# Analyse de données de séquençage haut débit

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- Sequencing is cheap
- Applications include
  - De novo genome assembly
  - Resequencing (SNPs, indels, rearrangements)
  - Transcriptome analysis (RNA-seq)
  - Protein-DNA interactions (Chip-seq)
  - Metagenomics
- The difficulty is not to generate data, but to analyse it
  - Many datasets are superficially analysed, or not analysed at all
  - Bioinformatics bottleneck





- Bioinformatics analysis of sequencing data starts by either
  - Mapping your reads to a reference genome
  - De novo assembling your reads
- Downstream analysis depends on the application (RNAseq, ChipSeq, etc)
- For each application, there is not yet a unique reference pipeline
- Choosing which pipeline to use requires to understand what it captures/misses



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- Choosing which pipeline to use requires to understand what are its limitation



# Complementarity of assembly-first and mapping-first approaches for alternative splicing annotation and differential analysis from RNAseq data

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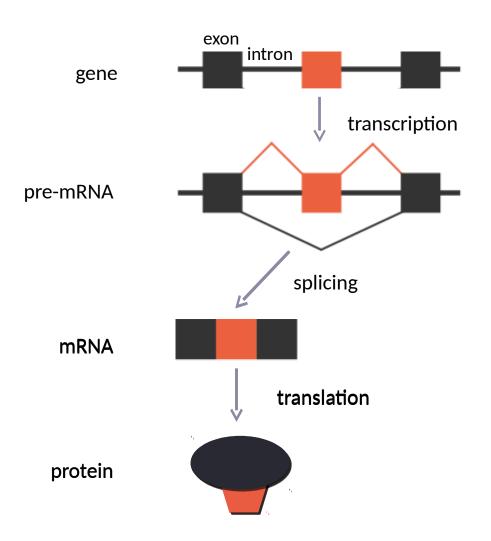








