

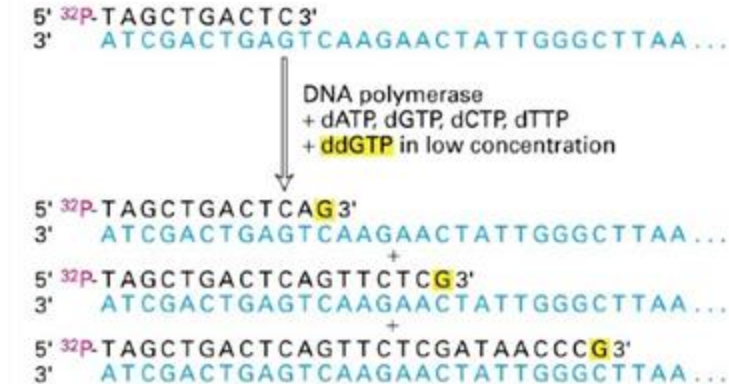
Long Read Sequencing

Chris Miller, Ph.D.
Washington University in St Louis

How to sequence a human genome: Sanger method

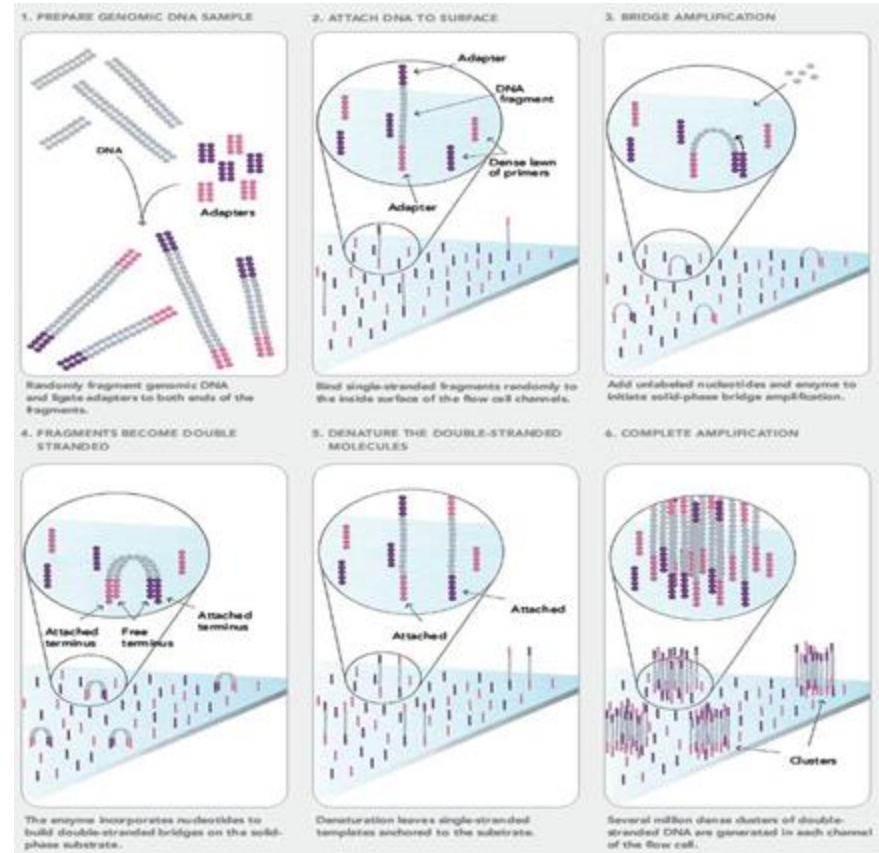
Key points:

- 1) sequencing by synthesis (not degradation)
- 2) primers hybridize to DNA
- 3) polymerase + dNTPS + ddNTP terminators at low concentration
- 4) 1 lane per base, visually interpret ladder

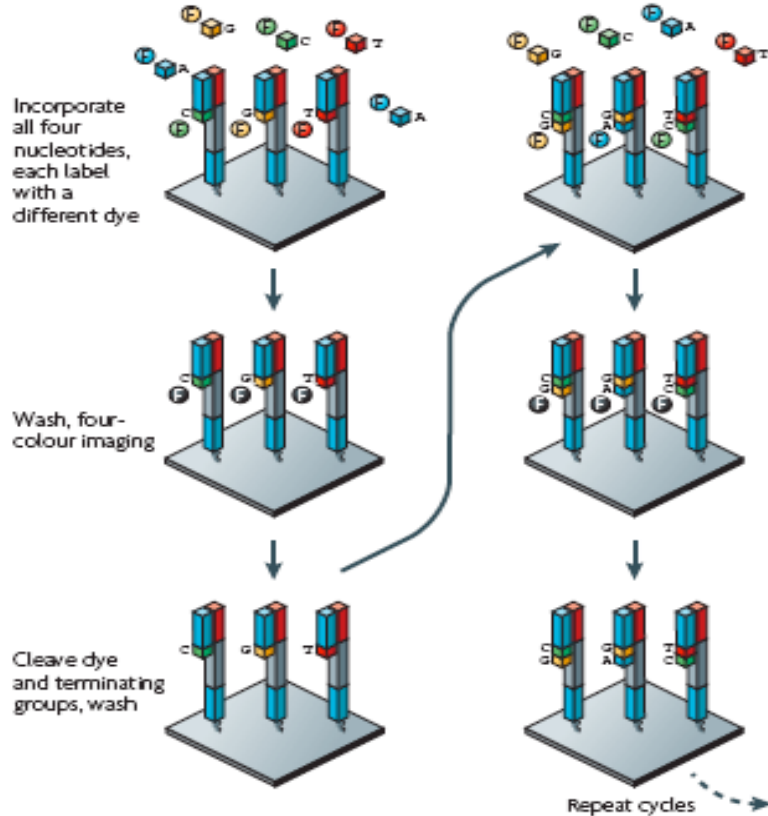


Solexa (Illumina) sequencing (2006)

