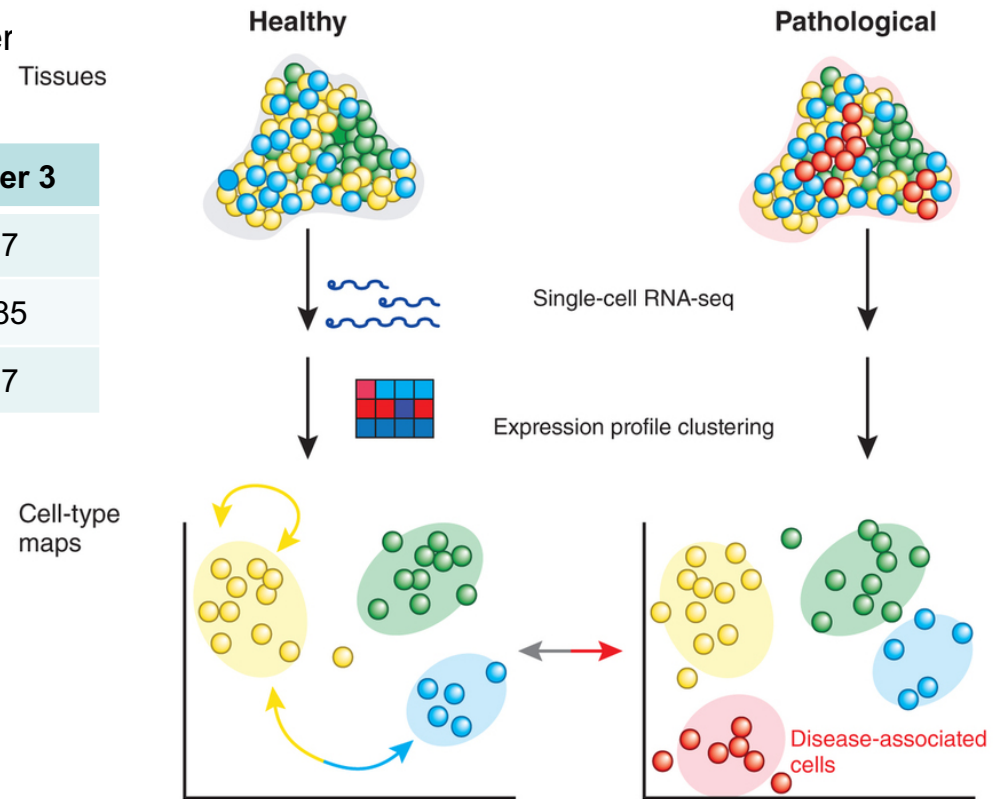


# Single cell 3' mRNA-Seq using 10X Chromium




Top differentially expressed genes per cell cluster  
(UMI counts/cell) Tissues

Gene name	Cluster 1	Cluster 2	Cluster 3
Abc1	28.52	0.03	14.7
Xyz2	4.56	8.33	30.85
Fgh3	8.94	17.44	1.27




Types of analyses

- 


**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



# 10X Chromium single-cell projects at CCR-SF



## 10X Chromium single cell projects at SF

Number of labs	13
Total number of samples processed	123
Data delivered	101
Total number of cells sequenced	145,982
Median number of cells / sample	1020
Median genes / cell	1,475
Median transcripts / cell	3,650
Total genes detected / sample (Median)	14,914

- Data analyzed by CCR-SF IFX team

## Different cell types processed at CCR-SF:

1. Human and mouse **cell lines** (HeLa cells and MEFs)
2. **B cells** (naïve and memory B cells)
3. Cells from the **dermis** layer of **skin**
4. **T cells** (naïve and activated, CD4+, CD8+, & thymus cells)
5. **Stem cells** (bone marrow cells)
6. Cutaneous **MAIT** (mucosal-associated invariant T) cells
7. **CAR-T** cells (cultured, activated)
8. **Epithelial cells** trypsinized from mouse **skin**
9. **Tumor cells** (**Prostrate cancer**) from organoid cultures
10. Cells from **patient urine** samples
11. **Hepatic cancer** patient-derived **frozen tumor** cells
12. **Thymic epithelial cells** from WT and mutant mice
13. **Neuronal cells** (cultured)
14. **Single nucleus sequencing** from frozen **post-mortem brain** tissues.

