



Genoma: Representación estática de una sistema biológico.
Conjunto completo de genes en un organismo o sus organelos.



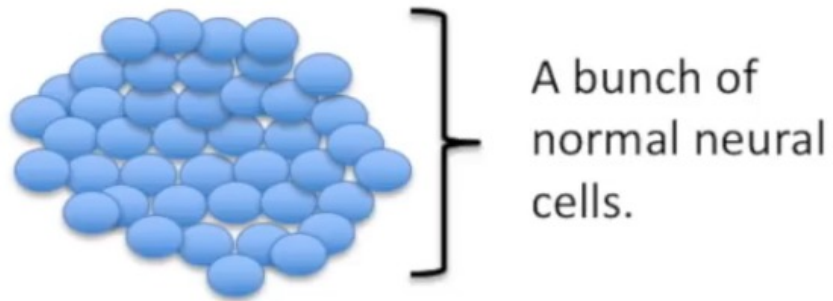
Transcriptoma: Representación de la expresión génica en estado fisiológico determinado.
Set completo de mARNs, presentes en una célula, tejido u órgano.



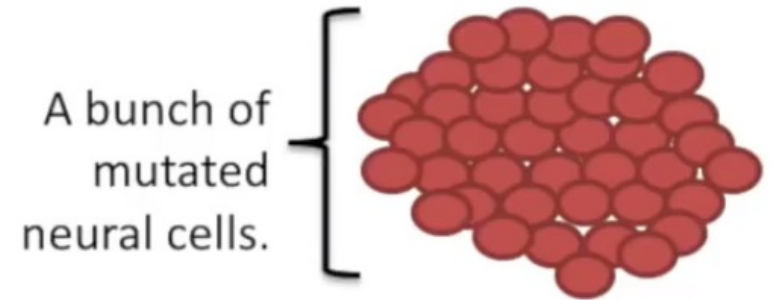
Proteoma: Visión integrada de las proteínas en un proceso biológico.
Son las proteínas codificadas por un genoma, presentes en una célula, tejido u órgano.

RNAseq

 = a normal neural cell

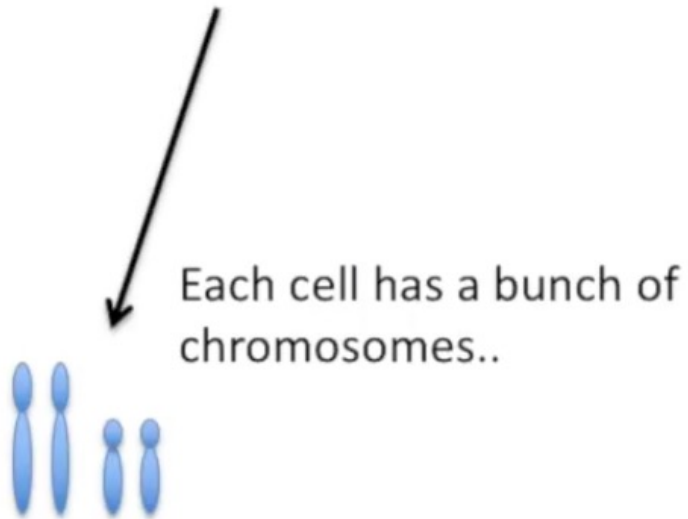
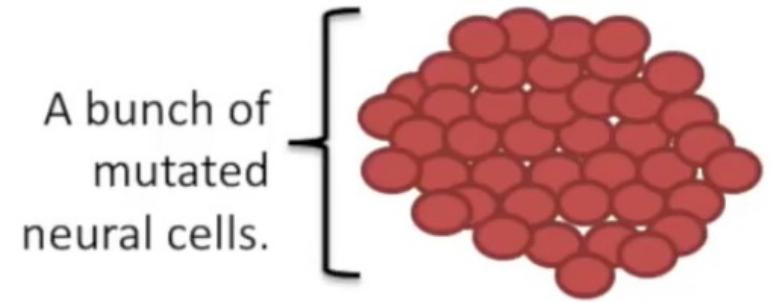
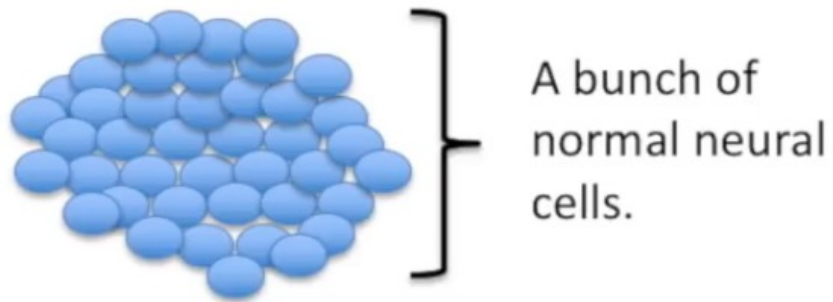


 = a mutated neural cell



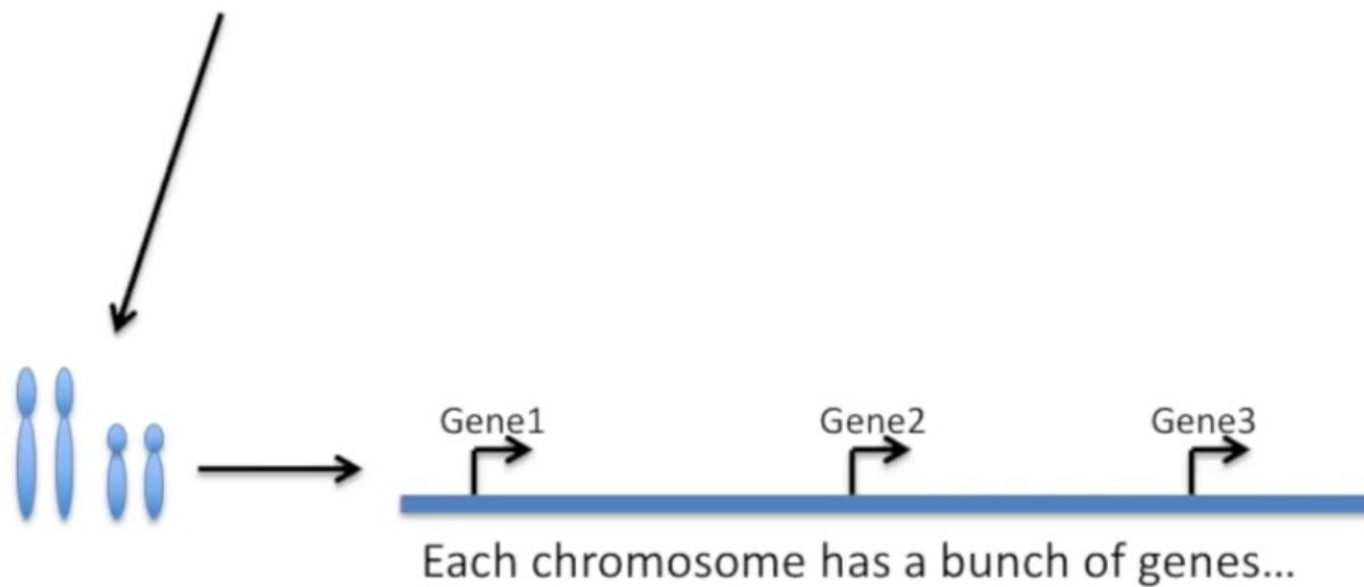
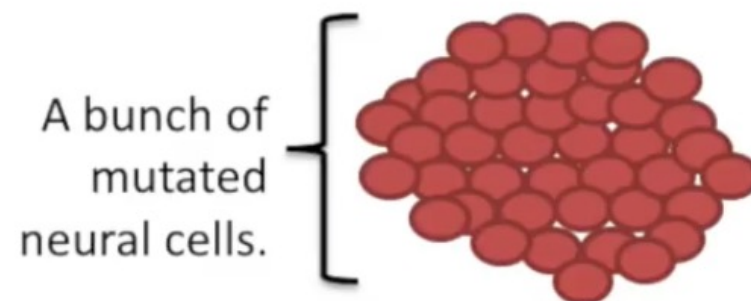
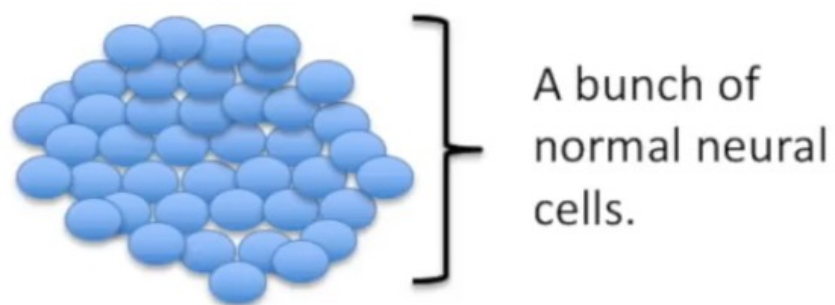
 = a normal neural cell

