

Bioinformatics for RNA-seq

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Rebecca Batorsky
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May 2020

Course Format

1-hour Zoom
Introduction
6/2 @ 11 am

~3 hours of self-guided
material on github,
suggested to be completed
over the **next week**:

[https://huoww07.github.io/
/Bioinformatics-for-RNA-
Seq/](https://huoww07.github.io/Bioinformatics-for-RNA-Seq/)

(working with a partner is
encouraged)

Piazza

- Please ask and answer questions liberally on [Piazza](#)
- Steps to enroll in class if you are not already enrolled:
 - <https://piazza.com/tufts>
 - Bioinformatics 2: Intro to RNA sequencing Bioinformatics
 - Join as student
- If you can't access Piazza for some reason please let us know Wenwen.Huo@tufts.edu or Rebecca.Batorsky@tufts.edu

Office Hours
6/4 @ 11 am
via zoom

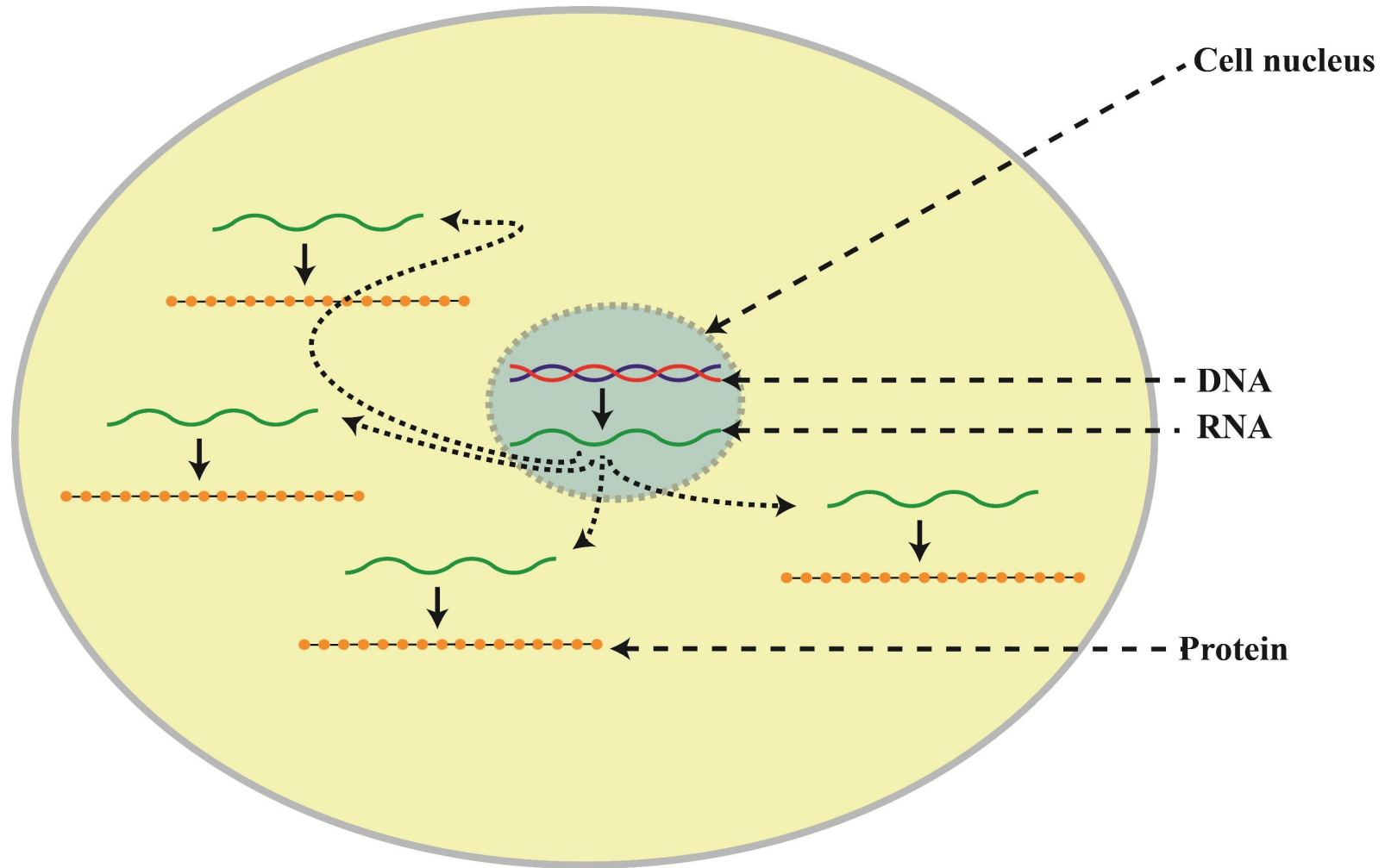
Requirements

- [HPC Cluster Account](#) available to Tufts affiliates
- [VPN](#) if working off campus
- Basic knowledge:
 - [Intro to Linux](#)
 - [HPC Quick Start guide](#) or [Intro to HPC](#)
 - [Introduction to R](#)

We'll test out access together during this session.

Depending on the number/type of questions, we may choose to follow up after the session.

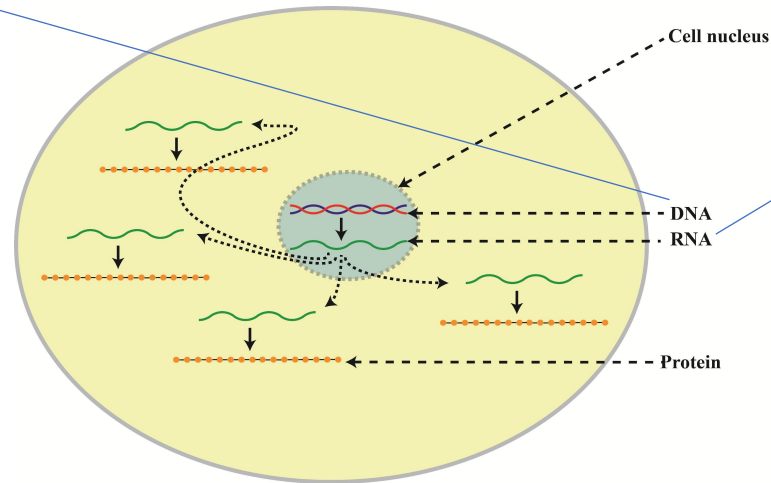
DNA and RNA in a cell



Two common analysis goals

DNA Sequencing

- Fixed copy number of a gene per cell
- Analysis goal: Variant calling and interpretation



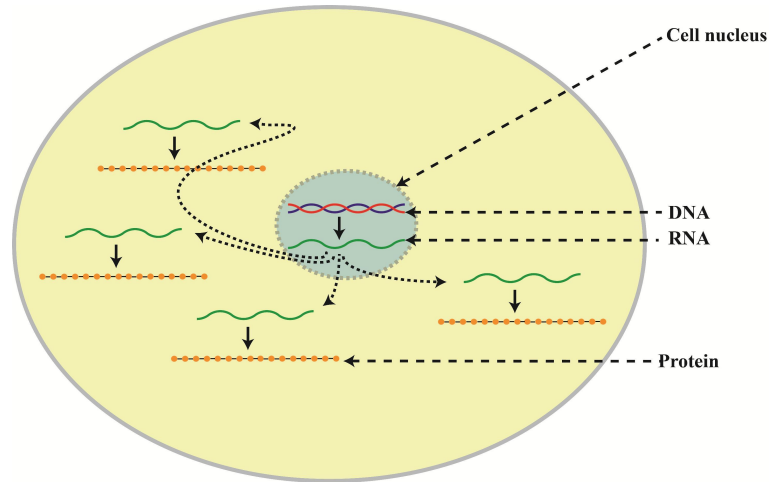
RNA Sequencing

- Copy of a gene (mRNA transcript) per cell depends on gene expression
- Analysis goal: Differential expression and interpretation