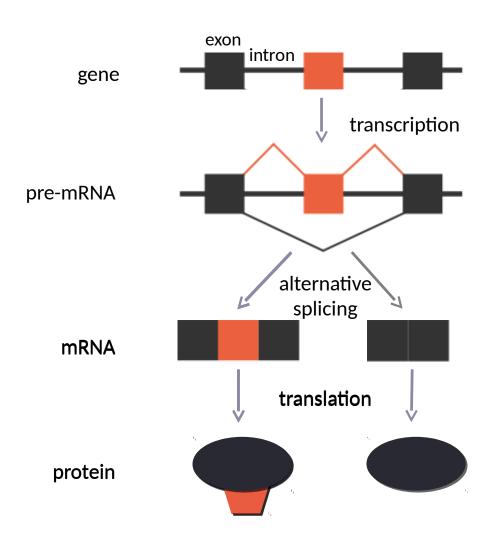
# **Alternative Splicing**







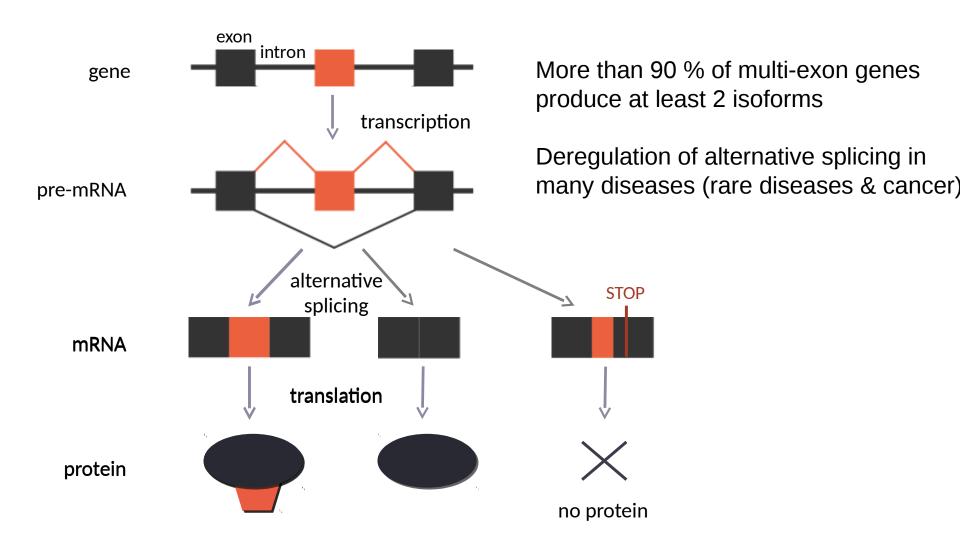
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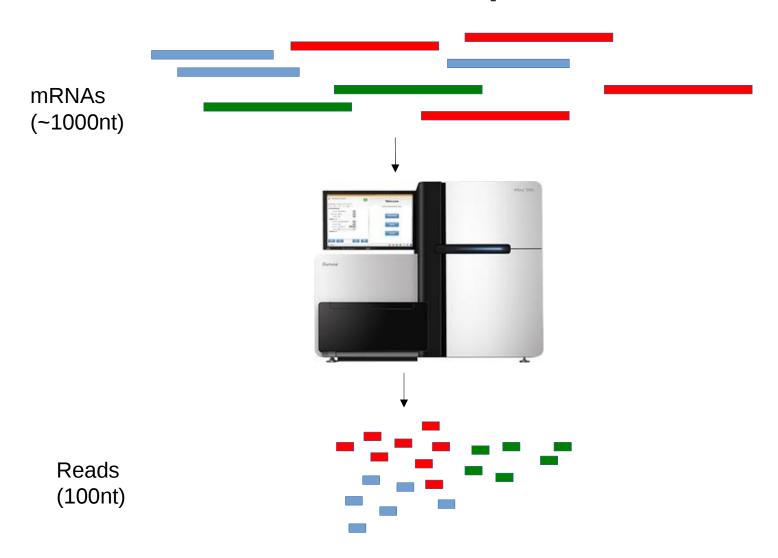
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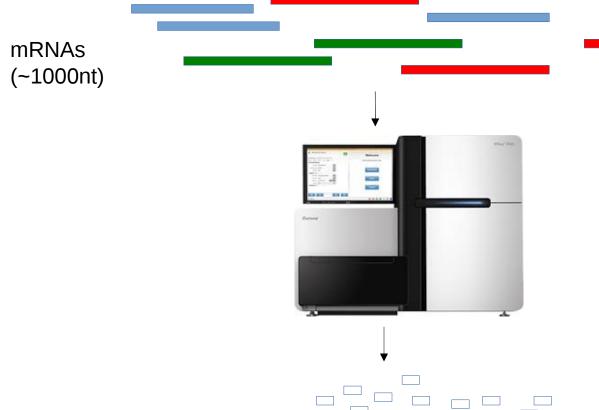
# RNAseq data







# RNAseq data



Reads (100nt)





# Alternative Splicing and RNA-seq data

- Gencode Annotations: 60 000 genes, 3 transcripts per gene.
- Assessing which gene/transcript is expressed in which tissue/condition can in principle be done through RNAseq
- The main challenges are:
  - Reads are short (100nt) and can be assigned to multiple transcripts (1000nt)
  - Some transcripts are annotated, some are novel
  - Some transcripts are highly expressed, many are poorly expressed





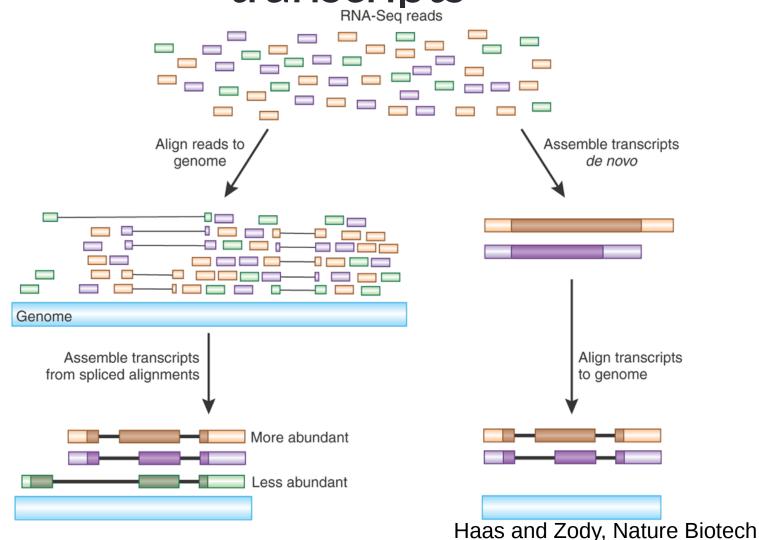
# **Annotation and Differential Analysis**

