

# Introduction to Bioinformatics

Cambridge Makerere Summer School 2024

Ashley D Sawle



Together we will beat cancer



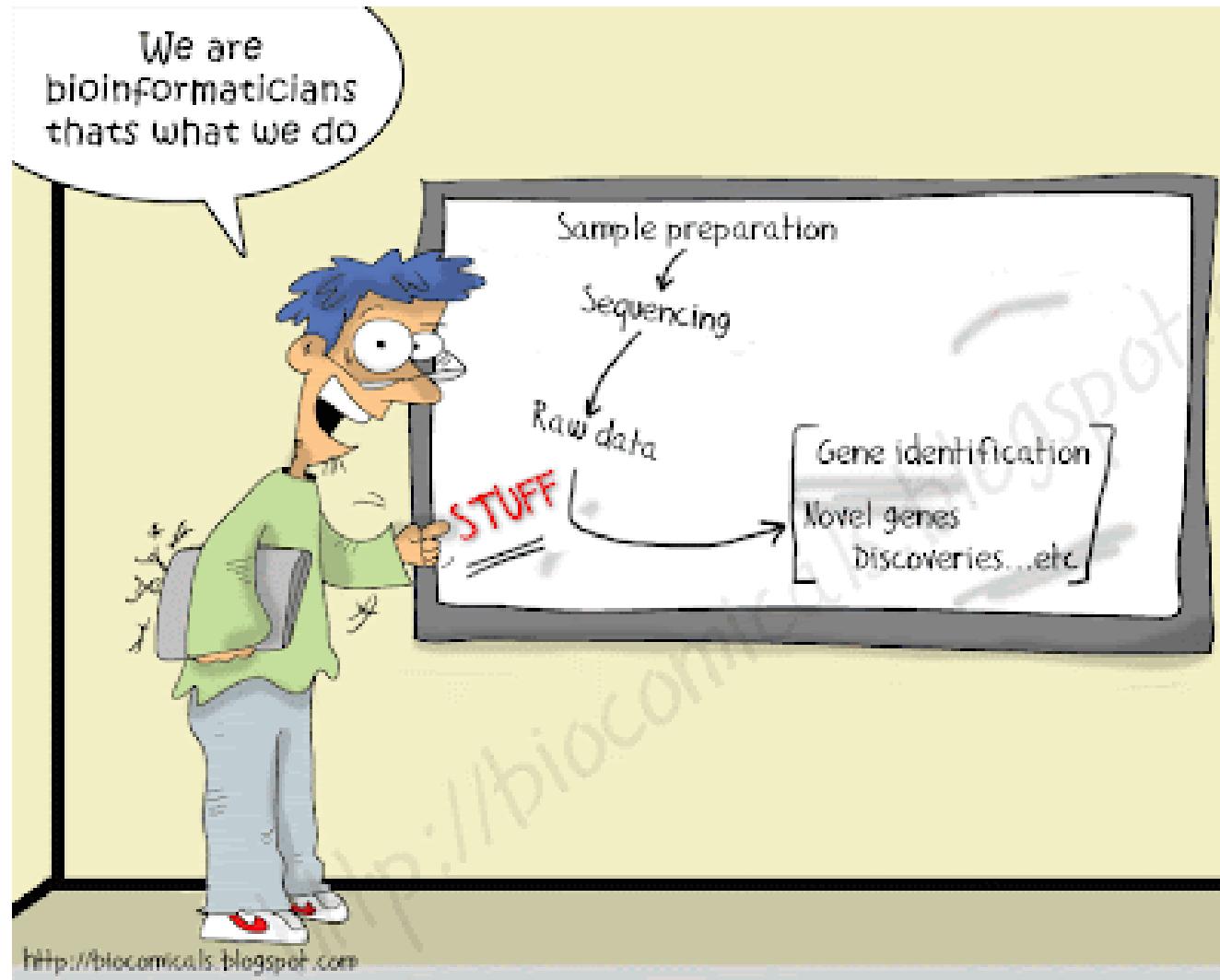
# Overview

- A Brief Overview of Bioinformatics
- Bioinformatic Analysis of Next Generation Sequencing Data

# Overview

- **A Brief Overview of Bioinformatics**
- Bioinformatic Analysis of Next Generation Sequencing Data

# What is Bioinformatics?



# What is Bioinformatics?

- Bioinformatics is a relatively new and evolving discipline that combines skills and technologies from computer science and biology to help us better understand and interpret biological data.
- Bioinformatics, as related to genetics and genomics, is a scientific subdiscipline that involves using computer technology to collect, store, analyze and disseminate biological data and information
- The mathematical, statistical and computing methods that aim to solve biological problems using DNA and amino acid sequences and related information.

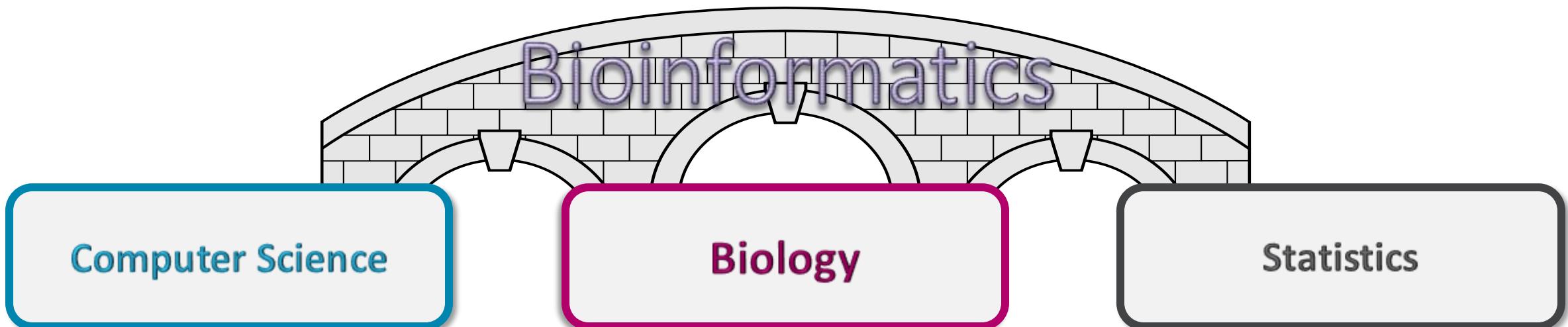
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# What is Bioinformatics?



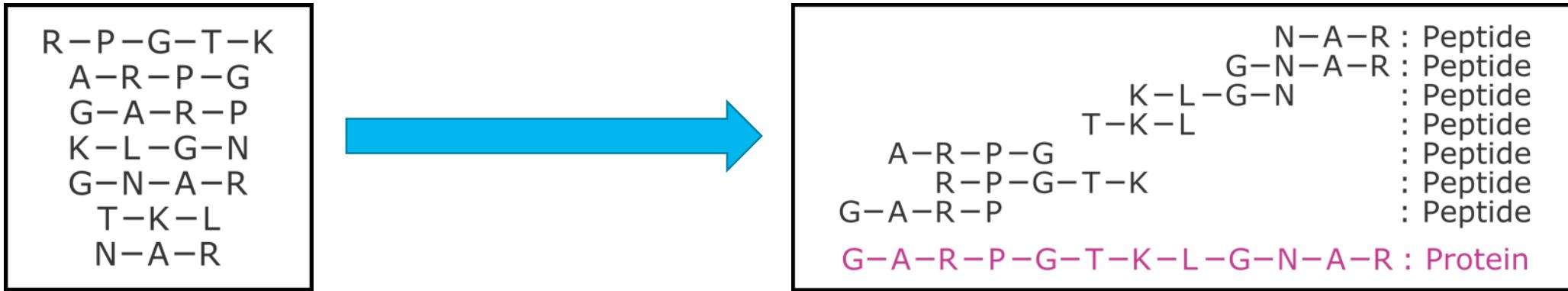
Bioinformatics is an **interdisciplinary** field of science that develops methods and software tools for understanding biological data, especially when the data sets are large and complex.



# It all started with proteins

- 1951 – Frederick Sanger sequenced the amino acid structure of insulin

# It all started with proteins



# It all started with proteins ...

- 1951 – Frederick Sanger sequenced the amino acid structure of insulin
- 1962 – Margaret Dayhoff and Robert Ledley publish COMPROTEIN
- 1965 – Margaret Dayhoff published the book “Atlas of Protein Sequence and Structure”
- 1970 – Needleman-Wunsch algorithm for sequence alignment published
- 1970s – Peter Chou and Gerald Fasman develop first protein structure prediction algorithm
- 1971 – The Protein Data Bank

# ... and it continues with proteins

The Nobel Prize in Chemistry 2024

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David Baker



“for computational protein design”

Demis Hassabis



“for protein structure prediction”

John Jumper



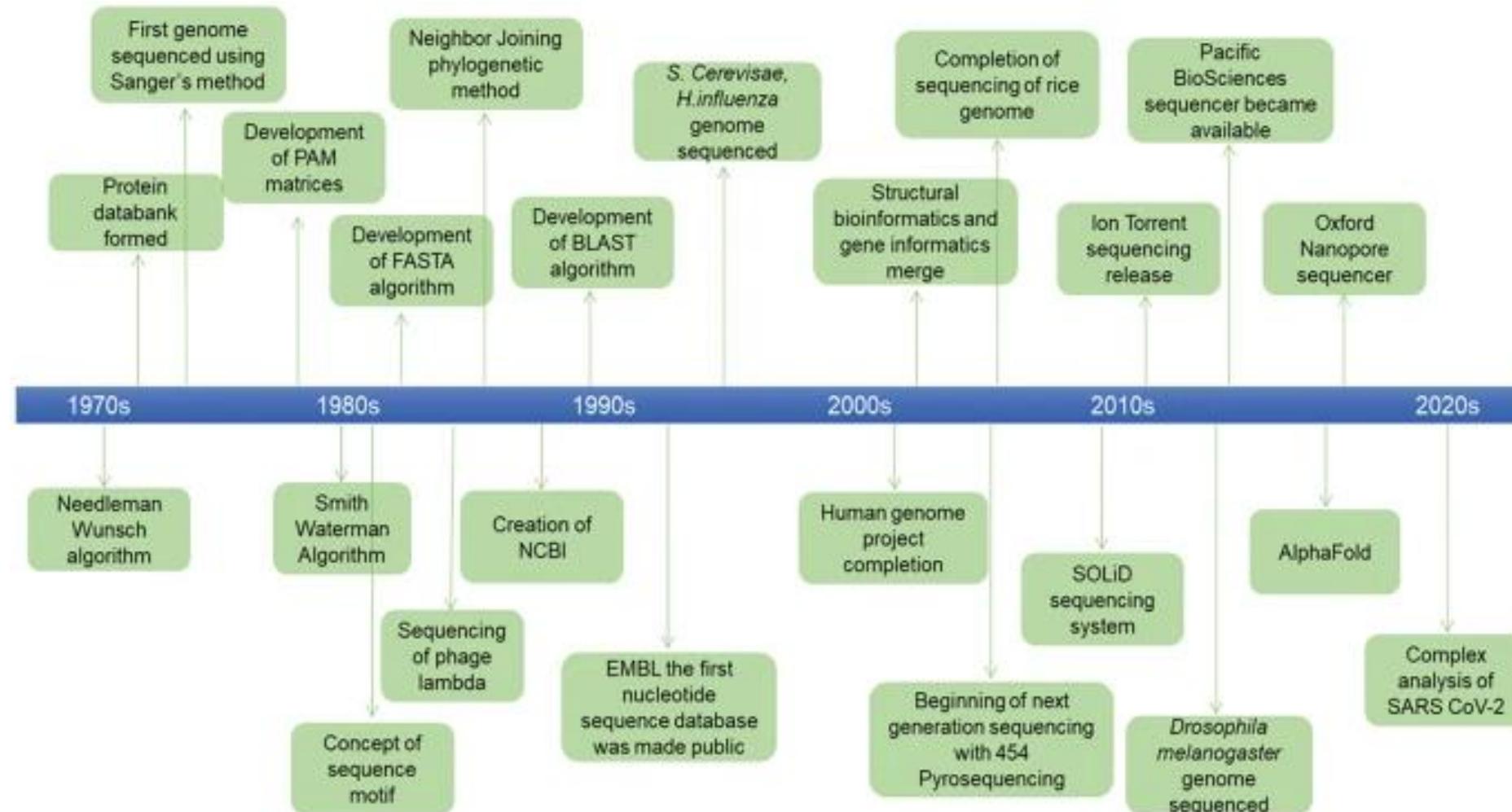
“for protein structure prediction”

David Baker. Ill. Niklas Elmehed © Nobel Prize Outreach

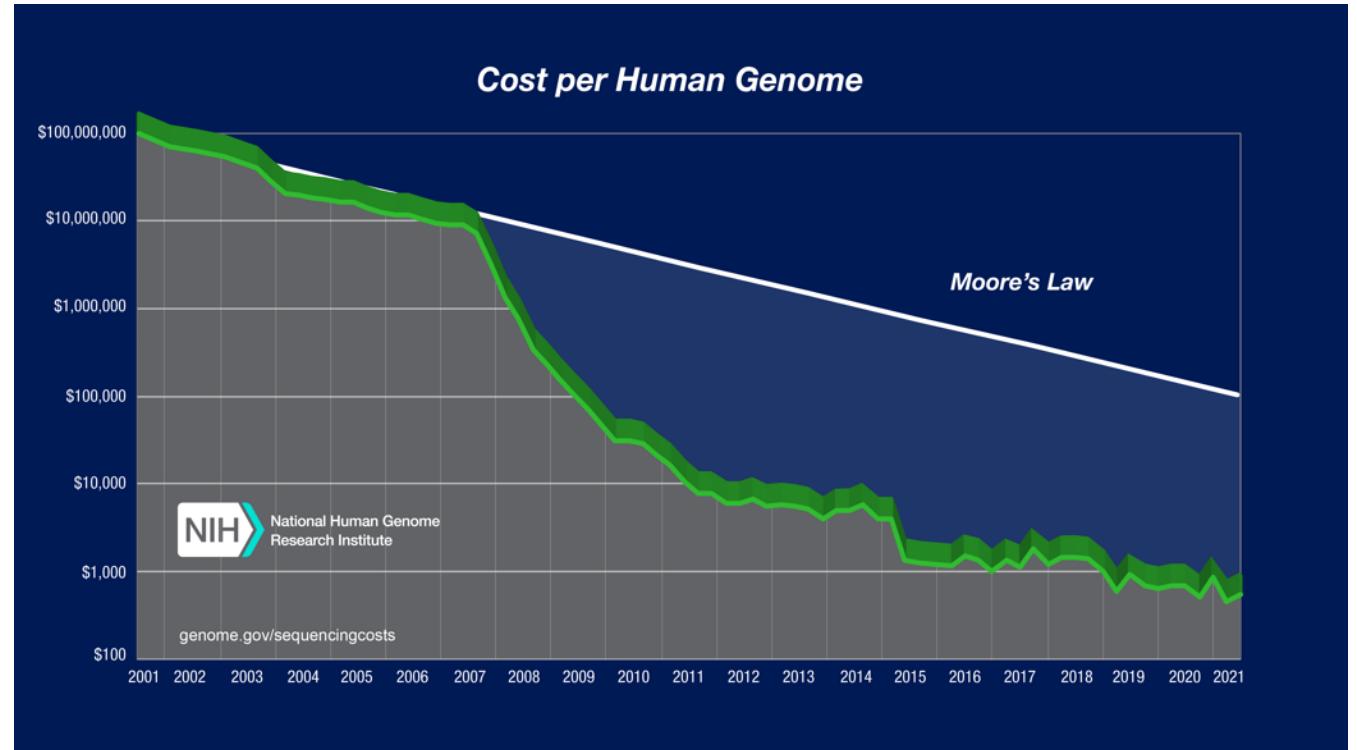
Demis Hassabis. Ill. Niklas Elmehed © Nobel Prize Outreach

John Jumper. Ill. Niklas Elmehed © Nobel Prize Outreach

# A history of nucleotide sequencing



# The era of big data



# The era of big data



# Data storage and Access

- Verification of results presented in papers requires access to the data used to generate them
- Data generated in one study can still be useful to other
- Open Data rather than siloed inaccessible data

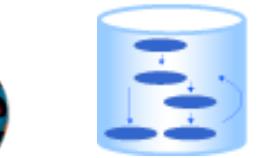
[Open Access](#) | Published: 15 March 2016

## The FAIR Guiding Principles for scientific data management and stewardship

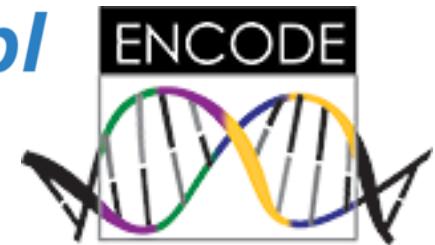
[Mark D. Wilkinson](#), [Michel Dumontier](#), [IJsbrand Jan Aalbersberg](#), [Gabrielle Appleton](#), [Myles Axton](#), [Arie Scientific Data](#) 3, Article number: 160018 (2016) | [Cite this article](#)

Findable Accessible Interoperable Reusable

# Data Repositories



Genome Aggregation Database



GTCGGCAATCCAAAGATAGCCAAATATTATTATTGTTAGATACTCAC  
AGCAGCAAGTGCAGATGCCCTTGAGTGTTGGAAATCAGTGAAATT  
ATTTGGCGAACGGGCTTCTTGGGTTAATGTTGTTGTTGTTGTTGTT  
ATTAATTAACCAAACAGTGTGAACACAGCAGGGCATCTTCATAGA



# A word on gene names

The screenshot shows the Ensembl gene page for the gene **FLT4** (ENSG00000037280). The top navigation bar includes links for BLAST/BLAT, VEP, Tools, BioMart, Downloads, Help & Docs, and Blog. The species is set to Human (GRCh38.p14). The location is 5:180,601,506-180,649,624. The main content area displays the gene's description as fms related receptor tyrosine kinase 4 [Source:HGNC Symbol;Acc:[HGNC:3767](#)], gene synonyms PCL, VEGFR-3, VEGFR3, and its location on Chromosome 5. A sidebar on the left provides gene-based displays for Summary, Sequence, and Comparative Genomics.

