

Outline

- Bioinformatics philosophy
 - We will be working in the bash or “command line” environment.
- Environmental genomics
 - We will focus on practice and principles critical to environmental genomics
- How is the data generated?
 - Critical to know limits and biases
- Data “structures”
 - Bioinformatics requires predictable data formats and conventions

Why are we pushing the Bourne Again Shell (BASH)?

- It is how real bioinformatics is done.
- Opens door to vast library of software.
- This is how people deal with all forms of “Big Data”.
- This is stable/durable knowledge.

Unfortunately:

- Most faculty are not experts.
- Hard to teach what you don't know.
- This is where we provide the focused help.

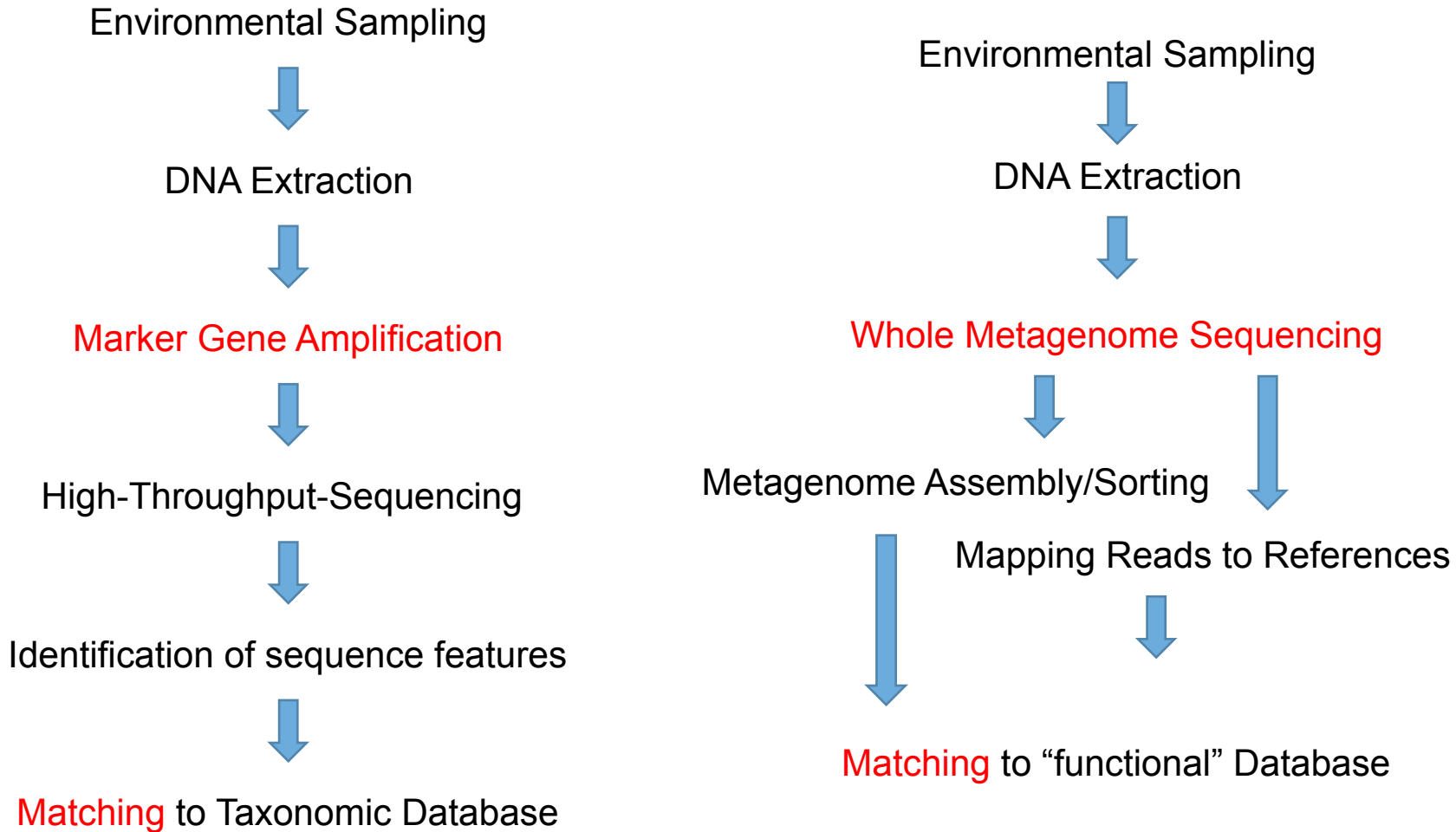
```
[14:05:49]-[devin@ron]-[~/Project_Ghosh/Sample_MP9/assembly]  
└─> gre█
```

Concepts for environmental genomics

- Metagenomics: The study of genetic information gathered directly from the environment. **AKA: environmental genomics, ecogenomics, or community genomics**
 - Sometimes “Whole metagenome shotgun sequencing”
- Metabarcoding: uses “Universal” PCR primers to identify DNA from a mixture of organisms **AKA: Amplicon or Marker gene sequencing.**
- eDNA: DNA does not have to come directly from the organism.

Major Metagenomic

Workflows



Every single step introduces biases

Major Workflows

Metabarcoding

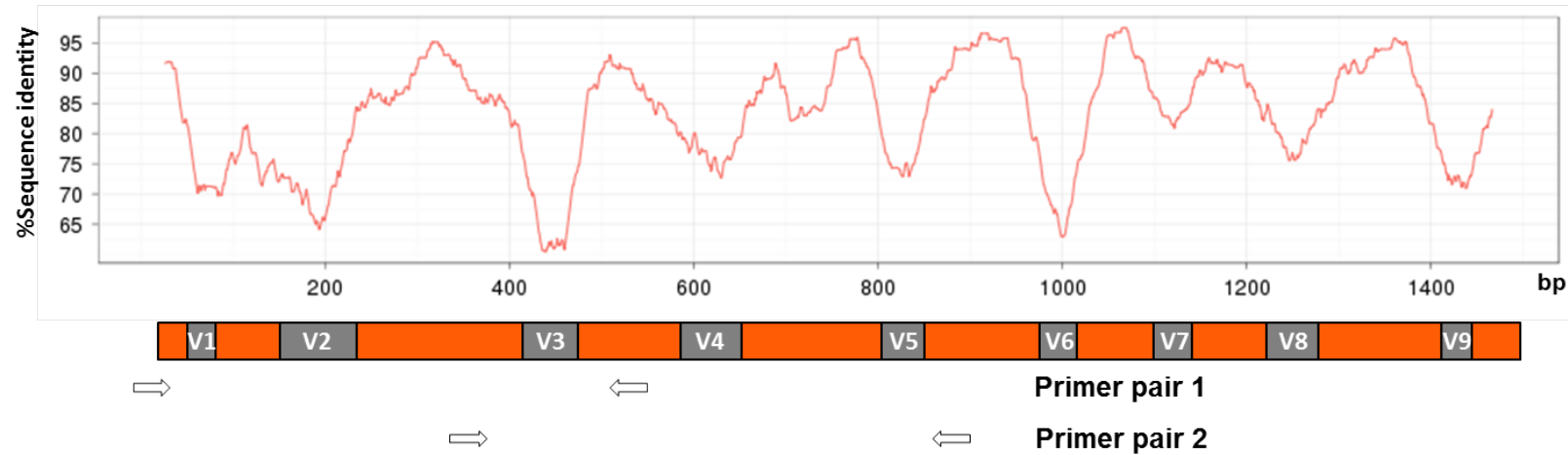
- Positive Aspects
 - Efficient
 - Possible to process 1000 of samples
- Negative Aspects
 - PCR Limitations
 - Limited by databases
 - Not directly related to ecosystem function

Whole Metagenome Shotgun

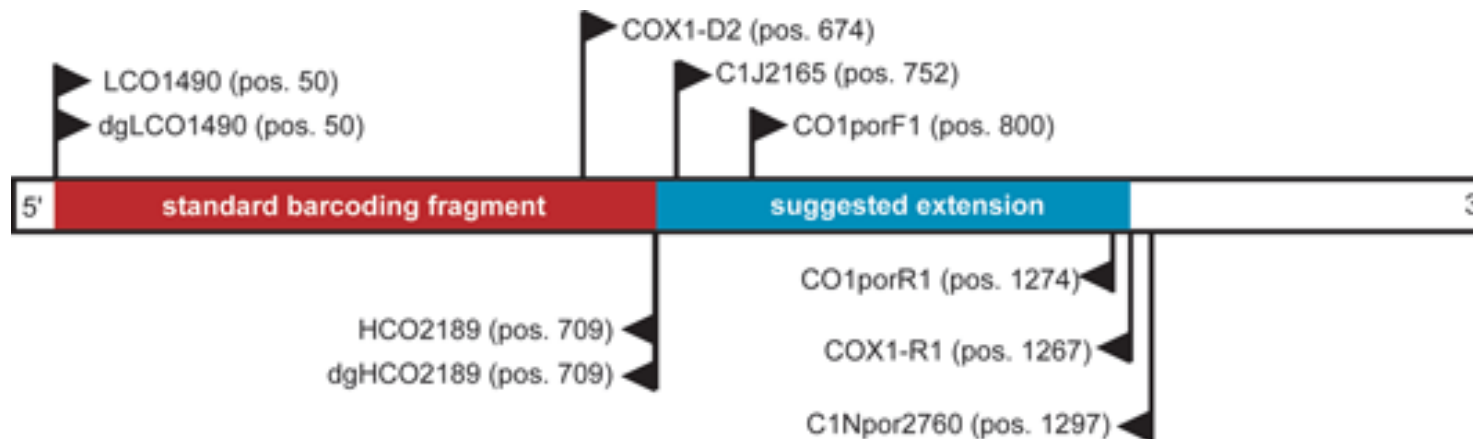
- Positive Aspects
 - Produces “functional” and taxonomic data
 - Not restricted by PCR All “organisms”
- Negative Aspects
 - Inefficient
 - Limited by databases

