

Module 2 : RNAseq bioinformatic
from sequence to expression level
smallRNA & isoform (eukaryote)

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RNAseq analyses

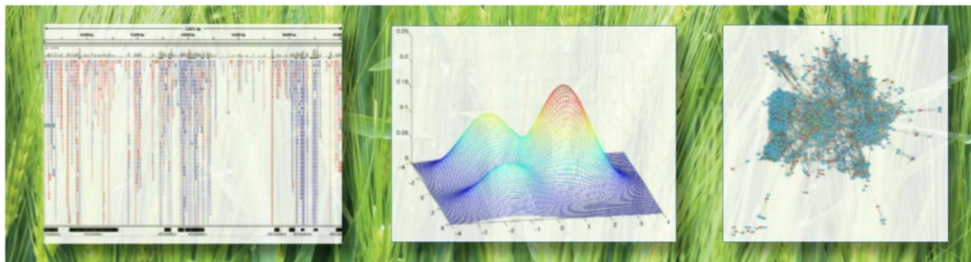
RNAseq: From sequence data (reads) to expression level (count)

Classical analyses of RNA-Seq:

- Check quality, Trimming
- Mapping / counts
- Assembly

Others usages of RNA-Seq:

- smallRNA study
- expression at isoform level



RNAseq analyses

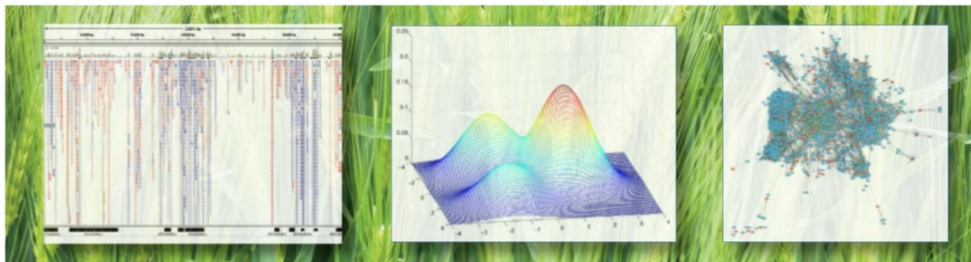
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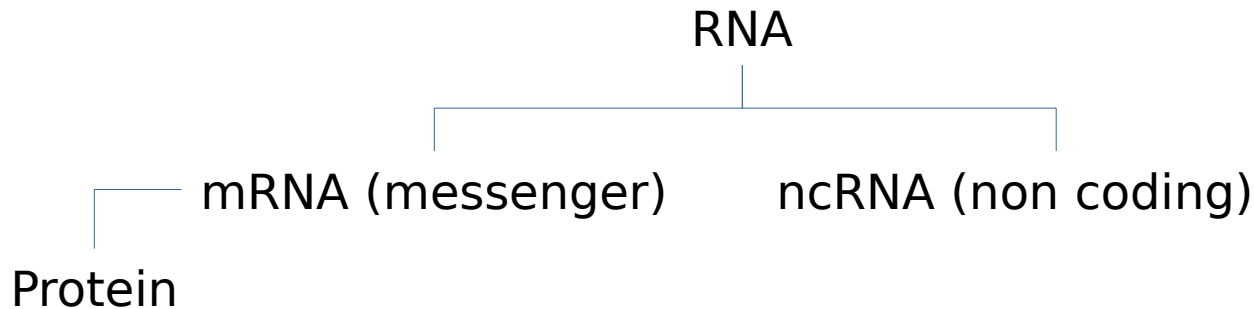
- Check quality, Trimming
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Others usages of RNA-Seq:

- **smallRNA study**
- expression at isoform level



RNA world



Not predicted by gene prediction tools

- No specific signal (start, stop, splicing sites...)
- Multiple location (intergenic, intronic, coding, antisense)
- Variable size
- No strong sequence conservation in general

Function related to structure : ncRNA of the same family have a conserved structure

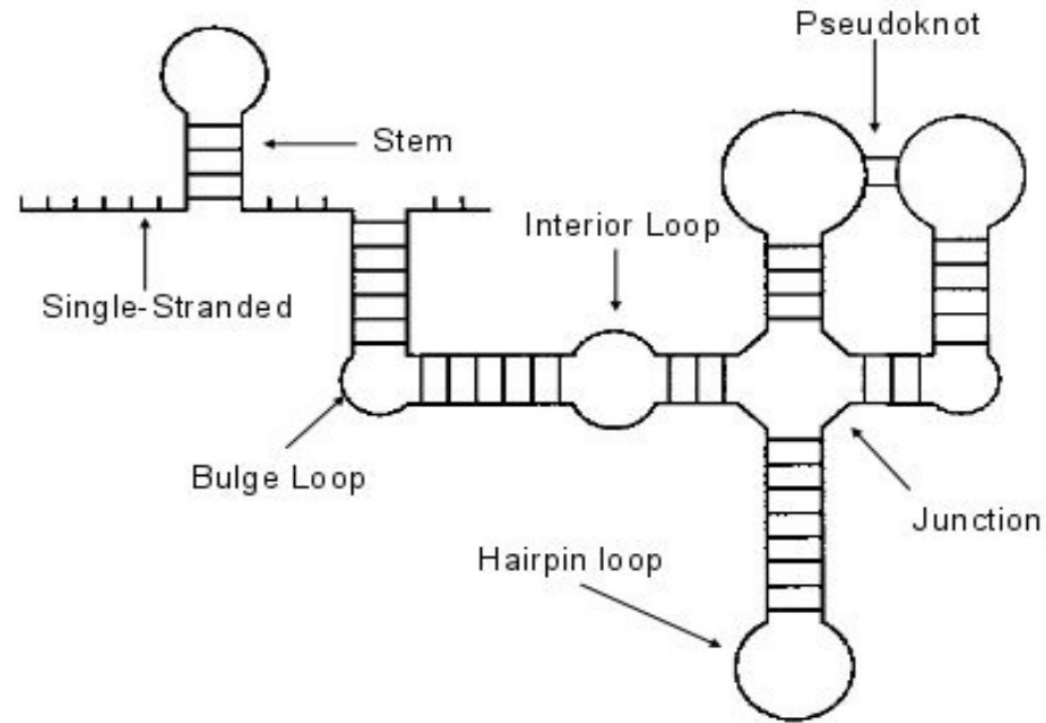
RNA folding

Folding = Secondary structure

RNA folds on itself by base pairing :

- A with U
- C with G
- sometimes G with U

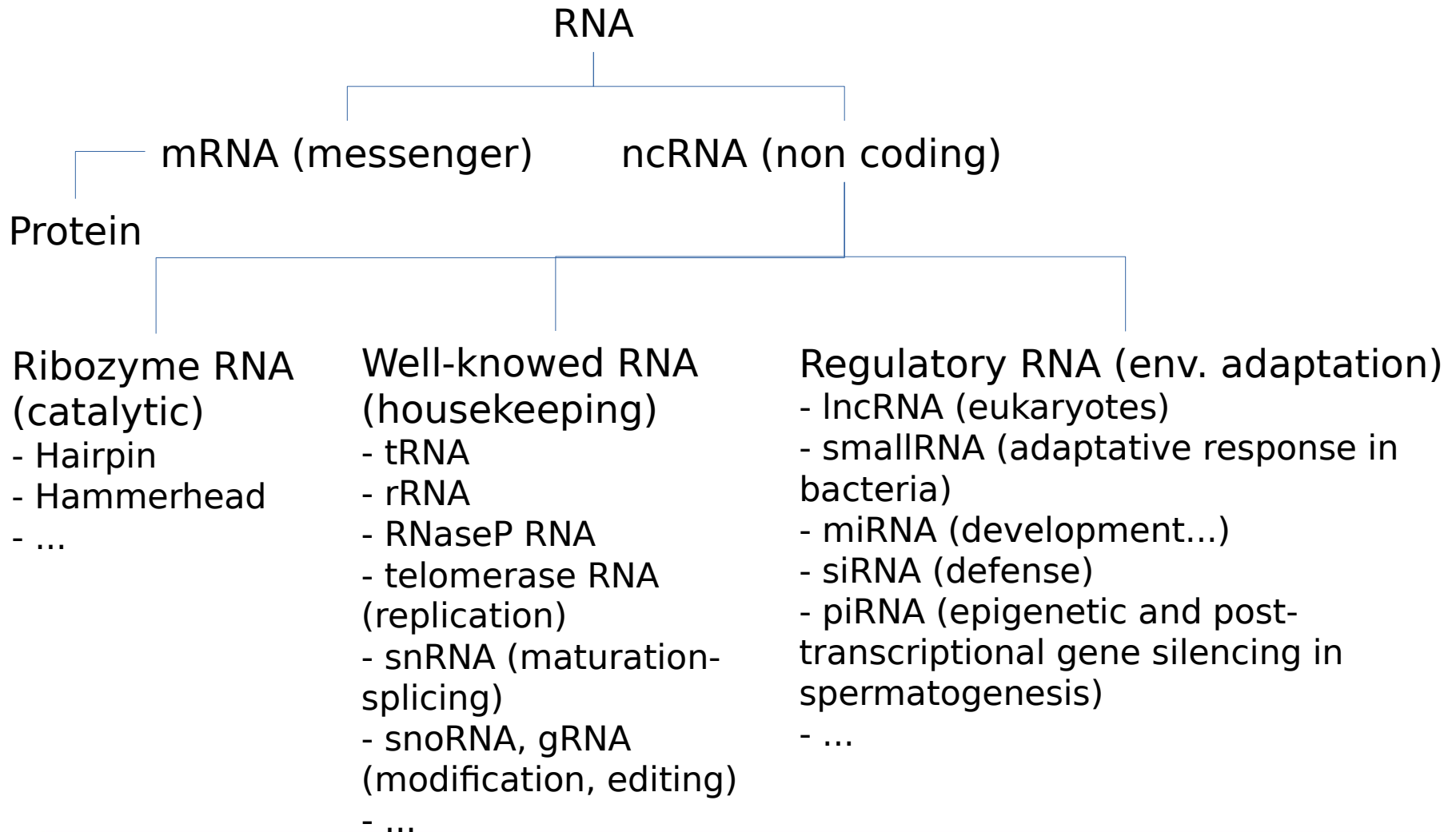
=> more combinatorial possibilities of folding than DNA



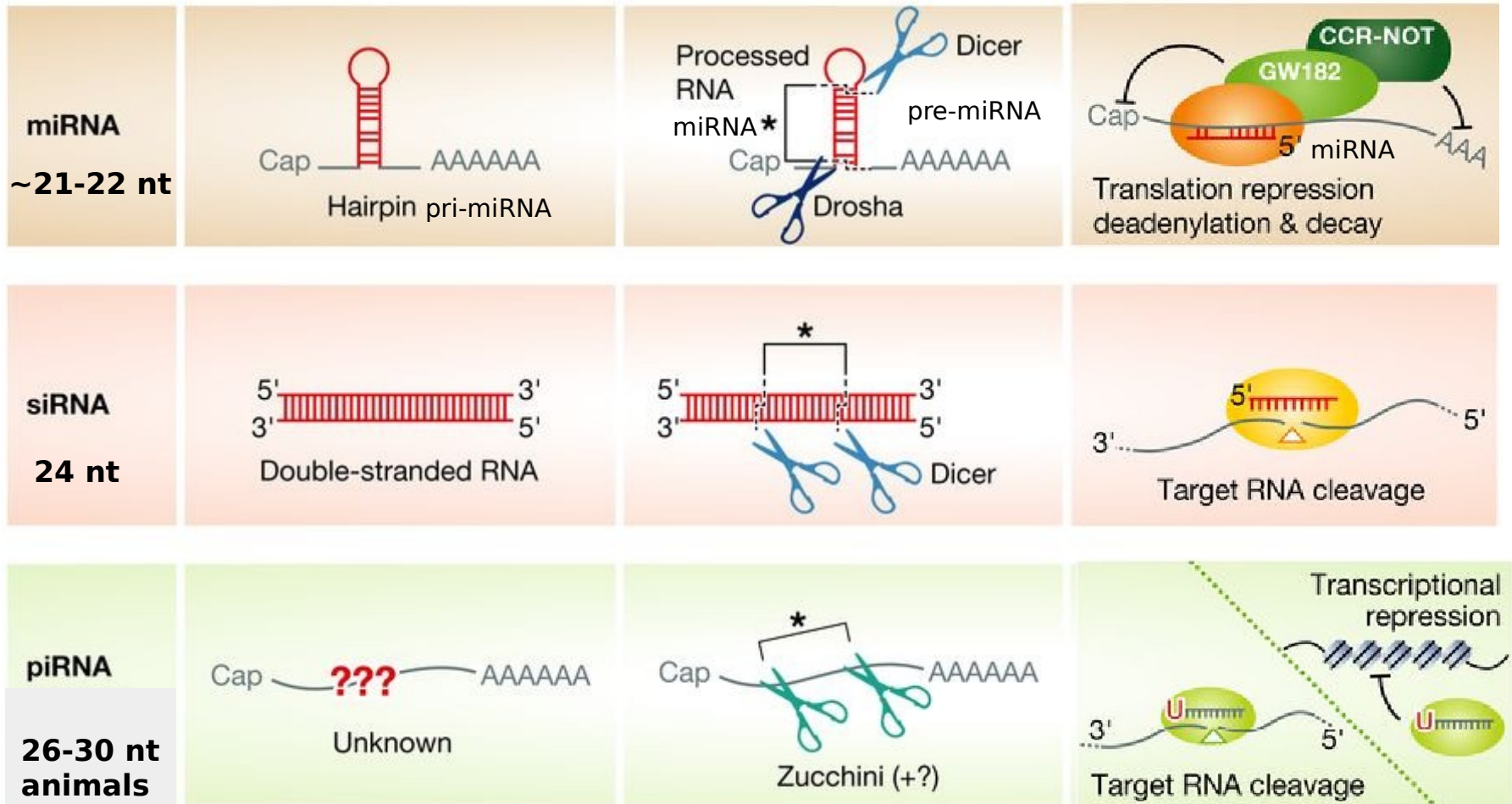
RNA folds:

- on itself
- with another: RNA duplex

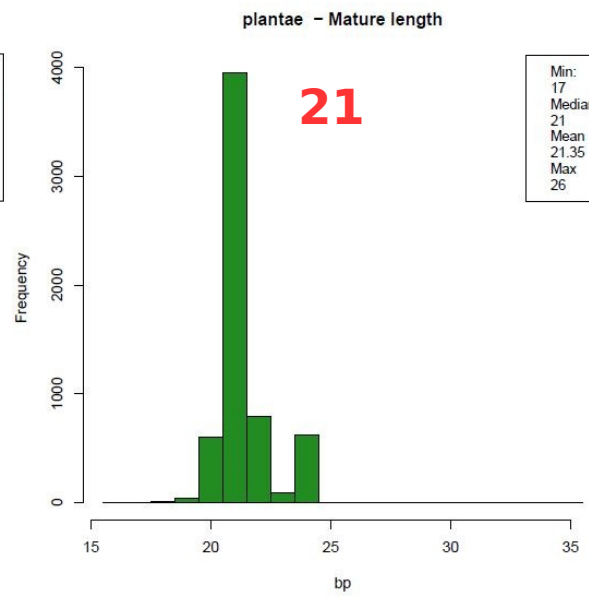
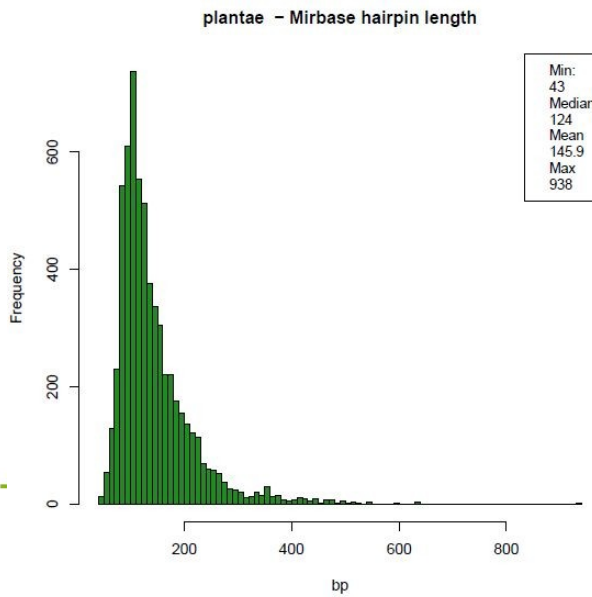
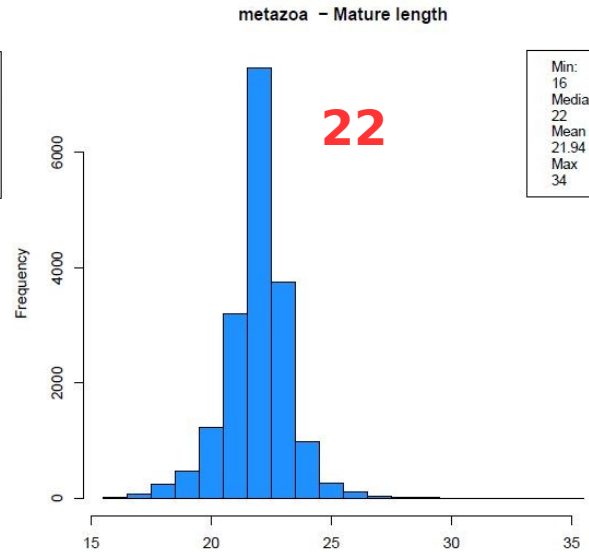
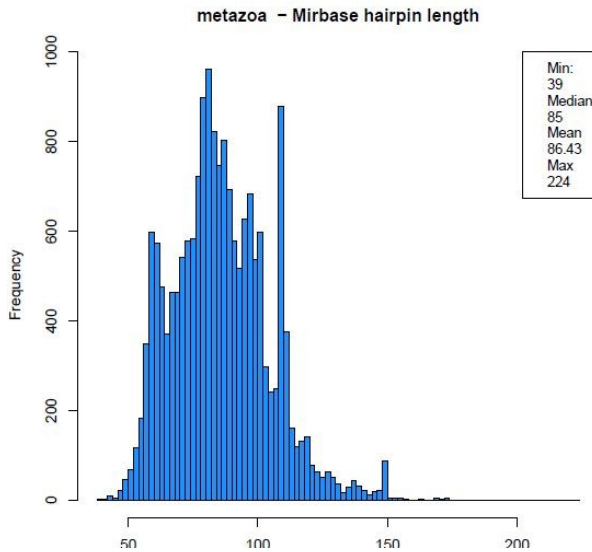
RNA world



Eukaryotic regRNA



Size matters



Griffiths-Jones S, Saini HK, van Dongen S, Enright AJ. miRBase: tools for microRNA genomics. *Nucleic Acids Res.* 2008 Jan;36(Database issue):D154-8

smallRNA & NGS

Classical RNAseq not suited for smallRNA (protocol and size)
smallRNAs lack a common sequence (e.g. a poly(A) tail) that can be used for selective enrichment or as a universal primer-binding site for reverse transcription

Strategies to enrich RNA sample in smallRNA (*):

- smallRNA cloning, deep sequencing : significant biases are introduced during small RNA cDNA library preparation (often more than 3 orders of magnitude for individual miRNAs ; one major source: RNA ligation)
- RNA Immuno-Precipitation (RIP-seq)
- Total RNA extraction + size selection

smallRNAseq pipeline

Experimental design

Module 1 : High Troughput Sequencing Technologies, available and futures...

Sequencing

Quality check

Cleaning

size selection

Module 2 : Bioinformatic of RNAseq

with reference

Mapping

w.o. reference

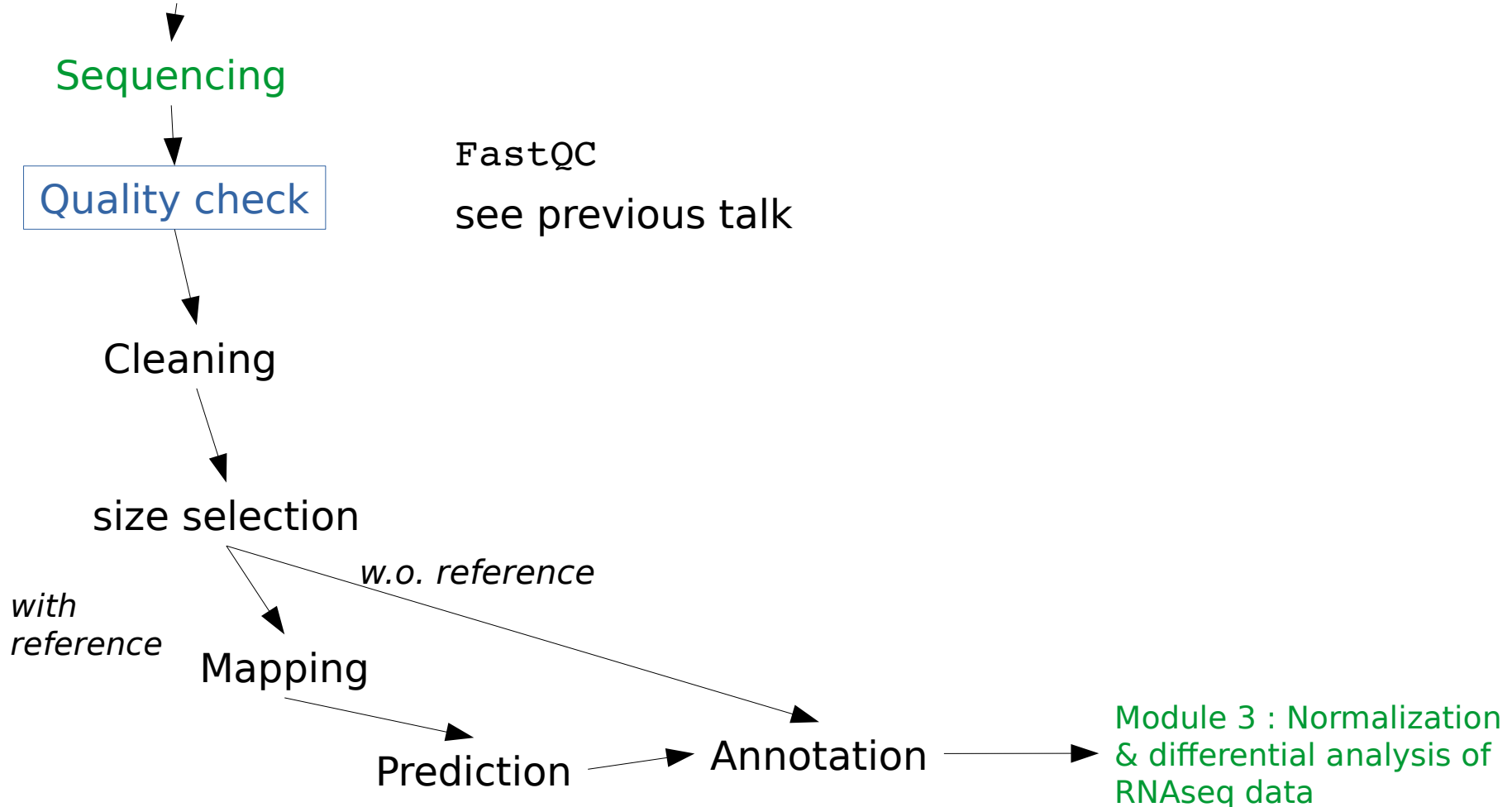
Prediction

Annotation

Module 3 : Normalization & differential analysis of RNAseq data

smallRNAseq pipeline

Experimental design



smallRNAseq pipeline

Experimental design

Read size > regRNA size

PCRprimer sequences & adapter are included in read

Sequencing

Quality check

Cleaning

size selection

with
reference

Mapping

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Module 3 : Normalization
& differential analysis of
RNAseq data

Overrepresented sequences

Sequence	Count	Percentage	Possible Source
GGGGATGTAGCTCAGAGATCGGAAGAGCACACGTCTGAACTCCAGTCAC	3865	1.546	Illumina Multiplexing PCR Primer 2.01 (100% over 34bp)
GCGTCTGTAGTCCAACCGTTAGGATAATTGAGATCGGAAGAGCACACGT	3021	1.2084	No Hit
GGGGATGTAGCTCAGATCGGAAGAGCACACGTCTGAACTCCAGTCACCA	2205	0.882	TruSeq Adapter, Index 7 (100% over 35bp)
AGATCGGAAGAGCACACGTCTGAACTCCAGTCACCAGATCATCTCGTATG	2047	0.8188000000000001	TruSeq Adapter, Index 7 (100% over 49bp)
AGGGCTATAGCTAGATCGGAAGAGCACACGTCTGAACTCCAGTCACCAGA	1478	0.5912	TruSeq Adapter, Index 7 (100% over 37bp)
CTAACAGACCGGTAGACTTGAAAGATCGGAAGAGCACACGTCTGAACTC	1222	0.4888	Illumina Multiplexing PCR Primer 2.01 (100% over 27bp)
CTTGAAAGATCGGAAGAGCACACGTCTGAACTCCAGTCACCAGATCATC	1155	0.462	TruSeq Adapter, Index 7 (100% over 42bp)

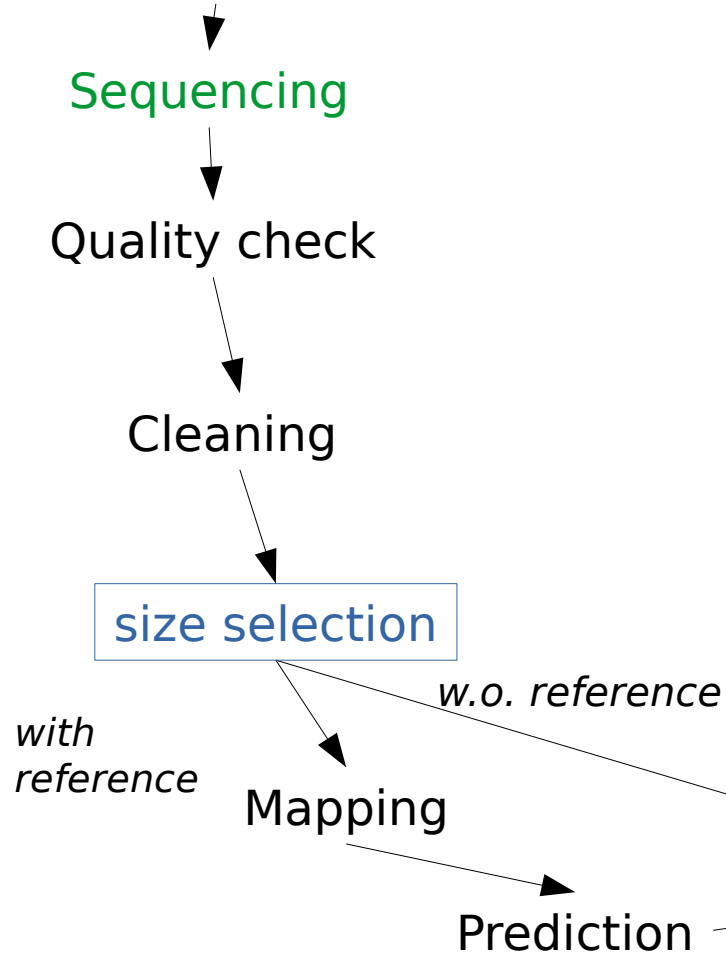
=> trimming step : remove adapter in 3', no quality trimming, remove rRNA, tRNA

trimgalore, cutadapt, trimmomatic, ...

