

Annotation of lncRNAs - FEELnc: FLEXible Extraction of LncRNAs

Thomas DERRIEN, Valentin Wucher
(tderrien@univ-rennes1.fr)

Canine Genetics Group
IGDR : Institute of Genetics and Development of Rennes
CNRS - UMR6290 - Université de Rennes 1

Fabrice Legeai
(fabrice.legeai@inra.fr)

Bioinformatics Platform for Agroecosystems Arthropods
IGEPP : Institute for Genetics, Environment and Plant Protection
INRA

Non-coding RNAs

- **80% of the variants** associated with disease (by GWAS) are localized **outside of protein-coding genes** (Manolio et al., Hindorf et al.)
- **>60% of the human genome** is transcribed into RNAs (~75% by primary transcripts) **with only 2% corresponding to proteins...**
(Human ENCODE Consortium; Djebali et al. 2012, Mouse ENCODE Consortium; 2015)

=> Need to annotate ncRNAs to ease the interpretation of genotype to phenotype relationships



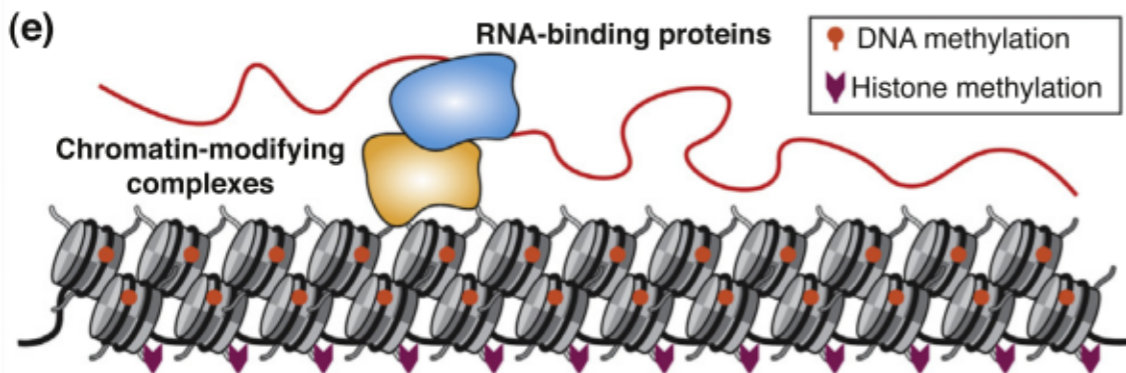
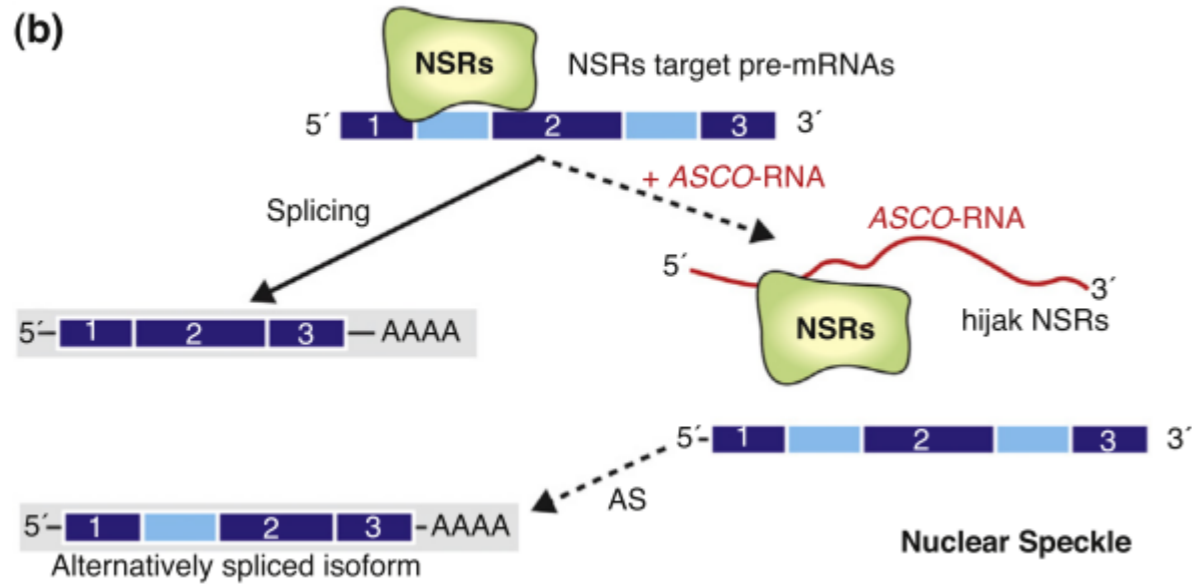
Non-coding RNAs

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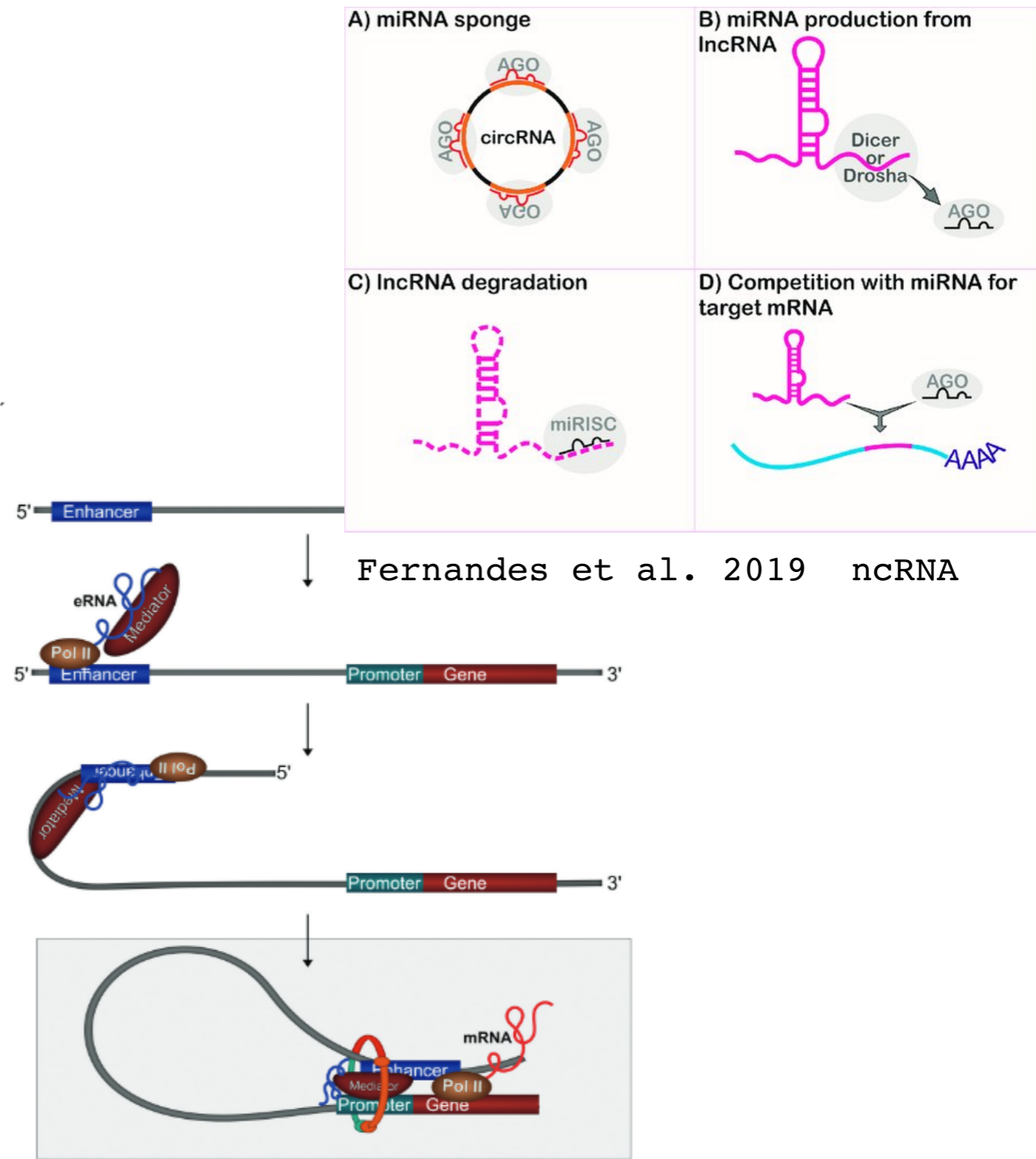
(bio)typeW	functions
mRNAs	many..
miRNAs	Regulation of gene expression
siRNAs	RNA interference pathway
snRNAs	Chemical modification of rRNA, tRNAs and small RNAs
piRNAs	transposon defense - regulate euchromatin formation
snRNA	splicing, regulation of TFs, telomere stability...
long ncRNAs (Xist, H19, Hotair..)	regulation of mRNAs expression, X chromosome inactivation, imprinting.