Marker Genes/Barcodes

Ribosomal RNA



Mitochondrial DNA



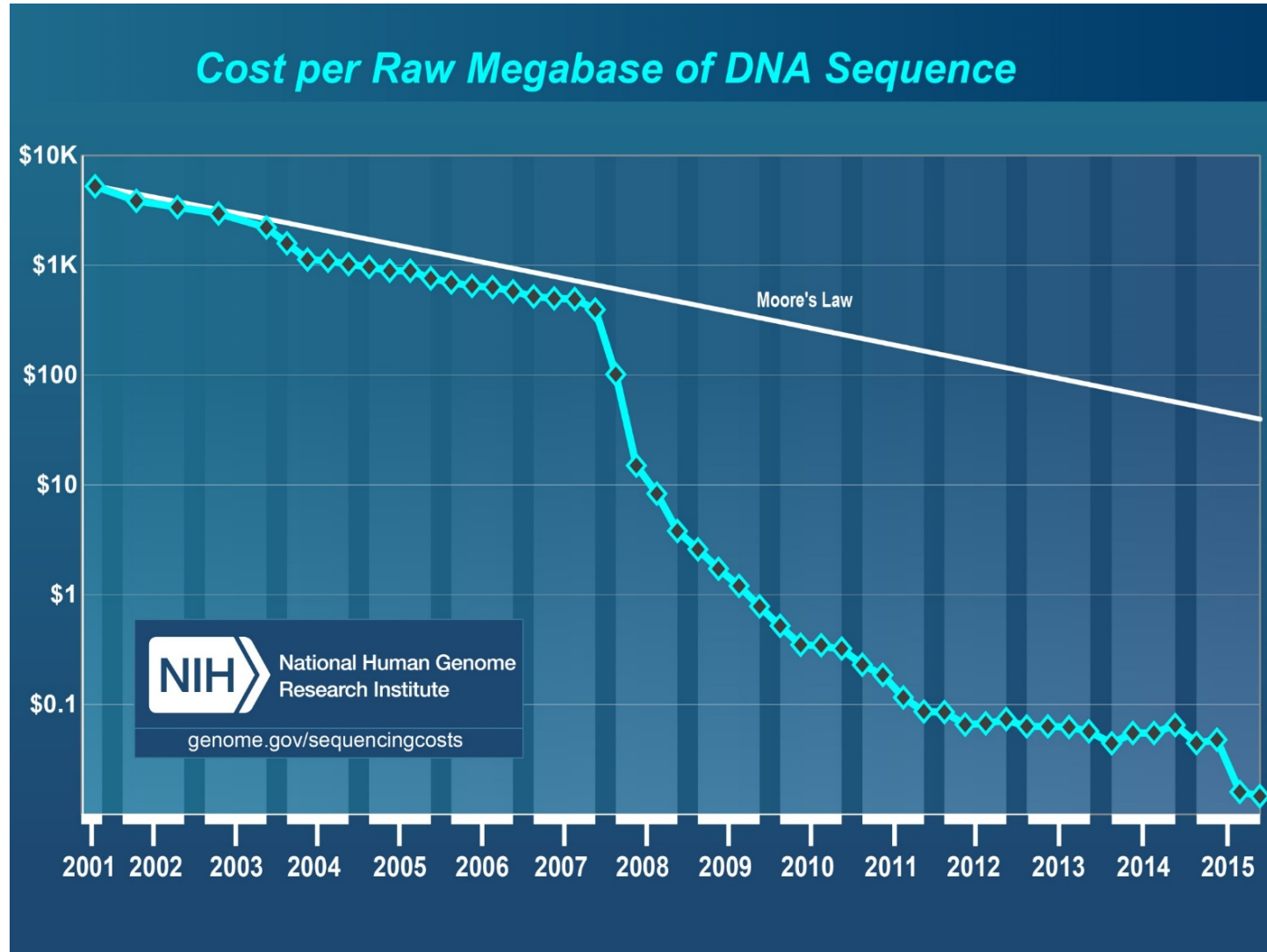
The generation of Genomic Data

Next Generation DNA sequencing technologies

A sea-change in genome-enabled biology

- The last several years have seen the development of fundamentally new sequencing technologies.
 - A process that continues...
- These technologies produce more data, and better data in many fewer steps.
- These changes make genomic analysis a core approach in diverse areas of biology

“Sequencing is now the cheapest part of sequencing” C. Titus Brown



Cost of Meta-barcoding

- Sampling \$10 -\$10,000 per sample
- DNA Isolation – \$5-\$10
- PCR \$1-\$5
- Sequencing \$ 1 for 50,000 reads
- Analysis: \$150/hour

Your experimental design should keep these relative costs in mind

Next Generation Sequencing Technologies

- The dominant sequencing approach today is Sequencing by Synthesis (SBS) from Illumina. This technology produces the vast majorities of datasets and will be the method used for your sequencing.
- Other interesting and useful approaches include the Single Molecule Real Time (SMRT) Sequencing approach by Pacific Biosciences and Nanopore Sequencing by Oxford Nanopore Technologies. Both of these technologies can produce much longer “reads” or continuous runs of sequence information.

