



Bioinformatics Summer School

Long-reads *Epi*-transcriptomics

Leda Katopodi

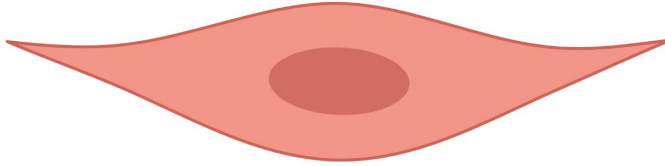
Centre for Genomic Regulation,
Barcelona, Spain

- 1 Introduction to Epitranscriptomics
- 2 Hands-on session: Theory & rationale
- 3 Hands-on session: Analysing Nano-tRNAseq data with AManITA
- 4 Discussion

Epitranscriptomics w/ Long-Read Sequencing Technologies

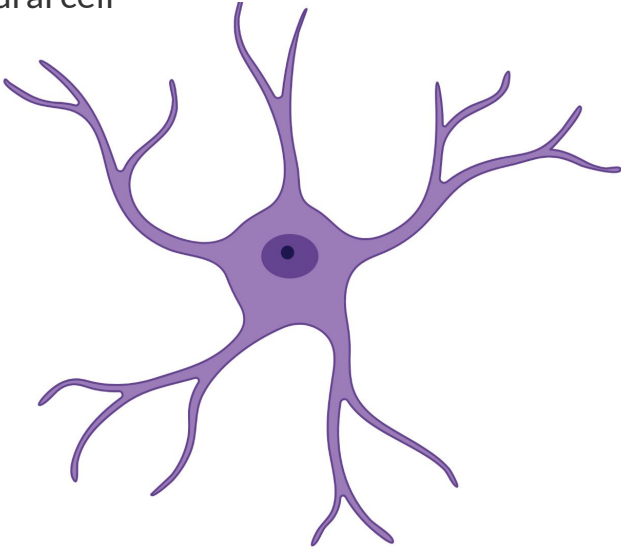
The promise of native RNA sequencing
for the detection of RNA modifications

muscle cell



How can the same DNA
give rise to different cell types?

neural cell

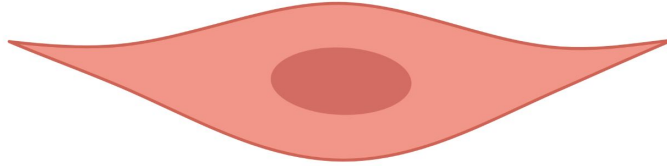


DNA

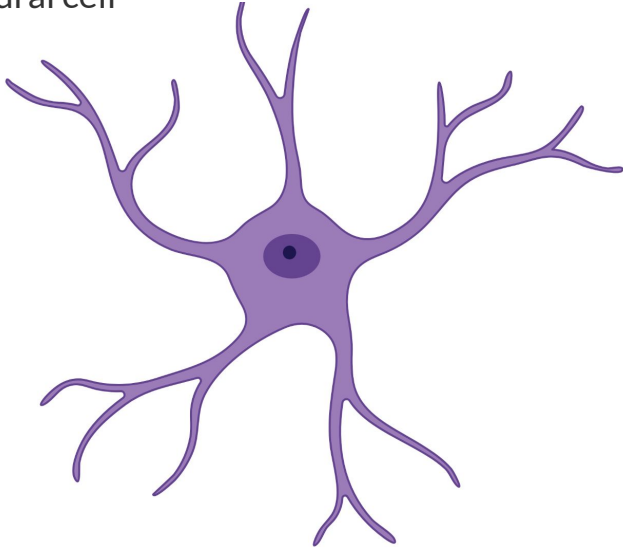
...ATGATCTCGTAA...ACTAGAGCAT...

Why do we care about epitranscriptomics?

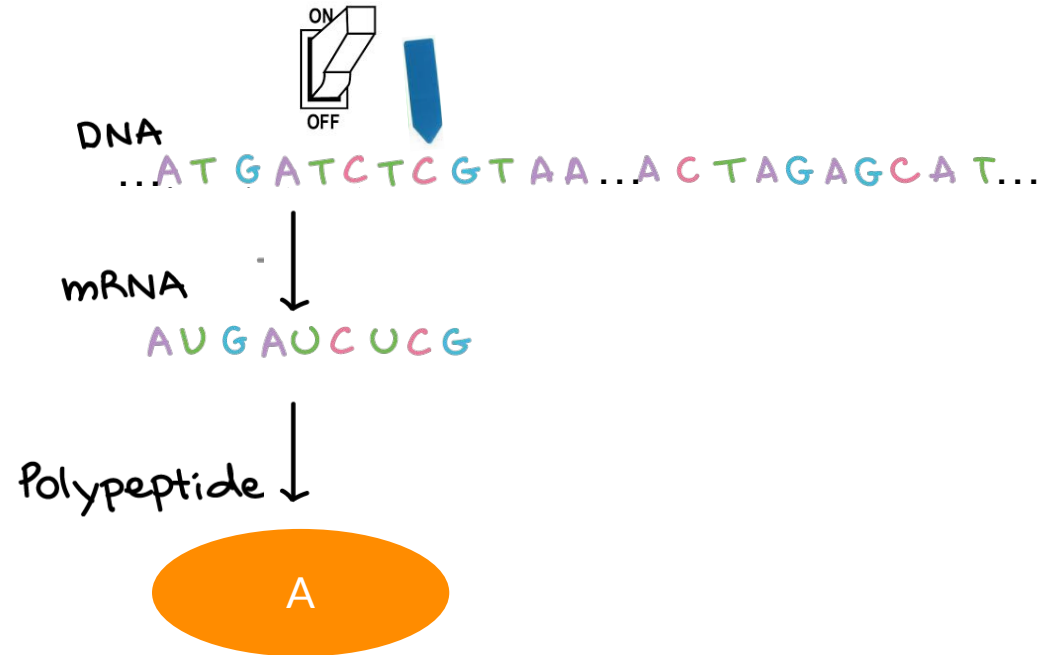
muscle cell



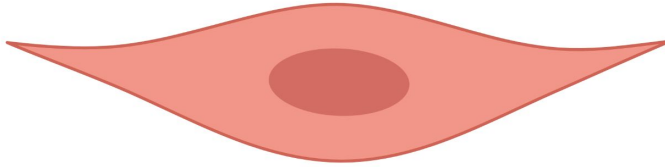
neural cell



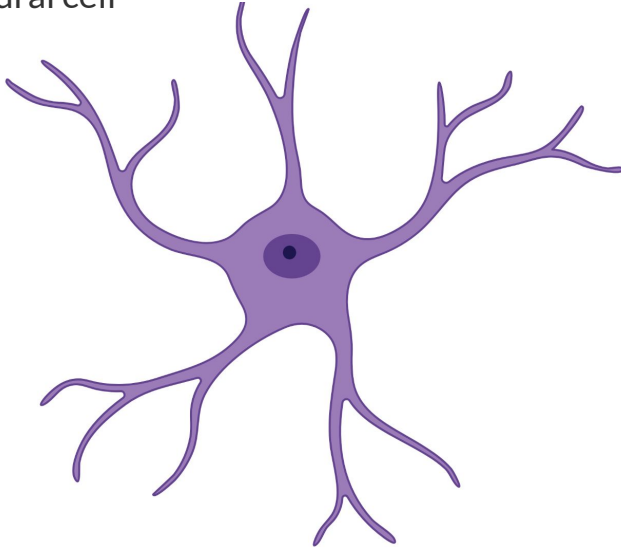
How can the same DNA give rise to different cell types?



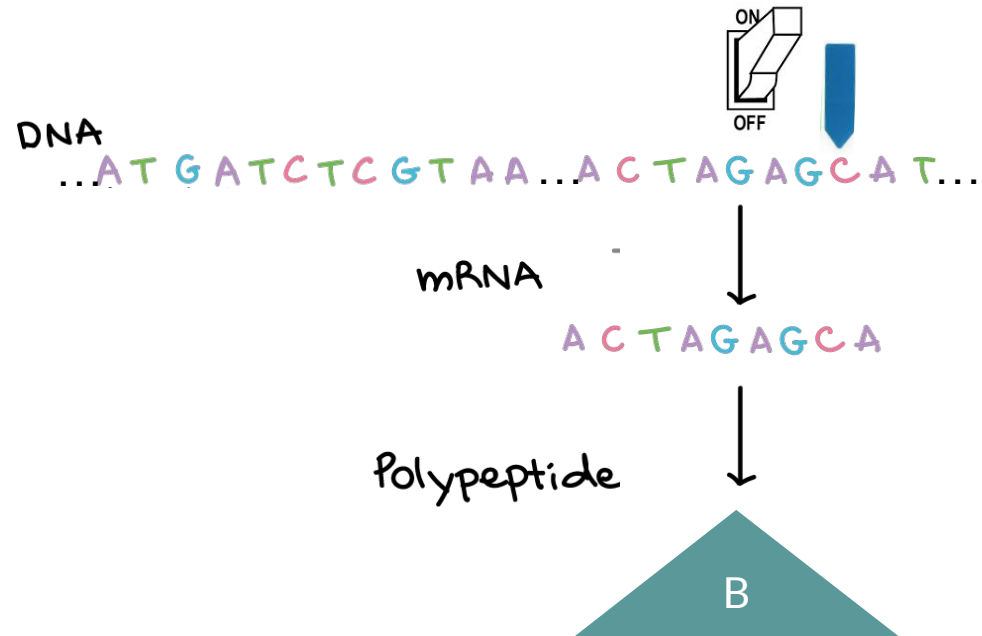
muscle cell



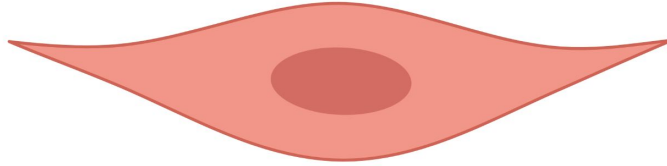
neural cell



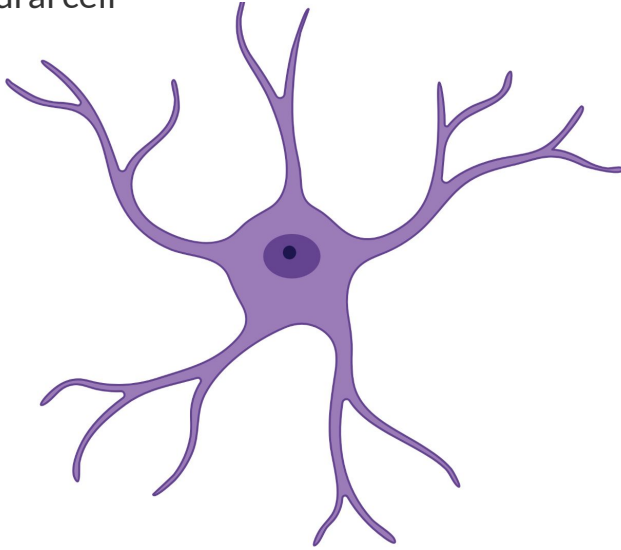
How can the same DNA give rise to different cell types?



muscle cell



neural cell



How can the same DNA give rise to different cell types?

