

# Genomics I

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Biomedical Data Science: Mining and Modeling  
CB&B 752 • MB&B 452  
Matt Simon  
January 15, 2020

# What is genomics?

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1. The **global** study of how biological **information** is encoded in genome sequence

Genes

Regulatory sequences

Genetic variation

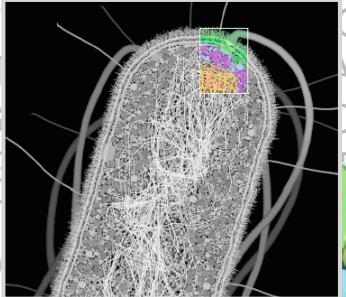
2. How this information is **read out** to produce distinct **biological outcomes**

Gene expression and regulation

Cellular identity, differentiation and development

Phenotypic variation among individuals and species

In practice, many experiments that involve **deep sequencing** are considered genomics.



# Overview

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- Genomics I (today's lecture): Focus on sequencing technology and genomes.
- Genomics II: (Friday's lecture): Focus on applications of sequencing technology.

Credit: Jim Noonan for many of the slides

# Overview

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- Sequencing data: from wet lab to fastq.
  - Applications to studying genomes and much much more.
- \*Sophisticated use of data from genomics requires an integrated understanding of the biological experiment, sample preparation and down stream computational analyses of the data.

Credit: Jim Noonan for many of the slides

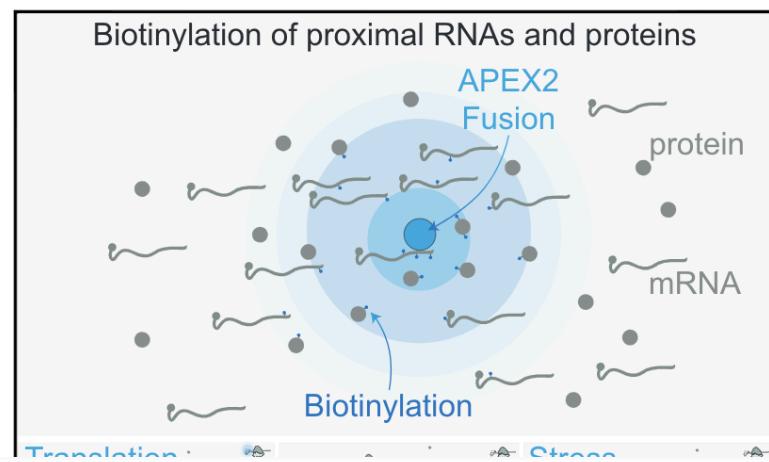
# Importance of genomics data: these data are central to most biomedical and biological sciences

Resource

## Molecular Cell

### Proximity RNA Labeling by APEX-Seq Reveals the Organization of Translation Initiation Complexes and Repressive RNA Granules

#### Graphical Abstract



#### Authors

Alejandro Padrón, Shintaro Iwasaki,  
Nicholas T. Ingolia

#### Correspondence

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#### In Brief

In this issue of *Molecular Cell*, Padrón et al. develop an RNA proximity labeling technique that maps subcellular RNA organization comprehensively. A powerful aspect of APEX-seq is the ability

#### DATA AND CODE AVAILABILITY

The raw sequencing data generated for this study are available at NCBI GEO GSE121575. Scripts to run the analyses mentioned above are available upon request.

Data can be found in genomics databases

- Most journals require authors to submit their data to a database (e.g., GEO) prior to publication.
  - These databases entries contain raw data and processed data.
  - These data can be used to examine the authors' claims, but also to test new hypotheses.

# Central questions

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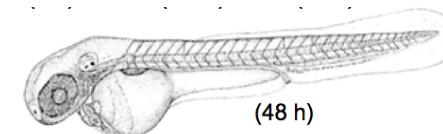
- Where do these data come from?
- How does the way we collect it influence what we know?

# Workflow

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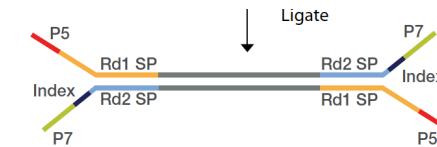
## 1. Isolation of sample.

e.g., Isolate DNA and shear.



## 2. Library preparation

e.g., Add known sequences to the ends.



## 3. Sequencing

e.g., Illumina Novaseq

## 4. Analysis

e.g., Map to genome and interpret.



# Metrics for evaluating sequencing technology

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- **Throughput:**

- Number of high quality bases per unit time
- Number of independent samples run in parallel
- Difficulty of sample preparation

- **Yield**

- Number of useful reads per sample
- Read length

- **Cost**

- Per run cost
- Per base cost
- Equipment
- Reagents
- Labor
- Analysis

# What is sequencing?

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## 1. First generation sequencing

- a. Maxam-Gilbert Sequencing
- b. Sanger Sequencing

## 2. Second generation sequencing

- a. Illumina Sequencing
- b. Ion Torrent

## 3. Third generation sequencing

- a. Nanopore based
- b. Pacific Bioscience Sequencing

The technology will change, but your need to critically understand the input and output will not.

# The steps of sequencing experiments

## 1. Sample preparation

- a. Isolation
- b. Library construction

## 2. Sequencing

- a. Flow cell loading
- b. Cluster generation
- c. Sequencing
- d. Processing image files
- e. De-multiplexing samples

## 3. Data analysis

- a. Read filtering
- b. Alignment to a genome
- c. Diverse analyses

The screenshot shows the Yale Center for Genome Analysis (YCGA) website. The header includes links for Yale School of Medicine, W.M. Keck Foundation, INFORMATION FOR, and a search bar. The main navigation menu has categories: Next-Gen Sequencing (selected), Bioinformatics, Microarrays, Services & Fees, Mendelian Center, and About YCGA. Below the menu, there are social sharing icons (share, tweet, email, print) and font size controls (T-, T, T+). The main content area is titled "Throughput" and contains a table of "Sequencer Lane Data Outputs".

Sequencer	Read length	# of Clusters per lane (millions)	Bases per lane (Gbp)
HiSeq 2500 Rapid	1x75	150	11.25
HiSeq 2500 Rapid	2x75	150	22.5
HiSeq 2500 Rapid	2x150	150	45
HiSeq 2500 High-output	1x75	200	15
HiSeq 2500 High-output	2x75	200	30
HiSeq 4000	2x100	300	60
HiSeq 4000	2x150	300	90
NovaSeq S2	2x100	1650	330
NovaSeq S2	2x150	1650	500
NovaSeq S4	2x150	2000	600

<http://ycga.yale.edu/sequencing/illumina/>

# What is the output from an Illumina sequencing experiment?

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One read (fastq format)

```
@HWI-ST1239:178:H0KPNADXX:2:1101:3120:1979 1:N:0:TGACCA
NCTGTAGGCTGCGTAGCCTCCCTGCAGGGTAAGTGGGAGGAGAGAGCAGAGGGACTTAGTGGGGCTCCCCAGGG
+
#1=DDFFFHHHHHIJIJJJJIJJJJJJJJ?FHIDGIJ=GIHGIIIHGIJHEHIHHGFFFFEEEDDDDDDDDDDDDD
```

