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

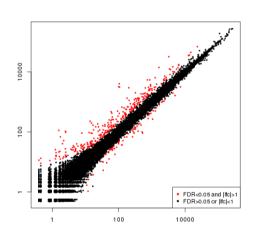
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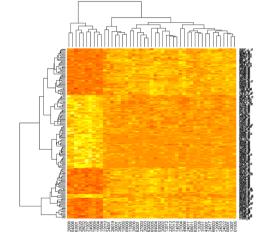
Transcriptome analysis: key aims

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

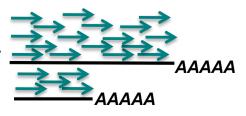
- Hybridization-based approach
 - Microarrays



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

Transcriptome analysis: different technologies

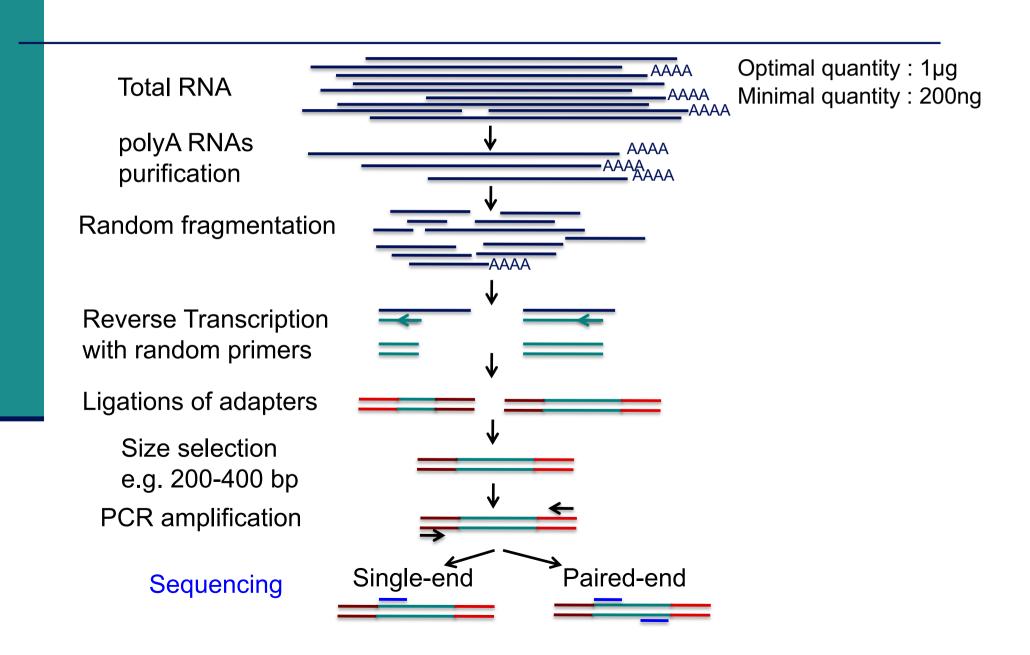
- Sequence-based approaches
 - EST (Expressed Sequence Tag)
 - Sequence of a cDNA fragment
 - Drawbacks
 - Sanger sequencing → low throughput
 - Generally not quantitative (normalized libraries)
 - SAGE (Serial Analysis of Gene Expression)
 - Sequence a tag : short fragment from a specific location of 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



RNA sequencing

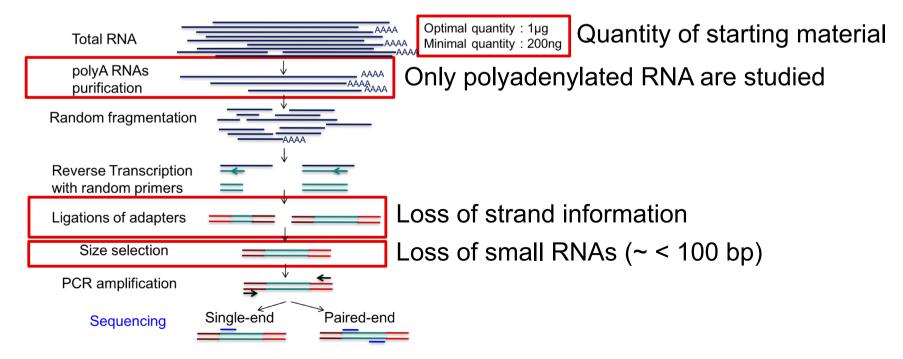
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RNA-seq library preparation