Illumina Sequencing by synthesis



Watch a Video showing the basics
of the technology

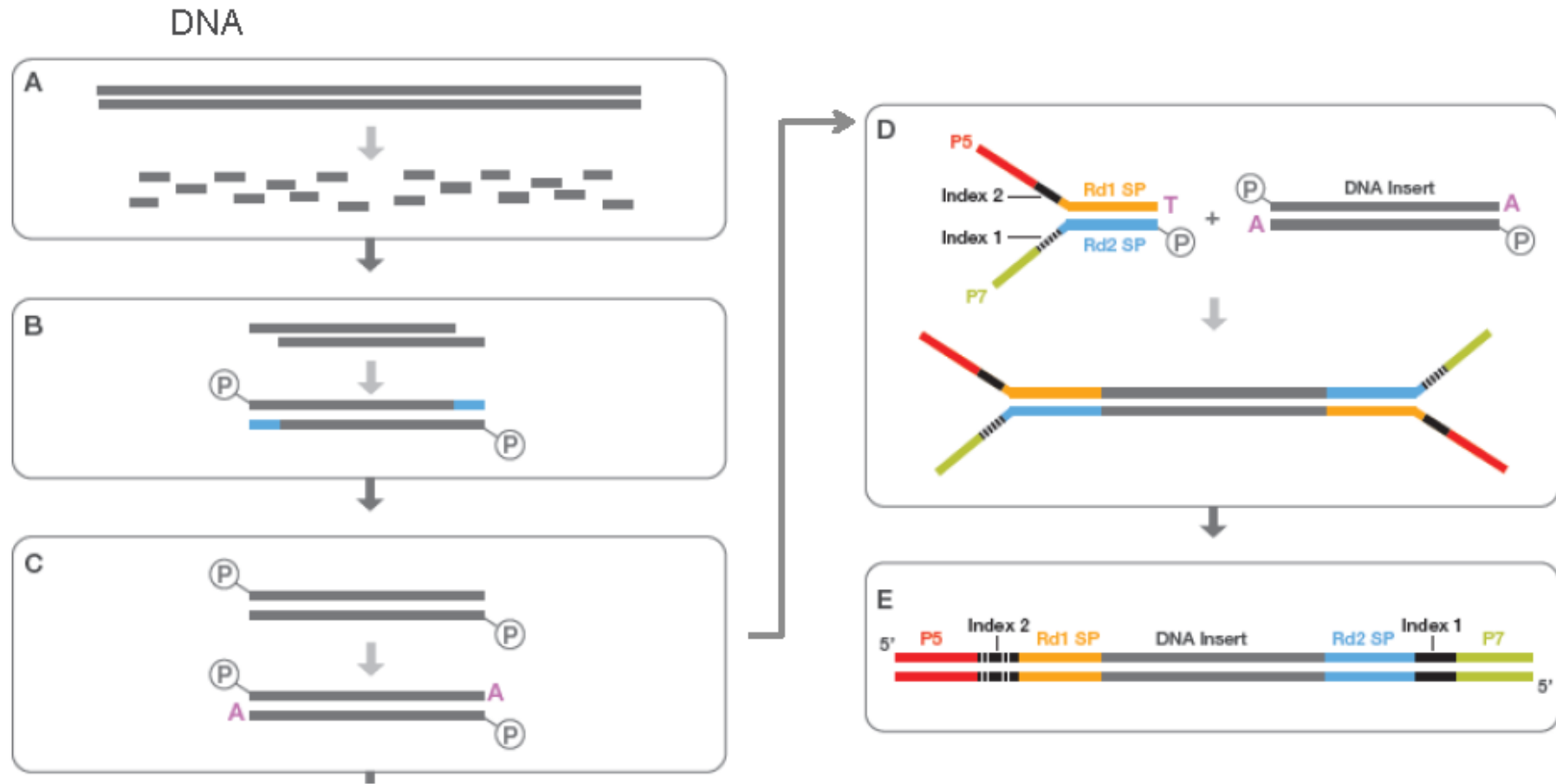
Serious

<https://www.youtube.com/watch?v=womKfikWlxM>

Not so serious

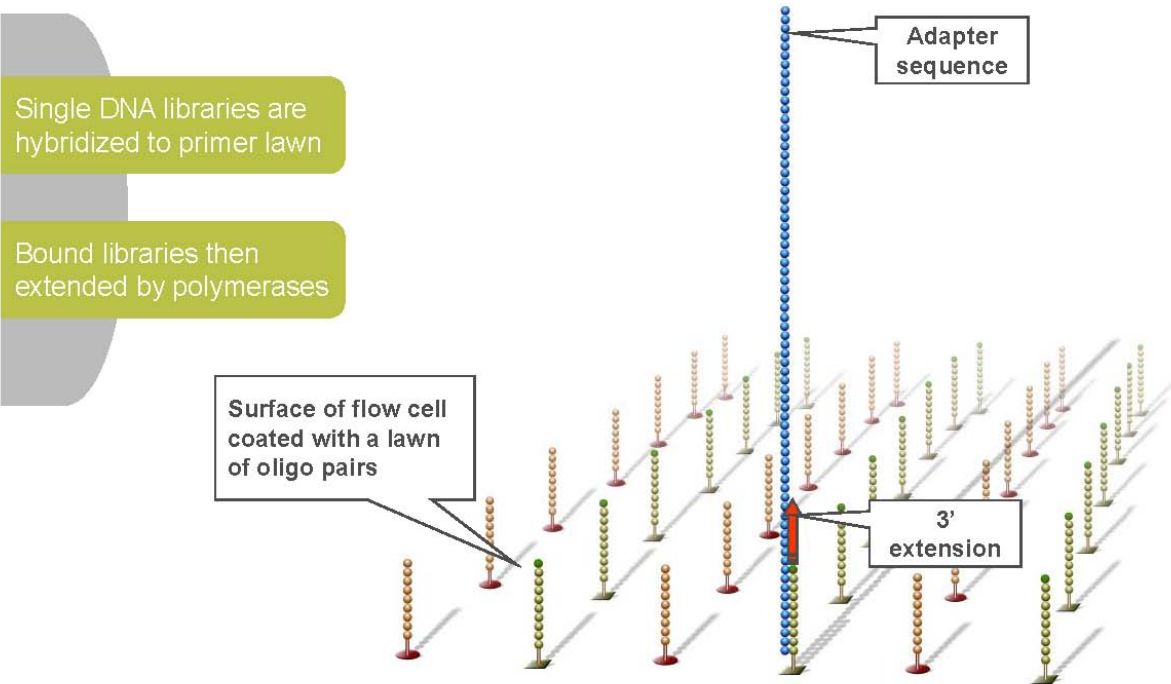
<https://www.youtube.com/watch?v=-7GK1HXwCtE>