DNA

...ATGATCTCGTAA...ACTAGAGCAT...

the RNA also carries modifications

mRNA

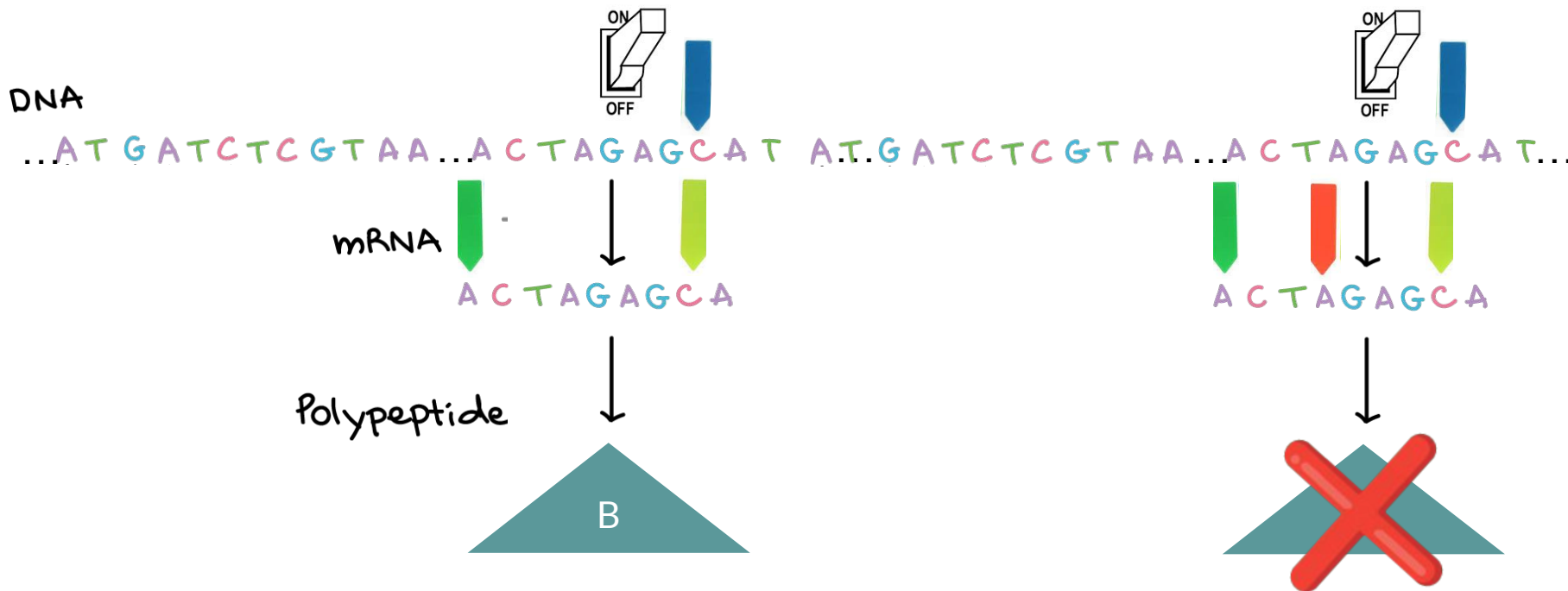
ACTAGAGCA

Polypeptide

B

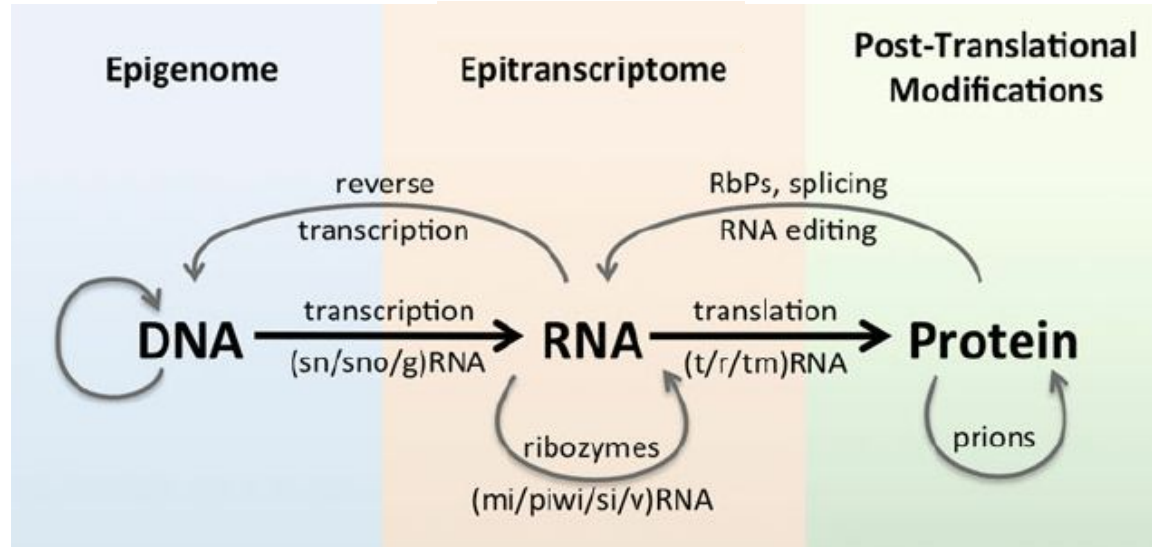


and those RNA modifications may overwrite DNA modifications



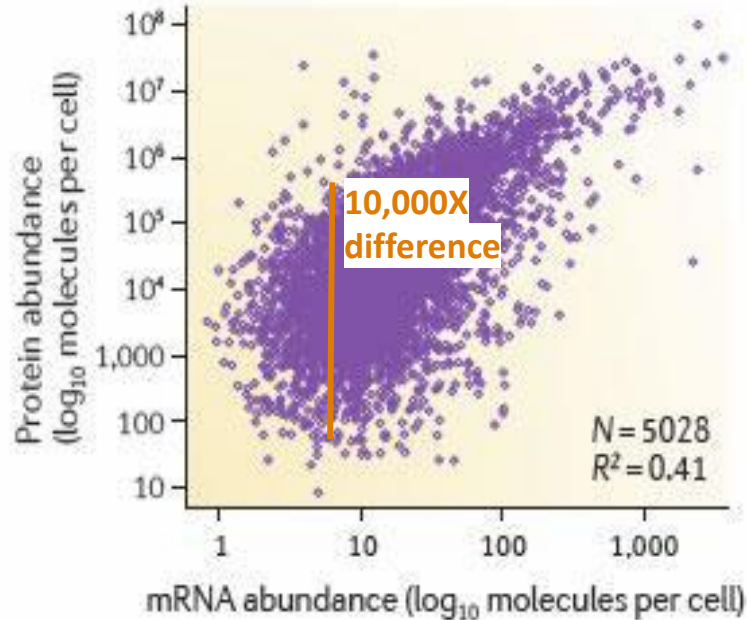


The central dogma of biology is a lie... or just an omission of the actual truth



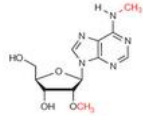
Saletore et al., *Genome Biology* 2012

- Translation efficiency
- tRNA availability
- mRNA half life
- Codon usage
- RNA structure

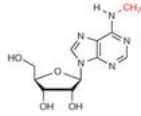


- miRNA activity
- RNA binding proteins
- 5' RNA degradation
- Ribosome specialization
- **RNA modifications (epitranscriptome)**

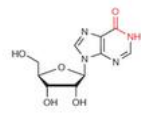
1950s: RNA modifications fine-tune structure/function



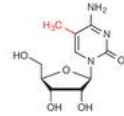
*N*⁶,2'-O-dimethyladenosine (m⁶Am)



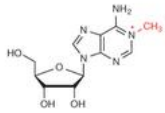
*N*⁶-methyladenosine (m⁶A)



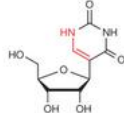
Inosine (I)



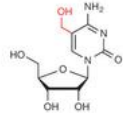
5-methylcytidine (m⁵C)



*N*¹-methyladenosine (m¹A)

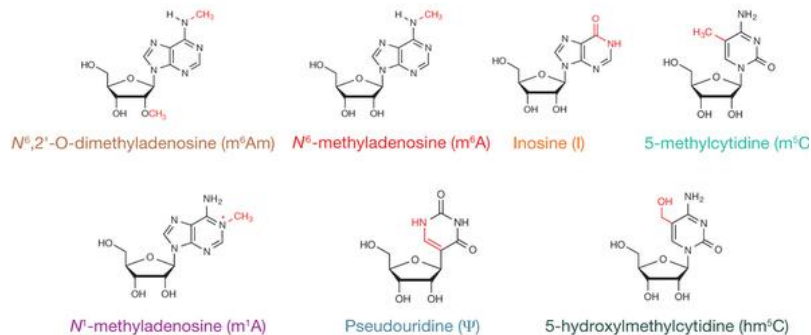


Pseudouridine (Ψ)

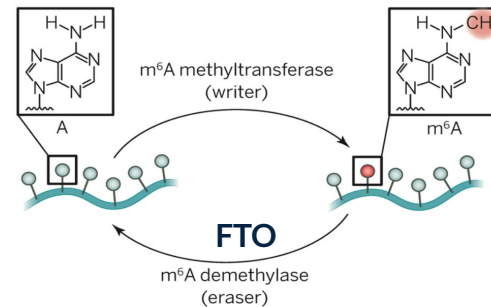


5-hydroxymethylcytidine (hm⁵C)

1950s: RNA modifications fine-tune structure/function



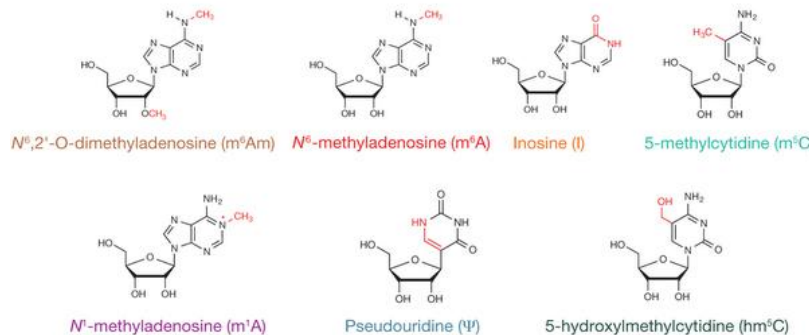
2011: RNA modifications are reversible!



