

Bioinformatic approaches to regulatory genomics and epigenomics

376-1347-00L - 2022 | week 03

Pierre-Luc Germain

Plan for today

- Debriefing on the assignments
- Overview of NGS basic analysis pipelines, file formats, etc.

How many protein-coding gene IDs, and how many gene symbols, does the mouse annotation have?

```
#Filter only the protein_coding genes from the original database  
#supportedFilters()  
mouse_pc <- genes(mouse_ensdb, filter = GeneBiotypeFilter("protein_coding"), columns = c("gene_id", "symbol"))  
  
#Get the number of different IDs from this filter  
length(unique(mouse_pc$gene_id))
```

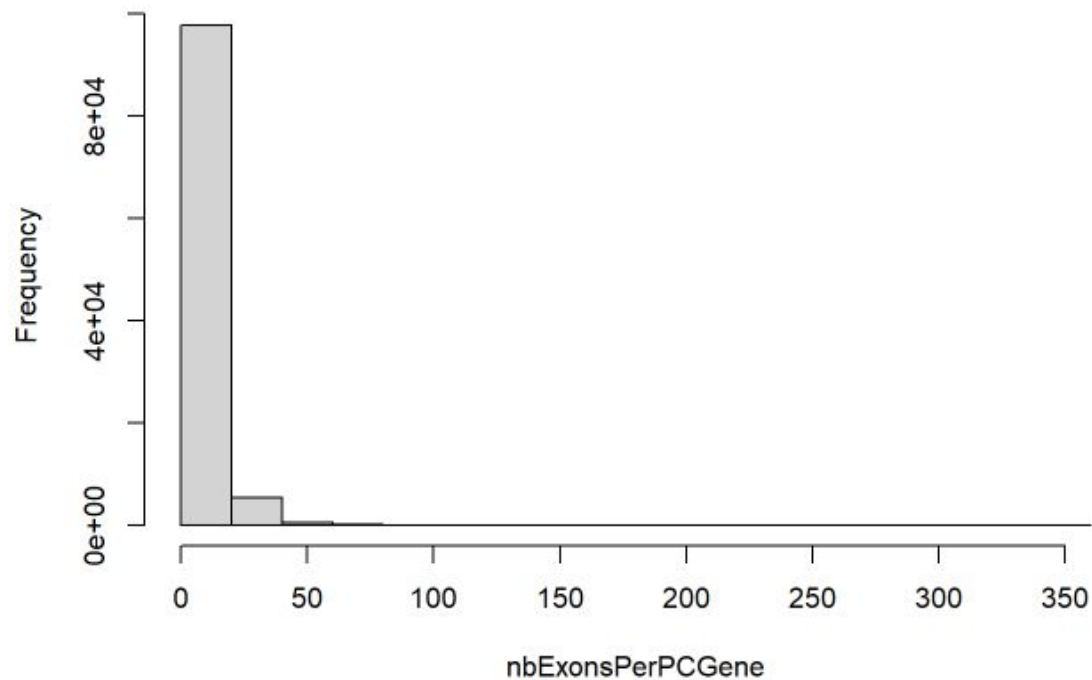
```
## [1] 22287
```

```
#Get the number of different gene symbols from this filter  
length(unique(mouse_pc$symbol))
```

```
## [1] 21964
```

```
#genes  
exsPerGene <- exonsBy(ensdb_m102, column=c("gene_id", "gene_biotype"),  
                      filter=GeneBiotypeFilter("protein_coding"))  
  
#exsPerGene  
nbExonsPerPCGene <- lengths(exsPerGene)  
hist(nbExonsPerPCGene)
```