**Gene: FLT4 ENSG00000037280**

**Description**: fms related receptor tyrosine kinase 4 [Source:HGNC Symbol;Acc:[HGNC:3767](#)]

**Gene Synonyms**: PCL, VEGFR-3, VEGFR3

**Location**: [Chromosome 5: 180,601,506-180,649,624](#) reverse strand.  
GRCh38:CM000667.2

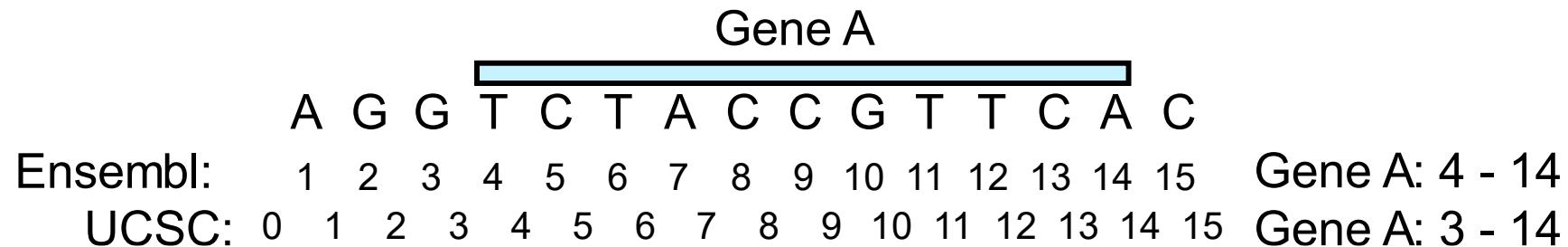
**Gene-based displays**

- Summary
  - Splice variants
  - Transcript comparison
  - Gene alleles
- Sequence
  - Secondary Structure
- Comparative Genomics
  - Genomic alignments

# Genomic Sequence Data

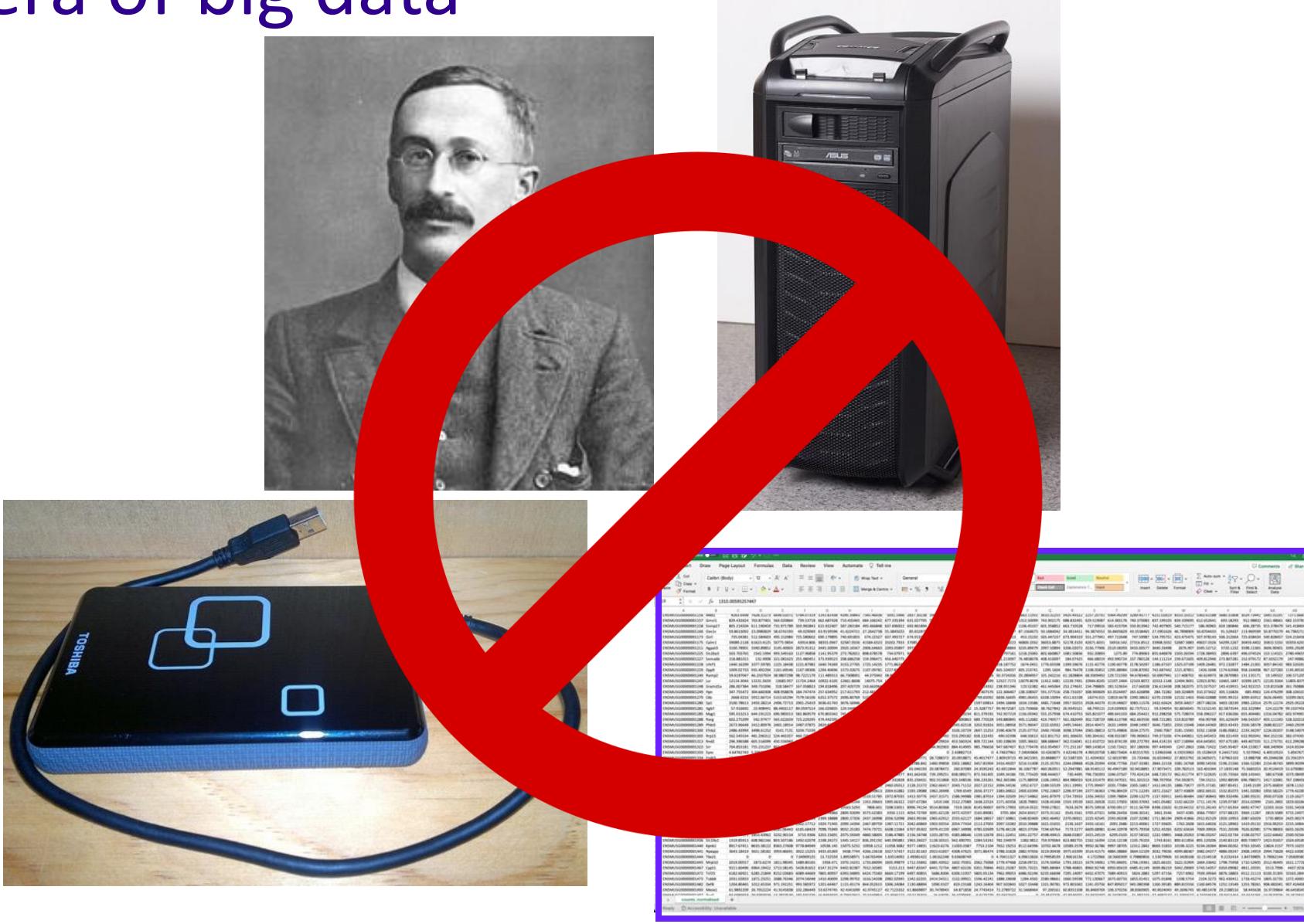
UCSC/NCBI versus Ensembl/Gencode/EBI

- Ensembl uses a one-based coordinate system, whereas UCSC uses a zero-based coordinate system.



- Ensembl/Gencode name sequences : 1, 2, 3 ... 22, X, Y, MT
- UCSC/NCBI name sequences: chr1, chr2, chr3 ... chr22, chrY, chrX, chrM
- Gene annotations differ significantly
- Gene IDs are different and do not map 1:1

# The era of big data



# Programming languages in Bioinformatics

- Bioinformatics creates huge quantities of data, and programming gives the means to analyse and interpret that data.
- The two most popular languages are **Python** and **R**
- Both are open source meaning they are freely available
- Both have large communities of users and developers
- Both have a wide range of bioinformatics resources and methods
- Both are cross-platform
- Others: Perl, Java, Ruby, Rust, Julia



- R is a language and environment for **statistical computing and graphics**
- R is available as Free Software under the terms of the Free Software Foundation's **GNU General Public License**
- **RStudio** provides a well developed integrated development environment (IDE) for R
- The Comprehensive R Archive Network (**CRAN**) repository features 19877 available general usage packages
- The **Bioconductor** project has 2230 bioinformatics specific packages





- Python is a **general-purpose**, open-source programming language used in various software domains, including data science, web development, and gaming
- Python is developed under an **OSI-approved open source license**, making it freely available
- Various IDEs are available, e.g. **Jupyter** or Spyder
- 100,000s of packages available via the **Python Package Index**
- Virtual environments easily managed with **Conda**
- Python is more suitable for **deep learning** applications

**BIOCONDA**<sup>®</sup>



# Excel tries to be helpful

Comment | [Open Access](#) | Published: 23 August 2016

## Gene name errors are widespread in the scientific literature

[Mark Ziemann](#), [Yotam Eren](#) & [Assam El-Osta](#) 

[Genome Biology](#) 17, Article number: 177 (2016) | [Cite this article](#)

NEWS | 13 August 2021 | Correction [25 August 2021](#)

## Autocorrect errors in Excel still creating genomics headache

Despite geneticists being warned about spreadsheet problems, 30% of published papers contain mangled gene names in supplementary data.

The Verge SCIENCE / TECH / MICROSOFT

## Scientists rename human genes to stop Microsoft Excel from misreading them as dates



Illustration by Alex Castro / The Verge

/ Sometimes it's easier to rewrite genetics than update Excel

By [James Vincent](#), a senior reporter who has covered AI, robotics, and more for eight years at The Verge.

Aug 6, 2020, 1:44 PM GMT+1 | □ 0 Comments / 0 New



If you buy something from a Verge link, Vox Media may earn a commission. [See our ethics statement.](#)

There are tens of thousands of genes in the human genome: minuscule twists of DNA and RNA that combine to express all of the traits and characteristics that make each of us unique. Each gene is given a name

MARCH1 → MARCHF1  
SEPT1 → SEPTIN1

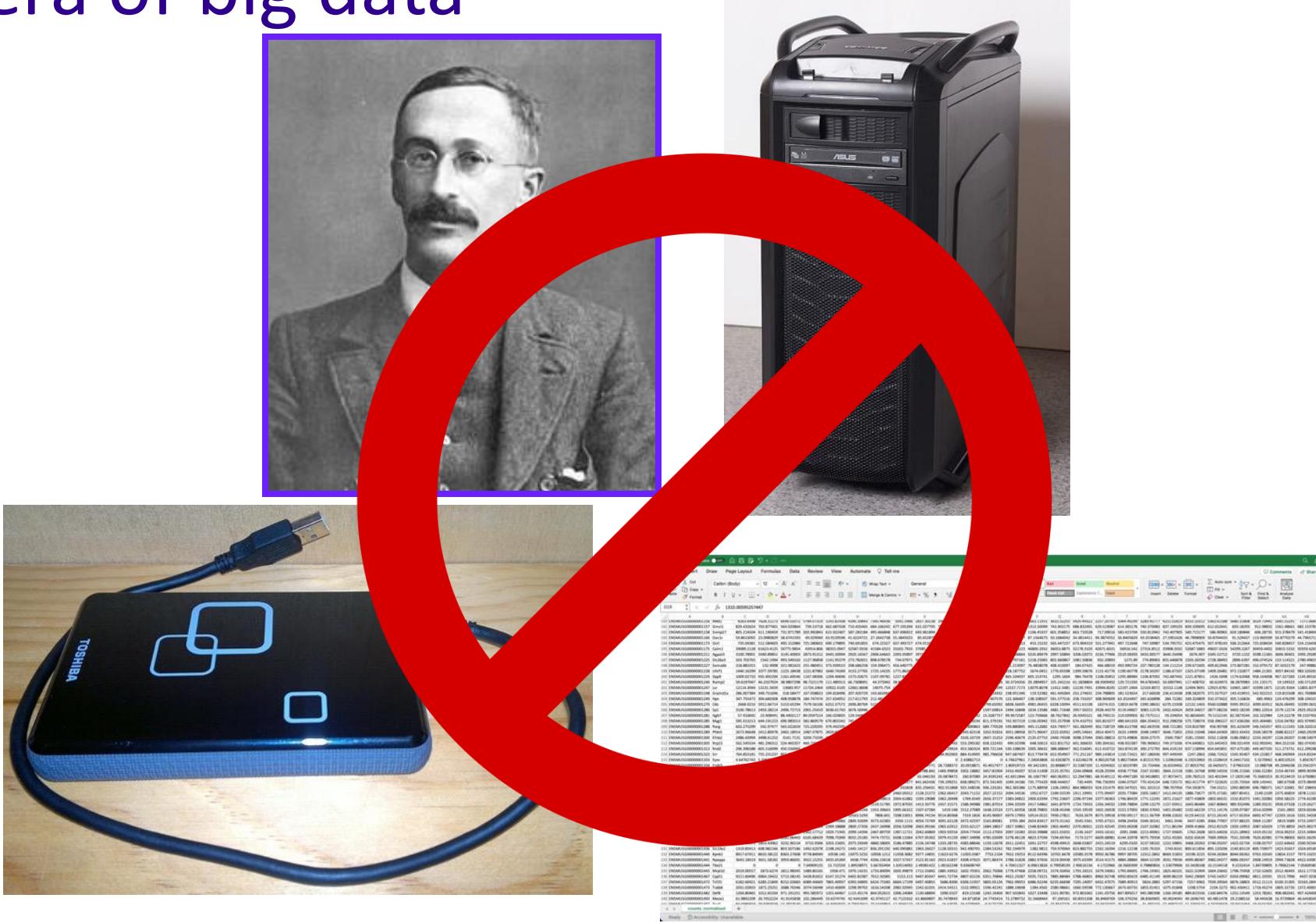
# The era of big data



# The need for more power

- Massive data sets require a lot memory to store and process
- Complex algorithms such as alignment need powerful processors to run in a reasonable time frame
- Having a lot of processors available means that many jobs can be run in parallel
- High-Performance Cluster – HPC (supercomputer)
- Cloud solution – Google Cloud, Amazon Web Services (AWS), Microsoft Azure
- Graphics Processing Units required for Deep Learning

# The era of big data



# Differing statistical approaches

- Single/Few measures in a simple experimental design – t-test
- More complex studies – Linear Model
  - Micro-arrays – Simple Linear Model with Normal Distribution
  - RNAseq data – Generalised Linear Model with Negative Binomial Distribution
- 10X Single Cell RNAseq – needs additional solutions to overcome missing data

# Complex Data Requires New Solutions

- Sequence alignment algorithms
- Clustering algorithms
- Hidden Markov Models (HMMs)
- Phylogenetic tree construction algorithms
- Molecular modelling algorithms
- Variant/Mutation calling algorithms
- Deep learning, large language models and machine learning

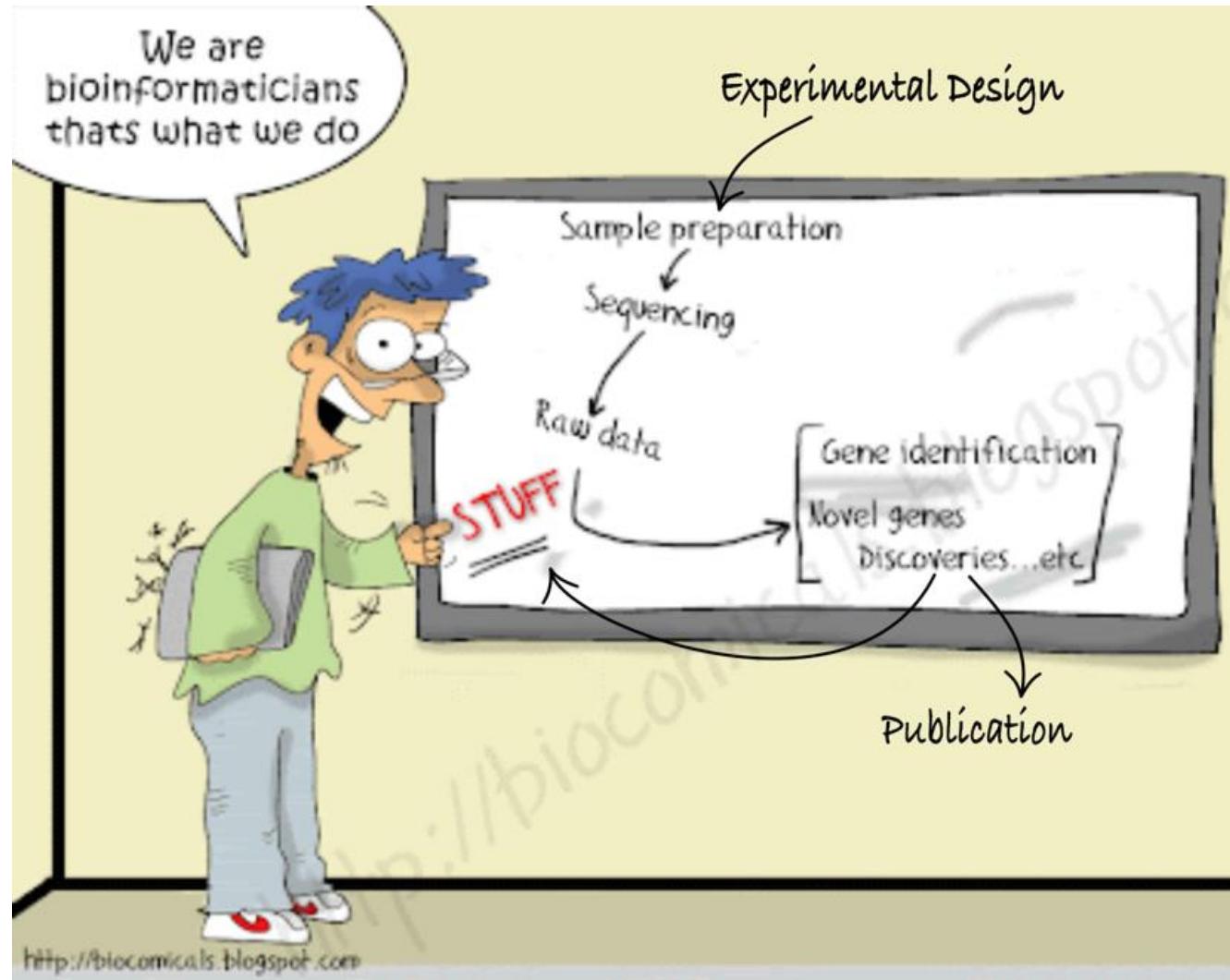
# Overview

- A Brief Overview of Bioinformatics
- **Bioinformatic Analysis of Next Generation Sequencing Data**