Basic Process

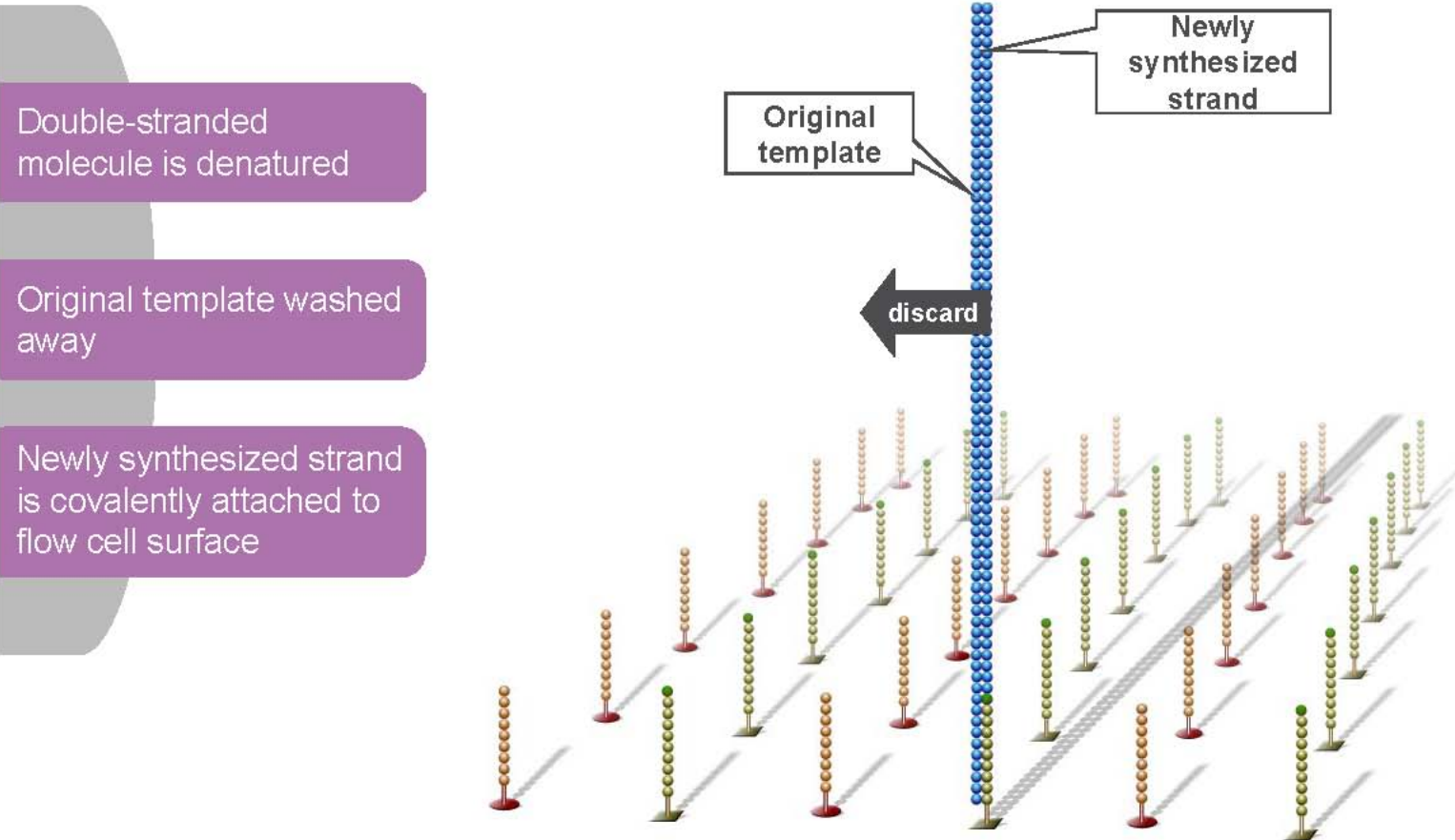


Sequencing on the inside surface of a flowcell

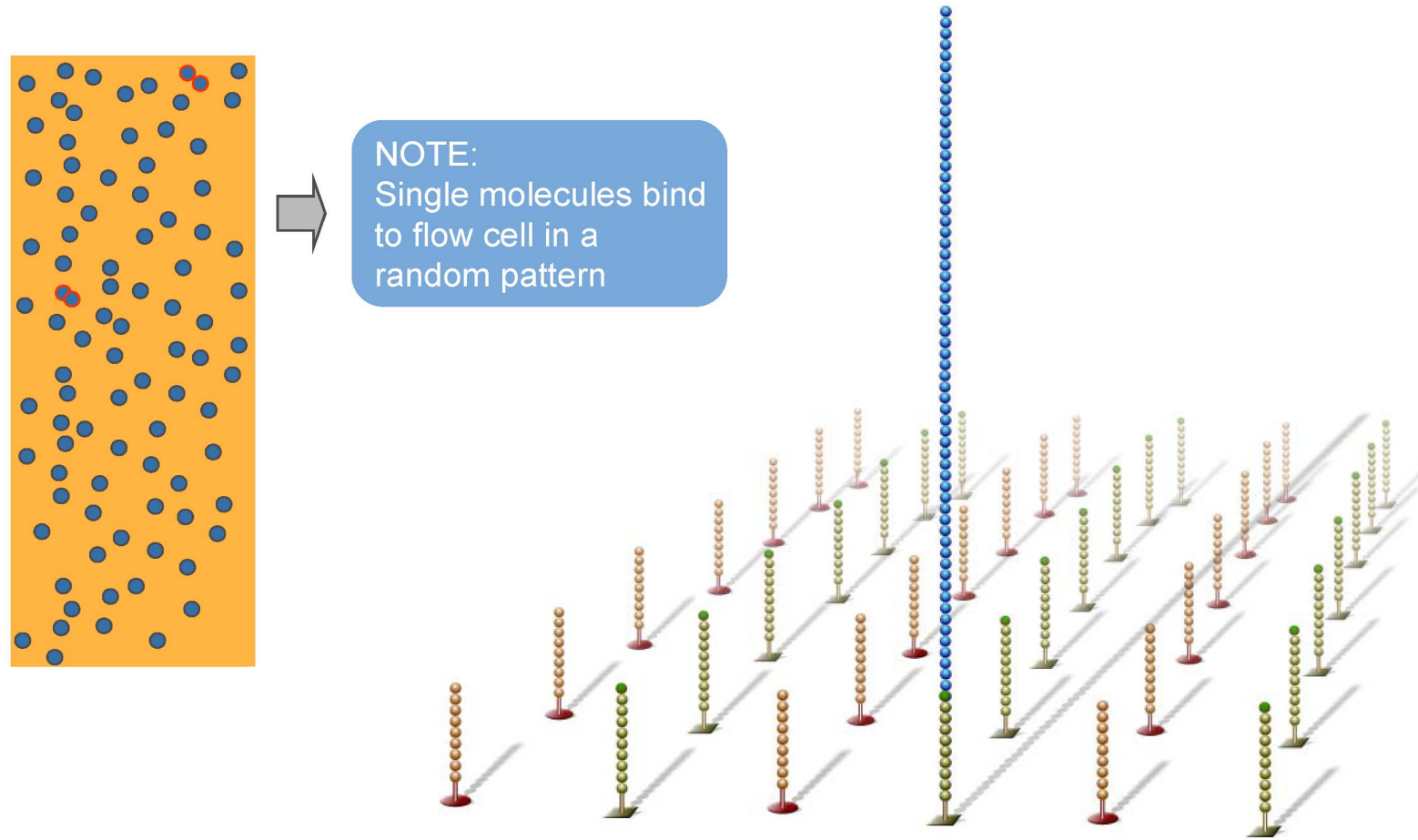
Hybridize Fragment & Extend



Denature Double-stranded DNA



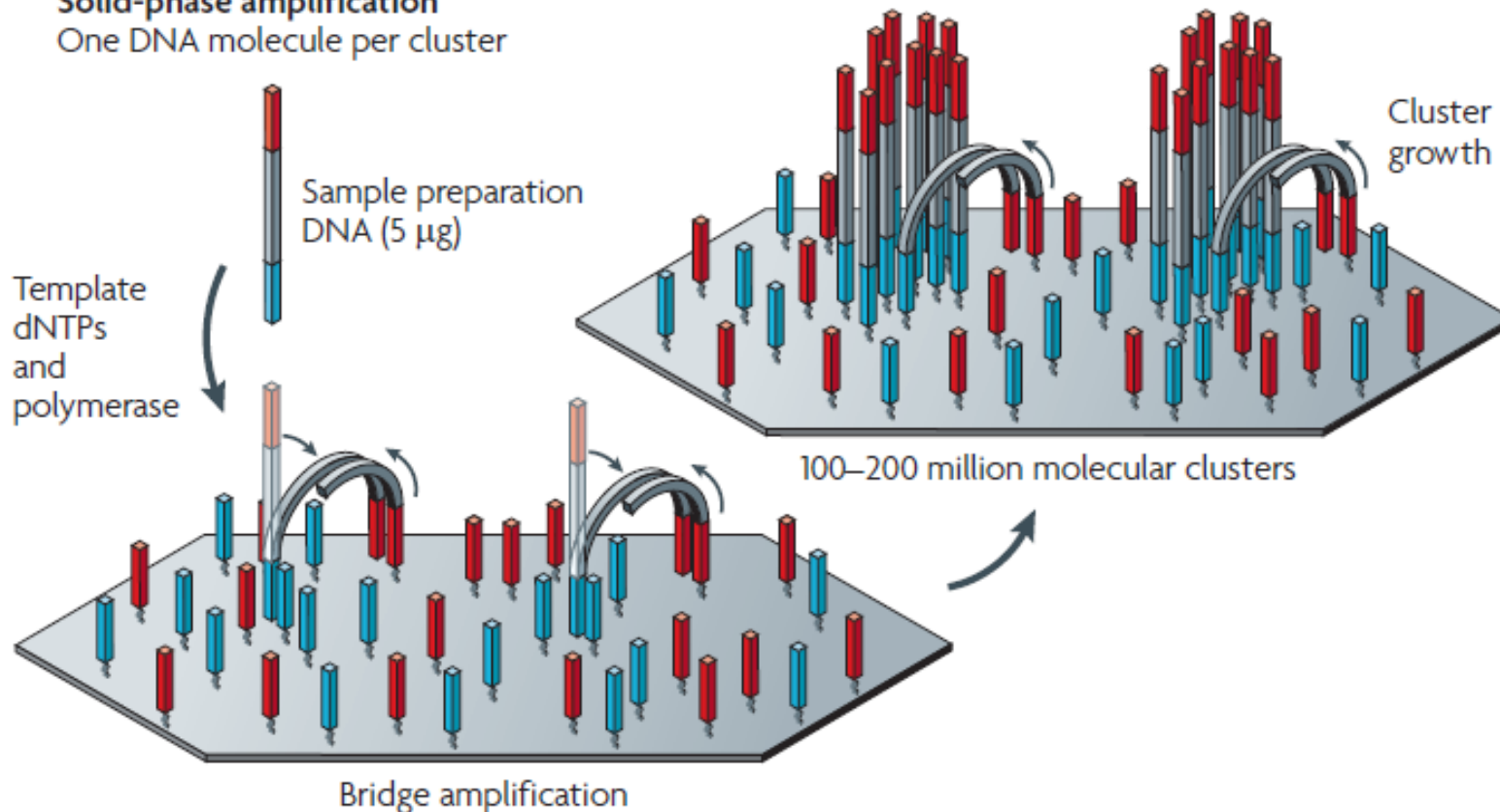
Hybridize Fragment & Extend



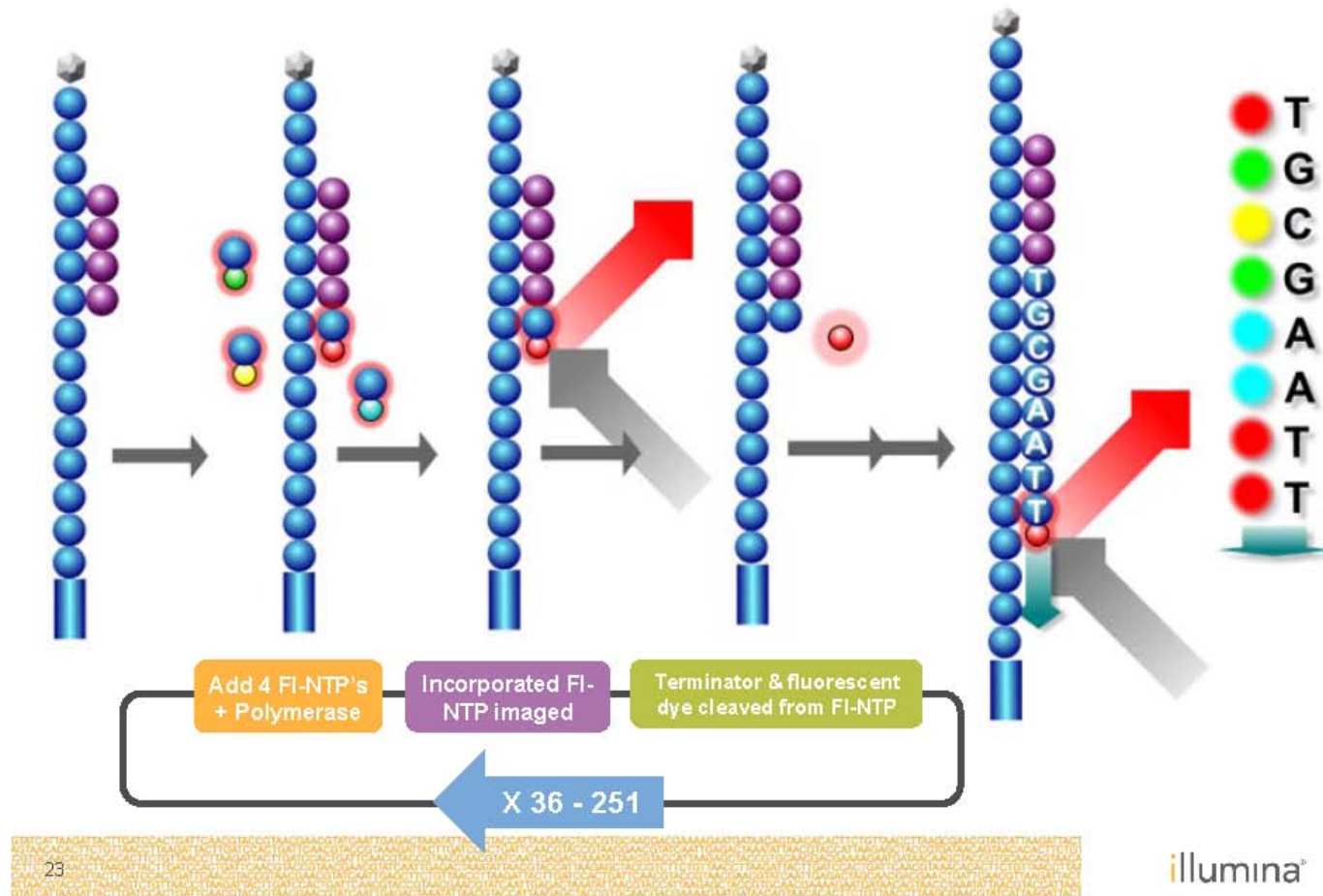
Making sequencing massively parallel

