



**University of  
Zurich** <sup>UZH</sup>



# **BIO634 - Day 2: RNA sequencing technologies**

**Carla Bello**, [carla.bello@ieu.uzh.ch](mailto:carla.bello@ieu.uzh.ch)

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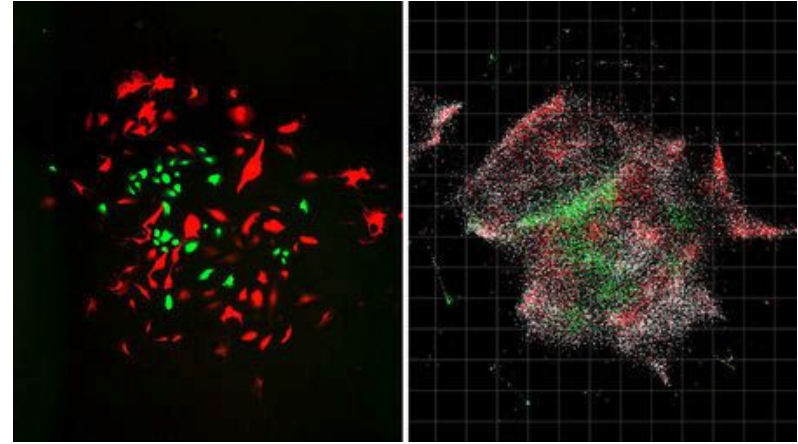
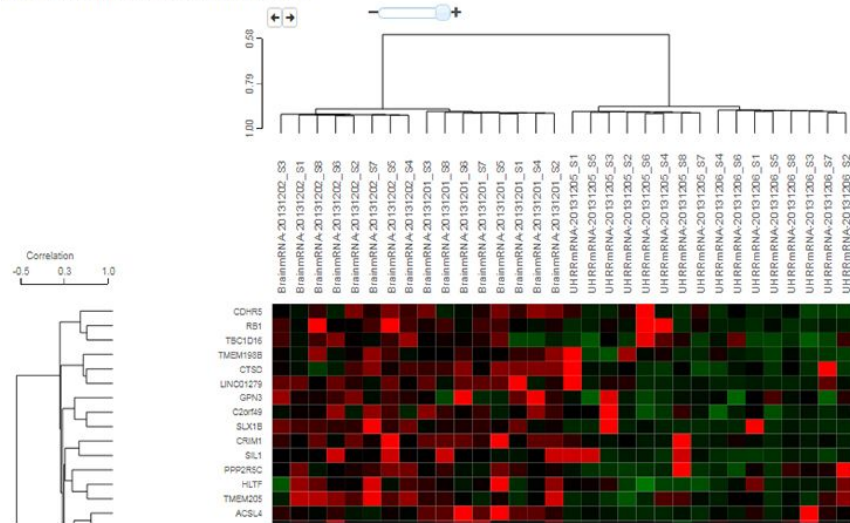
Zürich, CH

# Overview

- **Introduction** to RNA sequencing
- **RNA sequencing workflow**: steps to analyze the data
- **Important considerations**: technical, biological replicates, etc
- **Abundance quantification**: Gene, exons, transcripts
- **Differential expression analyses**: DEseq, edgeR, etc