LncRNAs Functions in plant epigenetics

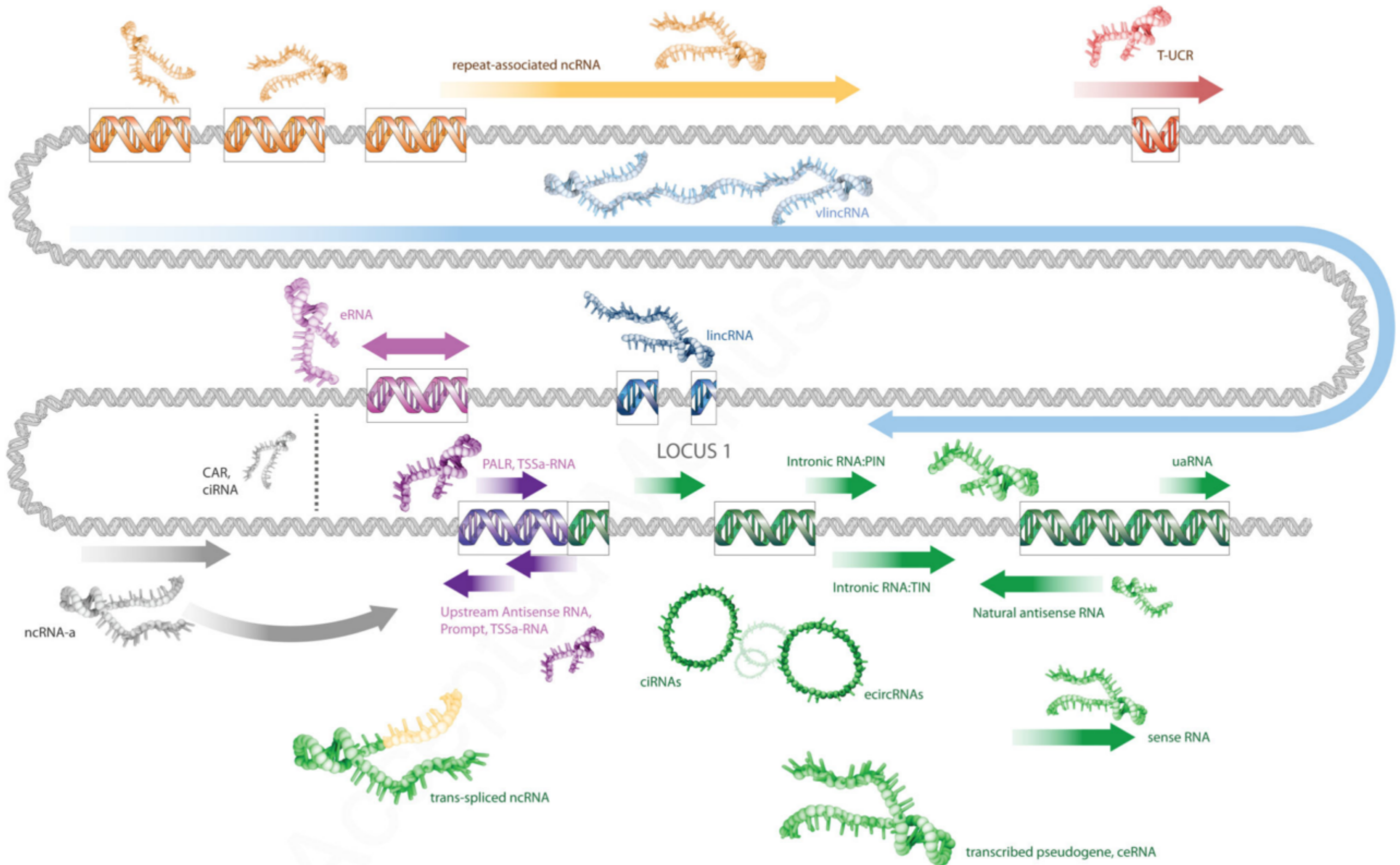


Chekanova 2015
Current Opinion in Plant Biology.



Shibayama et al. 2014 Transcription.

The several classes of lncRNAs



lncRNAs versus mRNAs

lncRNAs are not mRNAs...

- No coding potential => small ORFs
- Less conserved (20% bw Human-Mouse)
- Lowly expressed and more tissue-specific
- Overlap many Transposable Elements (TEs)

...but are "mRNA-like" transcripts:

- Transcribed by Pol_{II}
- Spliced
- Capped in 5'
- Most of them contain a polyA tail



**Transcripts without coding potential ,
>200 nt, spliced, polyA+/- (Derrien et
al., 2012)**

Structural definition of lncRNAs



First Annotation in human : e.g GENCODE reference annotation

(Harrow et al., 2012, 1000 genomes project)

LncRNAs annotation has been greatly increased by the use of whole transcriptome sequencing (RNA-Seq)



Human

Statistics about the current GENCODE Release (version 30)

The statistics derive from the [gtf file](#) that contains only the annotation of the main chromosomes.

For details about the calculation of these statistics please see the [README_stats.txt file](#).

General stats

Total No of Genes	58870	Total No of Transcripts	208621
Protein-coding genes	19986	Protein-coding transcripts	83688
Long non-coding RNA genes	16193	- full length protein-coding	57687
Small non-coding RNA genes	7576	- partial length protein-coding	26001
Pseudogenes	14706	Nonsense mediated decay transcripts	15550
- processed pseudogenes	10663	Long non-coding RNA loci transcripts	30369
- unprocessed pseudogenes	3525		
- unitary pseudogenes	221		
- polymorphic pseudogenes	42	Total No of distinct translations	61870
- pseudogenes	18	Genes that have more than one distinct translations	13709
Immunoglobulin/T-cell receptor gene segments			
- protein coding segments	408		
- pseudogenes	237		

<http://www.genencodegenes.org/human/stats.html>

Structural definition of lncRNAs

GreNC

A Wiki-database of plant lncRNAs (v1.12)



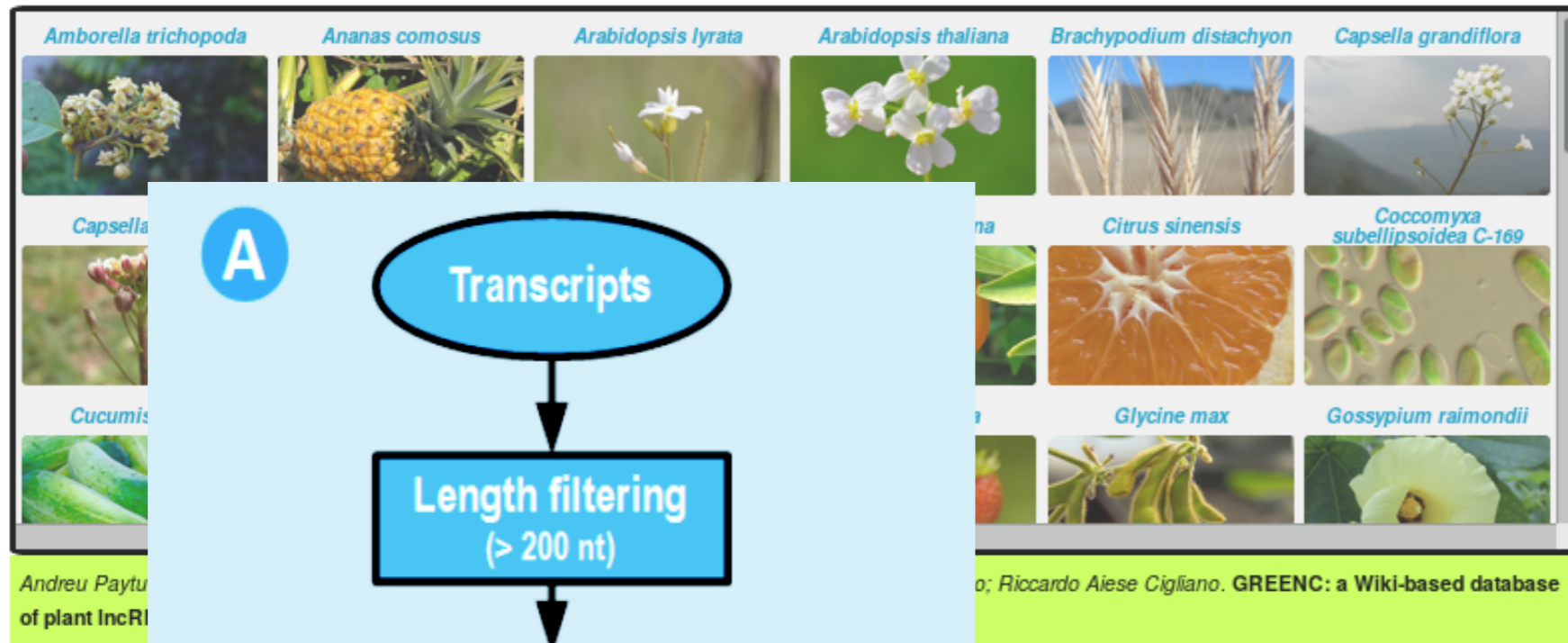
Statistics

Species	Assembly	Gene number	lncRNAs	High confidence	Low confidence	Repetitive elements	miRNA precursors
<i>Triticum_aestivum</i>	v2.2	23359	38820	21132	17688	5714	1039
<i>Zea_mays</i>	6a	16857	18110	10229	7881	4004	2192
<i>Physcomitrella_patens</i>	v3.3	6888	9690	8390	1300	907	299
<i>Medicago_truncatula</i>	Mt4.0v1	9373	9676	5793	3883	2567	286
<i>Glycine_max</i>	Wm82.a2.v1	5974	6689	4749	1940	832	133
<i>Solanum_tuberosum</i>	v3.4	5974	6680	2976	3704	2682	184
<i>Amborella_trichopoda</i>	v1.0	5698	5698	4156	1542	103	347
<i>Brachypodium_distachyon</i>	v3.1	4828	5584	3648	1936	870	1024
<i>Populus_trichocarpa</i>	v3.0	4997	5569	4111	1458	434	124
<i>Sorghum_bicolor</i>	v3.1	4624	5305	2682	2623	1737	1057
<i>Oryza_sativa_Japonica_Group</i>	v7.0	4995	5237	3601	1636	1148	119

Structural definition of lncRNAs

GreenNC

A Wiki-database of plant lncRNAs (v1.12)



Coding Potential Calculation

CPC : SVM based on 6 parameters

3 criteria based on the quality of the largest ORF (size, coverage, ATG)

3 criteria based on a result of a comparison to a protein databases (NR, Swissprot, ...) (number of hits, mean of the e-value, frameshifts)

FEELnc : FLEXible Extraction of LncRNAs

(<https://github.com/tderrien/FEELnc>)