1. Read identifier
2. **Sequence**
3. Quality score identifier “+”
4. Quality score

# What is the output from an Illumina sequencing experiment?

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Many reads...

```
@HWI-D00306:498:HBB89ADXX:1:1101:1180:1882 1:N:0:CGATGT
NCATCACTTCTGCACCAGCCATGACGTCAATCTCGTCCGAACCCAAACTCGAGATCGGAAGAGCACACGTCTG
+
#11BBDDDFDFBFFFIIIIIIIIIIIFEGIIIIFIGAGIIFIII=FEEEEFFFDDD=@9A@BBBBB=?BB<
@HWI-D00306:498:HBB89ADXX:1:1101:1167:1902 1:N:0:CGATGT
TATTGCAATATGTTAACAACTAACAGGAAAAAATACCCCACACAAAACACAAACCCCTAGAACTGTGCTG
+
B@@FFDFFHFHHHJJIJIGIIJJJJIJHFIJJJJJJJEHHJJIJJJJJGHHHFBDFFE>CEEC
@HWI-D00306:498:HBB89ADXX:1:1101:1190:1928 1:N:0:CGATGT
ACCAAGCCACAATAAGTTAGTGTTCATAGTACATGCTGAGTTATTGATCCGTATCTACACTGCTACTGTC
+
@<@DDDD8CDDGE?2<AFFBCCEEHEIEGHIEGEIDD@CDGFFFEIDGCFCDABFG>FBFGFGIEIFFFDDD
@HWI-D00306:498:HBB89ADXX:1:1101:1157:1931 1:N:0:CGATGT
CTGAGATTCTTGCCATAGCCTAACCACTACGCAACTGCAACCAACCACCTCCGTGGTTGCCCTCTCGATCG
+
CCCCFFFFFHIIJJIIJJJIIGHHIJGGJIGIJJJJJJIJJJJIIJGJJHCHFBDFFFDDECB
```

Generally ~ 2,000,000,000 reads/sequencing lane

Note: This is for an Illumina NovaSeq with current chemistry, but this number changes

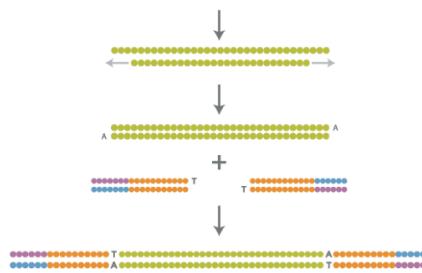
# How long are the reads?

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TATTGCAATATGTTAACAACTAACAGGAAAAAATACCCCACACAAAACACAACCCTTAGAACTGTGCTG  
← →  
**75 nt**

While there are other technologies that can give longer read lengths, Illumina reads are generally 50 nt - 250 nt

# Where do these reads come from?



Library Preparation  
~2 h [15 min hands-on (Nextera)]  
< 6 h [< 3 h hands-on (TruSeq)]



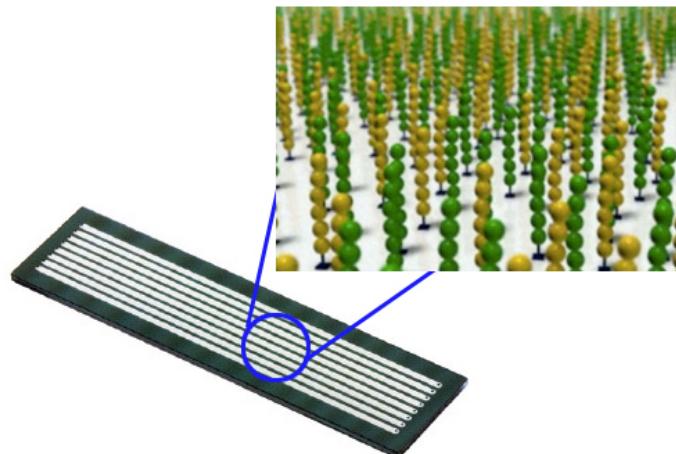
Cluster Generation  
~5 h (<10 min hands-on)



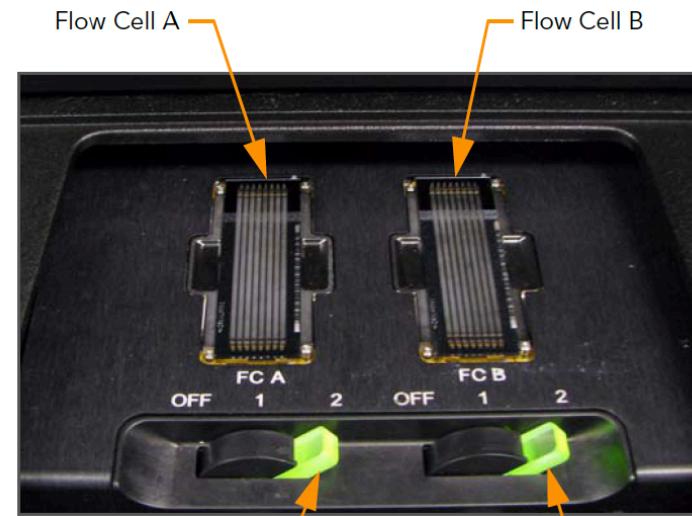
Sequencing by Synthesis  
~1.5 to 11 days



CASAVA  
2 days (30 min hands-on)



Flow cell



Flow Cell Lever A      Flow Cell Lever B

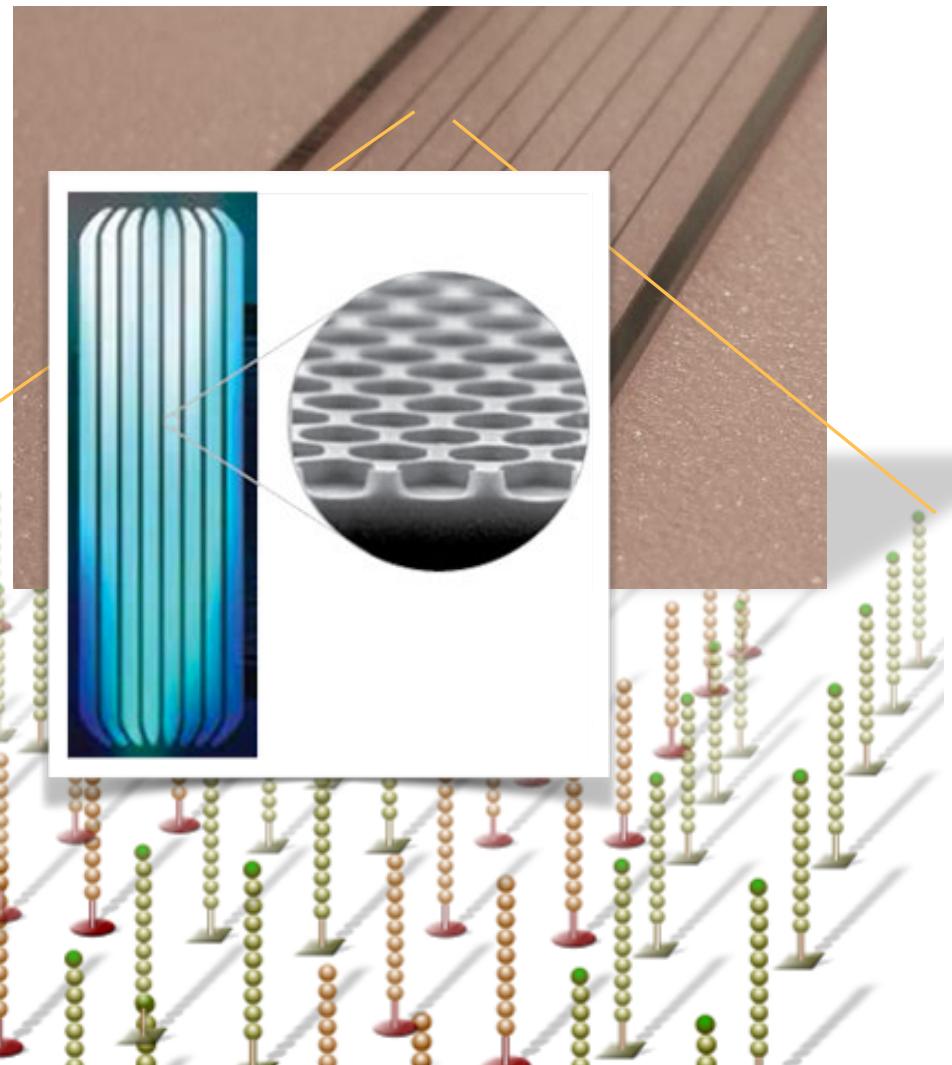
# What is a flow cell?