# RNAseq reveals the presence and quantity of RNA

DIFFERENTIAL EXPRESSION HEATMAP <sup>i</sup>

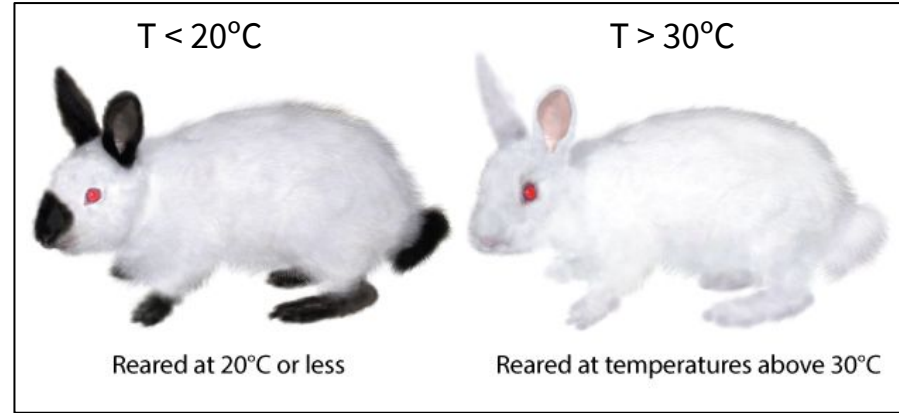


NGS technology that **quantifies the transcription / expression of the genome**

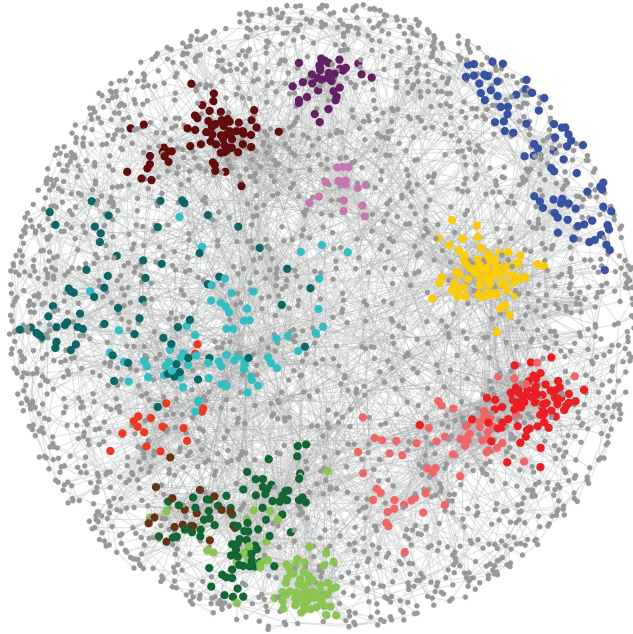
# Gene expression and phenotype

Image from Wikipedia

Himalayan rabbit

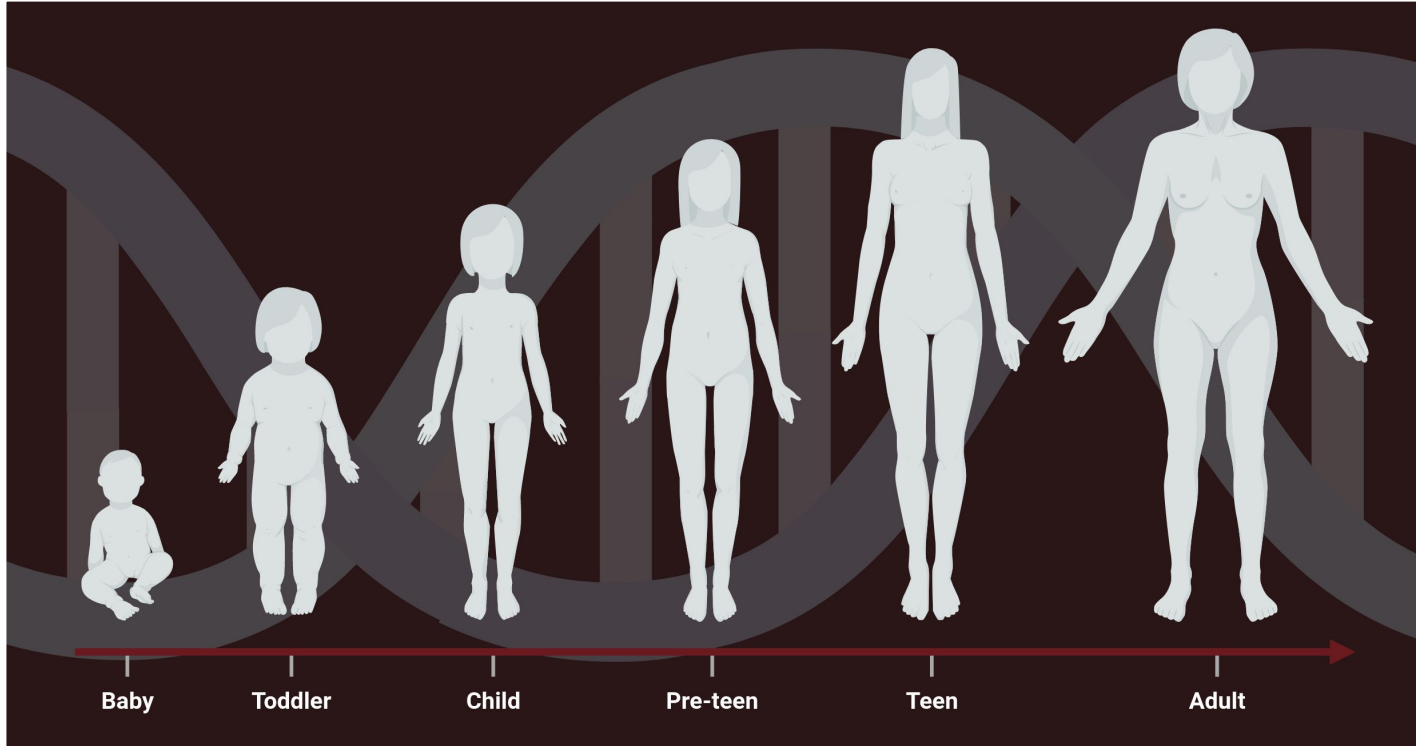