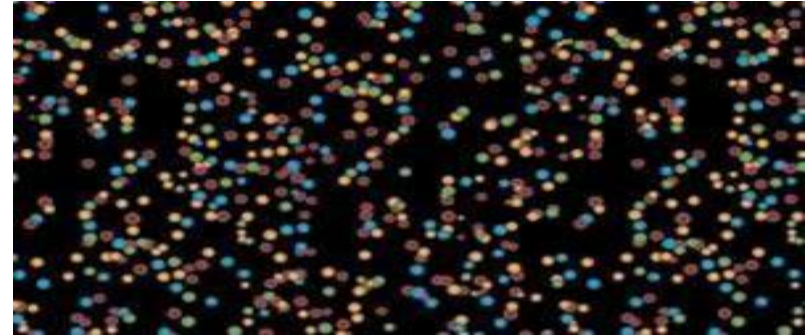
- PCR amplify sample (opt.)
- Immobilize and amplify single molecules on a solid surface
- Reversible terminator sequencing with 4 color dye-labelled nucleotides



Illumina sequencing (2005)



4 different images merged



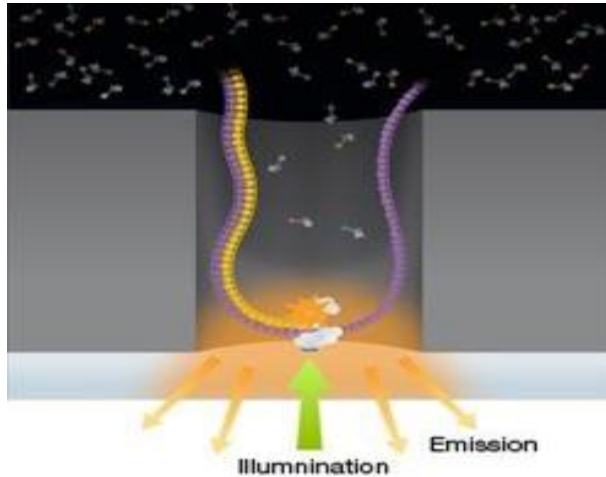
6 cycles w/ base-calling



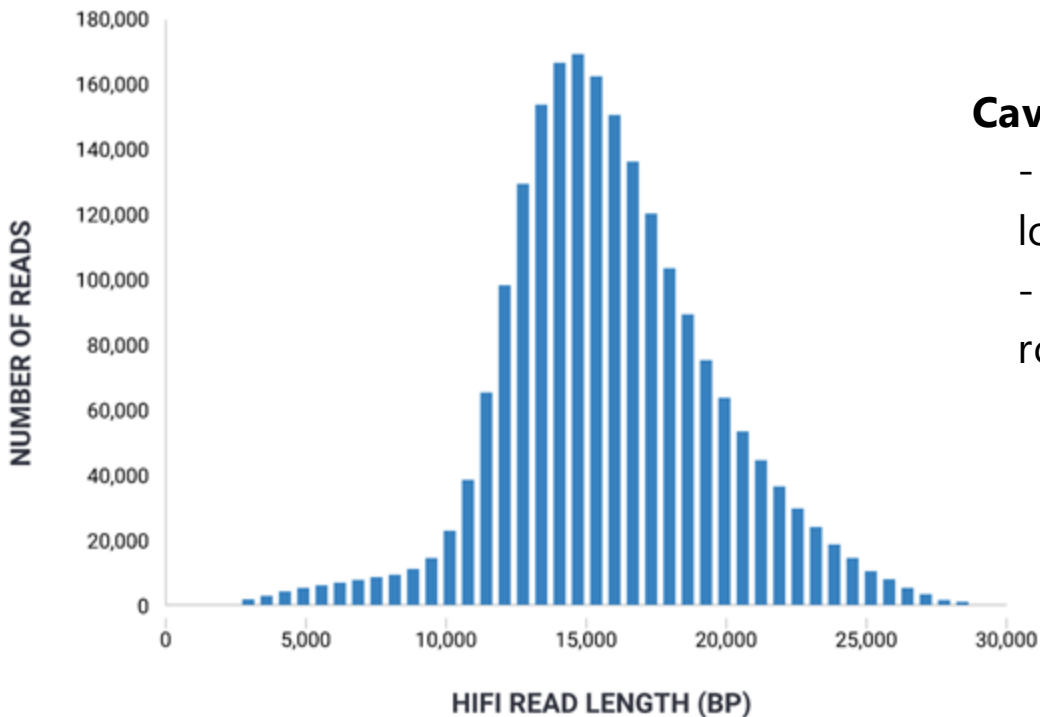
Pacific Biosciences

Key Points:

- 1 DNA molecule and 1 polymerase in each well (zero-mode waveguide)
- 4 colors flash in real time as polymerase acts
- Methylated cytosine has distinct pattern
- No *theoretical* limit to DNA fragment length