- Annotation: Identify and quantify all transcripts present in a given condition
- Differential Analysis: Assess which genes are differentially spliced across conditions (treatment / control, population 1 / population 2, disease / control)





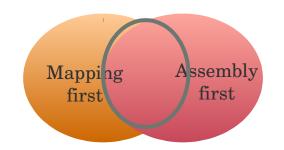
# Two approaches to assemble transcripts





# What is the overlap between the predictions of the two approaches?

Identify pros and cons of assembly-first and mapping-first methods

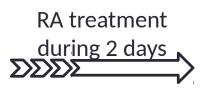


→ Comparison done on alternative skipped exon (ASE) events only



→ Public dataset (ENCODE) from neuroblastoma SK-N-SH cell line with o without retinoic acid (RA) treatment





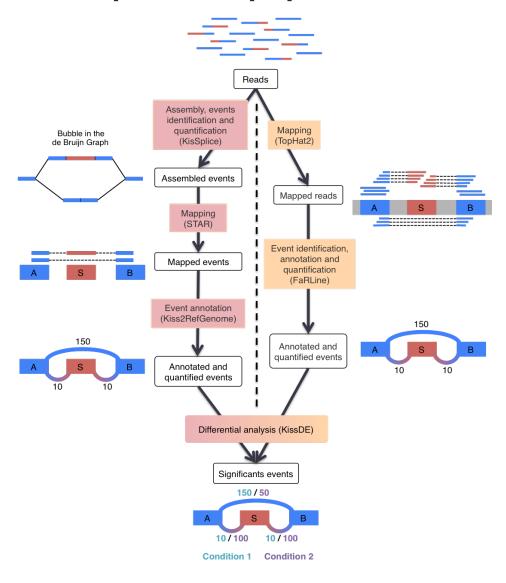






### Technologies pour la santé

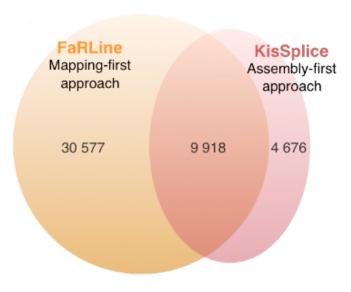
# Compared pipelines







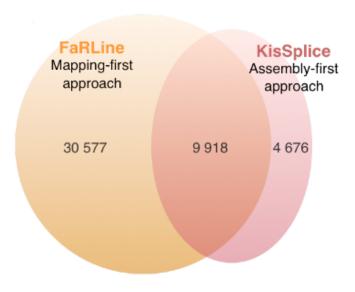
# Mapping-first approach finds many unfrequent variants

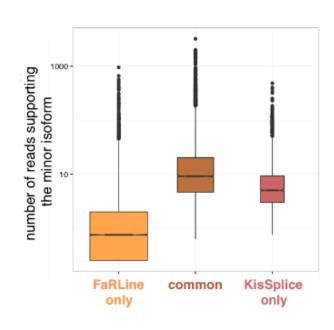






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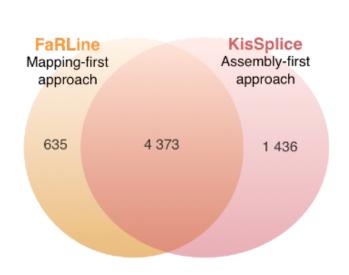