A flow cell is a thick glass slide with 8 channels or lanes.

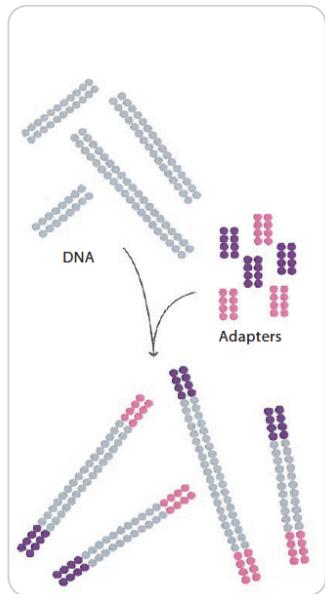
Each lane is randomly coated with a lawn of oligos that are complementary to library adapters

P5 oligo

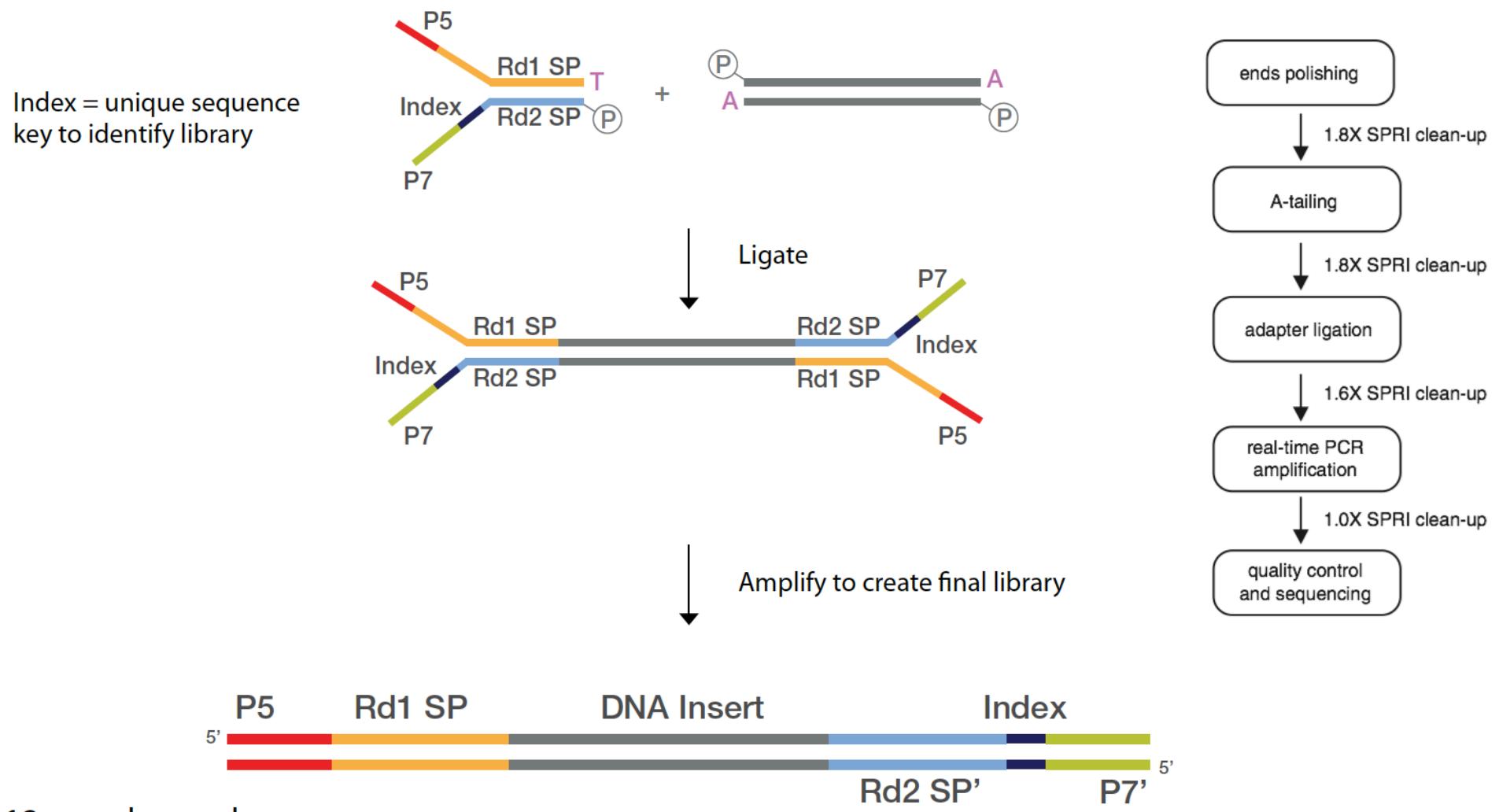
P7 oligo



Cluster PCR  
on flow cell  
(8 lanes)



*Optional:* How do you make a sequencing library?



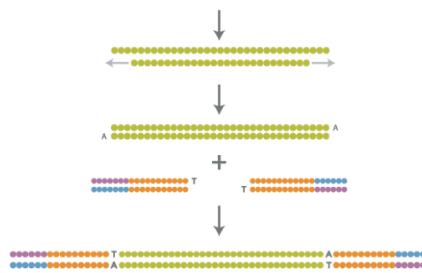
12 samples per lane

## Potential sources of bias:

1. Selective PCR amplification (issue of duplicates).
  2. Size selection.
  3. Enzyme specificities.

Challenging but possible to analyze pg quantities of DNA. (In humans, ~6 pg DNA/cell).

# Where do these reads come from?



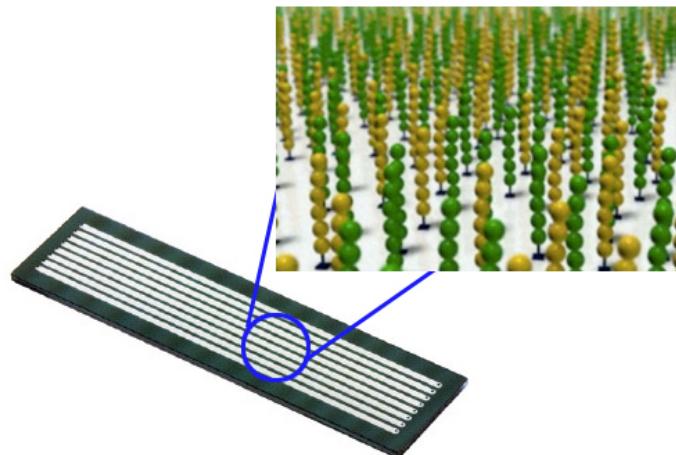
Library Preparation  
~2 h [15 min hands-on (Nextera)]  
< 6 h [< 3 h hands-on (TruSeq)]