Pacific Biosciences: long reads. Great for genome assembly

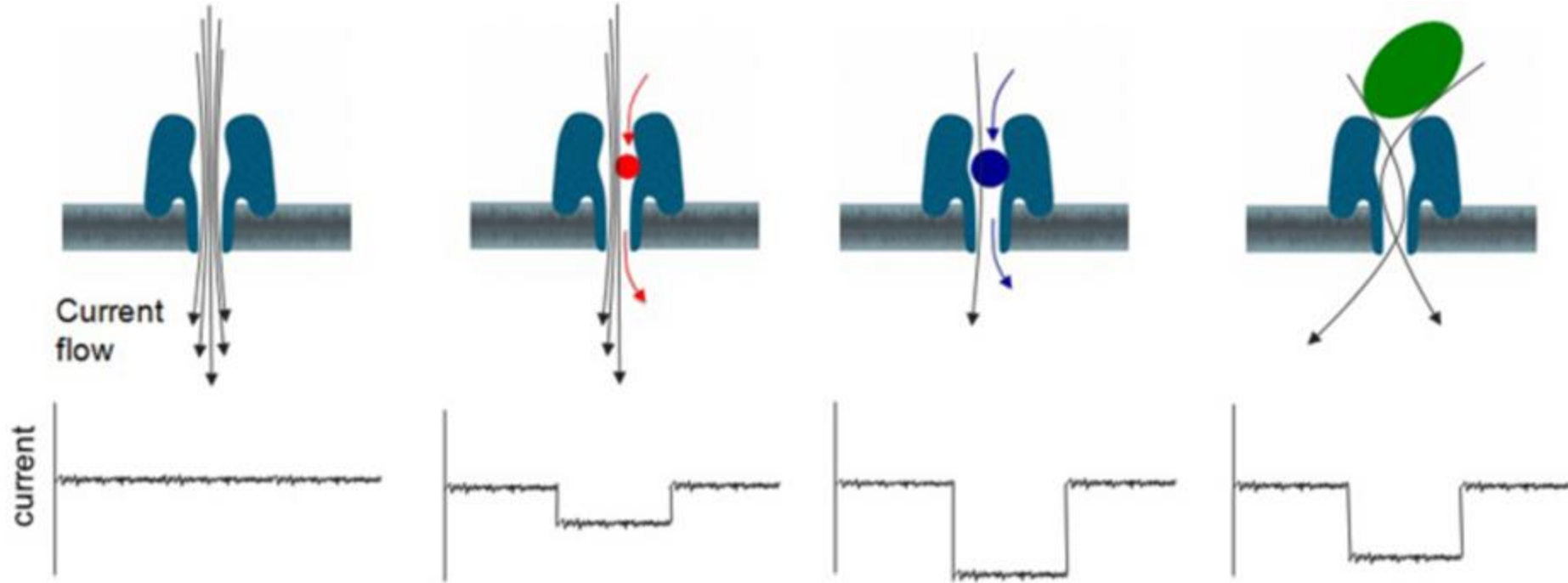


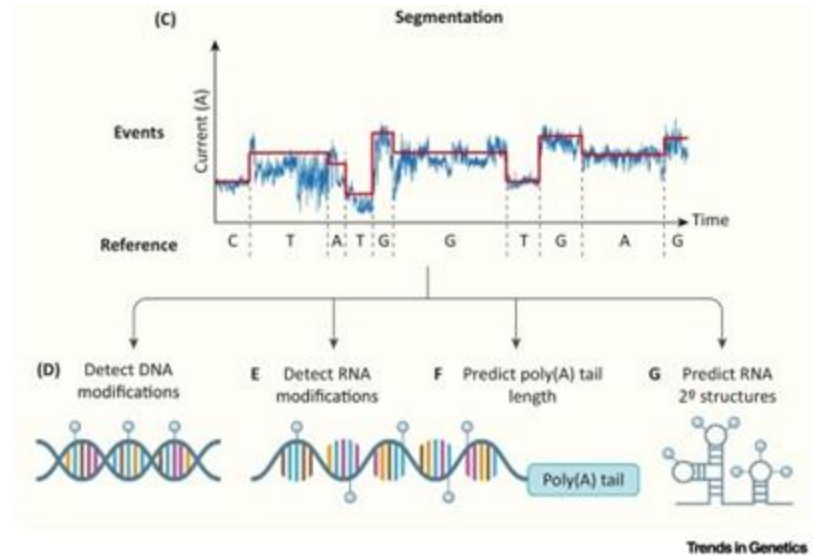
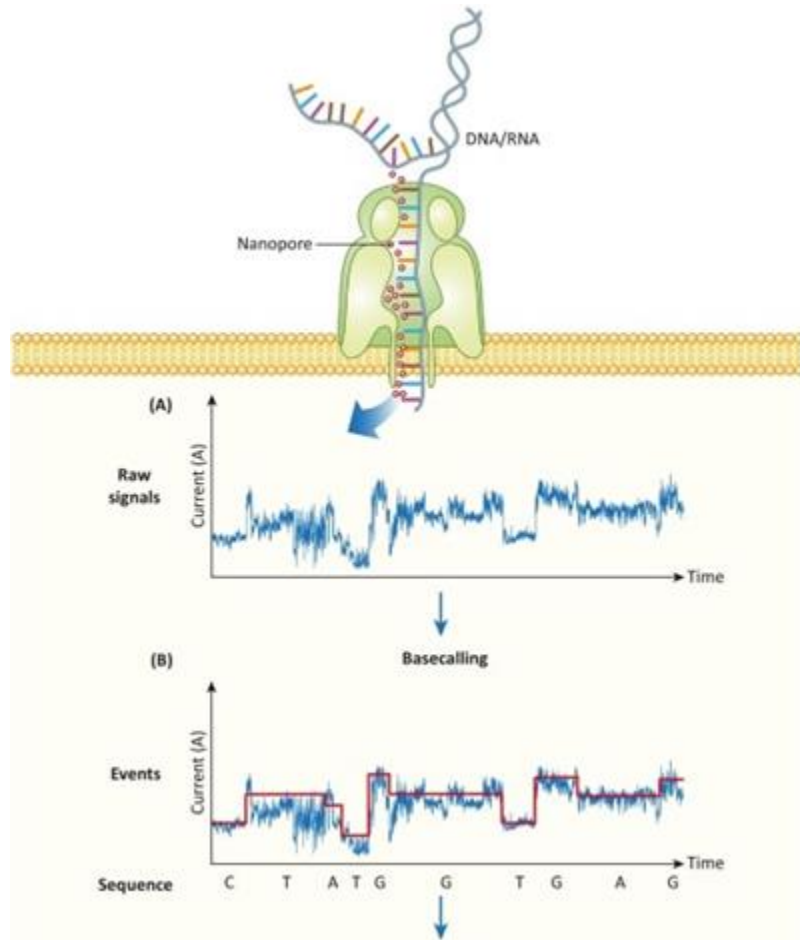
Caveats:

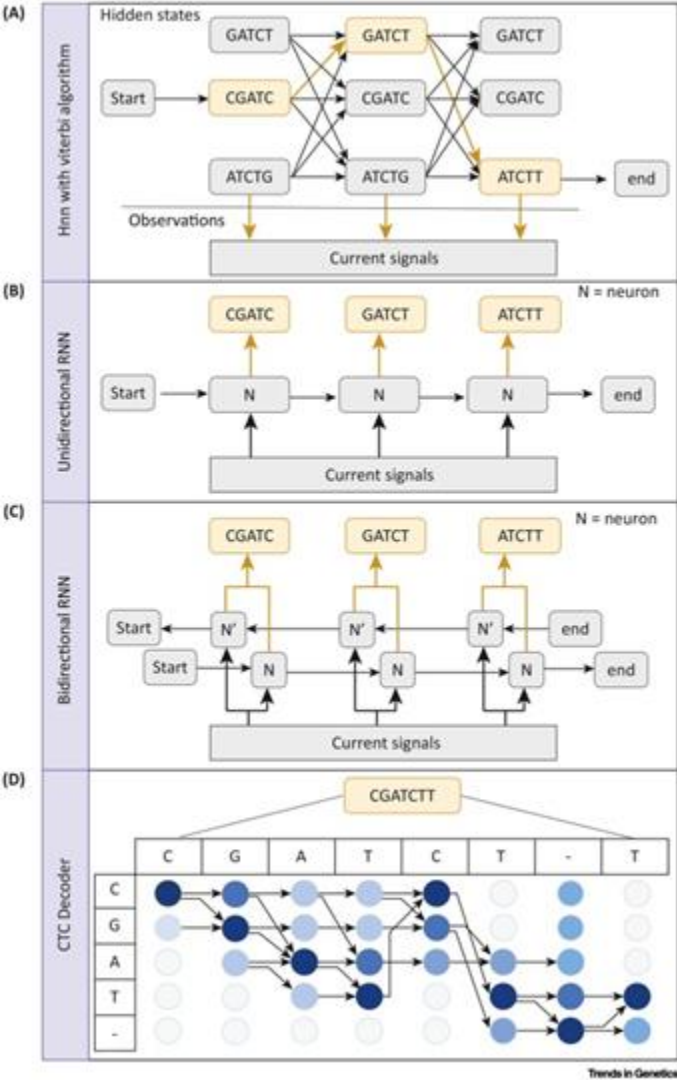
- higher error rate (1-2%), lower with Duplex runs
- lower throughput : roughly 90 gigabases per run