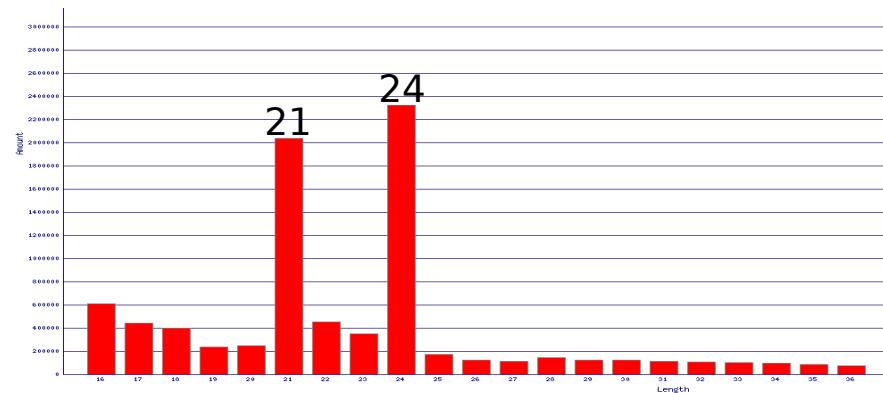
w.o. reference

smallRNAseq pipeline

Experimental design



Check the distribution of length reads
Plant: miRNA=21nt, siRNA=24nt



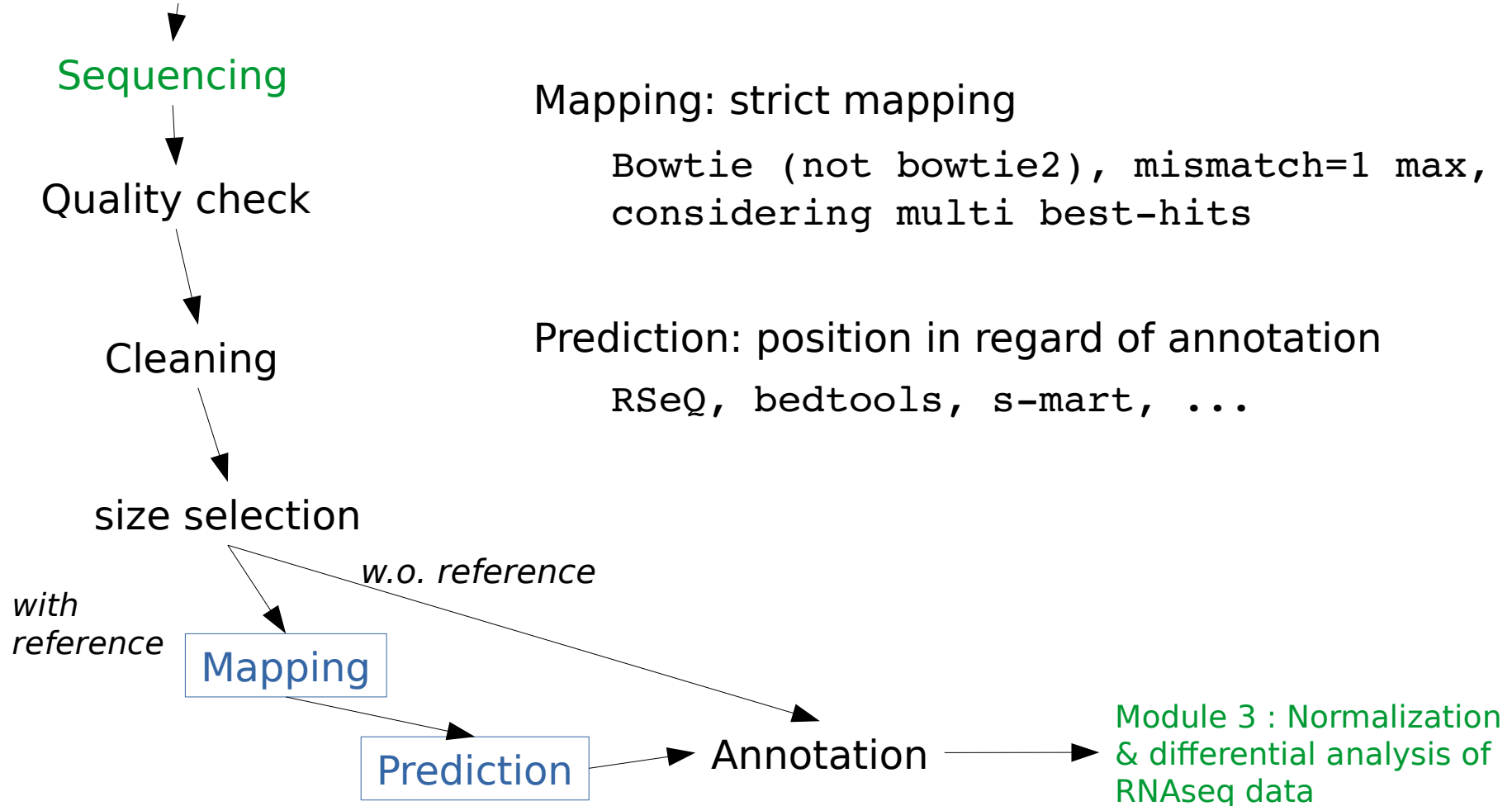
detect redundancy

`fasta_clipping_histogram,`
`duplicate fastx_collapser`
from `fastx-toolkit`

Module 3 : Normalization
& differential analysis of
RNAseq data

smallRNAseq pipeline

Experimental design



smallRNAseq pipeline

Experimental design

Sequencing

Quality check

Cleaning

size selection

with reference

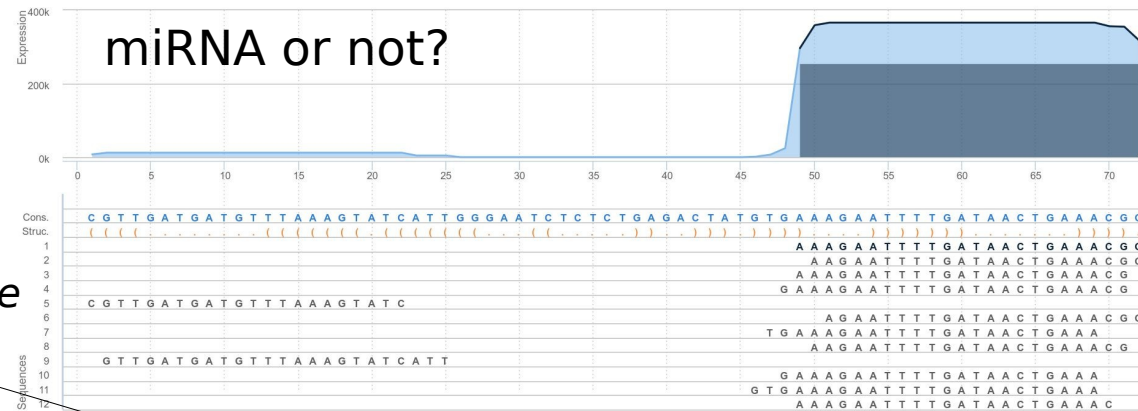
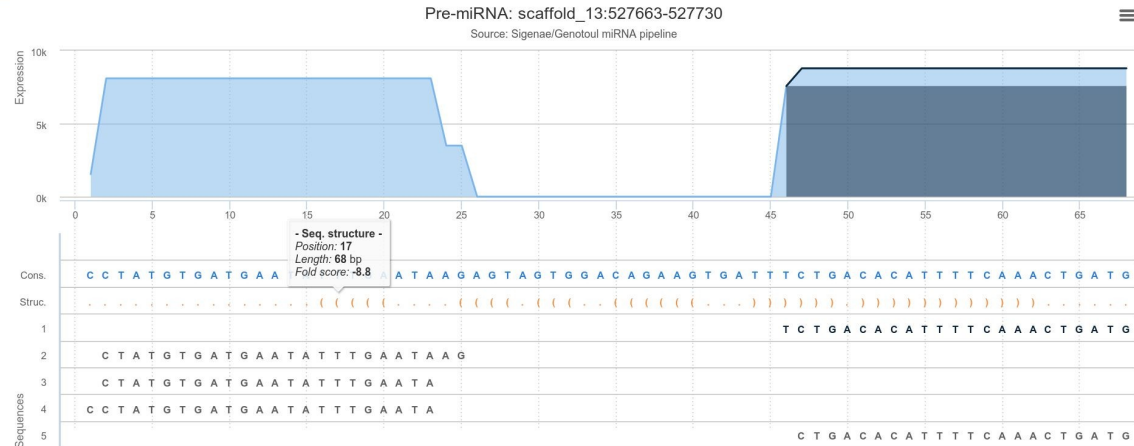
Mapping

Prediction

w.o. reference

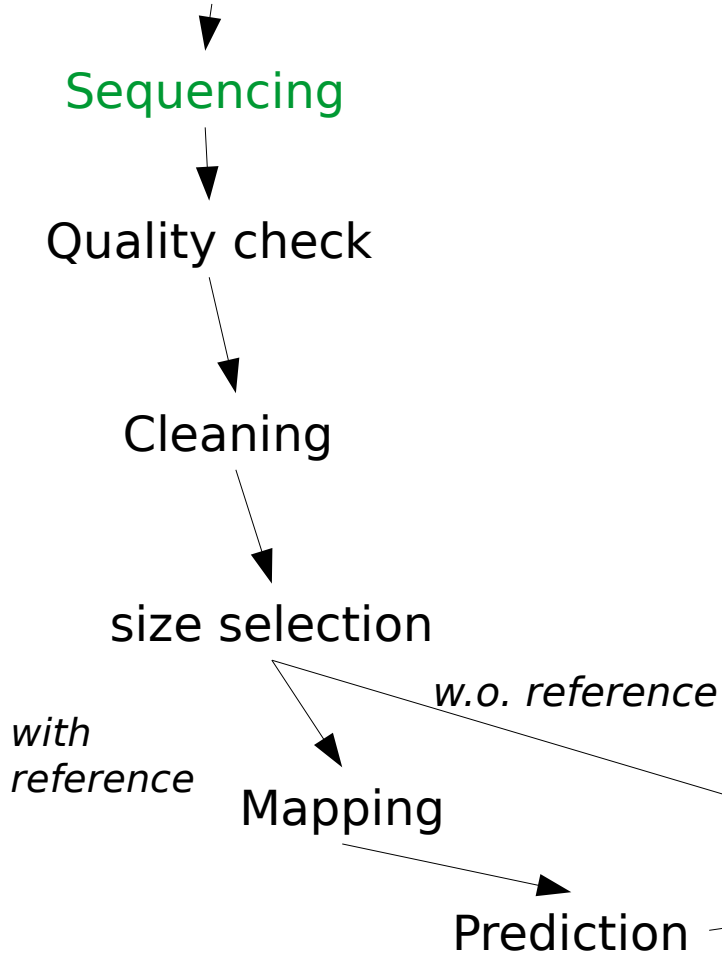
Annotation

Module 3 : Normalization & differential analysis of RNAseq data



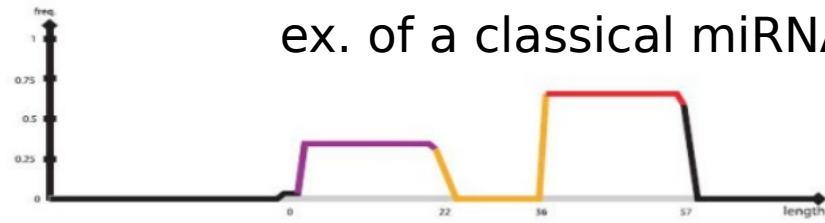
smallRNAseq pipeline

Experimental design



```

    Provisional ID : gi_89059864_ref_NT_011669.16_HsX_11826_19172
    Score total : 15.2
    Score for star read(s) : 3.9
    Score for read counts : 8.8
    Score for mfe : 1.6
    Score for randfold : 1.6
    Score for cons. seed : -0.6
    Total read count : 29
    Mature read count : 19
    Loop read count : 0
    Star read count : 10
  
```



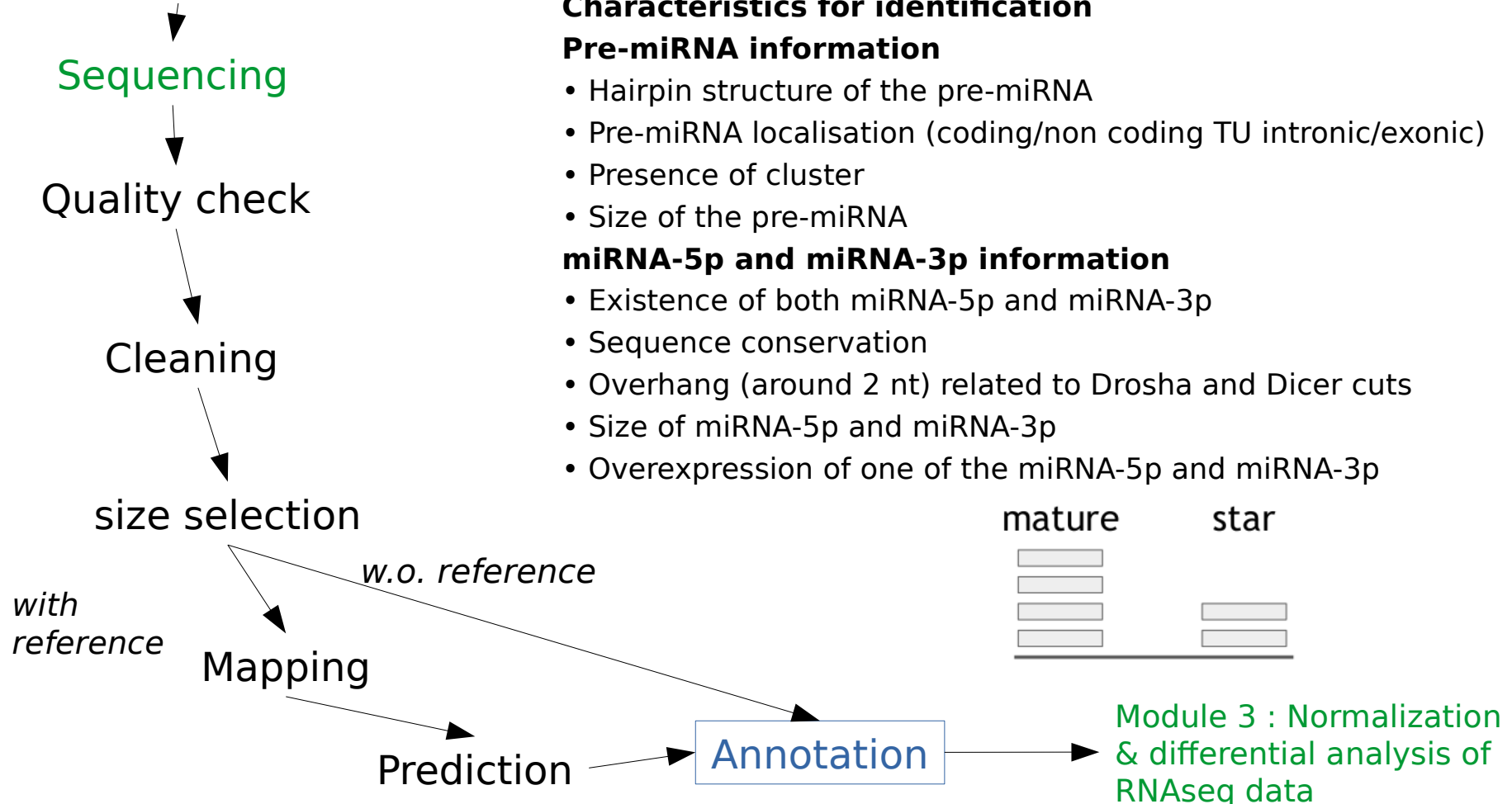
ex. of a classical miRNA

5'	Star	Mature	-3'	chr	exp	read	am	sample
.....	1	0	NS2
.....	1	0	NS2
.....	2	0	NS2
.....	2	0	NS2
.....	4	0	NS2
.....	1	1	NS2
.....	1	1	NS2
.....	1	0	NS2
.....
.....	1	0	NS3
.....	1	0	NS3
.....	1	1	NS3
.....	2	0	NS3
.....	1	0	NS3
.....	1	1	NS3
.....	2	1	NS3
.....
.....	1	0	NS1
.....	1	0	NS1
.....	1	1	NS1
.....	2	0	NS1
.....	2	0	NS1

Module 3 : Normalization & differential analysis of RNAseq data

smallRNAseq pipeline

Experimental design



smallRNAseq pipeline

Experimental design

Software:

Sequencing

Quality check

Cleaning

size selection

with reference

Mapping

Prediction

Annotation

Module 3 : Normalization & differential analysis of RNAseq data

Database:

- Rfam
- miRbase
- Silva
- GtRNAdb
- piRNA databank
- ...

H98-1976 Nucleic Acids Research, 2009, Vol. 37, Web Server issue
doi:10.1093/nar/gkp347

Published online 11 May 2009

miRAnalyzer: a microRNA detection and analysis tool for next-generation sequencing experiments

Michael Hackenberg¹, Martin Sturm², David Langenberger^{2,4},
Juan Manuel Falcon-Perez² and Ana M. Aransay^{1,2*}

¹Functional Genomics Unit, CIC bioGUNE, CIBERehb, Technology Park of Bizkaia, 48160 Derio, Bizkaia, Spain, ²Institute for Bioinformatics and Systems Biology, German Research Center for Environmental Health, Ingolstädter Landstrasse 1, D-85764 Nuerbingen, ³Department of Genome-Oriented Bioinformatics, Wissenschaftszentrum
Published online 16 May 2010 Nucleic Acids Research, 2010, Vol. 38, Web Ser

DSAP: deep-sequencing small RNA analysis

Published online 12 September 2011

Nucleic Acids Research, 2012, Vol. 40, No. 1 37-52
doi:10.1093/nar/gkr688

miRDeep2 accurately identifies known and hundreds of novel microRNA genes in seven animal clades

Marc R. Friedländer¹, Sebastian D. Mackow

BIOINFORMATICS APPLICATIONS NOTE

Sequence analysis

CPSS: a computational platform for the analysis of deep sequencing data

Yuanwei Zhang^{1,2}, Bo Xu^{1,3}, Yifan Yang², Rongjun Ban³, H Howard J. Cooke^{1,4}, Yu Xue^{5,6*} and Qinghua Shi^{1,4}

¹Hebei National Laboratory for Physical Sciences at Microscale and School of and Technology of China, Hele 230027, China, ²Department of Statistics, UI 40506, USA, ³Department of Computer Science & Technology, Nanjing Univ Genetics Unit, IGMM, University of Edinburgh, Edinburgh EH4 2XU, UK, and ⁴Huazhong University of Science and Technology, Wuhan 430074, China
Associate Editor: Ivo Haderik

shortran: A pipeline for small RNA-seq data analysis

Vikas Gupta^{1,2}, Katharina Markmann¹, Christian N. S. Pedersen³, Jens Stougaard¹ and Andersen^{1*}

¹Centre for Carbohydrate Recognition and Signalling, Department of Molecular Biology and Genetics, Aarhus Gustav Wieds Vej 10, 8000 Aarhus C, Denmark and ²Bioinformatics Research Centre, Aarhus University, C 8, 8000 Aarhus C, Denmark

BMC Bioinformatics



Software

miRExpress: Analyzing high-throughput sequencing data for profiling microRNA expression

Wei-Chi Wang¹, Feng-Mao Lin¹, Wen-Chi Chang^{1,5}, Kuan-Yu Lin^{2,3}, Hsien-Da Huang^{*1,4} and Na-Sheng Lin^{*2,3}

Address: ¹Institute of Biotechnology of Bioinformatics, National Central University, Chungli, Taiwan 31101, Taiwan, Republic of China

Address: ²Institute of Genome Biology 2010, 110101
http://genomebiology.com/2010/11/4/209



METHOD

Open Access

miRTRAP, a computational method for the systematic identification of miRNAs from high throughput sequencing data