**Assembled
transcripts**
(cufflinks/stringtie)

I- FEELnc_Filter

II- FEELnc_CodingPot

III- FEELnc_Classifier

LncRNAs

Standard pipeline for RNA-Seq analysis (genome-guided)

Input files :

- Reference genome
- Reference annotation

RNASeq_file (.fastq)

fastqc + trimmomatic **Cleaning**

Cleaned sequences (.fastq)

HISAT2/STAR/tophat2 **Mapping**

Mapped files (.bam)

Cufflinks/
Stringtie **Transcriptome
reconstruction**

(Many!) assembled transcripts models (.gtf)

I- Filtering transcripts

Assembled transcript models (.gtf)

Known and novel transcripts

I- FEELnc_Filters

```
* Mandatory arguments:
-i,--infile=file.gtf
-a,--mRNAfile=file.gtf

* Filtering arguments:
-s,--size=200
-b,--biotype
-l,--linonly
--monoex=-1|0|1
--biex=25

* Overlapping specification:
-f,--minfrac_over=0
-p,--proc=4
```

Candidate lncRNAs

Filtering out non-lncRNAs

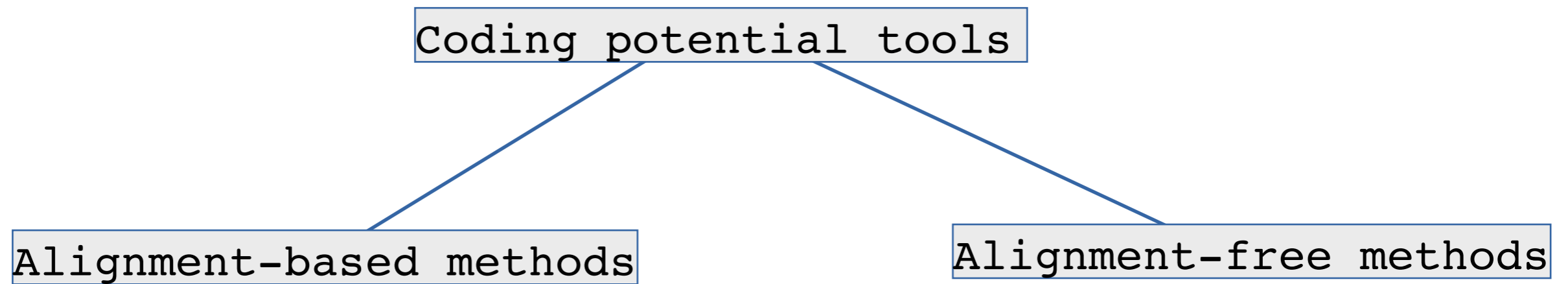
- every biotypes (small ncRNA, pseudogenes...) ?
- warning => removing potential lncRNAs host gene for small ncRNAs...
- only protein-coding biotype (=> considered as alternative isoform of mRNAs)
- < 200bp
- Monoexonic

Options :

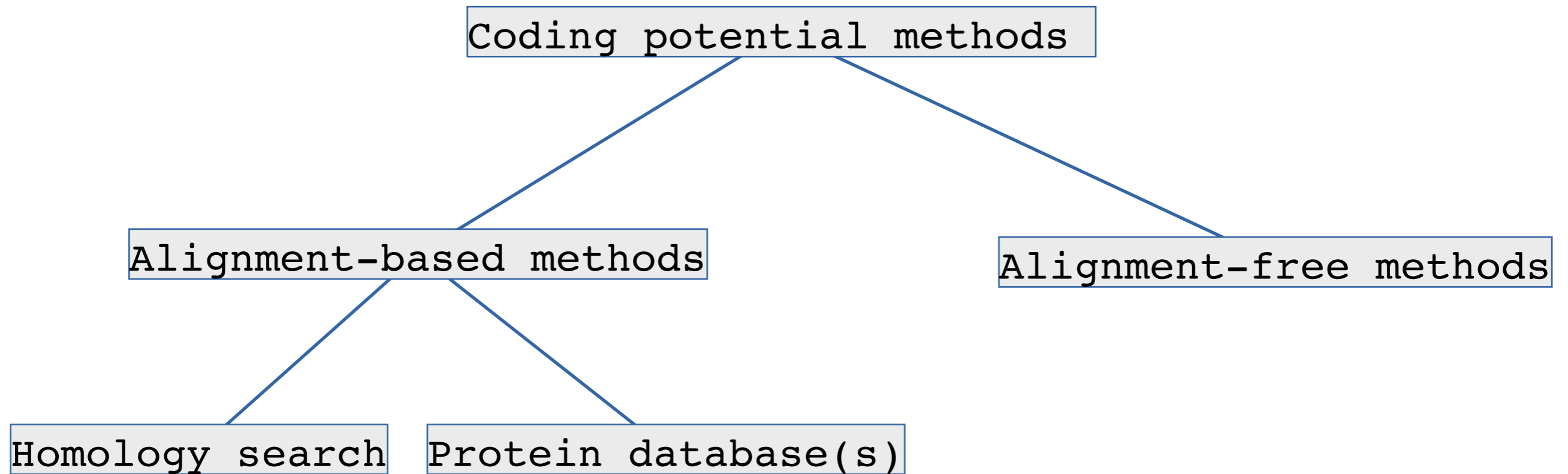
- Fraction of overlap
- Multithreaded

II- Computing Coding Potential

Aim : define a coding potential score (CPS) for candidate transcripts and then a cutoff/threshold to differentiate mRNAs from lncRNAs

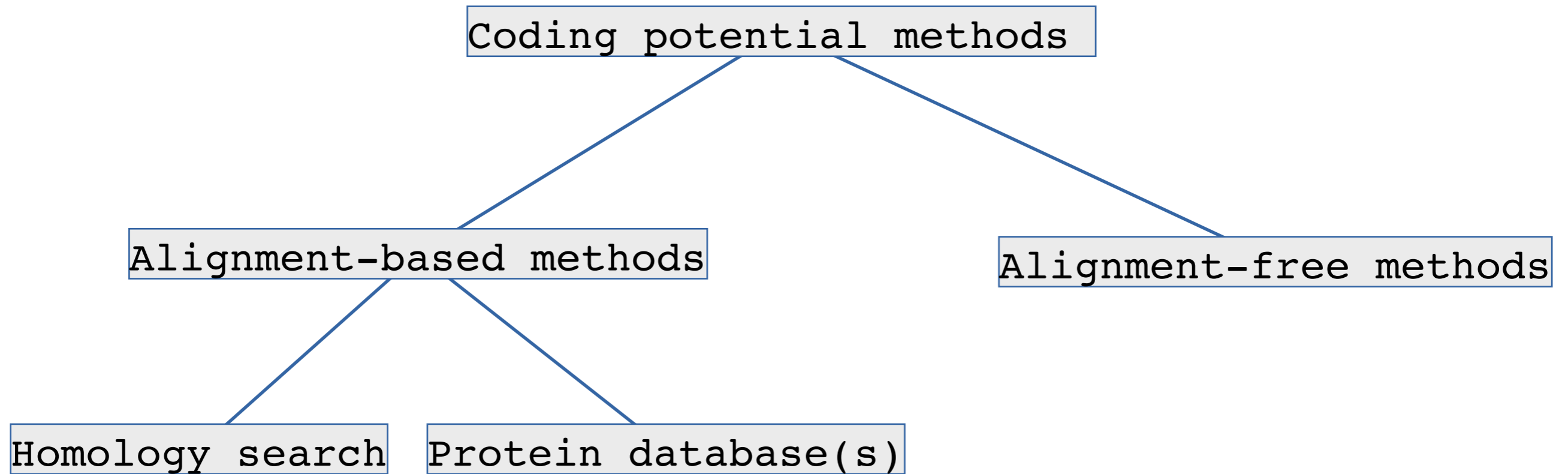


II- Computing Coding Potential



- **Phylo-CSF** (Lin, M. *et al.* 2011). **CPC** (Kong, L. *et al.* 2007).
- **RNACode** (Washietl, S. *et al.* 2011).

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- Advantages:

- high specificity
- introduce the notion of conserved lncRNAs

- Drawbacks:

- depends on which database/species to align with...
- quality of the alignments
- slow running time

II- Computing Coding Potential

Coding potential methods

Alignment-based methods

Homology search

Protein database(s)

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Alignment-free methods

- **CPAT** (Wang *et al.* 2013).
- **CNCI** (Sun *et al.* 2013).
- **PLEK** (Li *et al.* 2014).
- **LncRNA-ID** (Achawanantakun *et al.* 2015).
- ...
- **COME** (Hu *et al.* 2016).
- **lncScore** (Zhao *et al.* 2016).