About \$4,000 for a 30x human genome on the RevIO machine

Oxford Nanopore Technologies





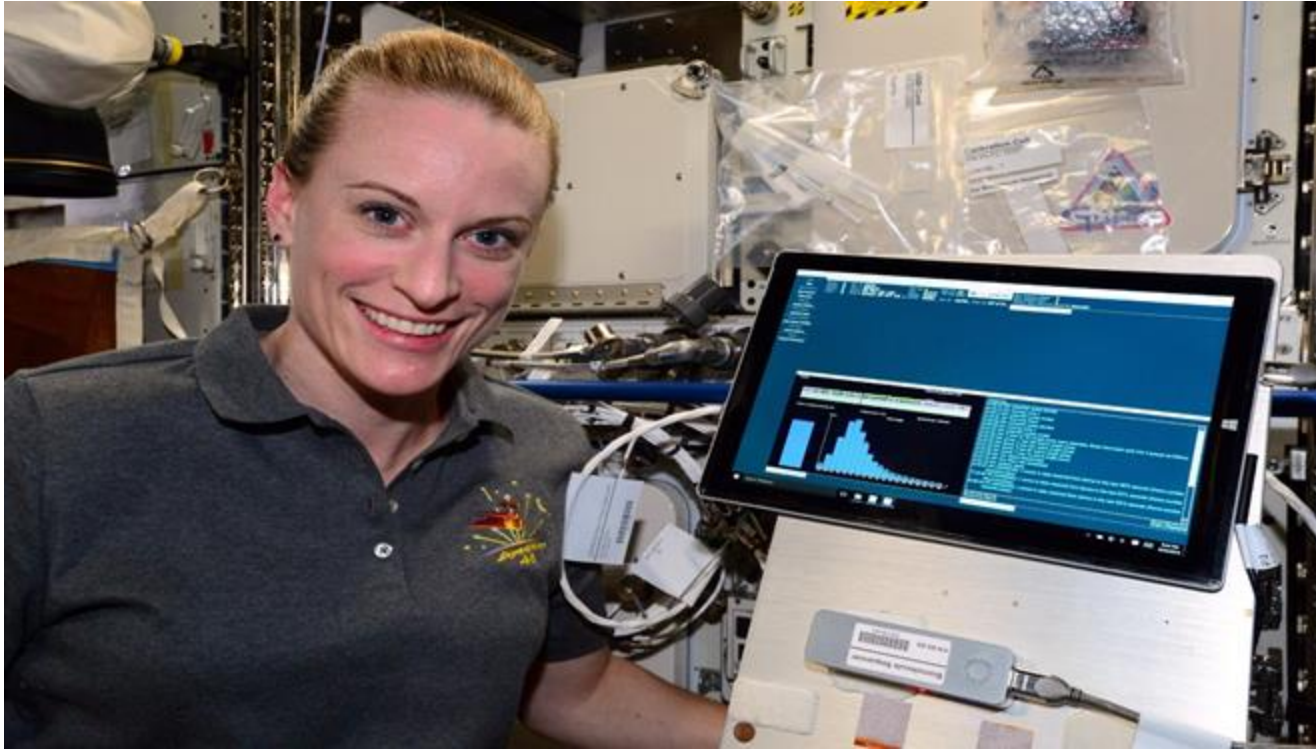


Neural networks to translate signal into base calls

- Guppy (many versions)
- Dorado (v0.4, eventual guppy replacement)
- many others

Practically, that means that we can't yet throw away our raw signal intensities. (1 Tb or more per run)

Nanopore sequencing is *extremely* portable



Kate Rubins sequencing DNA on the ISS

ONT sequence length distribution

Legend

Basecalled Estimated

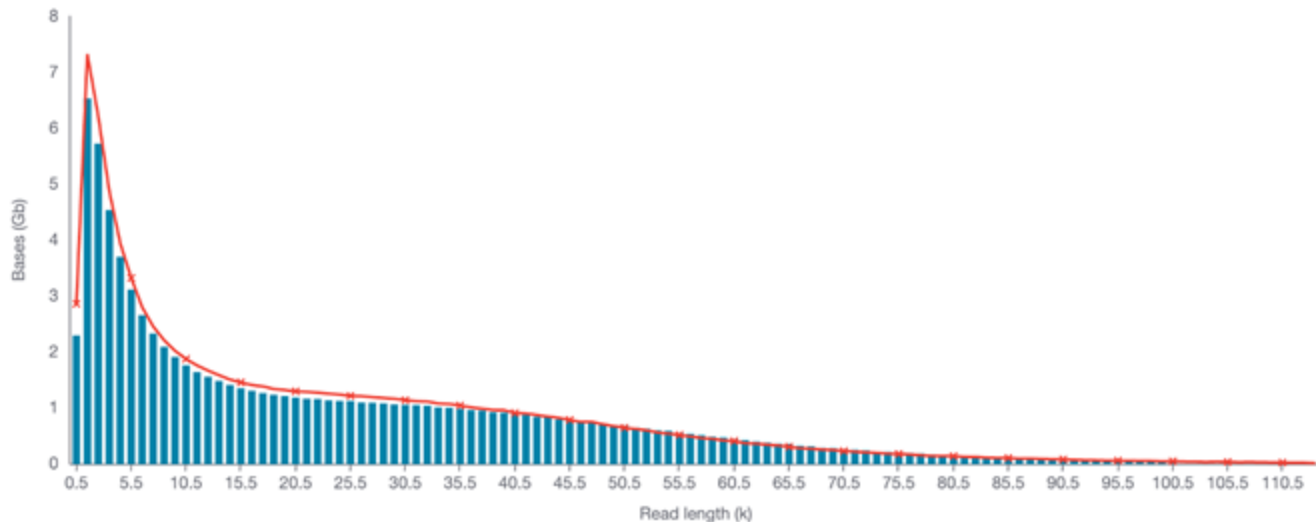
Estimated N50

17.75 kb

% Basecalled

100%

their relative amounts.