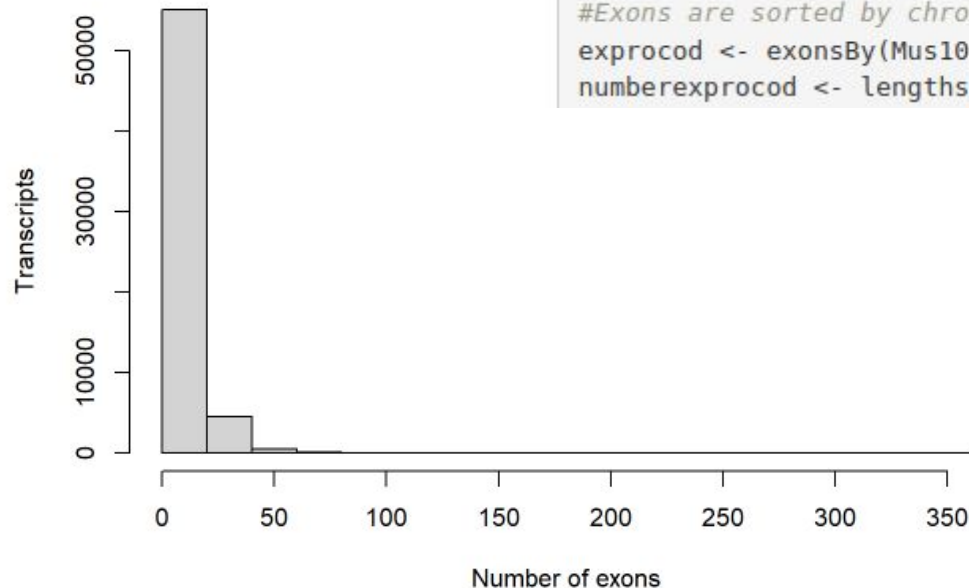
Histogram of nbExonsPerPCGene



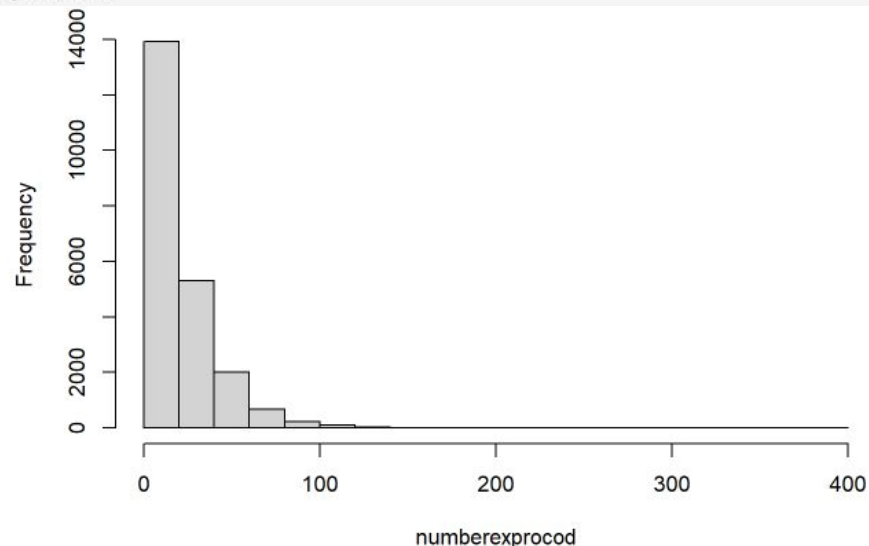
Plot the distribution of the number of exons for protein-coding genes

```
exsPerTx <- exonsBy(ensdb, column=c("tx_id", "tx_biotype"),  
                    filter=TxBiotypeFilter("protein_coding"))  
nbExonsPerPCTx <- lengths(exsPerTx)  
hist(nbExonsPerPCTx, main="Number of exons per transcript",  
     xlab="Number of exons", ylab="Transcripts")
```

Number of exons per transcript



```
#Exons are sorted by chromosome, strand, start and end values by using by = gene  
exprocod <- exonsBy(Mus102, by = "gene", filter=GeneBiotypeFilter("protein_coding"))  
numberexprocod <- lengths(exprocod)
```



Plot the distribution of the (spliced) length of protein-coding transcripts

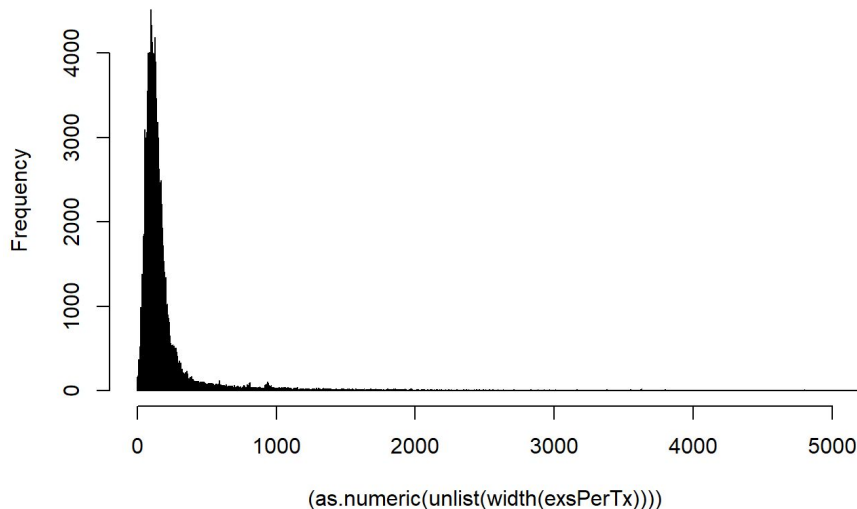
#Plot the distribution of the (spliced) length of protein-coding transcripts

```
#as.numeric(unlist(width(exsPerTx)))  
head(width(exsPerTx))
```

```
## IntegerList of length 6  
## [["ENSMUST00000000001"]] 259 43 142 158 129 130 154 210 203  
## [["ENSMUST00000000003"]] 215 140 68 111 102 52 214  
## [["ENSMUST00000000010"]] 602 1972  
## [["ENSMUST00000000028"]] 169 195 60 93 138 144 56 ... 162 1  
## [["ENSMUST00000000033"]] 109 163 149 3287  
## [["ENSMUST00000000049"]] 115 177 97 77 189 180 198 157
```

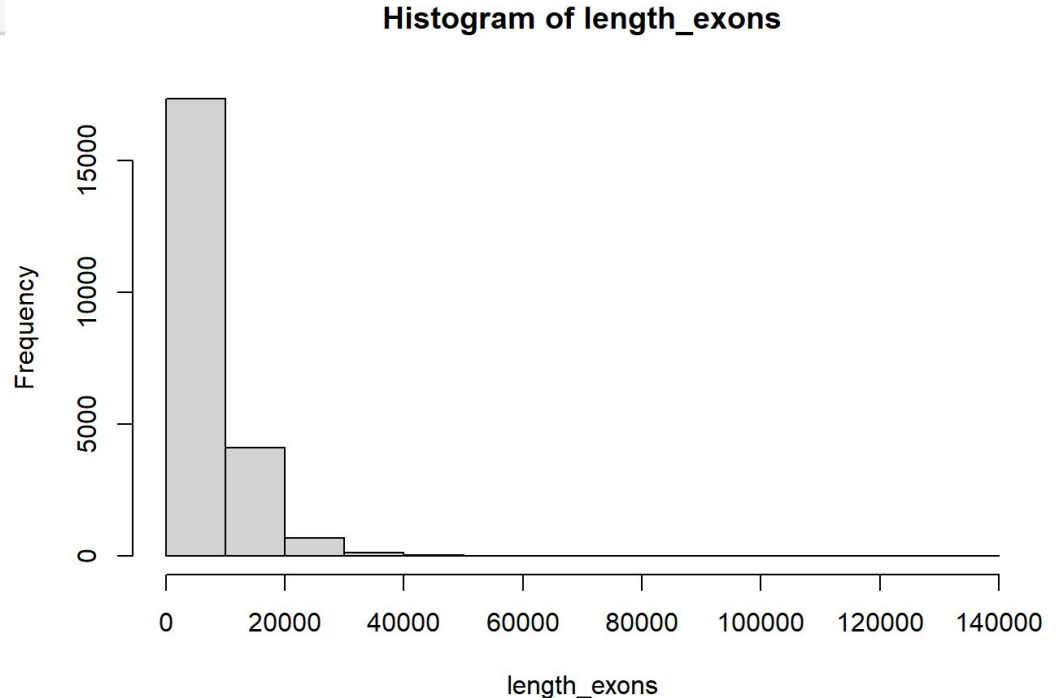
```
hist(as.numeric(unlist(width(exsPerTx))), xlim = c(0,6000))
```

Histogram of (as.numeric(unlist(width(exsPerTx))))



Plot the distribution of the (spliced) length of protein-coding transcripts

```
length_exons <- sum(width(exsPerGene))  
hist(length_exons)
```



misc...