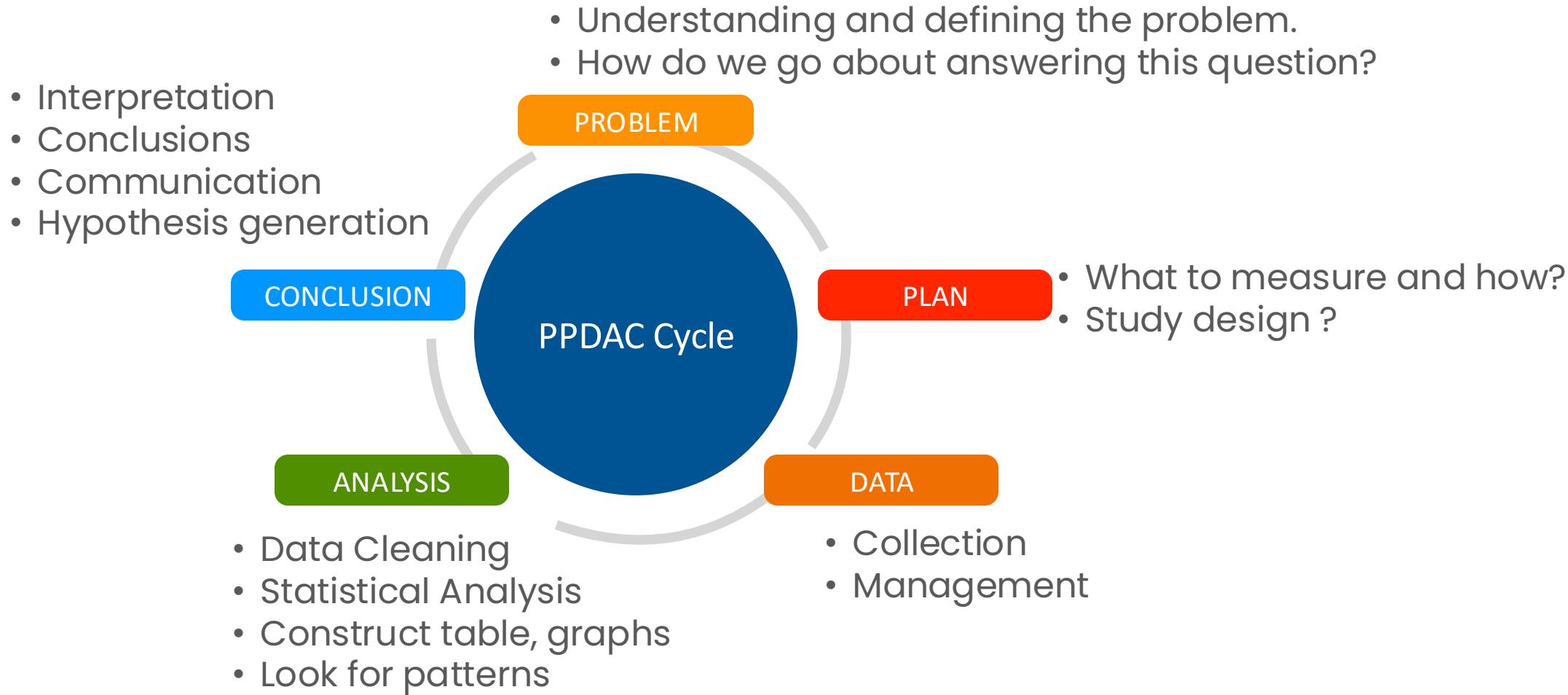
# Data literacy

The ability to not only carry out statistical analysis on real-world problems, but also to understand and critique any conclusions drawn by others on the basis of statistics.

# What we do...



# The PPDAC cycle



# Consequences of Poor Experimental Design

## **Inability to answer the questions we would like to answer**

- Cost of experimentation.
- Limited & Precious material, esp. clinical samples.
- Immortalization of data sets in public databases and methods in the literature. Our bad science begets more bad science.
- Ethical concerns of experimentation: animals and clinical samples.

# A Well-Designed Experiment

**Should have:**

- Clear objectives
- Focus and simplicity
- Sufficient power
- Randomised comparisons

**And be:**

- Precise
- Unbiased
- **Amenable to statistical analysis**
- Reproducible

# Sources of Variation

dependent variable =  $f$  ( independent variable ) + noise



## Biological “noise”:

- Biological processes are inherently stochastic
- Single cells, cell populations, individuals, organs, species....
- Timepoints, cell cycle, synchronized vs. unsynchronized

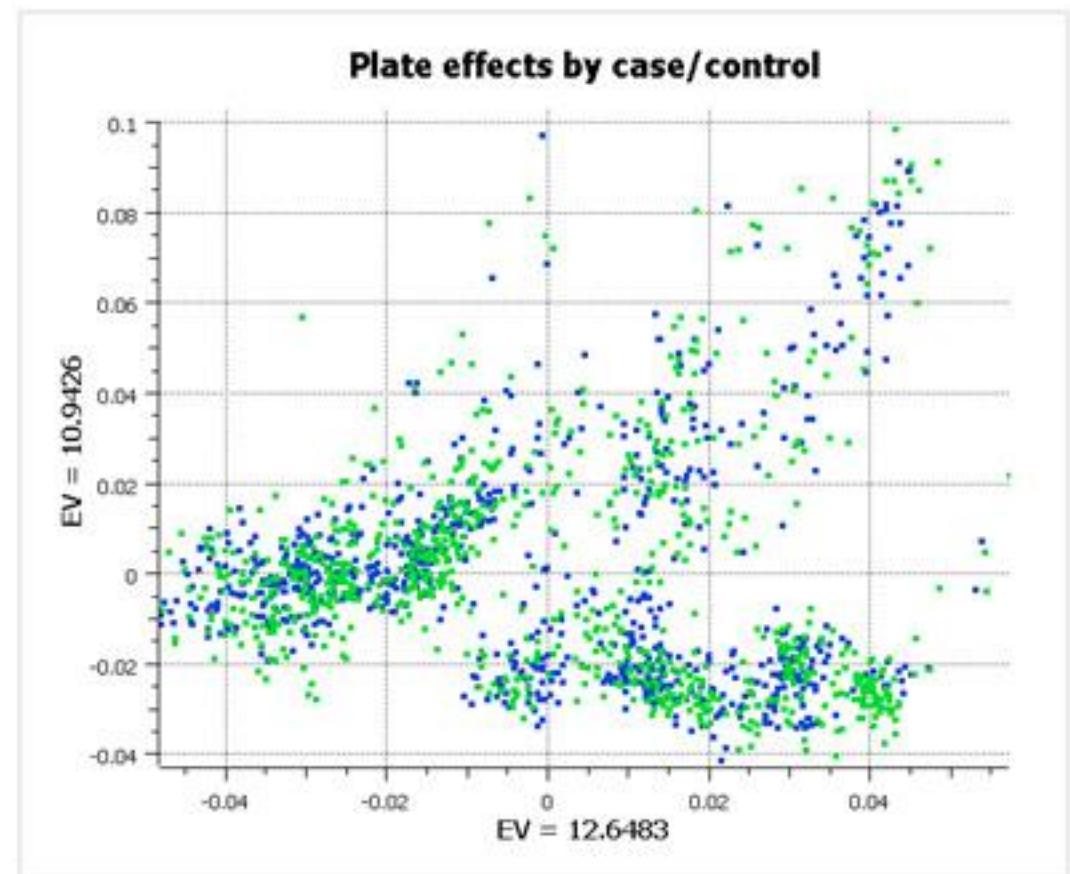
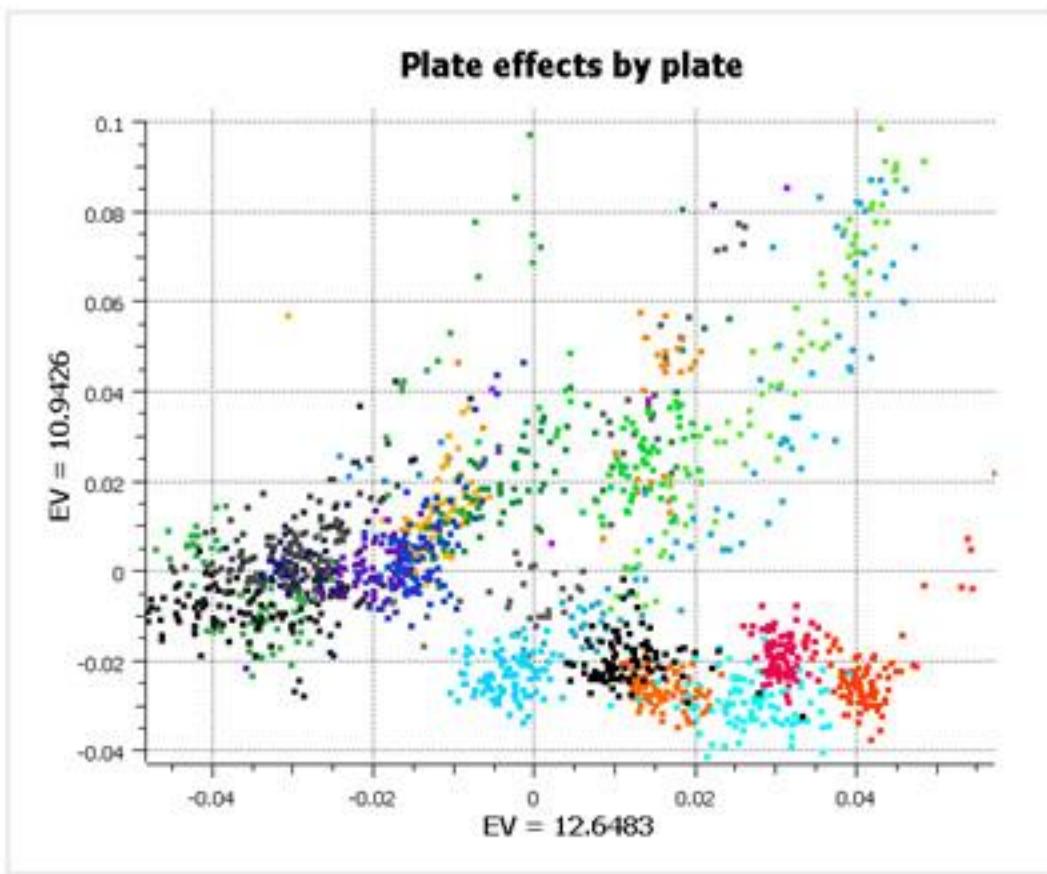
## Technical noise:

- Reagents, antibodies, temperatures, pollution
- Platforms, runs, operators

Replication is required to capture variance

Randomisation overcomes technical variation

# Randomisation

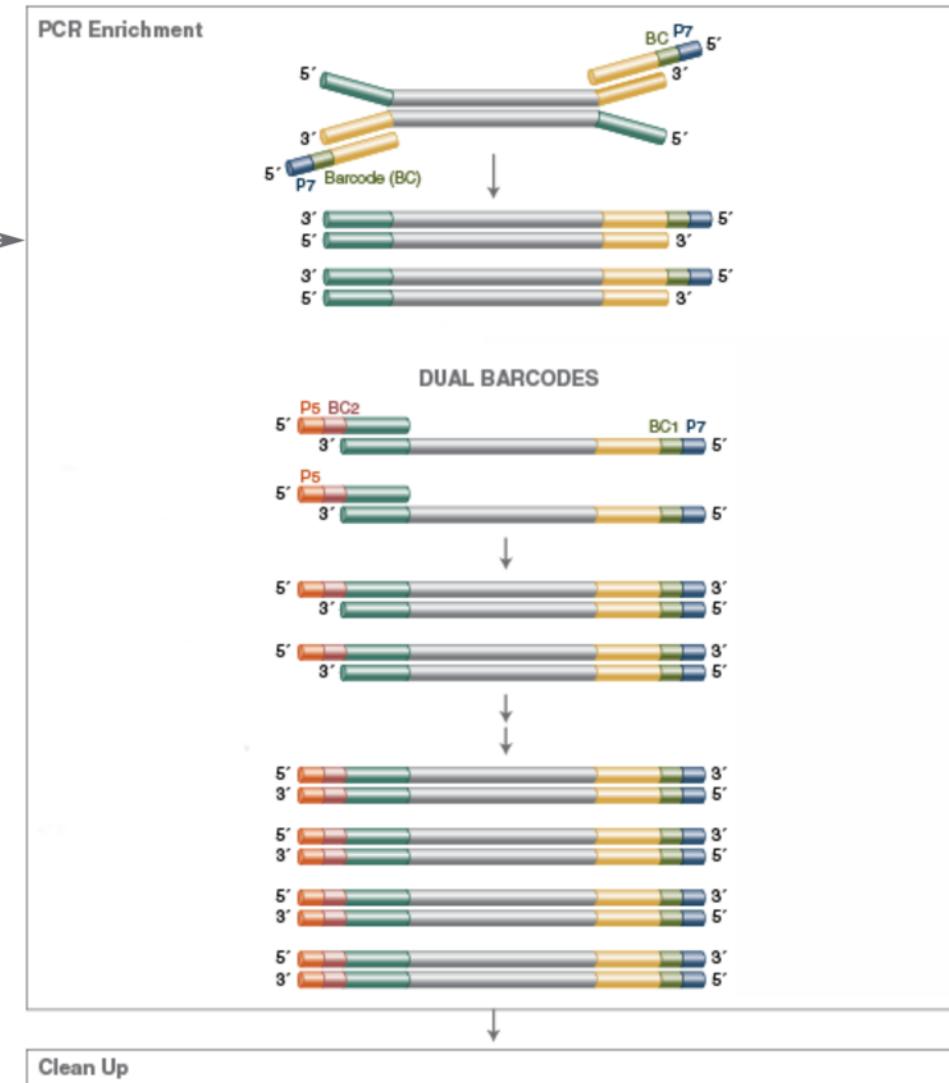
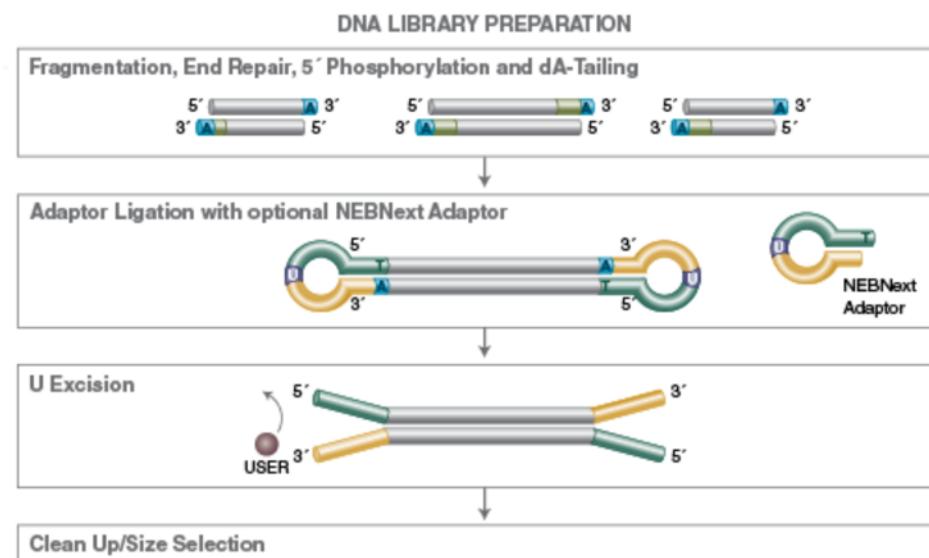


# Next Generation Sequencing

- We'll focus on Illumina short read sequencing as this is the most commonly used method at the moment

