# RNA sequencing: library preparation and experimental design

### RNA sequencing

- Introduction
- Preparation of RNA-seq libraries
- Design of RNA-seq experiments
- RNA-seq bias already identified

#### RNA sequencing

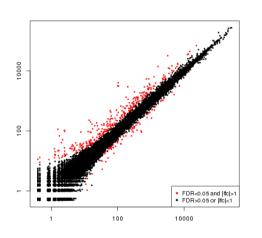
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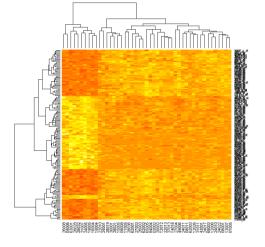
### Transcriptome analysis: key aims

#### Quantitative

Quantify the changes of expression level between different

conditions / time points





#### Qualitative

- Catalogue all different transcripts (mRNA, ncRNA)
- Determine the structure of these transcripts
  - TSS, 3' end, splicing patterns, post-transcriptional modifications

# Transcriptome analysis: different technologies

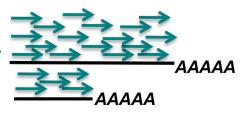
- Hybridization-based approach
  - Microarrays



- Drawbacks
  - Reliance upon existing knowledge on transcriptome
  - Poor quantification of lowly (background) and highly (saturation) expressed genes
  - Cross-hybridization

# Transcriptome analysis: different technologies

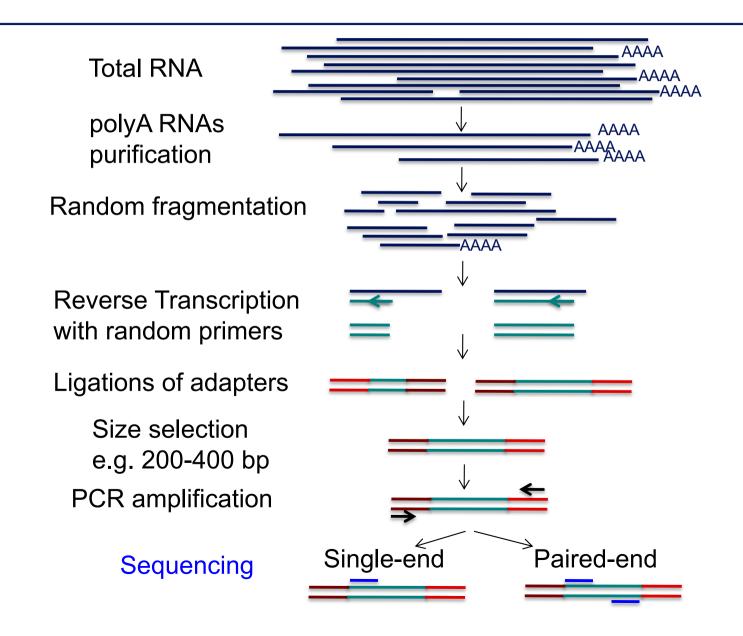
- Sequence-based approaches
  - EST (Expressed Sequence Tag)
    - Sequence of a cDNA fragment
    - Drawbacks
      - Sanger sequencing → low throughput
      - Generally not quantitative (normalized libraries)
  - SAGE (Serial Analysis of Gene Expression)
    - Sequence a tag : short fragment from a specific location of each transcript
    - Drawback : only a portion of the transcript is analysed (isoforms are generally indistinguishable from each other)
  - RNA-seq
    - Sequence cDNA fragments from the whole cDNA
    - Qualitative and quantitative information



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### RNA-seq library preparation