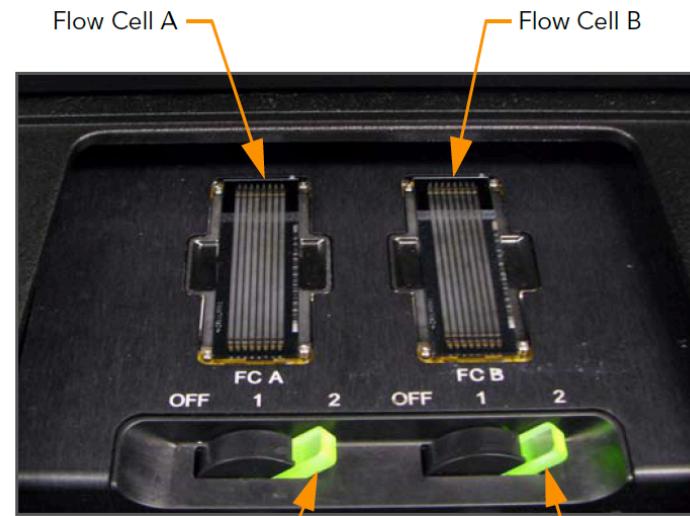
Cluster Generation  
~5 h (<10 min hands-on)

Sequencing by Synthesis  
~1.5 to 11 days

CASAVA  
2 days (30 min hands-on)



Flow cell



Flow Cell Lever A      Flow Cell Lever B

# What is the output from an Illumina sequencing experiment?

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Paired read (fastq format)

```
@HWI-ST1239:178:H0KPNADXX:2:1101:3120:1979 1:N:0:TGACCA
NCTGTAGGCTGCGTAGCCTCCCTGCAGGGTAAGTGGGAGGAGAGAGCAGAGGGACTTAGTGGGGCTCCCCAGGG
+
#1=DDFFFHHHHHIJIJJJIJJJJJJ?FHIDGIJ=GIHGIIIHGIFIHEHIHGFFFFEEEDDDDDDDDDDDDDDD
@HWI-ST1239:178:H0KPNADXX:2:1101:3120:1979 2:N:0:TGACCA
NNACCTAGCCATCTGCAGTCCTCGGTCTGTGTTAGACCAGAACTAGGTGCCAGGCCAGGTACCACTAACCTT
+
##4<@00000000?000?0@?????@0??@?????????????>?????????@>???000?0@?????
```

## 1. Read identifier

- a. Instrument
- b. Flow cell
- c. Read ID
- d. Coordinates
- e. Which read from a paired end sample
- f. Which index for multiplexed read

## 2. Sequence

- 3. Quality score identifier “+”
- 4. Quality score

# What limits the insert size and read length?

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## One read (fastq format)

```
@HWI-D00306:498:HBB89ADXX:1:1101:1180:1882 1:N:0:CGATGT
NCATCACTTCTGCACCAGCCATGACGTCAATCTCGTCCGAACCCAAACTCGAGATCGGAAGAGCACACGTCTG
+
#11BBDDDFDFBFFFIIIIIIIIIIIIIFEGIIIIFIGAGIIFIII=FEEEEFFFDDD=@9A@BBBBB=?BB<
```

- For each single end read: Incomplete incorporation of bases.
- For the size of the insert (especially for paired end analysis): Ability to get consistent clusters.

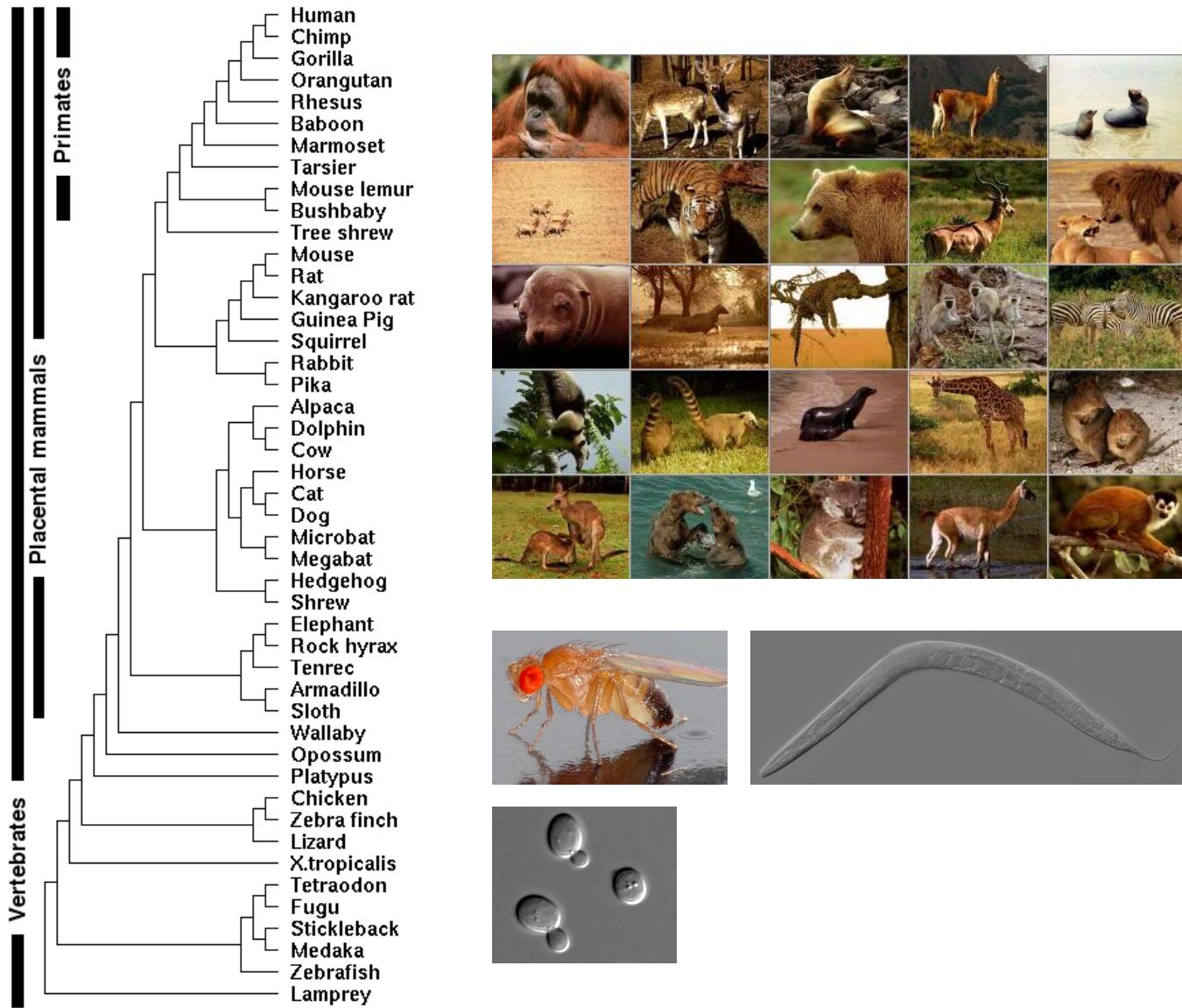
# What do I do with my sequencing reads?