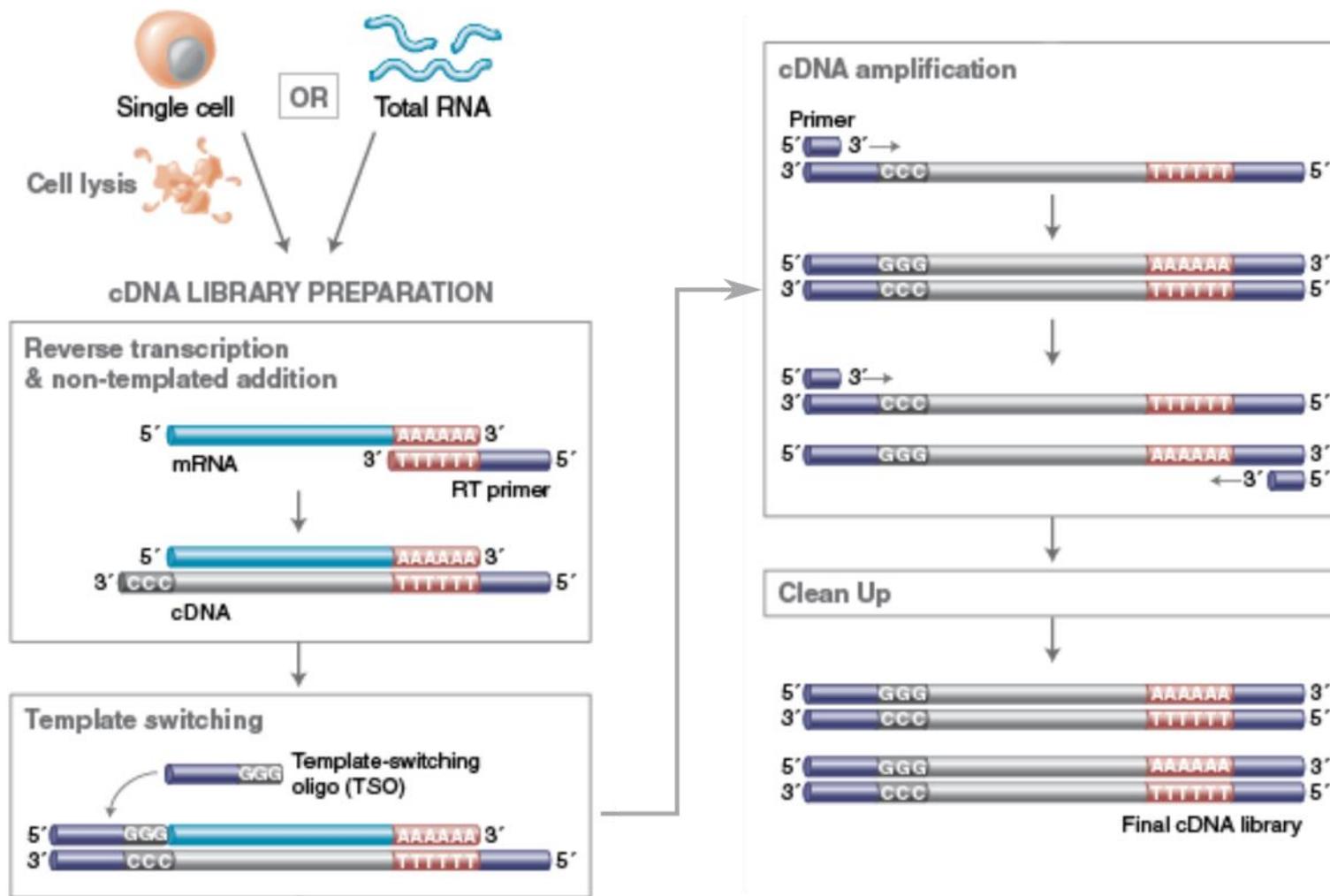
# Next Generation Sequencing

## Library preparation



# Next Generation Sequencing

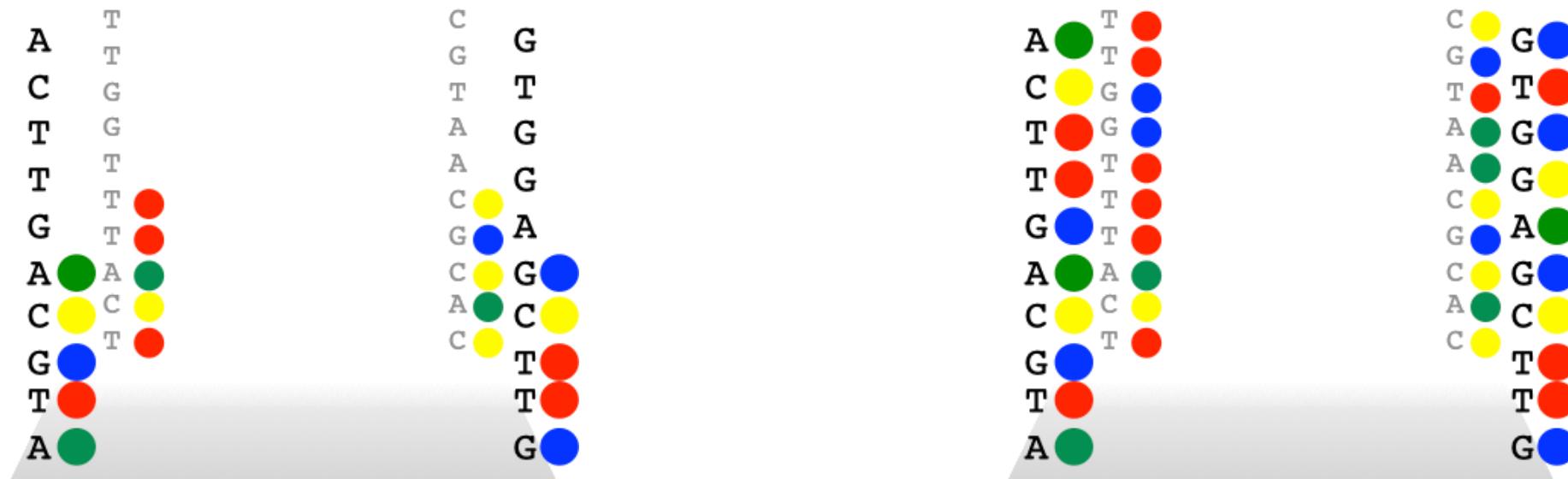
## Library preparation



# Next Generation Sequencing

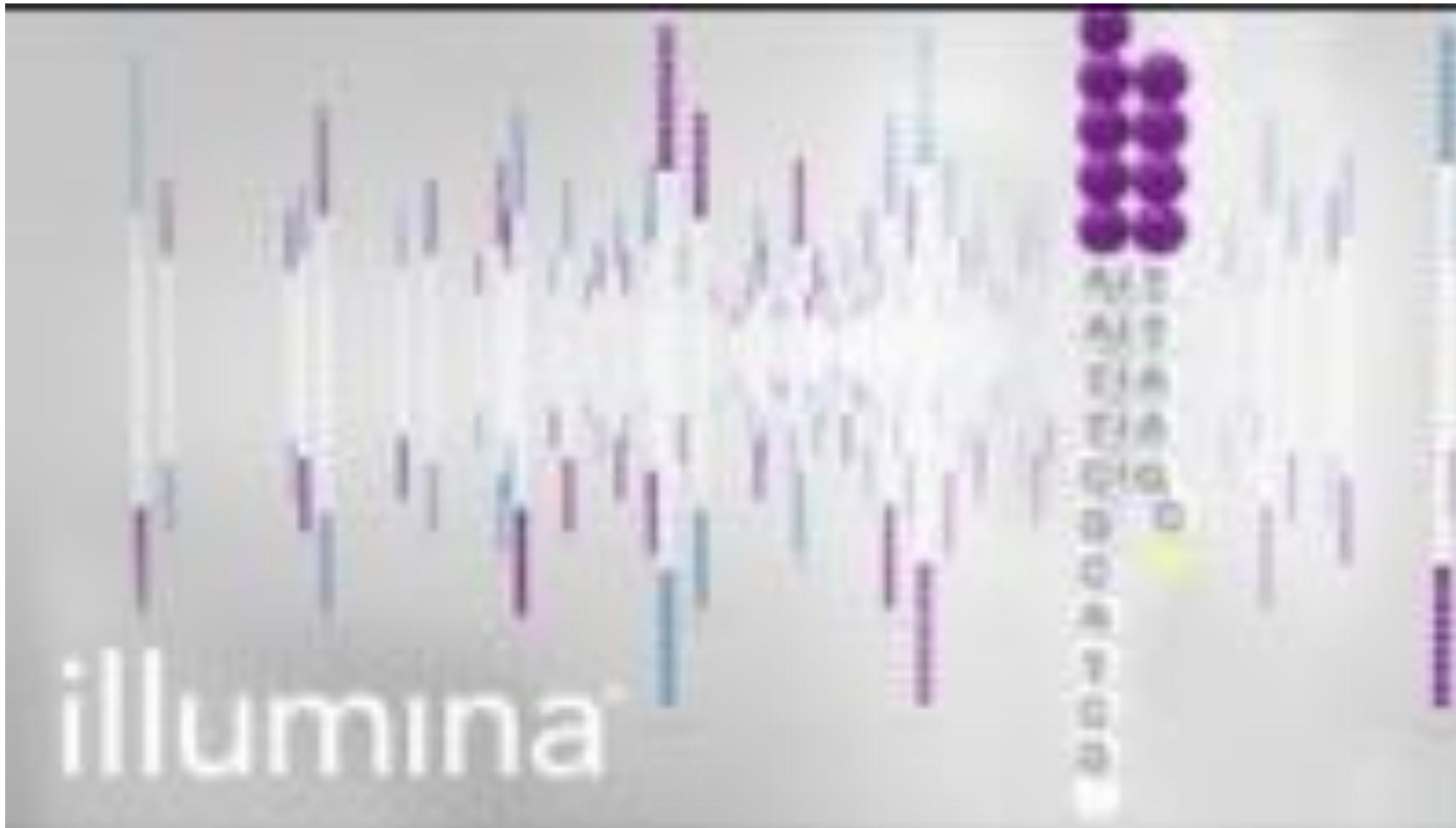
## Sequencing by synthesis

- A complementary strand is synthesized using the cDNA fragment as template.
- Each nucleotide includes a fluorescent tag and as the new strand is synthesized, the colour of the fluorescence indicates which base is being added.
- The sequencer records the order of these flashes of light and translates them to a base sequence.



# Next Generation Sequencing

Sequencing by synthesis



# Fastq file format

# Fastq file format - Headers

```
@LH00417:64:22MCGJLT3:3:1101:12206:1064 1:N:0:GATCAAGGCA+ATTAACAAGG  
CNTTGACATTGTACTCGGCAGAAGATGTGCCGGGCCAGGCTGCTGCCACTGGGCCTCAGGGATAGAGGTGCTGA  
+  
9#IIIIII9II9I-9IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII  
@LH00417:64:22MCGJLT3:3:1101:12410:1064 1:N:0:GATCAAGGCA+ATTAACAAGG  
TNCCAAACAACCAAACAAAAATCCCCAGCATTTCAGTGCATGAATATCTGAAAATCCTCCTGTGCAAGTAT/  
+  
9#9IIIIIII-IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII  
@LH00417:64:22MCGJLT3:3:1101:12798:1064 1:N:0:GATCAAGGCA+ATTAACAAGG  
TNTGACGATGTTAGCGGTTACTCTTCATTGCTTCTTAGCTGCCAGGATCAGTGGCGTCAAGCTAGACAATGGTT  
+  
9#IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII  
@LH00417:64:22MCGJLT3:3:1101:13963:1064 1:N:0:GATCAAGGCA+ATTAACAAGG  
CNTTCCTTCAGCCAGTGCCTGCACAGATGAAGCTGACATCTGATCCTCATGTCCCTGACAATTGCCGAGTAC  
+  
9#IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
```

# Fastq file format - Sequence

```
@LH00417:64:22MCGJLT3:3:1101:12206:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
CNTTGACATTGTACTCGGCAGAAGATGTGCAGGCCAGGCTGCTGCCACTGGGCCTCAGGGATAGAGGTGCTGA
+
9#IIIIII9II9I-9IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
@LH00417:64:22MCGJLT3:3:1101:12410:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
TNCCAAACAACCAAACAAAAATCCCCAGCATTTCAGTGCATGAATATCTGAAAATCCTCCTGTGCAAGTAT
+
9#9IIIIIII-IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
@LH00417:64:22MCGJLT3:3:1101:12798:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
TNTGACGATGTTAGCGGTTACTCTTCATTGCTCTTAGCTGCCAGGATCAGTGGCGTCAAGCTAGACAATGGTT
+
9#IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
@LH00417:64:22MCGJLT3:3:1101:13963:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
CNTTCCTTCAGCCAGTGCCTGCACAGATGAAGCTGACATCTGATCCTCATGTCCCTGACAATTGCCGAGTAC
+
9#IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
```

## Fastq file format – Line 3

```
@LH00417:64:22MCGJLT3:3:1101:12206:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
CNTTGACATTGTACTCGGCAGAAGATGTGCCGGCCAGGCTGCTGCCACTGGGCCTCAGGGATAGAGGTGCTGA
+
9#IIIIII9II9I-9IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
@LH00417:64:22MCGJLT3:3:1101:12410:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
TNCCAAACAACCAAACAAAAATCCCCAGCATTTCAGTGCATGAATATCTGAAAATCCTCCTGTGCAAGTAT/
+
9#9IIIIIII-IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
@LH00417:64:22MCGJLT3:3:1101:12798:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
TNTGACGATGTTAGCGGTTACTCTTCATTGCTCTTAGCTGCCAGGATCAGTGGCGTCAAGCTAGACAATGGTT
+
9#IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
@LH00417:64:22MCGJLT3:3:1101:13963:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
CNTTCCTTCAGCCAGTGCCTGCACAGATGAAGCTGACATCTGATCCTCATGTCCCTGACAATTGCCGAGTAC
+
9#IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
```

# Fastq file format - Headers

```
@LH00417:64:22MCGJLT3:3:1101:12206:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
CNTTGACATTGTACTCGGCAGAAGATGTGCCGGCCAGGCTGCTGCCACTGGGCCTCAGGGATAGAGGTGCTGA
+
9#IIIIII9II9I-9IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
@LH00417:64:22MCGJLT3:3:1101:12410:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
TNCCAAACAACCAAACAAAAATCCCCAGCATTTCAGTGCATGAATATCTGAAAATCCTCCTGTGCAAGTAT/
+
9#9IIIIIII-IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
@LH00417:64:22MCGJLT3:3:1101:12798:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
TNTGACGATGTTAGCGGTTACTCTTCATTGCTCTTAGCTGCCAGGATCAGTGGCGTCAAGCTAGACAATGGTT
+
9#IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
@LH00417:64:22MCGJLT3:3:1101:13963:1064 1:N:0:GATCAAGGCA+ATTAACAAGG
CNTTCCTTCAGCCAGTGCCTGCACAGATGAAGCTGACATCTGATCCTCATGTCCCTGACAATTGCCGAGTAC
+
9#IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII
```

# (Phred) Quality Scores

Sequence quality scores are transformed and translated p-values

Sequence bases are called after image processing (base calling):

- Each base in a sequence has a p-value associated with it
- p-values range from 0-1 (e.g.: 0.05, 0.01, 1e-30)
- p-value of 0.01 implies 1 in 100 chance that called base is wrong

# (Phred) Quality Scores ...

How do we assign p-values to bases in the fastq file?

- P-values can be many characters long (e.g.:0.000005)
- Transform to Phred quality scores –  $Q = -10 \times \log_{10}(\text{pvalue})$  :  
If  $p = 0.01 \rightarrow \log_{10}(0.01) = -2 \rightarrow Q = 20$
- Translate Q values to ASCII characters (adding 33):  
Q value of 2 = #, Q value of 40 = !
- This gives us a single digit quality score code for each base that fits nicely in the fastq format

Dec	Hex	Chr	Dec	Hex	Chr	Dec	Hex	Chr	Dec	Hex	Chr
0	00	NUL	32	20	Space	64	40	@	96	60	'
1	01	SOH	33	21	!	65	41	A	97	61	a
2	02	STX	34	22	"	66	42	B	98	62	b
3	03	ETX	35	23	#	67	43	C	99	63	c
4	04	EOT	36	24	\$	68	44	D	100	64	d
5	05	ENQ	37	25	%	69	45	E	101	65	e
6	06	ACK	38	26	&	70	46	F	102	66	f
7	07	BEL	39	27	'	71	47	G	103	67	g
8	08	BS	40	28	(	72	48	H	104	68	h
9	09	HT	41	29	)	73	49	I	105	69	i
10	0A	LF	42	2A	*	74	4A	J	106	6A	j
11	0B	VT	43	2B	+	75	4B	K	107	6B	k
12	0C	FF	44	2C	,	76	4C	L	108	6C	l
13	0D	CR	45	2D	-	77	4D	M	109	6D	m
14	0E	SO	46	2E	.	78	4E	N	110	6E	n
15	0F	SI	47	2F	/	79	4F	O	111	6F	o
16	10	DLE	48	30	0	80	50	P	112	70	p
17	11	DC1	49	31	1	81	51	Q	113	71	q
18	12	DC2	50	32	2	82	52	R	114	72	r
19	13	DC3	51	33	3	83	53	S	115	73	s
20	14	DC4	52	34	4	84	54	T	116	74	t
21	15	NAK	53	35	5	85	55	U	117	75	u
22	16	SYN	54	36	6	86	56	V	118	76	v
23	17	ETB	55	37	7	87	57	W	119	77	w
24	18	CAN	56	38	8	88	58	X	120	78	x

# QC is important

Check for any problems before we put time and effort into analysing potentially bad data