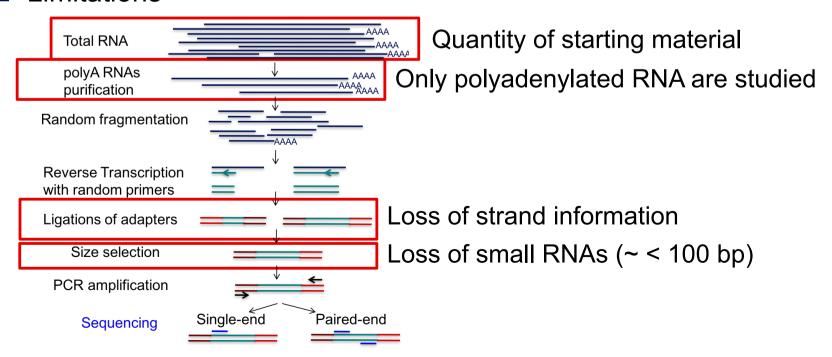


#### RNA-seq library preparation

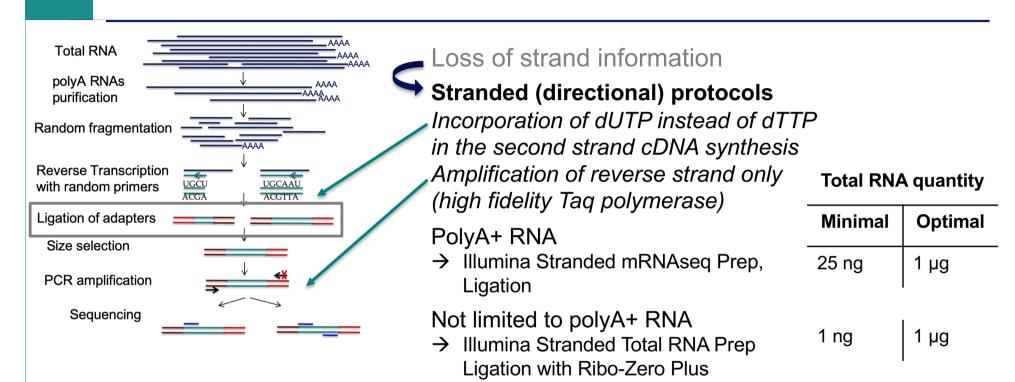
#### Advantages

- Highly reproducible
- High sensitivity
- Allows to study both coding and non-coding polyA+ RNAs expression
- Allows transcript discovery

#### Limitations



# RNA-seq library preparation : stranded protocols



Advantage: preserves the strand information

- → Allows to determine transcript orientation
- → Important for novel transcript discovery and annotation, especially for overlapping transcripts

# RNA-seq library preparation : stranded protocols

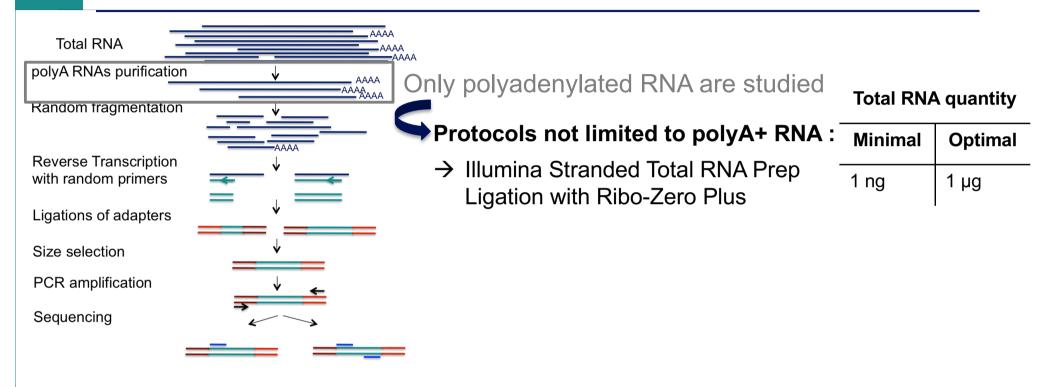
Good quality of strand-specificity

e.g. Results obtained on spike-in RNAs added in 4 libraries prepared with both standard and directional polyA+ RNA-seq protocols (GenomEast Platform)

Proportion of reads from each cDNA strand:

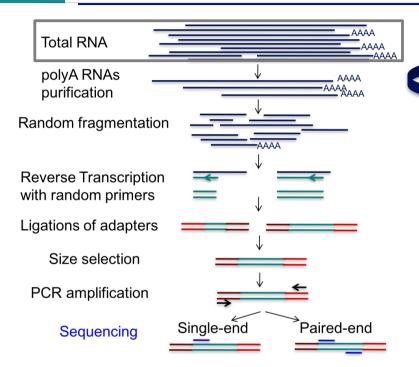


# RNA-seq library preparation: protocols not limited to polyA+ RNA



- Advantage
  - Allows to study non-polyadenylated transcripts
- Drawbacks
  - Efficiency of rRNA removal ≠ between samples
  - Higher number of RNA molecules sequenced compared to standard RNA-seq
    - → More reads needed to achieve the same coverage on polyadenylated RNAs

# RNA-seq library preparation : protocol with amplification



Quantity of starting material

#### **Protocol with amplification:**

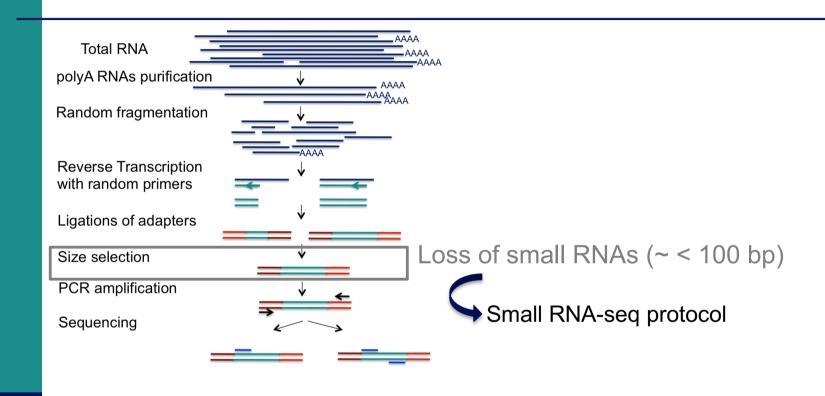
Clontech SMART-Seq v4 UltraLow input RNA cDNA preparation : cDNA synthesis + amplification Illumina Nextera XT DNA sample preparation : library preparation

**Total RNA quantity** 

Minimal	Optimal
150 pg (or 100 cells)	10 ng

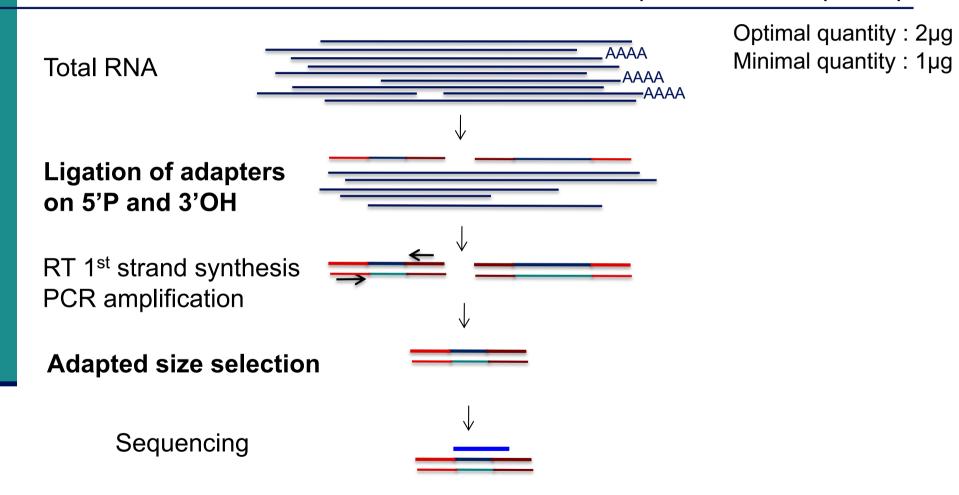
- Advantage
  - Low quantity of starting material
- Drawback
  - Bias due to the amplification