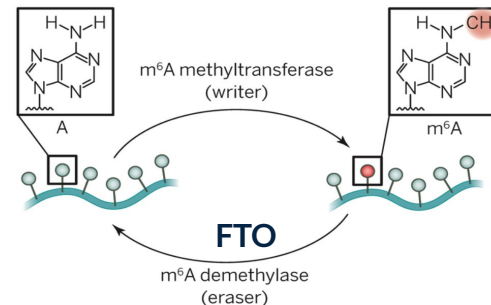
>> **function**

Jia et al., Nature Chemical Biology 2011

1950s: RNA modifications fine-tune structure/function



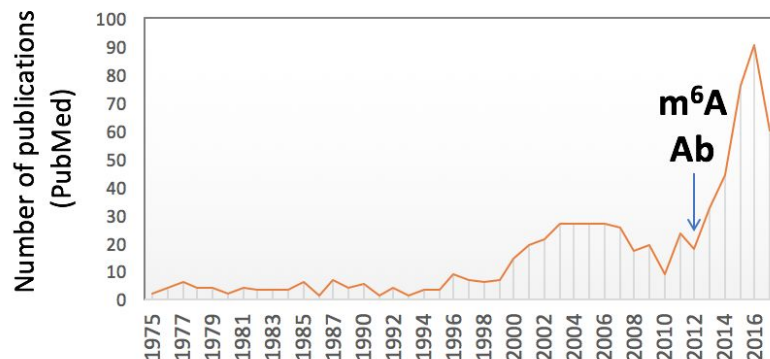
2011: RNA modifications are reversible!



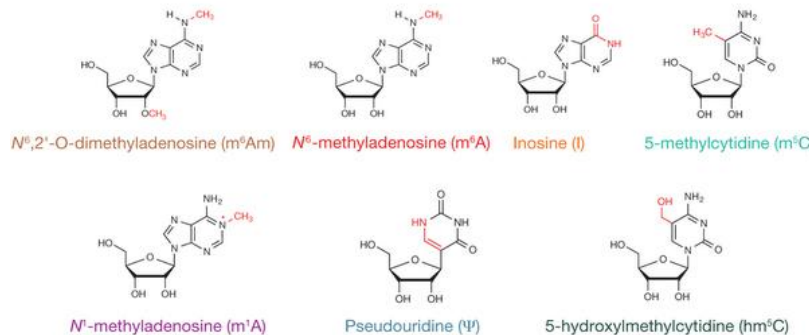
>> **function**

Jia et al., *Nature Chemical Biology* 2011

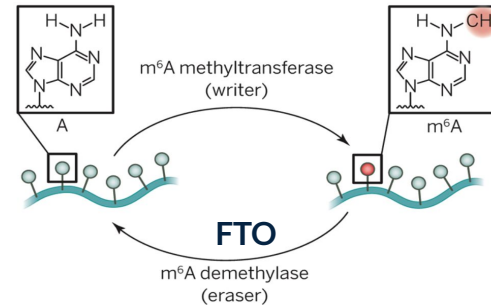
2012: First genome-wide method (m^6A -Seq)



1950s: RNA modifications fine-tune structure/function



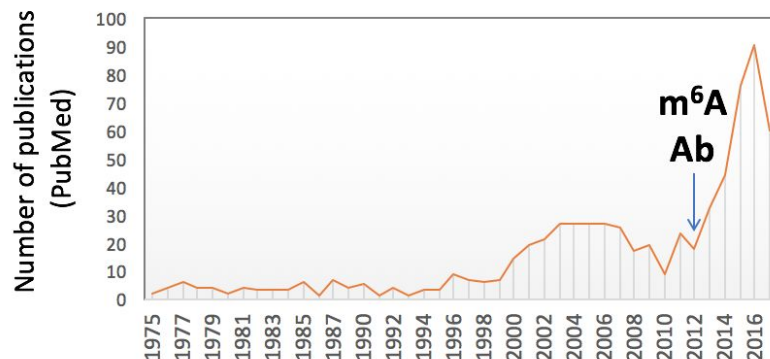
2011: RNA modifications are reversible!



>> function

Jia et al., *Nature Chemical Biology* 2011

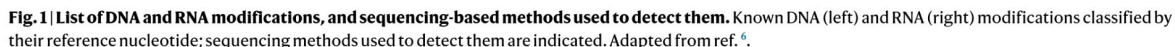
2012: First genome-wide method (m6A-Seq)



2013+: Pivotal roles of m6A in cellular functions

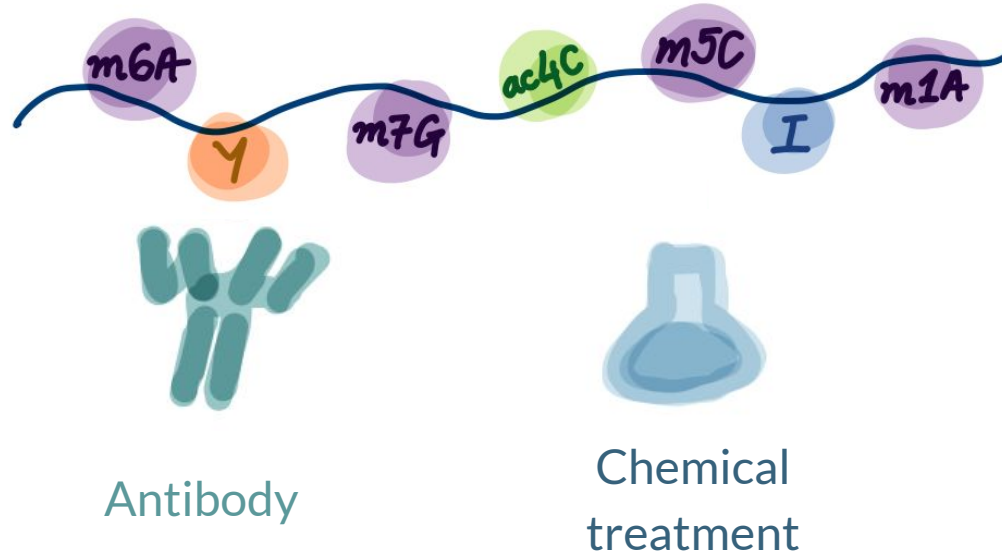
- Cell differentiation (2014)
- Stress responses (2015)
- mRNA half lives /RNA stability (2013)
- Sex determination (2016)
- Embryonic development (2017)
- Splicing (2024)

But... what about other modifications?



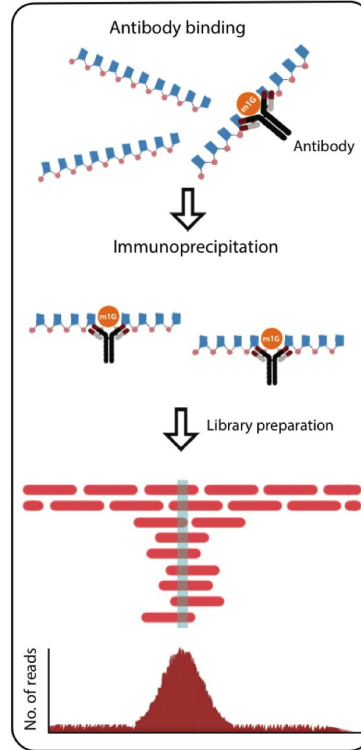
16

How can RNA modifications be detected with short-read sequencing?

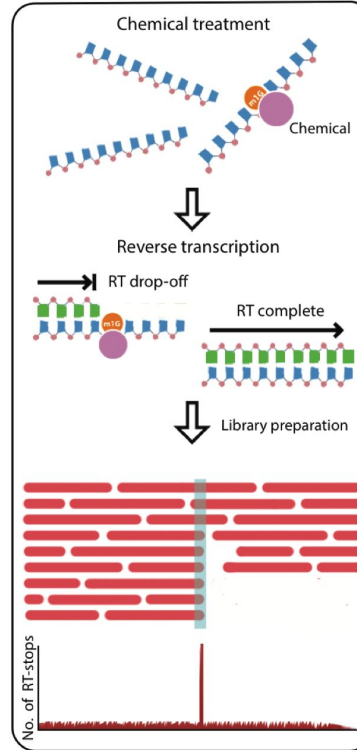


How can RNA modifications be detected with short-read sequencing?

Antibody-based
RIP-Seq
e.g. m6A, m1A



Chemical-based
Chem-Seq
e.g. m5C, Y

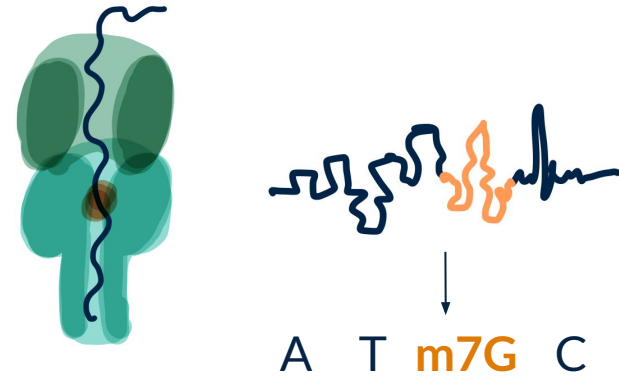


Jonkhout et al., RNA 2017



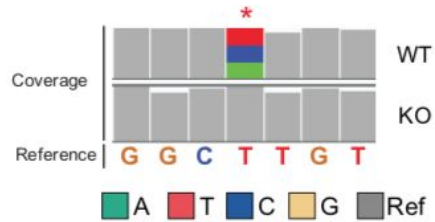
- No PCR bias
- **In principle, all RNA modifications**
- No custom protocols
- Stoichiometry information (% modified)
- Isoform-specific RNA modifications
- Multiple RNA modification types (m6A, m1A, m5C, etc) within the same molecule
- PolyA tail lengths
- **Single molecule resolution**





1

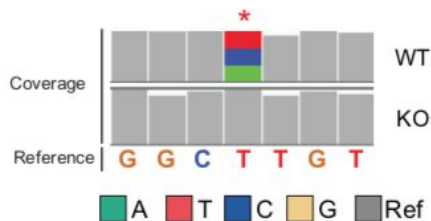
Differential Basecalling 'errors'



Liu et al., *Nature
Communications* 2019
(Epinano)

1

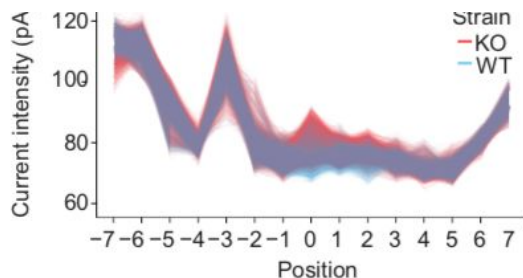
Differential Basecalling 'errors'



Liu et al., *Nature Communications* 2019
(Epinano)

2

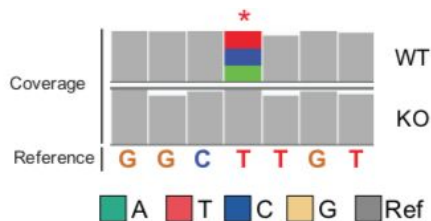
Alterations in Current intensity



Begik et al., *Nature Biotechnology* 2021
(NanoRMS)