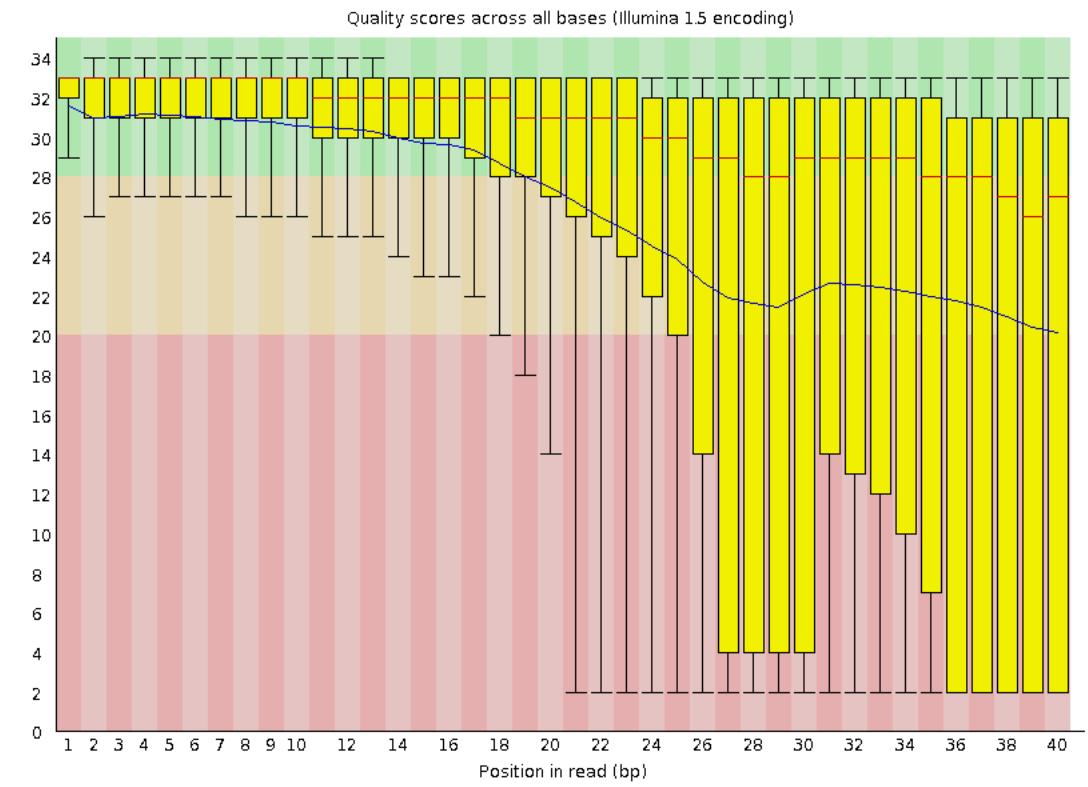
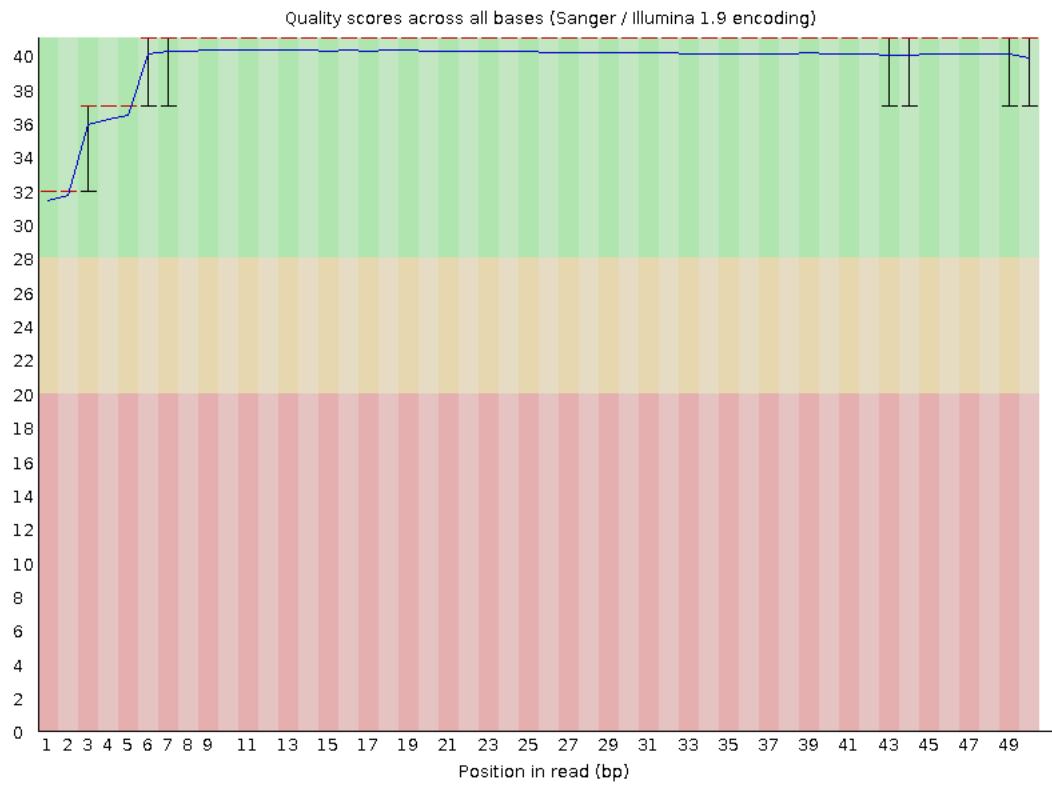
Start with FastQC:

- Quick
- Outputs an easy-to-read html report

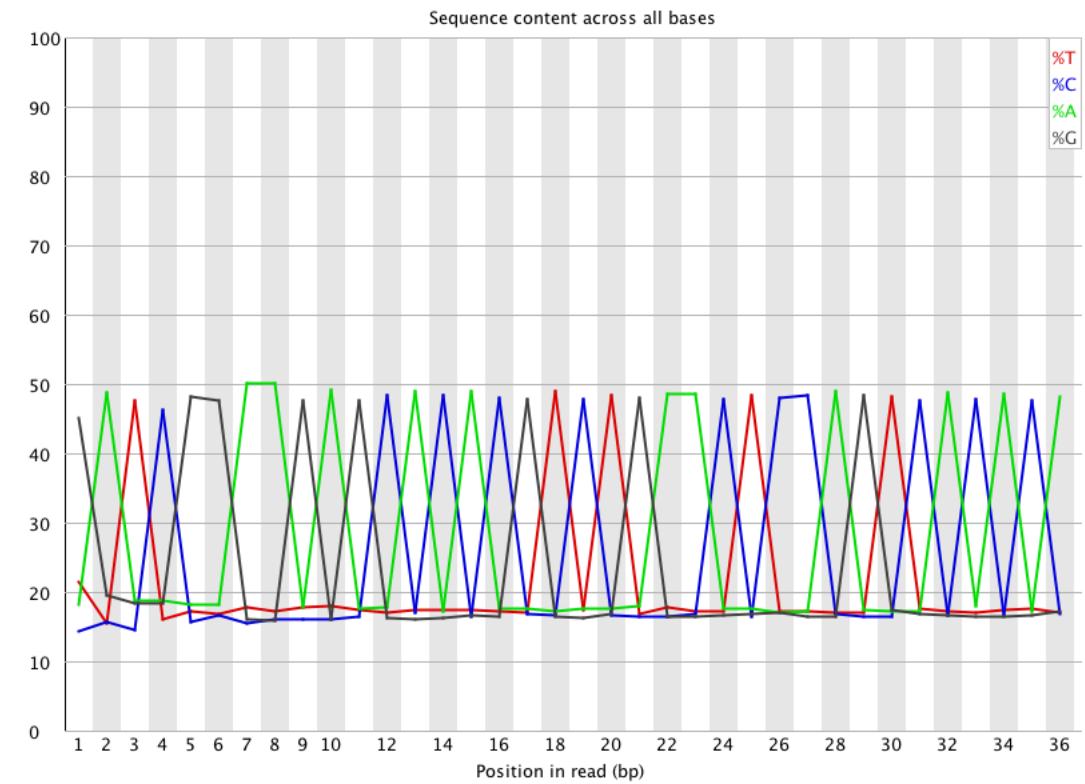
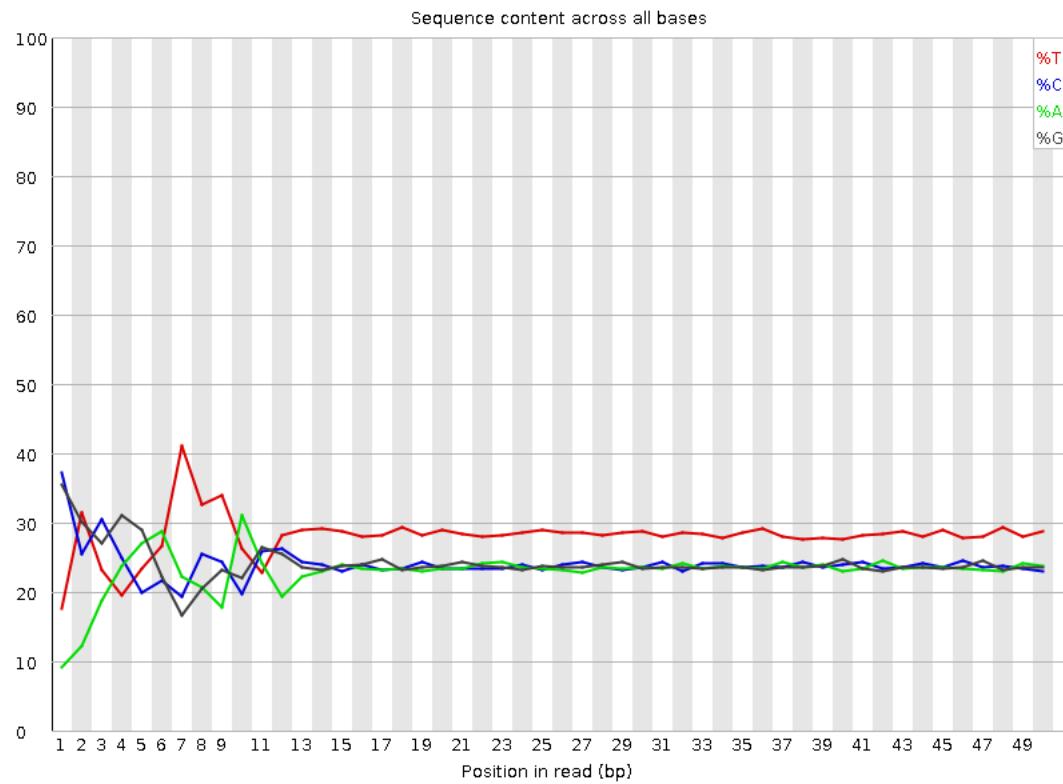
<https://www.bioinformatics.babraham.ac.uk/projects/fastqc/>



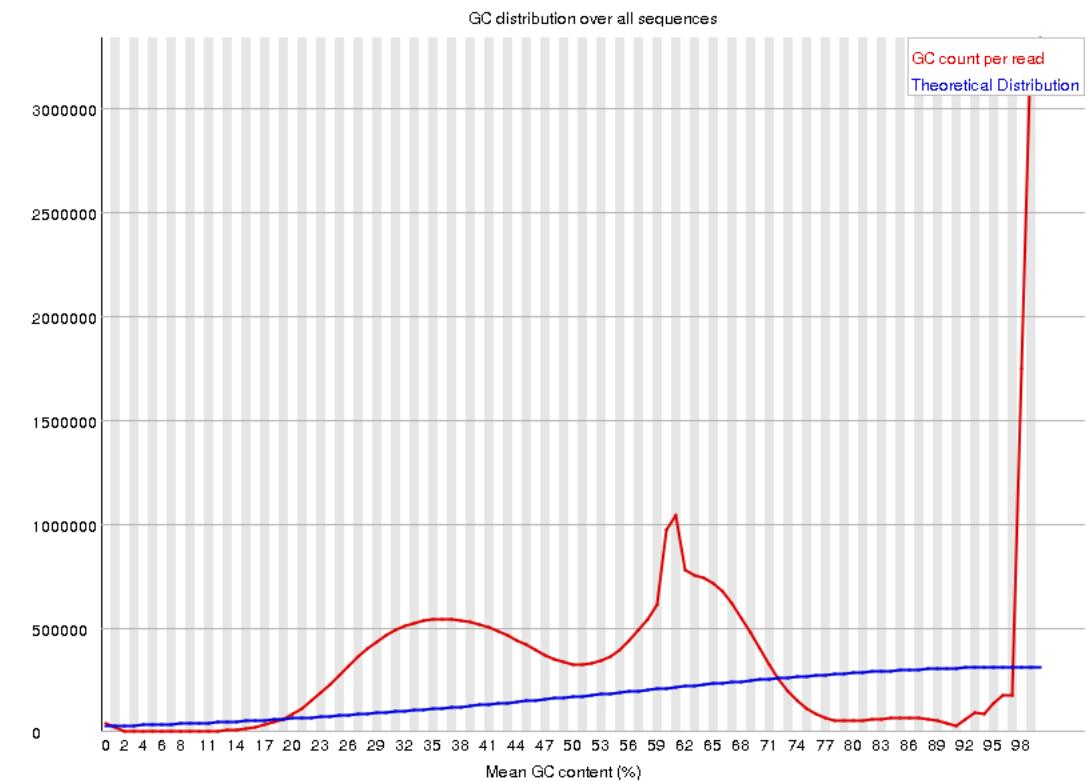
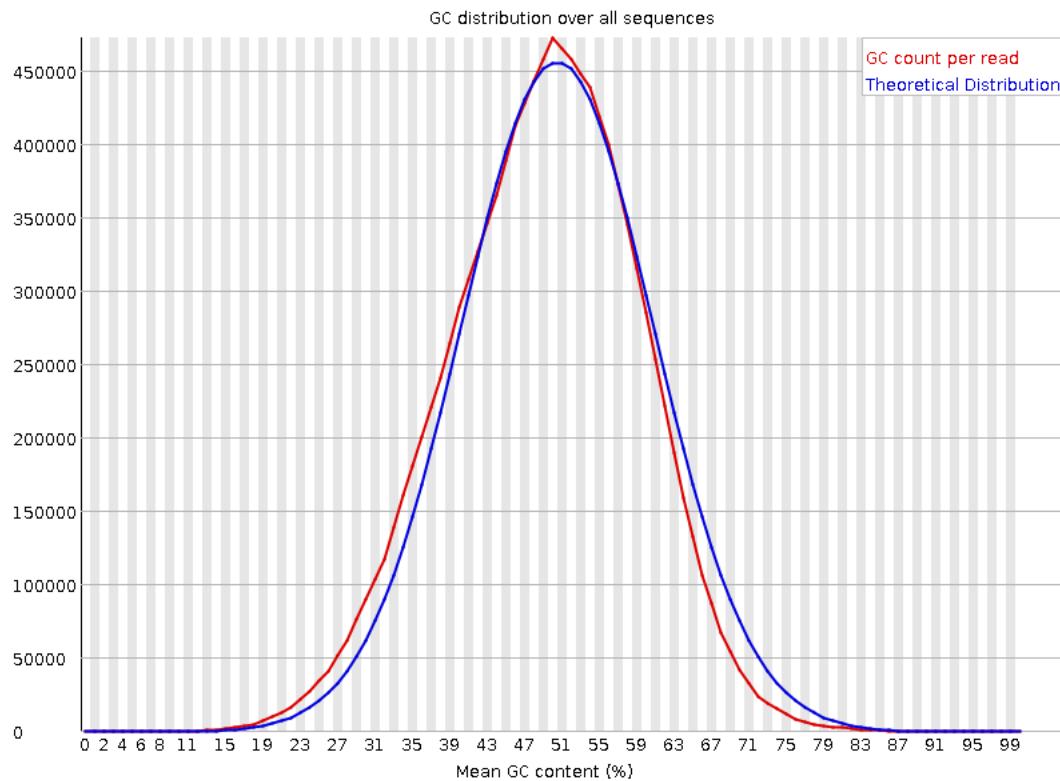
# Per base sequence quality



# Per base sequence content



# Per sequence GC content



# Read Alignment

AIM: Given a reference sequence and a set of short reads, align each read to the reference sequence finding the most likely origin of the read sequence.

**Reference:** ...GCTGATGTGCCGCCCTCACTTCGGTGGTACGCT...

**Reads:** {

- GATGTGCCGCCCTCACTTCGG
- TGTGCCG**G**CTCACTTCGGTG
- CTGATGTGCCG**G**CTCACTTC
- GG**CTCACTTCGGTGGTACCGC
- CCGCCTCACTTCGGTGGTAC
- CCGCCTCACTTCGGTGGTAC

# Read Alignment

- 1970 – Needleman-Wunsch algorithm for sequence alignment published
- The Smith-Waterman algorithm was the first alignment algorithm to include the concept of gaps
- Faster less computationally intensive alignment algorithms have been developed
- Aligners also use indexes of the genome to speed up alignment

Initialize the scoring matrix

	T	G	T	T	A	C	G	G
T	0	0	0	0	0	0	0	0
G	0							
G	0							
T	0							
T	0							
G	0							
A	0							
C	0							
T	0							
A	0							

Substitution matrix:

$$S(a_i, b_j) = \begin{cases} +3, & a_i = b_j \\ -3, & a_i \neq b_j \end{cases}$$

Gap penalty:  $W_k = kW_1$   
 $W_1 = 2$

# Short Read Aligners

There are a lot of short read aligners, which use a variety of indexing algorithms and alignment algorithms.

The most popular are probably:

- The Burrows-Wheeler Aligner (BWA) - <https://github.com/lh3/bwa>
- Bowtie2 - <https://github.com/BenLangmead/bowtie2>
- STAR - <https://github.com/alexdobin/STAR>
- HISAT2 - <http://daehwankimlab.github.io/hisat2/>

# Alignment

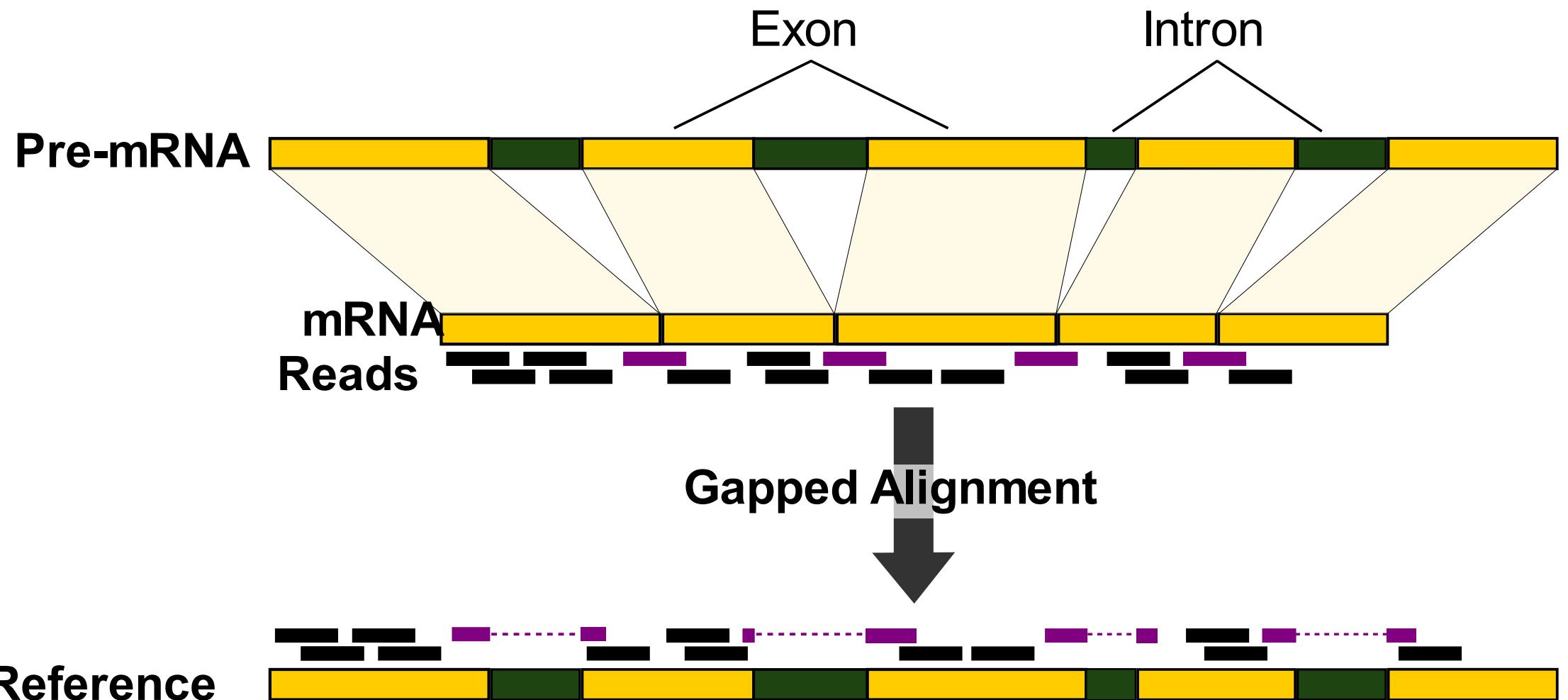
Ref AACTTGGCCATGCAGGCCGTTAGACAACTTAGTACTGACACCGAAG

Read TTGGCCACGTAAACCGTTACGCTAAGTACT

# Alignment

Ref AACTTGCCATGCAGGCCGTTAGACAACTTAGTACTGACACCGAAG  
↑↑↑  
Read TTGCCCACGTAAACCGTTA DEL CGCTAAGTACT

# mRNAseq Alignment



# Short Read Aligners

There are a lot of short read aligners, which use a variety of indexing algorithms and alignment algorithms.