**Example:** A pigment gene is influenced by temperature.  
When the temperature is **< 20°C** the **gene is inactive**



Nothing in the genome makes sense except in the light of the transcriptome

# RNA sequencing of whole genomes

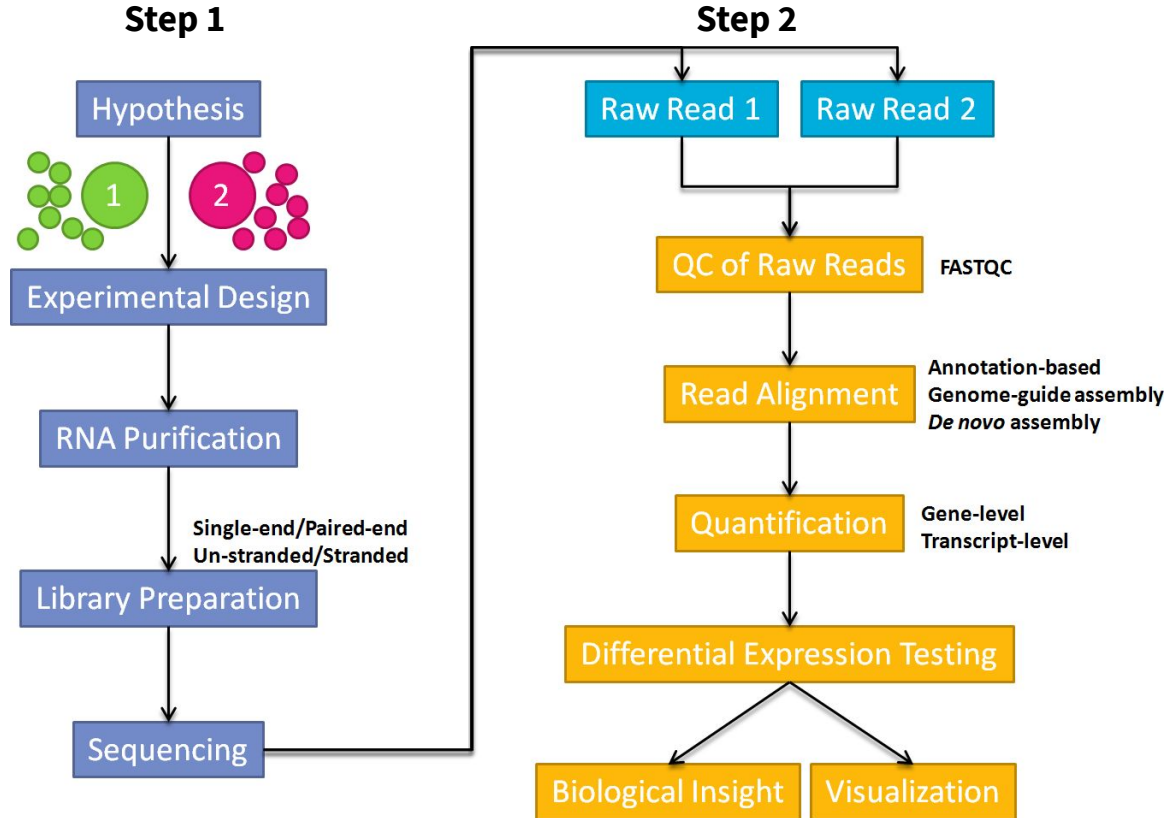


Expression of all the genes in the genome at different stages of aging

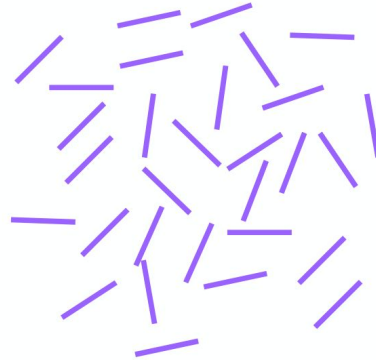
# Applications of RNA sequencing

- **Gene** expression/**differential** gene expression