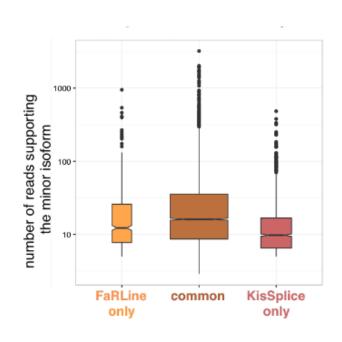






# The overlap between methods increases when unfrequent variants are filtered out

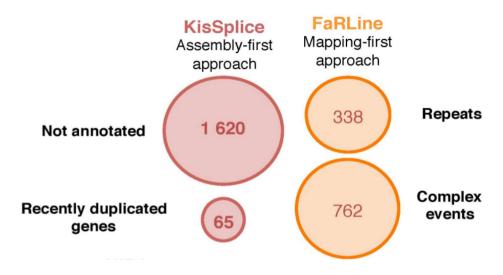








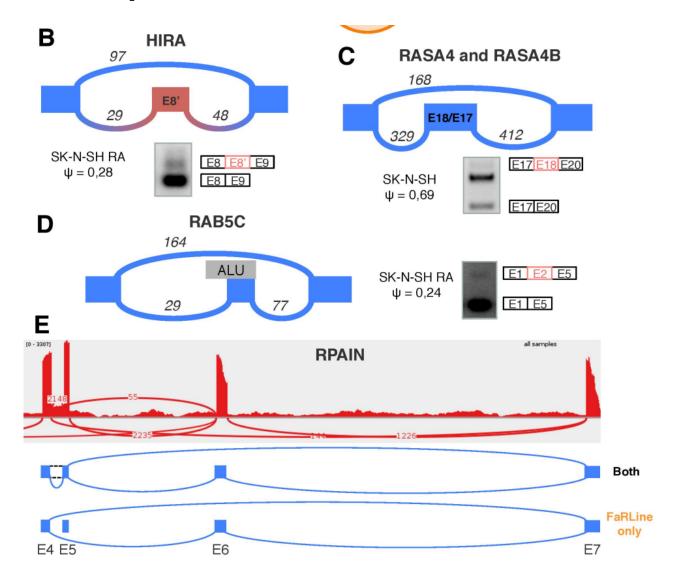
# Some abundant transcripts are systematically missed by one approach







### **Experimental Validations**







### **Annotation summary**

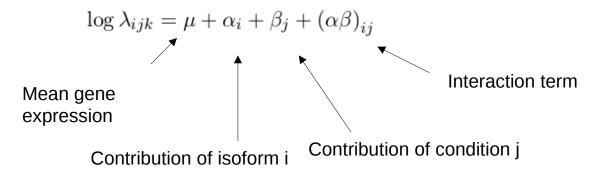
Mapping-first is stronger for rare variants and exonised Alus Assembly-first is stronger for novel variants and recent paralogs

Should I care about these differences?

Does it have an impact on my differential analysis?

# **Statistical Analysis**

- Count regression with negative binomial distribution
- Generalised linear model, 2 way design, with interaction

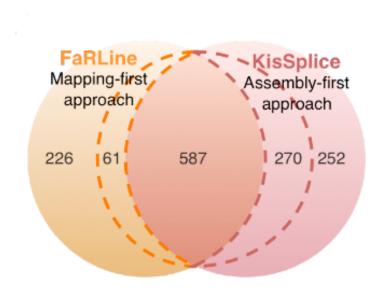


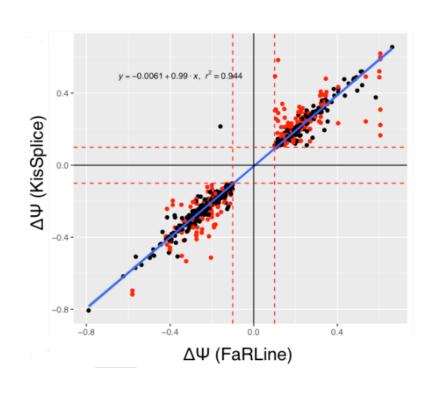
- Target hypothesis:  $H_0 : \{(\alpha \beta)_{ij} = 0\}$
- Likelihood ratio test
- P-values adjusted with benjamini-hochberg procedure





# Comparison after differential analysis



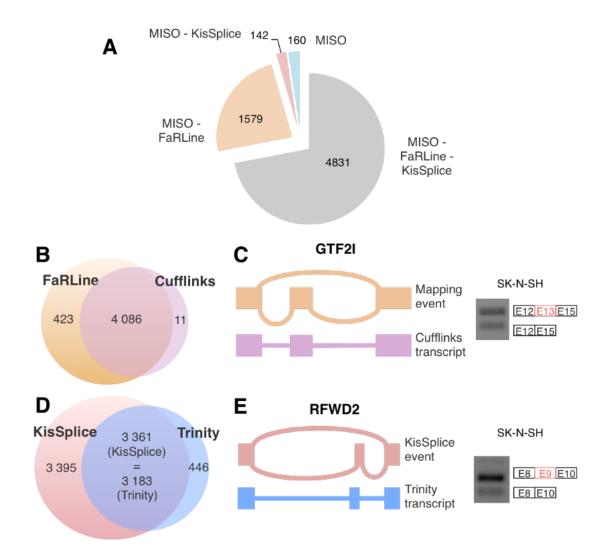


AS events predicted by both pipelines have some quantification differences, especially for complex events (red dots)