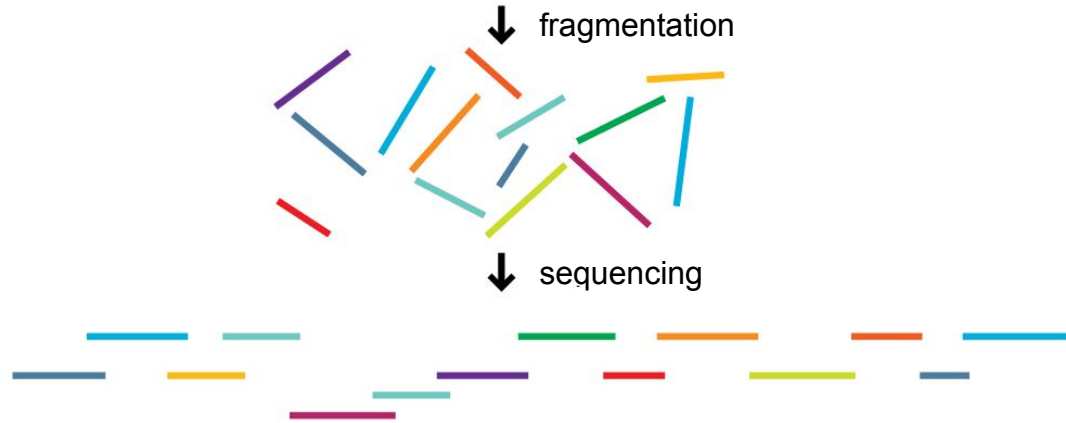
#Plot the distribution (histogram) of how many exons each protein-coding gene has:

```
# look at exons with 'exonsBy'  
# Find how many exons each protein coding gene has  
# The by tells you whether exons should be fetched by transcript or by gene (in TranscriptsBy it tells you whether to fetch by genes or by exons)  
exsPerGene <- exonsBy(ensdb, by = "gene", filter = GeneBiotypeFilter("protein_coding"))  
exsPerGene
```


Next Generation Sequencing (NGS)

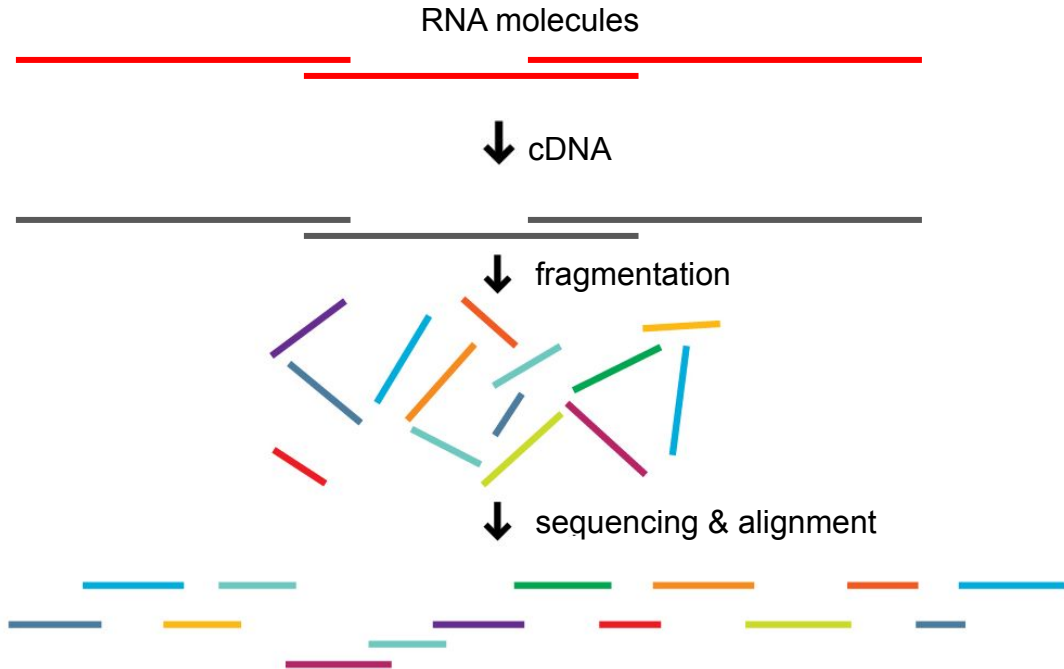
Shotgun sequencing:

Large DNA molecule



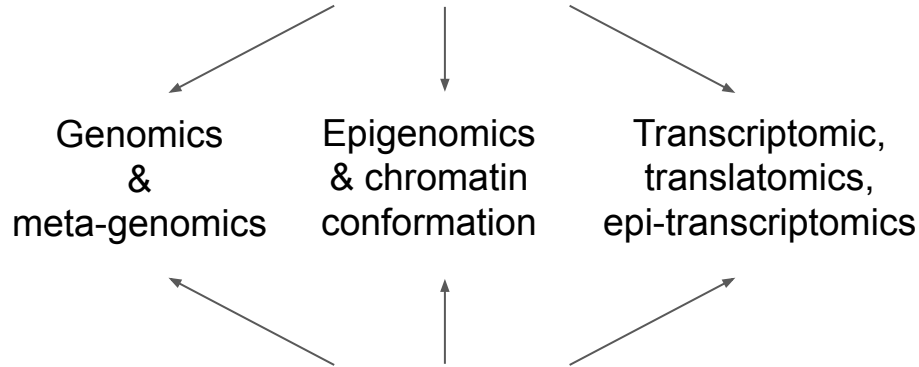
Next Generation Sequencing (NGS)