### small RNA-seq library preparation



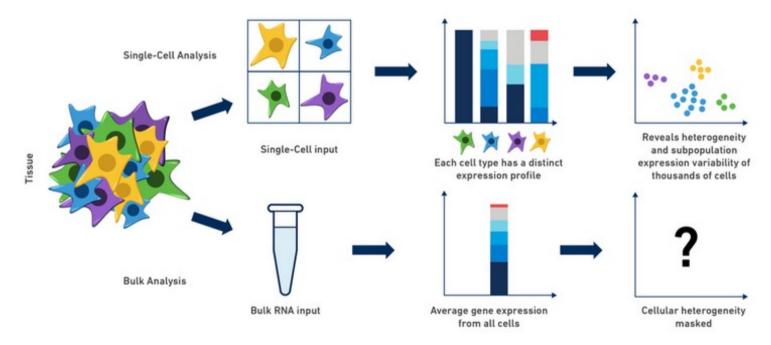
### small RNA-seq library preparation

Illumina Truseq smallRNA SamplePrep



### Single-cell RNA-seq

- Majority of RNA-seq experiments : study of a cell population
- Overlooks differences within a cell population that may be important for maintaining normal tissue function or facilitating disease progression
- Single-cell RNA-seq provides the expression profiles of individual cells
  - Allows to characterize the subpopulation structure
  - Allows to study cell heterogeneity



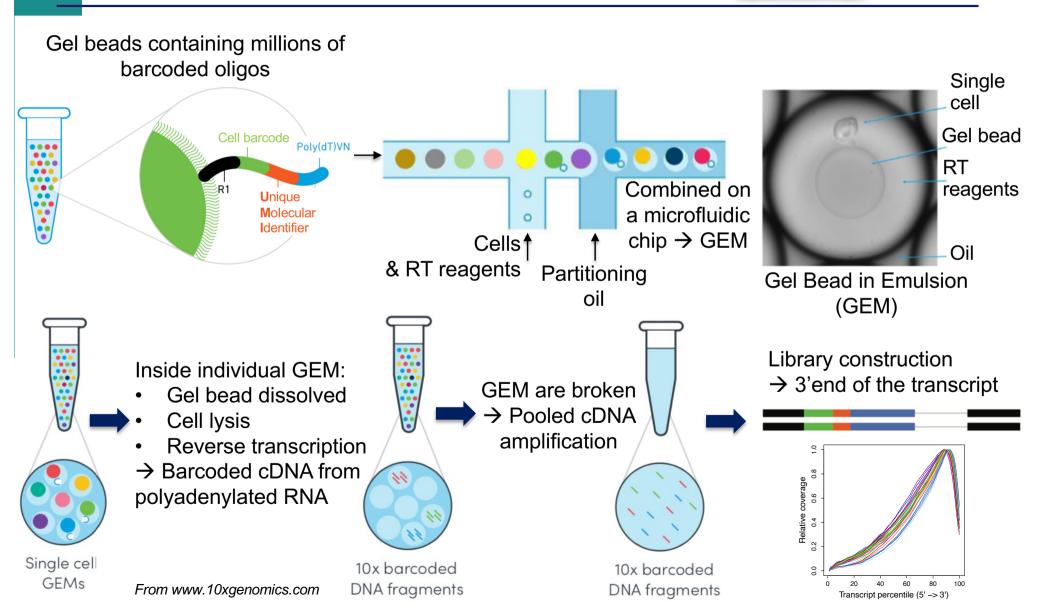
#### Single-cell RNA-seq

- Different protocols for single-cell RNA-seq
  - 3' counting or full-length
  - Unique Molecular Identifiers (UMI)
    - Random sequences used to tag each molecule prior to library amplification
    - 2 reads from the same cell align to the same location + have the same UMI
       → highly likely PCR duplicates
- Limits
  - Technical noise due to amplification and dropout