NATURE BIOTECHNOLOGY VOLUME 26 NUMBER 4 APRIL 2008

NOTE Vol. 26 no. 20 2010, pages 2615-2616
doi:10.1093/bioinformatics/btq493

Discovering microRNAs from deep sequencing data using miRDeep

Marc R. Friedländer¹, Wei Chen², Catherine Adamidi³, Jonas Maackola⁴, Ralf Einspanier⁵, Signe Knespel⁶ & Nikolaus Rajewsky^{1*}

The capacity of highly parallel sequencing technologies to detect small RNAs at unprecedented depth suggests their value in systematically identifying microRNAs (miRNAs). However, the identification of miRNAs from the large pool of sequenced transcripts from a single deep sequencing run remains a major challenge. Here, we present an algorithm, miRDeep, which uses a probabilistic model of miRNA

and 454 Life Sciences/Roche, can sequence DNA orders of magnitude faster and at lower cost than Sanger sequencing and are evolving so rapidly that increases in sequencing speed by at least another order of magnitude seem likely over the next few years. Although the Solexa/Illumina system can produce ~32 million sequencing reads in one run, read length is currently limited to 35 bp. In contrast, the current 454 platform yields reads up to 200 bases each, although the number of reads
DOI: 10.1093/bioinformatics/btq493

Advance Access publication August 27, 2010

ep sequencing analysis
v², Gideon Dror², Eran Halperin^{3,4}

¹Academy, Tel Aviv University, ²The Academic Science Institute, Berkeley, CA, USA and ³School of Bioinformatics, George Washington University

smallRNAseq conclusion

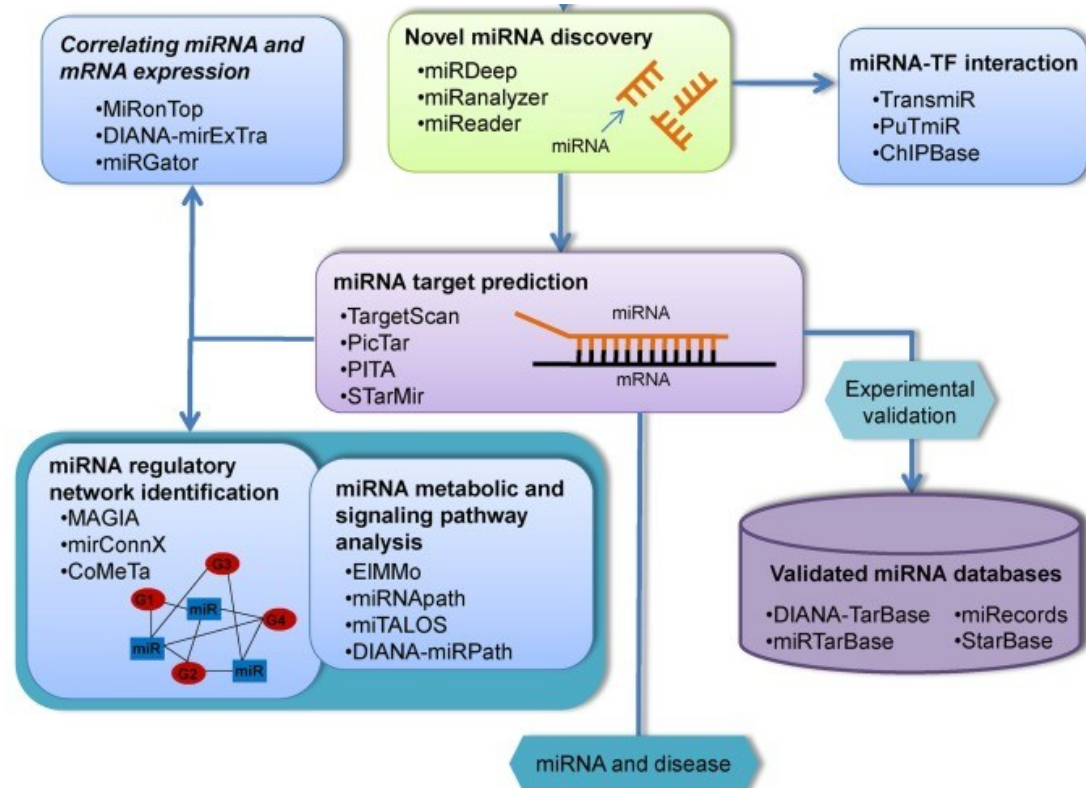
Presented example:
miRNA

Experimental design

Sequence characteristics
to select the expected
smallRNA

Annotation: first step of
the biological analysis

RNAseq or high-
throughput qPCR?



RNAseq analyses

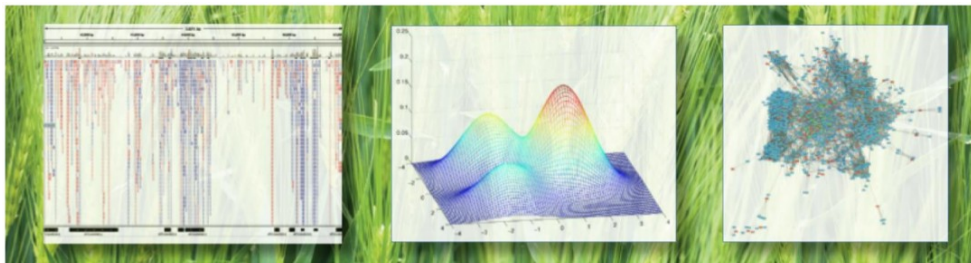
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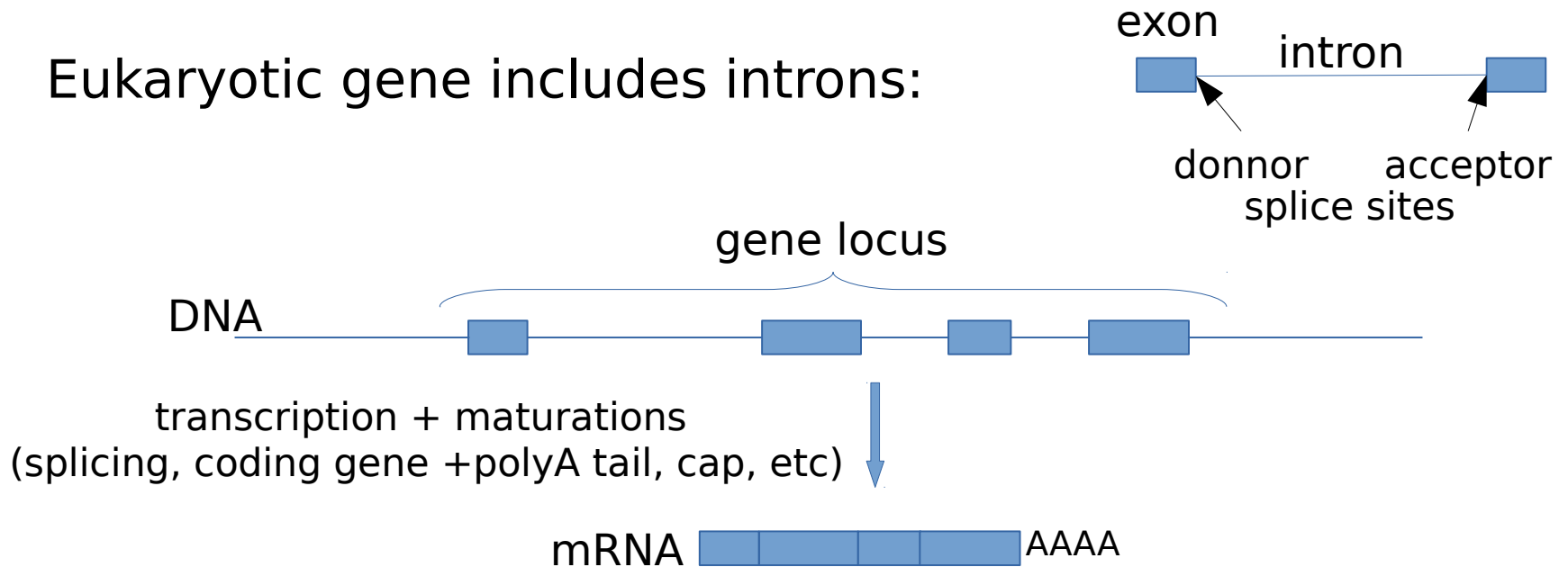
Others usages of RNA-Seq:

- smallRNA study
- **expression at isoform level**



Isoform

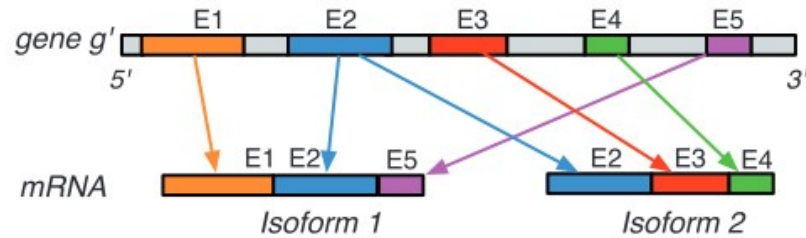
Eukaryotic gene includes introns:



One gene locus may rise diverse transcripts with different usages of exons

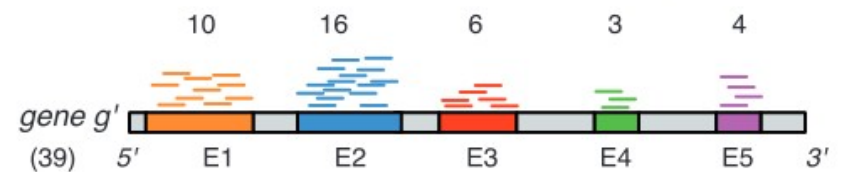
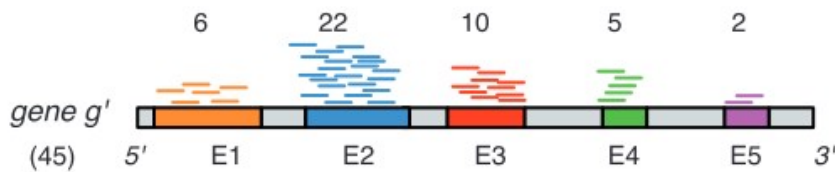
Alternative Splicing Event => isoforms

Gene level

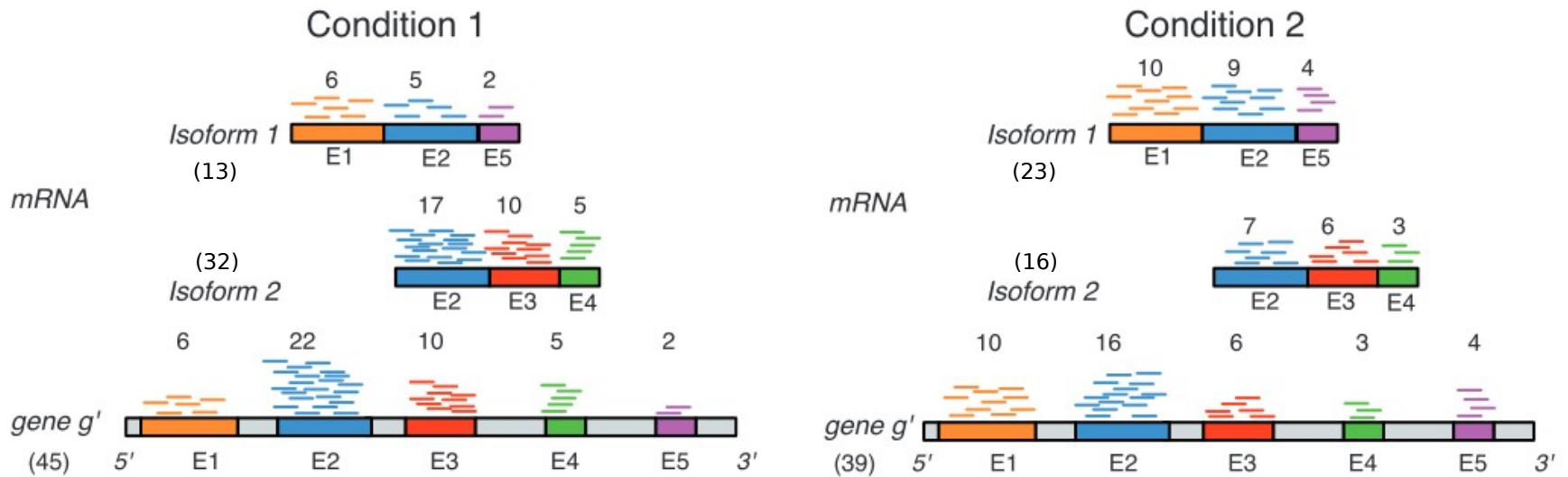
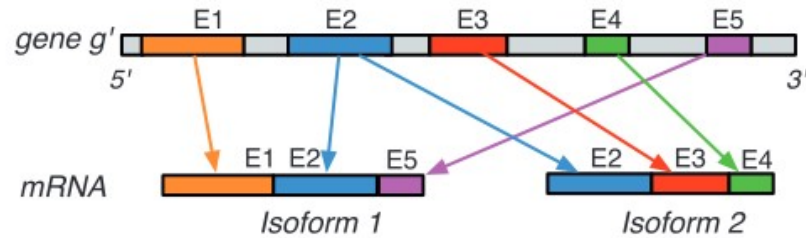