### Comparison to other methods





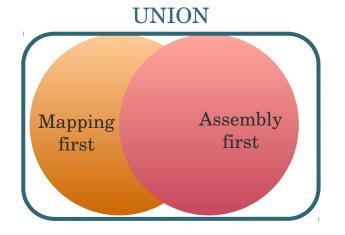


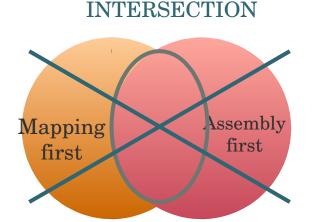
# **Conclusion & Perspectives**

Annotating alternative splicing with a single approach leads to missing a large number of candidates.

These candidates cannot be neglected, since many of them are differentially regulated across conditions.

We advocate for the use of a combination of both mappingfirst and assembly-first approaches for annotation and differential analysis of alternative splicing from RNA-seq data.









# Software availability

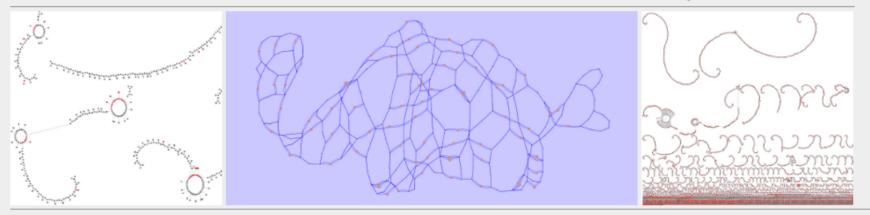
http://kissplice.prabi.fr

#### **KisSplice**

A local transcriptome assembler for SNPs, indels and AS events

HOME PUBLICATIONS DOWNLOAD TOOLS GALAXY

DOCUMENTATION TRAINING CONTRIBUTORS CONTACT FAQ



#### **Latest News**

- 2017-05-12: **kissDE version 1.5.0** Release
- 2017-02-28: Our <u>AMB paper</u> is out.
- 2016-07-31: Our NAR paper is out. Full protocol to reproduce the results is available here





### Recent paralogs

#### **RASA4 and RASA4B**





Missed by FaRLine

RASA4 and RASA4B are recent paralogs

Multi-mapping reads are discarded by mapping-first approaches

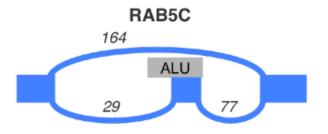
KisSplice co-assembles the two paralogs, and states that they collectively produce two transcripts

Confirmed experimentally by RT-PCR





### Exons overlapping repeats





Missed by KisSplice RAB5C contains an exonised Alu Since this exon is annotated, FaRLine finds it KisSplice fails to assemble it, because the bubble has more than 5

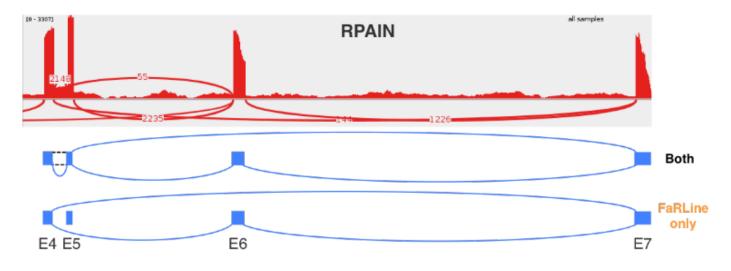
branches

(i.e. too many Alu copies in the dataset)
Confirmed experimentally by RT-PCR





### Complex events



Missed by KisSplice

The skipping of E6 with E4 and E7 as flanking exons is reported only by FaRLine

KisSplice discards E4-E6 junction because it is supported only by 55 reads, which is less then 2 % of the read flow leaving E4