This workshop will cover RNA sequencing

Sequencing

- Fixed copy of gene per cell
- Analysis goal: Variability and interpretation



RNA Sequencing

- Copy of a gene per cell depends on gene expression
- Analysis goal: Differential expression and interpretation

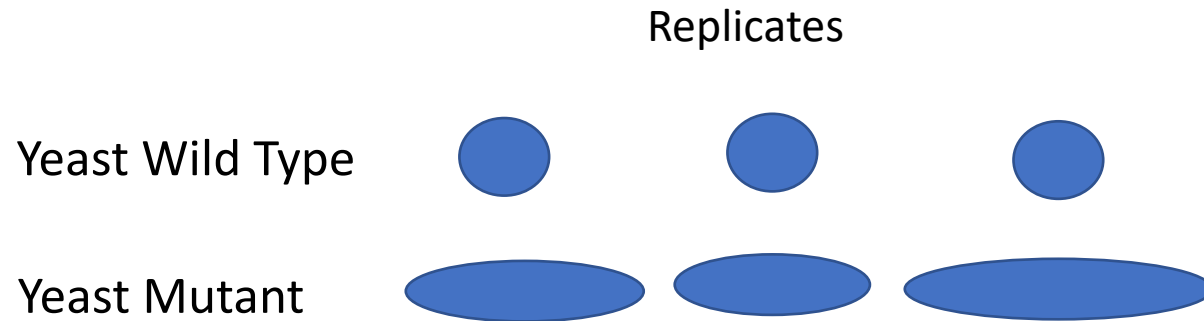
Not today!

Check out our “Intro to NGS” workshop:

<https://rbatorsky.github.io/intro-to-ngs-bioinformatics/>

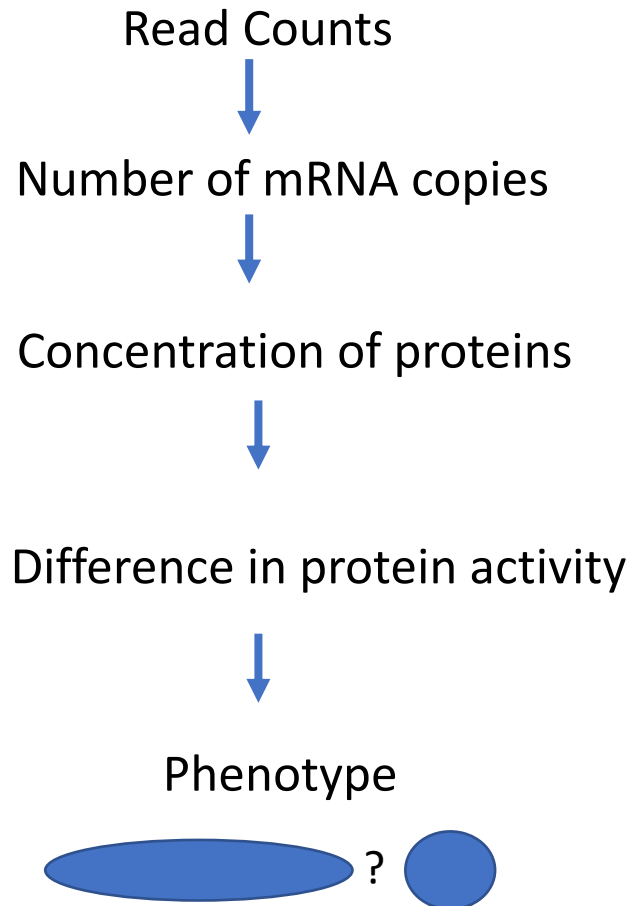
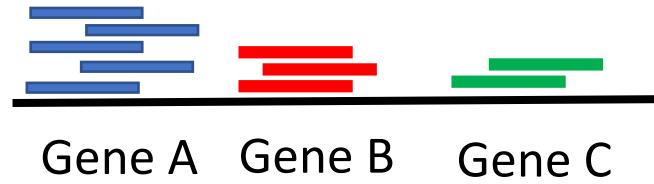
Why is differential expression useful?

We're looking for an explanation of observed phenotypes:

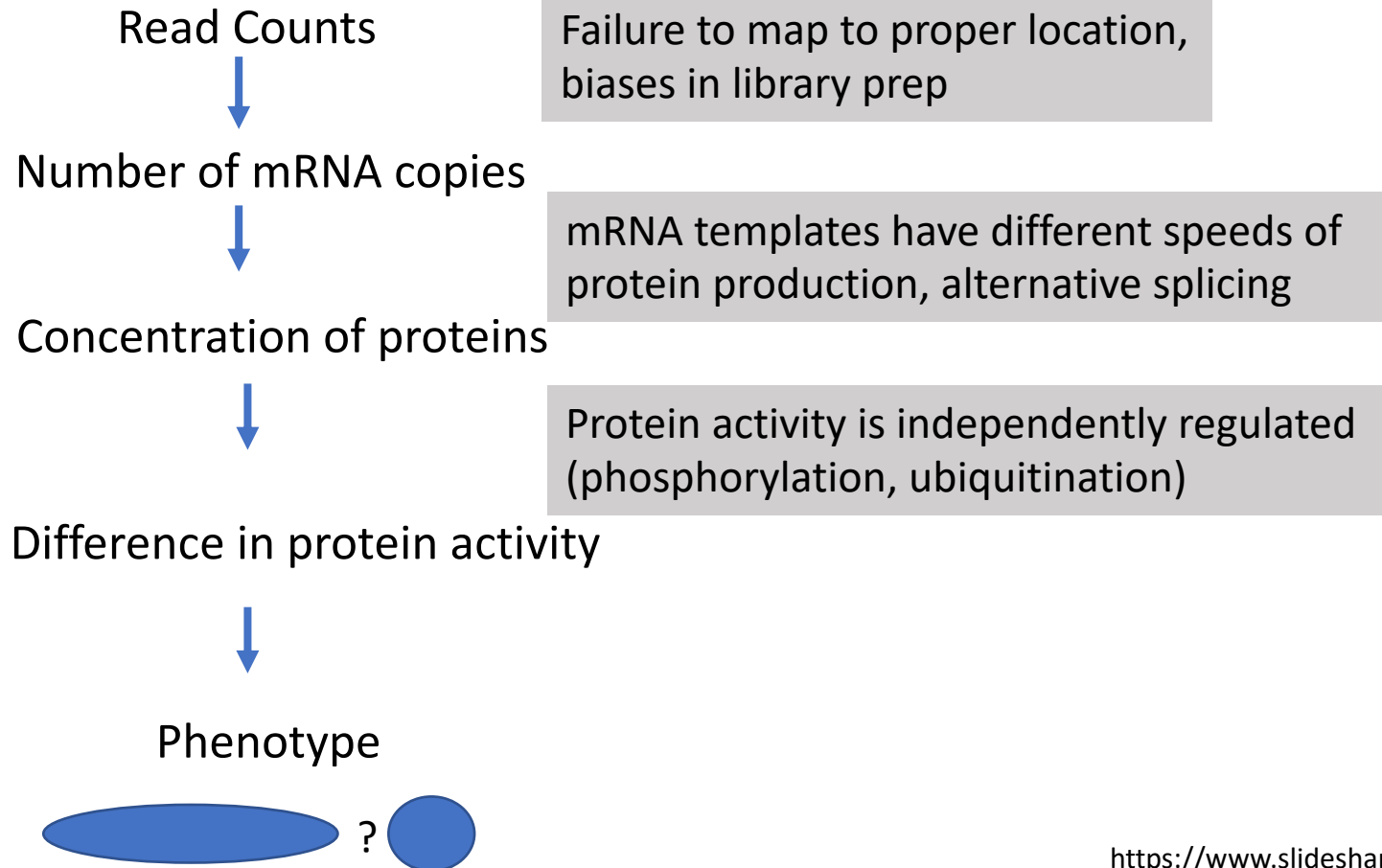
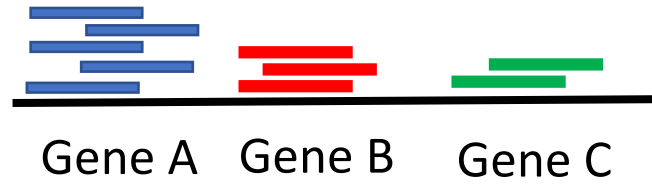