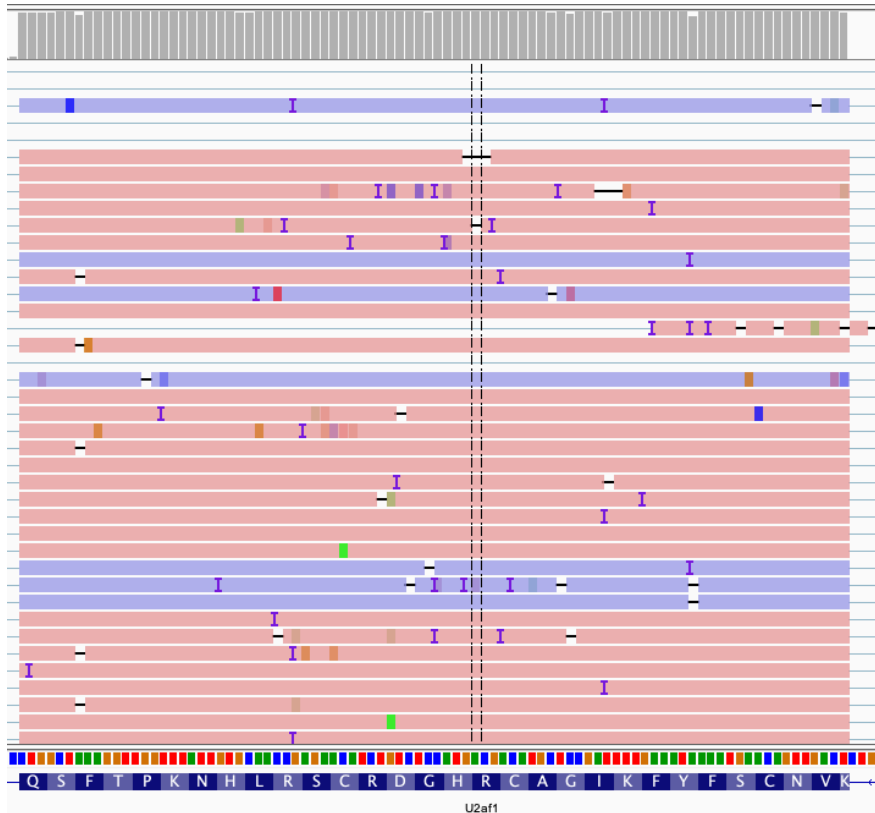


Read length (kb)	Aggregated reads (Mb)
100 - 164	886.98
164 - 228	36.06
228 - 292	4.02
292 - 344	0.35

Recent run of a tumor sample

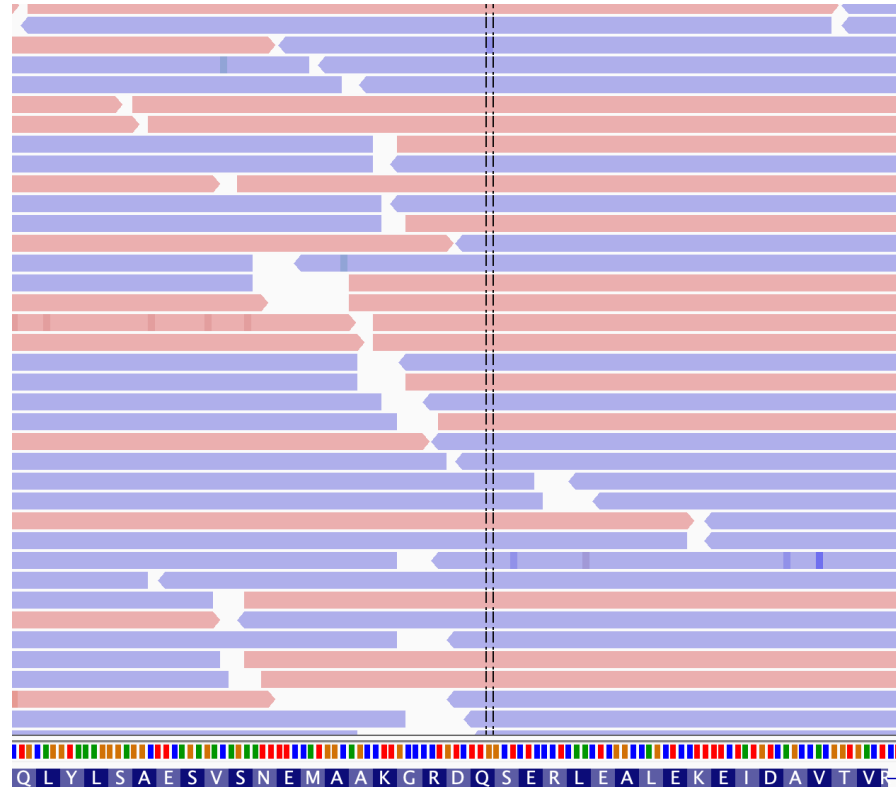
About \$3,500 for a
30x human genome
on a PromethION

What does the data look like?