- Solid Phase

b Illumina/Solexa
Solid-phase amplification
One DNA molecule per cluster

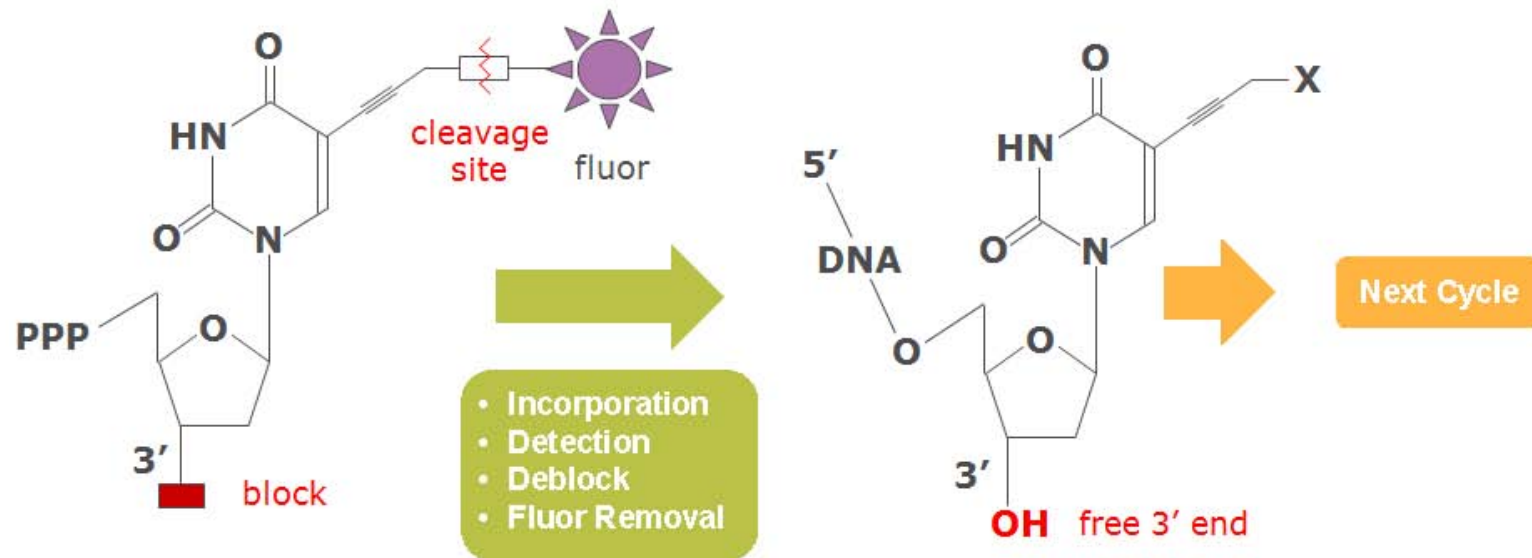


Sequencing By Synthesis



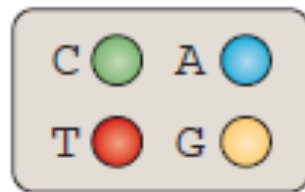
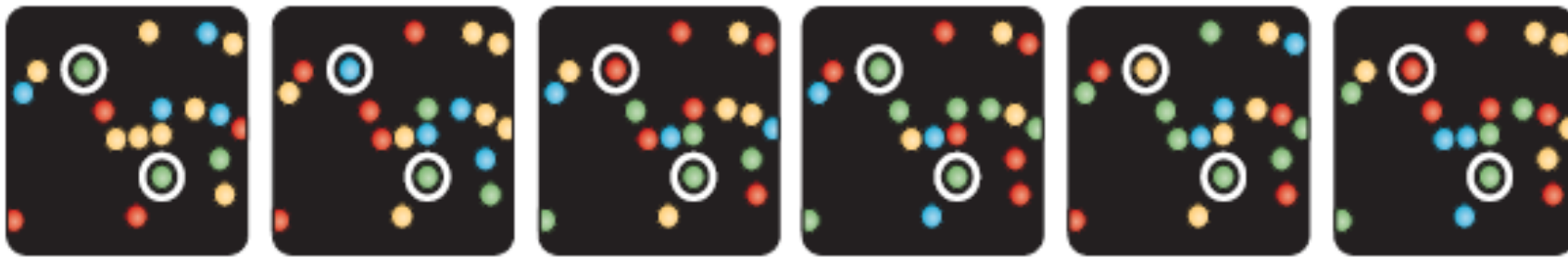
Reversible Terminator Chemistry

- All 4 labeled nucleotides in 1 reaction
- Higher accuracy
- No problems with homopolymer repeats