What causes difference in phenotype? Difference in protein activity!

mRNA is easier to measure than protein, so we use it as a proxy

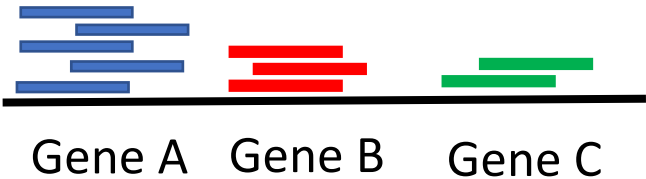


Though our assumptions about correlation are often violated

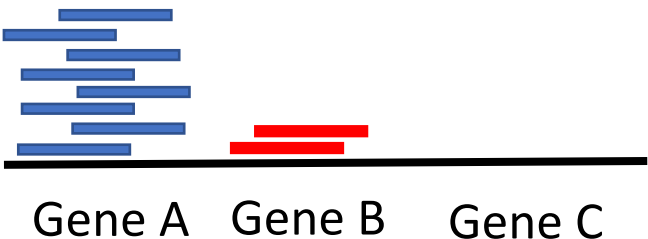


As a consequence, we look at comparisons

Wild Type 



Mutant 



The final test will look at ratios:

$6/8$ $3/2$ $2/0$

Experiment design

How deep to sequence? How many biological replicates to choose?

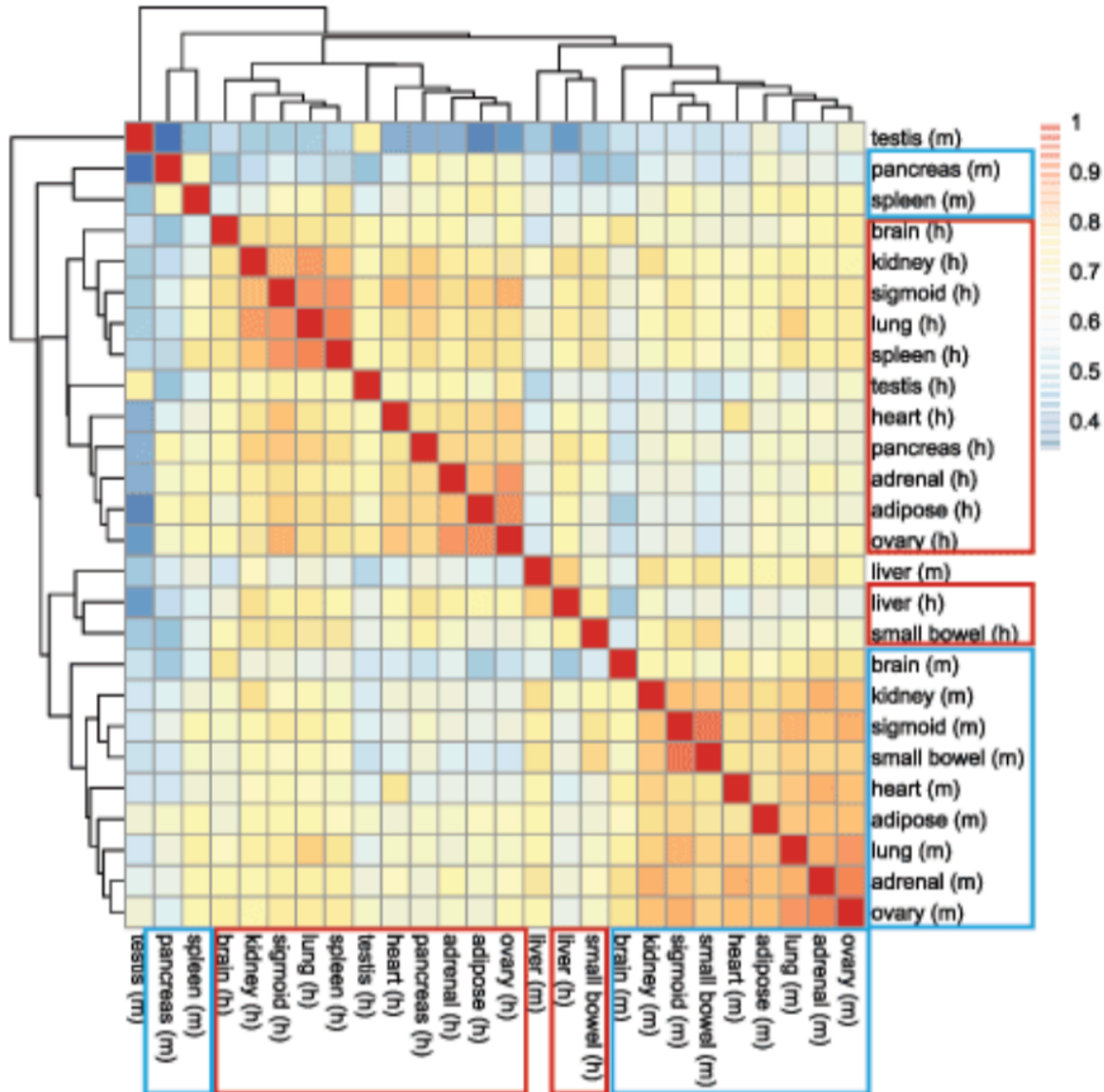
- Difficult to answer in general but certainly ≥ 3 replicates and ~ 20 M reads/replicate for strongly expressed genes
- Pilot studies are recommended to determine the number of replicates needed to capture the variability (e.g. 2 bio replicates, 10-20 M reads)
- Talk to the sequencing core!

Lessons from the mouse ENCODE study (2014)

This study was designed to test “the common notion that major developmental pathways are highly conserved across a wide range of species, in particular across mammals.”

How close are mouse and human in terms of gene expression across multiple tissues?