Long-read ONT

~5% base error rate

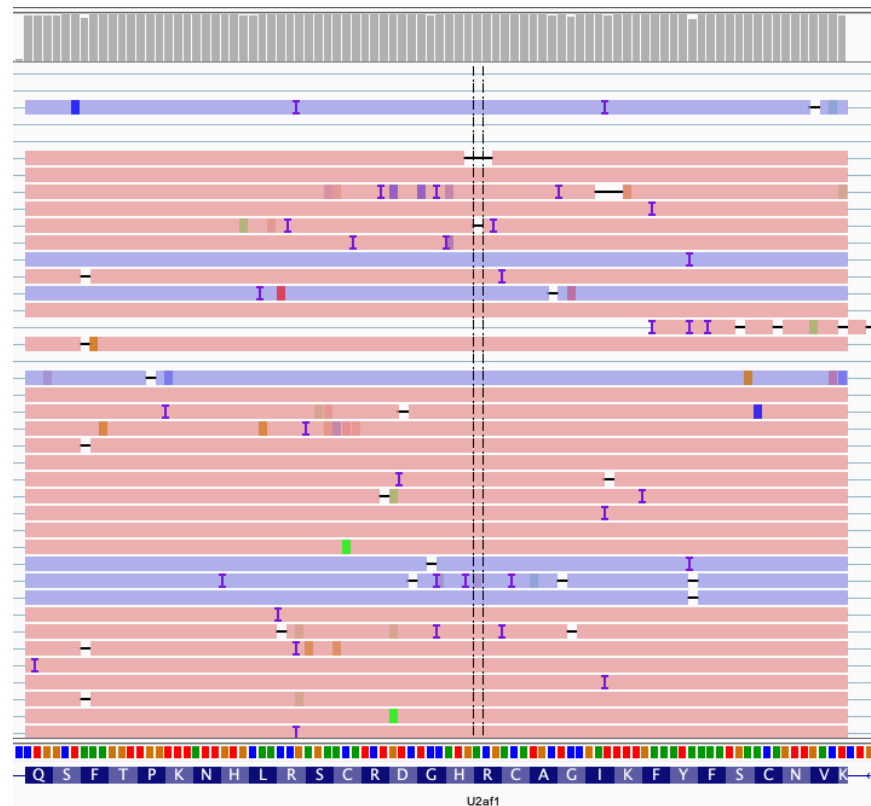


Short-read Illumina

~0.3% base error rate

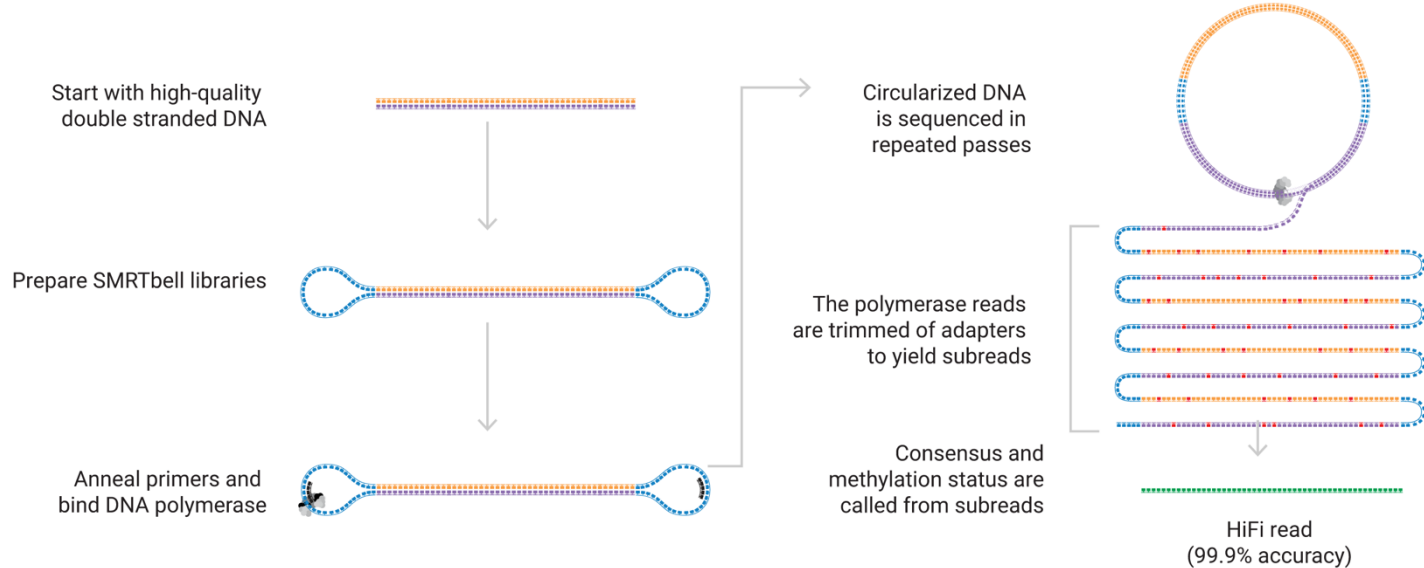
Error rates are contentious and confusing

- How do you calculate error?
Per base?
Per read?
Per variant call?
(after collapsing all of the data?)



PacBio HiFi Sequencing

How are HiFi reads generated?