1

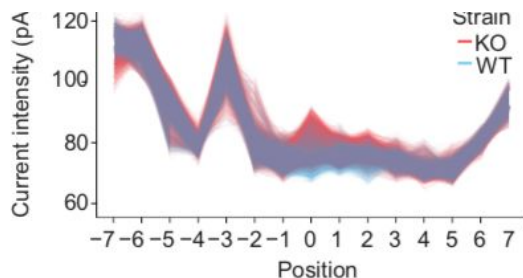
Differential Basecalling 'errors'



Liu et al., *Nature Communications* 2019
(Epinano)

2

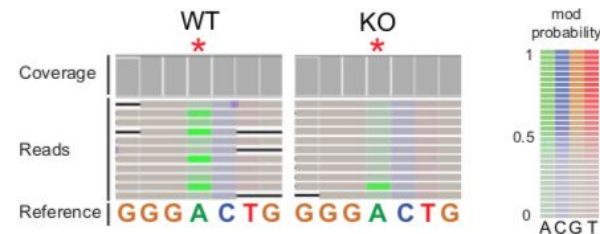
Alterations in Current intensity



Begik et al., *Nature Biotechnology* 2021
(NanoRMS)

3

Modification-aware Basecalling models



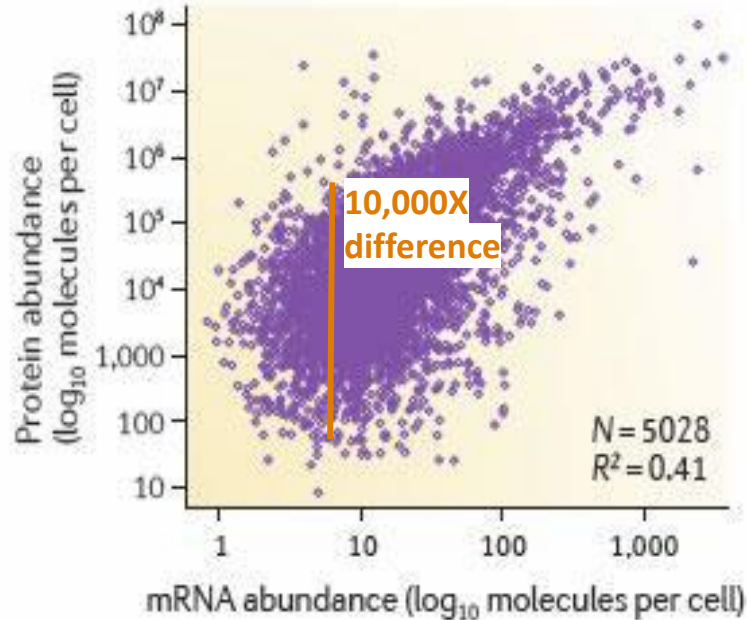
Cruciani et al., *Genome Biology* 2025
(m6ABasecaller)

- 1 Why do we care about RNA modifications?
- 2 What is the most well-studied RNA modification and what are some with which biological functions has it been associated with?
- 3 What are the advantages of ONT DRS over short-read sequencing for modification detection?
- 4 What are the 3 main strategies in RNA modification detection?
Which is the significantly superior one?

Hands-on Session: Theory & Rationale

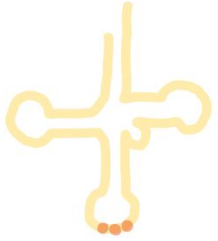
Working with DRS tRNA modification data

- Translation efficiency
- tRNA availability
- mRNA half life
- Codon usage
- RNA structure

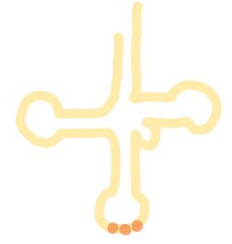


- miRNA activity
- RNA binding proteins
- 5' RNA degradation
- Ribosome specialization
- RNA modifications (epitranscriptome)

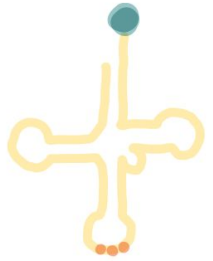
tRNA



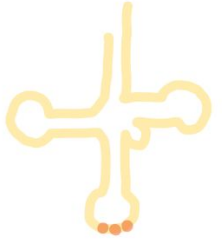
tRNA



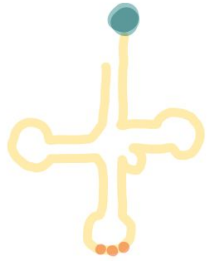
Amino acid



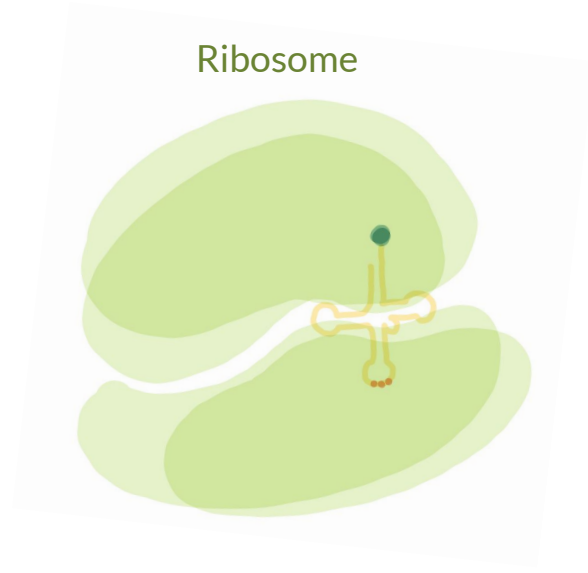
tRNA



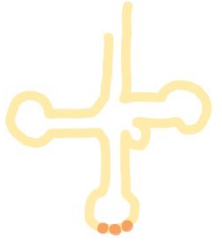
Amino acid



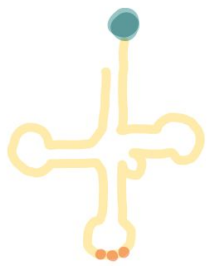
Ribosome



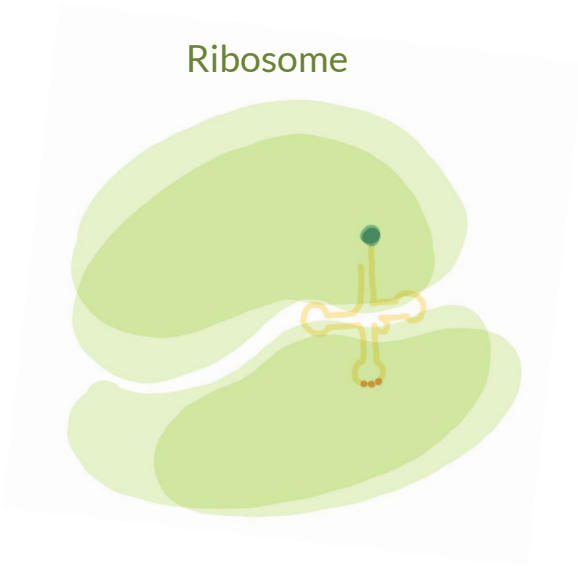
tRNA



Amino acid

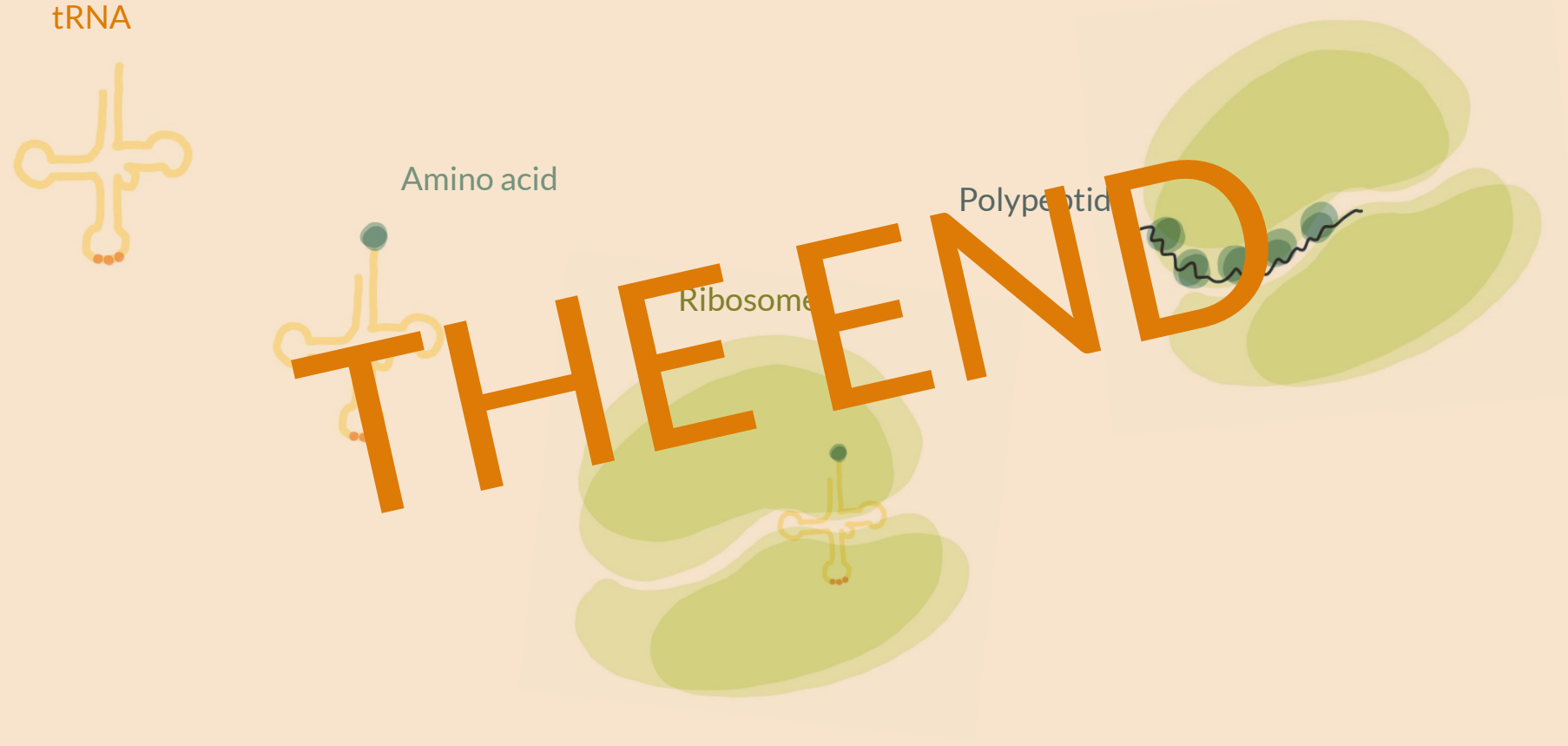


Ribosome

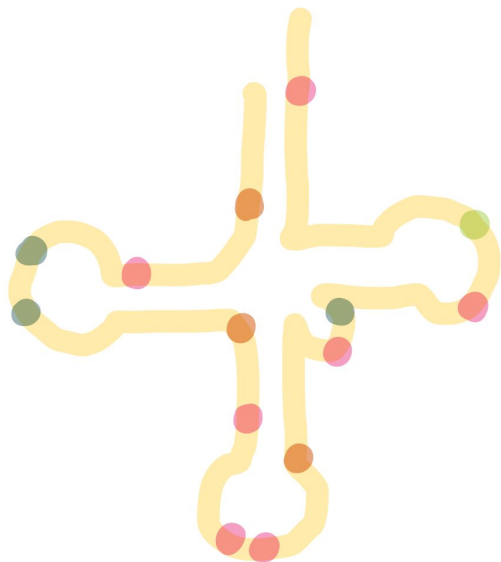


Polypeptide





tRNA



- typically **70-90** nts long
- most **abundant** RNA species
- heavily post-transcriptionally modified **~13 mods/molecule**