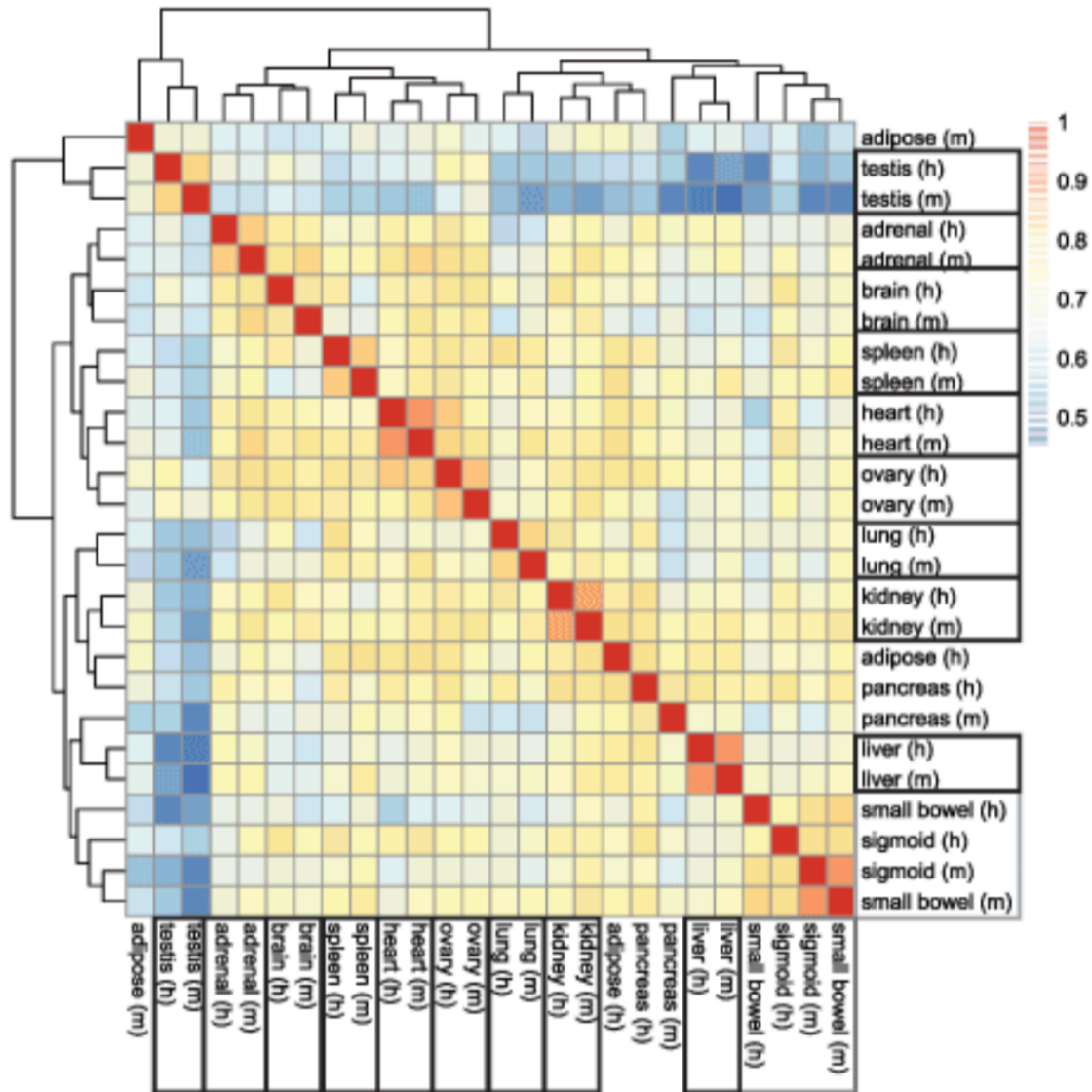
Initial publication showed mouse and human cluster separately



“Overall, our results indicate that there is considerable RNA expression diversity between humans and mice, well beyond what was described previously, likely reflecting the fundamental physiological differences between these two organisms.”

Lin, Lin, and Snyder (2014). PNAS 111:48

Once batch effects were accounted for: clustering by tissue



“Once we accounted for the batch effect (...), the comparative gene expression data no longer clustered by species, and instead, we observed a clear tendency for clustering by tissue.”

Gilad & Mizrahi-Man (2015). F1000Research 4:121

ENCODE* study design was not optimal

Most human samples were sequenced separately from the mouse samples:

D87PMJN1 (run 253, flow cell D2GUAACXX, lane 7)	D87PMJN1 (run 253, flow cell D2GUAACXX , lane 8)	D4LHBFN1 (run 276, flow cell C2HKJACXX , lane 4)	MONK (run 312, flow cell C2GR3ACXX , lane 6)	HWI-ST373 (run 375, flow cell C3172ACXX , lane 7)
heart	adipose	adipose	heart	brain
kidney	adrenal	adrenal	kidney	pancreas
liver	sigmoid colon	sigmoid colon	liver	brain
small bowel	lung	lung	small bowel	spleen
spleen	ovary	ovary	testis	● Human
testis		pancreas		● Mouse

Many tissues were not sex-matched

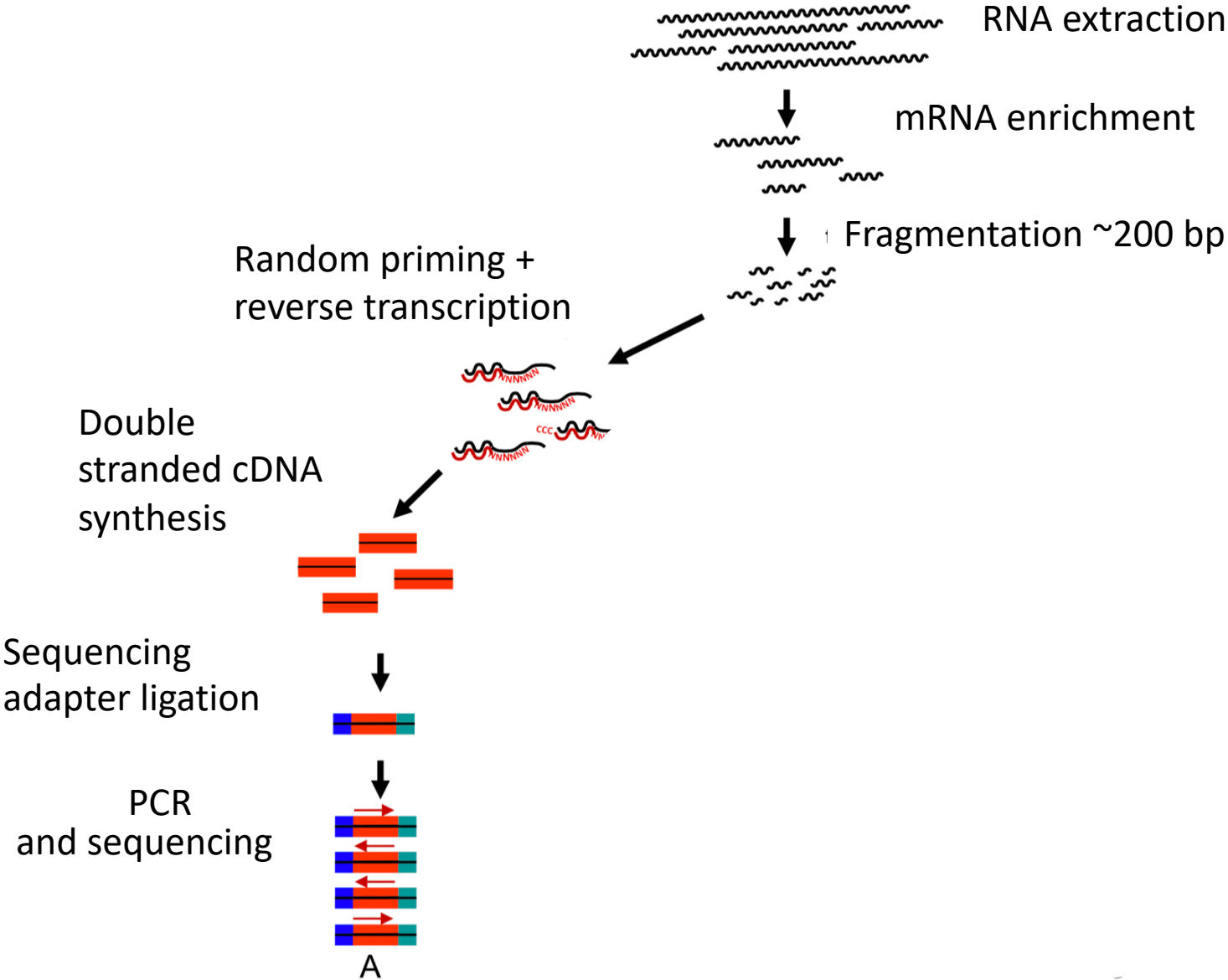
Tissue	Human	Mouse
adipose	FEMALE	MALE
adrenal	MALE	FEMALE
brain	FEMALE	MALE
heart	FEMALE	FEMALE
kidney	MALE	FEMALE
liver	MALE	FEMALE
lung	FEMALE	FEMALE
ovary	FEMALE	FEMALE
pancreas	FEMALE	FEMALE
sigmoid colo	MALE	FEMALE
small bowel	FEMALE	FEMALE
spleen	FEMALE	MALE
testis	MALE	MALE

- Avoid batch effects when possible!
- Account for unavoidable batch effects in your differential expression analysis.