Improved error rates

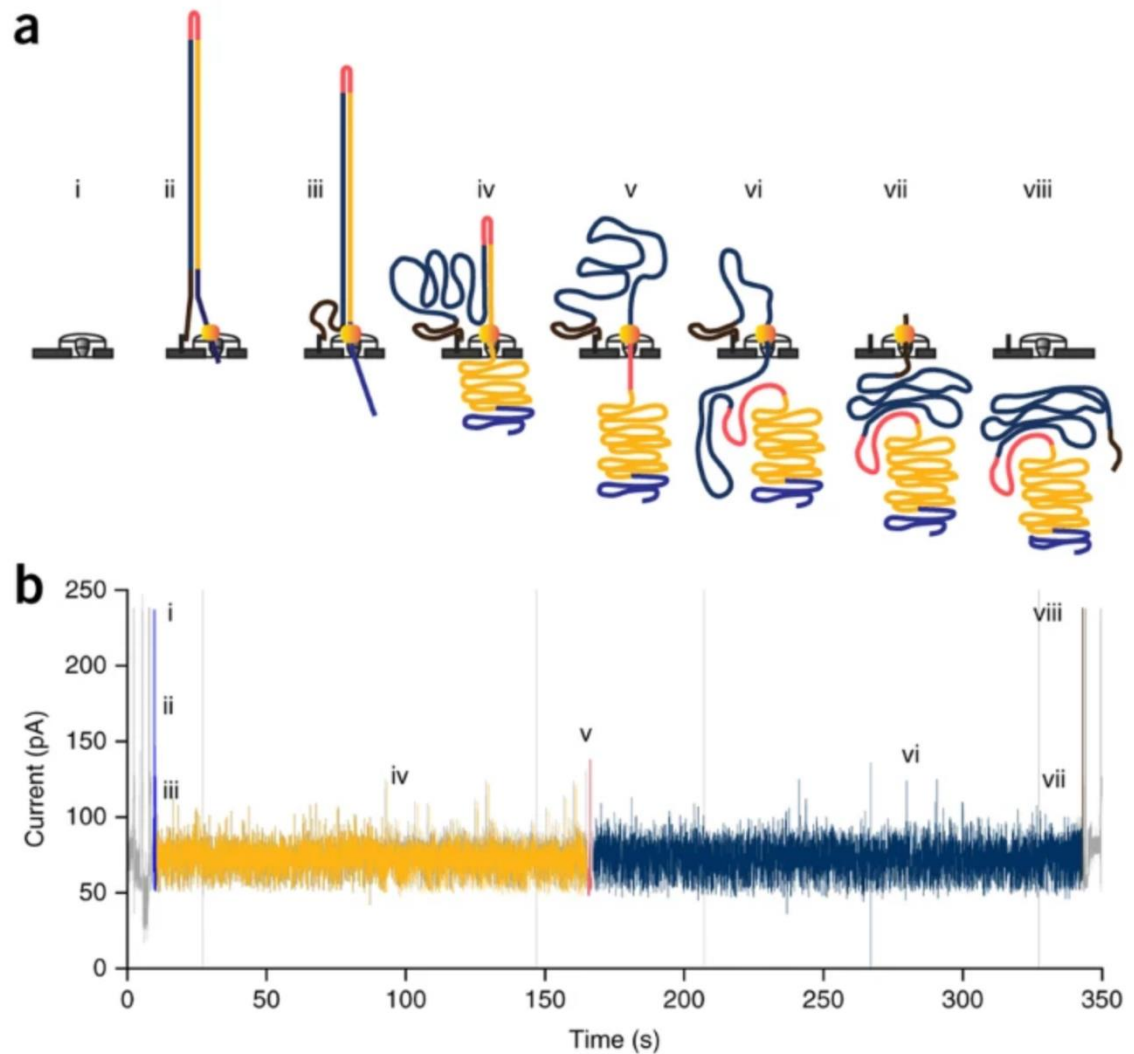
higher cost/lower throughput

ONT Duplex sequencing

Improved error rates

higher cost/lower throughput

<https://www.nature.com/articles/nmeth.3290>



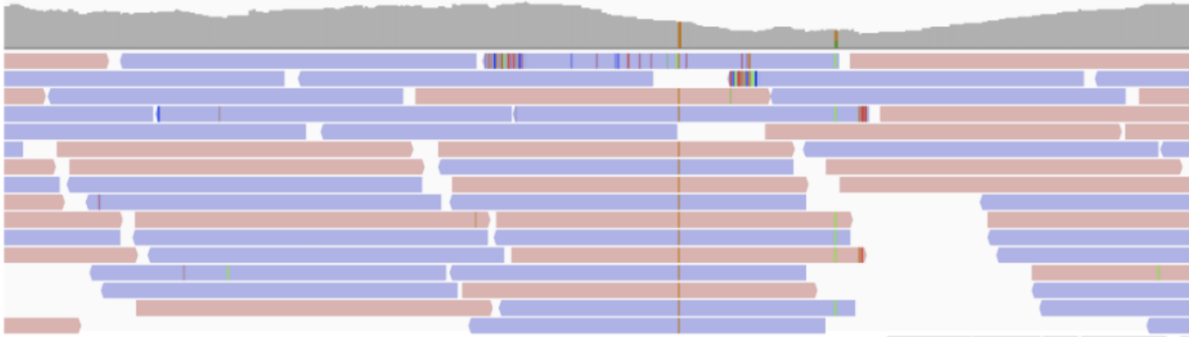
Genomic DNA advantages



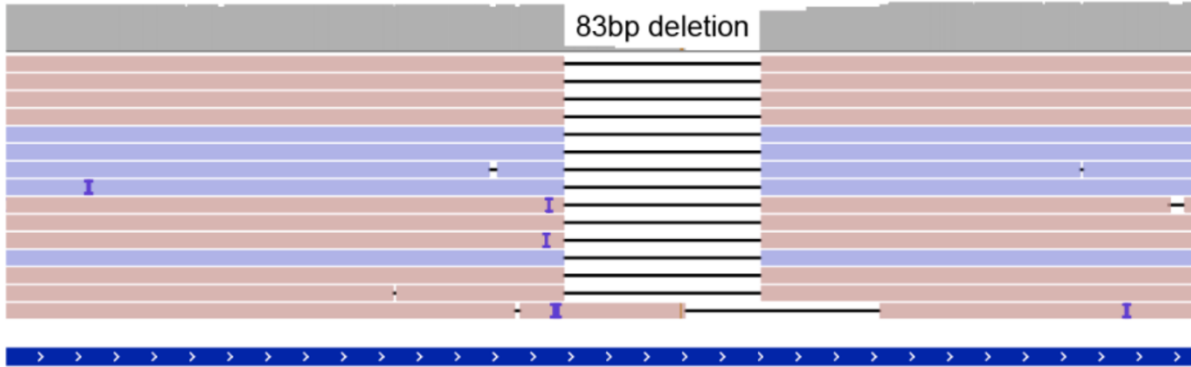
Figure 1: Blue-labeled genomic regions are accessible to long reads but not short, and have functional annotations (e.g. genes or enhancers)

Large Indel detection

No indel detectable - Short-read sequencing - Illumina



83bp deletion - Long-read sequencing - Oxford Nanopore



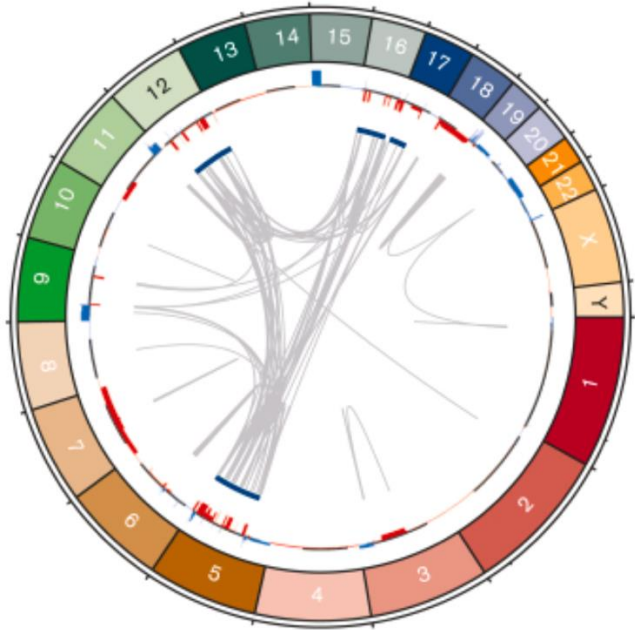
chr3: 31990200-31990700

ZNF860

(protein-coding sequence)