Further reading:

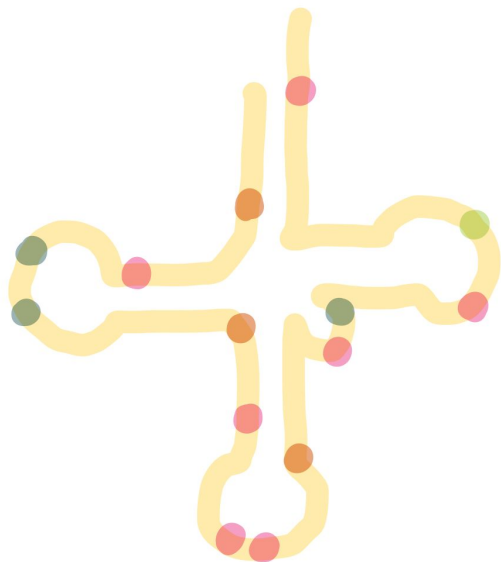
[Suzuki et al., *Nature Reviews Molecular Cell Biology* 2021](#)

[Pan et al., *Cell Research* 2018](#)

[Goodarzi et al., *Cell* 2016](#)

[C. Close et al., *Cancer and Noncoding RNAs* \(Chapter 10\) 2018](#)

tRNA



- typically 70-90 nts long
- most abundant RNA species
- heavily post-transcriptionally modified ~13 mods/molecule
- chemical modifications affect **translational dynamics**
 - modifications affect structure and function of tRNA
 - modifications modulate direct tRNA-rRNA interactions

Further reading:

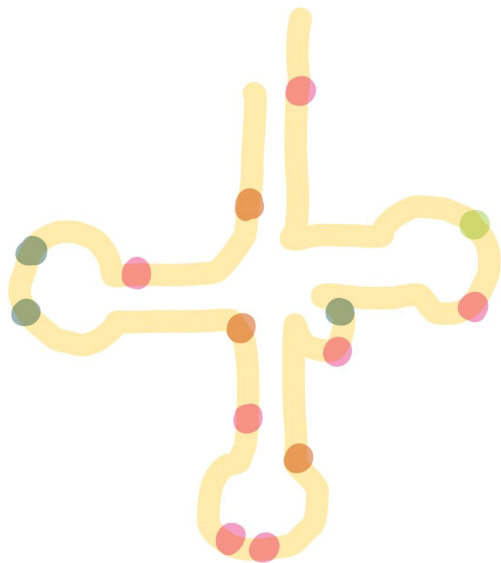
[Suzuki et al., *Nature Reviews Molecular Cell Biology* 2021](#)

[Pan et al., *Cell Research* 2018](#)

[Goodarzi et al., *Cell* 2016](#)

[C. Close et al., *Cancer and Noncoding RNAs* \(Chapter 10\) 2018](#)

tRNA



- typically 70-90 nts long
- most abundant RNA species
- heavily post-transcriptionally modified ~13 mods/molecule
- chemical modifications affect translational dynamics
- tRNA abundance and modification dysregulation involved in
 - mitochondrial diseases
 - neurological disorders
 - cancer

Further reading:

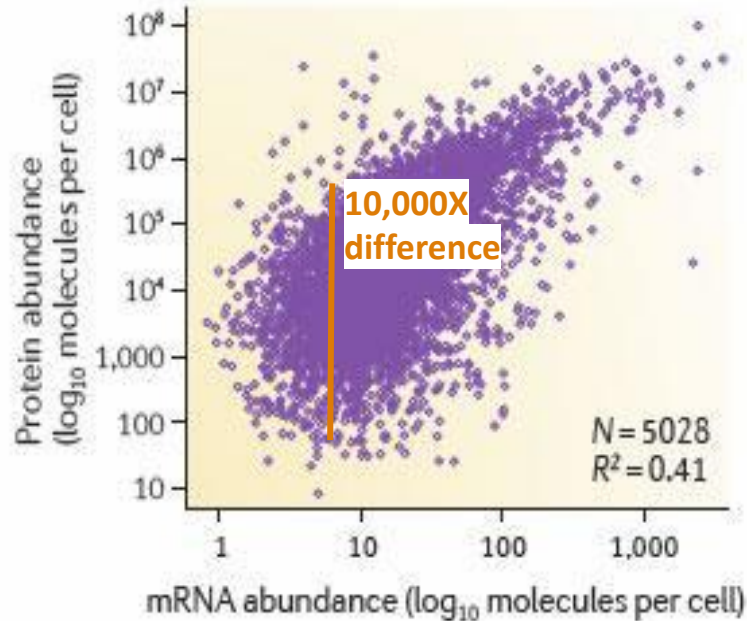
[Suzuki et al., *Nature Reviews Molecular Cell Biology* 2021](#)

[Pan et al., *Cell Research* 2018](#)

[Goodarzi et al., *Cell* 2016](#)

[C. Close et al., *Cancer and Noncoding RNAs* \(Chapter 10\) 2018](#)

- Translation efficiency
- tRNA availability
- mRNA half life
- Codon usage
- RNA structure



- miRNA activity
- RNA binding proteins
- 5' RNA degradation
- Ribosome specialization
- RNA modifications (epitranscriptome)

the challenges:

- DRS caters towards longer reads; inefficient capturing of <200nt-long transcripts, unable at <100nt
- first 15nt at 5' typically lost due to an increase in RNA translocation speed
- clover-like secondary structure of tRNA molecules

ACS Nano > Vol 15/Issue 10 > Article

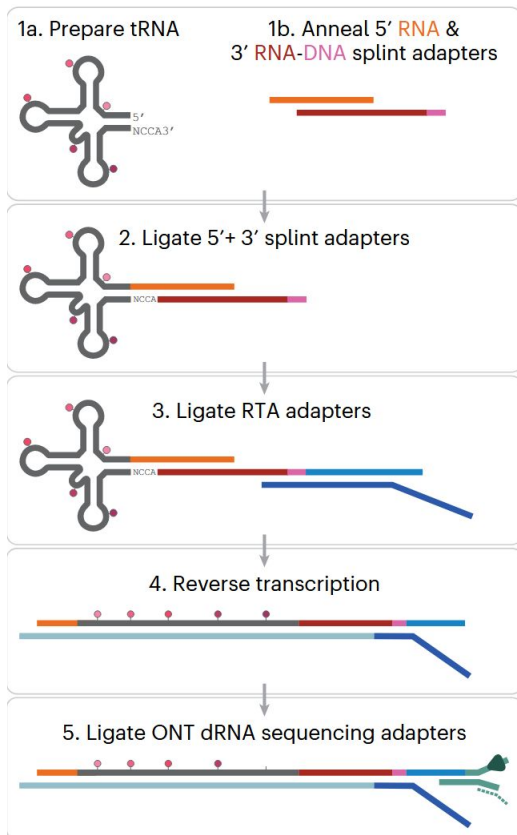
Open Access

ARTICLE | October 7, 2021

Direct Nanopore Sequencing of Individual Full Length tRNA Strands

Niki K. Thomas, Vinay C. Poodari, Miten Jain, Hugh E. Olsen, Mark Akeson, and Robin L. Abu-Shumays*

Thomas et al., *ACS Nano* 2021



Nano-tRNAseq protocol

Article | [Open access](#) | [Published: 06 April 2023](#)

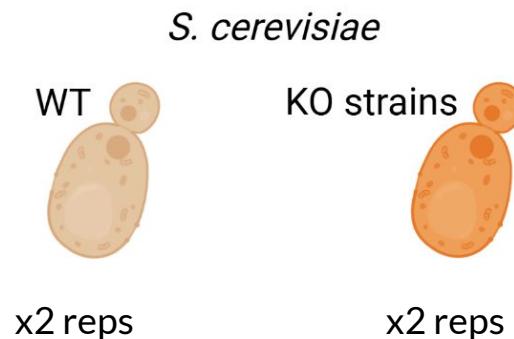
Quantitative analysis of tRNA abundance and modifications by nanopore RNA sequencing

[Morghen C. Lucas](#), [Leszek P. Pryszcz](#), [Rebeca Medina](#), [Ivan Milenkovic](#), [Noelia Camacho](#), [Virginie Marchand](#), [Yuri Motorin](#), [Lluís Ribas de Pouplana](#) & [Eva Maria Novoa](#) 

[Nature Biotechnology](#) **42**, 72–86 (2024) | [Cite this article](#)

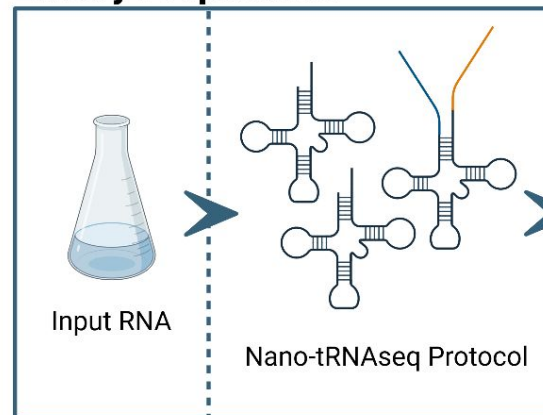
[Lucas & Pryszcz et al., Nature Biotechnology 2023](#)

- 5 teams of 5-6 people
- 6 **secret** datasets

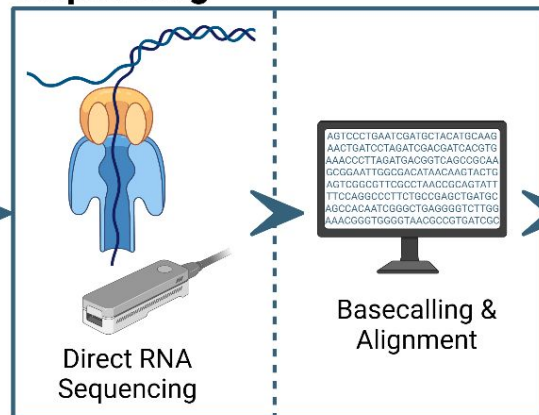


- get familiarized with the data and tool
- run the analysis on your selected dataset
- interpret and discuss your results with teammates
- have me judge you