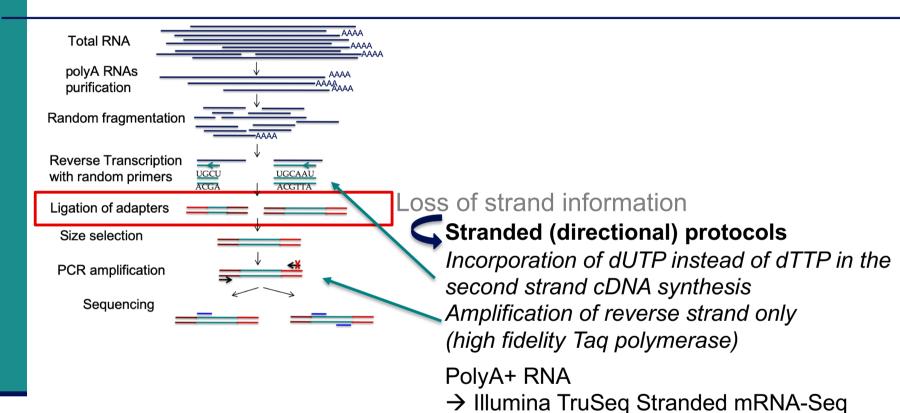


RNA-seq library preparation

- 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



Advantages

→ Illumina TruSeq Stranded Total RNA-Seq

Optimal quantity: 1 µq

Minimal quantity: 200 ng

Not limited to polyA+ RNA

- 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

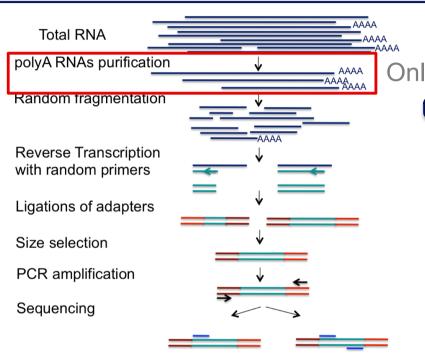
Good quality of strand-specificity

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



Only polyadenylated RNA are studied

Protocols not limited to polyA+ RNA:

With amplification

→ NuGEN Ovation RNA-seq

(RT primers specific to non rRNA sequences)

Optimal quantity: 10 ng Minimal quantity: 500 pg

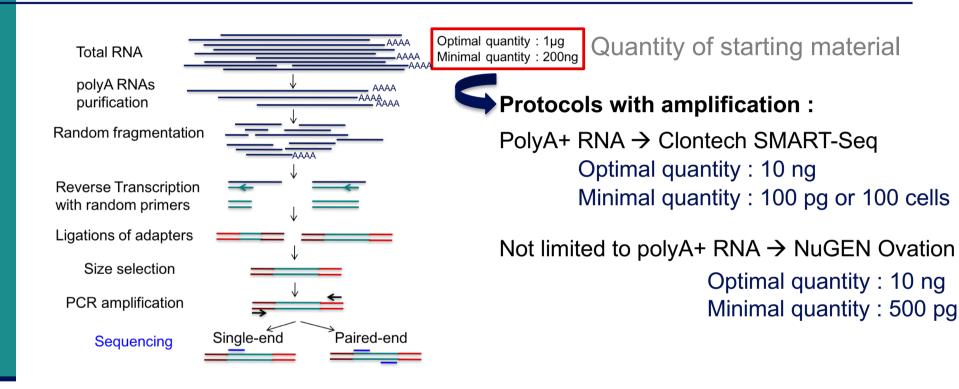
Without amplification

→ Illumina TruSeq Stranded Total RNA-seq (RiboZero probes used for rRNA depletion)