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Source: Slate via Noonan

# Many reference genomes are available



# There is a wide range of genome sizes.

kb = 1000 bp

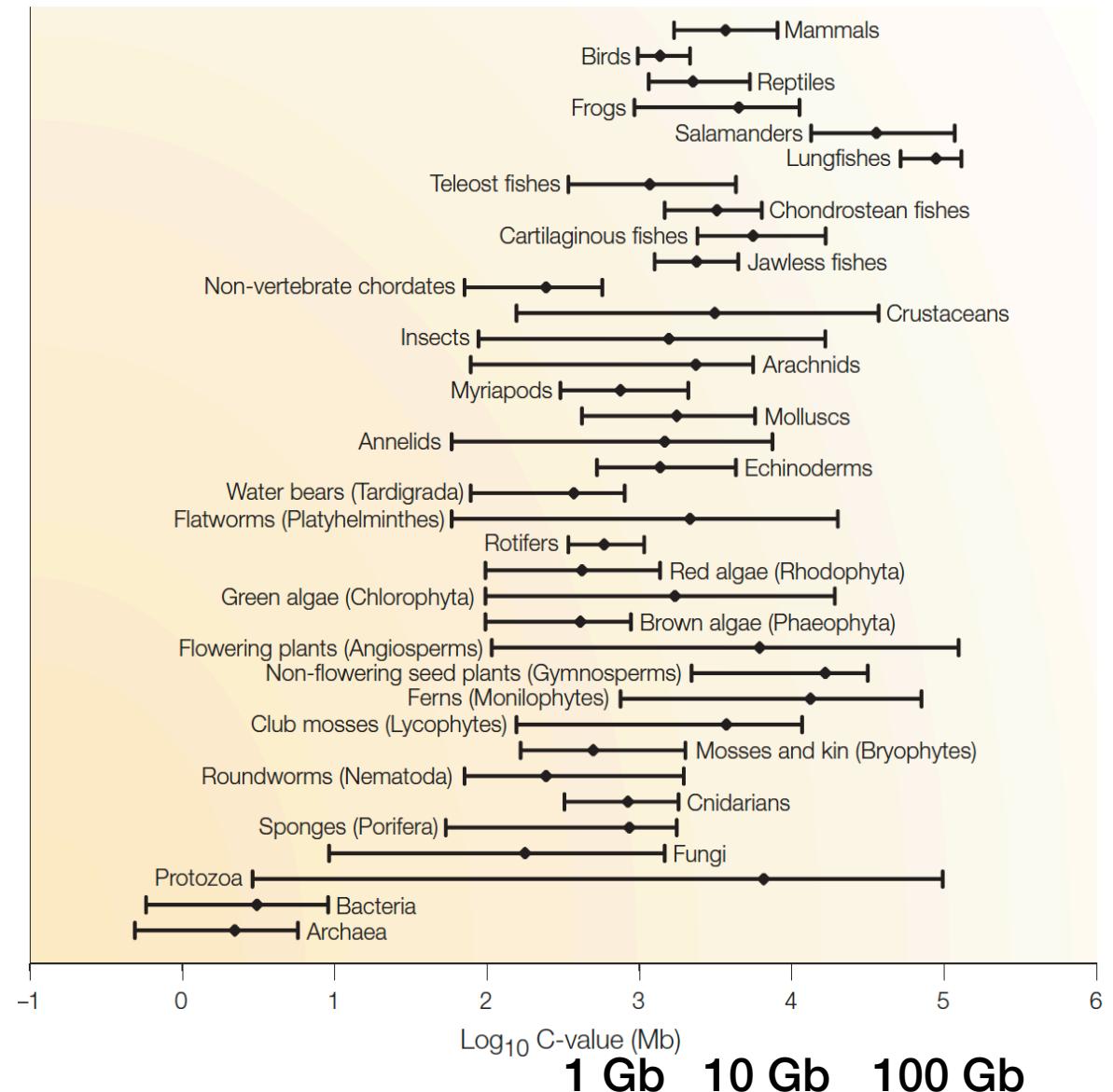
Mb =  $1 \times 10^6$  bp

Gb =  $1 \times 10^9$  bp

Tb =  $1 \times 10^{12}$  bp

## Human haploid genome ~ 3 Gb

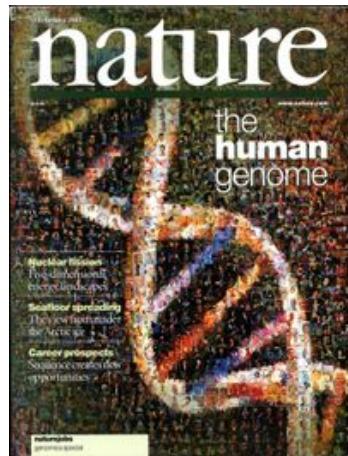
75 nt x  $3 \times 10^8$  reads/lane is about the right scale, but the amount of **coverage** necessary depends on application.



# Sequencing of the human genome

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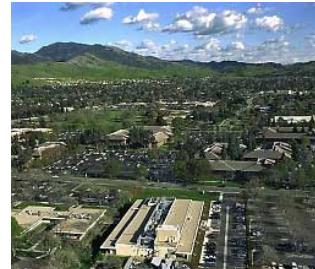
Victory declared **2003**