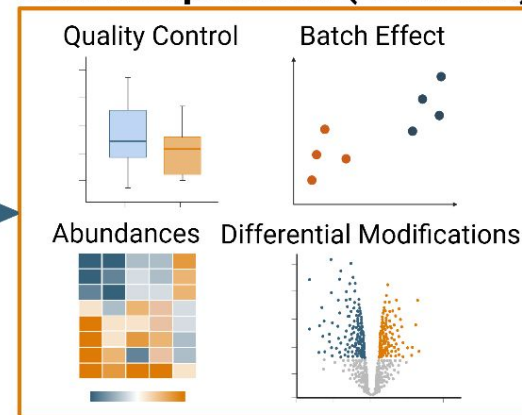
Library Preparation



Sequencing



Data Interpretation (AMaNITA)

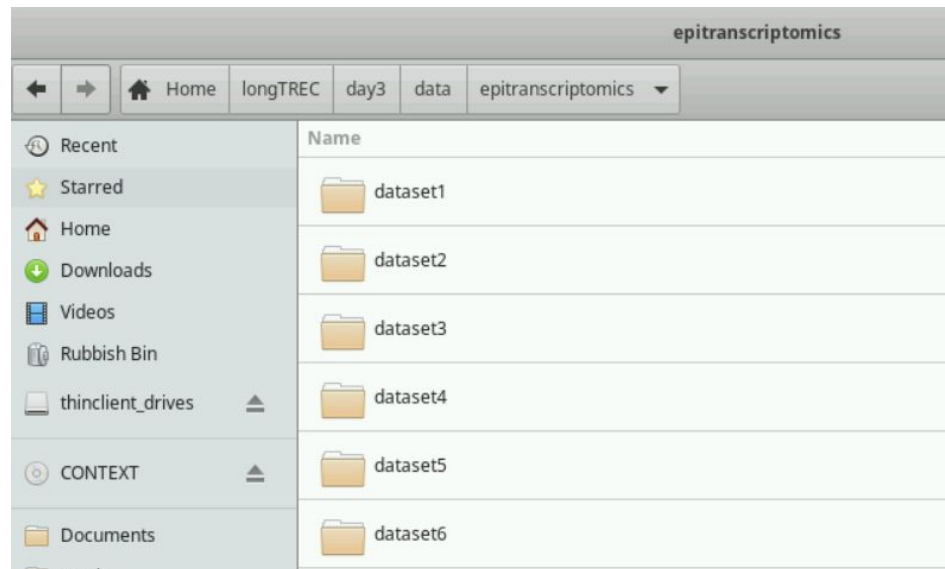
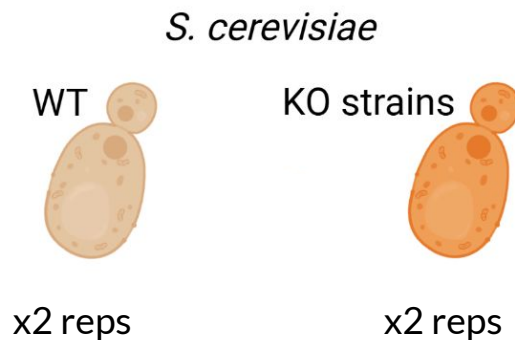


- 1 Why do we care about tRNAs?
- 2 Why is ONT DRS better for the sequencing of tRNAs?
- 3 What are the current limitations of studying tRNAs with ONT DRS?

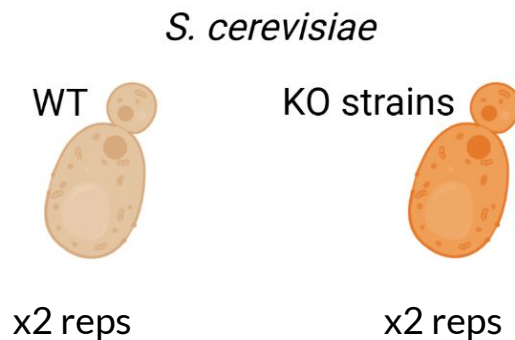
Hands-on Session: Downstream Interpretations with AMaNITA




Analyzing yeast Nano-tRNAseq data

- 5 teams of 5-6 people
- 6 *secret* datasets

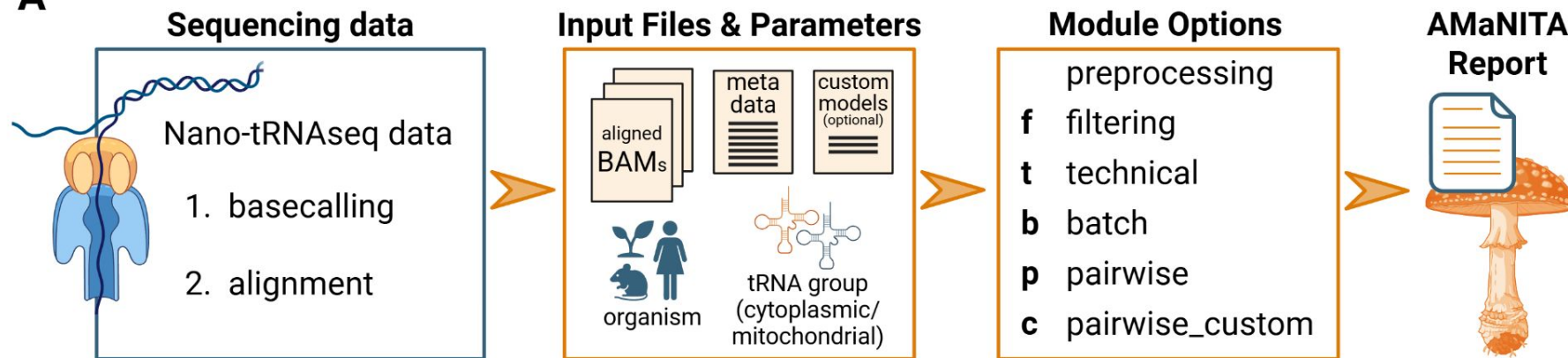


- 5 teams of 5-6 people
- 6 *secret* datasets

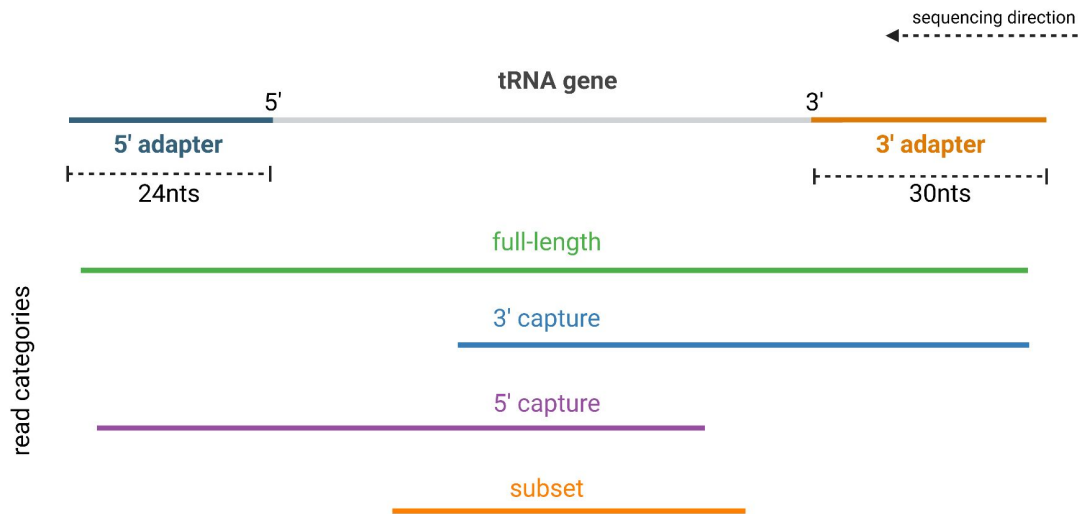


dataset1				
TREC	day3	data	epitranscriptomics	dataset1 ▼
Name				
	reads.KO_Rep1.bam			
	reads.KO_Rep2.bam			
	reads.WT_Rep1.bam			
	reads.WT_Rep2.bam			

A



filtering module

**Filters**

1. All **5'** and **subset** reads are discarded
2. All **3'** reads spanning < 25b of the tRNA gene are discarded
3. Any type of reads with deletions/non-matches > 10b are discarded

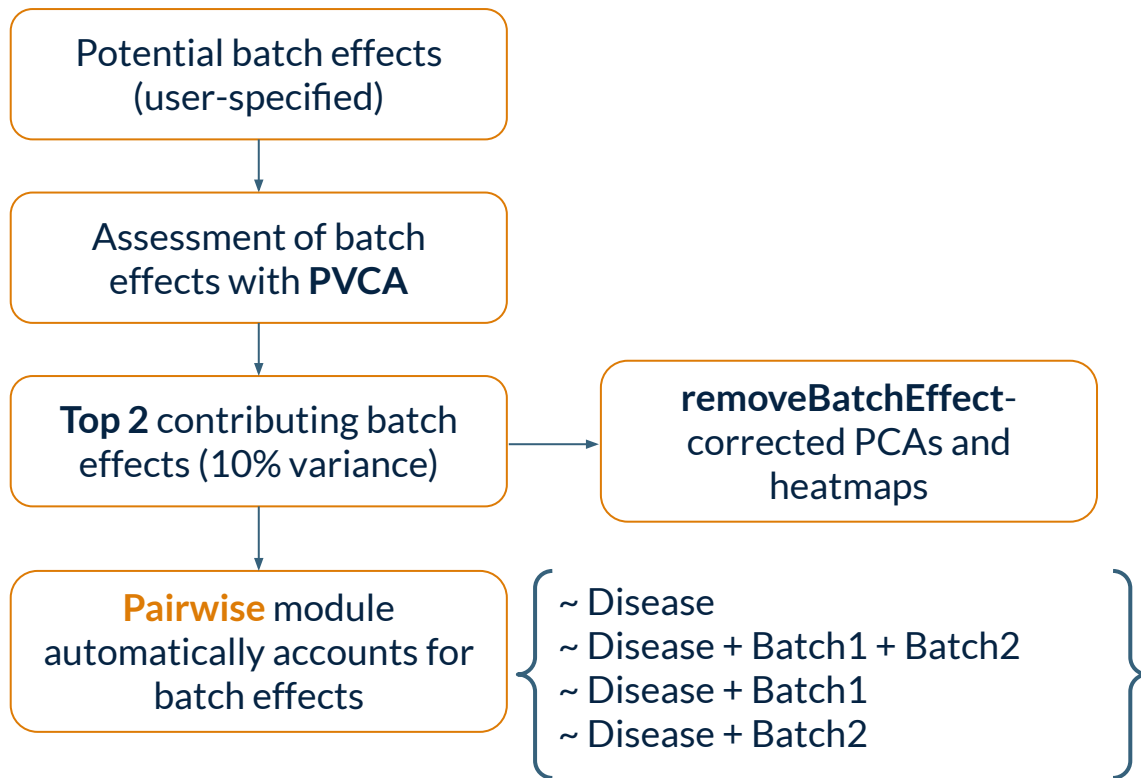
batch module

1. PVCA > identify potential batch effects, based on the variance explained by each variable
2. limma::removeBatchEffect > plotting ONLY (heatmaps and PCA plots)
3. Linear model accounting for batch variables > differential analyses
e.g. ~ **Tissue** vs ~ **Tissue + RunDate**