> Optimal quantity: 1 µg Minimal quantity: 100 ng

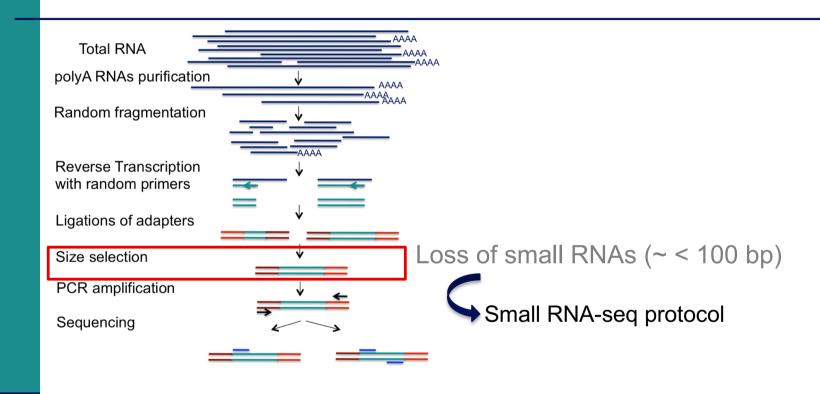
- Advantage
 - Allows to study non-polyadenylated transcripts
- Drawbacks
 - Efficiency of rRNA removal ≠ between samples
 - Higher number of RNA molecules sequenced compared to standard RNA-seq
 - → More reads needed to achieve the same coverage on polyadenylated RNAs

RNA-seq library preparation : protocols with amplification



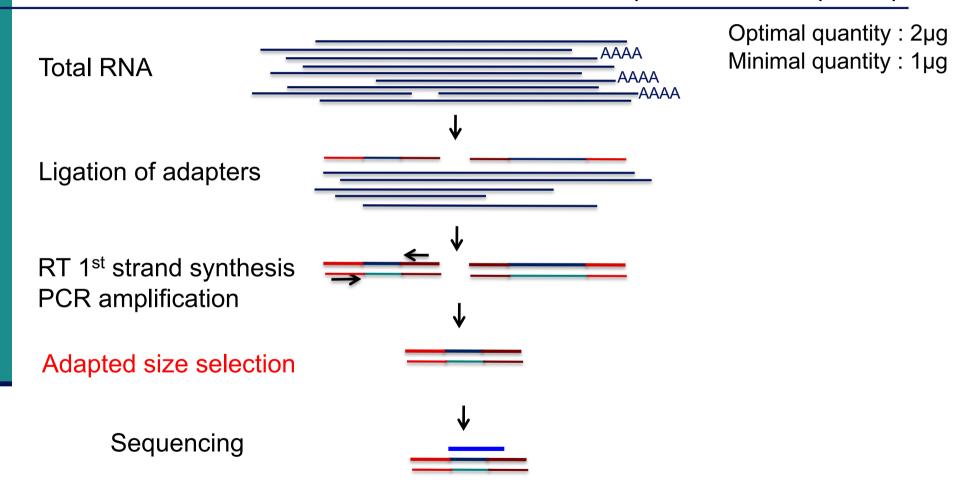
- Advantage
 - Low quantity of starting material
- Drawback
 - Bias due to the amplification