RNA sequencing:

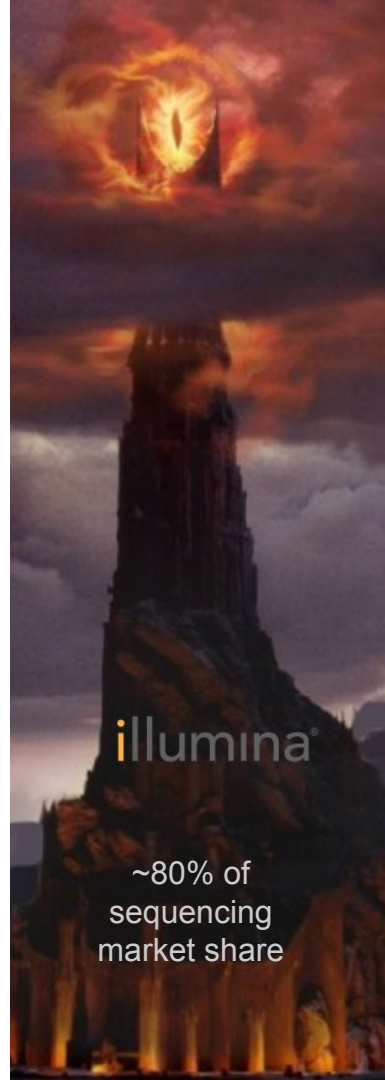


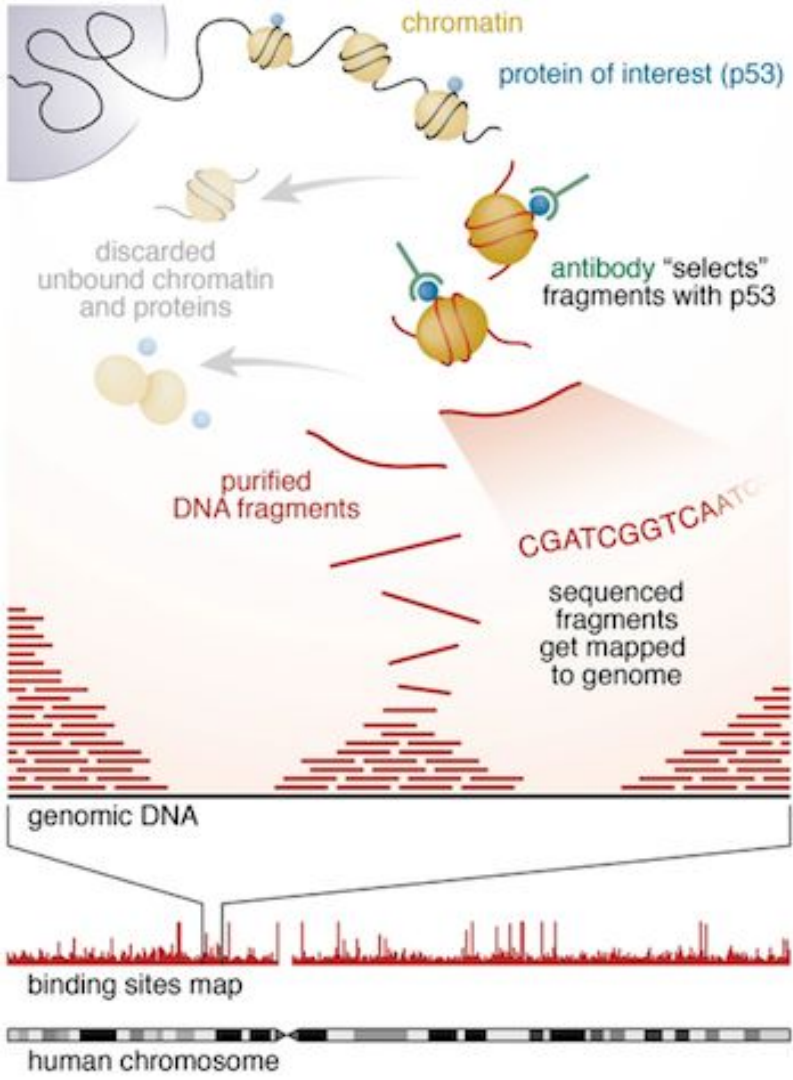


Next Generation Sequencing: one technology to rule them all

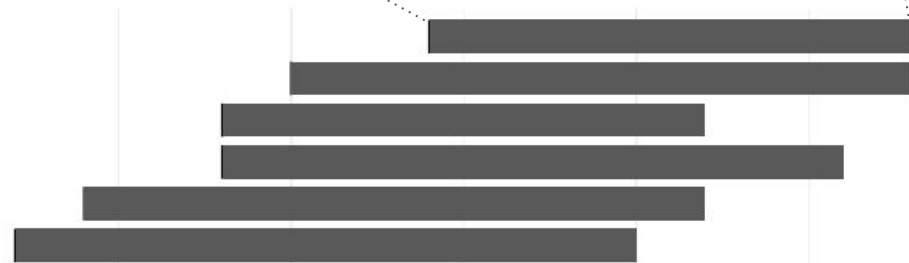