- Advantages:

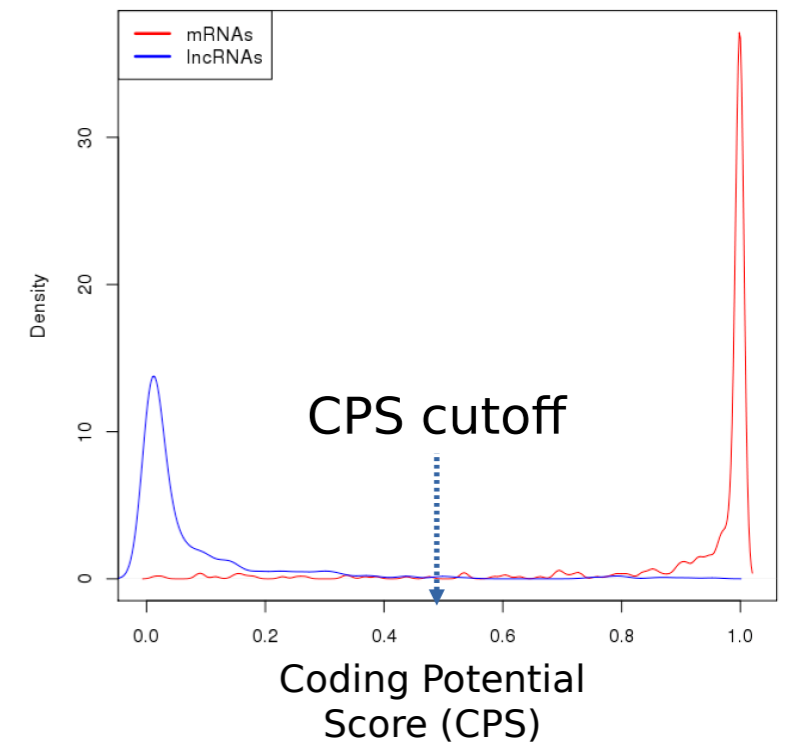
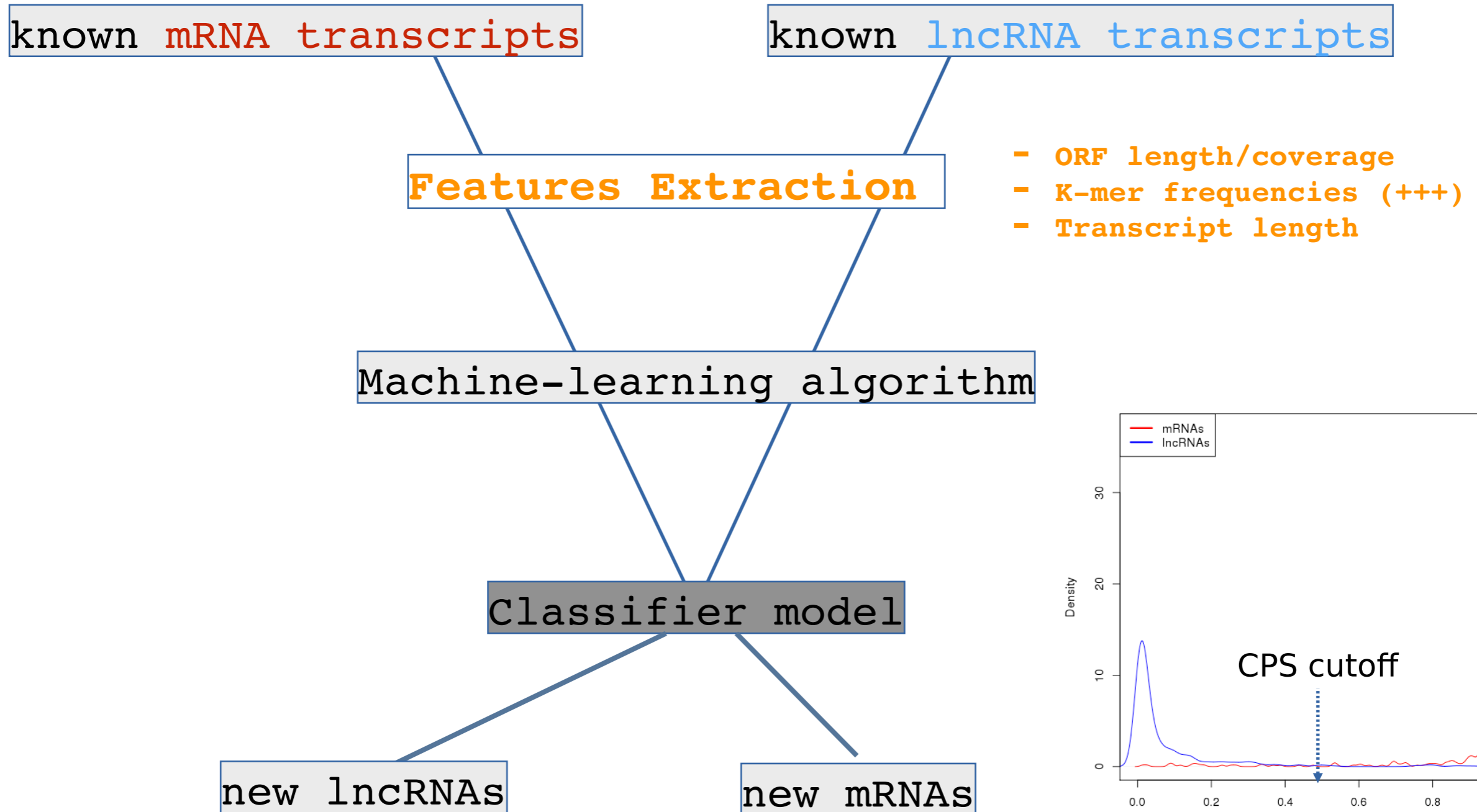
- fast and (relatively) easy to use
- independent of any alignment
- lineage-specific lncRNAs (or weakly conserved)

- Drawbacks:

- +++ designed for model organisms

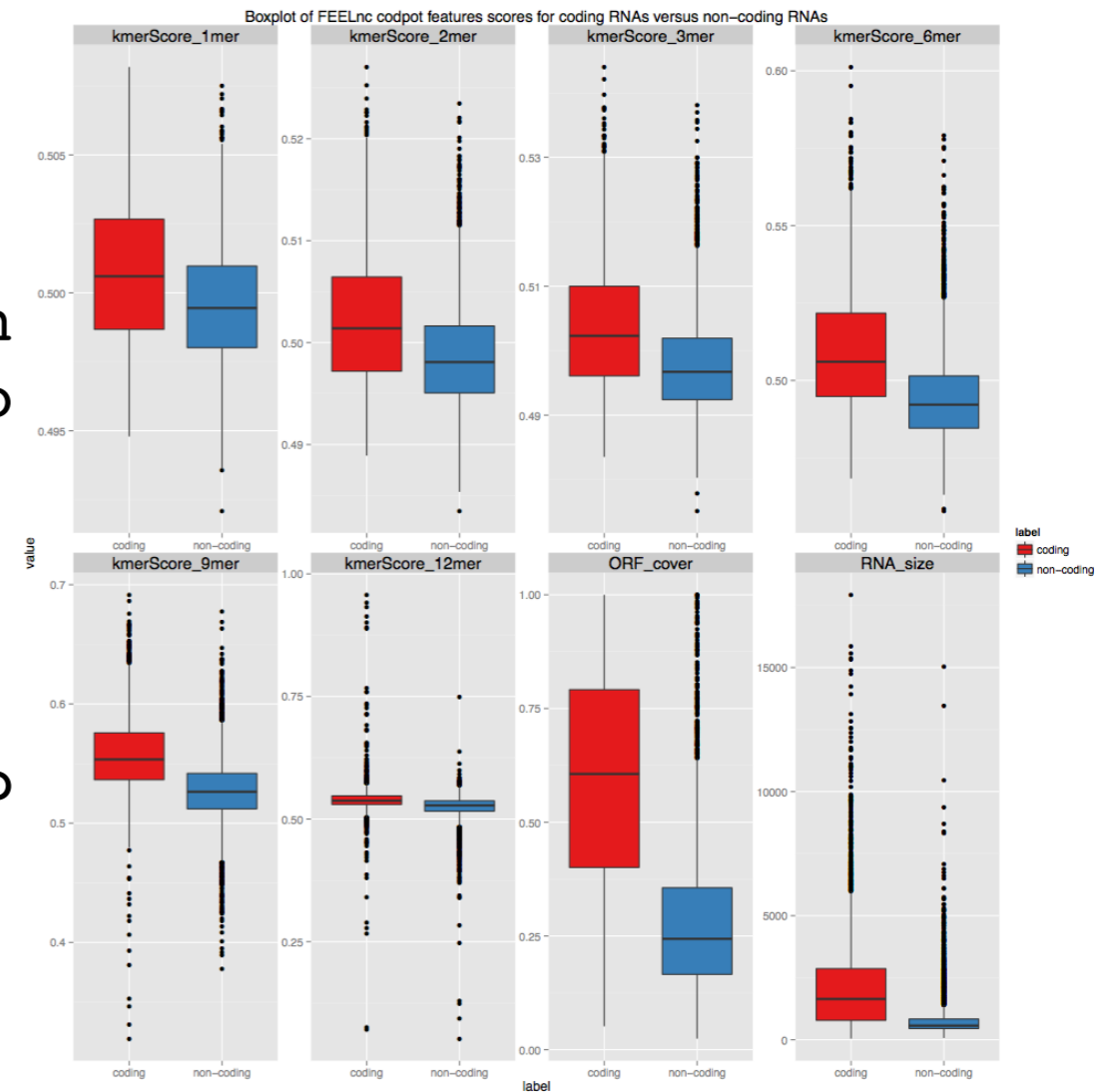
II- Computing Coding Potential

Extracting Features and Machine learning



3 main classifiers/features

1. Multi- k -mer scores ($k= 1$ to 12)
 - **KmerInShort (KIS)** developed from GATB tools (Drezen et al.) (<https://github.com/rizkg/KmerInShort>)
 - very fast and parallel extraction of k -mer profiles (not limited to one k -mer i. hexamer)
2. ORF coverage (ORF defined wrt 3 modes)
 - strict : requires start && stop
 - moderate : requires start || stop
 - relaxed : RNA sequence
3. RNA size