Condition 1

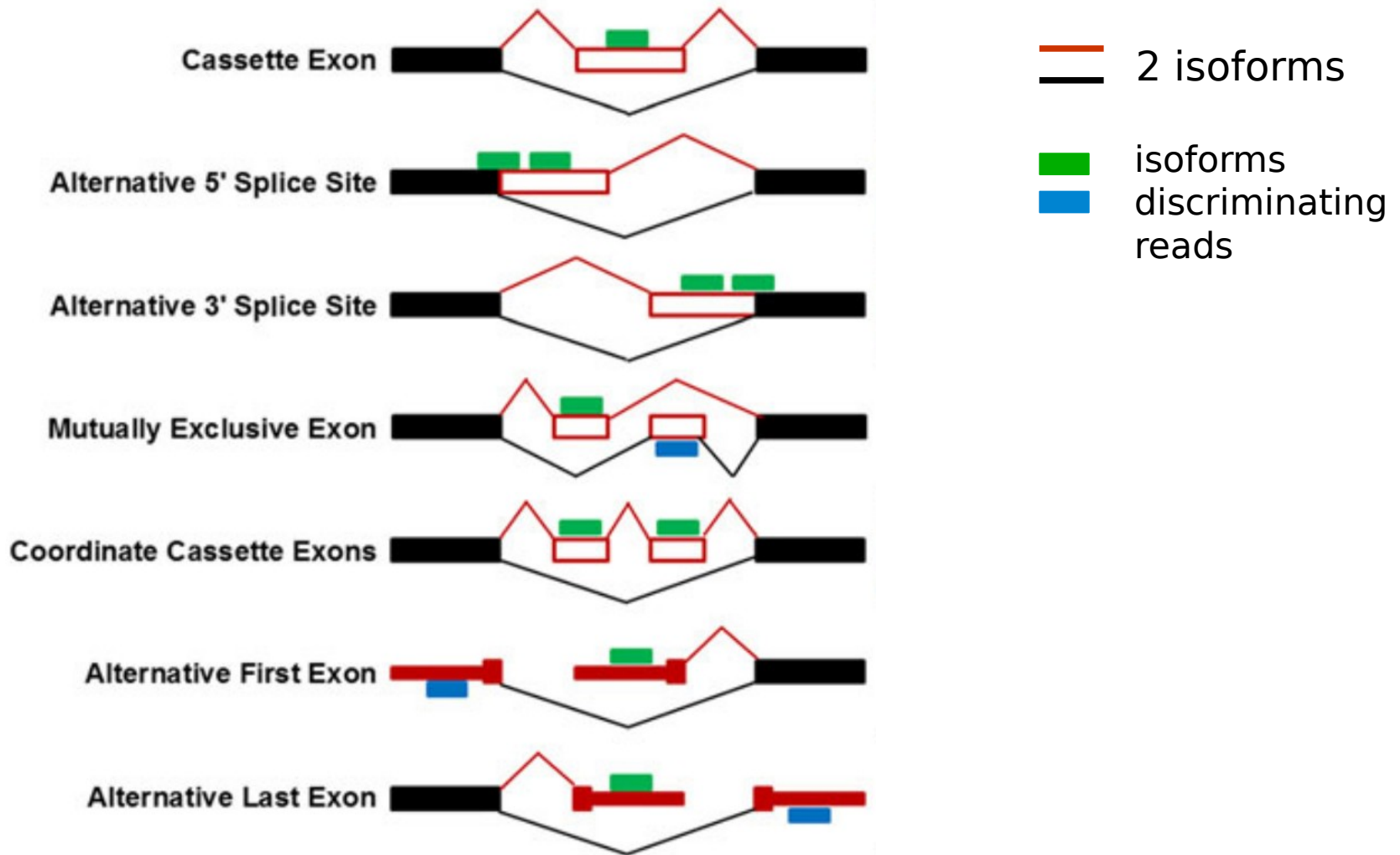
Condition 2



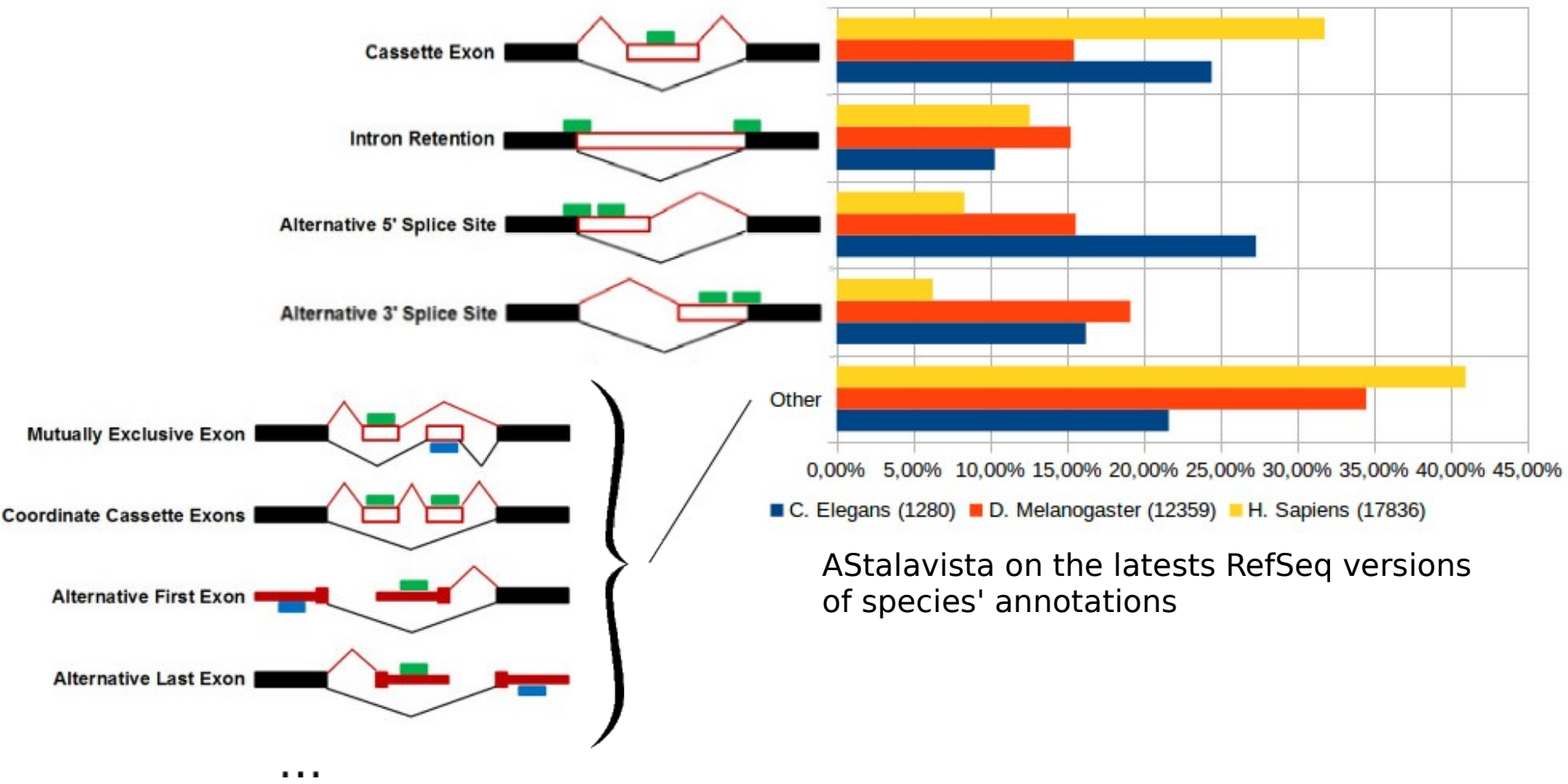
Transcript level



ASE-Alternative Splicing Events

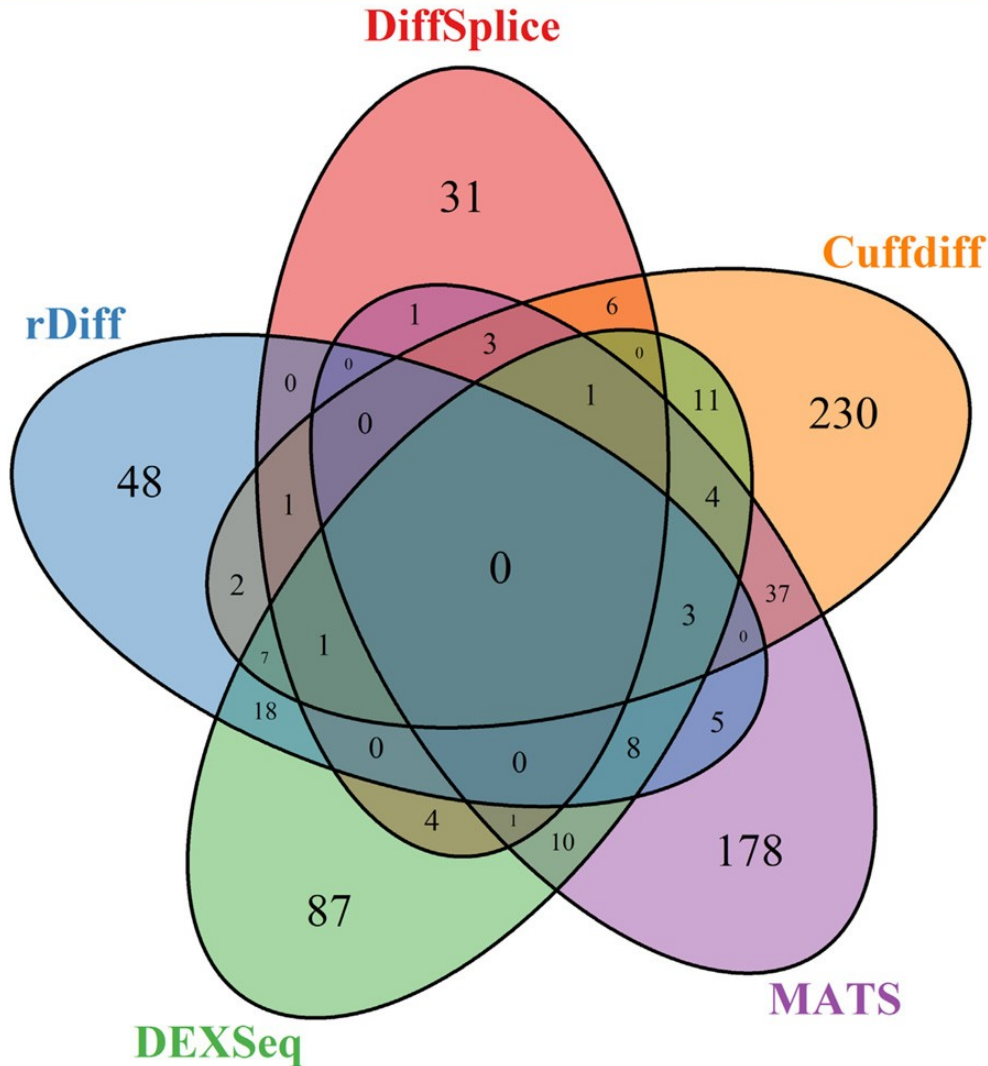


Organism specificity



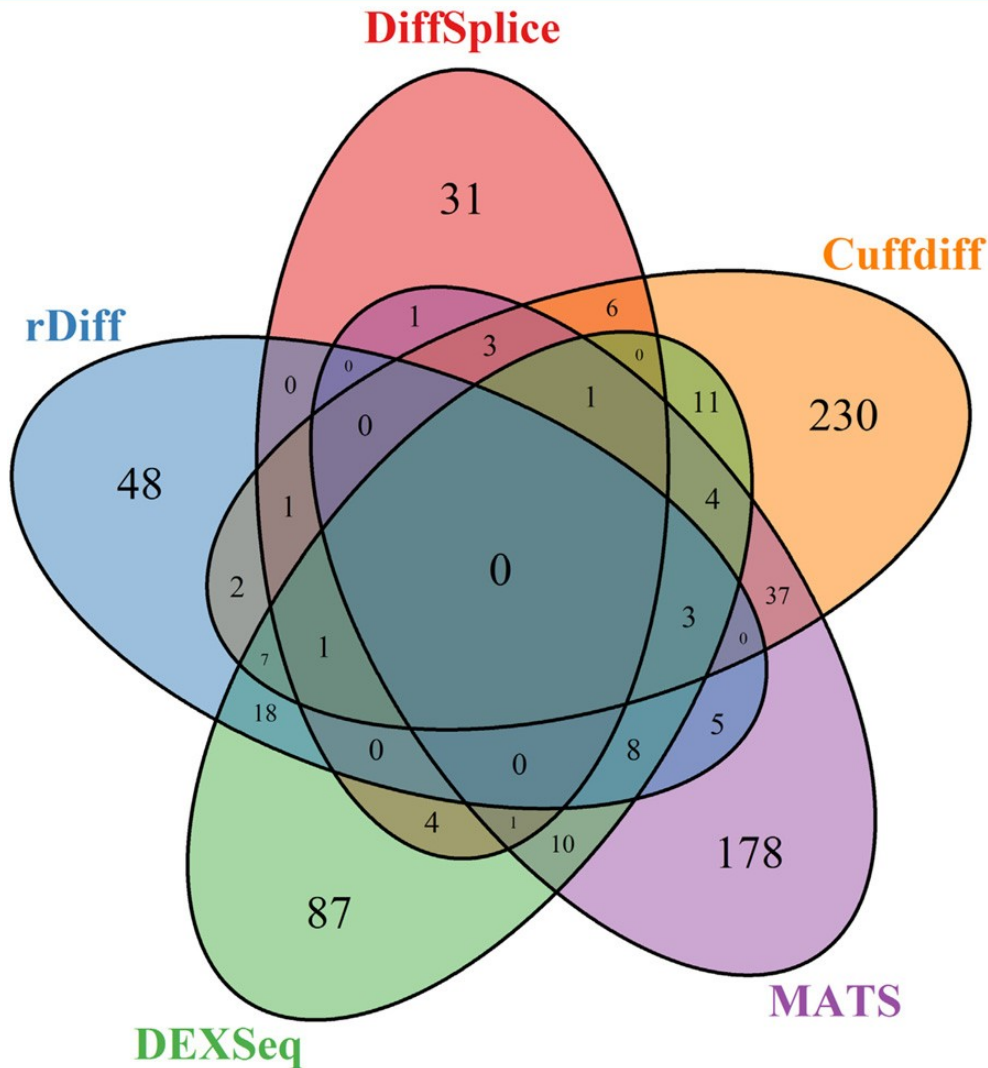
AStalavista on the latests RefSeq versions of species' annotations

Comparing DE tools



- 5 common tools for DE
- RNAseq data:
 - simulated data, *A.thaliana*
- number of common DE genes

Comparing DE tools



- 5 common tools
- RNAseq data:
 - simulated data, *A.thaliana*
- number of common DE genes

Hypotheses :

- level of analysis:
 - transcript \neq exon \neq region
- organism specificity
- different methods/algorithms:
 - mapping, counting, and DE

Benchmarking

Softwares?

- the number of tested methods is limited

Data?

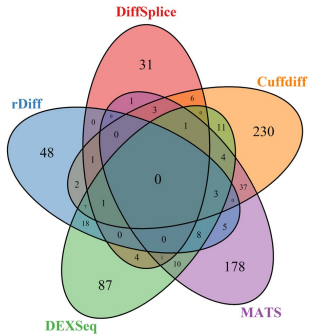
- RNAseq => how to know the truth?
 - qPCR studies? Only small set of genes & cross-hybridization between isoforms
 - RNA spikes? exogene sequences in controled quantity
- Simulated data => how to be as close as possible to the variability of the real data?

Task?

Some benchmarking tasks

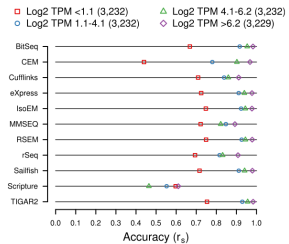
DE tools

How many common DE genes?

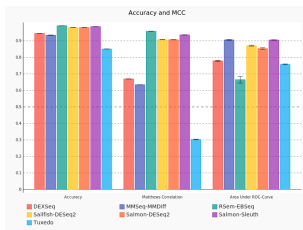


Isoform quantification

Expression rate, Exon number/transcript
Number of isoform/gene, ASE type



ASE detection



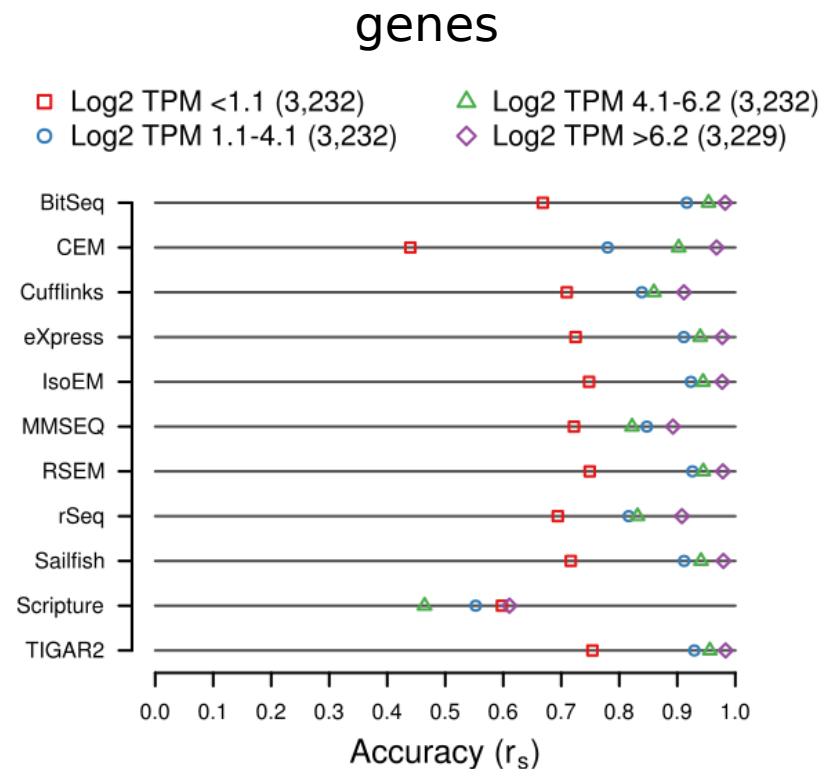
Quantification: expression rate

Simulation:
Flux simulator
Human data set
RNAseq single-end
sequencing depth: 30 million reads

restricted on expressed transcripts
(10% of human transcripts)

Spearman correlation coefficient (r_s)
between the estimates and the
known input levels