A lot of convergence in terms of analysis
tools and techniques

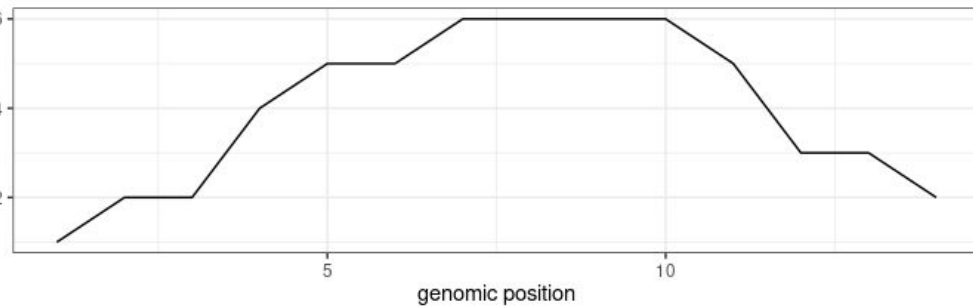




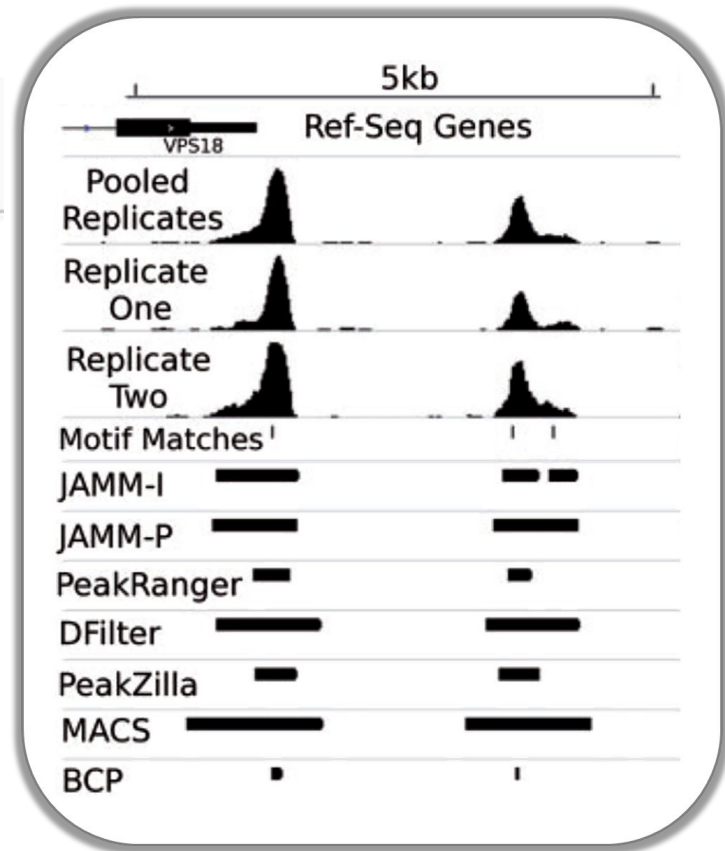
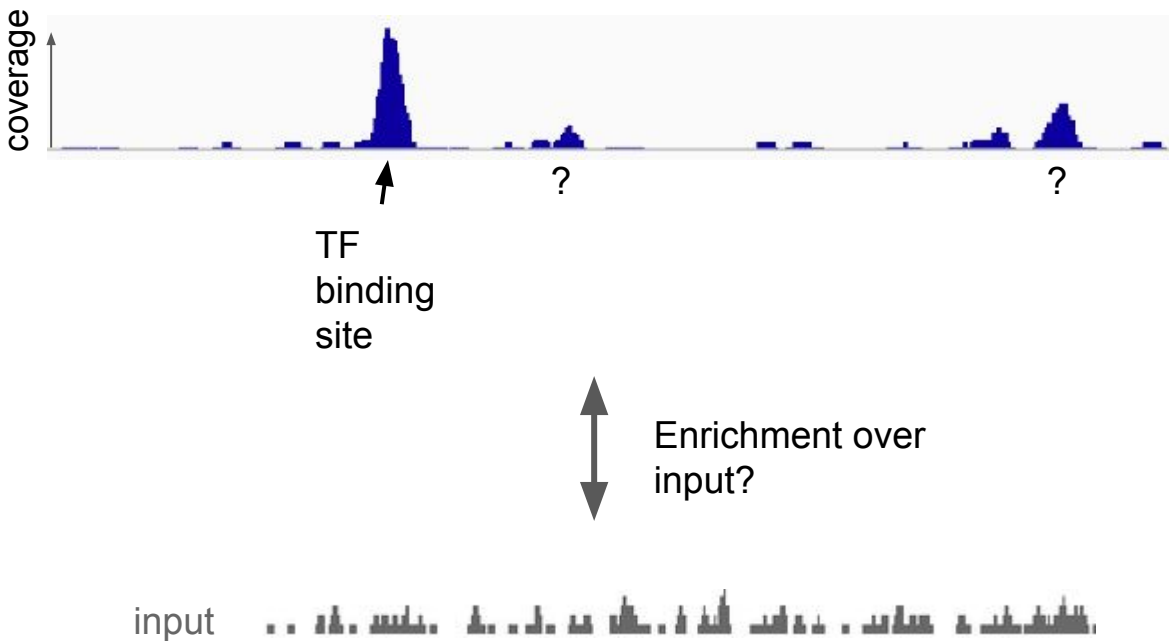
reads / fragments