 = a mutated neural cell



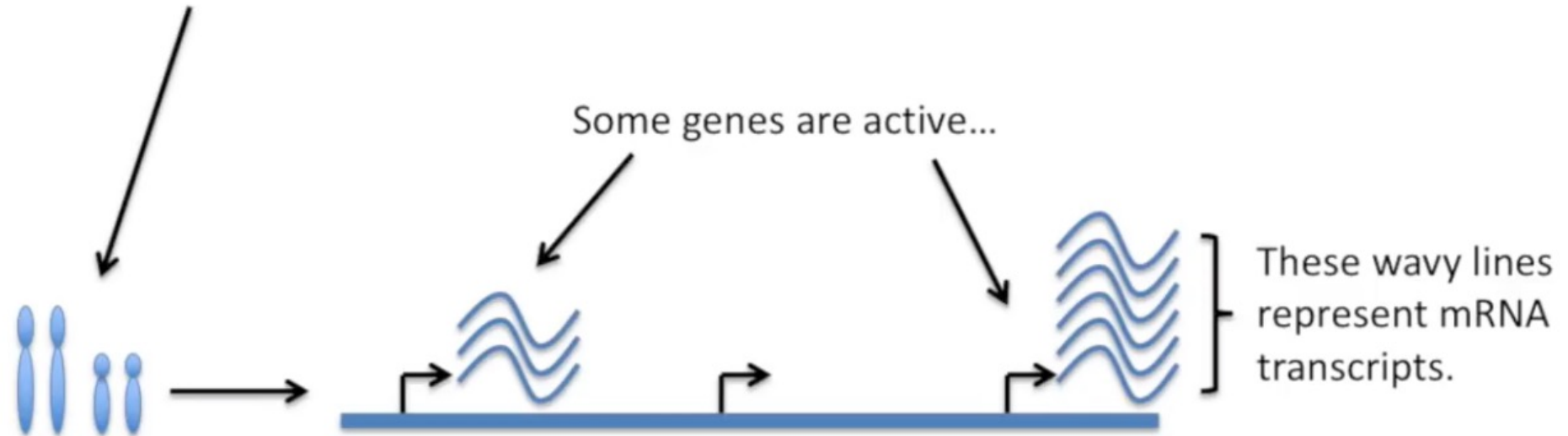
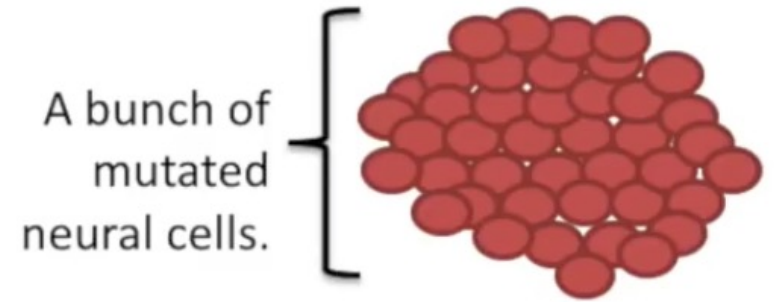
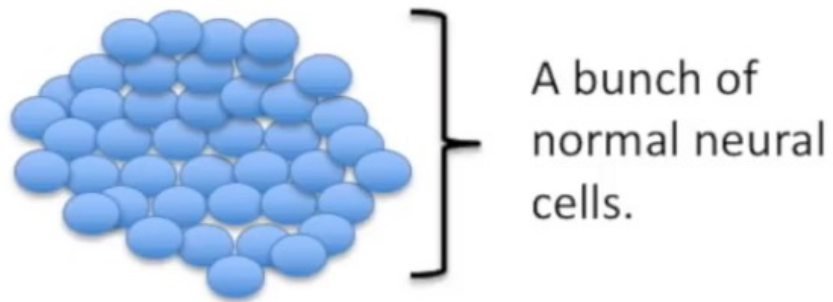
 = a normal neural cell

 = a mutated neural cell



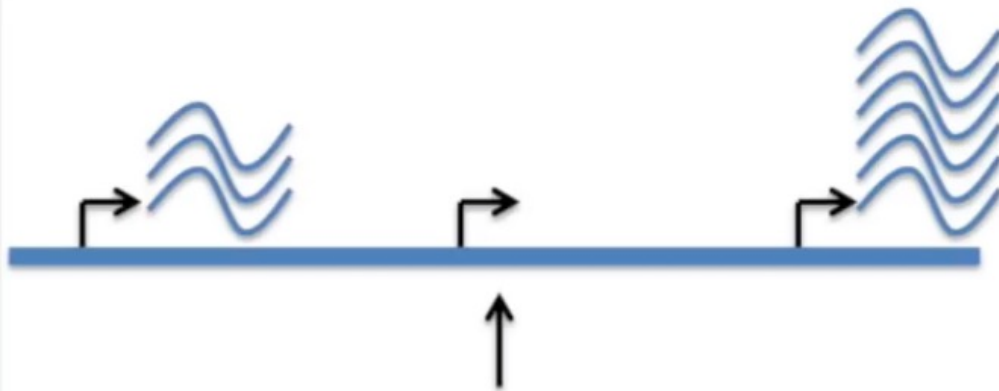
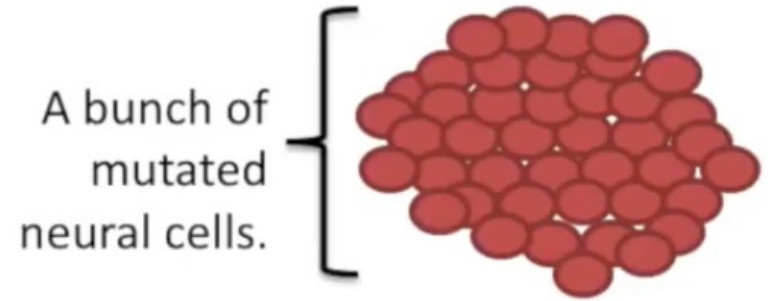
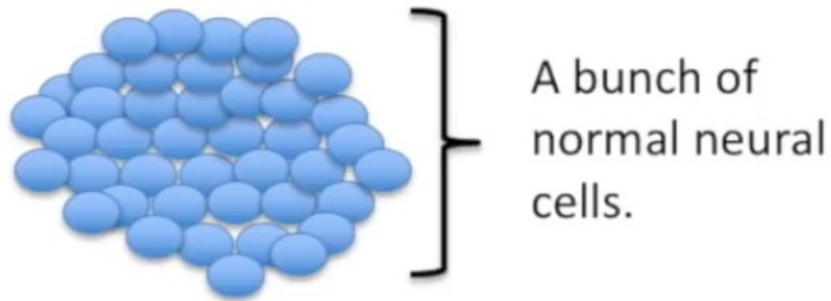
● = a normal neural cell