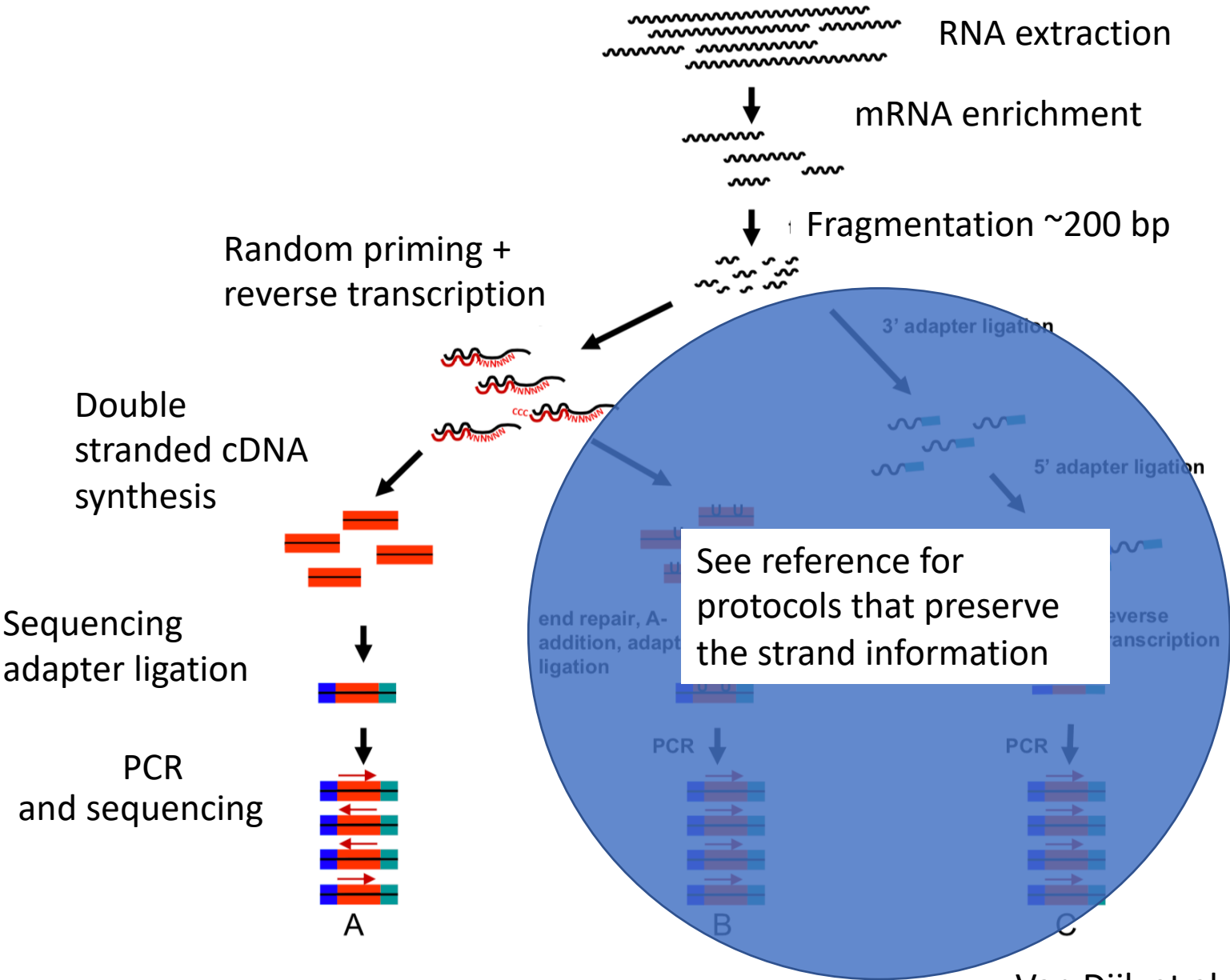
* Not just ENCODE! Good review! <https://f1000research.com/articles/4-121>

Credit: <http://chagall.med.cornell.edu/RNASEQcourse/>

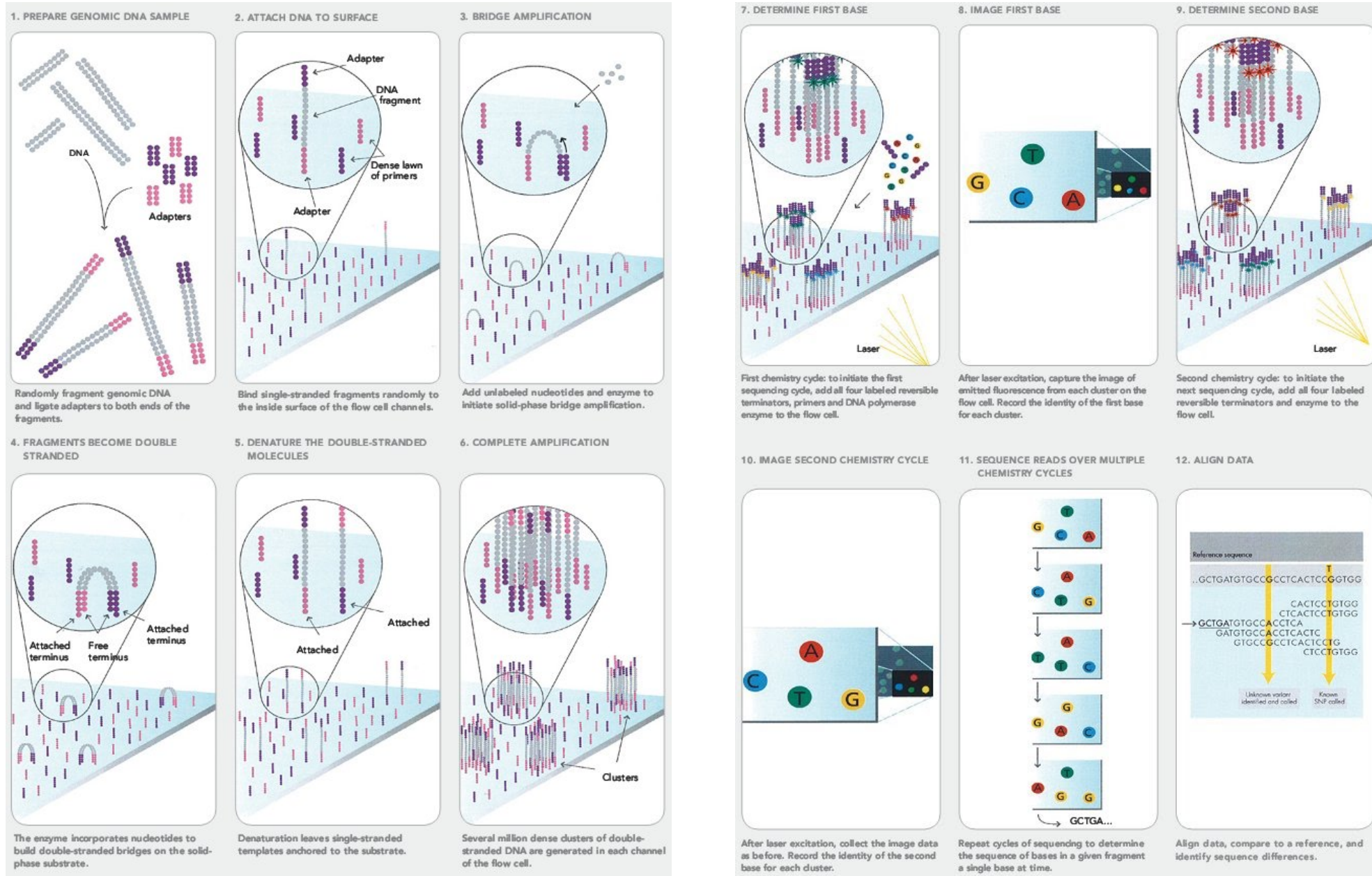
RNAseq Library Preparation and Sequencing (Classic Illumina)



RNAseq Library Preparation and Sequencing (Classic Illumina)



Next Generation Sequencing (NGS)