II- Computing Coding Potential

Feature scores on known lncRNAs and mRNAs



RandomForest

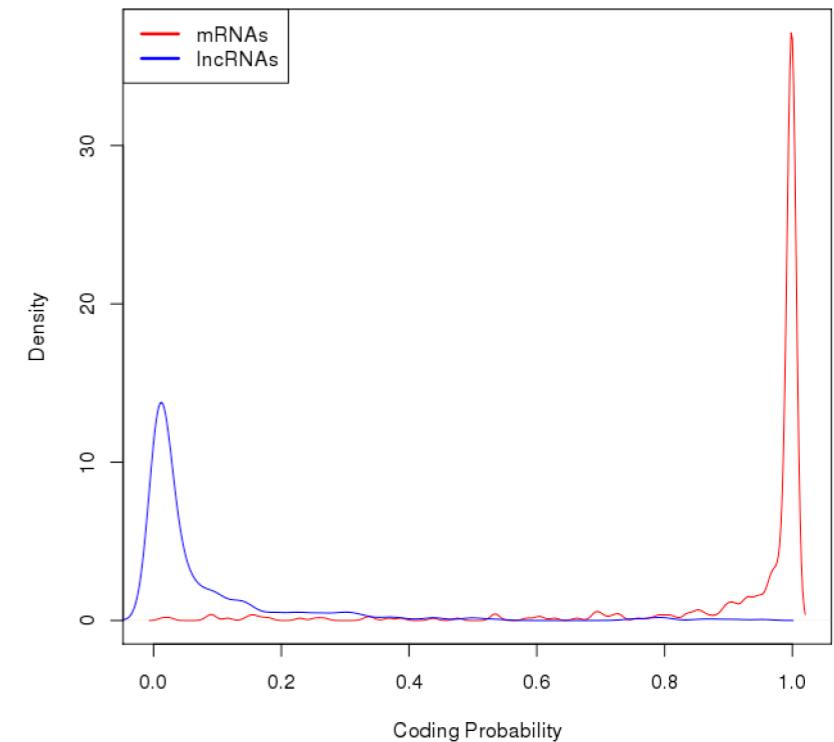
- fast and easy to optimize
- can deal with unbalanced training set (+++)
- could deal with missing data



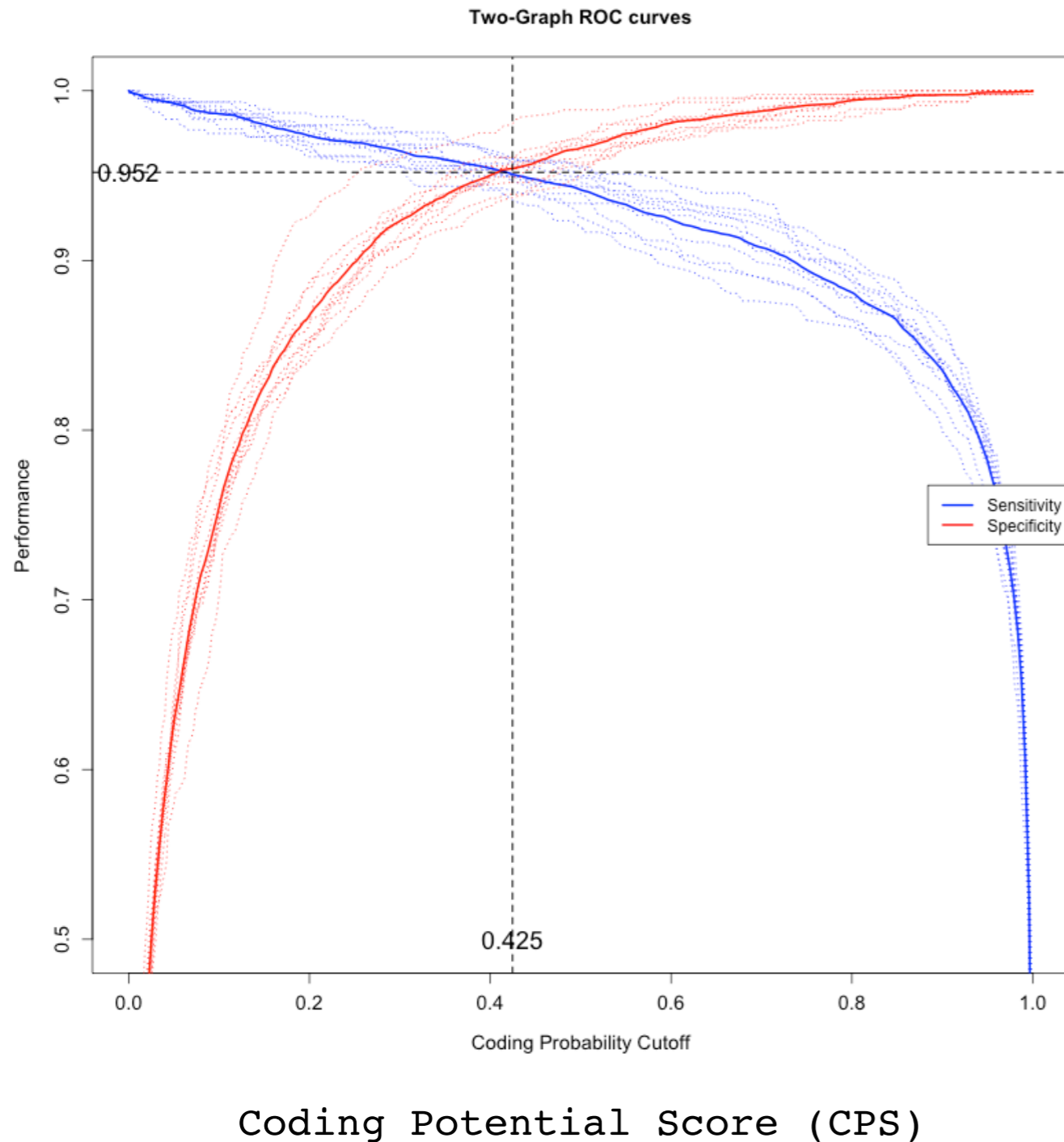
Defines a coding potential score (CPS) for all transcripts from training sets



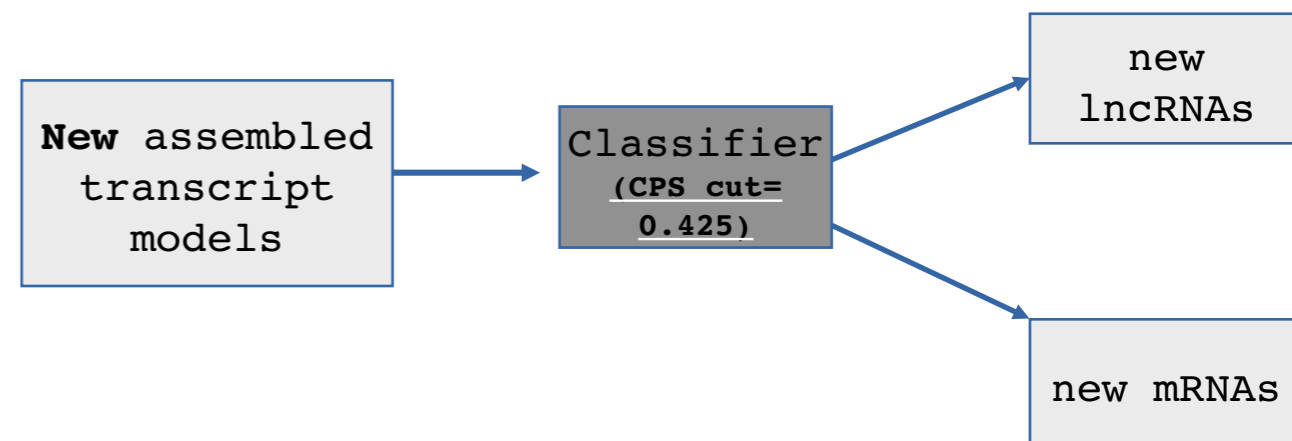
What is the best CPS cutoff?



Optimal CPS threshold

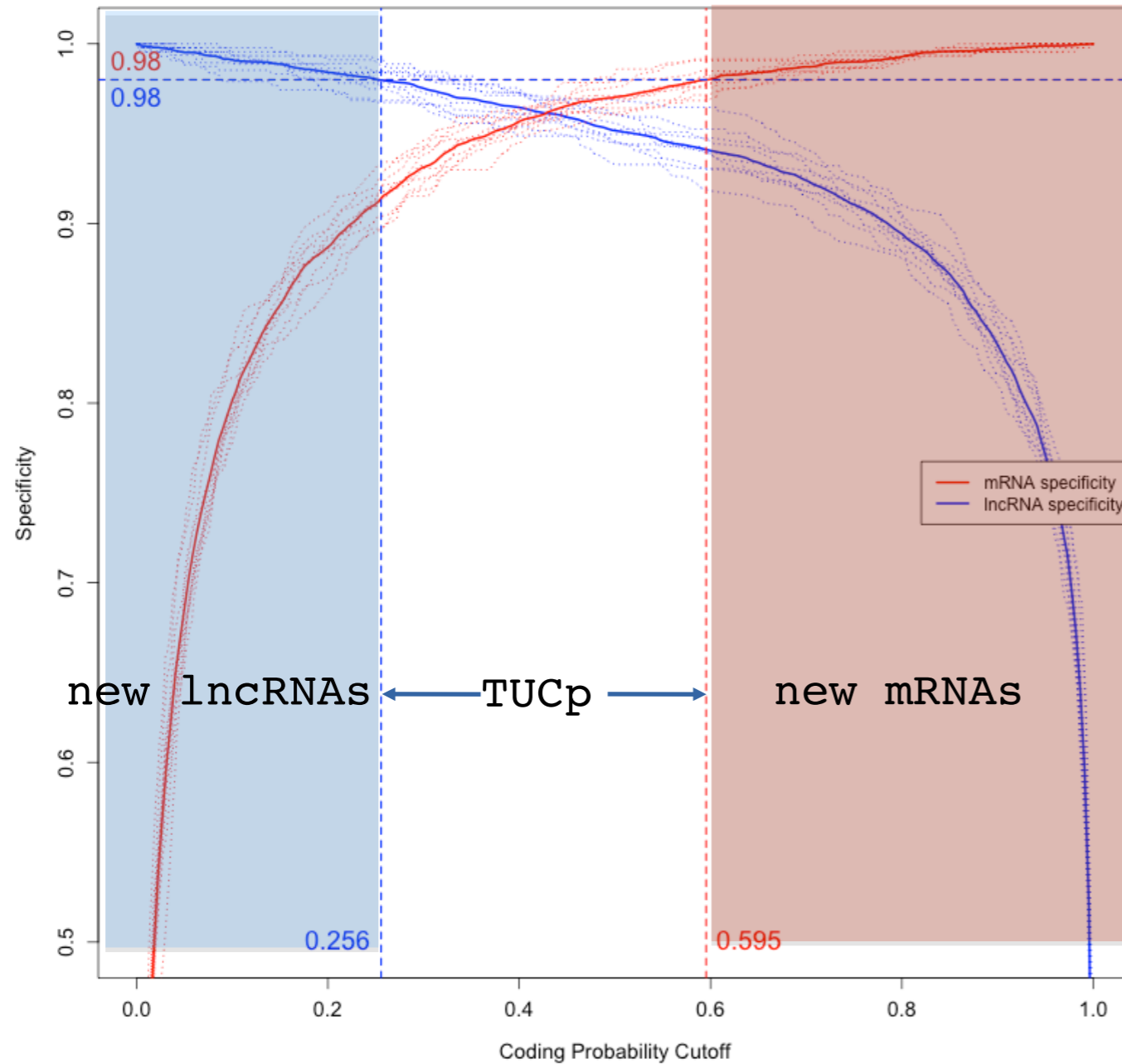


- Automatic Two-Graph ROC curves to define an optimal cutoff to separate mRNAs from lncRNAs
- FEELnc defined optimal CPS cutoff via :
 - Sensitivity == Specificity (= 0.952)
 - OR
 - Max (Sn & Sp)