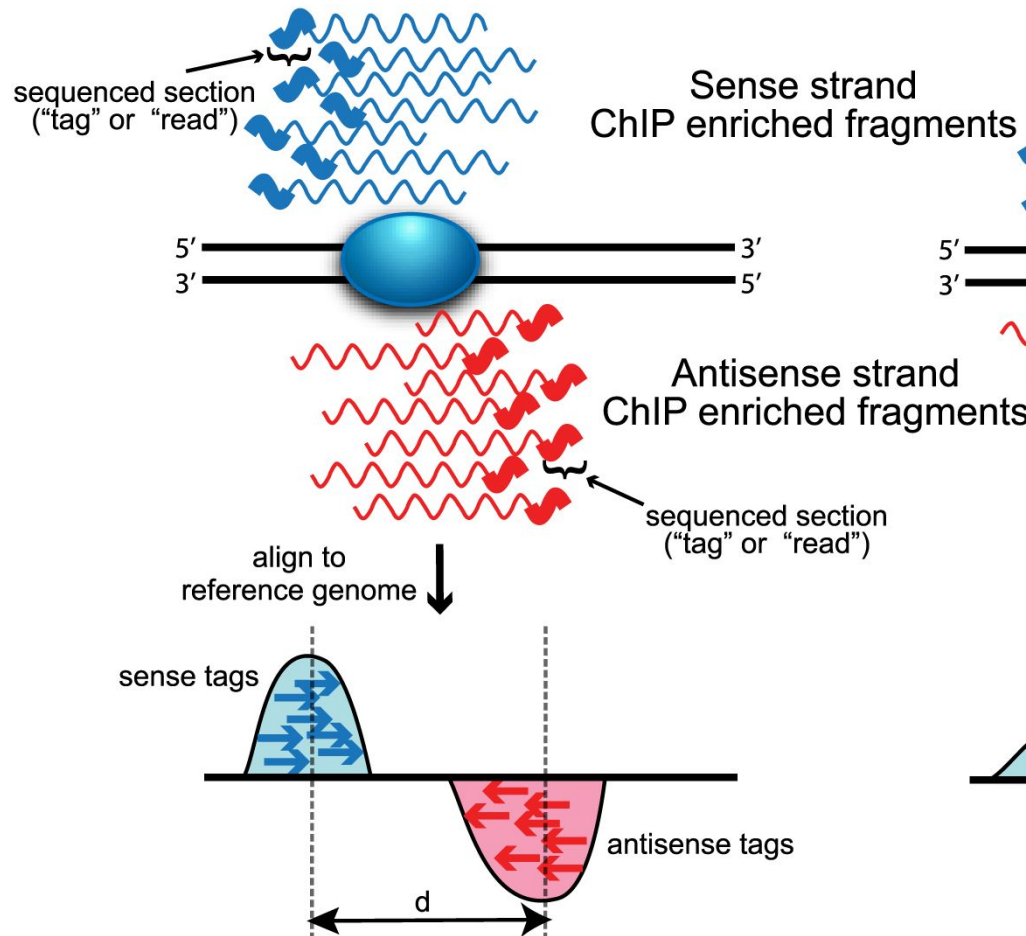
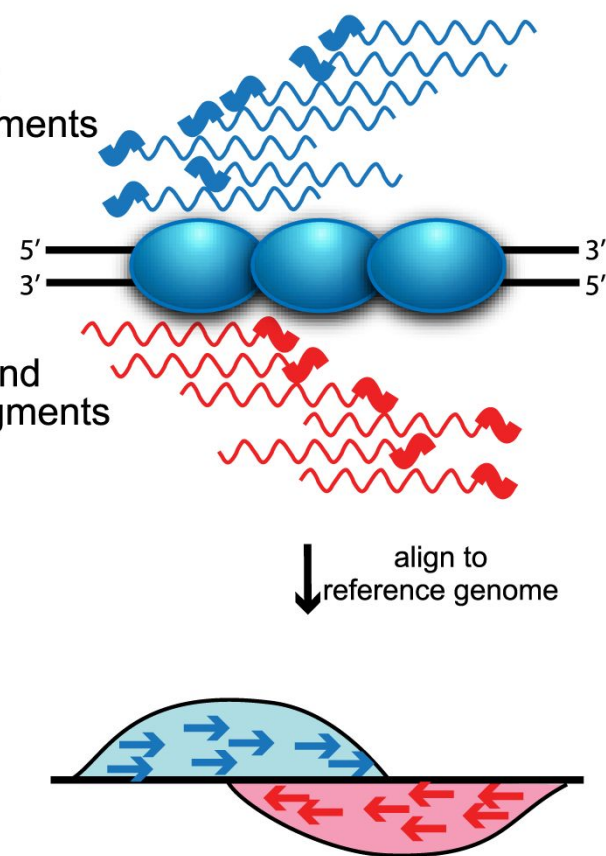


coverage (aka pileup)