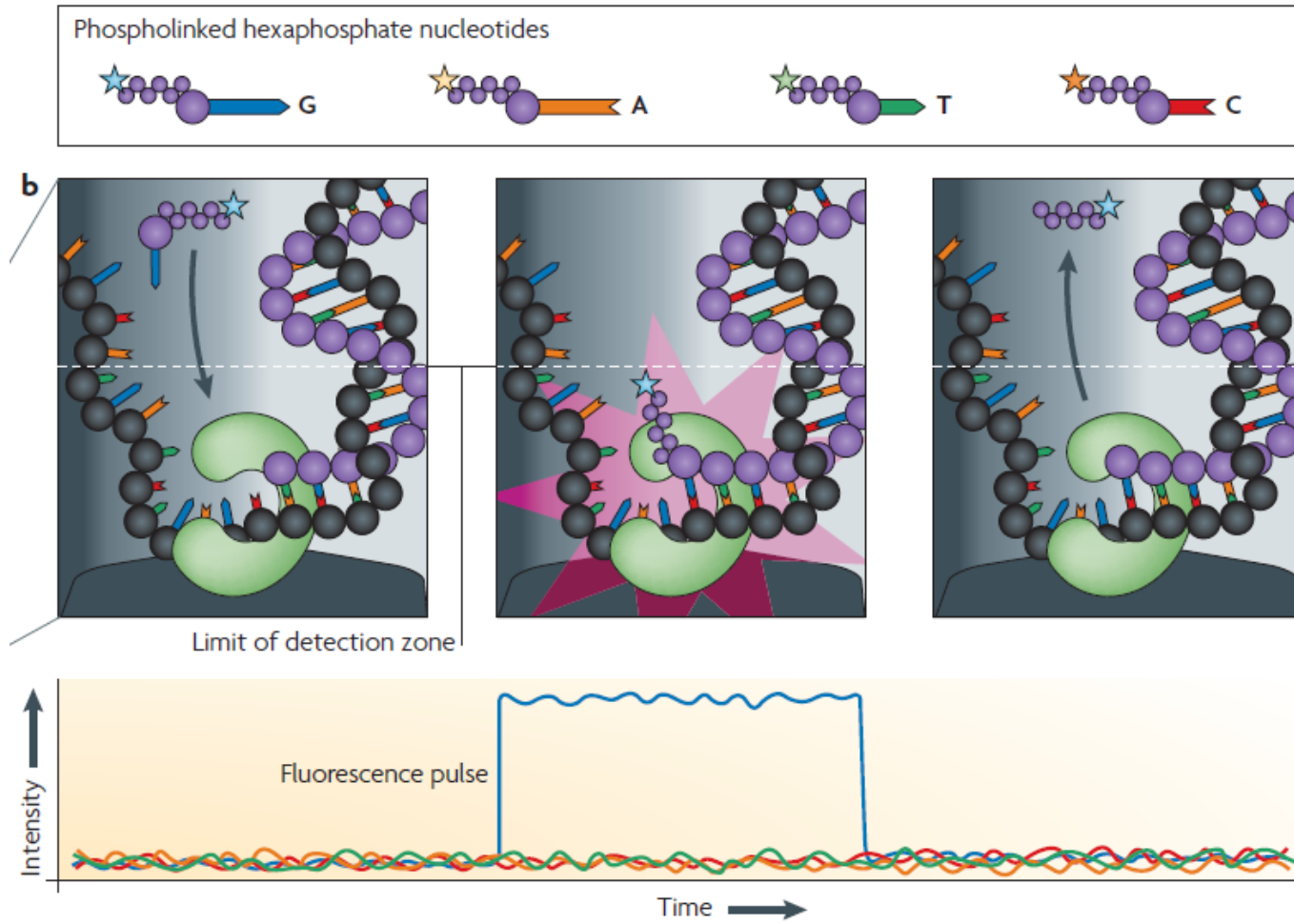
Illumina Stargazing

b



Top: CATCGT
Bottom: CCCCCC

Pac Bio

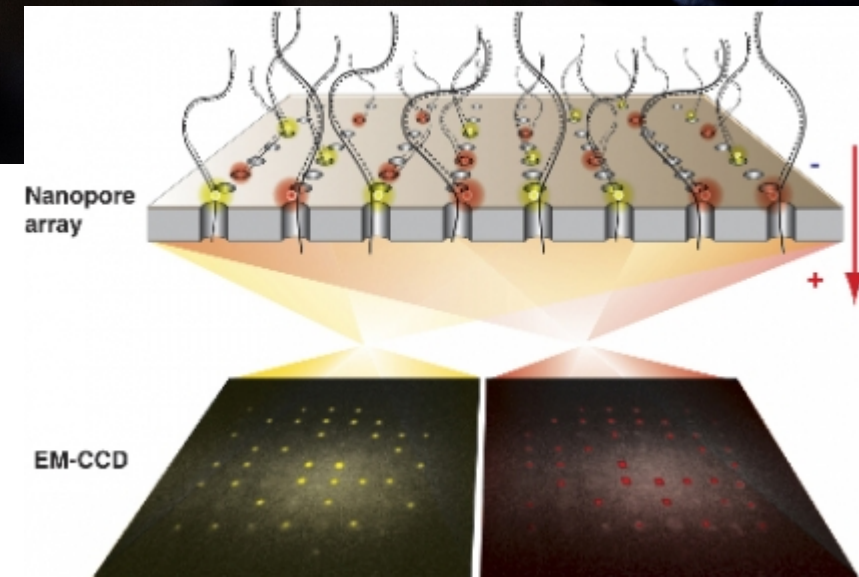
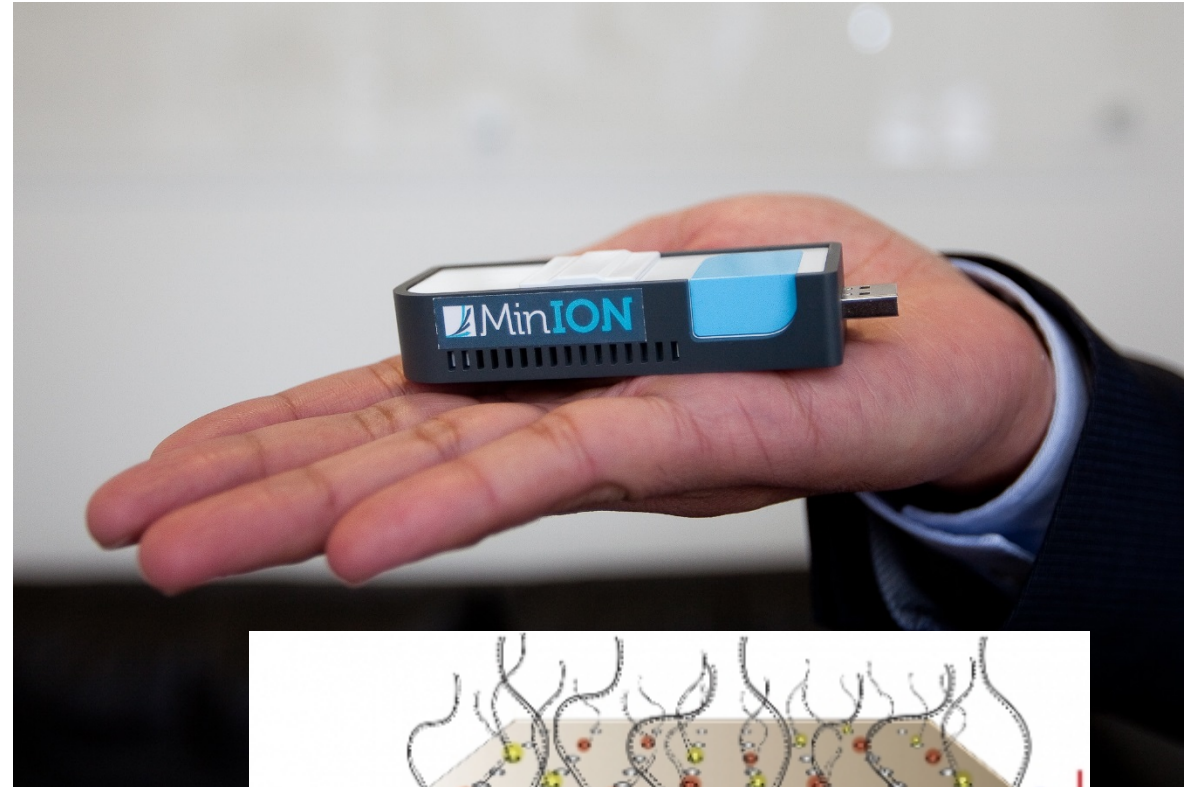
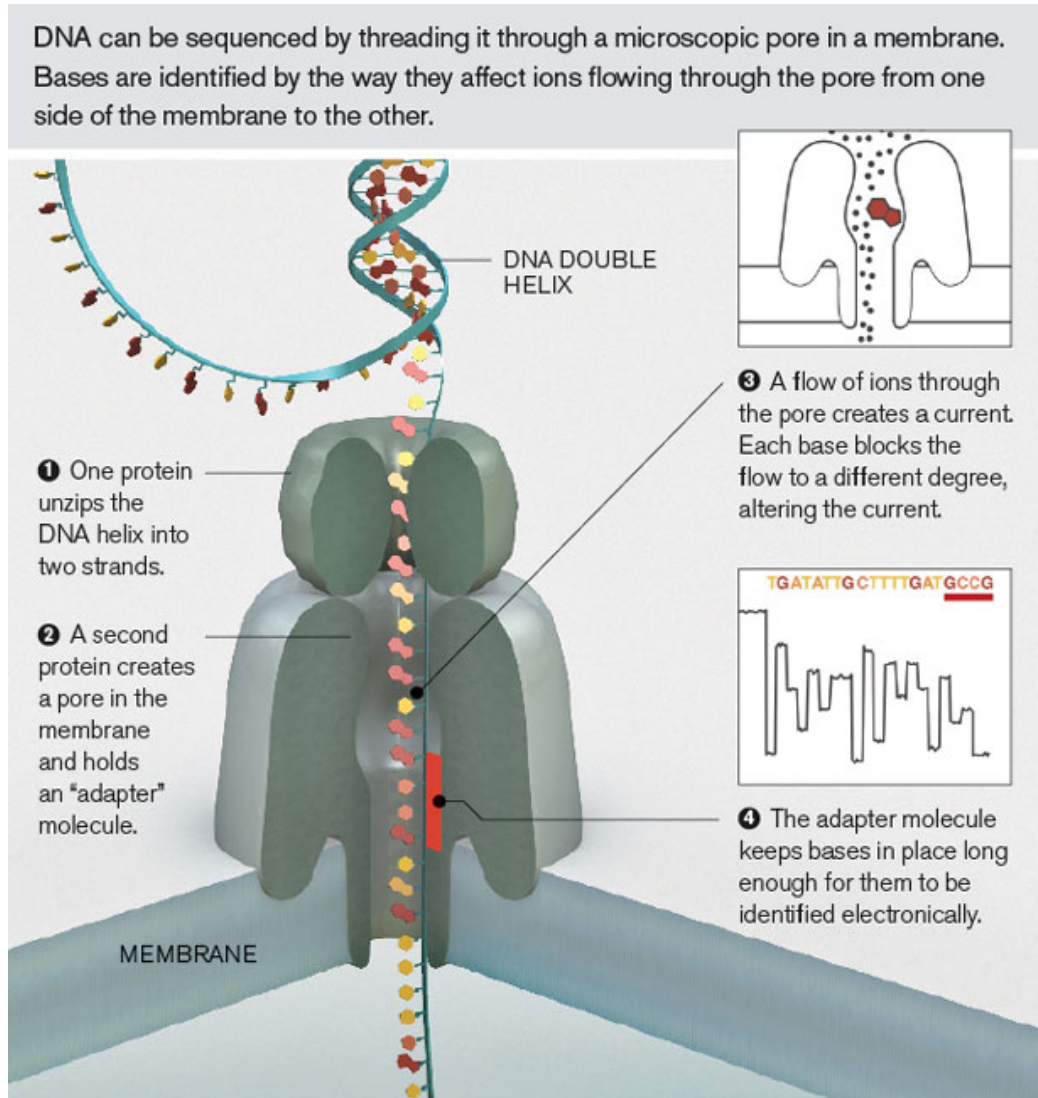


Watch a Video showing the basics of the technology

<https://www.youtube.com/watch?v=hBr0TJg-N6U>

Nanopore Sequencing

<https://www.youtube.com/watch?v=3UHw22hBpAk>



Data structures and conventions

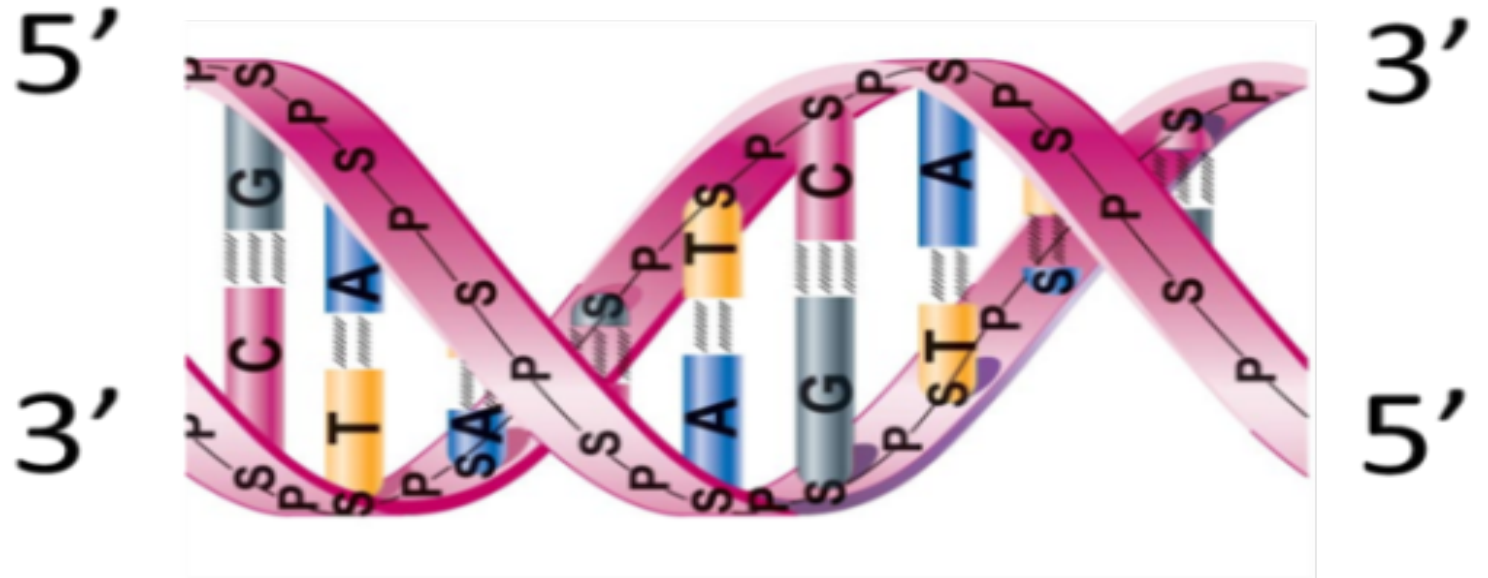
- The interaction of computers and data requires that we follow strict conventions for how we communicate genomic data.
- Bioinformatics relies on the use of common file formats.

Some simple concepts about DNA sequence data

In bioinformatics, DNA or RNA is depicted in a single string from 5' to 3' and only one strand shown (saves space). Similarly, protein sequences are always written from the N-terminus to the C-terminus.

“It has not escaped our notice that the specific pairing we have postulated immediately suggests a possible copying mechanism for the genetic material.”

In J.D. Watson and F.H.C. Crick, 'A Structure for Deoxyribose Nucleic Acid,' Letter in *Nature* (25 Apr 1953)



The FASTA File Format

Protein FASTA

```
>gi|129295|sp|P01013|OVAX_CHICK GENE X PROTEIN (OVALBUMIN-RELATED)
QIKDLLVSSSTDLDTTLLVLVNAIYFKGMWKTAFAEDTREMPFHVTKQESKPVQMMCMNNSFNVATLP
AEKMKILELPFASGDL SMLVLLPDEVSDLERIEKTINFEKLTEWTNPNTMEKRRVKVYLPQMKIEEKYN
LTSVLMALGMTDLFIPSANLTGISSAESLKISQAVHGAFMELSEDGIEMAGSTGVIEDIKHSPESEQFR
ADHPFLFLIKHNPTNTIVYFGRYWSP
```

Nucleotide FASTA

```
>SRR014849.1 EIXKN4201CFU84 length=93
GGGGGGGGGGGGGGGGGGGGCTTTTTTTGTTTGGAACCGAAAGG
GTTTTGAATTTCAAACCCTTTTCGGTTTCCAACCTTCCAA
AGCAATGCCAATA
```

Quality Files with PHRED quality scores are often created in parallel

```
>SRR014849.1 EIXKN4201CFU84 length=93
18 10 5 3 2 1 1 1 1 1 1 1 1 1 1 22 37 31 22 16 11 6 1 26 34 30 11 33 26 30 21
33 26 25 36 32 16 36 32 16 36 32 20 6 24 33 25 30 25 2 24 36 32 15 35 31 17
36 32 20 6 25 29 20 30 25 4 32 26 32 23 32 26 30 24 33 26 35 31 14 28 27 30 22
28 24 27 17 32 23 28 28
```

Extension	Meaning
.fna	fasta nucleic acid
.ffn	FASTA nucleotide of gene regions
.fasta (.fas)	generic fasta
.faa	fasta amino acid
.fq (.fastq)	FASTQ file

Quality scores and meaning. Quality scores = $-\log_{10}(\text{probability of error})$

Phred Quality Score	Probability of incorrect base call	Base call accuracy
10	1 in 10	90 %
20	1 in 100	99 %
30	1 in 1000	99.9 %
40	1 in 10000	99.99 %
50	1 in 100000	99.999 %

FASTQ file format

The general format is similar in style to FASTA but has quality score attached in same file and given as ASCII characters to save space. This is the common format for raw sequence data.