National Human  
Genome Research  
Institute

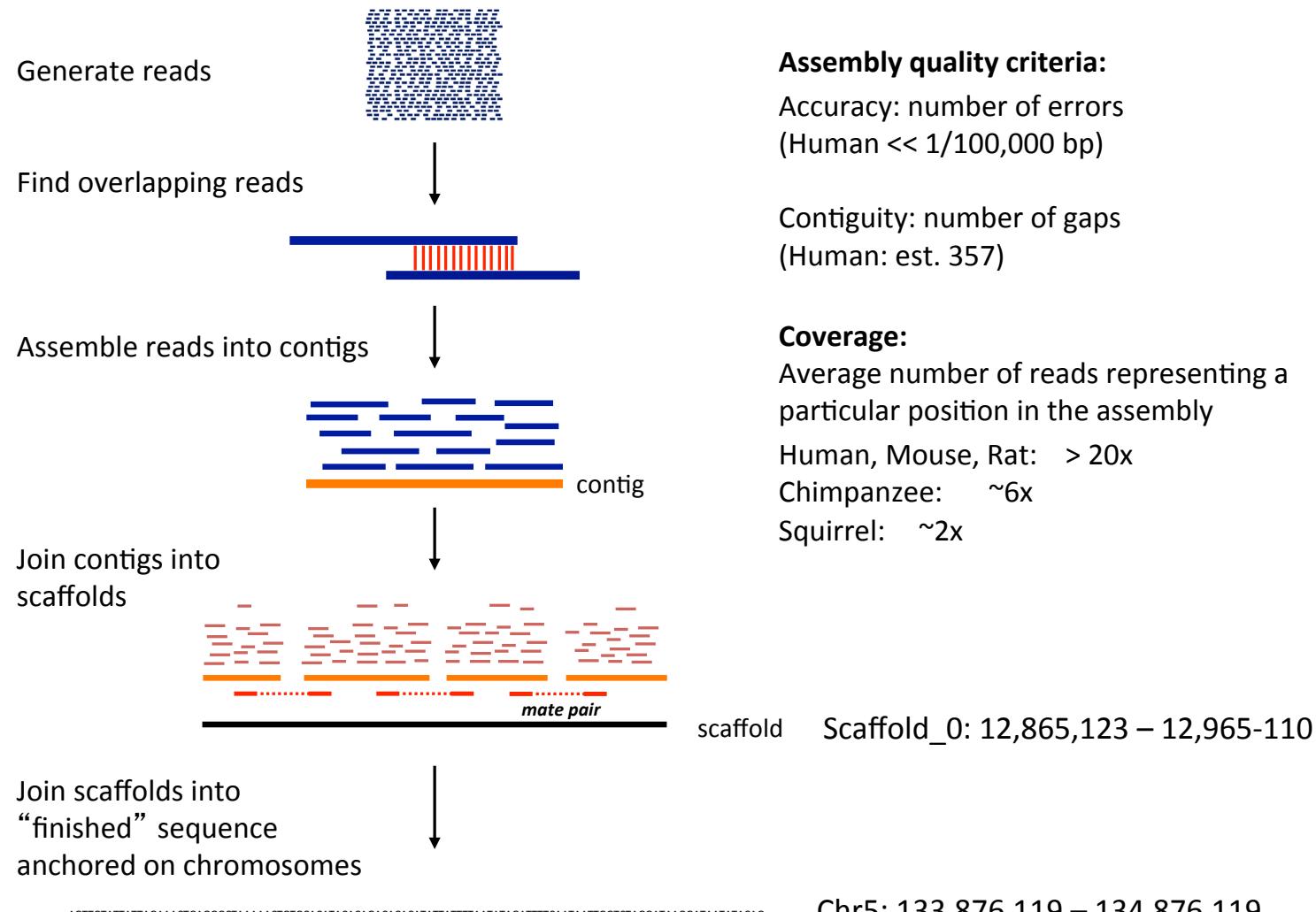


- Industrialization of Sanger sequencing, library construction, sample preparation, analysis, etc.
- \$3 billion total cost
- 1 Gb/month at largest centers (2005)



Novaseq 20 billion reads 2x150 bp. \$1000 -> \$100/genome.

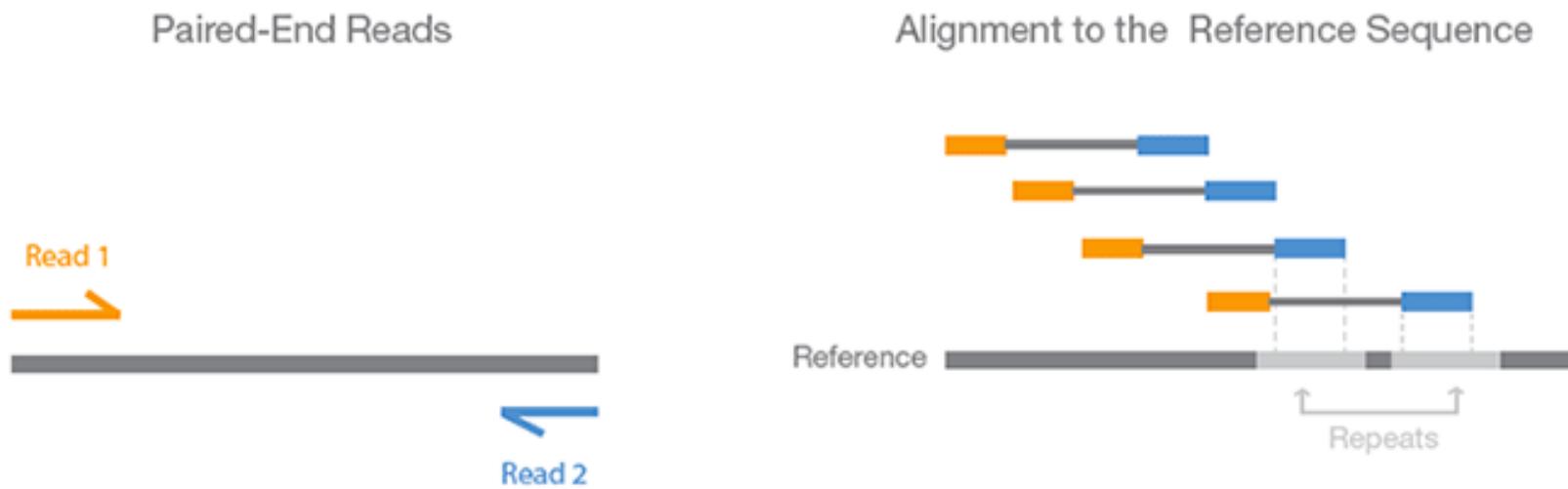
# How to assemble a genome



There are various

# The importance of paired end reads

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- Increase coverage of the insert.
- Particularly helpful when one read maps to multiple places in the genome.