Optimal CPS threshold

Two-Graph ROC curves

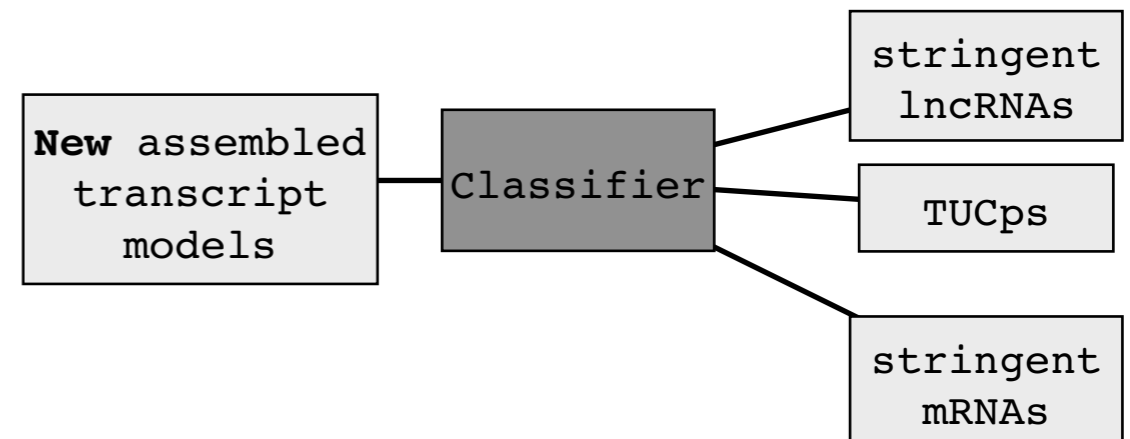


Coding Potential Score (CPS)

"The (CPS) threshold is (...) somewhat arbitrary, and transcripts that reside in questionable regions of the distribution should be annotated as transcripts of unknown coding potential (TUCPs)"

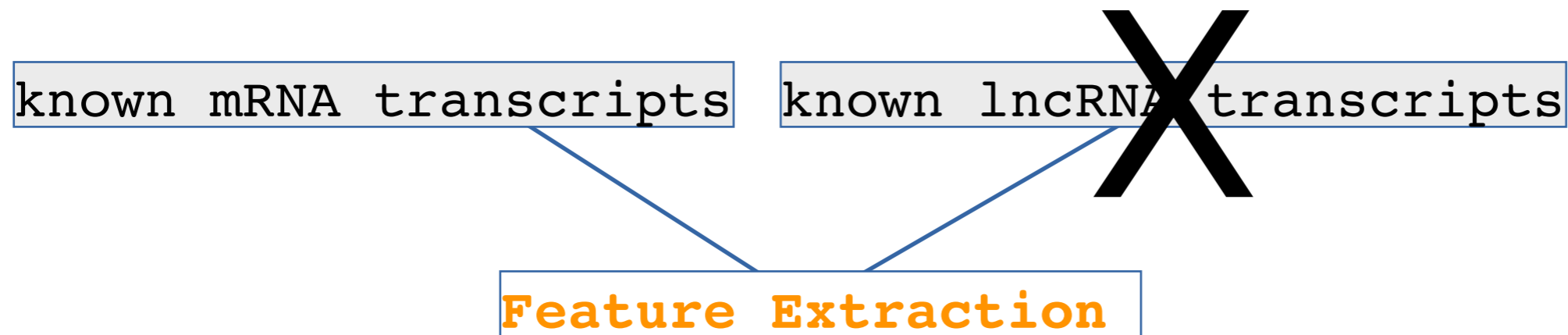
J.S. Mattick, J.L. Rinn, Discovery and annotation of long noncoding RNAs. *Nature Structural Molecular Biology*, 22:5–7, 2015.

- TUCPs = ambiguous transcript given their CPS



FEEInc : without lncRNA training set

Training



For non-model organisms, known lncRNAs are (often) not available. FEEInc implements 2 options to model ncRNAs:

- **intergenic module** : extract random intergenic sequences
- **mRNA shuffle module** : shuffle mRNA sequences while preserving a certain *k*-mer frequency (using UShuffle program)

II- Computing Coding Potential

Tools performance on the GENCODE human datasets.

F-score/MCC capture the global performance of the tools in a single measure

HUMAN dataset	Program	Sensitivity	Specificity	Precision	Accuracy	F-score	MCC
CPC	FEELnc	<u>0.923</u>	0.915	0.916	<u>0.919</u>	<u>0.919</u>	<u>0.838</u>
	CPAT	0.899	0.924	0.922	0.912	0.910	0.823
	CPAT_train	0.920	0.901	0.903	0.910	0.911	0.821
	CNCI	0.829	0.979	0.975	0.904	0.896	0.817
	PLEK	0.732	<u>0.985</u>	<u>0.981</u>	0.858	0.838	0.741
	PhyloCSF	0.906	0.802	0.820	0.854	0.861	0.712
	PLEK_train	0.582	0.960	0.936	0.770	0.718	0.584
		0.728	0.719	0.713	0.438		

Bold-underlined values correspond to the highest values of each metrics.

CPAT_train and PLEK_train correspond to program versions trained with the human training dataset.

Programs are sorted by MCC values.

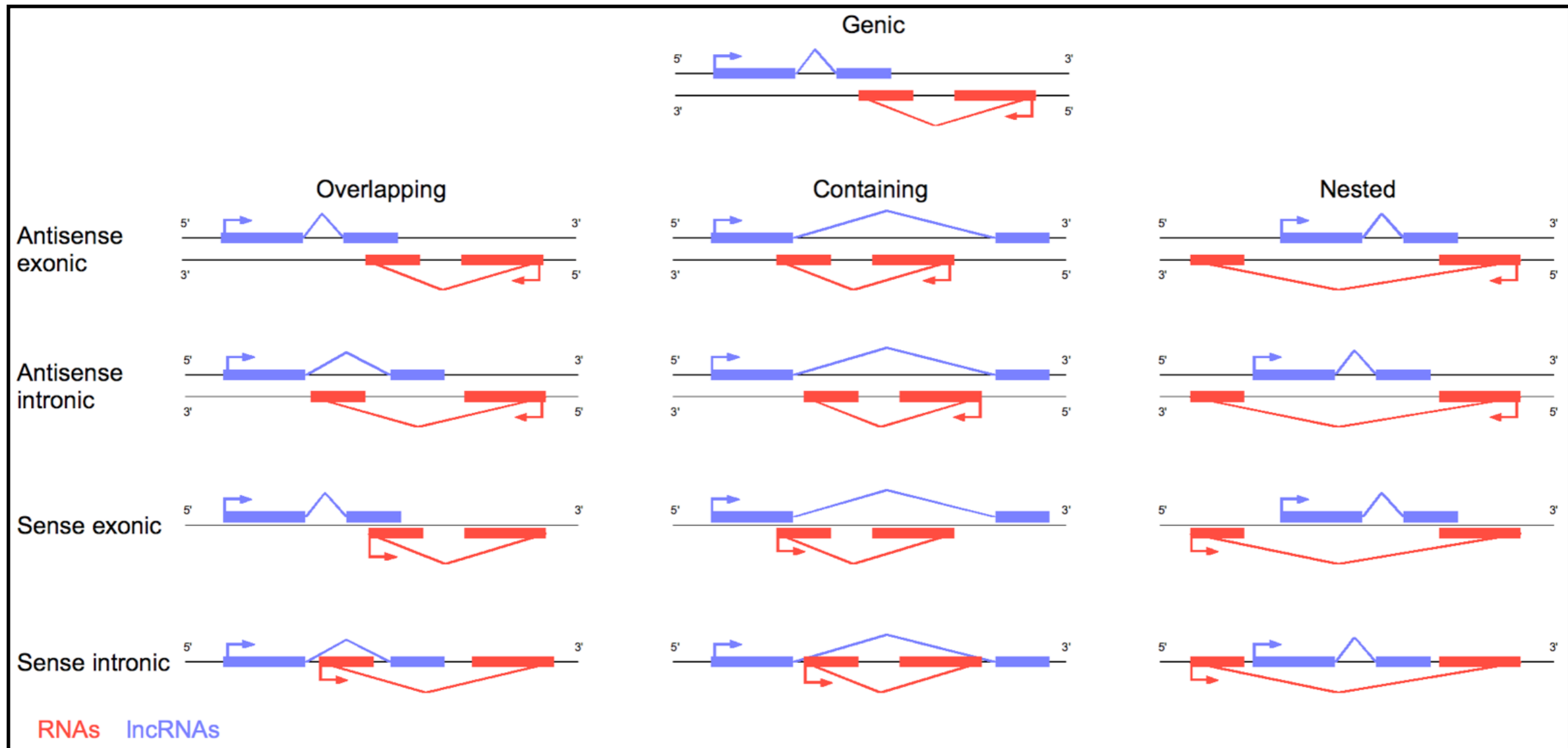
III- Classifying

- Classifying lncRNAs genomic context wrt to mRNAs could help predict functionality
- FEELnc_Classifier uses a sliding window around lncRNA
- Classify with closest RNA_partner (mRNAs or ncRNAs) according to
 - orientation of transcription
 - location of the interaction