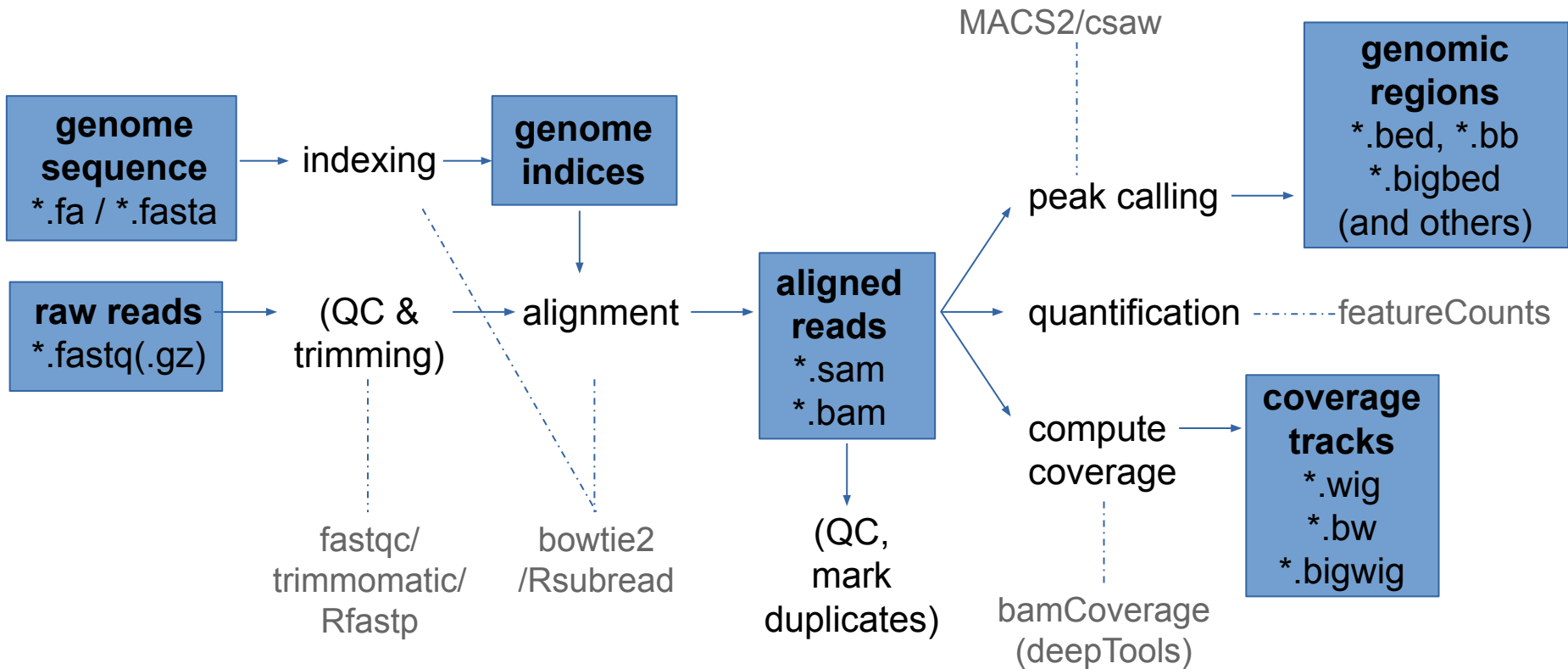


Peak calling



A**B**

Overview of a primary analysis pipeline (ChIP-seq and the likes)



Alternative toolsets for (DNA) primary analysis

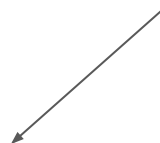
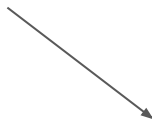
- The most standard one:

- [fastqc](#)
- [trimmomatic](#)
- [bowtie2](#)
- [picard](#)
- [deeptools](#)

- Pure R-based

- [rfastp](#)
- [Rsubread](#)

[QuasR](#)

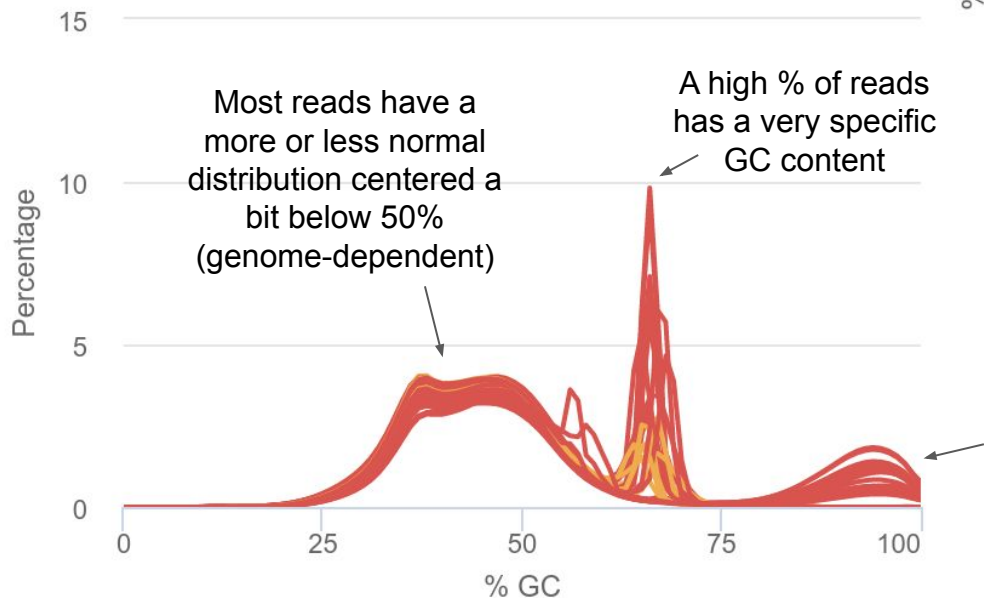


Downstream analysis (R)

- [epiwraps](#)