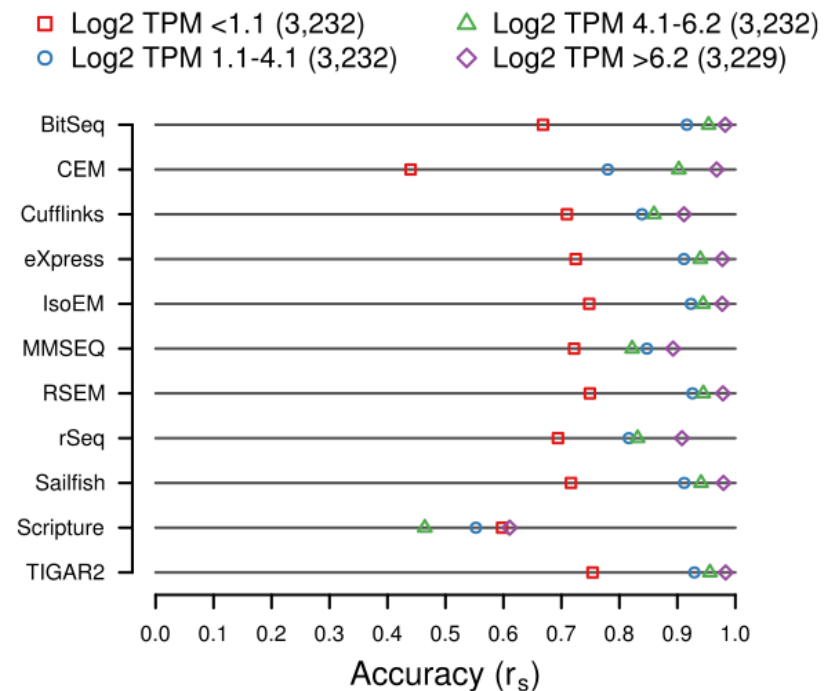
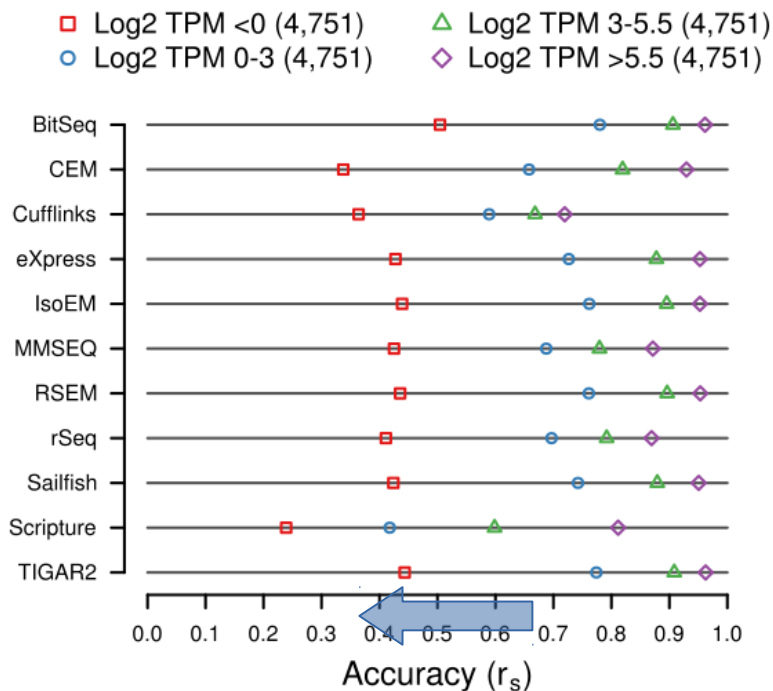
4 bins of expression levels
(Log2 TPM)



Quantification: expression rate

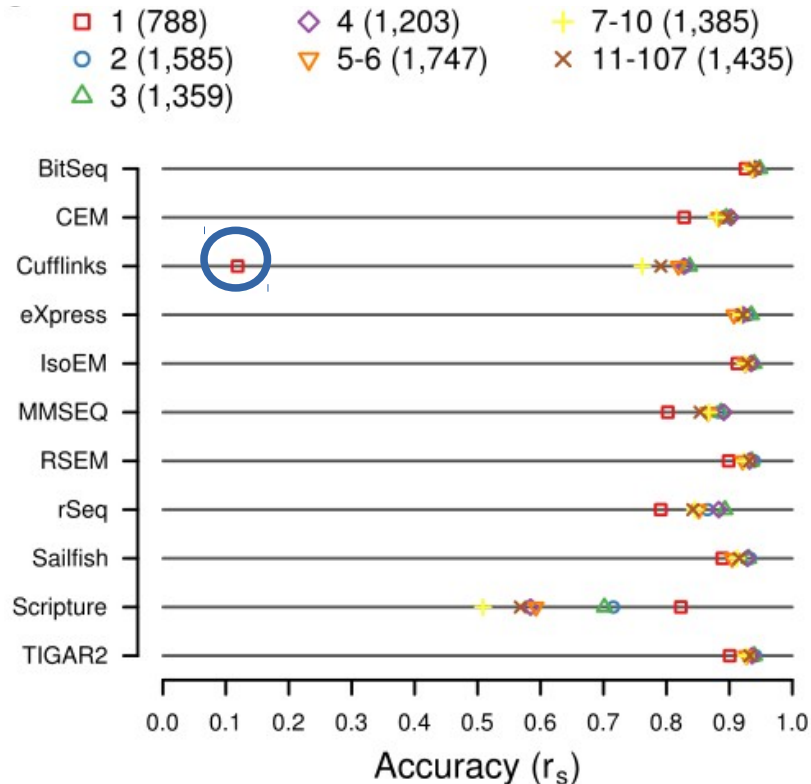
transcripts

genes



Quantification: exon number

#exons / transcript



Median expression levels:
 $0 < \text{Log}_2 \text{TPM} < 5.5$

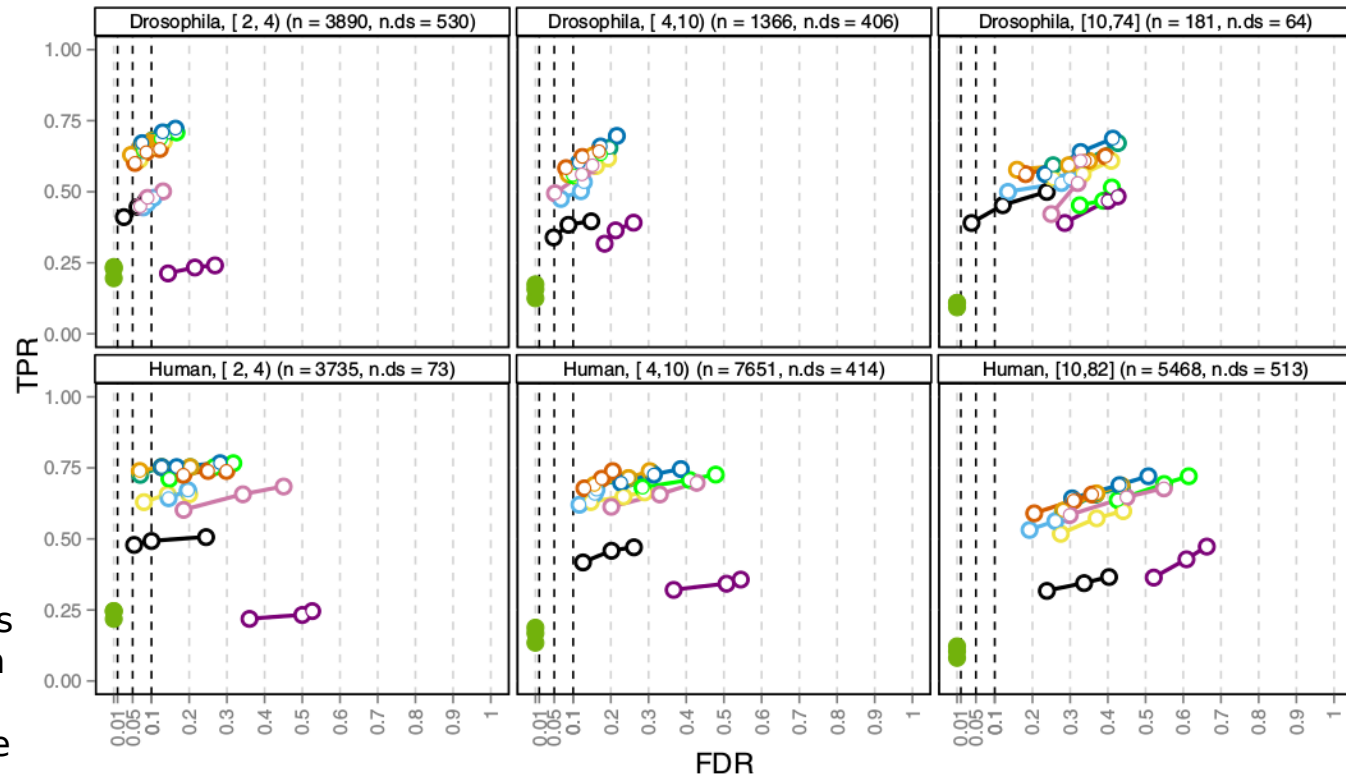
Cufflinks uses read-overlapping junction

Quantification: isoform number

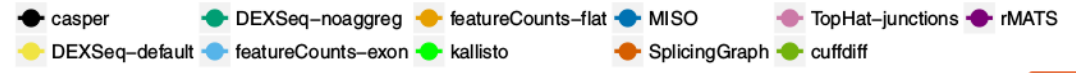
Simulation:
RSEM model from real data set + DTU for 1000 genes (switch of the relative abundances for the 2 most abundant isoforms between the conditions)

[i,j] i to j isoforms
n: gene number
n.ds: with DTU

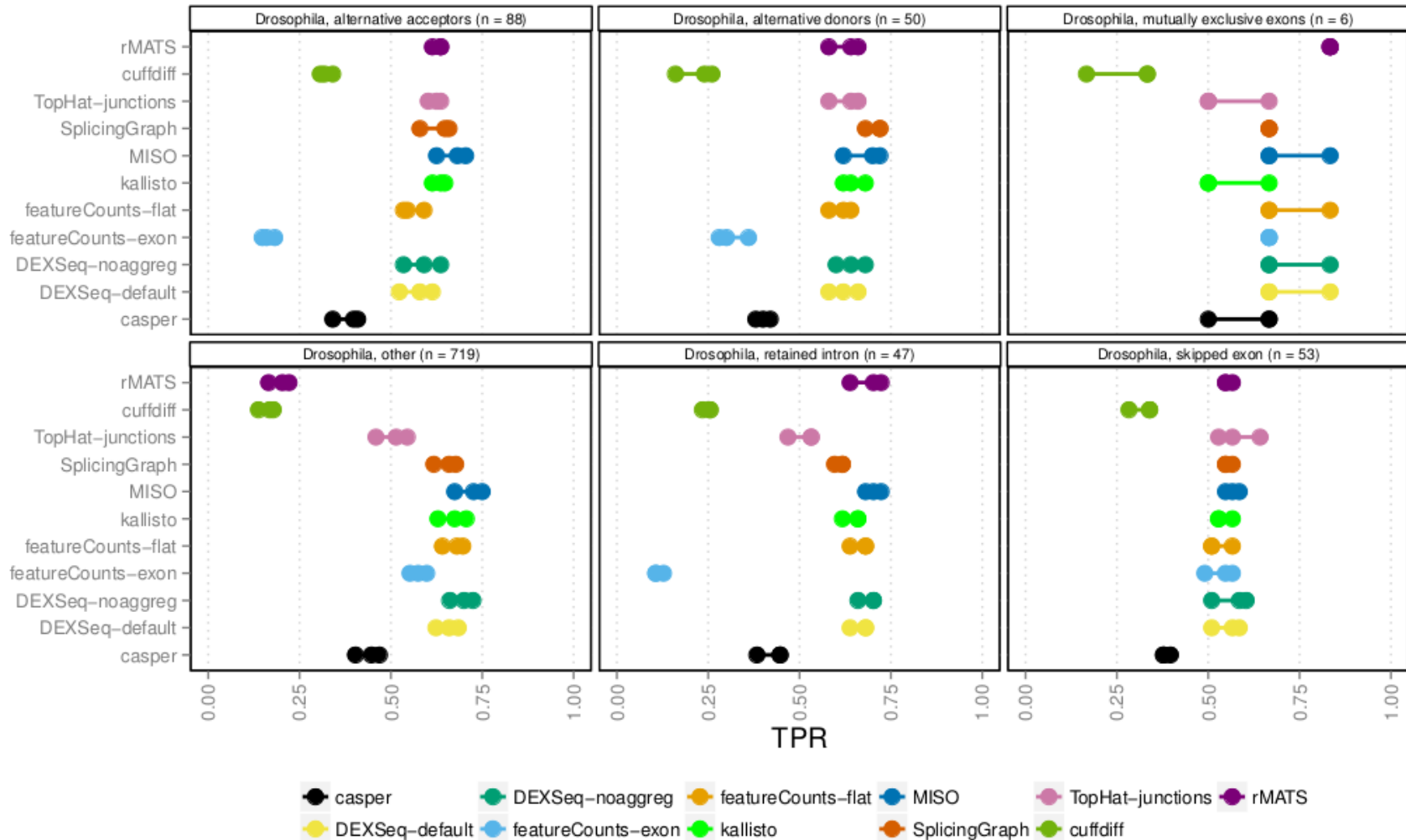
3 circles, usual FDR thresholds (0.01, 0.05, 0.1): ideally, each circle should fall to the left of the corresponding vertical line



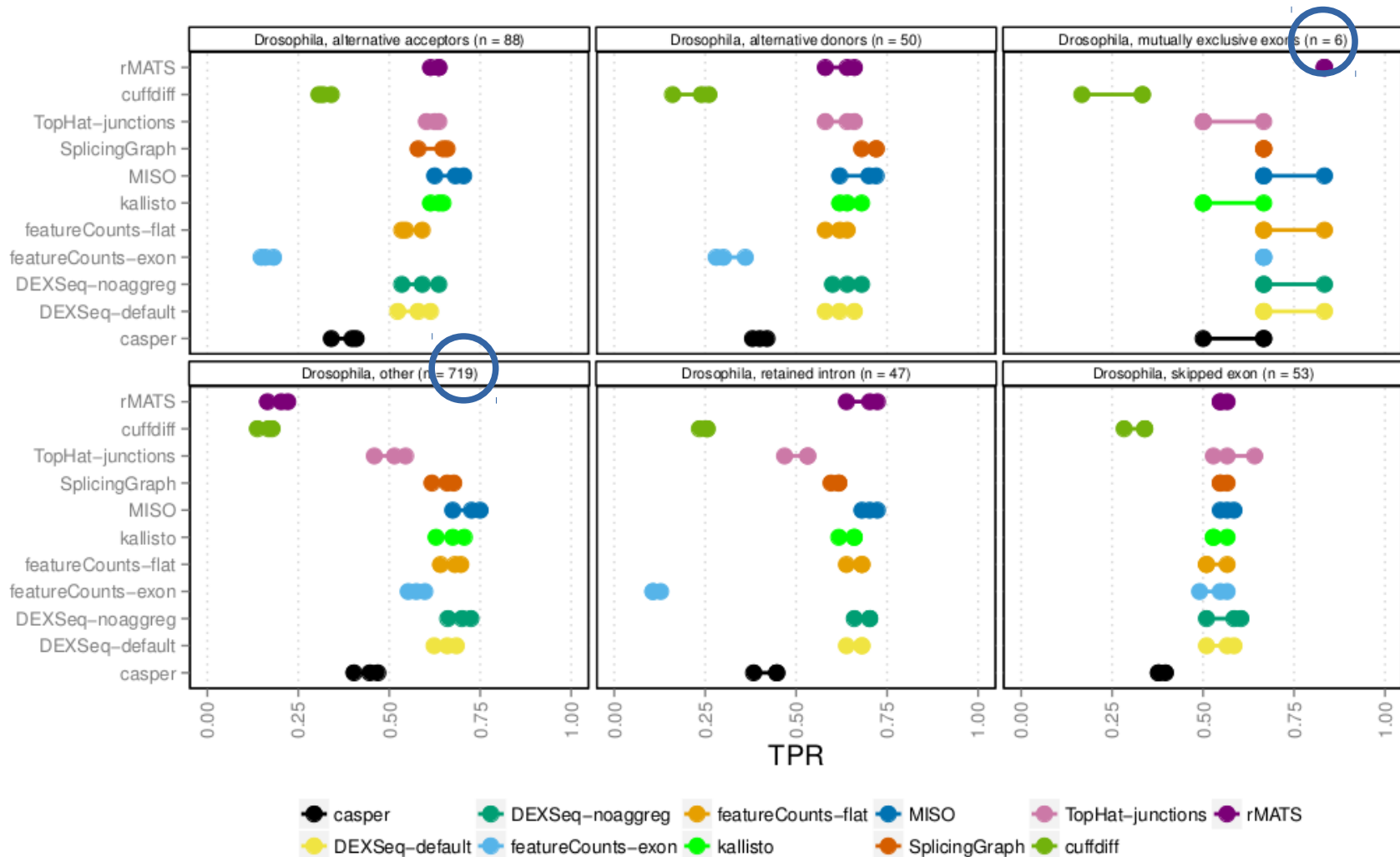
DTU: Differential Transcript Usage
FDR: False Discovery Rate
TPR: True Positive Rate



Quantification: ASE ?