# How many cells?

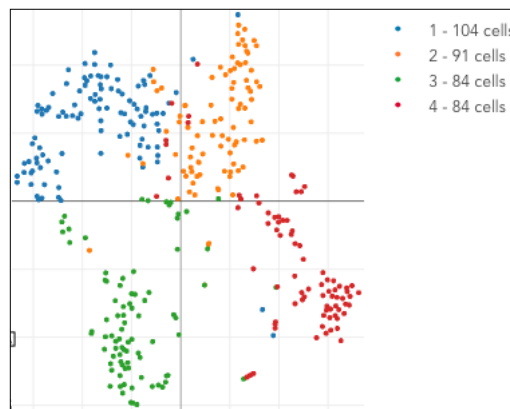
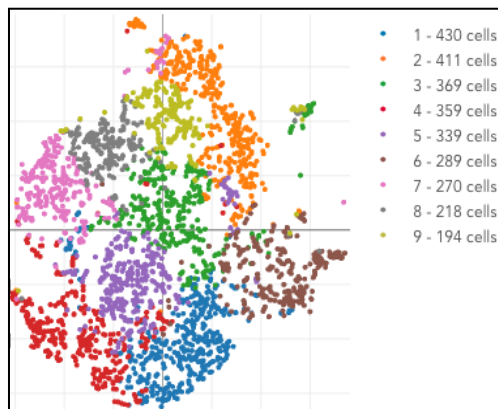
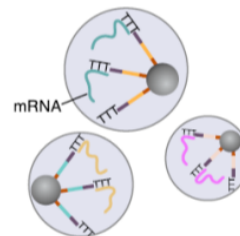


~ 10-15% transcripts captured per cell

- PolyA capture
- RT conversion within the cell lysate

More cells – higher likelihood of more representative data

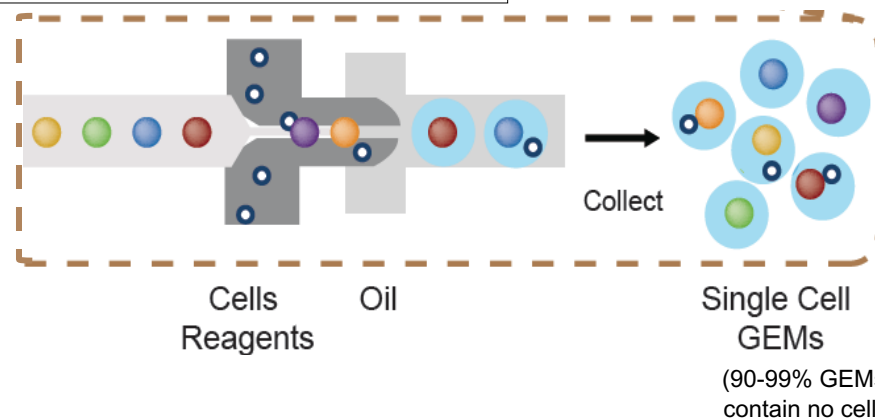
More cells – more confidence in heterogeneity analysis



More cells – more probability of multipllets

10X Chromium:

- ~65% cell processing efficiency
- 500-10,000 cells per channel, ~0.4-7.6% doublet rate



# How many cells?



Multiplet Rate (%)	# of Cells Loaded	# of Cells Recovered*
~0.4%	~870	~500
~0.8%	~1700	~1000
~1.6%	~3500	~2000
~2.3%	~5300	~3000
~3.1%	~7000	~4000
~3.9%	~8700	~5000
~4.6%	~10500	~6000
~5.4%	~12200	~7000
~6.1%	~14000	~8000
~6.9%	~15700	~9000
~7.6%	~17400	~10000

\* For good quality samples

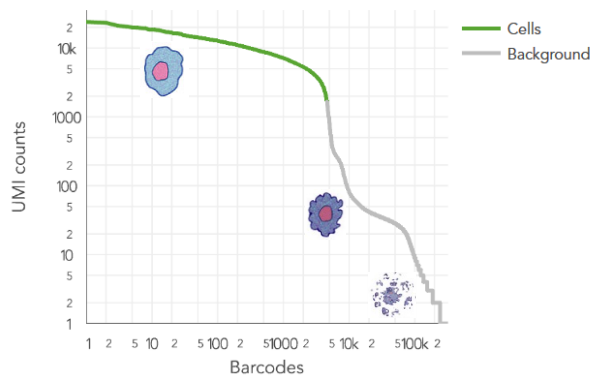
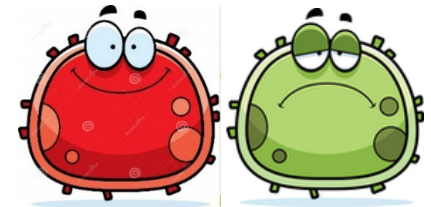
- Accurate cell count
- Single cell suspensions – no aggregates
- Enough cells for washes – removal of debris, nucleic acids, RT inhibitors