- FEELnc classifier output file:

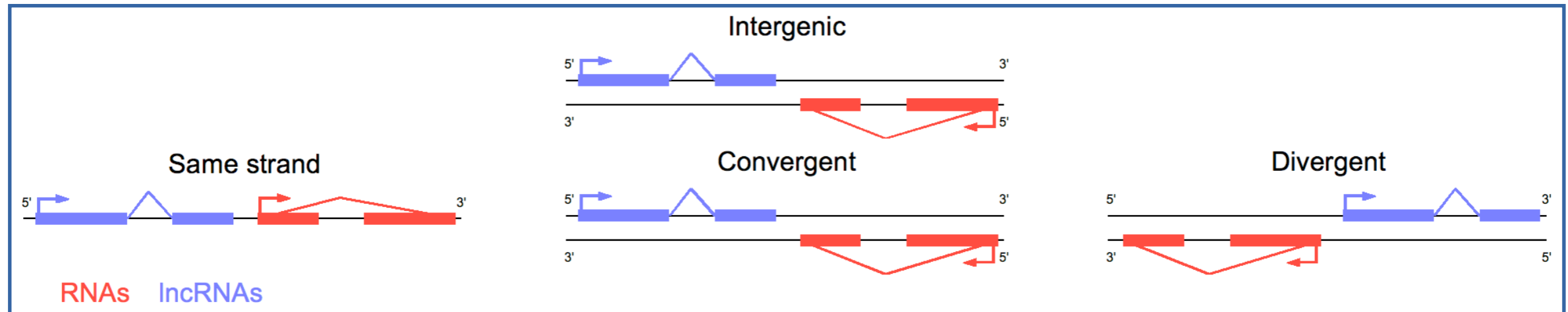
```
cat {INPUT}_classes.txt
isBest lncRNA_gene lncRNA_transcript partnerRNA_gene partnerRNA_transcript direction type distance subtype location
1 XLOC_090743 TCONS_00232056 ENSCAFG00000013346 ENSCAFT00000021186 antisense intergenic 377 divergent upstream
1 XLOC_090720 TCONS_00231943 ENSCAFG00000026373 ENSCAFT00000040656 sense intergenic 66670 same_strand upstream
1 XLOC_090678 TCONS_00231794 ENSCAFG00000010781 ENSCAFT00000017151 antisense genic 0 nested intronic
0 XLOC_090678 TCONS_00231794 ENSCAFG00000010794 ENSCAFT00000017171 sense intergenic 8293 same_strand upstream
```


III- Classifying



Such as gencode classes (i.e sense_intronic, antisense...)

III- Classifying



dubious lincRNAs
(i.e UTRs of mRNAs)

lincRNAs sharing bi-
directional promoter
with mRNAs

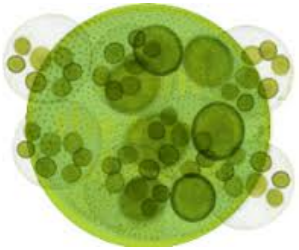
Applications



- Various tissues (n=16 in 7 different breeds)
- ~2,500 new lncRNAs loci (lincRNAs)
- Wucher V. *et al*, NAR 2017



- Adipose and liver tissues
- 2,200 novel lncRNA genes
- Muret K. *et al*, Genet Sel Evol 2017



- male and female gametophytes
- 717 novel lncRNA genes
- Cormier A. *et al*, New Phytol 2017

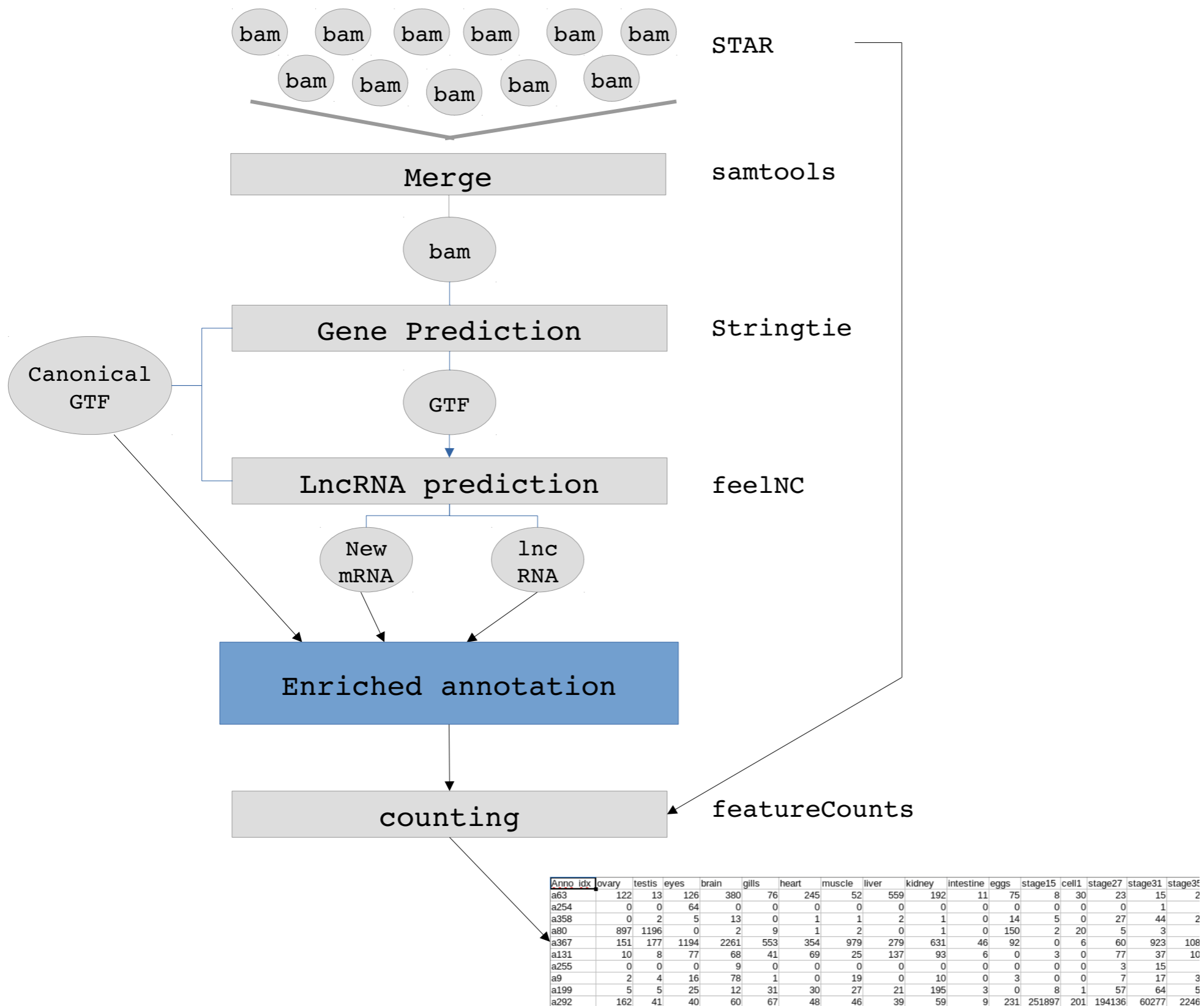


- Sexual and asexual embryos ; salivary glands of plant host biotype
- 2625 novel lncRNA genes
- In preparation