● = a mutated neural cell

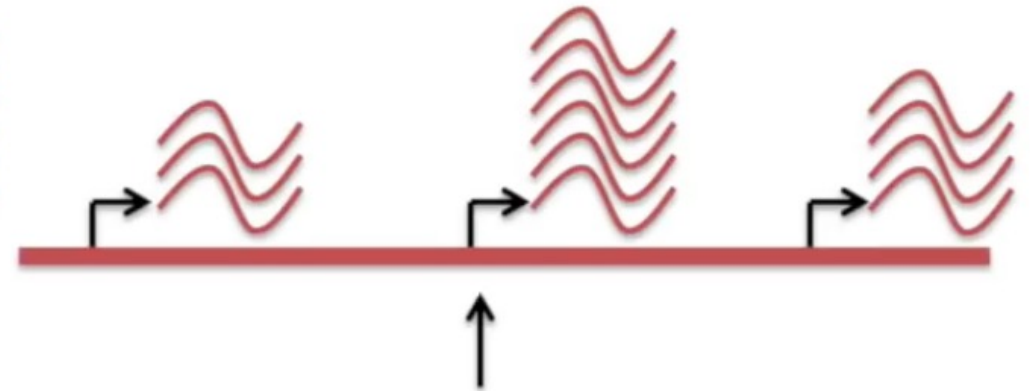


● = a normal neural cell

● = a mutated neural cell



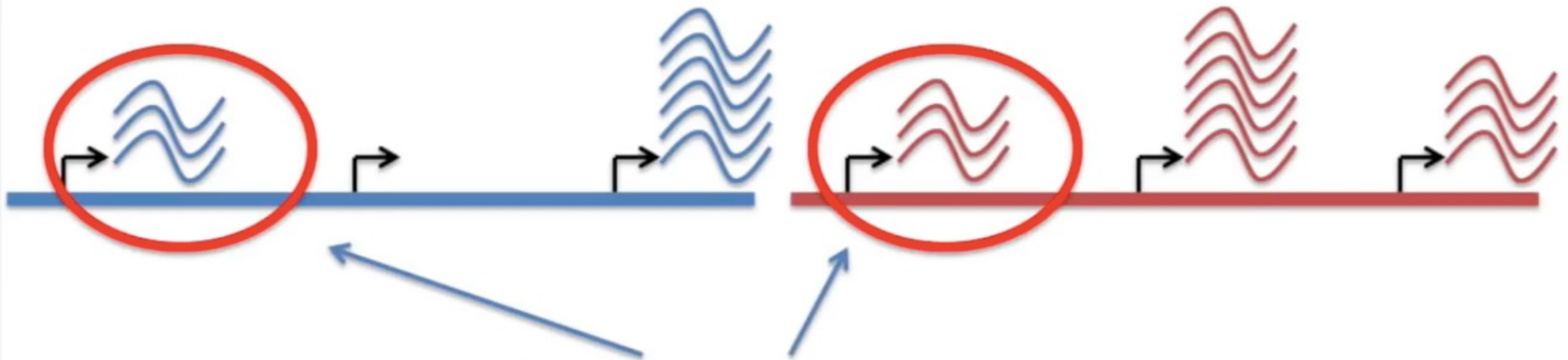
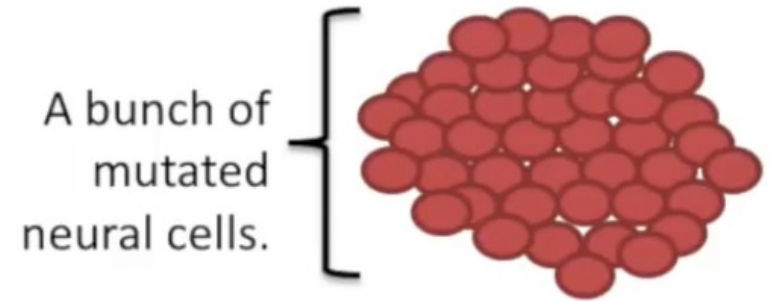
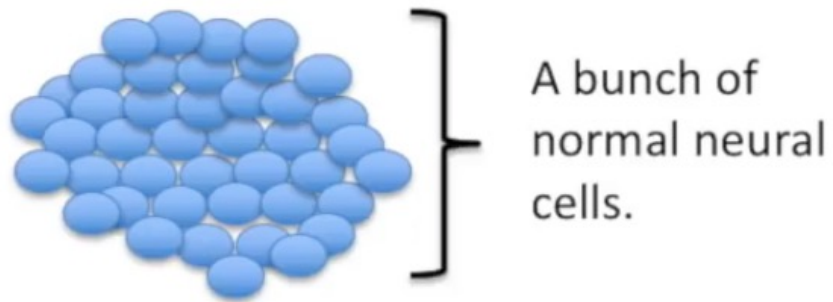
We can use RNA-seq to measure gene expression in normal cells...



... then use it to measure gene expression in mutated cells...

● = a normal neural cell

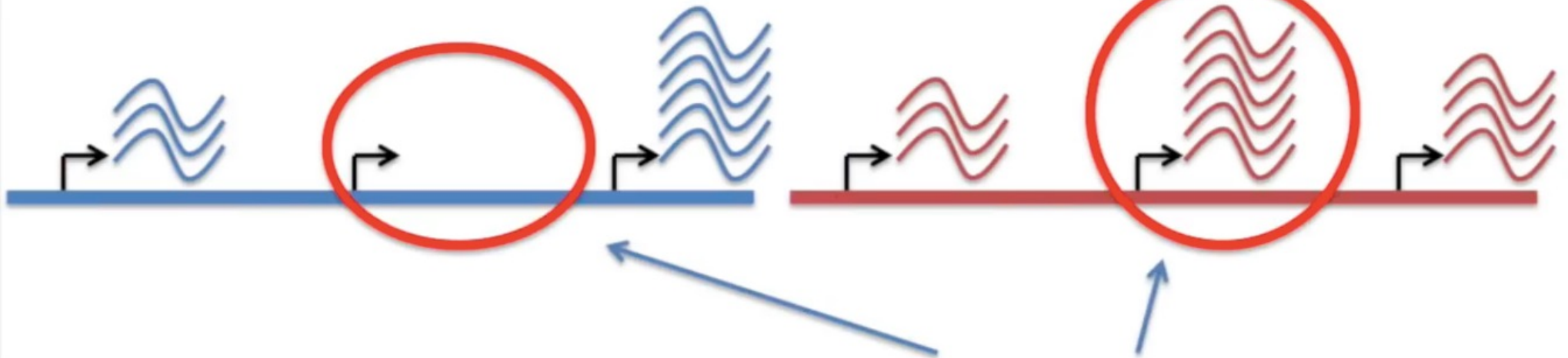
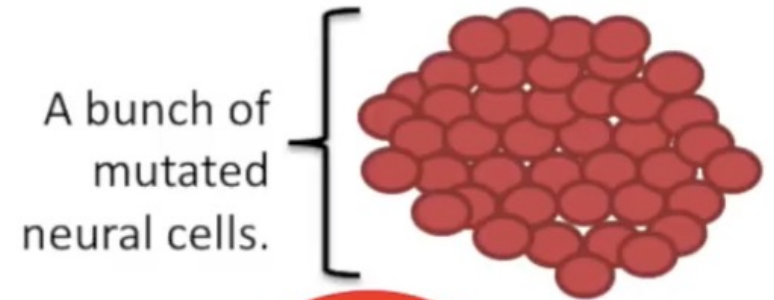
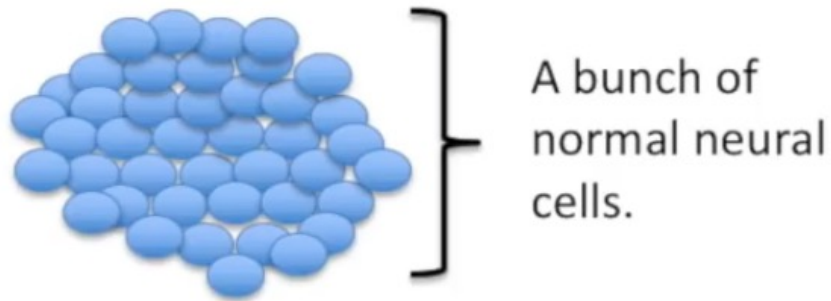
● = a mutated neural cell



Gene1: No difference between normal and mutated cells.