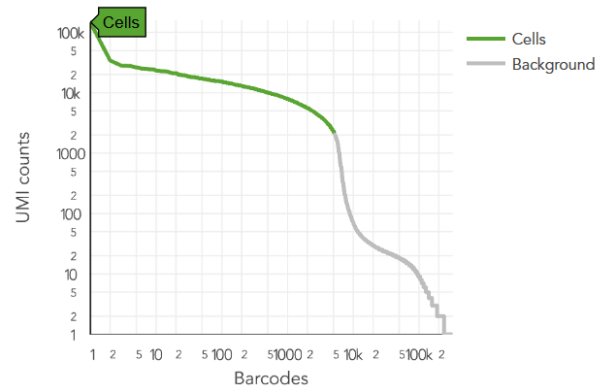
# Quality matters!



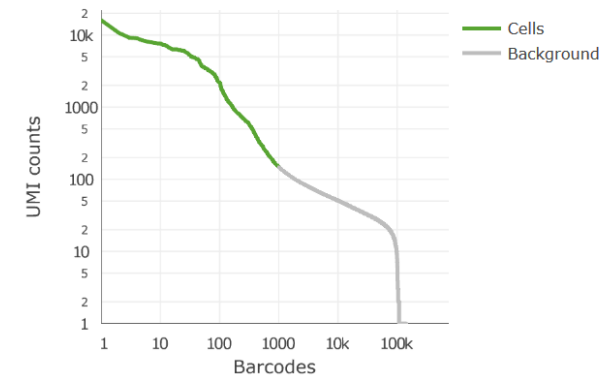
- Low viability / dead cells
  - mRNA degraded, excluded from analysis – data from fewer cells
  - mRNA enters other GEMs – incorrect data
  - Unhealthy/dying cells may interfere with subpopulation analysis
- Ideal viability > 90%; >70% is good
- Fresh cells > Frozen; some cell types more susceptible to degradation
- Removal of dead cells, if enough number of cells available



Estimated number of cells: 4,601  
Estimated viability: ~87%



Estimated number of cells: 5,234  
Estimated viability: ~90%



Estimated number of cells: 1,011  
Estimated viability: ~20%

- **Tips: Gentle handling of the cells; low speed spins, wide bore tips, maintain cells in ice, shorter processing time, no RT inhibitors**

# Sequencing Depth

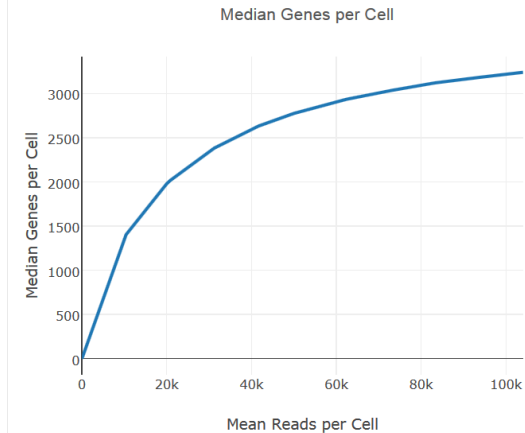
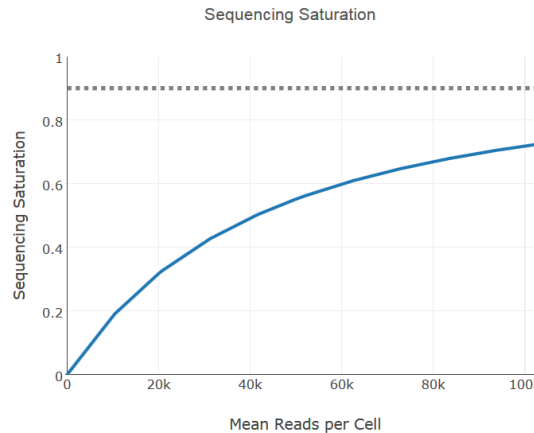