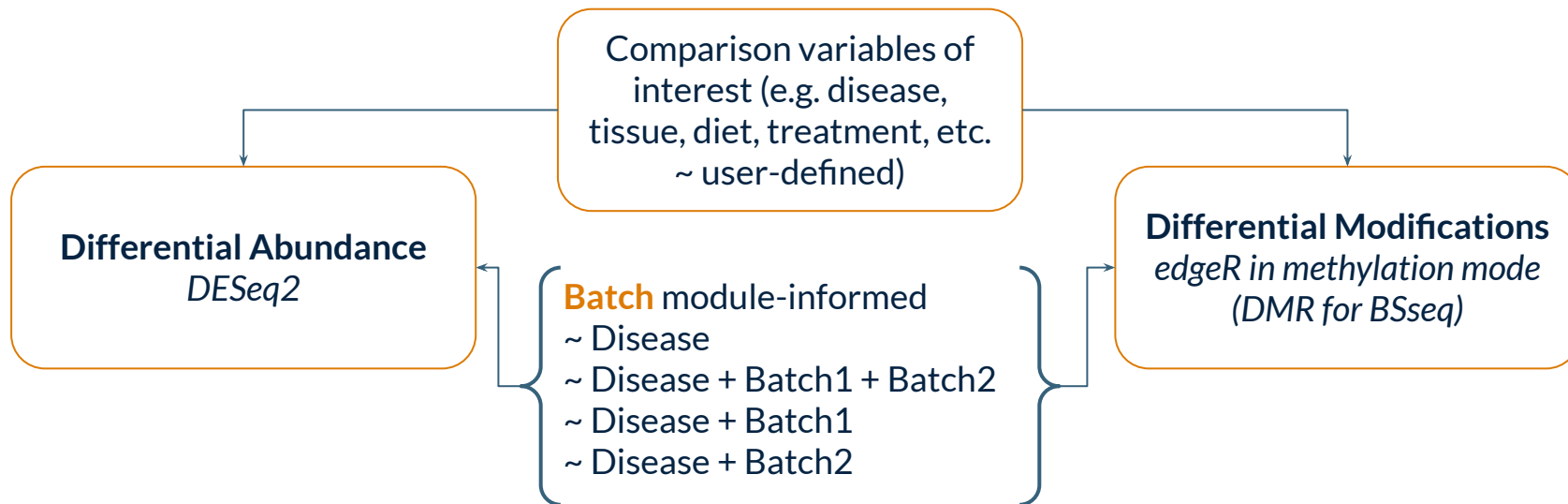
 **Batch effect correction should be treated with caution!**
Unbalanced batches and/or using batch corrected data for purposes other than intended is very dangerous and can lead to wrong conclusions! 

Nygaard et al., Biostatistics 2016

batch module



pairwise module



summer_school / practicals / day3 / epitranscriptomics / AMaNITA_docs /

 **LedaKatopodi** Add files via upload ✓

Name	Last commit message
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wiki	Add files via upload
README.md	Create README.md

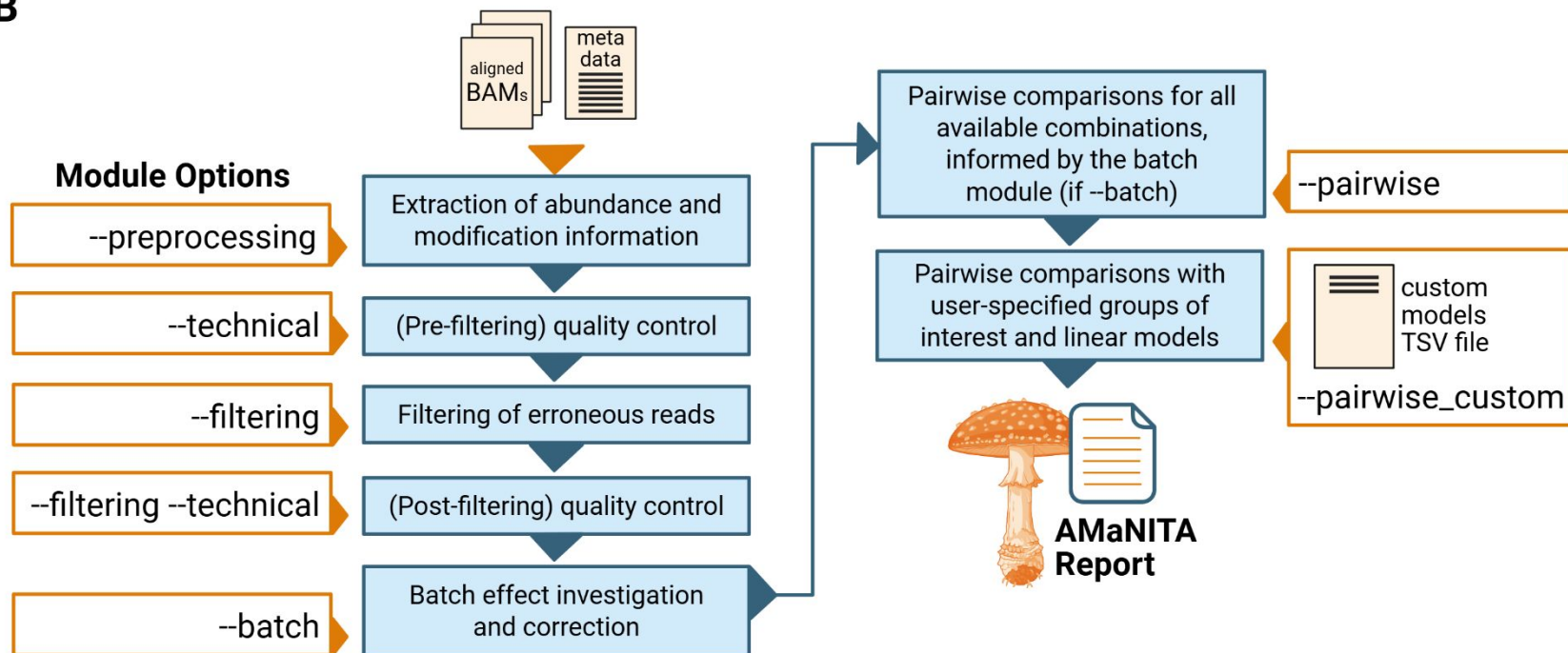
README.md

AMaNITA

Abundance, Modifications, and Nanopore Intensity Toolbox/Application



B



How to properly build your Metadata file

AMaNITA is built around a properly defined metadata table that provides all information required for an end-to-end analysis of the input data. This is a walkthrough on how to build your metadata table, things you should be careful, tips and tricks.

Important

The metadata file is a tab-separated file, expected to be named `{project_id}.metadata.tsv`.


The metadata file is expected to be under the project directory, e.g. `{wrk_dir}/{project_id}/{project_id}.metadata.tsv`.

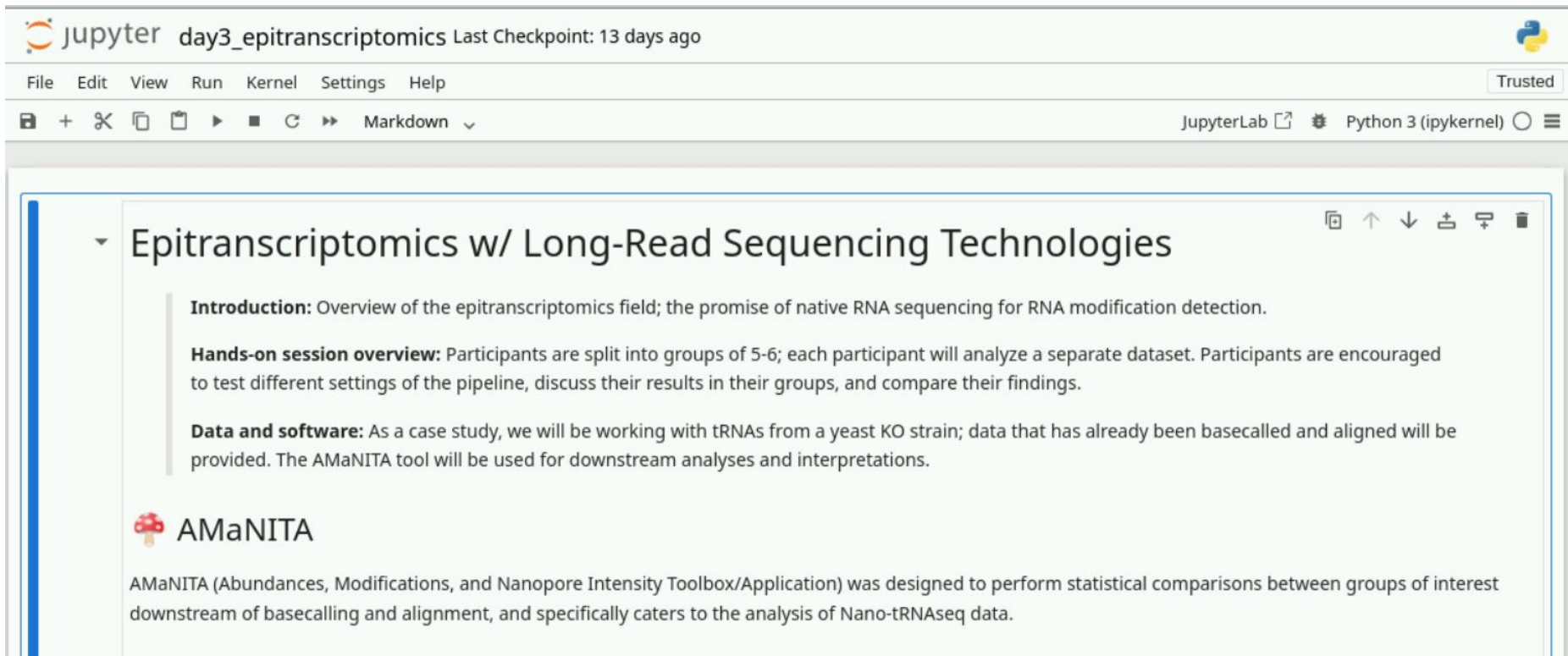
Each **line** contains information on **one** sample.