● = a normal neural cell

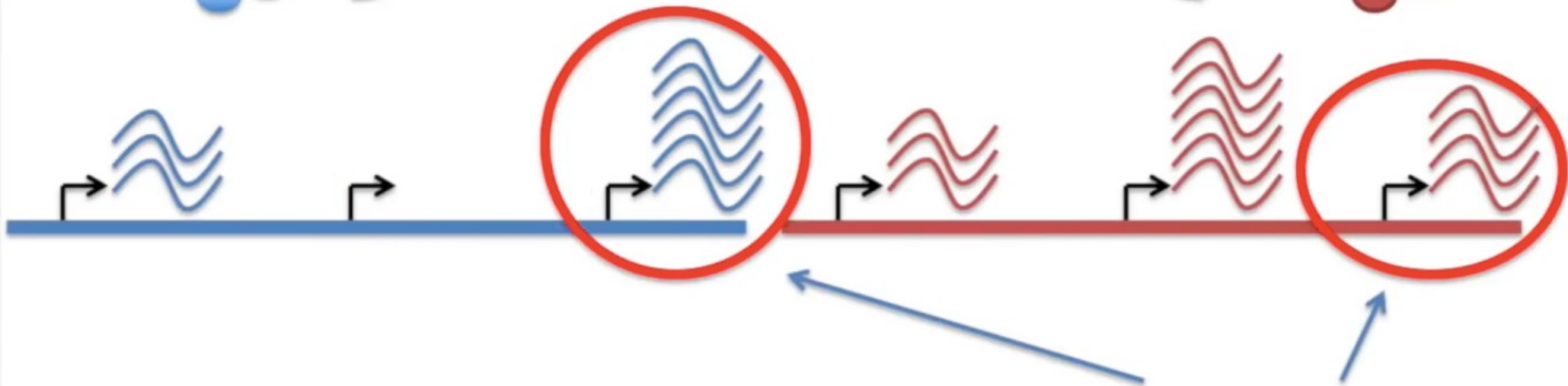
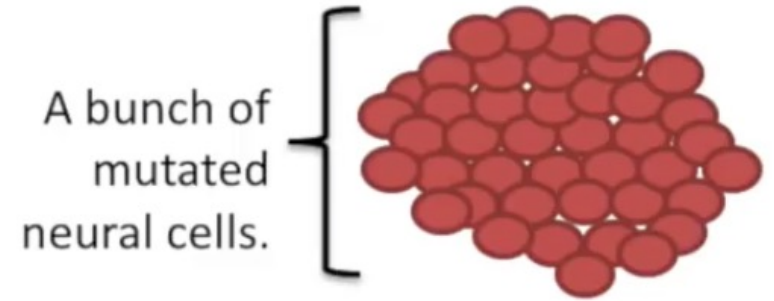
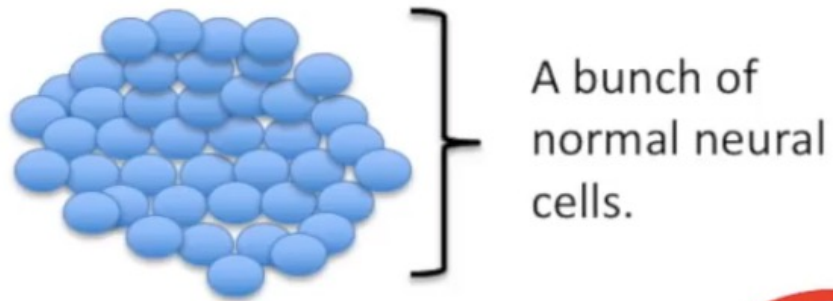
● = a mutated neural cell



Gene2: A big difference between normal and mutated cells.

● = a normal neural cell

● = a mutated neural cell



Gene3: A subtle difference between normal and mutated cells.

3 Main Steps for RNA-Seq:

- 1) Prepare a sequencing library
- 2) Sequence
- 3) Data analysis

Preparing an RNA-seq library

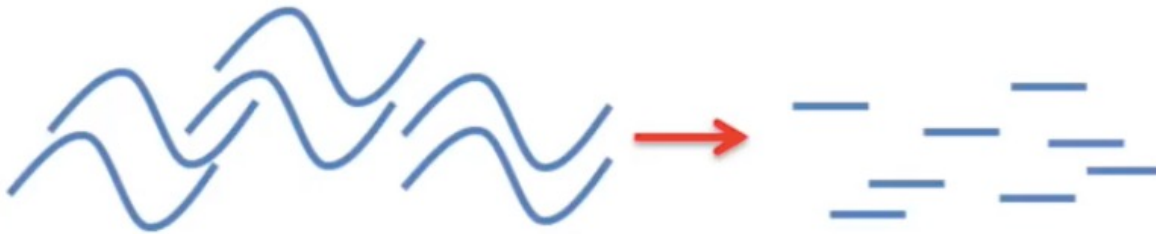
Step 1: Isolate the RNA



Preparing an RNA-seq library

Step 1: Isolate the RNA

Step 2: Break the RNA into small fragments.



We do this because RNA transcripts can be thousands of bases long, but the sequencing machine can only sequence short (200-300 bp) fragments

Preparing an RNA-seq library

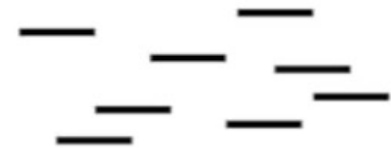
Step 1: Isolate the RNA



Step 2: Break the RNA into small fragments.



Step 3: Convert the RNA fragments into double stranded DNA.



Double stranded DNA is more stable than RNA and can be easily amplified and modified. This leads us to the next step...

Preparing an RNA-seq library

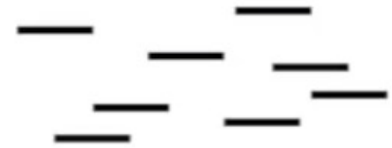
Step 1: Isolate the RNA



Step 2: Break the RNA into small fragments.



Step 3: Convert the RNA fragments into double stranded DNA.



Step 4: Add sequencing adaptors.



Preparing an RNA-seq library

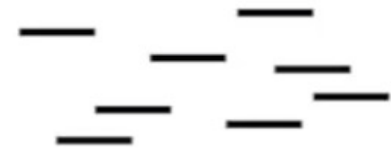
Step 1: Isolate the RNA



Step 2: Break the RNA into small fragments.



Step 3: Convert the RNA fragments into double stranded DNA.



Step 4: Add sequencing adaptors.



The adaptors do two things:

- 1) Allow the sequencing machine to recognize the fragments.
- 2) Allow you to sequence different samples at the same time, since different samples can use different adaptors.