Depends on

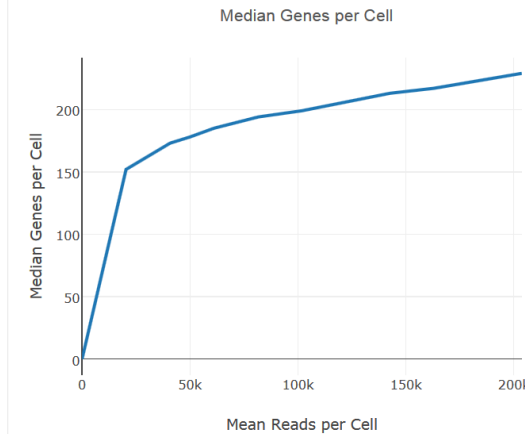
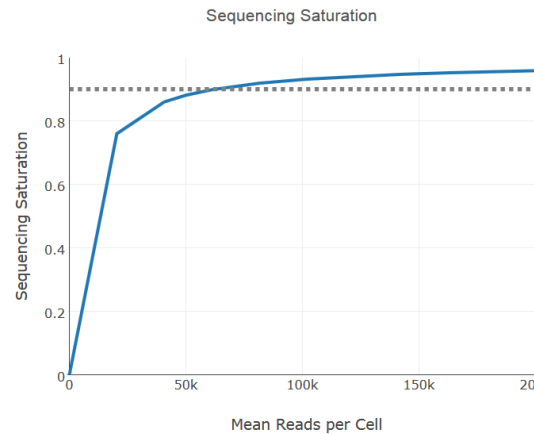
- Cell type: More RNA vs less RNA per cell
- Cell number: More cells, more reads
- Cell quality: More healthy cells, more reads

Typical: 100,000 reads per cell

- 2-3 samples per NextSeq lane (SR)



Estimated number of cells: 2,355  
Median genes per cell: 3,241



Estimated number of cells: 989  
Median genes per cell: 484

# Tips

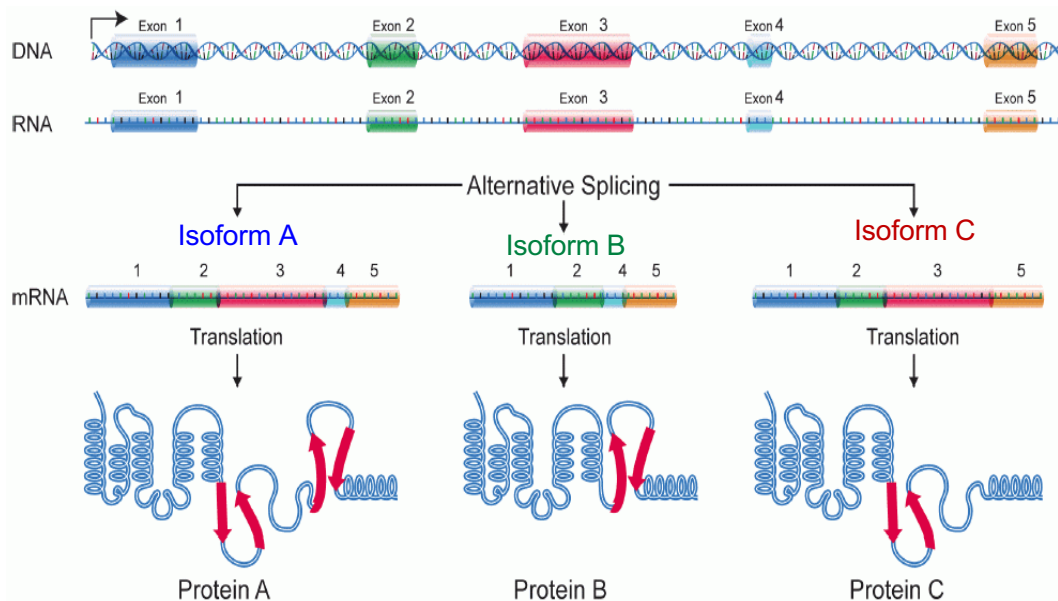


- Cells should have high viability
- Remove dead/unhealthy cells, if possible
- Cell number: enough to enable at least a couple of washes and cell counting before loading
- Cell handling: as gentle as possible. Use low speed spins, wide bore tips
- Minimize cell stress: maintain cells in ice
- Processing time: as short as possible. Capture as quickly after extraction as possible.
- If delays are inevitable, keep cells in the medium they are happy in
- Fresh cells better than frozen.
- Single cell suspensions should be free from cell aggregates and debris
- No RT inhibitors in the cell suspensions
- Accurate cell count

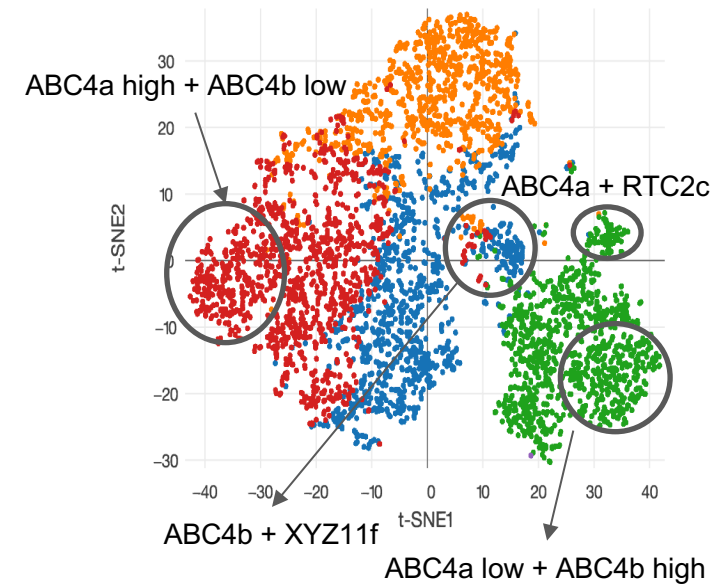
# Coming soon – Full-length RNA-seq from single cells