# 10X Genomics single-cell RNA-seq

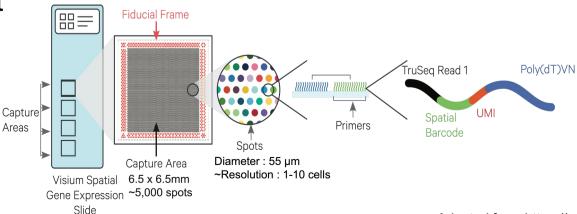


Chromium iX

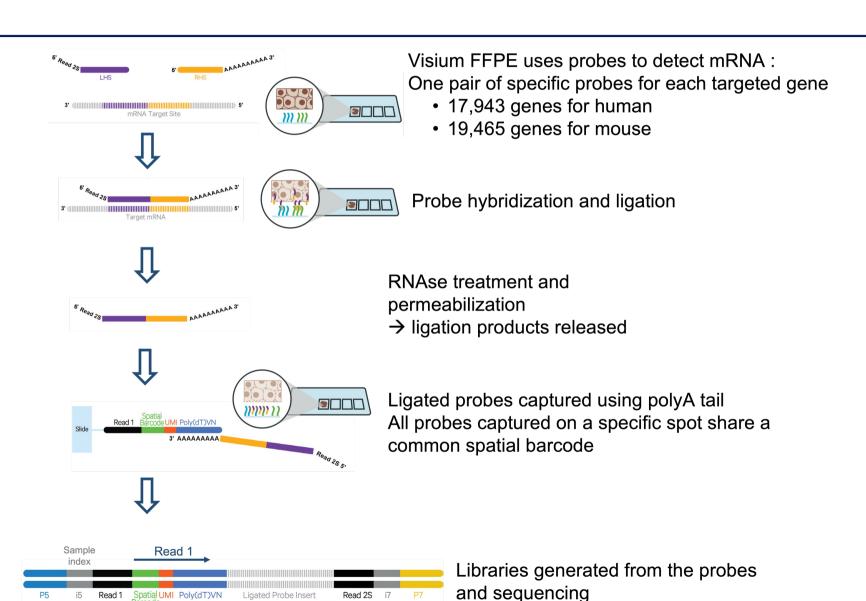


### Spatial transcriptomics

- Organization of cells within a tissue
  - Crucially tied to their biological activity
  - Interesting to analyse transcriptomic profiles while retaining spatial context
- 10X Genomics Visium
  - → To measure RNA levels in FFPE or fresh frozen tissue sections
  - Allows to map the transcriptome within the morphological context



#### 10X Genomics Visium FFPE



Read 2

#### RNA sequencing

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#### Experimental design

- 1. Define your biological questions of interest
- 2. Define the best appropriate experimental design to answer these questions:
  - Library preparation protocol
  - Sequencing strategy
  - Number of reads
  - Number of replicates
- Define a detailed experimental plan in advance of doing the experiment
- Try to reduce batch effects
- ENCODE guidelines (mammalian tissues)

https://www.encodeproject.org/about/experiment-guidelines/

### Which protocol for which application?

- Choice depend on
  - Quantity of starting material
  - Type of RNA studied (small/long, polyA+/-)
  - Biological questions of interest
    - e.g. new transcript identification → directional protocol
- Keep the same protocol for all samples within a project

#### Which protocol for which application?

Kit used by the	Total RNA	A quantity	Type of studied	Cturous de d
platform	Minimal	Optimal	RNA	Stranded
Truseg Stranded mRNA Prep Illumina	200 ng	1 μg	Only polyA+ RNA of size > 100 b	Yes
Illumina Stranded mRNA Prep, Ligation	25 ng	1 μg	Only polyA+ RNA of size > 100 b	Yes
SMART-Seq v4 <u>UltraLow</u> Input RNA kit ( <u>Clontech</u> ) + <u>Nextera</u> XT DNA sample preparation Kit (Illumina)	100 cells	10 ng	Only polyA+ RNA of size > 100 b	No
QuantSeq 3' mRNA-Seq Li- brary Prep Kit for Illumina (FWD) (Lexogen)	1 ng	500 ng	Only 3' end of polyA+ RNA	Yes
Chromium Next GEM Sin- gle Cell 3' Reagent Kits	1 cell	1 cell	Only 3' end of polyA+ RNA	No
Truseq Stranded Total RNA Sample Prep Illumina	100 ng	1 μg	All RNA of size > 100 b	Yes
Illumina Stranded Total RNA Prep Ligation with Ribo-Zero Plus	1 ng	1 μg	All RNA of size > 100 b	Yes
Truseg SmallRNA Sample Prep (Illumina)	1 µg	2 µg	All small RNAs with 5'P and 3'OH (desired size can be chosen by the project manager)	Yes