- **Detecting novel** or **alternative** transcripts
- **De-novo transcriptome** assembly
- **SNP analysis**, e.g disease association studies
- **Allele-specific expression**
- **RNA studies**: miRNA, tRNAs, snRNA, lncRNAs, etc.

# RNA sequencing workflow



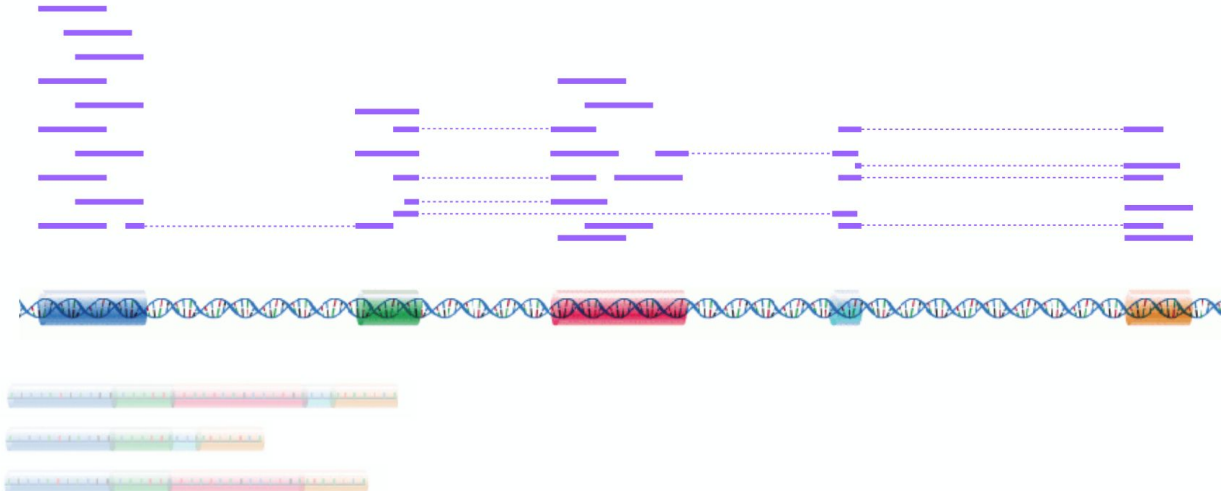
# Abundance quantification





# Abundance quantification

Genome alignment of RNA-seq requires a splice-aware aligner (STAR, HISAT2)