**Alternate Splicing:** Increases informational diversity, regulates various cellular processes, commonly dysregulated in cancer, thought to result in functional diversity of most cancer-relevant genes.



Subpopulation 1 (20% cells): *ABCa* + *ABCb*  
Subpopulation 2 (50% cells): *ABCa* + *ABCc*  
Subpopulation 3 (30% cells): *ABCb*



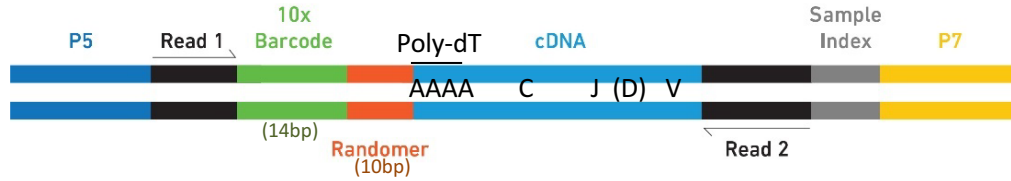
- Using long-read sequencing technology to obtain full-length sequence and isoform information from single cells

# Coming soon – Analysis of V(D)J rearrangement in B- and T-cells

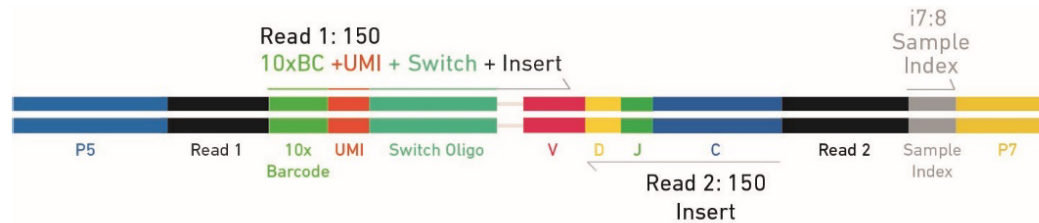


- Single cell sequencing to reveal the true clonality and diversity of the immune repertoire
- Simultaneous analysis of a single sample for cellular heterogeneity and phenotype, as well as T-cell receptor and B cell immunoglobulin repertoires.