The most popular are probably:

- The Burrows-Wheeler Aligner (BWA) – DNA
- Bowtie2 – DNA
- STAR – Either
- HISAT2 – Either

# SAM format

Sequence Alignment/Map (SAM) format is the standard format for files containing aligned reads.

Two main parts:

- Header - meta data (source of the reads, reference genome, aligner, etc.)
- Alignment section:
  - 1 line for each alignment
  - contains details of alignment position, mapping, base quality etc.
  - 11 required fields, but other content may vary depending on aligner and other tools used to create the file

# SAM format

LH00417:64:22MCGJLT3:3:1101:24078:1064 163 X 20546017 60 1S5M1N89M236N5M = 20547099  
1416 ATGTTGATGACAGCCGTCTTGAGGAGCTCAAAGCCACATTGCCAGCCCAGACAAGTTACCCGGATTAAAGATGTACCCATTGAT  
TTTGAGAAGGATGA -II  
IIIIIIIIIIIIIIIIII NH:i:1 HI:i:1 AS:i:200 nM:i:0  
LH00417:64:22MCGJLT3:3:1101:24078:1064 83 X 20547099 60 12M236N86M2S = 20546017  
-1416 GATTTGAGAAGGATGATGACAGCAATTCCACATGGATTTCATTGTGGCTGCATCCAATCTCGGGCCGAAAATGATATT  
CCCTGCAGACCGNG IIIIIIIIIIIII-9III9II  
IIIIIIIIIIIIII#9 NH:i:1 HI:i:1 AS:i:200 nM:i:0  
LH00417:64:22MCGJLT3:3:1101:25280:1064 163 1 34964132 60 53D21M2D26M = 4964187 155 AAG  
AACAGAAAATTGCCTACTTCAGGGGCCTAGATACTATTGCCATTGAGTACAGAGCCGAGATTTTTTTAAGTGAGGGGTGTGTGTCTG  
TG -II-IIII9I-I-  
I9III-III NH:i:1 HI:i:1 AS:i:175 nM:i:0  
LH00417:64:22MCGJLT3:3:1101:25280:1064 83 1 34964187 60 21M2D77M2S = 4964132 -155  
AGAGCCGAGATTTTTTAAGTGAGGGGTGTGTGTCTGTGTGTGTGCGTGCCTGCGTGTGCGTGTGCGTGTGTGT  
GTGTNG IIIIII-II  
IIIIIIIIII#9 NH:i:1 HI:i:1 AS:i:175 nM:i:0

# SAM format

LH00417:64:22MCGJLT3:3:1101:24078:1064 163 X 20546017 60 1S5M1N89M236N5M = 20547099  
1416 ATGTTGATGACAGCCGTCTTGAGGAGCTCAAAGCCACATTGCCAGCCCAGACAAGTTACCCGGATTTAAGATGTACCCATTGAT  
TTT GAGAAGGATGA -II  
IIIIIIIIIIIIIIIIII NH:i:1 HI:i:1 AS:i:200 nM:i:0  
LH00417:64:22MCGJLT3:3:1101:24078:1064 83 X 20547099 60 12M236N86M2S = 20546017  
-1416 GATTTGAGAAGGATGATGACAGCAATTCCACATGGATTCATTGTGGCTGCATCCAATCTCGGGCCGAAAATATGATATTG  
CCCTGCAGACCGNG IIIIIIIIIIIII-9III9II  
IIIIIIIIIIIIIIII#9 NH:i:1 HI:i:1 AS:i:200 nM:i:0  
LH00417:64:22MCGJLT3:3:1101:25280:1064 163 1 34964132 60 53D21M2D26M = 4964187 155 AAG  
AACAGAAAATTGCCTACTTCAGGGGCCTAGATACTATTGCCATTGAGTACAGAGCCGAGATTTTTTTAAAGTGAGGGGTGTGTGTCTG  
TG -II-IIII9I-I-  
I9IIII-IIII NH:i:1 HI:i:1 AS:i:175 nM:i:0  
LH00417:64:22MCGJLT3:3:1101:25280:1064 83 1 34964187 60 21M2D77M2S = 4964132 -155  
AGAGCCGAGATTTTTTTAAAGTGAGGGGT  
GTGTNG IIIII-II  
IIIIIIIIII#9 NH:i:1 HI:i:1 AS:i:175 nM:i:0

# SAM format

# SAM format

# SAM format

# SAM format

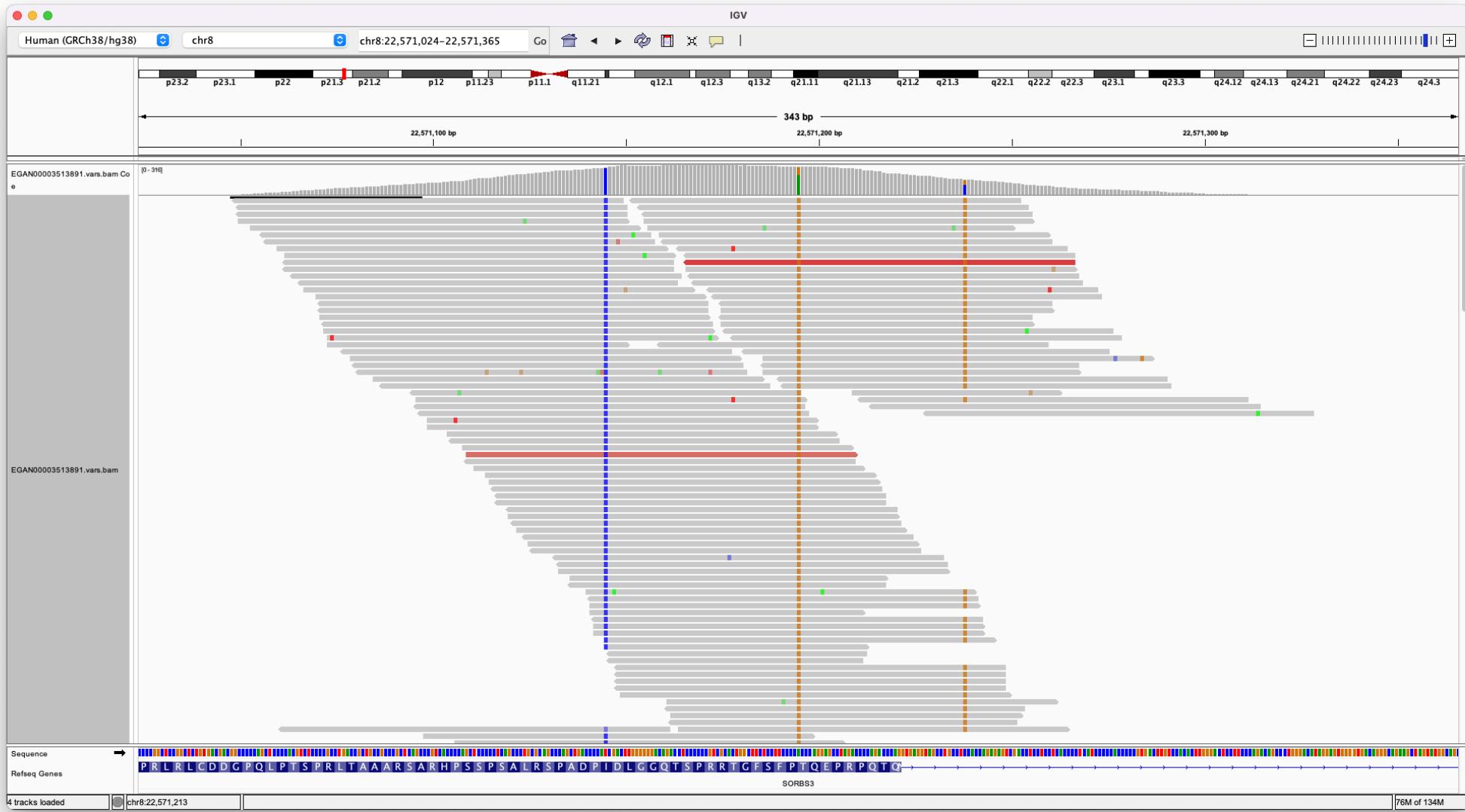
```
QNAME LH00417:64:22MCGJLT3:3:1101:24078:1064
FLAG 163
RNAME X
POS 20546017
MAPQ 60
CIGAR 1S5M1N89M236N5M
RNEXT =
PNEXT 20547099
TLEN 1416
SEQ ATGTTGATGACAGCCGTCTTGAGGAGCTC...
QUAL -IIIIIIIIIIIIIIIIIIIIIIIIII...
NH:i:1
HI:i:1
AS:i:200
nM:i:0
```

# DNAseq – Somatic Variant Calling

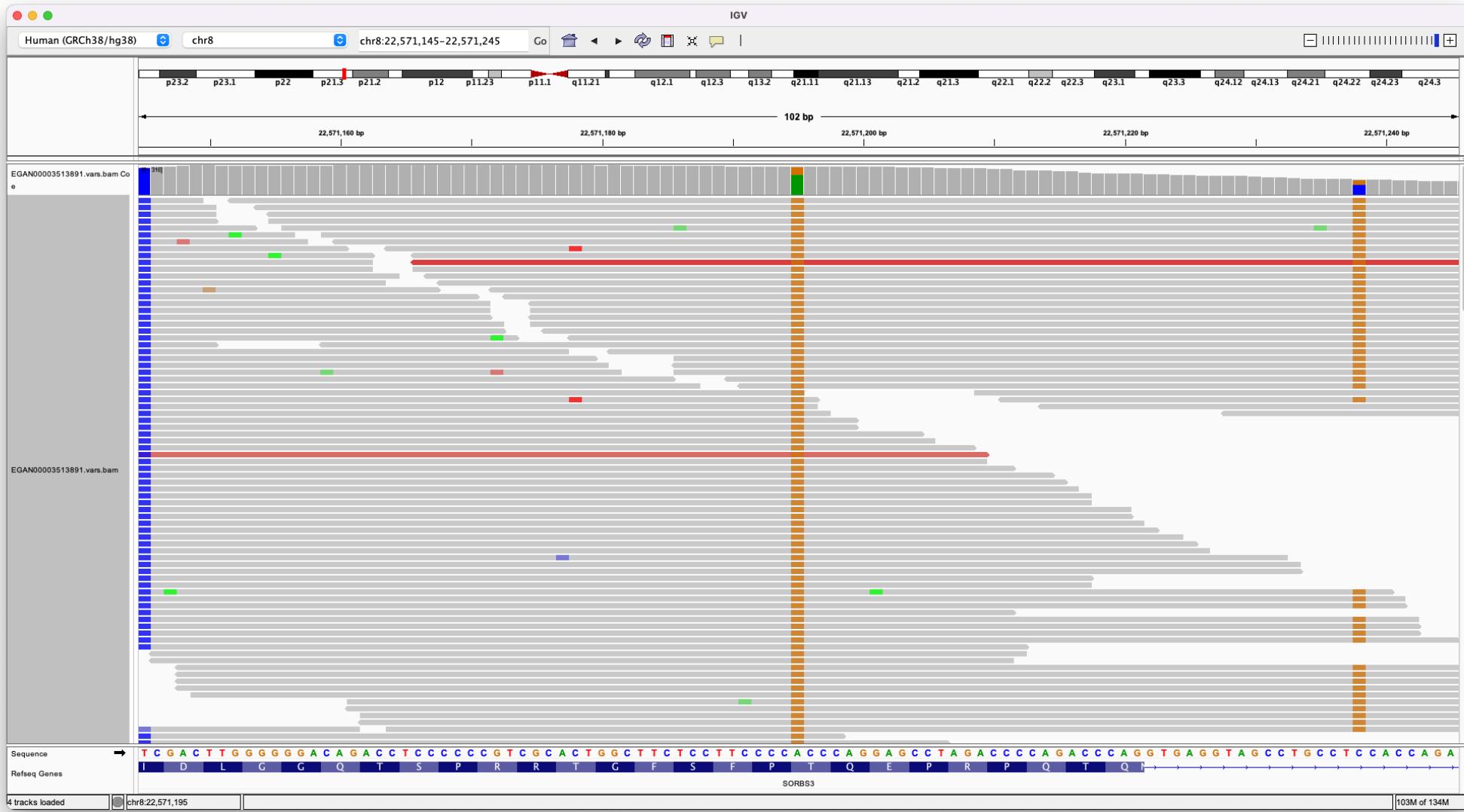
**AIM: To identify mutations in the genome**

- Whole Genome Sequencing, Whole Exome Sequencing, Targeted Panels
- Sequence paired end with long reads ( $>=150$  bp)
- Remove or mark duplicate reads
- Align to genome including viral decoy sequences, e.g. human cytomegalovirus (CMV), Epstein-Barr virus (EBV)
- Call Single Nucleotide Variants (SNV) and small Insertions/Deletions (Indels)
- Many somatic mutations may have very low variant allele frequency – sequence to high depth – 100x coverage
- Differentiate germline variants by including both a tumour sample and a normal sample

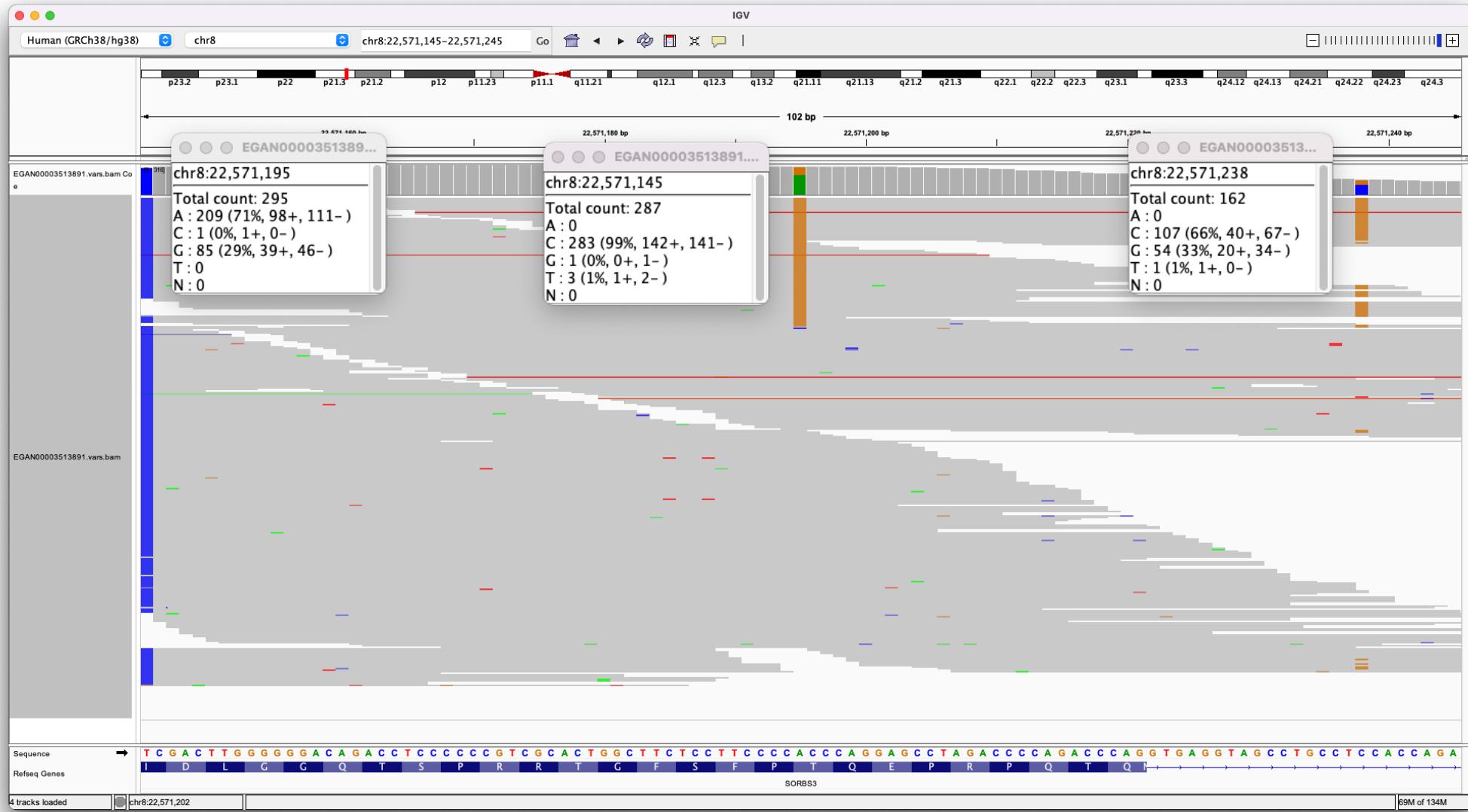
# DNAseq – Somatic Variant Calling



# DNAseq – Somatic Variant Calling



# DNAseq – Somatic Variant Calling



## Several factors complicate somatic SNV calling

- Low cellularity (tumour DNA content)
- Intra-tumour heterogeneity in which multiple tumour cell populations (subclones) exist
- Aneuploidy
- Unbalanced structural variation (deletions, duplications, etc.)
- Matched normal contaminated with cancer DNA
  - adjacent normal tissue may contain residual disease or early tumour-initiating somatic mutations
  - circulating tumour DNA in blood normals
- Sequencing errors
- Alignment artefacts