@title and optional description

sequence line(s)

+optional repeat of title line

quality line(s)

@SRR014849.1 EIXKN4201CFU84 length=93

GGGGGGGGGGGGGGGGGGGGCTTTTTTTGTTTGGAACCGAAAGG

GTTTTGAATTTCAAACCCTTTTCGGTTTCCAACCTTCCAA

AGCAATGCCAATA

+SRR014849.1 EIXKN4201CFU84 length=93

3+&\$#"7F@71,";C?,B;?6B;:EA1EA

1EA5'9B:?:#9EA0D@2EA5':>5?:%A;A8A;?9B;D@

/=<?7=9<2A8==

Sequences are expected to be represented in the standard IUB/IUPAC amino acid and nucleic acid codes.

The nucleic acid codes are:

A adenosine	C cytidine	G guanine
T thymidine	N A/G/C/T (any)	U uridine
<i>K G/T (keto)</i>	<i>S G/C (strong)</i>	<i>Y T/C (pyrimidine)</i>
<i>M A/C (amino)</i>	<i>W A/T (weak)</i>	<i>R G/A (purine)</i>
<i>B G/T/C</i>	<i>D G/A/T</i>	<i>H A/C/T</i>
<i>V G/C/A</i>	<i>- gap of indeterminate length</i>	

The amino acid codes are:

A alanine	P proline
B aspartate/asparagine	Q glutamine
C cystine	R arginine
D aspartate	S serine
E glutamate	T threonine
F phenylalanine	<i>U selenocysteine</i>
G glycine	V valine
H histidine	W tryptophan
I isoleucine	Y tyrosine
K lysine	Z glutamate/glutamine
L leucine	X any
M methionine	* translation stop
N asparagine	<i>- gap of indeterminate length</i>

GFF files describe the position of genes in a FASTA file.

The **general feature format** (**gene-finding format**, or **generic feature format**, **GFF**) is a [file format](#) used for describing [genes](#) and other features of [DNA](#), [RNA](#) and [protein](#) sequences. The [filename extension](#) is .GFF.

In a GFF file there are 9 columns of information for each feature in a DNA sequence (FASTA file).

Position	Name	
1	sequence	
2	source	
3	feature	
4	start	
5	end	
6	score	
7	strand	
8	frame	
9	attributes.	

Sequence Alignment Map (SAM) Files or BAM for binary version

11 mandatory fields in each row: separated by spaces in a text file

1=Read Name, 2=bitwise flag, 3=Reference Sequence,
4=start position of read in reference, 5= Map Quality, 6=
CIGAR string, 7= paired end read reference, 8=position of
mate, 9=distance between reads, 10=Read sequence,
11=read quality.

```
1:497:R:-272+13M17D24M 113 chr1 497 37 37M chr15 100338662 0
```

```
CGGGTCTGACCTGAGGAGAACTGTGCTCCGCCTTCAG
```

```
0;==--==9;>>>>>=>>>>>>>>>>>=>>>>>>>>>>>
```

Sequencing and assembling a genome

The process of sequencing genomes typically involves breaking the genome up and then attempting to put Humpty-Dumpty back together again.



Why Sequence a Genome?

To establish a gene catalogue: The parts list needed for all functions in a cell or organism.

To establish a reference platform for:

- functional analysis (e.g. gene expression)
- investigating DNA sequence variation

To Investigate Biodiversity and ecosystem function
(e.g. Metagenomics)

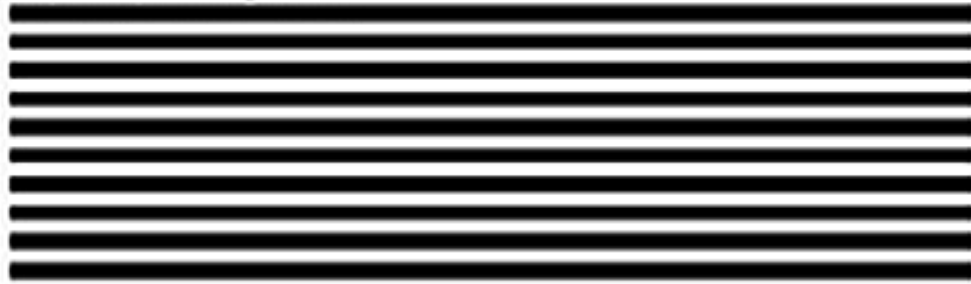
To explore broader issues such as the Ethical, Legal and Social Implications (ELSI).

There are many diverse reasons for sequencing genomes.



Sequencing a genome

Multiple copies of genome



Sheared random fragments



Size fractionated fragments



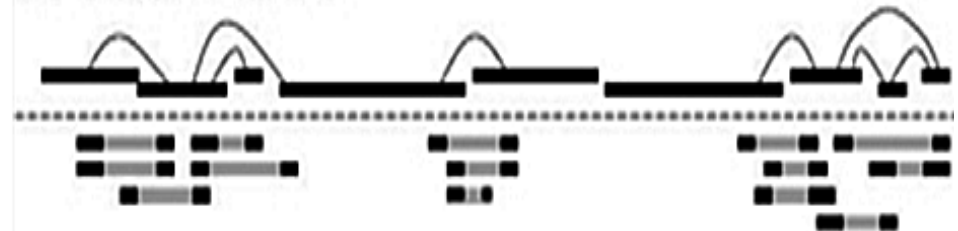
Reads



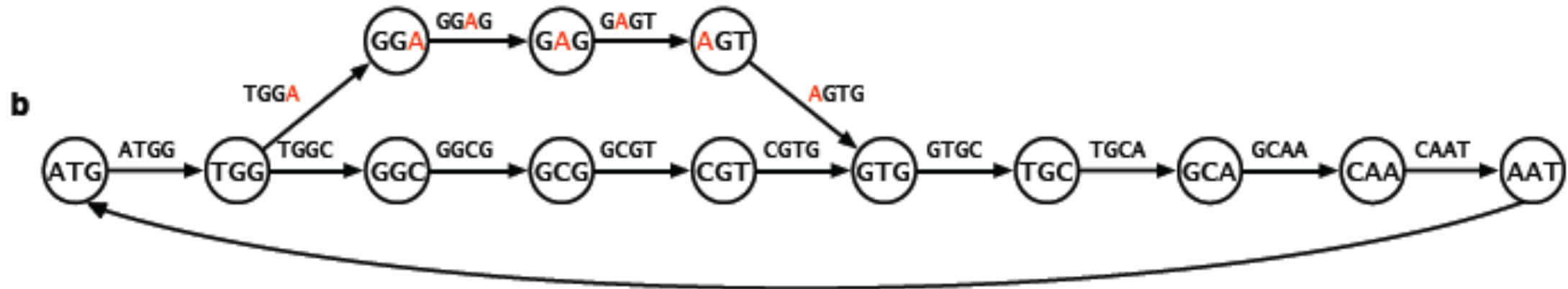
Contigs



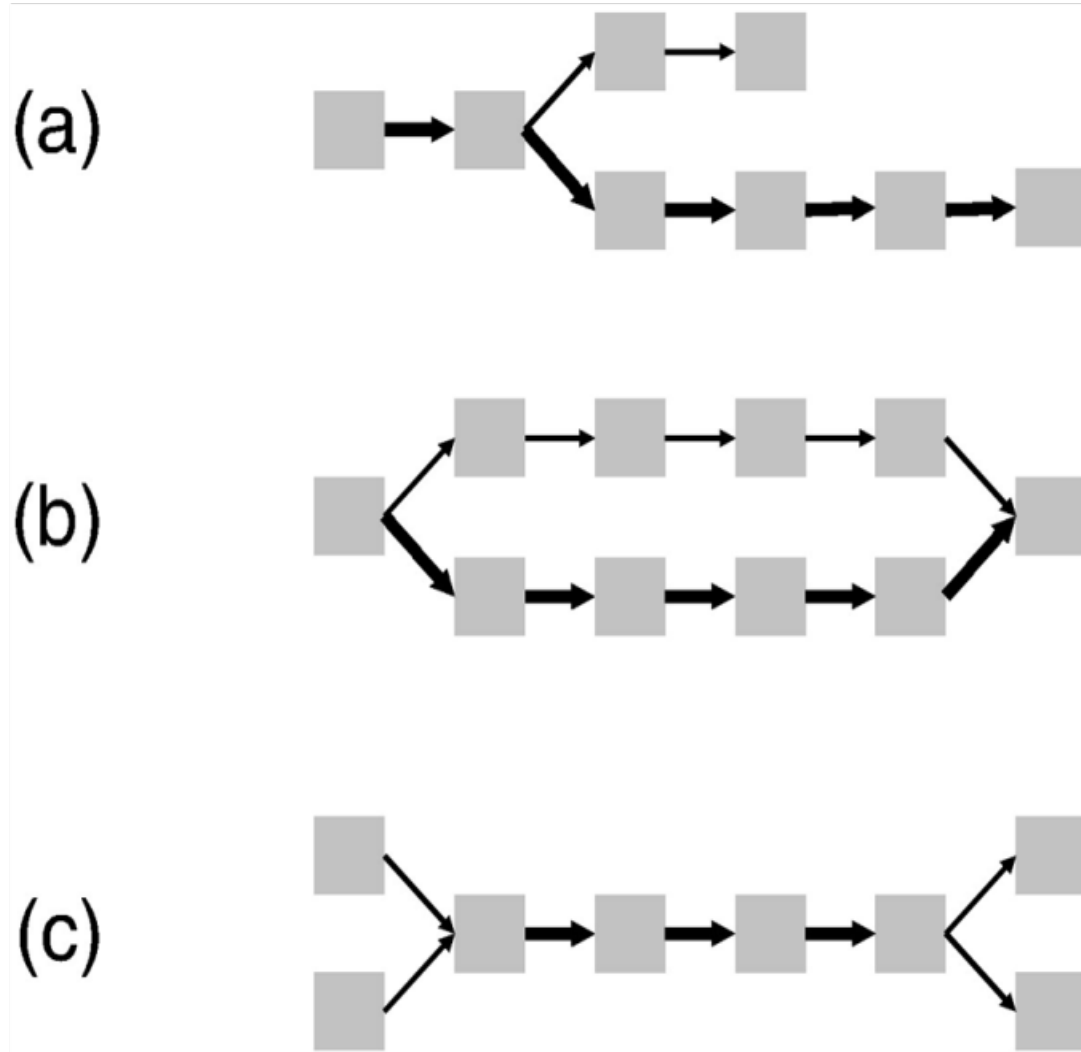
Scaffolds(Super contigs)