## 3' Gene Expression Library Structure:



## V(D)J Enriched Library Structure:



## 5' Gene Expression Library Structure:

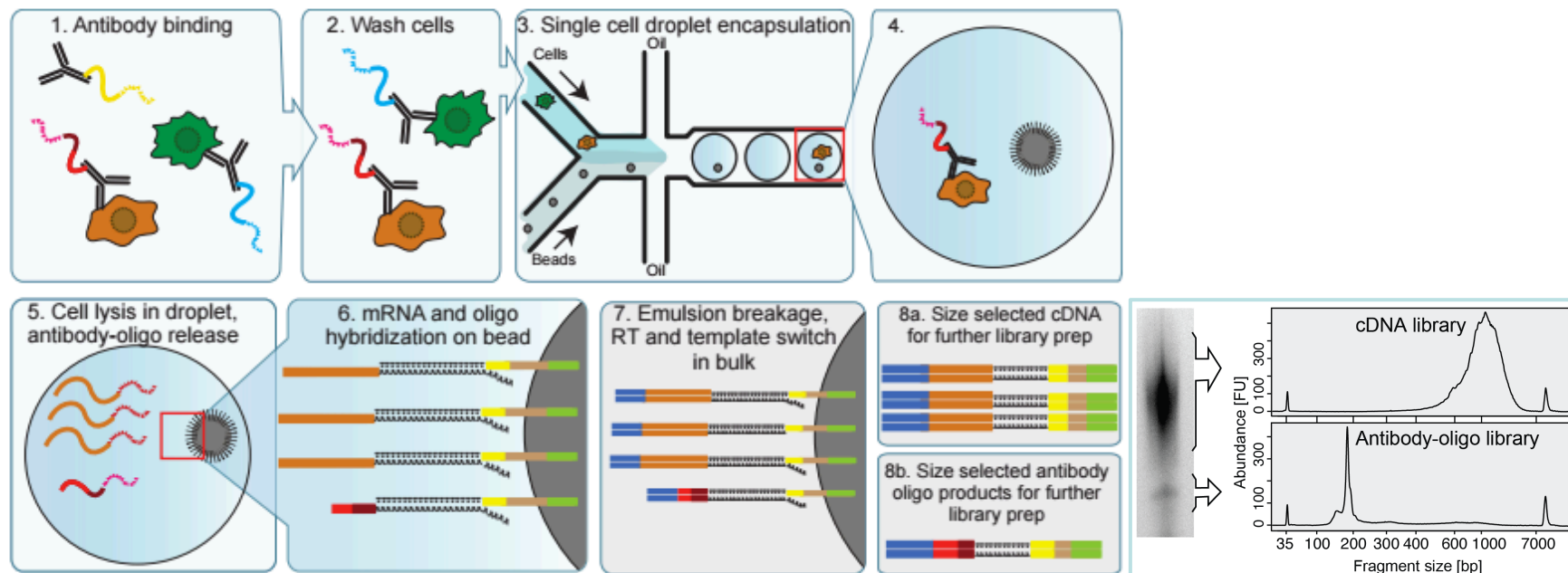




# Coming soon – CITE-seq (Cellular Indexing of Transcriptomes and Epitopes by sequencing)



- Couples measurement of surface protein markers on thousands of single cells with simultaneous mRNA sequencing - NYGC
- Correlation between mRNA and protein levels
- Clustering of cells with similar transcriptional profiles