Mwenifumbo & Marra, Nat Rev Genet. 2013

# DNAseq – Somatic Variant Calling

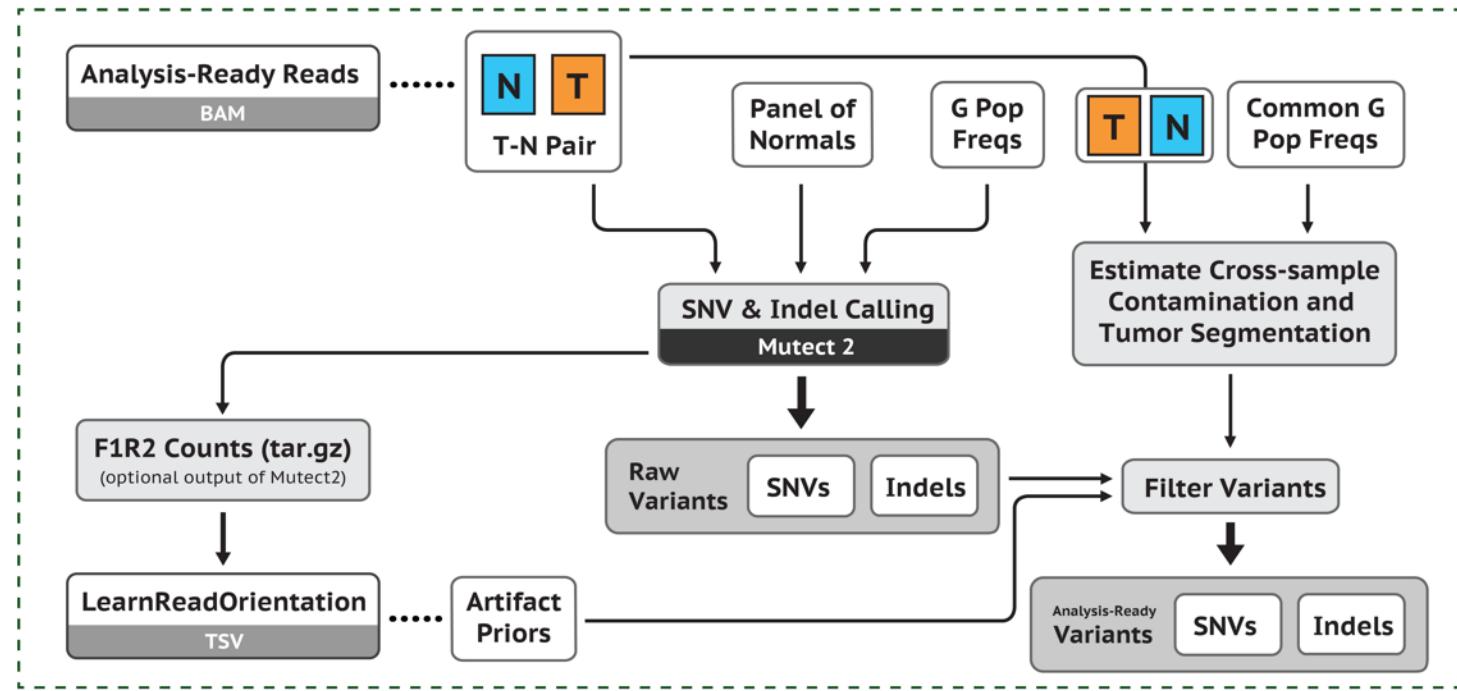
There are a lot of tools for somatic variant calling.

Some of the most popular are:

- Mutect2 (GATK)
- Strelka
- FreeBayes
- VarDict
- VarScan2

They take different approaches, and it is not uncommon to use multiple tools to call variants and then take a consensus

# Using Mutect2 from GATK



References required:

- Genome Reference – same as used to align
- Germline Resource – e.g. Gnomad
- If not WGS then intervals list

# Variant Call Format (VCF) output

```
##fileformat=VCFv4.1

##FILTER=<ID=base_quality,Description="Average base quality for variant alleles < 25">
##FILTER=<ID=germline_risk,Description="Evidence suggests that the site may be germline, not somatic">

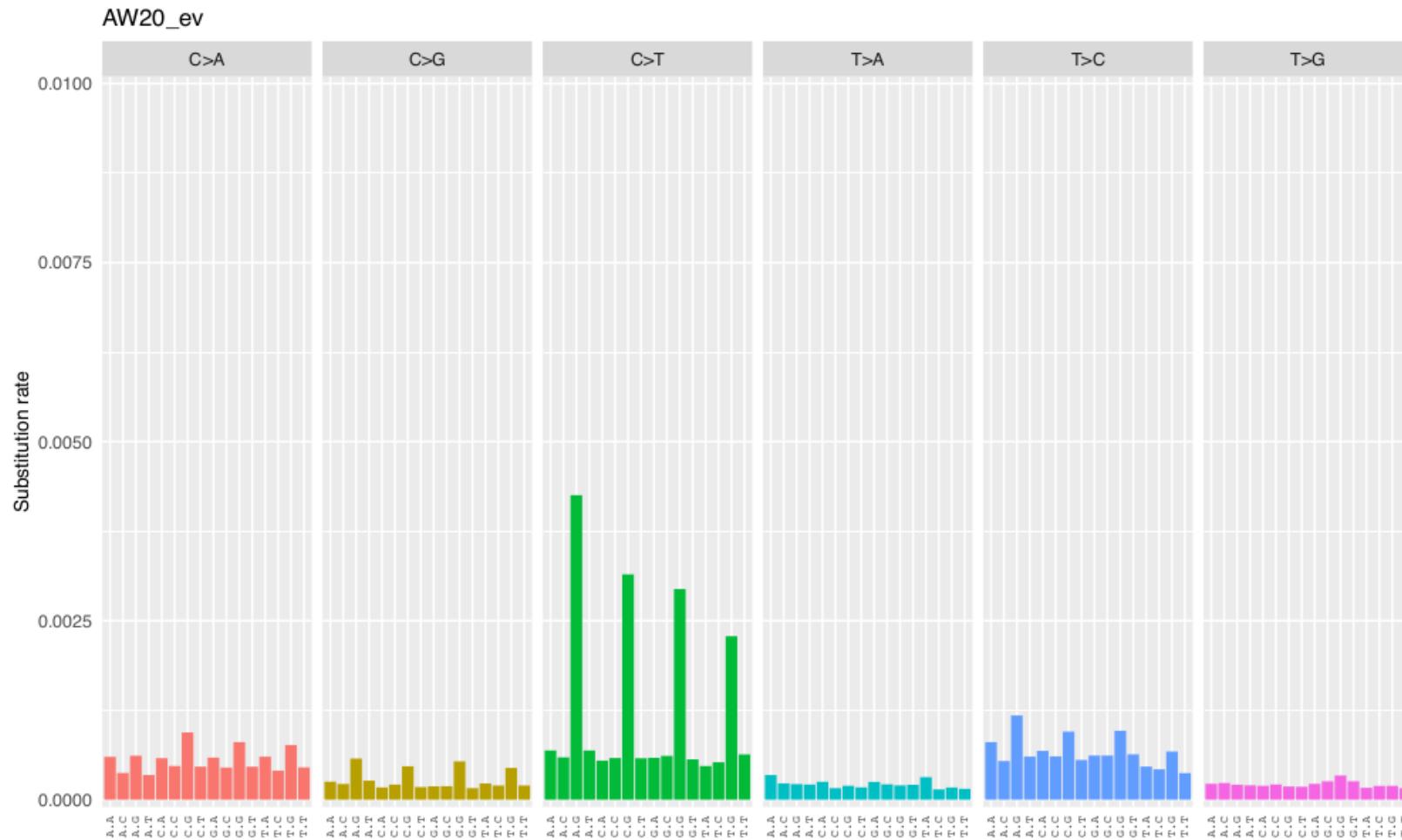
##FORMAT=<ID=GT,Number=1,Type=String,Description="Genotype">
##FORMAT=<ID=DP,Number=1,Type=Integer,Description="Read depth at this position in the sample">
##FORMAT=<ID=AD,Number=R,Type=Integer,Description="Allelic depths for the ref and alt alleles in the order listed">
##FORMAT=<ID=BQ,Number=.,Type=Integer,Description="Average base quality for reads supporting alleles">

#CHROM POS ID REF ALT QUAL FILTER INFO FORMAT NORMAL TUMOUR

8 142478336 . A C germline_risk, . GT:DP:AD:BQ 0/0:24:23,1:38,31 0/1:41:35,6:32,14
base_quality

8 142486034 . C T . PASS . GT:DP:AD:BQ 0/0:43:43,0:36,0 0/1:52:44,8:36,40
```

# Hypermutation Signatures in Glioblastoma



# Single Cell RNAseq

## Average expression level

- Comparative transcriptomics
- Disease biomarker
- Homogenous systems



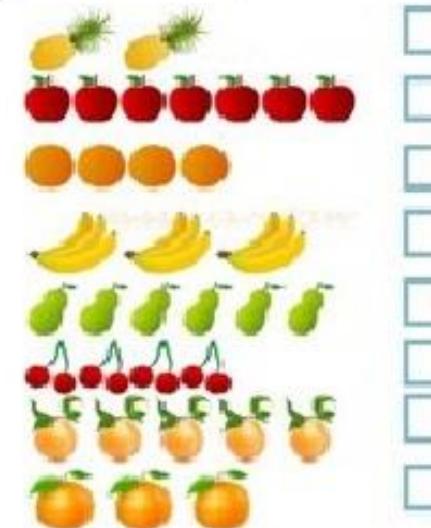
RNA-Seq



## Separate populations

- Define heterogeneity
- Identify rare cell populations
- Cell population dynamics

scRNA-Seq



# Single Cell RNAseq

nature medicine

Letter | Published: 08 June 2020

## A single-cell atlas of the peripheral immune response in patients with severe COVID-19

Aaron J. Wilk, Arjun Rustagi, Nancy Q. Zhao, Jonasel Roque, Giovanny J. Martínez-Colón, Julia L. McKechnie, Geoffrey T. Ivison, Thanmayi Ranganath, Rosemary Vergara,

LETTER

<https://doi.org/10.1038/s41586-018-0394-6>

## A single-cell atlas of the airway epithelium reveals the CFTR-rich pulmonary ionocyte

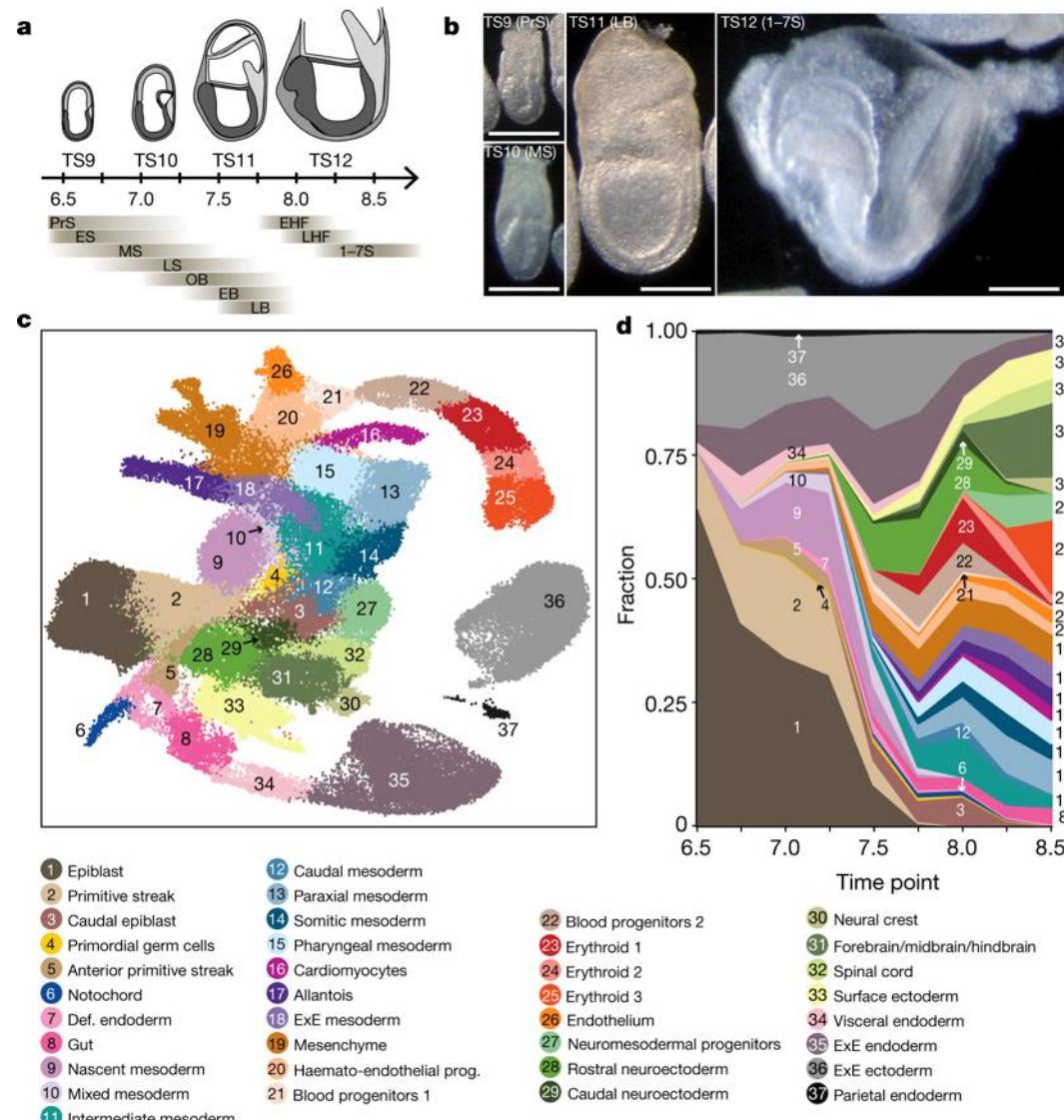
Lindsey W. Plasschaert<sup>1,5,7</sup>, Rapolas Zilionis<sup>2,3,7</sup>, Rayman Choo-Wing<sup>1,5</sup>, Virginia Savova<sup>2,6</sup>, Judith Knehr<sup>4</sup>, Guglielmo Roma<sup>4</sup>, Allon M. Klein<sup>2,6</sup> & Aron B. Jaffee<sup>1,5\*</sup>

nature

Article | Published: 20 February 2019

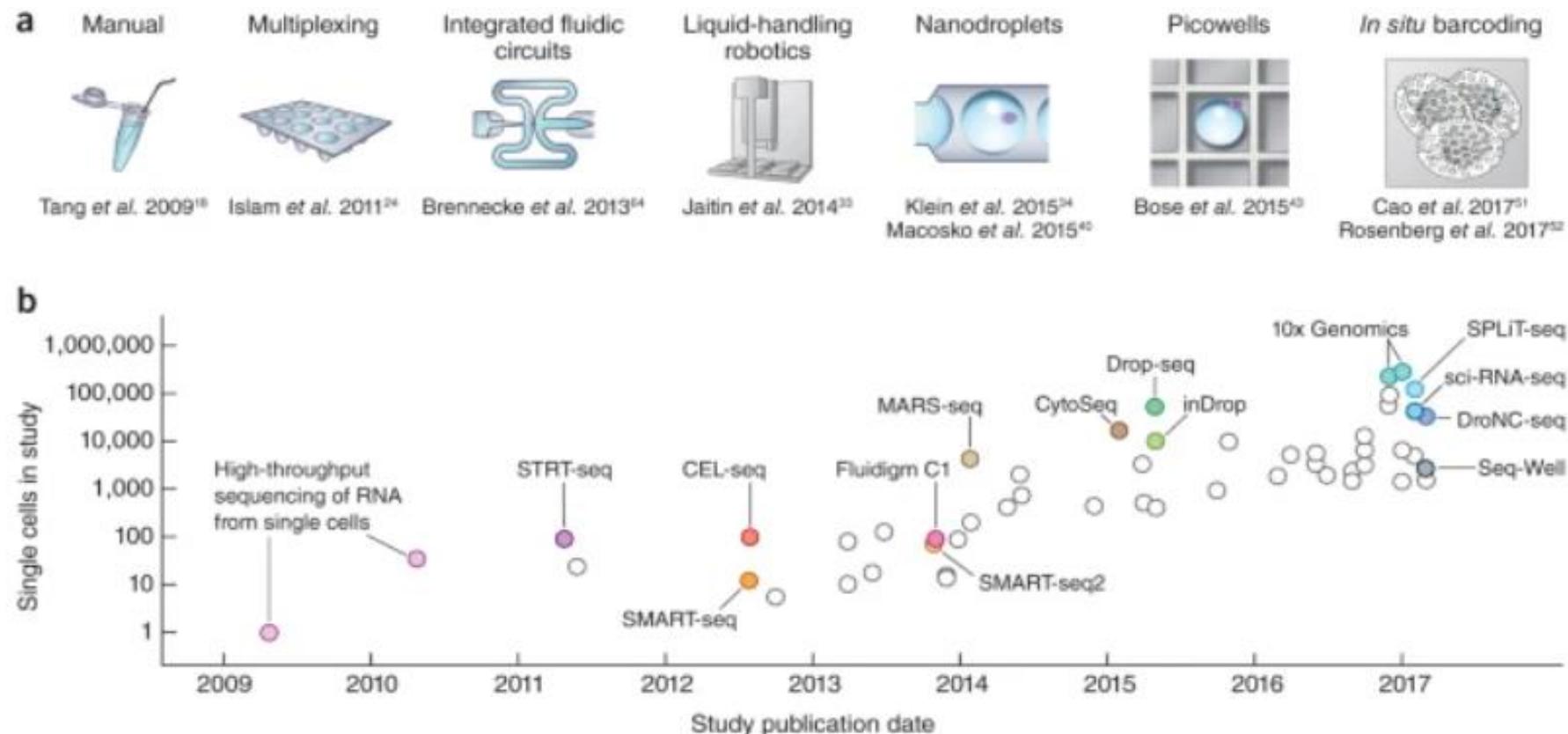
## A single-cell molecular map of mouse gastrulation and early organogenesis

Blanca Pijuan-Sala, Jonathan A. Griffiths, Carolina Guibentif, Tom W. Hiscock, Wajid Jawaid, Fernando J. Calero-Nieto, Carla Mulas, Ximena Ibarra-Soria, Richard C. V.