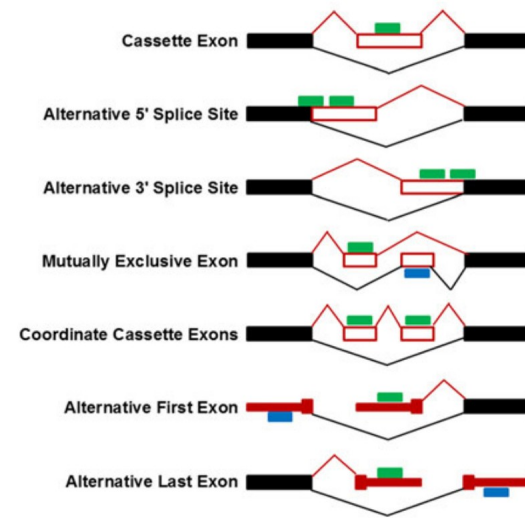
Quantification: ASE ?



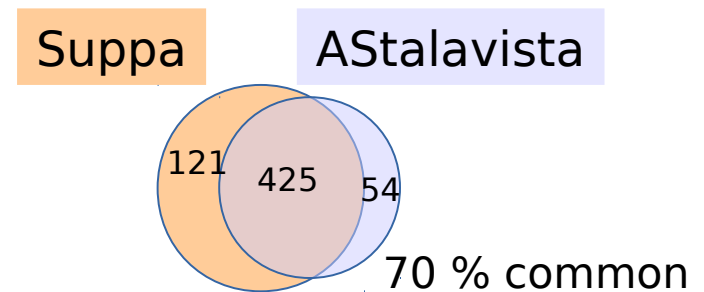
ASE analysis

Annotation file
(gtf, gff, gbk)
gene
exon
transcript
CDS

→ ASE list
Suppa
AStalavista



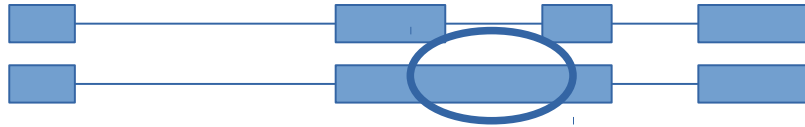
chr22 hum.	Suppa	AStalavista
with ASE	539	479
without ASE	733	793
total	1272	1272



ASE analysis need reference

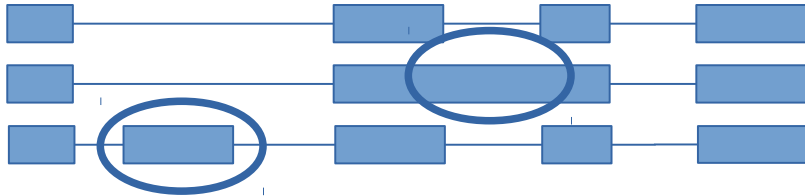


ASE analysis need reference



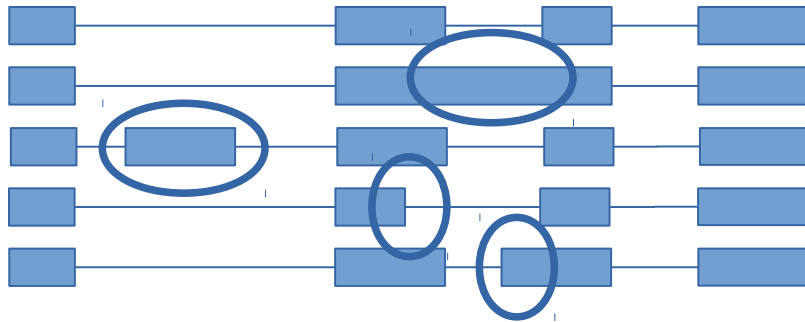
Retained Intron

ASE analysis need reference



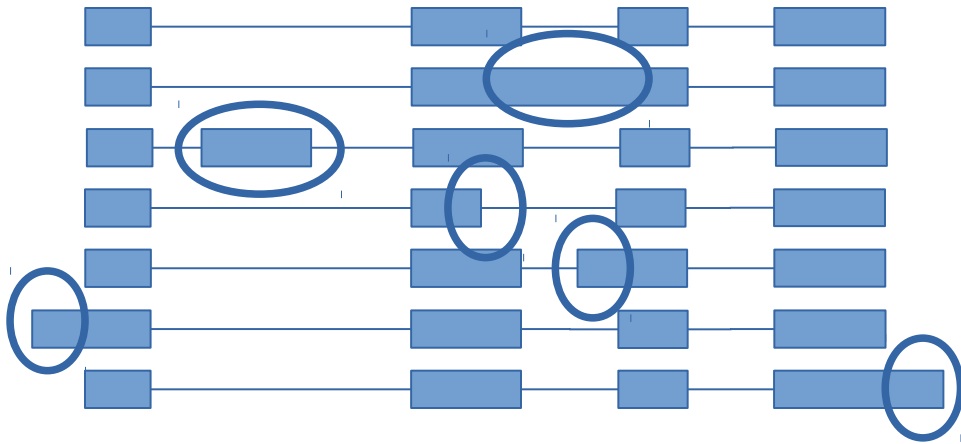
Retained Intron
Skipped Exon

ASE analysis need reference



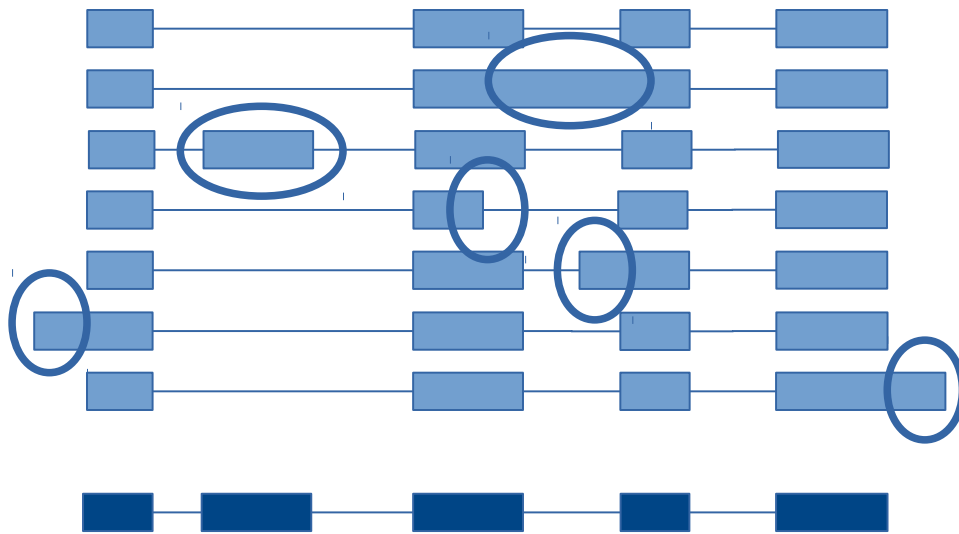
Retained Intron
Skipped Exon
Alternative Donnor/Acceptor
Splicing Site

ASE analysis need reference



Retained Intron
Skipped Exon
Alternative Donnor/Acceptor
Splicing Site
Alternative First/Last Exon

ASE analysis need reference



Retained Intron
Skipped Exon
Alternative Donnor/Acceptor
Splicing Site
Alternative First/Last Exon

Define a Reference Transcript :

- 1 - **largest set** of **non-overlapping** exons
- 2 - that appends the **most frequently** among isoforms
- 3 - that covers the **widest area** over the gene region

=> may be a non “real” transcript
=> specific for each project

Benchmark: ASE detection

Evaluate tools in their capacity to detect ASE from RNAseq data (neither the « right » rate of transcript expression, nor discovery of new expressed loci)

Simulated data: controlled expression rate of each isoform & the presence of each type of ASE :

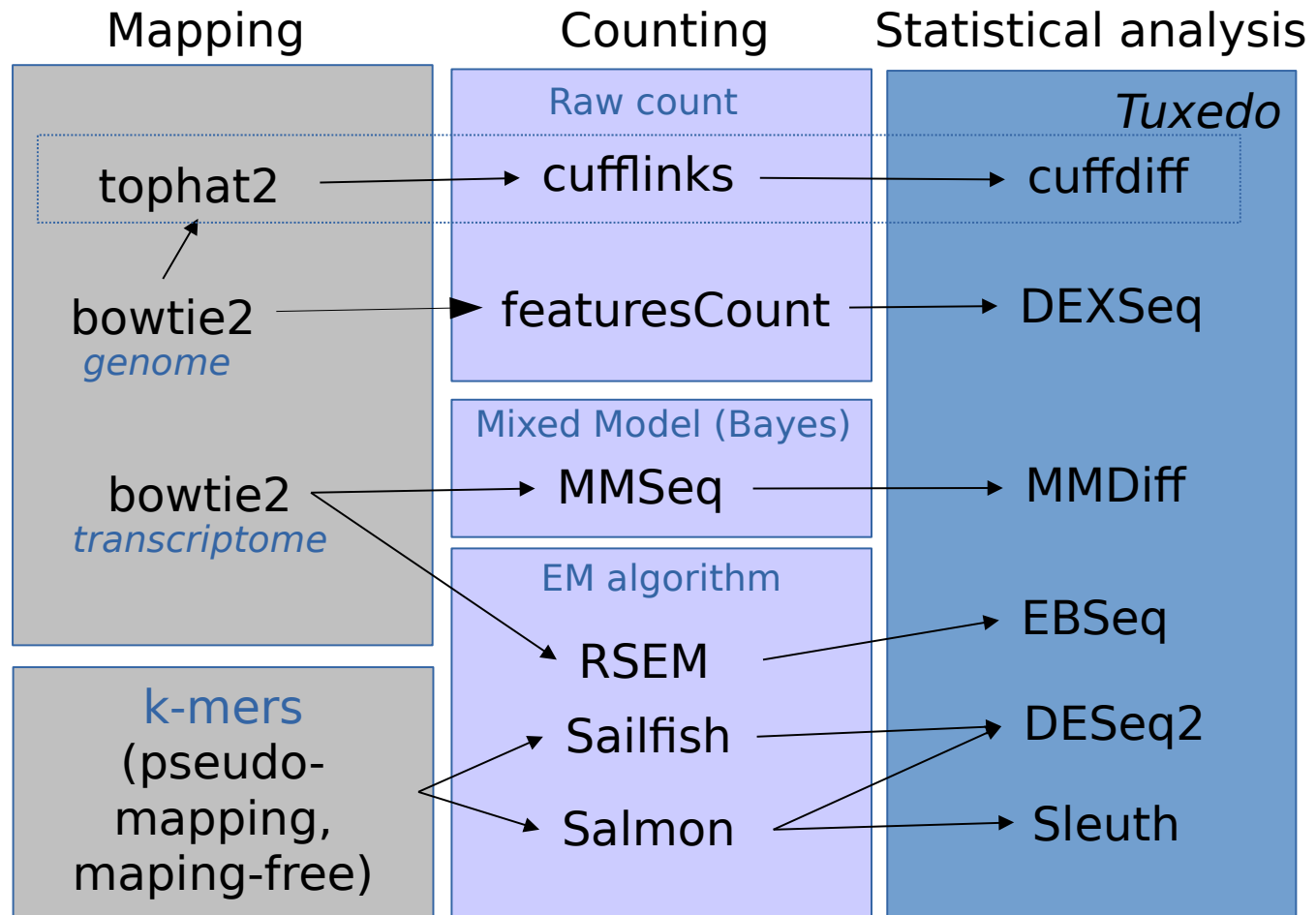
- human chromosome 22, 744 genes
- 2 conditions x 3 replicates
- reads : paired-end, 2 x 100 bp
- expression : 100 reads / transcript (no DE variation)
- For each type of ASE :
 - 10% of the transcript in one condition
 - Only reference transcripts in the other condition
 - 1 ASE/transcript/gene

Condition 1
100% references

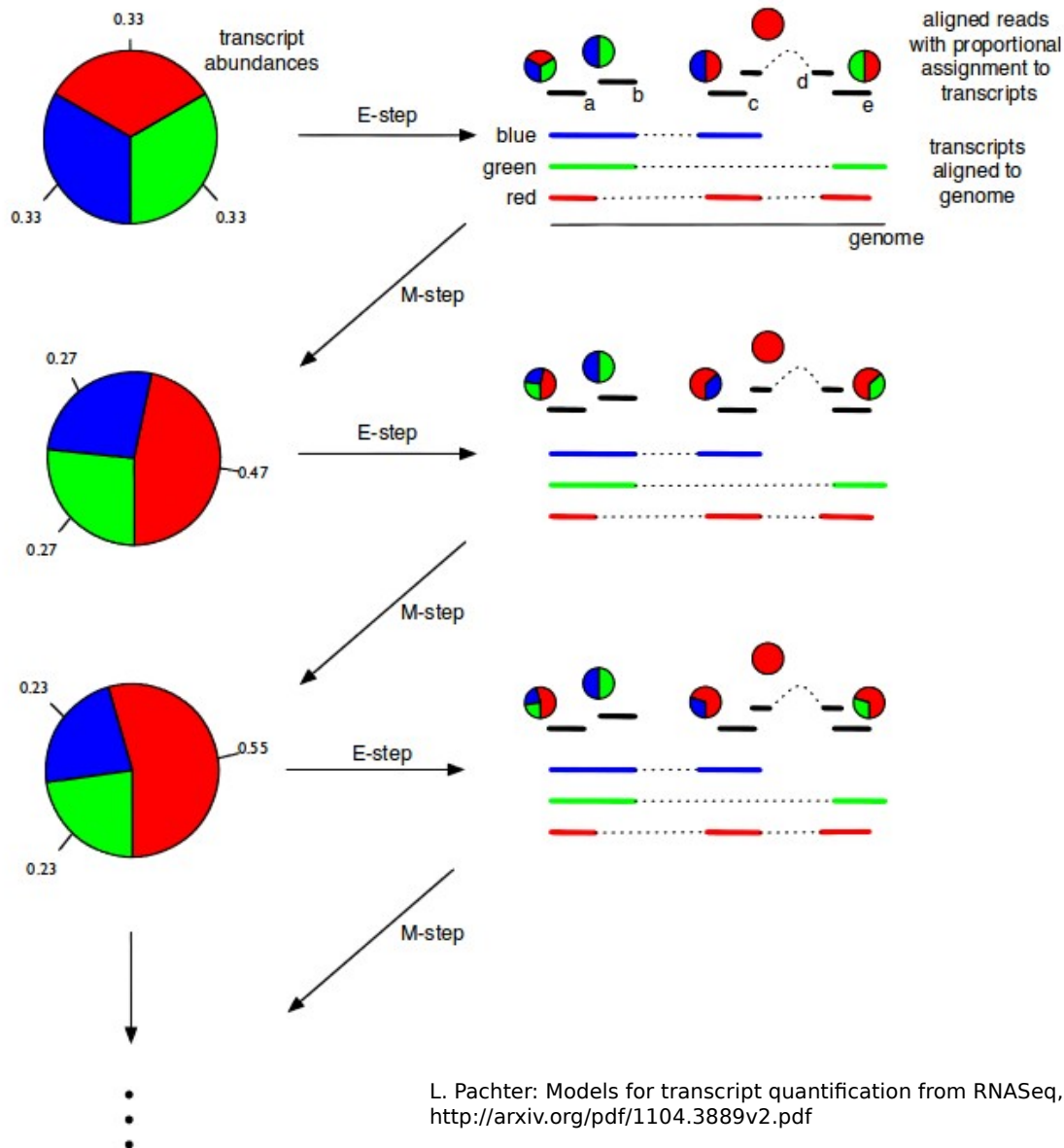
Condition 2
90% references
10% with 1 ASE