Preparing an RNA-seq library

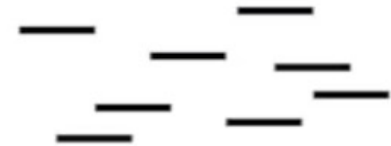
Step 1: Isolate the RNA



Step 2: Break the RNA into small fragments.



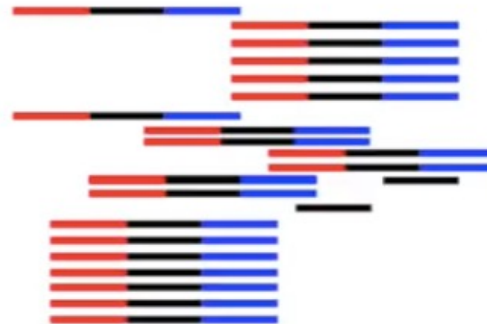
Step 3: Convert the RNA fragments into double stranded DNA.



Step 4: Add sequencing adaptors.



Step 5: PCR amplify.



Preparing an RNA-seq library

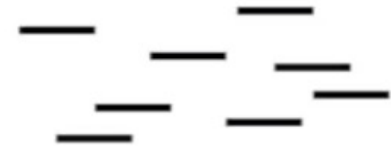
Step 1: Isolate the RNA



Step 2: Break the RNA into small fragments.



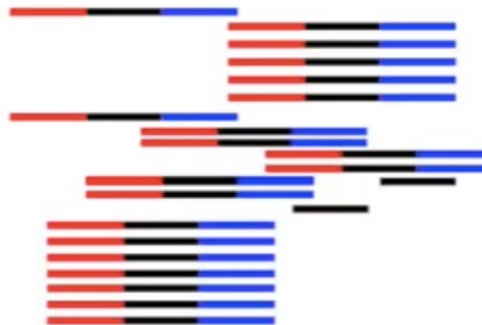
Step 3: Convert the RNA fragments into double stranded DNA.



Step 4: Add sequencing adaptors.



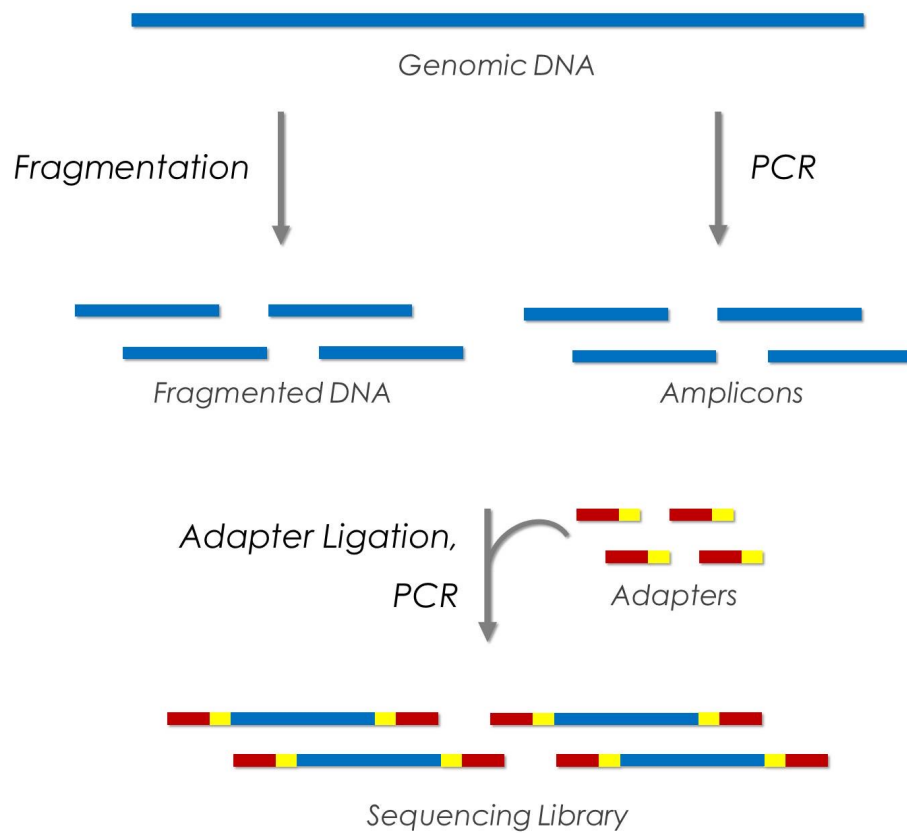
Step 5: PCR amplify.



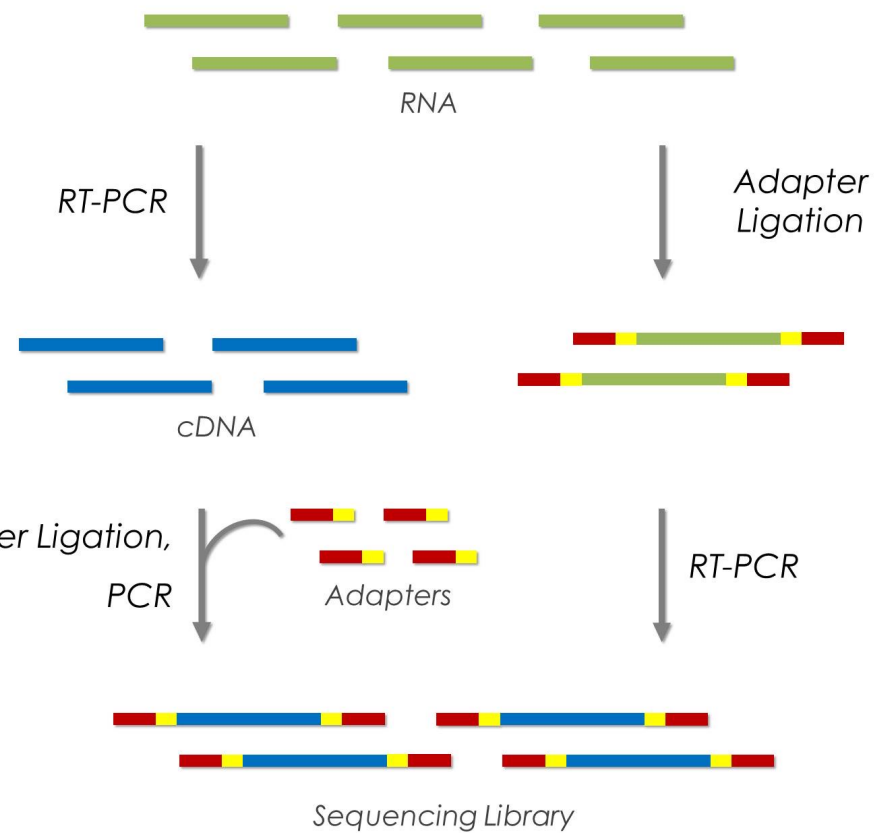
Step 6: QC

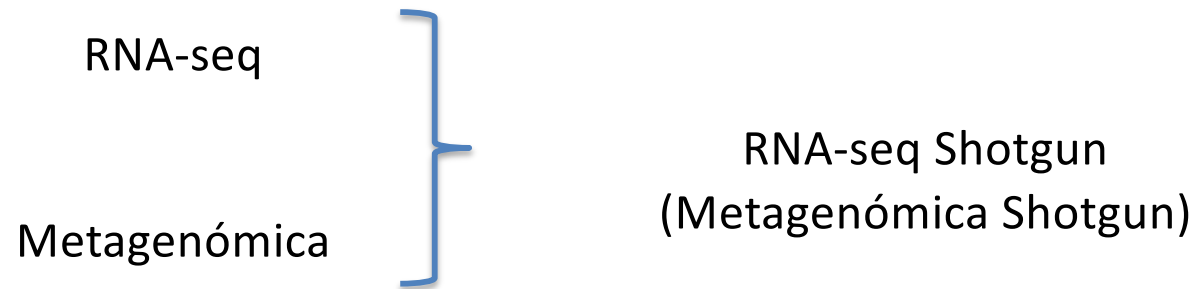
- 1) Verify library concentration
- 2) Verify library fragment lengths