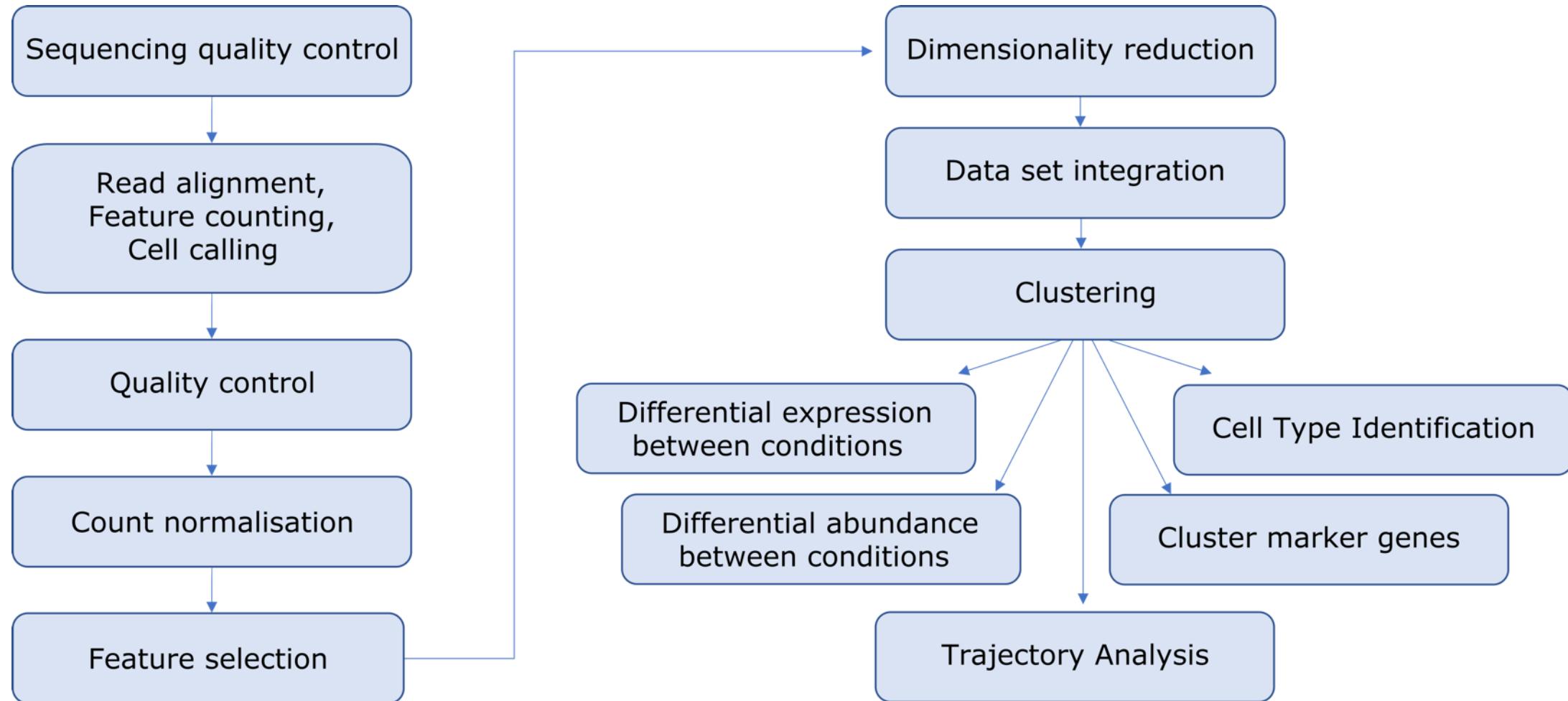


# Single Cell RNAseq

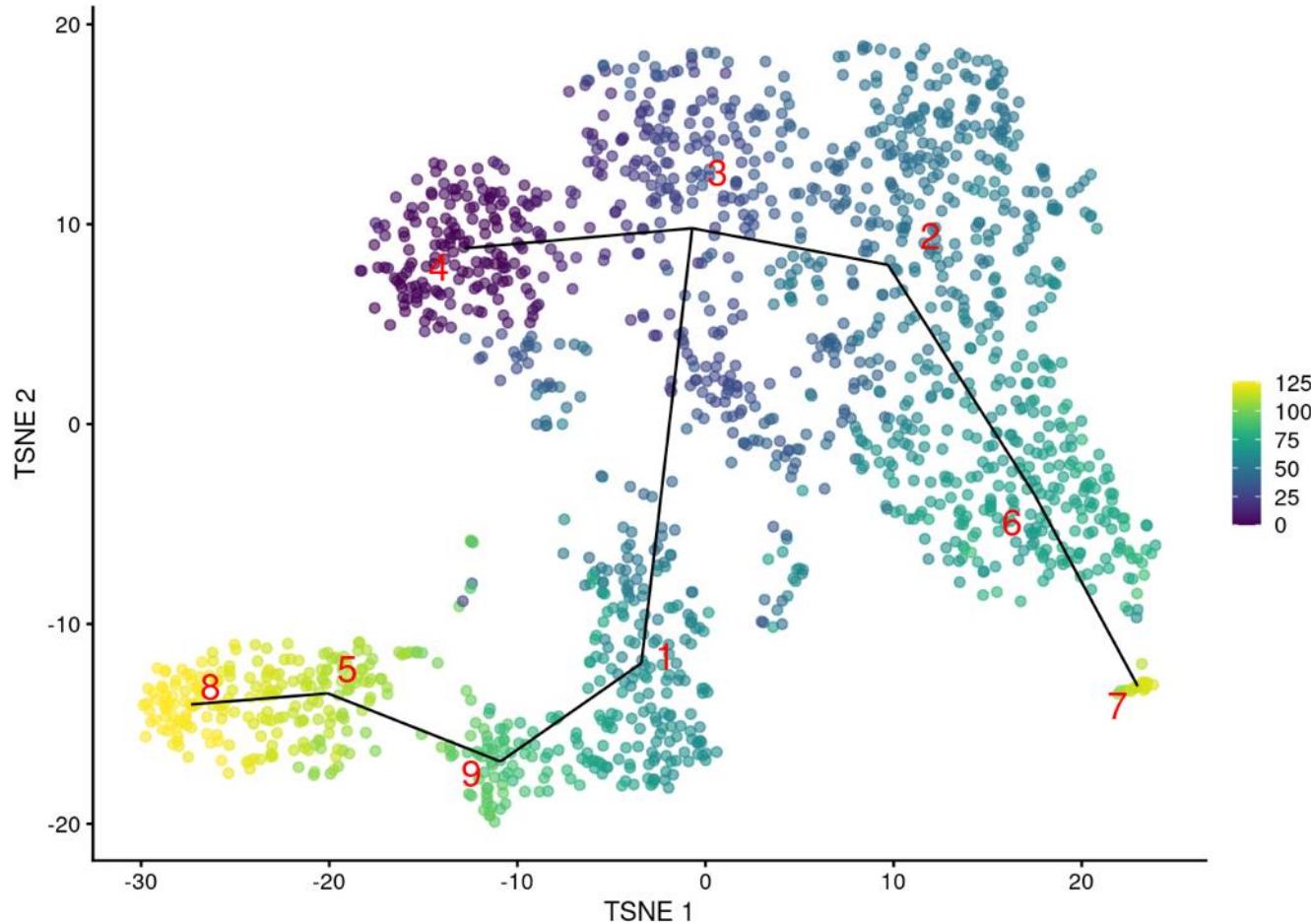
**Figure 1: Scaling of scRNA-seq experiments.**



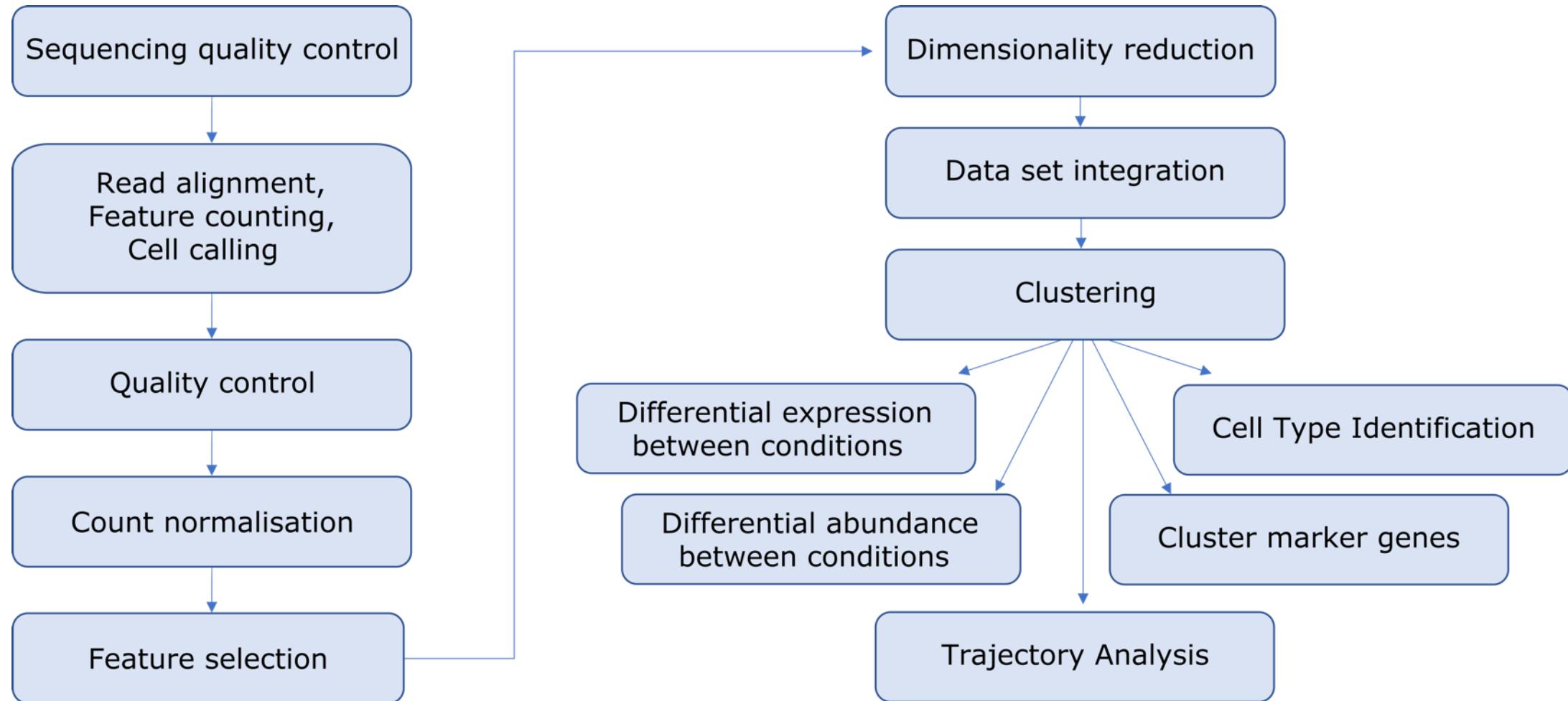
# scRNAseq analysis workflow



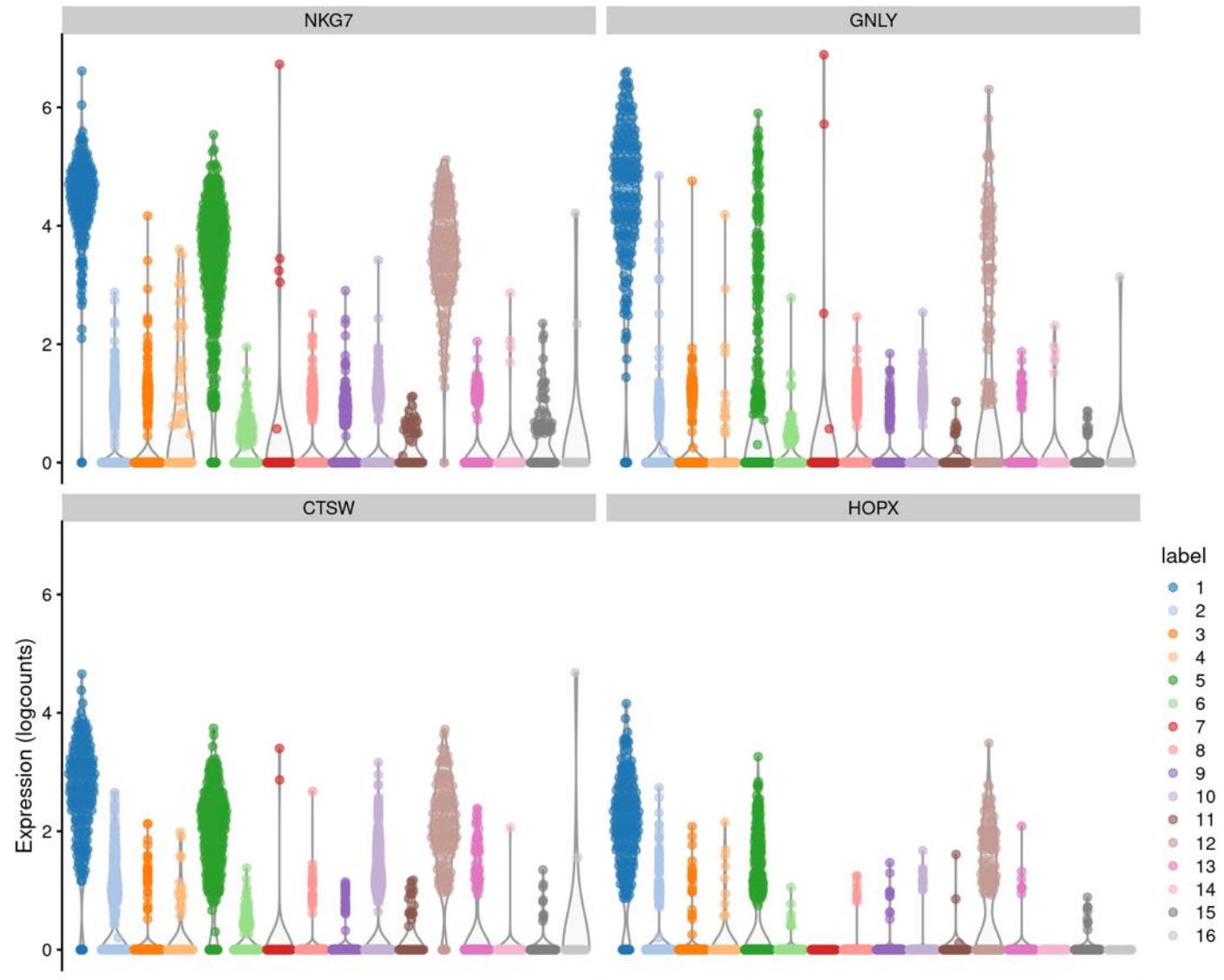
# Trajectory Analysis



# scRNAseq analysis workflow



# Cluster Marker Genes



# scRNAseq analysis tools

- Alignment, Gene Expression Quantification, Cell Calling, QC:
  - CellRanger (10x), STARsolo, Alevin
- Data Exploration:
  - Loupe Browser (10x)
- Downstream Analysis:
  - R – Bioconductor packages: scran, scater, bluster, MiloR, SingleR
    - See the OSCA book at <https://bioconductor.org/books/release/OSCA/>
  - R – Seurat
    - See the Seurat documentation at <https://satijalab.org/seurat/>
  - Python – Scanpy
    - See the Scanpy documentation at <https://scanpy.readthedocs.io/en/stable/>

# What is Bioinformatics?

In the beginning of the 1970s, Ben Hesper and I started to use the term “bioinformatics” for the research we wanted to do, defining it as **“the study of informatic processes in biotic systems”**.

Paulien Hogeweg, <https://doi.org/10.1371/journal.pcbi.1002021>

# THANK YOU