CCAAATCAAACAGTTGATTATTAGAAACTGAGGGCTAAAACGTGCACATAACAGACACACATATTATTTAATATAGATTTCAATAATTGGTAGGATAAG  
AAGCAAGAAGAAAACAAGACTGTTACTATGGAAAAATGAAAATAGATTTAAAACATGTTAATTGACGTTACTTTTGTAAATTACTTTCTTCACCTCTT  
AATAAATCACATTAATTCTTATCTCATGTGAAATTCTATTTGATTGATACCTTAAATGTCATTGTTGAAGGAAGATTATTCAATTTCATTCAATAAAATTT  
CAGTATTATGTTCTAGGCATTGGGATACCAGTTCACAAGACAGACTATGATTACAGGATCAGATGTGGACTCTCAAATTGACTGAGAATAAAACAGACACT  
TAATTGATGCTAGAAAGACAATGAAACAGAGGCCATGTGACCAATGAGAGAGATGAGGGTGGCAGCAGCCTGTTAGATAAGGTACCTGATTGGTGGATTGG  
ATGCCCTAATGATATGAAAGAACCATGGGAGGCCTAGCATTAAAACCGCTAGGCAGAATGAGCAGCAAGTGCAAGGGCCTGGATAGGAATGAGC  
ATGGAAAAATGAAAATAGATTTAAAACATGTTAATTGACGTTACTTTGTTAAATTACTTTCTTCATTCAATAAAATTTTTAGAATAATAAGTCCCAGGCACAAGA  
CATGTTCACAGACAGACTATGATTACAGGATCAGATGTGGACTCTCAAATTGACTGAGAATAAAACAGACACTAAACAAGTAAATAAGTTAATTCAAGTT  
AGATTTAAAACATGTTAATTGACGTTACTTTGTTAAATTACTTTCTTCATTCAATAAAATTTTTAGAATAATAAGTCCCAGGCACAAGACCAGTATTGTTCTAG  
ATTGATACCTTAAATGTCATTGTTGAAGGAAGATTTCATTCAATAAAATTTTTAGAATAATAAGTCCCAGGCACAAGACCAGTATTGTTCAAGTTGTAATTGATGCTATCCC  
ACTATGATTACAGGATCAGATGTGGACTCTCAAATTGACTGAGAATAAAACAGACACAAACAAGTAAATAAGTTAATTCAAGTTGTAATTGATGCTATCCC  
TGGGGATACCATTACCTGCAATGTTATTAAATTTAGAACAATAACATTAAATTCCAAACATGCAAAGAGGAATCTCCATATCATGCTTGTCAATTGTT  
GTGTGAAAACATTCTCAGAATTAAACAATAACAAATCAGGGCTGAATGTGGCCAACATGCAAAGAGGAATCTCCATCTGTCATAACAGTTGATT  
CATACACAGACACACATATTAAATATAGATTTCAATAATTGGCTAGGATAAGGATAATACAGAGAACATGCCAAAGTTAAGCAAGAAGAAAACAAAG  
TTAAACATGTTAATTGACGTTACTTTGTTAAATTACTTTCTTCATTCTTACCTGTCATGTTATTAAATTTAGAACAATAACATTAAATTCTTATCTCATGTGAAATTCAATTCT  
TACCTTAAATGTCATTGTTGAAGGAAGATTTCATTCAATAAAATTTTTAGAATAATAAGTCCCAGGCACAAGACCAGTATTGTTCTAGGCATTGGGATACCAGT  
GATTACAGGATCAGATGTGGACTCTCAAATTGACTGAGAATAAAACAGACACTAAACAAGTAAATAAGTTAATTCAAGTTGTAATTGATGCTAGAAAGACA  
AGATGAGGGTGGCAGCAGCCTGTTAGATAAGGTACCTGATTGGGGATTGGAAGACCTCTGAGATTAGTGTCTTCAGATATGCCATTGATGATATGAAAG  
AACCGTCTAGGCAGAATGAGCAGCAAGTGCAAGGGCCTGGATAGGAATGAGCTGGATATACTCAAGGAAGAAAGAGAAACTATGAAAATGAAAATGATT  
AAATTACTTTCTTCATTCTTACCTTACCTGTCATGTTATTAAATTTAGAACAATAACATTAAATTCTTATCTCATGTGAAATTCAATTGATACCTTAAATGTCATT  
TTATTCAATTCTTCAATAAAATTTTTAGAATAATAAGTCCCAGGCACAAGACCAGTATTGTTCTAGGCATTGGGATACCAGTTCACAAGACAGACTATGATTACAGGATCAG  
ATTGACTGAGAATAAAACAGACACTAAACAAGTAAATAAGTTAATTCAAGTTGTAATTGATGCTATCCCAGGCACAAGACCAGTATTGTTCTAGGCATTGGGATACCATTACCTGTC  
TTCACTCTTACCTGTCATGTTATTAAATTTAGAACAATAACATTAAATTCTTATCTCATGTGAAATTCAATTGATACCTTAAATGTCATTGTT  
CAATAAAATTTTTAGAATAATAAGTCCCAGGCACAAGACCAGTATTGTTCTAGGCATTGGGATACCAGTTCACAAGACAGACTATGATTACAGGATCAGATGTC  
AACAGACACAAACAAGTAAATAAGTTAATTCAAGTTGTAATTGATGCTATCCCAGGCACAAGACCAGTATTGTTCTAGGCATTGGGATACCATTACCTGTC  
CACATTAAATTCCAAACATGCAAAGAGGAATCTCCATATCATGCTTGTCAATTGTTATTCAAGGGCCAATGTTTCTTGTAAACGTGTAAACATTCTCAGA  
GTGGCCAACATGCAAAGAGGAATCTCCATCTGTCATAACAGTTGATTATTAGAACAATAACATTAAATTCTTATCTCATGTGAAATTCAATTGATACCTTAAATGTCATTGTT  
GATAAGGATAATACAGAGAACATGCCAAAGTTAAGCAAGAAGAAAACAAAGACTGTTACTATGAAAAATGAAAATGATTAAACATGTTAATTCAACGTT  
CTTCTTACCTGTCATGTTATTAAATTTAGAACAATAACATTAAATTCTTATCTCATGTGAAATTCTCATGTTAAACATGTTAATTCAACGTT  
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ACACTAAACAAGTAAATAAGTTAATTCAAGTTGTAATTGATGCTAGAAAGACAATGAAACAGAGCCATGTGACCAATGAGAGAGATGAGGGTGGCAGCAGC  
ATTGGAAGACCTCTGAGATTAGTGTCTTCAGATATGCCATTGATGAAAGAACATTGATGGGCTAGCATTAAAACCGCTAGGCATTGGGATACCATTACCTGTC  
GAGCTGGATATACTCAAGGAAGAAAAGAGAAACTATGAAAAATGAAAATGATTAAACATGTTAATTCAACGTTACTTTGTTAAATTACTTTCTTCATTGTC  
GGAACAATAAAATCACATTAAATTCTTATCTCATGTGAAATTCAATTGATACCTTAAATGTCATTGTTGAAGGAAGATTTCATTGTTCAATTGTC  
AAGACCAGTATTGTTCTAGGCATTGGGATACCAGTTCACAAGACAGACTATGATTACAGGATCAGATGTGGACTCTCAAATTGACTGAGAATAAAACAGACACAAACAAGTAA  
AGTTGTAATTGATGCTACTATGAAAAATGAAAATGATTAAACATGTTAATTCAACGTTACTTTGTTAAATTACTTTCTTCATTGTC  
ATTAAATTCTTATCTCATGTGAAATTCAATTGATACCTTAAATGTCATTGTTGAAGGAAGATTTCATTGTTCAATTGTC  
TTCTAGGCATTGGGATACCAGTTCACAAGACAGACTATGATTACAGGATCAGATGTGGACTCTCAAATTGACTGAGAATAAAACAGACACAAACAAGTAA  
ATCCCAGGCACAAGACCAGTATTGTTCTAGGCATTGGGATACCATTACCTGTCATGTTATTAAATTTAGAACAATAACATTAAATTCAACATGCA  
CGTTTATCAGAGGCCAAATGTTTCTTGTAAACGTGTAAACATTCTCAGAATTAAACAAATAACAAATTGAGGGCTGAATGTGGCCAACATGCAAAGAG  
TGTATTAGAAACTGAGGGCTAAAACGTGTCACATACAGACACACATATTAAATAGATTTCATAATTGGTCTAGGATAAGGATAATACAGAGA  
CAAAGACTGTTACTATGAAAAATGAAAATGATTAAACATGTTAATTCAACGTTACTTTGTTAAATTACTTTCTTCATTCAATAAAATTTTTAGAATAATAAGT  
CTCCTTATCTCATGTGAAATTCAATTGATACCTTAAATGTCATTGTTGAAGGAAGATTTCATTGTTCAATAAAATTTTTAGAATAATAAGT

# What types of annotation do we have/want?

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**~3 billion bp**

```
ACAATAAATCACATTAATTCTTATCTCATGTGAAATTCAATTATGATTG  
ATACCTTTAAATGTCATTGTTGAAGGAAGGATTATTCATTTCATTCAAT  
AAATTTTTAGAATAAAGTCCCAAGGCACAGCAGTATTATGTTCT  
AGGCAATTGGGATACCATGTTCAACAGACAGACTGATTACAGGATC  
AGATGTGGACTCTCAAATCGACTGAGAATAAAACAGACACTAAACAG  
TAATAAAAGTTAATTCAAGTTAATTGATGCTGAGAAAGACAA  
GAGGCAATTGGGATACCATGAGAGAGATGAGGGTGCAGCAGCTGTTA  
GATAAGGCTCTGATTGGTGGATTGGAAGACCTCTGAGATTGTTG  
CTTCAGATATGCCATTATGATGATGAAAGAACCATTCATGGAGGCCTAG  
CATTAAGGAAACCGTCTAGGCAGAACTGAGCAGCAAGTGCAGGGCTGG  
ATAGGAATGAGCTGATATACTCAAGGAAAGAGAAACTATGAAAAA  
ATGAAAATAGATTTAAACATGTTAATTCACTGTTACGTTACCTTTGTTAAATT  
CTTTCTCTTCACTTCACTGCAATGTTAAATATTGAAACA  
ATAAAITCACATTAATCCTTATCTCATGTGAAATTCAATTATGATTGATA  
CCTTTAAATGTCATTGTTGAAGGAAGGATTATCATTTCATTCAAA  
TATTTTAAAGAATAAAGTCCCAAGGCACAGCAGTATTATGTTCTAGG  
CATGGGGATACCATGTTCAAAAGACAGACTATGATTACAGGATCAGAT  
GTGGACTCTCAAATCGACTGAGAATAAAACAGACACTAAACAAGTAAT  
AAAGTTAATTCAGTGTAATTGATGCTACTATGAAAAAAATGAAAATAGA  
TTTTAAACATGTTAATTCACTGTTACCTTTGTTAAATTACTTTCTCTTT  
CACTTCTTACCTGCAATGTTAAATATTGAAACAATCACATT  
AATTCTTATCTCATGTGAAATTCAATTATGATTGATAACCTTAAATGT  
CATTTGTTGAAGGAAGGATTATCATTTCATTCAATAAATATTGAA  
ATAATAAGTCCCAAGGCACAGCAGTATTATGTTCTAGGATGGGAT  
ACCATGTTCAAAAGACAGACTATGATTACAGGATCAGATGTTGACTCTC  
AAATTGACTGAGAATAAAACAGACACAAACAAGTAATAAAGTTAATT  
CAAGTTGTAATTGATGCTATCCCAAGGCACAGACCA....
```