When creating the metadata table, make sure that the newline character is Unix-based (i.e. not Windows- or MacOS-based)

Expected columns:

- 1: `Project`
- 2: `Run`
- 3: `BAMID`
- 4: `BarcodeID`
- 5: `SampleID`
- 6: (optional) `ReplicateID`
- 6/7+: variables of interest, tagged with either of the following: `comp_`, `batch_`, or `extra_`

 Columns 1-6 (when ReplicateID is included) require the column names to be as shown above (`Project` , `Run` , `BAMID` , `BarcodeID` , `SampleID` , `ReplicateID`)



The screenshot shows a JupyterLab environment. At the top, the title bar reads 'jupyter day3_epitranscriptomics Last Checkpoint: 13 days ago'. Below this is a menu bar with 'File', 'Edit', 'View', 'Run', 'Kernel', 'Settings', and 'Help'. To the right of the menu bar is a 'Trusted' button. Below the menu bar is a toolbar with various icons for file operations and execution. The main area displays a notebook with the title 'Epitranscriptomics w/ Long-Read Sequencing Technologies'. The notebook content includes an introduction, a hands-on session overview, and data and software information. The AMaNITA logo is also present.

Epitranscriptomics w/ Long-Read Sequencing Technologies

Introduction: Overview of the epitranscriptomics field; the promise of native RNA sequencing for RNA modification detection.

Hands-on session overview: Participants are split into groups of 5-6; each participant will analyze a separate dataset. Participants are encouraged to test different settings of the pipeline, discuss their results in their groups, and compare their findings.

Data and software: As a case study, we will be working with tRNAs from a yeast KO strain; data that has already been basecalled and aligned will be provided. The AMaNITA tool will be used for downstream analyses and interpretations.

AMaNITA

AMaNITA (Abundances, Modifications, and Nanopore Intensity Toolbox/Application) was designed to perform statistical comparisons between groups of interest downstream of basecalling and alignment, and specifically caters to the analysis of Nano-tRNAseq data.

Challenge 1: What are the KO?

Challenge 2: What are the differences between same-KO datasets?

- 1 How sensitive is modification detection with DRS and to which factors?
- 2 What are the challenges, advantages, and limitations of using basecalling errors as a proxy for modification detection?

Discussion

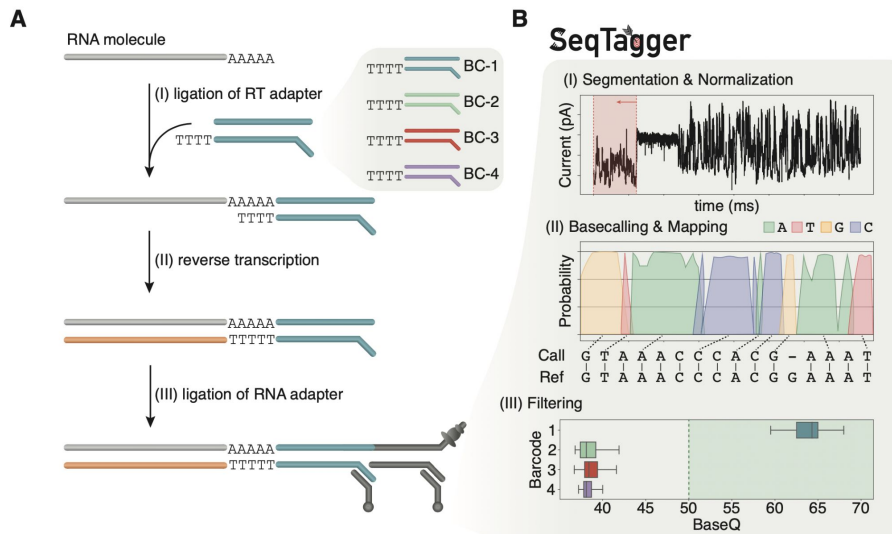
Last comments; take-home messages; Q&A

⚠ Batch effect correction should be treated with caution!
Unbalanced batches and/or using batch corrected data for purposes other than intended is very dangerous and can lead to wrong conclusions! ⚠

Nygaard et al., *Biostatistics* 2016

Priorities when dealing with batches:

1. Don't introduce batches! >>
multiplexing
(Pryszcz & Diensthuber et al.,
Genome Research 2025)



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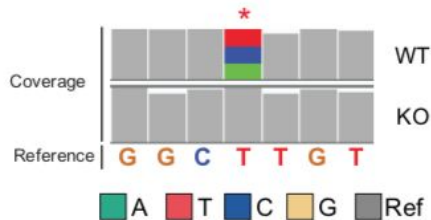
Nygaard et al., *Biostatistics* 2016

Priorities when dealing with batches:

1. Don't introduce batches! >>
multiplexing
(Pryszcz & Diensthuber et al.,
Genome Research 2025)
2. Make sure your batches are balanced during the experimental design step
3. When performing batch effect correction, make sure you are not over-correcting (and therefore losing biological information), and that you are using batch-corrected data for its intended purposes

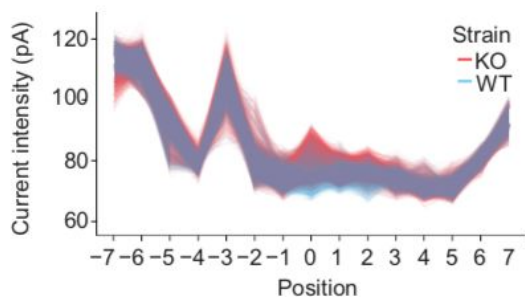
1

Differential Basecalling 'errors'



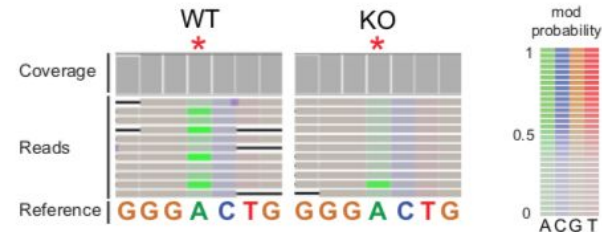
2

Alterations in Current intensity



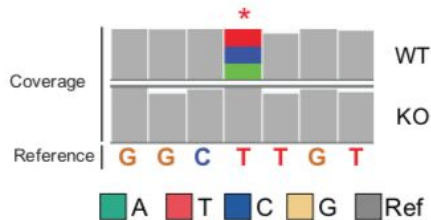
3

Modification-aware Basecalling models



1

Differential Basecalling 'errors'

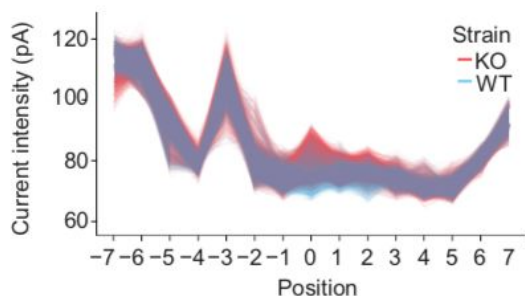


Epinano Liu et al., 2019
DiffErr Parker et al., 2020
Eligos Jenjaroenpun et al., 2021
Drummer Price et al., 2020

NanoConsensus Delgado-Tejedor et al., 2023

2

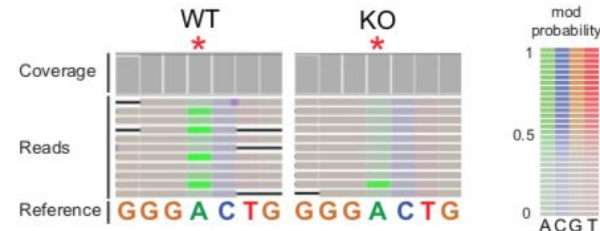
Alterations in Current intensity



Nanopolish Simpson et al., 2017
Tombo Stoiber et al., 2017 (*)
Nanocompore Leger et al., 2021
NanoRMS Begik et al., 2021

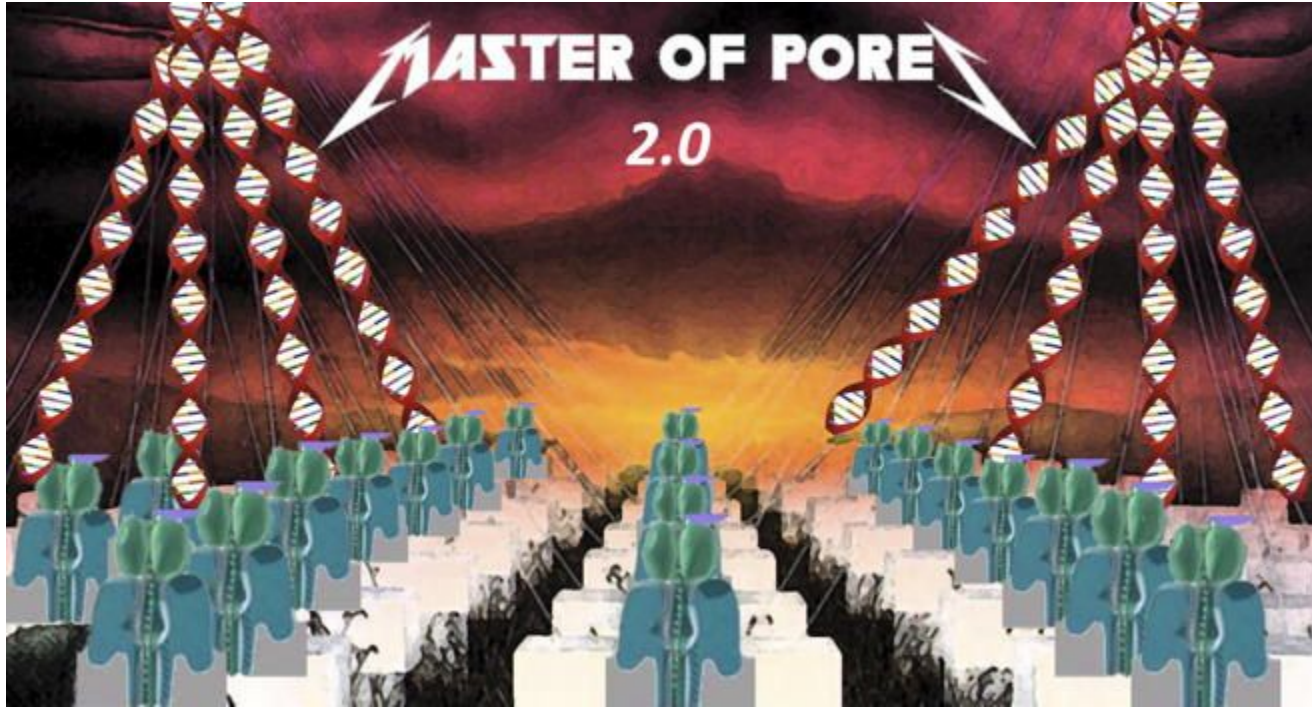
3

Modification-aware Basecalling models



Dorado modification-aware settings
m6ABasecaller Cruciani et al., 2025
Uncalled4 Kovaka et al., 2025

Comprehensive review: **Furlan et al., RNA Biology 2021**



Cozzuto et al., *Frontiers in Genetics* 2020





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[@secuenciasdecienicia](#)

Thank You!



For more information about the LongTREC Summer School:

<https://longtrec.eu>