DNA Library Construction



RNA Library Construction





- **Análisis metagenómico mejorado:** Se han desarrollado nuevos métodos, como PT-seq, para analizar la epigenética de los fosforotioatos en las comunidades microbianas, lo que proporciona una visión más profunda de las funciones e interacciones microbianas.
- **Procesamiento de datos mejorado:** Herramientas como Lemur y Magnet se han optimizado para elaborar perfiles taxonómicos ligeros y precisos a partir de conjuntos de datos metagenómicos de “Shotgun” de lectura larga, reduciendo los falsos positivos y mejorando la precisión de los datos.
- **Perfiles completos del microbioma:** Los estudios han utilizado la metagenómica Shotgun para elaborar perfiles de comunidades microbianas en diversos entornos, como lagos periurbanos y el intestino humano, revelando importantes conocimientos ecológicos y funcionales.
- **Integración con la multiómica:** Integración de metagenómica Shotgun con otros datos ómicos (por ejemplo, la metabolómica) para proporcionar una visión holística de los ecosistemas microbianos y sus rutas metabólicas.

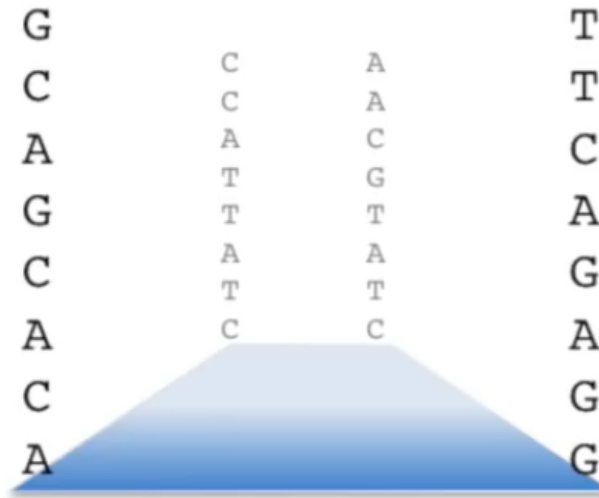
SECUENCIACIÓN DE NUEVA GENERACIÓN (NGS)



Examples of NGS systems

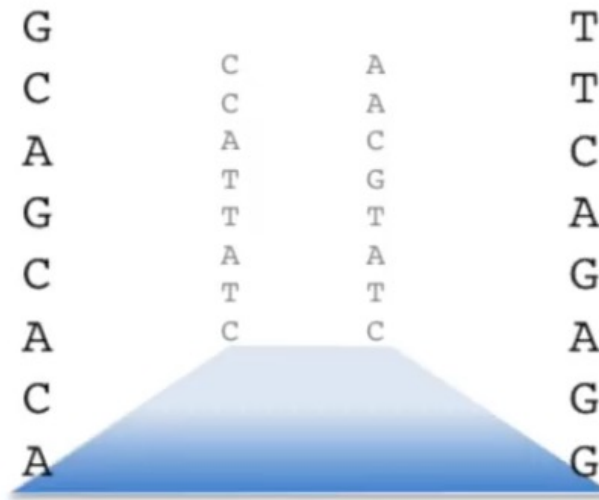
ILLUMINA

Actually, there are about 400,000,000 fragments laid out vertically in a grid.

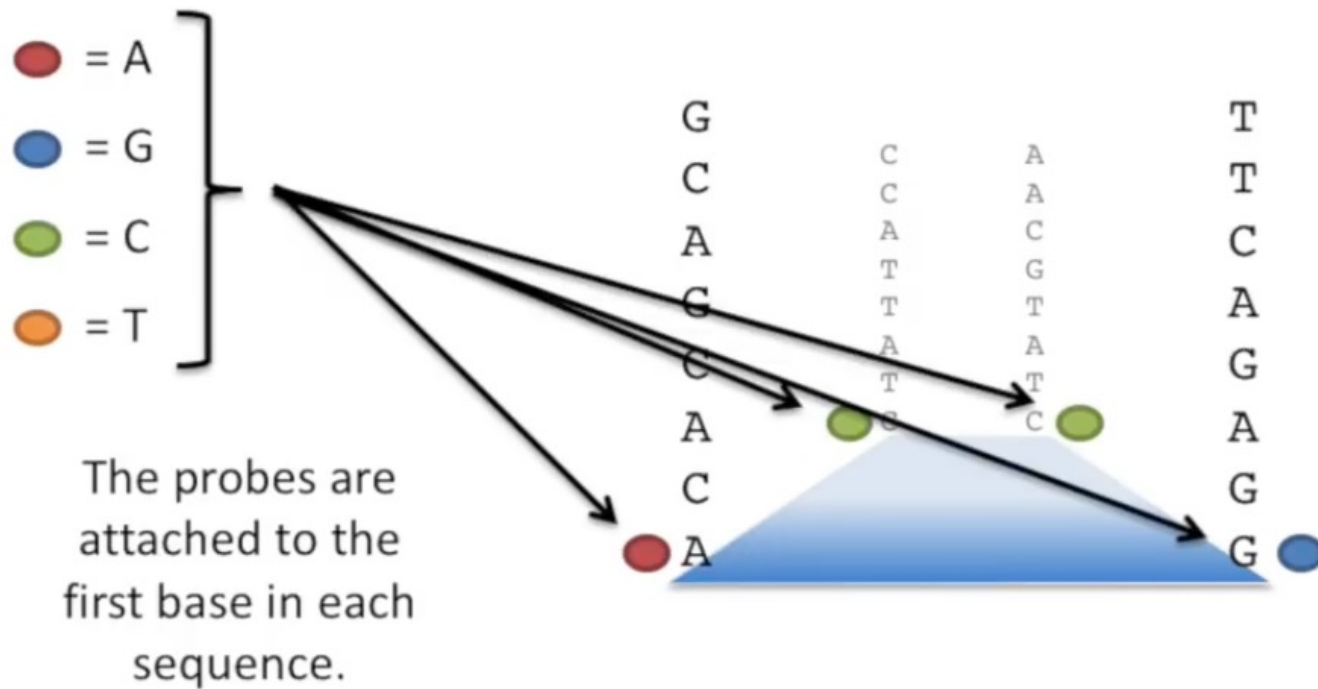


The machine has fluorescent probes that are color coded according to the type of nucleotide they can bind to.