## Genes:

- Coding, noncoding, miRNA, etc.
- Isoforms
- Expression

## Genetic variation:

- SNPs and CNVs

## Sequence conservation

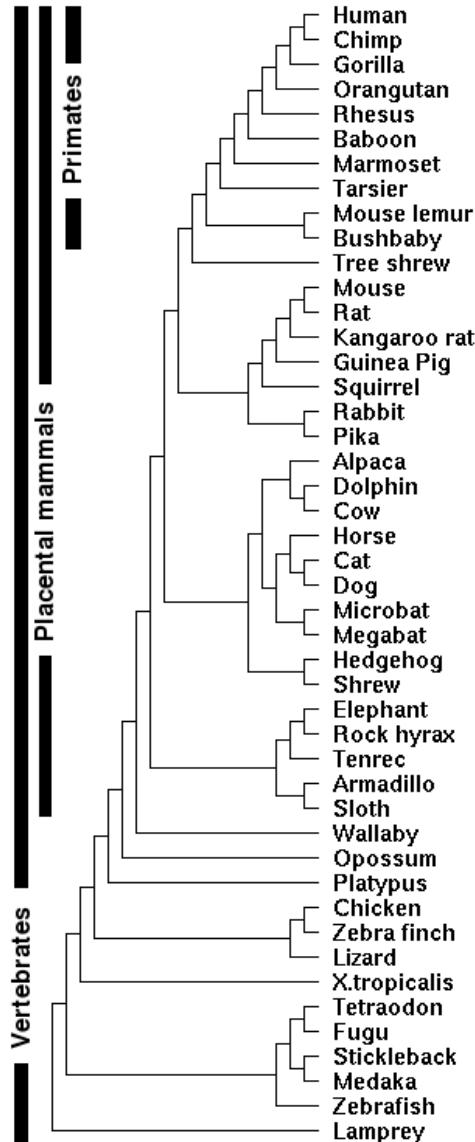
## Regulatory sequences:

- Promoters
- Enhancers
- Insulators

## Epigenetics:

- DNA methylation
- Chromatin

# Degrees of genomic annotation vary widely



## ENCODE and modENCODE

### Human, Mouse (Fly, Worm, Yeast):

- Chromosome assemblies
- Dense gene and regulatory maps, variation, etc.

### Other models (Dog, Chicken, Zebrafish):

- Chromosome assemblies
- Partial gene maps; variation; little regulatory data

### Low coverage vertebrate genomes:

- Scaffold assemblies
- Few annotated genes
- Used for comparative purposes

# Where do you look for existing annotations?

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**UCSC Genome Browser** ([genome.ucsc.edu](http://genome.ucsc.edu)):

Visualization, data recovery, simple analysis  
(also <http://genome-preview.ucsc.edu/>)

**ENSEMBL** ([ensembl.org](http://ensembl.org)):

Visualization, data recovery, simple analysis

**Integrative Genomics Viewer**

([broadinstitute.orgsoftware/igv/](http://broadinstitute.org/software/igv/)):

Local genome viewer (visualize local and remote data)

**Galaxy** ([main.g2.bx.psu.edu](http://main.g2.bx.psu.edu)):

Complex data analysis and workflows

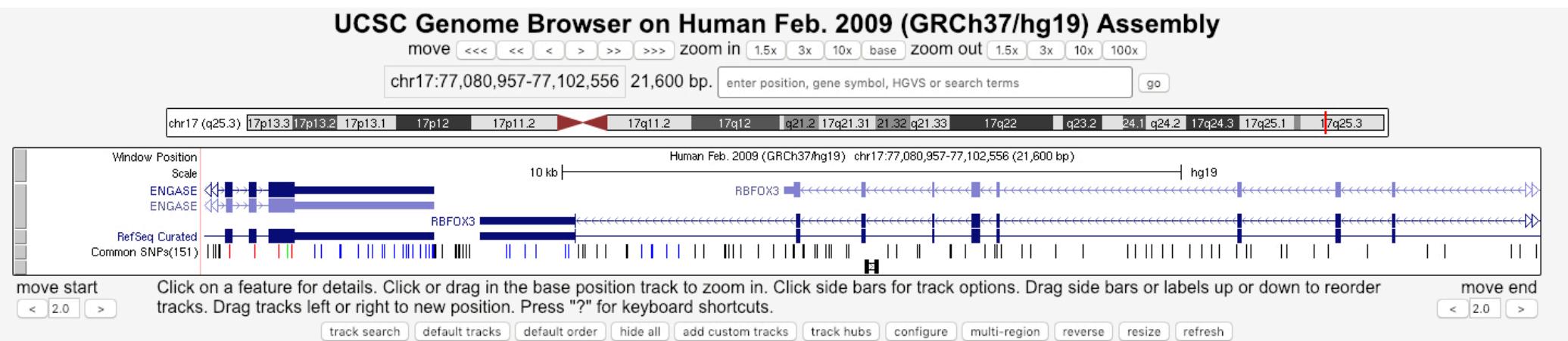
# Example of a genome browser track (UCSC)

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Chr5: 133,876,119 – 134,876,119

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## Our specific example:

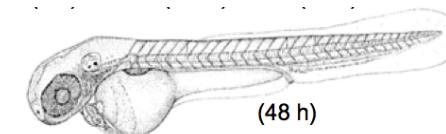


# Workflow

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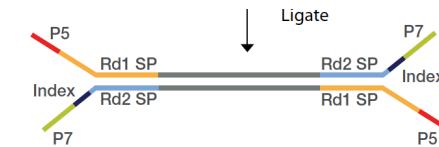
## 1. Isolation of sample.

e.g., Isolate DNA and shear.



## 2. Library preparation

e.g., Add known sequences to the ends.



## 3. Sequencing

e.g., Illumina Novaseq

## 4. Analysis

e.g., Map to genome and interpret.



# Using sequencing to annotate the genome

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1. Where are the cis-acting regulatory elements in DNA?
  - A. DNase I hyper-sensitivity mapping (DNase-Seq).
  - B. FAIRE to map regulatory elements.
2. Where do transcription factors bind?
  - C. ChIP-seq of transcription factors (or in high res, ChIP-exo)
  - D. Nucleosome mapping (MNase-Seq).
3. Where are different histone modifications found?
  - E. ChIP-Seq of histone modifications.
  - F. ChIP-Seq of chromatin writers, readers and erasers.
4. Where is RNA polymerase transcribing?
  - G. ChIP-Seq of polymerase.
  - H. GRO-Seq, NET-Seq and TT-Seq to measure RNA in the polymerase active site..
5. How is the genome organized in 3D?
  - I. 4C/5C/Hi-C to measure chromatin conformation.

Applications of sequencing technology next week.

## Conclusions

- High-throughput sequencing has become democratized - moved out of industrial-scale genome centers
- Sequence is no longer limiting - next generation of sequencers will make sequencing very inexpensive
- Earlier methods for counting / resequencing applications are largely obsolete
- Scale of data production outstripping our ability to store and analyze it