In real data, there will be sequencing errors and polymorphisms. In the figure below, a single base difference results in two paths that diverge and then converge. This could be caused by a sequencing error in the middle of a read or polymorphisms. If this represents heterozygosity, the paths may have equal representation.



In the diagram below, the path complexities include spurs that will result from a sequencing error at the end of a read, bubbles as shown above and “rope ends”. Rope ends depict two different paths that share a common set of k-mers. These are the result of repeats that are greater than the length of a k-mer.



Spur

Bubbles

Rope ends

Assembly output and assessing the quality of an assembly:

De novo assembly produces two main outputs.

- **Contig file** (FASTA or multi FASTA)
- **SAM file:** SAM (Sequence Alignment/Map) format is a generic format for storing large nucleotide sequence alignments.

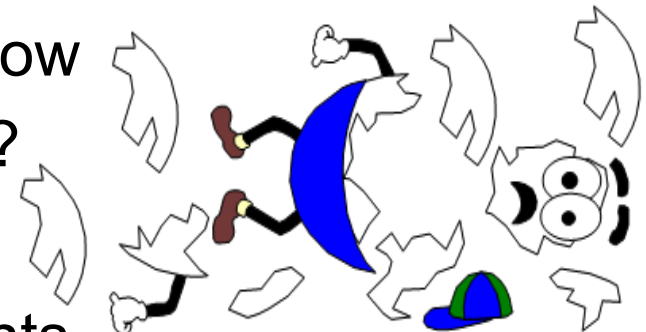
How do we assess the quality of an assembly? There are three basic measures of assembly quality:

- 1.N50:** A measure of average contig size. Specifically, $\frac{1}{2}$ of the genome is assembled in contigs of this size or greater.
- 2.Depth of coverage:** A measure of how much information is available for each base call.
- 3.Completeness of the gene catalogue:** What percentage of the genes are assembled into contigs?

Key challenges for genome assembly:

Intrinsic Challenges:

- 1. *Heterozygosity*:** The alleles of a gene are not the same, yet we typically force them into a single consensus sequence.
- 2. *Paralogy vs. Alleleism*:** Genes come from other genes by a process of duplication. This results in two or more similar genes in an organism. There are two alleles in a diploid organism that are very similar. How do you tell a duplicated gene from alleles of a gene?
- 3. *Sequence complexity*:** Simple sequence repeats (SSR), large-scale repeats like transposable elements (TEs).



Extrinsic Challenges:

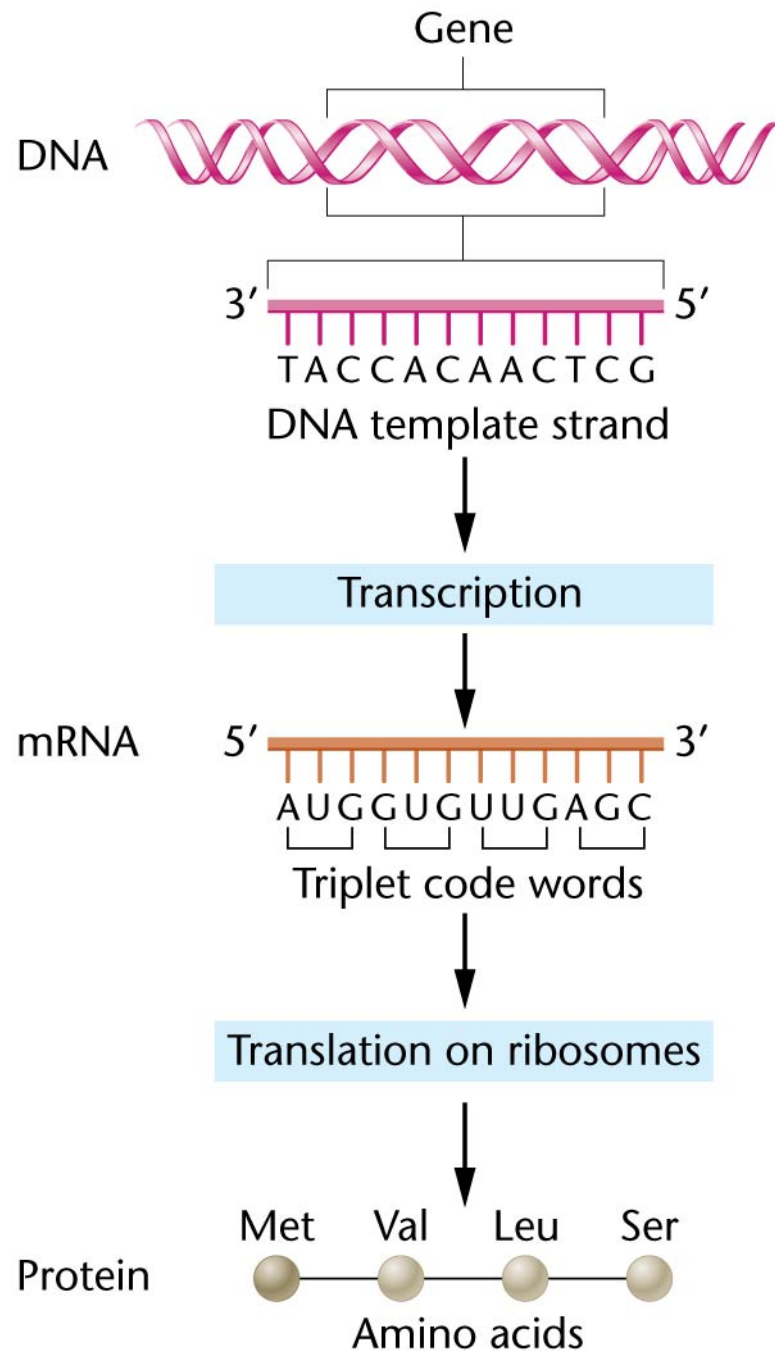
- 1. *Quality of DNA sequences (sequencing errors):*** Each sequencing technology has specific patterns of error. For example, pyrosequencing typically has high error rates associated with runs of a single nucleotide.
- 2. *Length of DNA sequence reads:*** Shorter reads are less likely to be unique or to include many unique K-mers (see below).
- 3. *Coverage:*** Depth of Coverage is a random process at best. Consequently some regions of the genome will have low levels of coverage.
- 4. *Memory intensive:*** Inherently requires large amounts of RAM for assembly and storage for input and output.
- 5. *Software:*** Need for approaches that are flexible, user friendly and powerful.

Genome annotation and inferring function

Once we have assembled a genome into one or more large “contigs” how do we “read” the DNA sequences and predict the genes and their functions

Inferring Function from a DNA Sequence

- We use our understanding of cellular processes and evolution to predict the existence and function of genes in DNA sequences.
 - The near universal nature of the genetic code makes it possible to predict what protein sequences can be encoded by any DNA molecule.
 - Evolution allows us to compare genes and their proteins from one species to another.
 - When we have demonstrated the function of a gene in a model organism we often assume it will serve the same or similar role in other species



Most concepts in Bioinformatics rely on core knowledge of Genetics

		Second position					
		U	C	A	G		
First position (5'-end)	U	UUU <i>phe</i>	UCU <i>ser</i>	UAU <i>tyr</i>	UGU <i>cys</i>	U	Third position (3'-end)
		UUC	UCC	UAC	UGC	C	
		UUA	UCA	UAA <i>Stop</i>	UGA <i>Stop</i>	A	
		UUG	UCG	UAG <i>Stop</i>	UGG <i>trp</i>	G	
	C	CUU <i>leu</i>	CCU <i>pro</i>	CAU <i>his</i>	CGU <i>arg</i>	U	
		CUC	CCC	CAC	CGC	C	
		CUA	CCA	CAA <i>gln</i>	CGA	A	
		CUG	CCG	CAG	CGG	G	
	A	AUU <i>ile</i>	ACU <i>thr</i>	AAU <i>asn</i>	AGU <i>ser</i>	U	
		AUC	ACC	AAC	AGC	C	
		AUA	ACA	AAA <i>lys</i>	AGA <i>arg</i>	A	
		AUG <i>met</i>	ACG	AAG	AGG	G	
	G	GUU <i>val</i>	GCU <i>ala</i>	GAU <i>asp</i>	GGU <i>gly</i>	U	
		GUC	GCC	GAC	GGC	C	
		GUA	GCA	GAA <i>glu</i>	GGA	A	
		GUG	GCG	GAG	GGG	G	

Initiation Termination

Beyond protein coding genes

- Not all genes encode proteins
- How would you find the genes for transfer RNAs and ribosomal RNAs in a DNA molecule?
 - You can look for similar sequences identified in related organisms
 - You can consider their special features like secondary structures