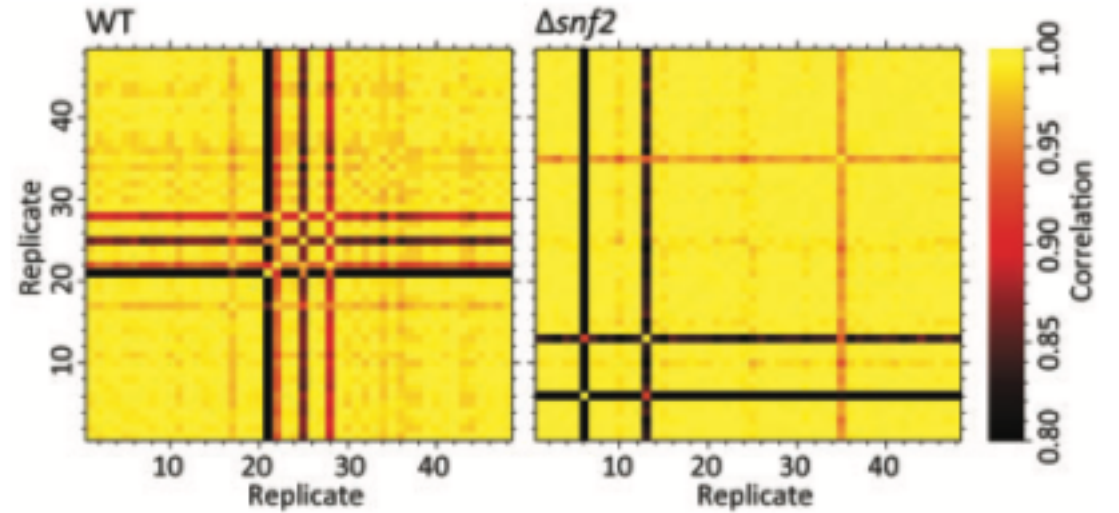
This [Illumina Video](#) is helpful for visualization!

Dataset for this course

“Statistical models for RNA-seq data derived from a two-condition 48-replicate experiment”

[Gierlinski et al Bioinformatics 2015](#)

- mRNA data from 48 biological replicates of two *Saccharomyces cerevisiae* populations
- Wildtype (WT) and SNF2 knock-out ($\Delta snf2$)
- Unusually comprehensive analysis of variability in sequencing replicates

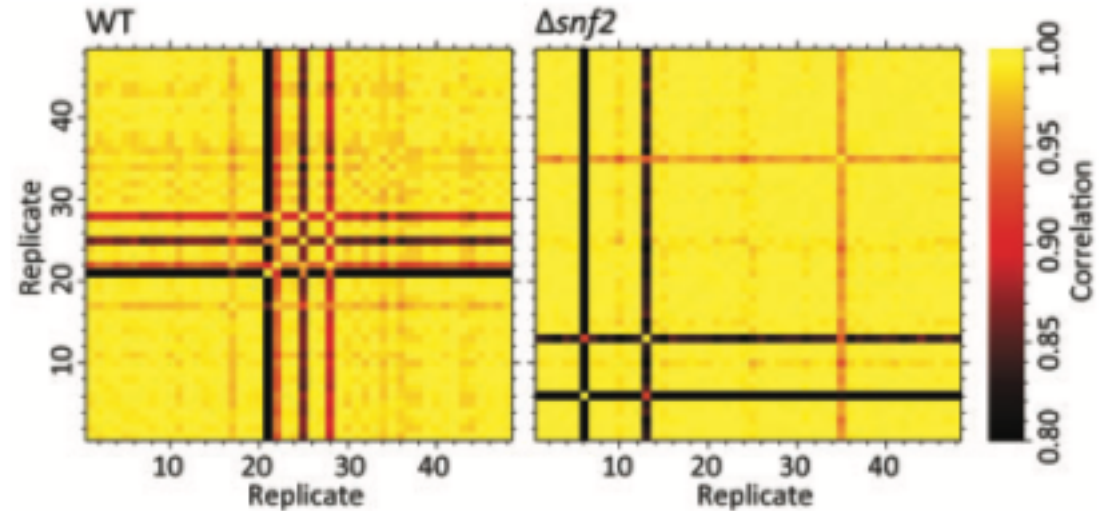


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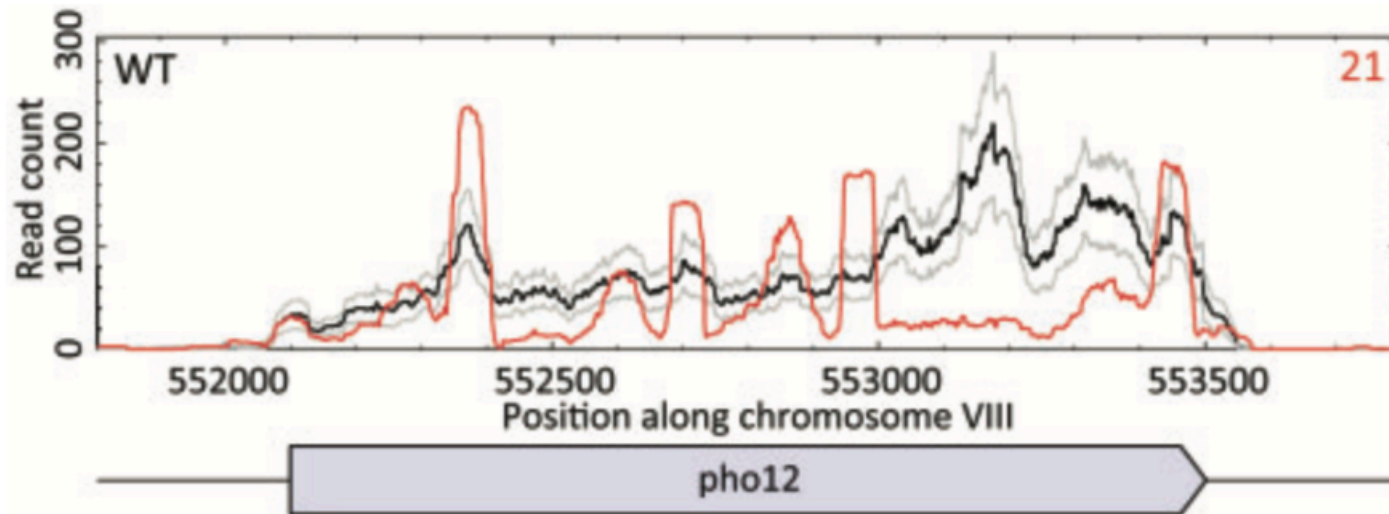
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- Unusually comprehensive analysis of variability in sequencing replicates



**Course dataset will consider 7 subsamples of one WT replicate and one SNF2 mutant, to demonstrate differences between populations and details of processing batches from different conditions

Invest in replicates!