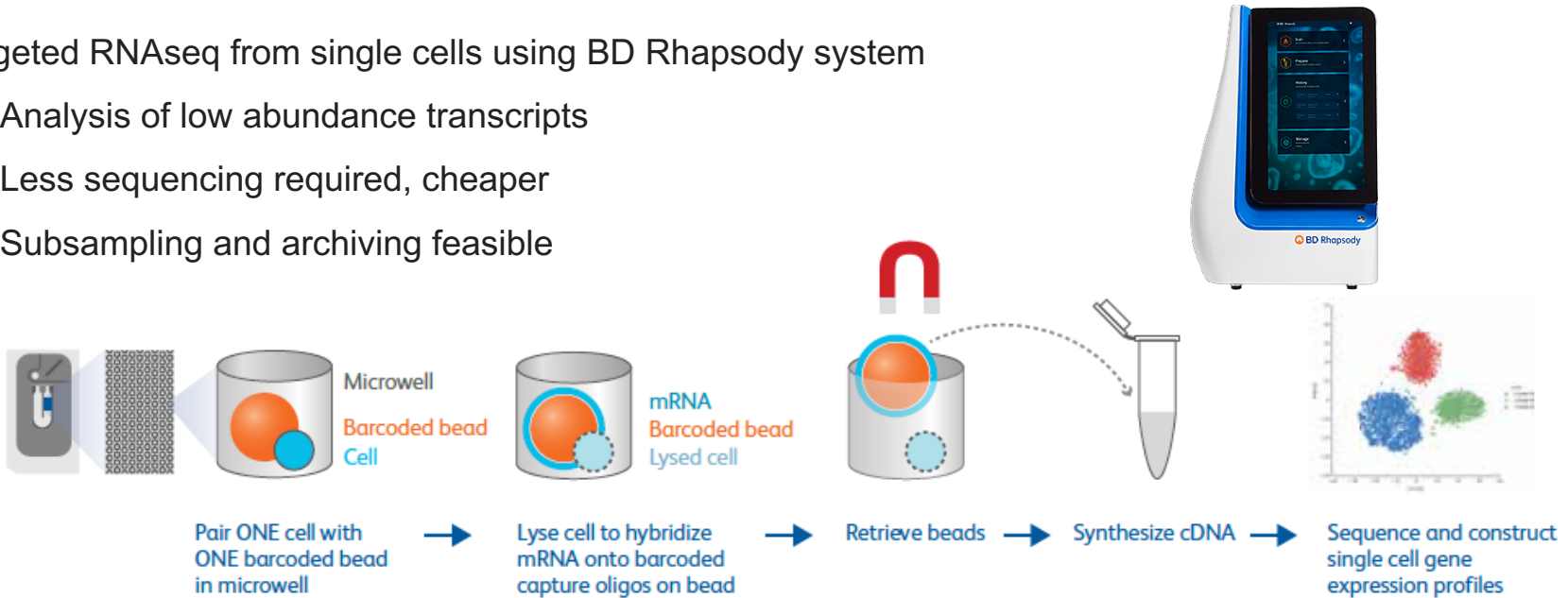


# Coming soon – targeted RNAseq from single cells

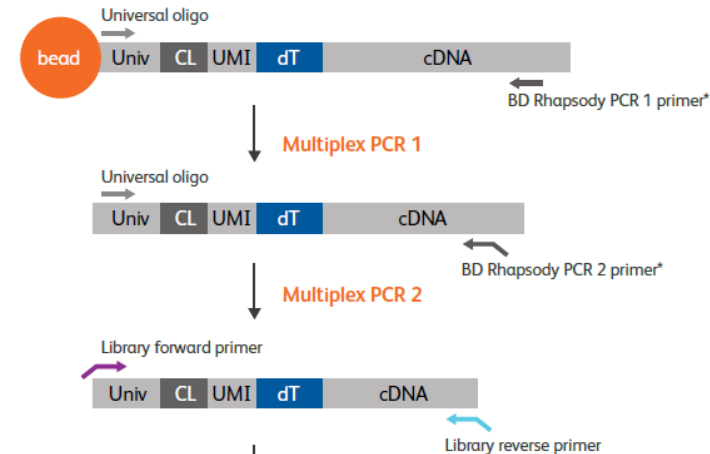


## Targeted RNAseq from single cells using BD Rhapsody system

- Analysis of low abundance transcripts
- Less sequencing required, cheaper
- Subsampling and archiving feasible



## Amplification of cDNA using targeted primer panels



- Panels – immune cell panels available, custom option in future.

## Whole Genome Sequencing from single cells

# Acknowledgements



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Illumina

# Thank you!



Contact us

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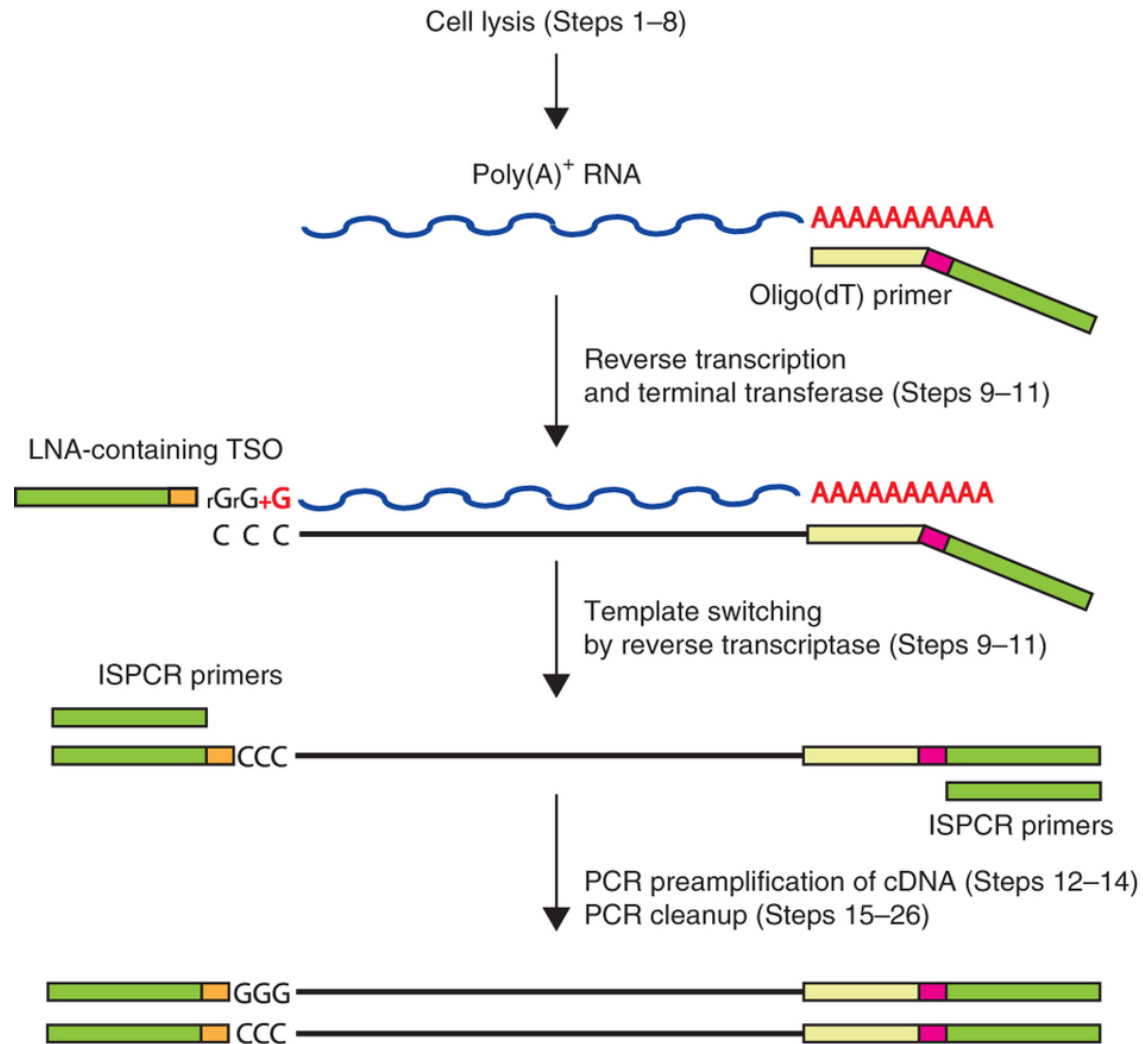
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Frederick National Laboratory for Cancer Research