# Abundance quantification

**Transcript-level** counts, e.g. obtained by  
“alignment-free” estimation methods

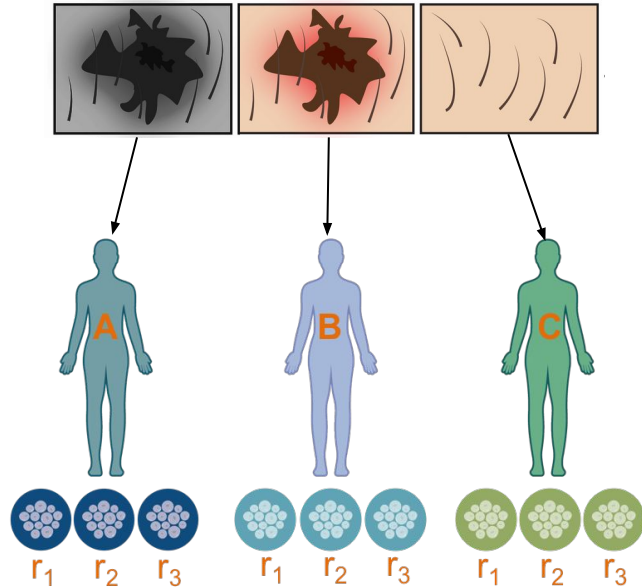


- **Sailfish** (Patro et al, *Nat Biotechnol* 2014)
- **Salmon** (Patro et al, *Nat Methods* 2017)
- **kallisto** (Bray et al, *Nat Biotechnol* 2016)

Pre-indexed datasets of  
transcripts

# Important considerations before sequencing

3 different samples  
3 different conditions



3-12 replicates per sample/condition

**Technical and biological replicates are important** and should be taken into consideration **when planning the experiments**

# Technical replicates

1. Technical replicates: Biological material is the same but the technical steps used to measure gene expression are repeated.

In particular **RNA-seq library preparation** (RNA fragmentation, cDNA synthesis and PCR amplification) **may introduce biases in the data**.

# Biological replicates and statistical power

2. **Biological replicates**: Are different biological samples that are processed separately. They are **required if inference on the population is to be made**, with **three** biological replicates **being the minimum for any inferential analysis**.
3. **Desired statistical power**, that is the capacity for detecting statistically significant differences in gene expression between experimental groups.

# Different methods to analyze RNA-seq data

- There are different packages for **differential expression analysis**, such as **edgeR** and **DESeq** based on negative binomial (NB) distributions or **baySeq** and **EBSeq** which are **Bayesian** approaches based on a **negative binomial model**.
- These packages work mostly by **estimating the variance mean dependance** in **count** data.

# Factors to consider for RNAseq analysis

## 1) Within sample

- Gene/transcript length
- Relative expression (a few highly expressed genes)

## 2) Between samples

- Sequencing depth (library size)
- Sequencing biases

3) Raw read counts **are NOT directly** comparable **between** samples:

**Solution:** Normalize read counts

# RNA-Seq Read Count Normalization

- **RPKM**: Reads per kilobase of transcript per million reads of library (single-end reads)
- **FPKM**: Fragments per kilobase of transcripts per million reads of library (paired-end reads)
- **TPM**: Transcripts per million reads of library