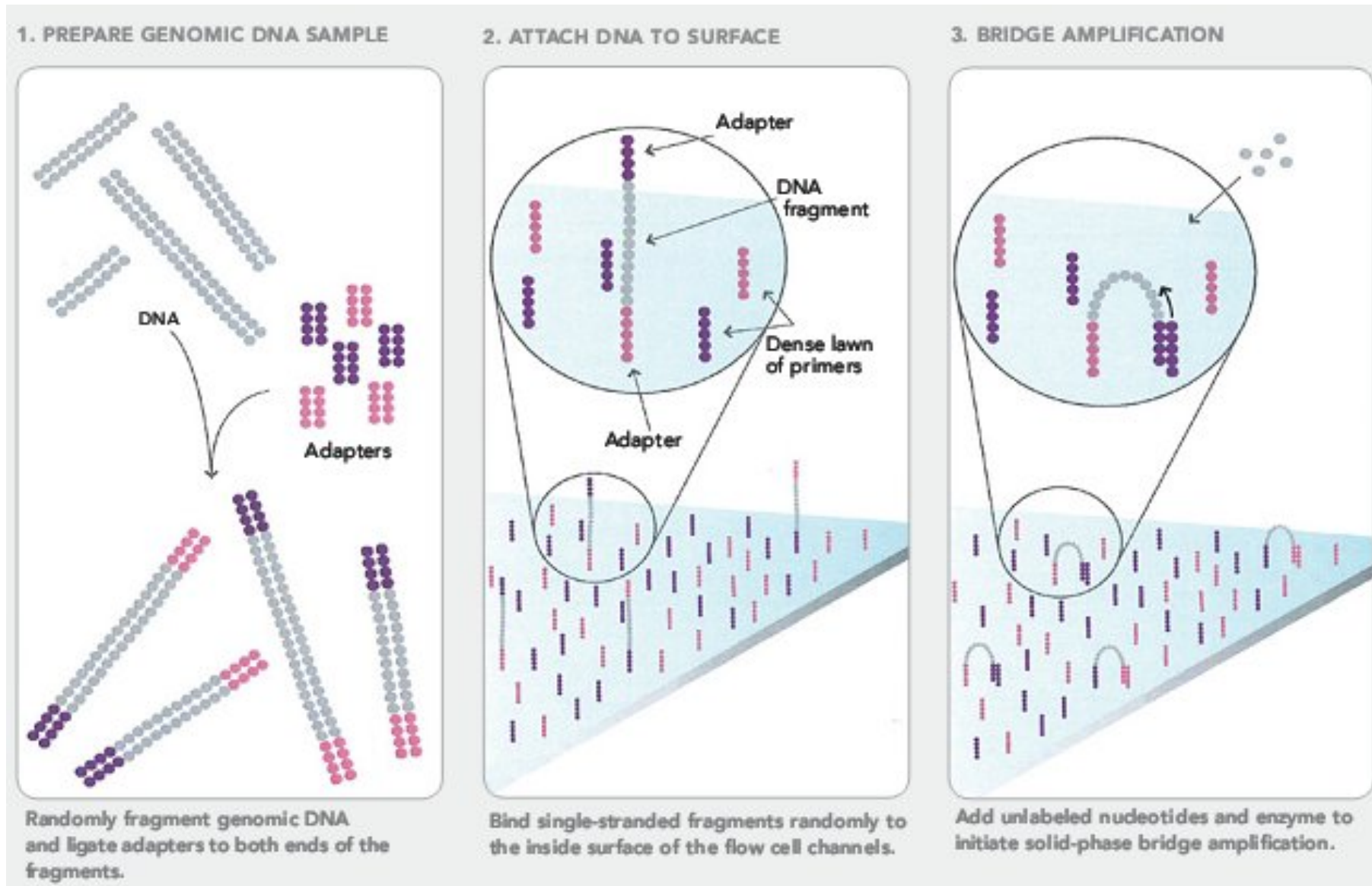
- The most effective way to improve detection of differential expression in low expression genes is to add more replicates, rather than adding more reads
- The following figure from **Gierlinski et al** shows coverage variation among replicates of a relatively simple yeast transcriptome (black is average of good replicates, grey is standard deviation)
- The paper concludes that we should invest in 6 **biological** replicates per condition