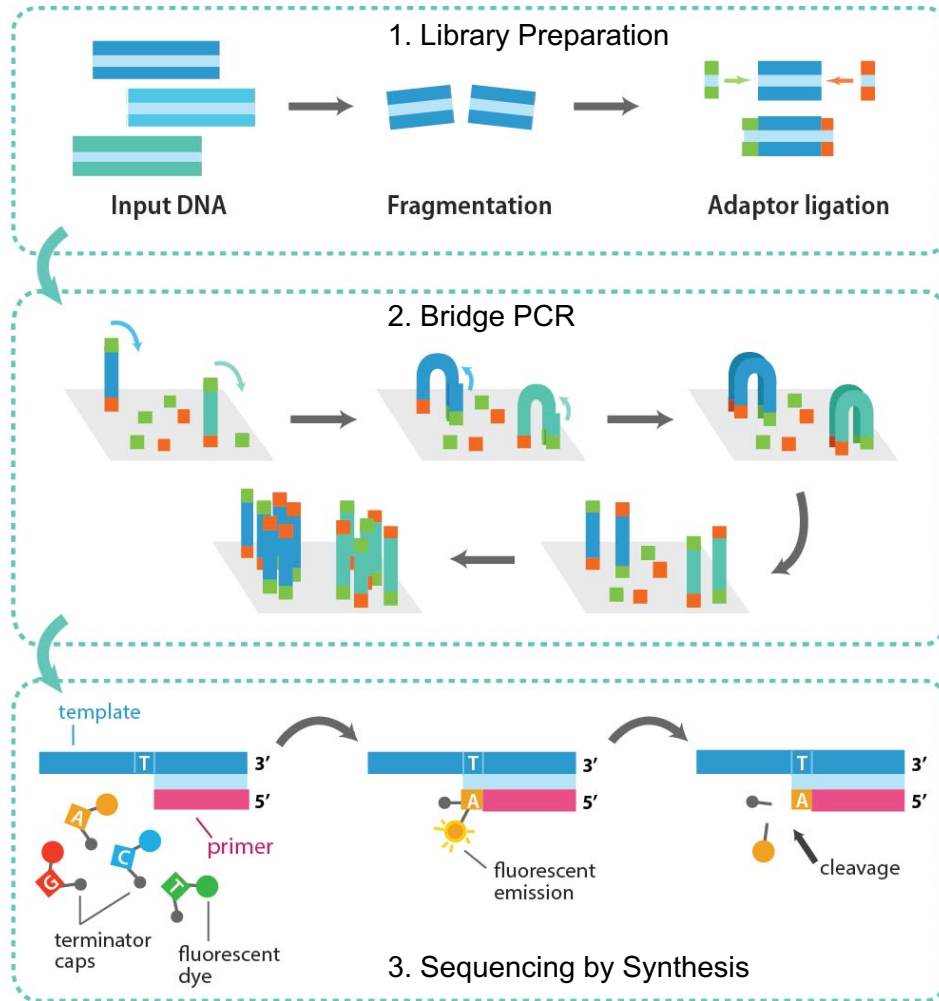


# Short-read Sequencing : Sequencing by Synthesis



## Illumina Sequencers (MiSeq, NextSeq, HiSeq)

- Max read length: 300bp
- Variety of sequencing applications: genome assembly, transcriptome analysis, SNP detection, DNA methylation analysis, metagenomic studies.
- High level multiplexing of samples possible.
- Big advantage: scale.
- Limitations of short read sequencing – genome assembly, isoform analysis, structural variation, haplotypes and SNP phasing.



# Library preparation for Short-read Sequencing