Identification of a DE transcript => the method detects the ASE type

Methods & tools benchmark



Count estimation with EM



Based-on the Expected-Maximisation algorithm

First 3 cycles of EM algorithm :

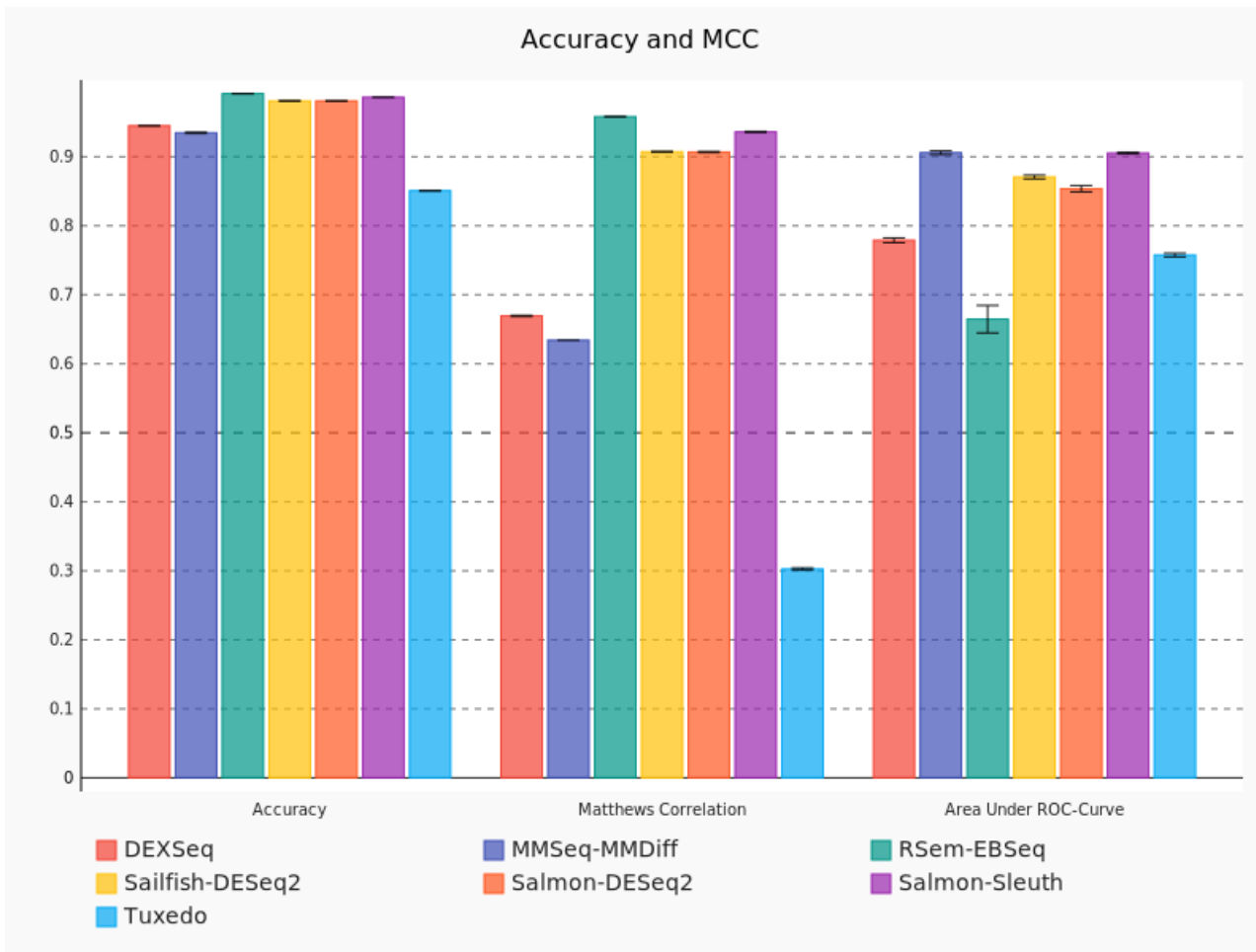
ex. : abundance of **red** transcript estimated after the 1st M-step:
 $(1/3 \text{ read a} + 1/2 \text{ read b} + 1 \text{ read d} + 1/2 \text{ read e}) / (\text{total read number})$

or $(0.33 + 0.5 + 1 + 0.5) / 5 = 0.47$

- proved to converge

- stop criterion implementation:
 when all probabilities that a fragment is derived from a transcript $\geq 10^{-7}$ have a relative change of $\leq 10^{-3}$

Results, alternative donor site



Accuracy: are tool predictions correct?

RSEM/EBSeq, Sailfish, Salmon

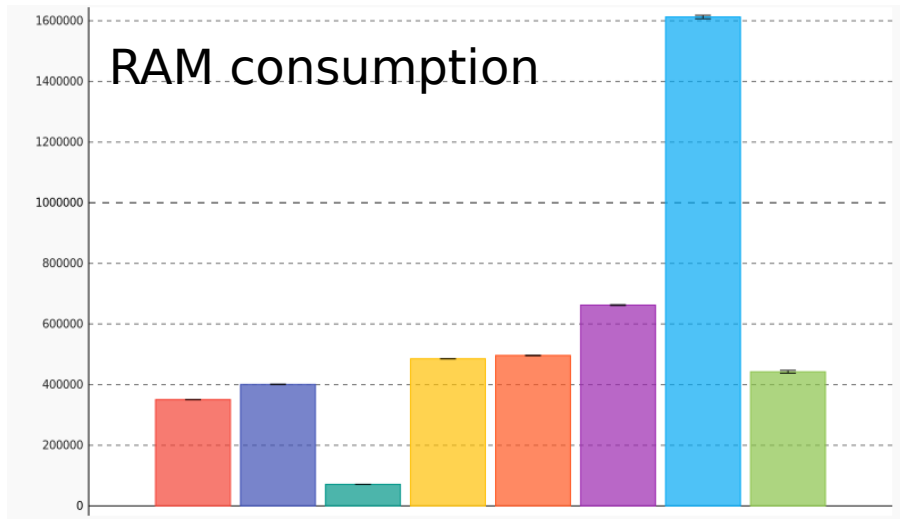
MCC: if >0.5 then using tool is better than random

RSEM/EBSeq, Salmon-Sleuth
Tuxedo

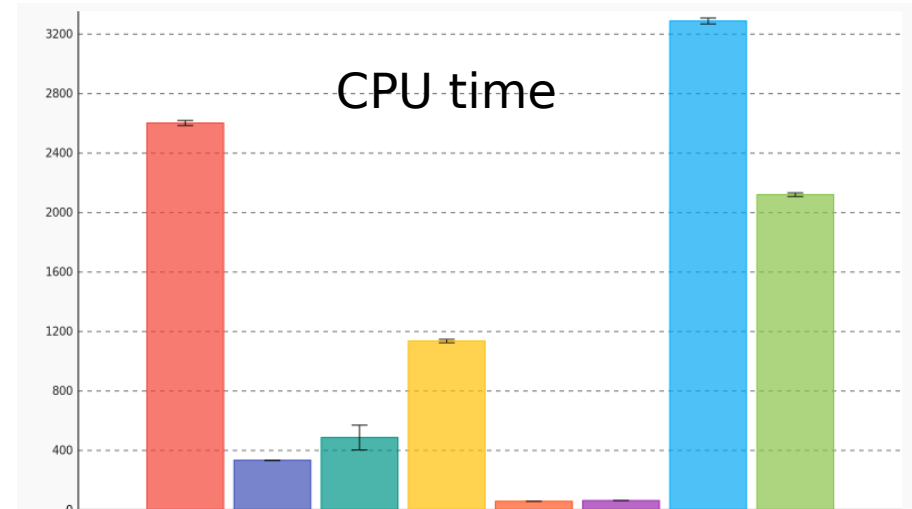
AUC of ROC curve: confidence in the results of the tool
Salmon-Sleuth

Performances (ADss)

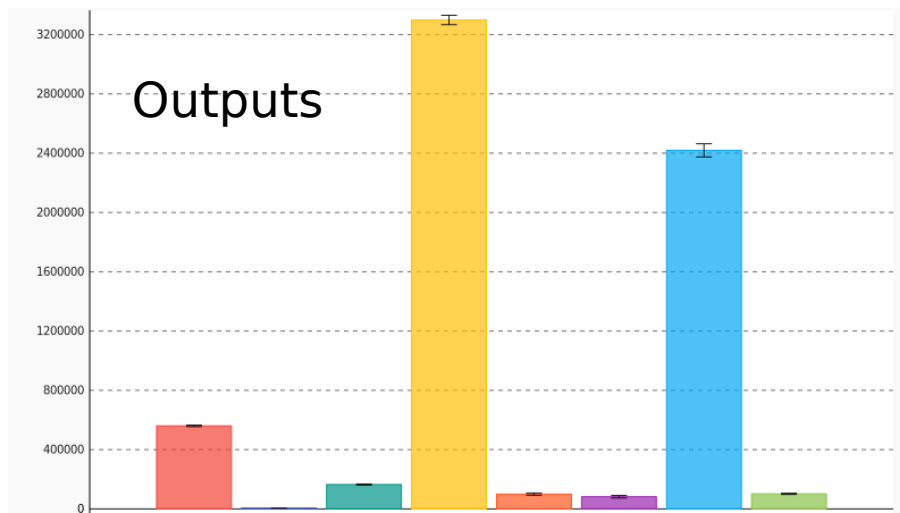
Max RAM usage (Ko)



CPU time (seconds)



Number of outputs



- Bowtie2
- DEXseq
- MMSeq/MMDiff
- RSEM/EBseq
- Sailfish/DESeq2
- Salmon/Sleuth
- TopHat2
- Cufflinks/CuffDiff

Tuxedo : bowtie2+Tophat2+cufflinks/cuffdiff

Choice of the method

For our benchmark

human chr22, 10 % of ON/OFF transcripts with 1 ASE
between 2 conditions, 3 replicates, 100 reads/genes

Salmon/Sleuth

(RSEM/EBSeq, Sailfish)

developped pipeline: fastq \Rightarrow DE transcripts

appliance for IFB cloud, <https://cloud.france-bioinformatique.fr>

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SIGN IN



ASE Benchmarking conclusions

A benchmark is always limited:

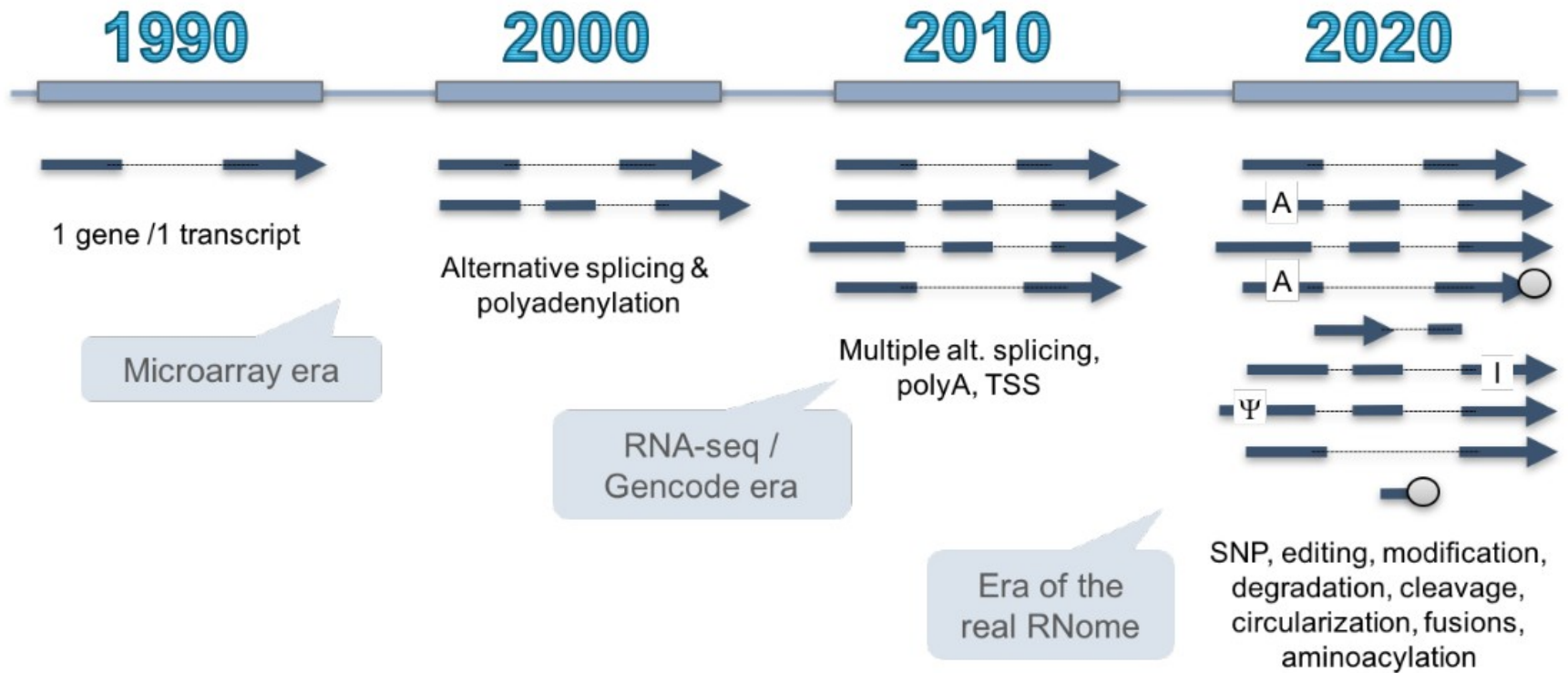
- › Software
- › Simulation (human, plant, ...)
 - Skipped exon, Retained Intron, Alternative First/Last Exon, Alternative 3'/5' Splicing Site
 - 100 reads / gene (3-4 days, 6 cores, 13 G RAM)
 - 1000 reads / gene (to check if the wrong predictions result from the coverage deepness or from the algorithm)
 - On/off condition : 1 ASE / transcript / gene :
 - more than 1 ASE / transcript
 - combination of ASE in the same transcript

Several benchmark studies should be conducted to have a global overview

The conclusions should be regularly update

RNomics evolution

Part1-smallRNA: search for the method corresponding to the smallRNA
Part2-isoform level: will be easier with full length RNAseq technology



The more there is technical advances, the more we are going towards the unknown biology

Thanks



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Eukaryotic small RNA



ifb

AVIESAN-IFB



P. Bardou, C. Gaspin, S. Maman, J. Mariette, O. Rué, M. Zytnicki
<http://www.france-bioinformatique.fr/sites/default/files/sRNA-Seq.pdf>

Isoforms



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