small RNA-seq library preparation

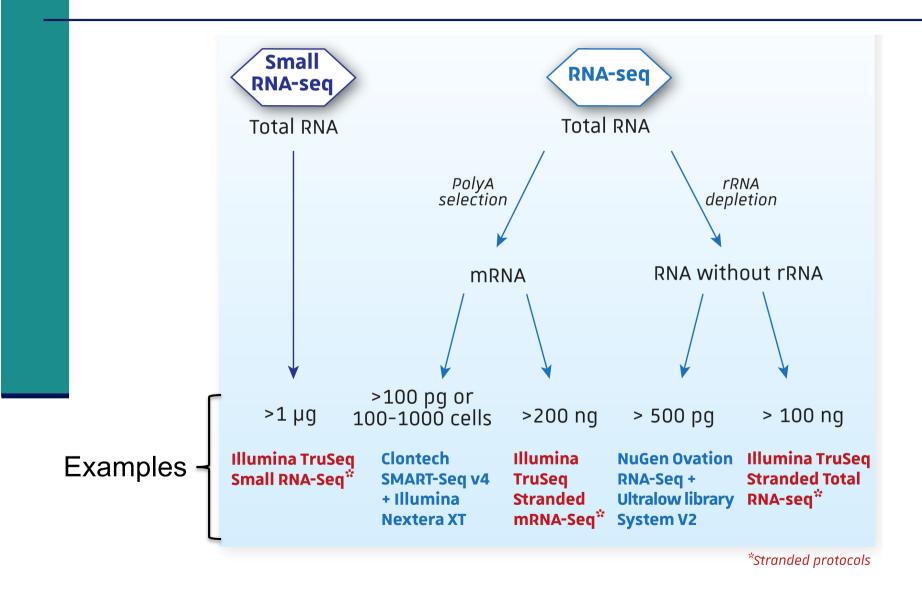


small RNA-seq library preparation

Illumina Truseq smallRNA SamplePrep

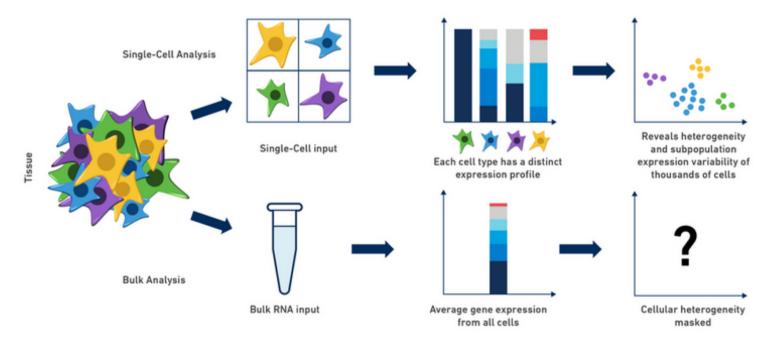


RNA-seq library preparation protocols (bulk)



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 technologies for single cell collection
 - Droplets (e.g. Chromium, 10X Genomics)
 - Microfluidics (e.g. C1 single-cell auto-prep system, Fluidigm)
 - Microwells (e.g. Rhapsody single-cell analysis System, BD)
- Different protocols for RNA-seq
 - 3' counting or full-length
 - With or without Unique Molecular Identifiers (UMI)
 - Random sequences used to tag each molecule prior to library amplification
 - 2 reads align to the same location and have the same UMI
 - → highly likely PCR duplicates
- Limits
 - Technical noise due to amplification and dropout

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Experimental design

- 1. Define your biological questions of interest
- 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?

Library preparation	Kit used by the platform	Total RNA quantity		Type of studied RNA	Stranded
preparation		Minimal	Optimal	MVA	
mRNAseq/ standard quantity	TruSeq RNA Sample Prep	200 ng	1μg	Only polyA+ RNA of size > 100 b	No
Stranded mRNAseq/ standard quantity	Directional mRNA-Seq SamplePrep	200 ng	1μg	Only polyA+ RNA of size > 100 b	Yes
mRNAseq/low input (Smarter)	SMART-Seq v4 UltraLow Input RNA kit + Nextera XT DNA sample preparation Kit	100 cells	10 ng	Only polyA+ RNA of size > 100 b	No
mRNA-seq/ single cell	SMARTer Ultra Low RNA Kit for the Fluidigm C1 System + Nextera XT DNA sample preparation Kit	1 cell	1 cell	Only polyA+ RNA of size > 100 b	No
Total RNAseq Ribozero/standard quantity	Truseq Stranded Total RNA SamplePrep	100 ng	1 μg	All RNA of size > 100 b	Yes
Total RNAseq/ low input (Ovation)	Ovation RNA-Seq System V2 + Ovation SP Ultralow Library systems	500 pg	10 ng	All RNA of size > 100 b	No
Small RNA-seq	Truseq SmallRNA SamplePrep	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

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on	our	Hiseq	4000

Application	Suggested multiplexing for standard experiments on mammalian genomes
small RNA-seq	20 samples / lane
mRNA-seq with polyA selection → for gene expression quantification → for alternative splicing analysis	8 samples / lane 3 samples / lane
RNA-seq with ribodepletion → for gene expression quantification	5 samples / lane

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

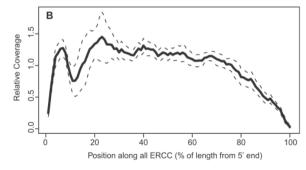
- As all techniques, RNA-seq present bias affecting expression estimates and subsequent statistical analysis
- Identification of bias in RNA-seq protocol
 - Use of synthetic spike-in standards
 (Jiang et al. Genome Research 2011;21(9):1543-51)
 - Provided by ERCC (External RNA Control Consortium)
 - 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



- 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