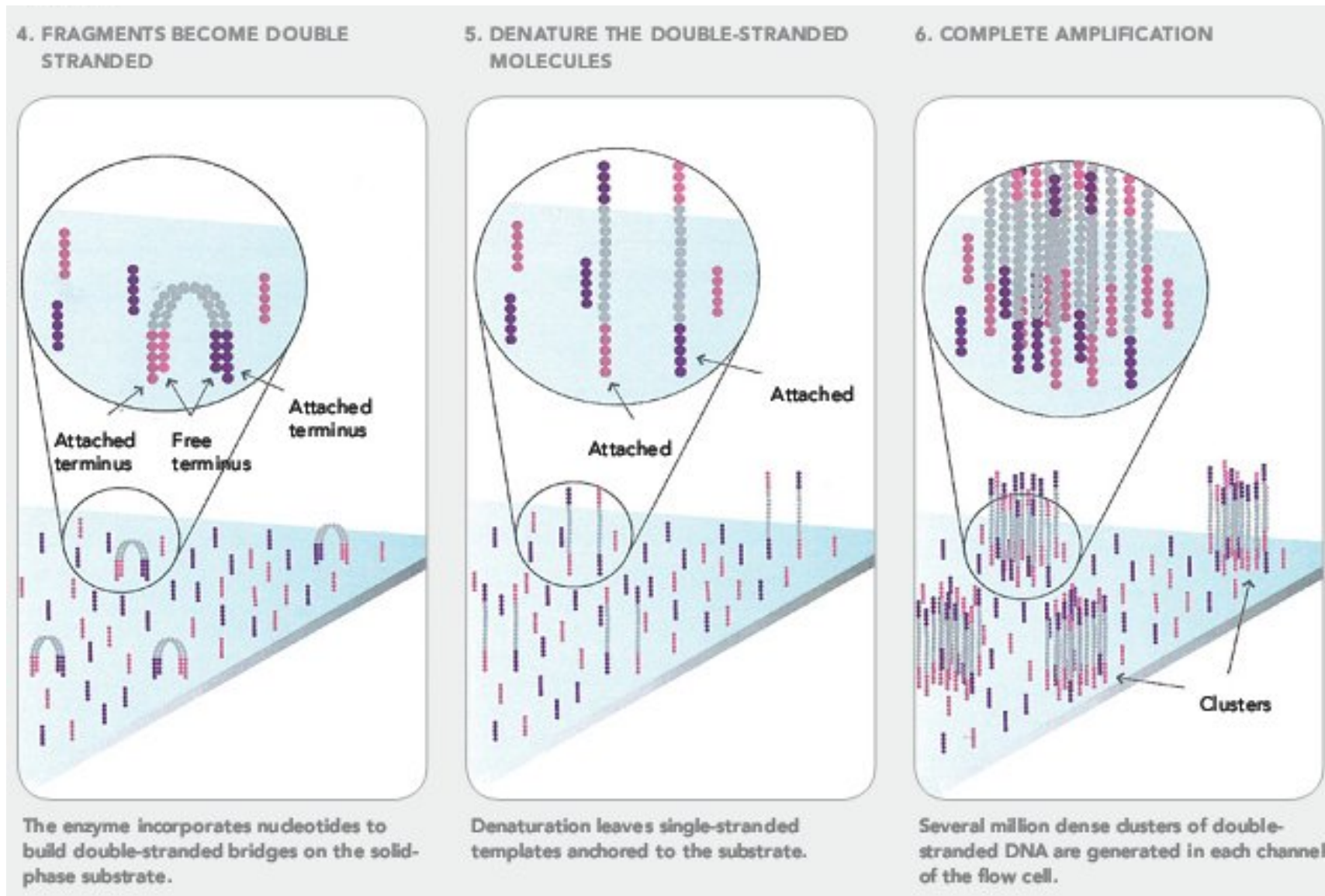
Gierlinski et al Bioinformatics 2015

<https://www.ncbi.nlm.nih.gov/pmc/articles/PMC4754627>

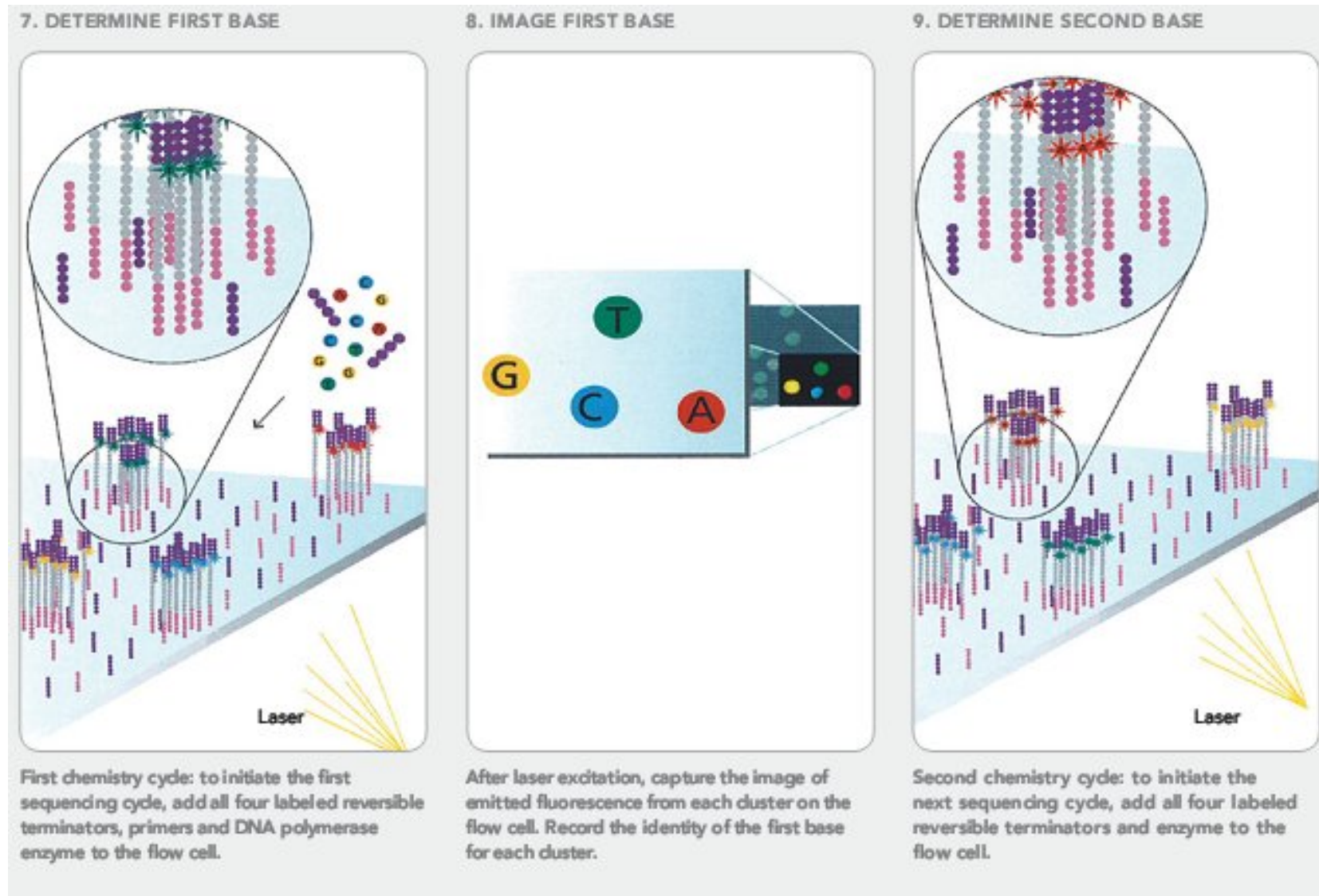
Next Generation Sequencing (NGS)



Next Generation Sequencing (NGS)

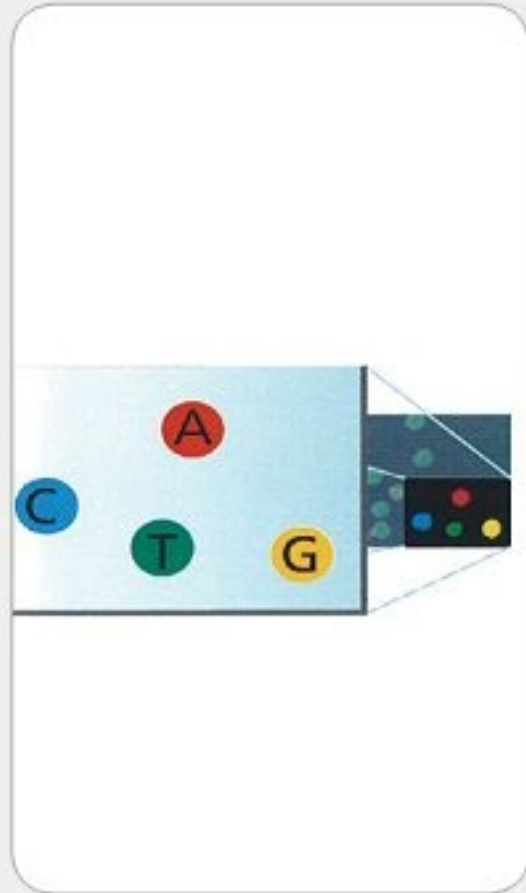


Next Generation Sequencing (NGS)



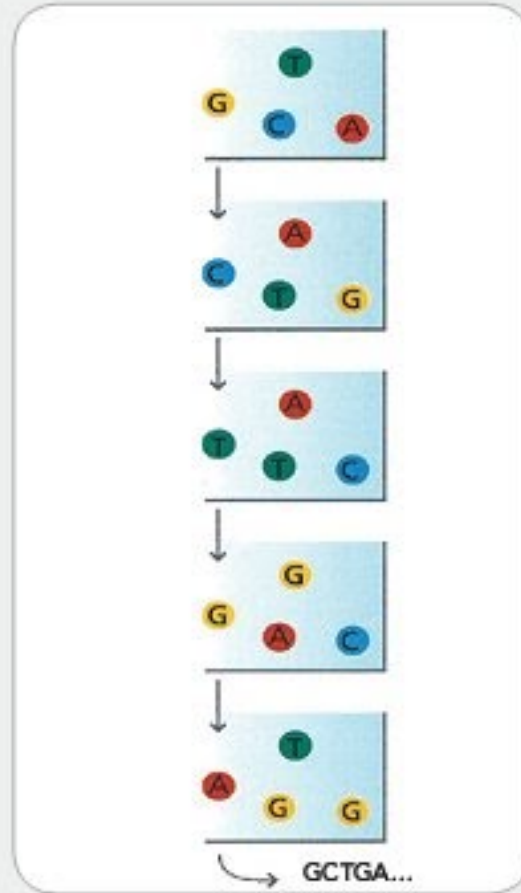
Next Generation Sequencing (NGS)

10. IMAGE SECOND CHEMISTRY CYCLE



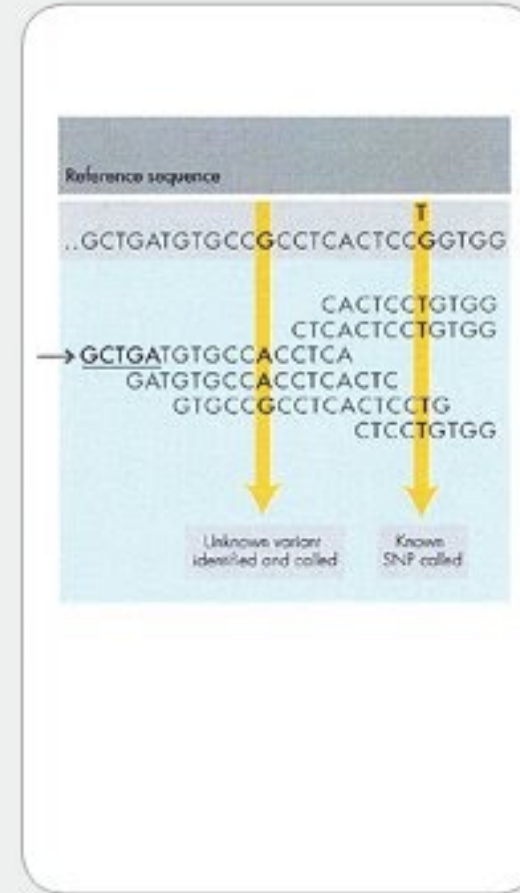
After laser excitation, collect the image data as before. Record the identity of the second base for each cluster.

11. SEQUENCE READS OVER MULTIPLE CHEMISTRY CYCLES



Repeat cycles of sequencing to determine the sequence of bases in a given fragment a single base at time.

12. ALIGN DATA



Align data, compare to a reference, and identify sequence differences.

RNAseq workflow