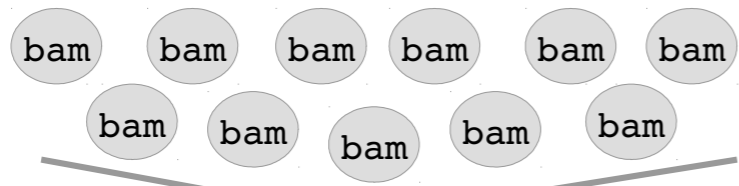


- Ovogenesis medaka
- 1131 novel lncRNA genes
- In preparation

Applications



Applications



STAR

Merge

samtools

bam

Gene Prediction

Stringtie

GTF

LncRNA prediction

feelNC

New mRNA

lnc RNA

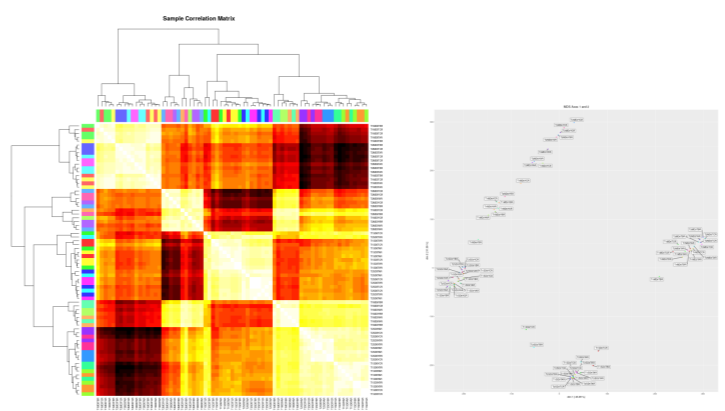
Enriched annotation

Canonical GTF

counting

featureCounts

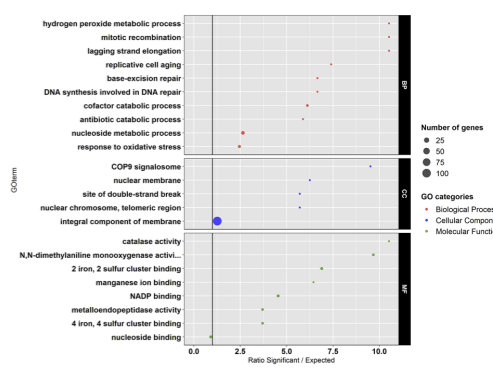
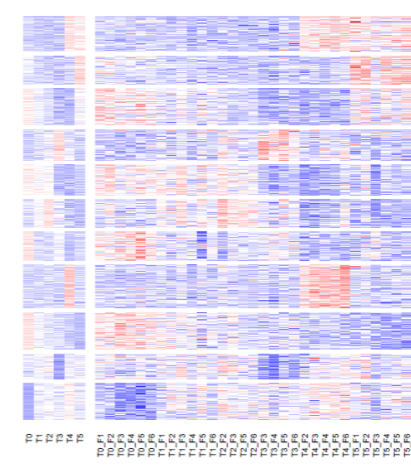
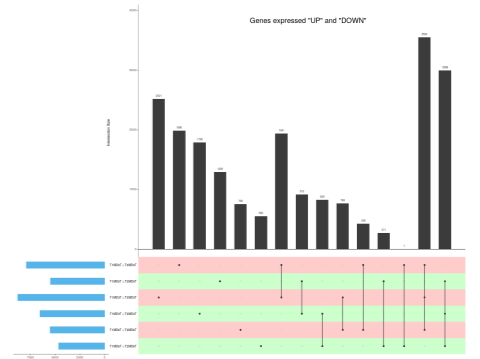
Anno_idx	ovary	testis	eyes	brain	gills	heart	muscle	liver	kidney	intestine	eggs	stage15	cell1	stage27	stage31	stage3f
a63	122	13	126	380	76	245	52	559	192	11	75	8	30	23	15	2
a254	0	0	64	0	0	0	0	0	0	0	0	0	0	0	0	1
a358	0	2	5	13	0	1	1	2	1	0	14	5	0	27	44	2
a80	897	1196	0	2	9	1	2	0	1	0	150	2	20	5	3	
a367	151	177	1194	2261	553	354	979	279	631	46	92	0	6	60	923	108
a131	10	8	77	68	41	69	25	137	93	6	0	3	0	77	37	10
a255	0	0	0	9	0	0	0	0	0	0	0	0	0	3	15	
a9	2	4	16	78	1	0	19	0	10	0	3	0	0	7	17	3
a199	5	5	25	12	31	30	27	21	195	3	0	8	1	57	64	5
a292	162	41	40	60	67	48	46	39	59	9	231	251897	201	194136	60277	2246



Differential expressions analysis

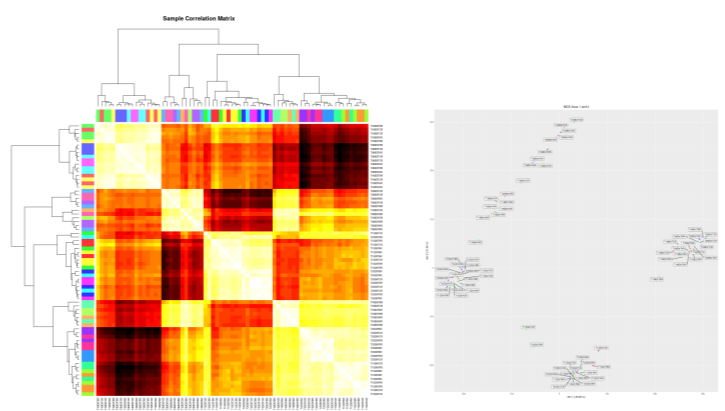
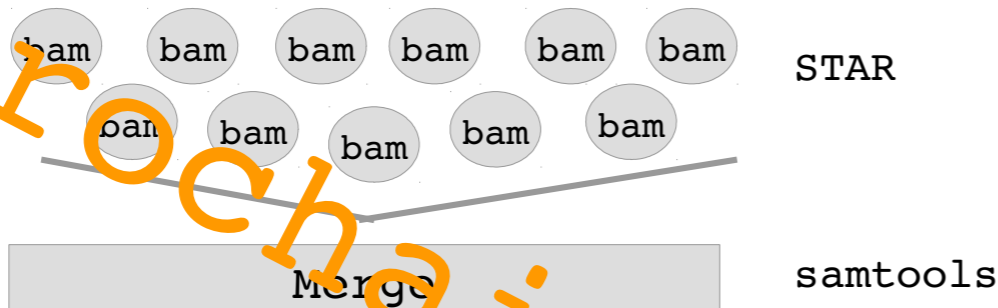
```

T1SD 0T< T1SD 3T T1SD 0T= T1SD 3T T1SD 0T> T1SD 3T
15 64633 189
T1SD 0T< T1SD 6T T1SD 0T= T1SD 6T T1SD 0T> T1SD 6T
1180 63479 178
T1SD 3T< T1SD 6T T1SD 3T= T1SD 6T T1SD 3T> T1SD 6T
1002 63825 10
T2SD 0T< T2SD 3T T2SD 0T= T2SD 3T T2SD 0T> T2SD 3T
440 63225 1172
T2SD 0T< T2SD 6T T2SD 0T= T2SD 6T T2SD 0T> T2SD 6T
1767 59938 3132
T2SD 3T= T2SD 6T
64837
T1SD 0Y= T1SD 3Y
64837
T1SD 0Y< T1SD 6Y T1SD 0Y= T1SD 6Y
5 64832
    
```



Applications

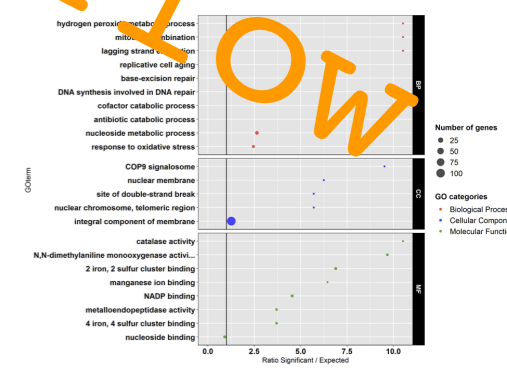
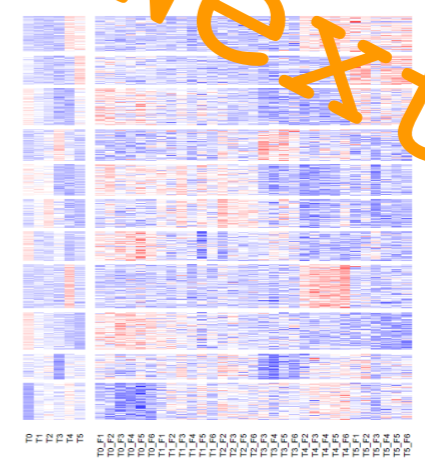
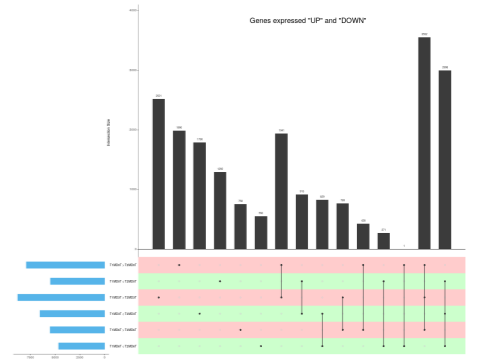
Prochainement



Differential expressions analysis

```

T1SD 0T< T1SD 3T T1SD 0T= T1SD 3T T1SD 0T> T1SD 3T
15 64633 189
T1SD 0T< T1SD 6T T1SD 0T= T1SD 6T T1SD 0T> T1SD 6T
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1767 59938 3132
T2SD 3T= T2SD 6T
64837
T1SD 0Y= T1SD 3Y
64837
T1SD 0Y< T1SD 6Y T1SD 0Y= T1SD 6Y
5 64832
  
```



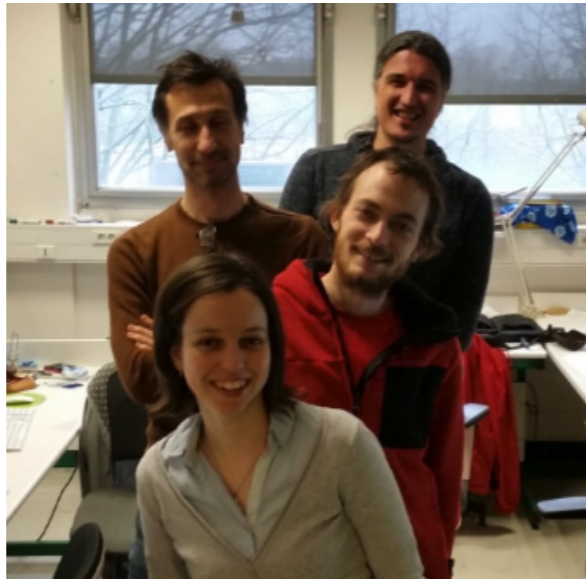
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a254	0	0	64	0	0	0	0	0	0	0	0	0	0	0	0	1
a358	0	2	5	13	0	1	1	2	1	0	14	5	0	27	44	2
a80	897	1196	0	2	9	1	2	0	1	0	150	2	20	5	3	
a367	151	177	1194	2261	553	354	979	279	631	46	92	0	6	60	923	108
a131	10	8	77	68	41	69	25	137	93	6	0	3	0	77	37	10
a255	0	0	0	9	0	0	0	0	0	0	0	0	0	3	15	
a9	2	4	16	78	1	0	19	0	10	0	3	0	0	7	17	3
a199	5	5	25	12	31	30	27	21	195	3	0	8	1	57	64	5
a292	162	41	40	60	67	48	46	39	59	9	231	251897	201	194136	60277	2246

Nextflow

Remerciements



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