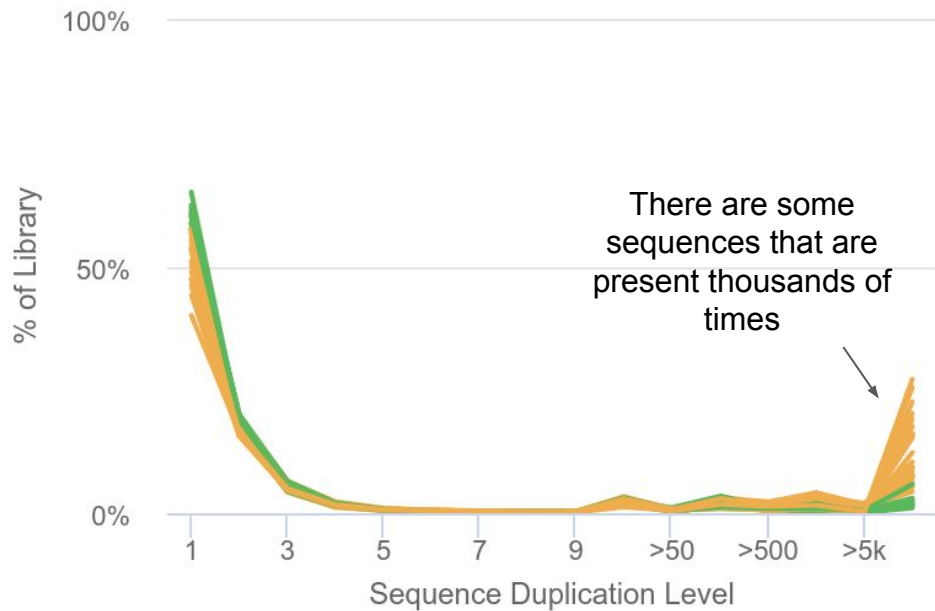
Example (rather extreme) QC problems

FastQC: Per Sequence GC Content



Created with MultiQC

FastQC: Sequence Duplication Levels

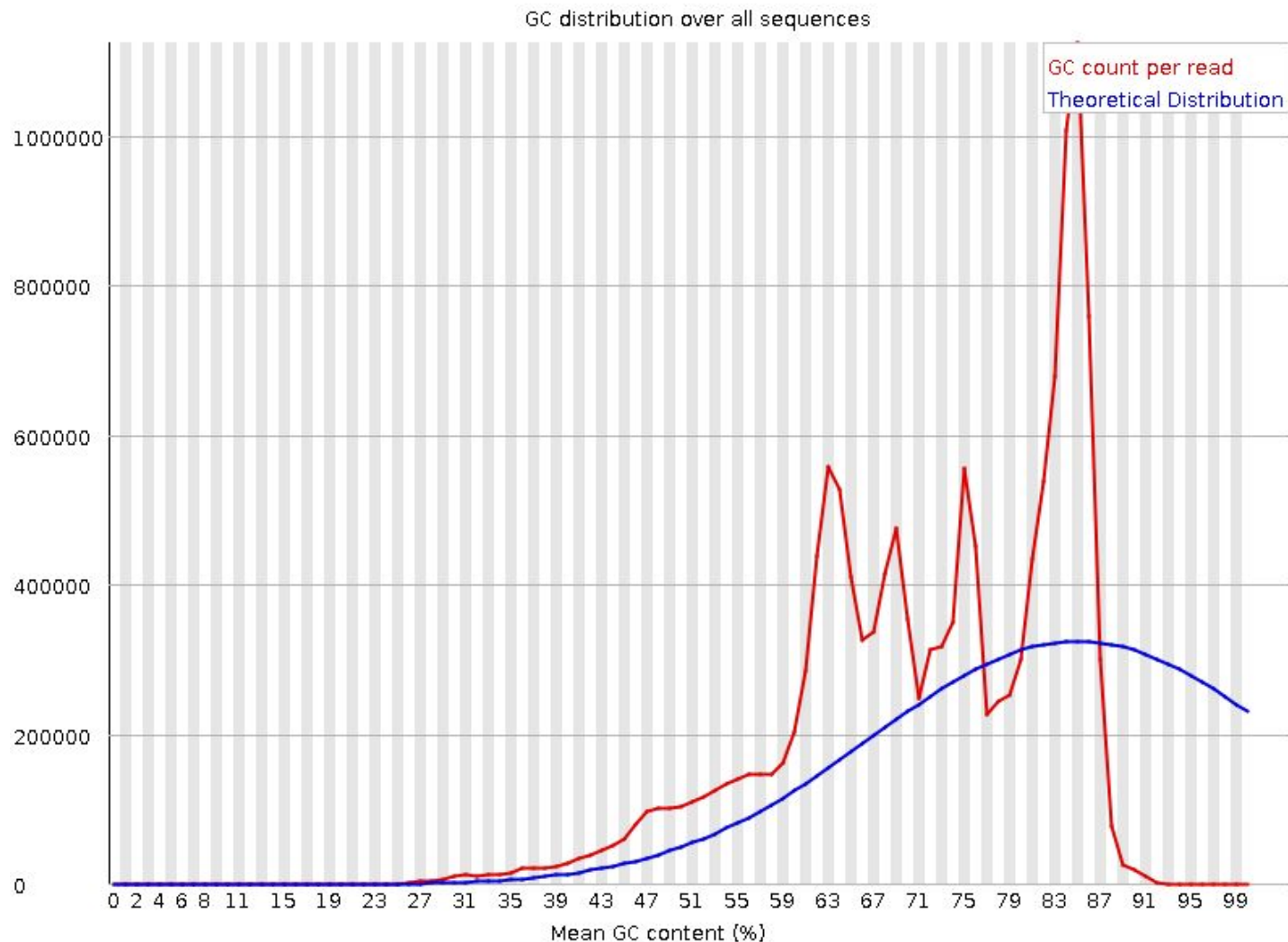


Created with MultiQC

Example (rather extreme)

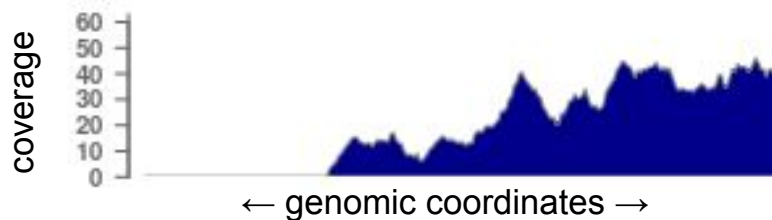
QC problems:

Bias from overamplification



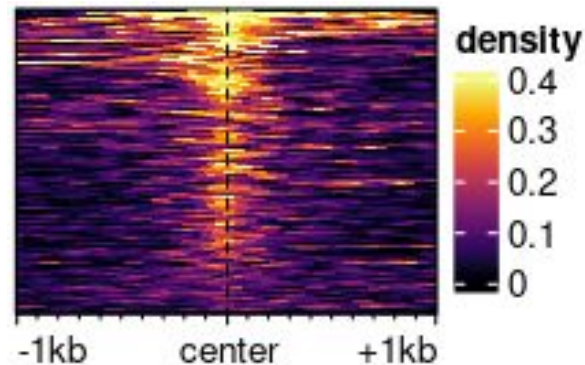
Visualizations available in *epiwraps*

- Signal across one genomic region:
`plotSignalTracks`



(Based on the *Gviz* R package)

- Signal across several genomic regions:
`signal2Matrix` →
`plotEnrichedHeatmaps`



(Mainly based on the *EnrichedHeatmap* R package, itself based on *ComplexHeatmap*)

Assignment

- Download a mouse ChIPseq dataset
 - Download and process it from the raw data, obtaining:
 - bam file, along with number and percentage of mapped reads
 - bigwig file
 - peaks
 - How many peaks do you find?
 - Plot the signal round one of the peaks
 - Please make sure that you name your final file **assignment.html** !!
- Suggested dataset:
 - p300 in mESC:
<https://www.encodeproject.org/files/ENCFF001LJN/@@download/ENCFF001LJN.fastq.gz>
 - the corresponding input control would be:
<https://www.encodeproject.org/files/ENCFF001KEU/@@download/ENCFF001KEU.fastq.gz>