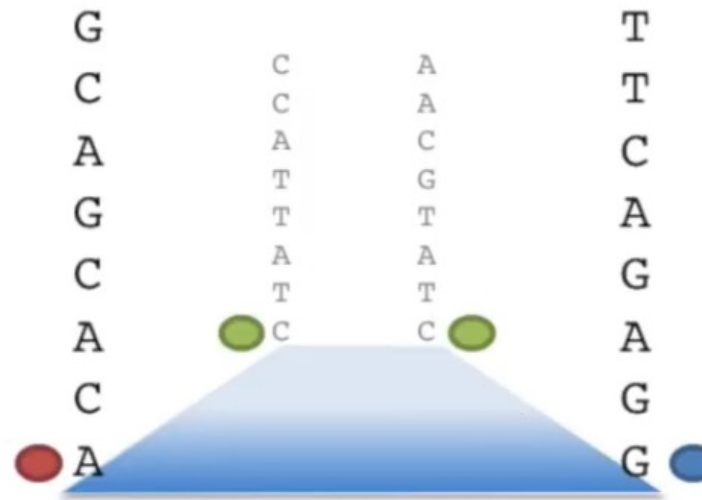
- = A
- = G
- = C
- = T



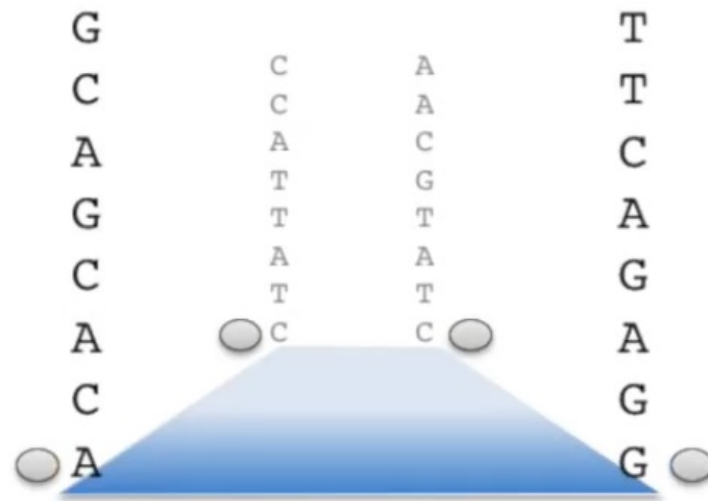
The machine has fluorescent probes that are color coded according to the type of nucleotide they can bind to.



Once the probes have attached, the machine takes a picture of the flow cell from above that looks like this...

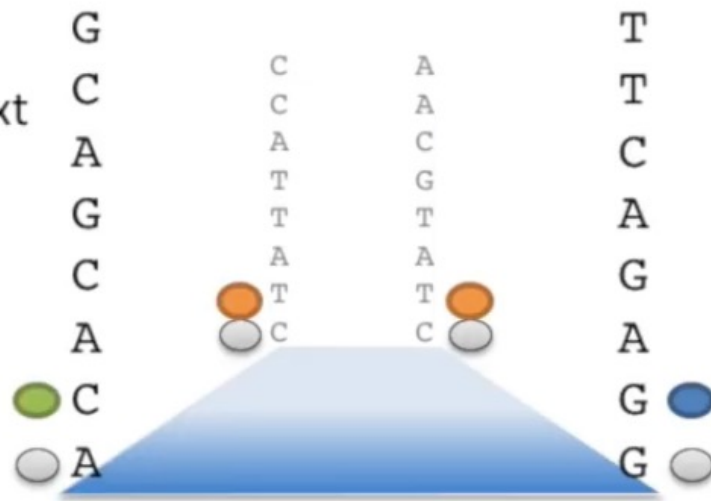


Then the machine washes the color off of the probes....

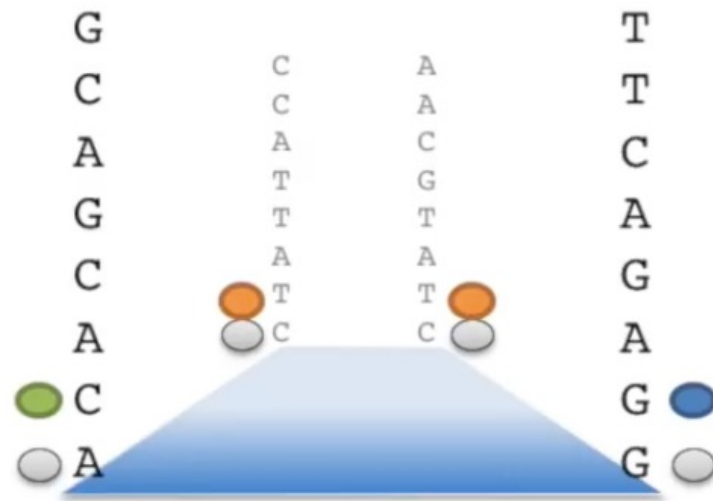
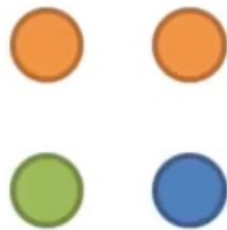


● = A
● = G
● = C
● = T

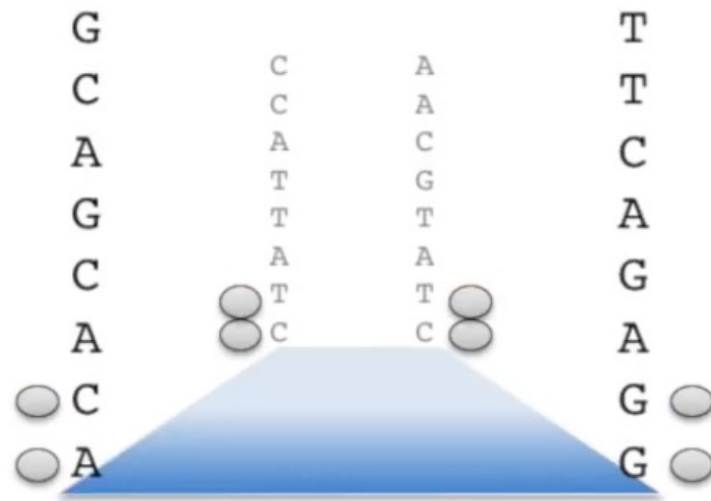
Then probes are
 bound to the next
 base in each
 fragment.



The machine takes a picture
from above...

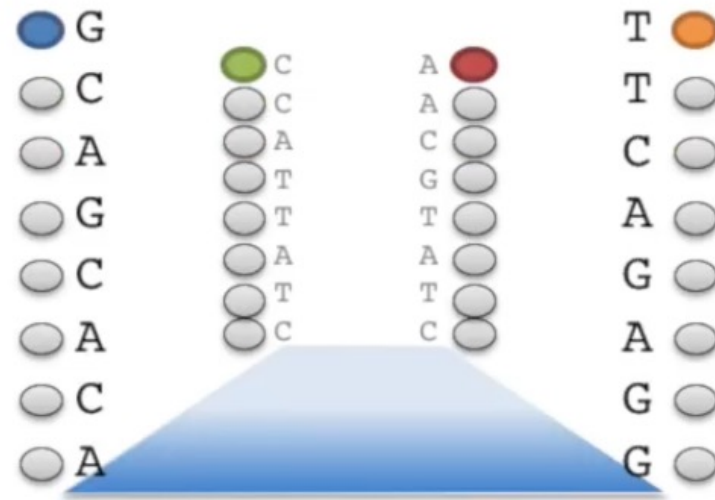


Then the machine washes the color off of the probes....



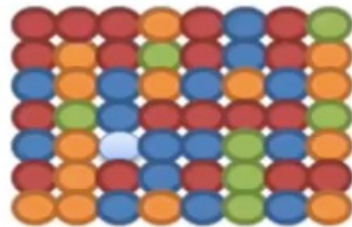
● = A
 ● = G
 ● = C
 ● = T

And the process repeats until the machine has determined each sequence of nucleotides.



This is how it works with 4 DNA fragments.

This matrix still isn't 400,000,000 DNA fragments, but it illustrates one type of problem that can occur.



The raw data...

```
@NS500177:196:HFTTAFXX:1:11101:10916:1458 2:N:0:CGCGGCTG
ACACGACGATGAGGTGACAGTCACGGAGGATAAGATCAATGCCCTCATTAAGCAGCCGGTGTAA
+
AAAAAEEEEEEEEEEEEEE//AEEEEEEEEEEEEEEEEEE/EE/<<EE/AEEEAEE//EEEEAEAEAEA<
```

Each sequencing “read” consists of 4
lines of data.