# RNA-Seq Read Count Normalization

## RPKM:

**Reads Per Kilobase and Million mapped reads**

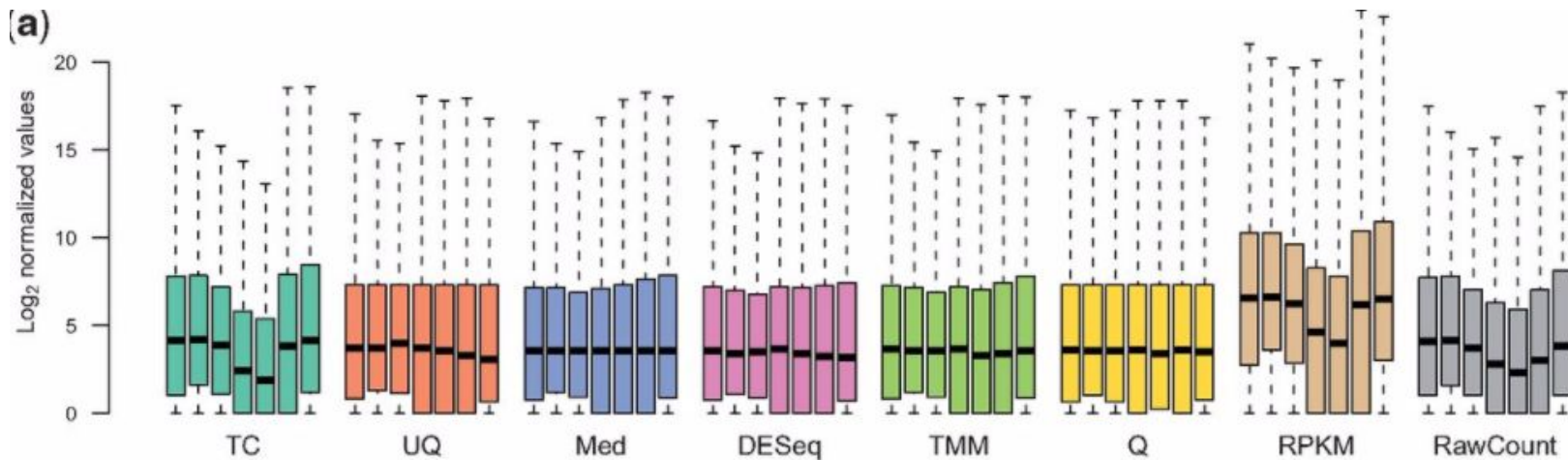
Unit of measurement

$$RPKM = \#MappedReads * \frac{1000bases * 10^6}{length\ of\ transcript * Total\ number\ of\ mapped\ reads}$$

- RPKM reflects the molar concentration of a transcript in the starting sample by normalizing for
  - RNA length
  - Total read number in the measurement
- This facilitates transparent comparison of transcript levels within and between samples

# RNA-Seq Read count normalization

RPKM/FPKM are normalized counts. **DESeq/edgeR** requires raw counts as **input** as they have their **own normalisation methods**



# Differential expression analyses

- **Many statistical methods available**
  - T-test
  - Poisson Distribution
  - Negative binomial
- **No clear consensus yet.**
- **Tools shown to perform well (under certain circumstances):**
  - LIMMA (TMM)
  - DESeq (RLE)
  - edgeR (TMM)
  - Cuffdiff (FPKM)
  - RSEM (EM)
  - Trinity

**Identify genes that show differences in expression level between conditions (samples)**

# Differential expression analyses

1. Mapping **RNAseq** data with **Salmon**
2. Importing **transcript abundance** with the ***tximport*** library
3. **Differential analysis:** A comparison between **DEseq** and **edgeR**

# Later in the afternoon

1. **List of differentially expressed genes**
2. **Biological context**
3. **Pathway Analysis (differentially expressed biological pathways)**
4. **Gene Set Enrichment Analysis (GSEA) (functional enrichment between two biological groups)**
5. **Co-expression analysis**

# Hands-on session - Part II: RNA-seq

**Please, go here and follow the instructions:**

**<https://github.com/carlalbc>**

**[https://github.com/carlalbc/BIO634\\_2021/](https://github.com/carlalbc/BIO634_2021/)**

~~ Lunch break from 12:00 to 13:30 ~~



Enjoy! See you in a bit! :)