### Which sequencing strategy?

- Expression quantification on annotated transcripts
  - Single-end sequencing provides good results
- Alternative splicing analysis, fusion transcript detection, mapping over repetitive regions, de novo transcriptome assembly

#### How many reads are needed?

- Transcriptome coverage as a function of sequencing depth: highly dependant on transcriptome complexity
- Sequencing depth should be determined by the goals of the experiment
- General recommendations for typical mammalian tissues
  - > 30 million reads with polyA+ protocols
  - > 50 million reads with total protocols
  - ... if the goal is to quantify expression of annotated genes
- Higher sequencing depth needed if
  - the sensitivity of detection is important
  - the purpose is to discover novel transcripts
  - the purpose is to precisely quantify transcript isoforms

#### How many replicates are needed?

- Low technical variability and technical variability << biological variability (Marioni et al. Genome Research 2008. Bullard et al. BMC Bioinformatics 2010)
  - → Technical replicates not required
- But "sequencing technology does not eliminate biological variability" (Hansen et al. Nat Biotechnol. 2011)
  - Biological replicates are fundamental!
  - How many ?
    - Highly dependant on the correlation between replicates and on the difference between the compared conditions
    - If possible, prepare more samples for low-input RNA-seq

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#### RNA-seq bias / sources of variability

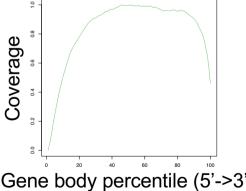
- As all techniques, RNA-seq present bias affecting expression estimates and subsequent statistical analysis
- Identification of bias in RNA-seq protocol
  - Use of synthetic spike-in standards
     (Jiang et al. Genome Research 2011;21(9):1543-51)
  - Provided by ERCC (External RNA Control Consortium)
  - 92 sequences
  - Minimal sequence homology with endogenous transcripts from sequenced eukaryotes
  - Various lengths and GC content, large range of concentrations

### RNA-seq bias / sources of variability

Composition bias of the first 13 nucleotides due to a non-random hexamer priming

(Hansen et al. 2010;38(12):e131. Li et al. Genome Biology 2010;11(5):R50)

- Bias during library amplification (Kozarewa et al. 2009;6(4):291-5)
  - Over-amplification of GC-rich regions
  - Generation of duplicate sequences
- Read coverage bias (Jiang et al. Genome Research 2011;21(9):1543-51)
  - Unevenness in read coverage along transcripts



Gene body percentile (5'->3')

- Variability in RNA-seq data (Marioni et al. Genome Research 2008;18(9):1509-17. Bullard et al. BMC Bioinformatics 2010;11:94)
  - Biological condition >> library preparation > run > lane

#### RNA-seq bias / sources of variability

#### Transcript abundance

- Low abundance transcripts more affected by sampling error : more bias in the estimation of their expression level
- Highly dependant on the sequencing depth :
  - A question of cost, not due to the technique
- Transcript length (Oshlack et al. Biology Direct 2009;4:14)
  - The ability to call differentially expressed genes between samples is associated with the length of the transcript :
    - more statistical power to detect differential expression for long transcripts compared to short ones

#### Mappability bias

- Uniquely mapping reads are typically summarized over genomic regions → regions with lower sequence complexity will tend to end up with lower sequence coverage
- Reads corresponding to longer transcripts have a higher mappability