A typical sequence run with 400,000,000 reads
will generate a file containing 1.6 billion lines
of data!!!

Count reads per gene



Once we know the chromosome and position for a read, we can see if it falls within the coordinates of a gene (or some other interesting feature.)

Xkr4 – Chromosome 1, position: 3204563-3661579

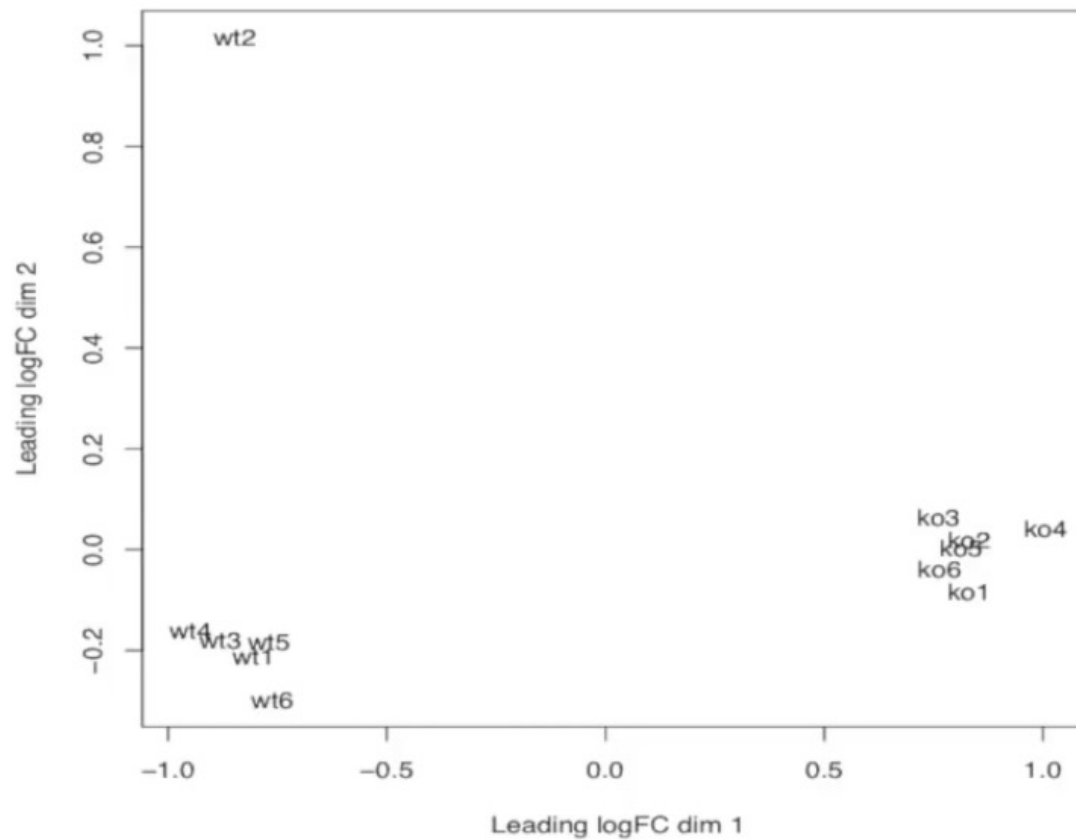
Rp1 – Chromosome 1, position: 4280927-4399322

etc.. (for all 20,000 genes in the genome)

Gene	Sample1	Sample2	Sample3...
A1BG	30	5	13...
A1BG-AS1	24	10	18...
A1CF	0	0	0...
A2M	5	9	7...
A2M-AS1	3563	5771	4123...
A2ML1	13	8	7...
...

If this were a Single Cell RNA-seq experiment, we would have
 20,000 rows (genes) by 800+ columns (samples), giving us at
 least 16 million values to keep track of

This is a PCA plot from a real RNA-seq experiment done on neural cells.
The “wt” samples are “normal”.
The “ko” samples are samples that were mutated by the researchers.

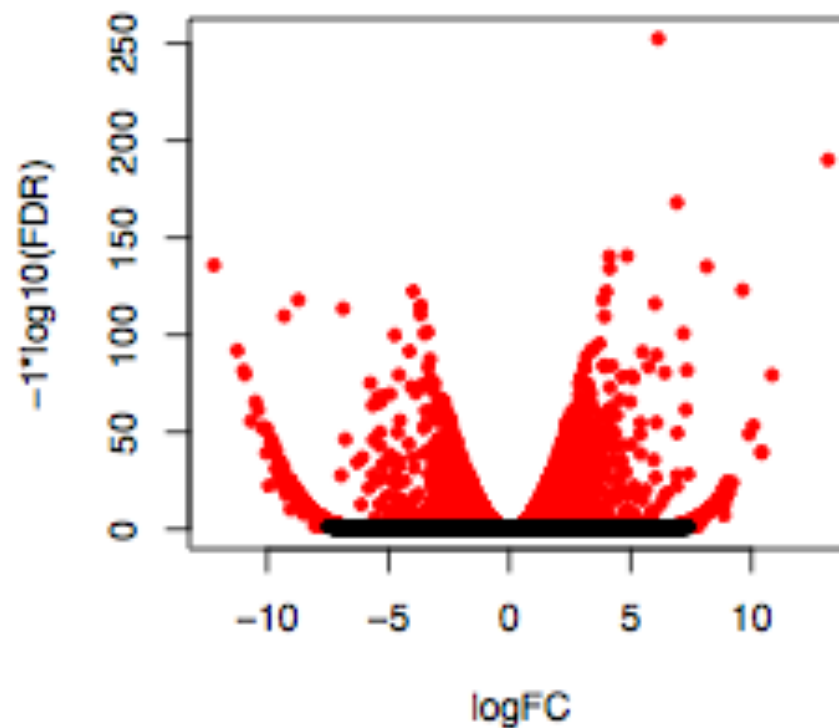


Step 2) Identify differentially expressed genes between the “normal” and “mutant” samples.

This is typically done using R with either **edgeR** or **DESeq2**, and the results are generally displayed using this sort of graph.



Volcano plot



Muestra 1

CCCACATCTTCTCCATCTCCGACAACGCCTATCAGTACATGCTGACAGGTGAGAGGCCCTGGAA

[illegible]

Muestra 2

ACGGATCAATGTAATGAACCGTGGGGATGACACCCCTCTGCATCTGGCAGCCAGTCATGGA
ACGGATCAATGTAATGAACCGTGGGGATGACACCC
ACGGATCAATGTAATGAACCGTGGGGATGACACCC
ACGGATCAATGTAATGAACCGTGGGGATGACACCC
ACGGATCAATGTAATGAACCGTGGGGATGACAC

Muestra 1

CCCACATCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGCCCTGGAA

ATCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAG

ATCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAG

ATCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAG

ATCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAG

ATCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

TCTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGG

CTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGC

CTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGC

CTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGC

CTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGC

CTTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGC

TTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGCC

TTCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGCC

TCTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGCC

CTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGCCCT

CTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGCCCT

CTCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGCCCT

TCCATCTCCGACAACGCCATCAGTACATGCTGACAGGTGAGAGGCCCTG

Muestra 2

ACGGATCAATGTAATGAACCGTGGGGATGACACCCCTCTGCATCTGGCAGCCAGTCATGGA

ACGGATCAATGTAATGAACCGTGGGGATGACACCC

ACGGATCAATGTAATGAACCGTGGGGATGACACCC

ACGGATCAATGTAATGAACCGTGGGGATGACACCC

ACGGATCAATGTAATGAACCGTGGGGATGACAC