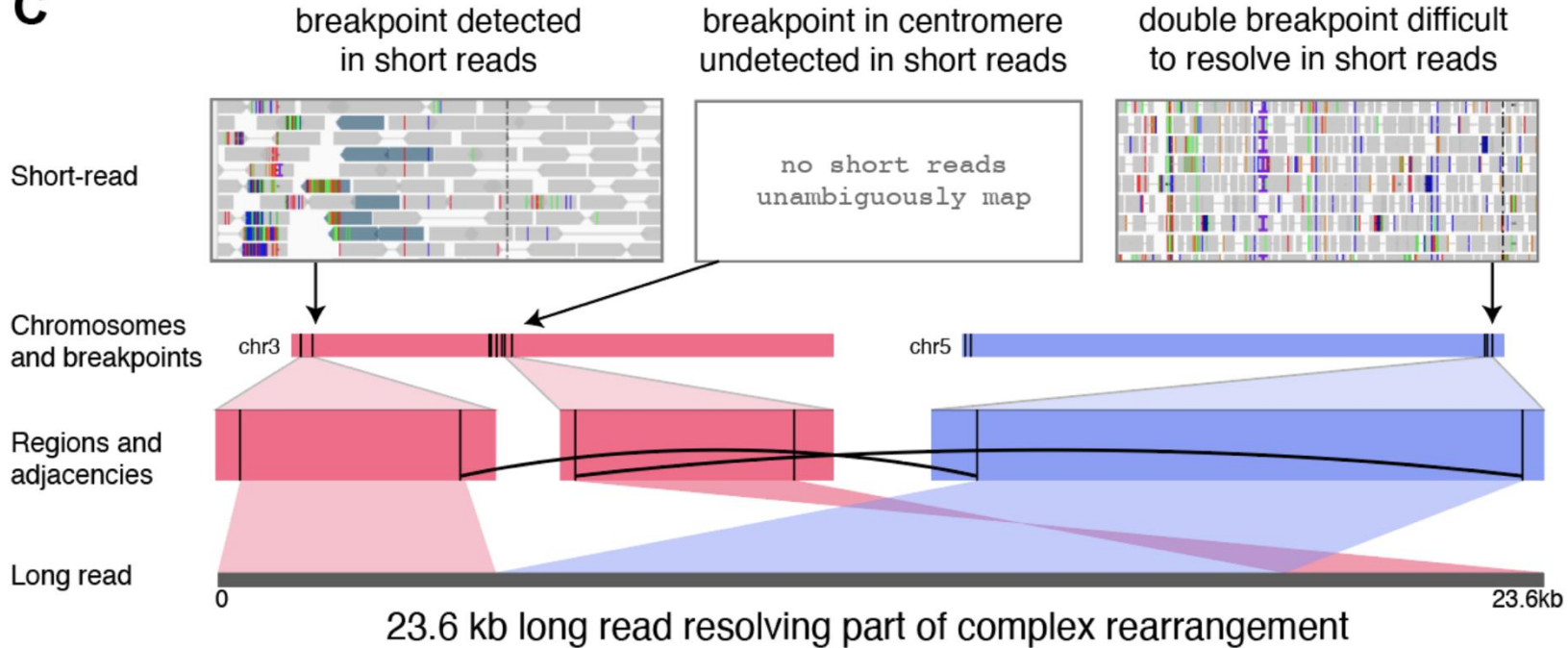
Structural variant resolution

TP53-mutated AML

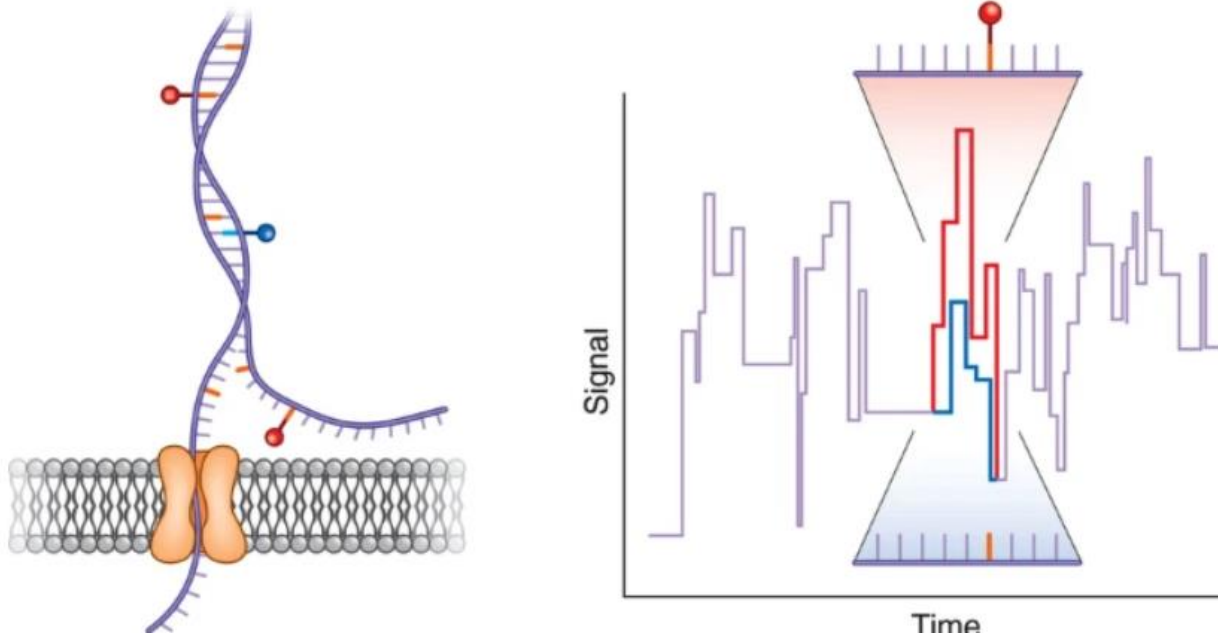


Structural variant resolution

C



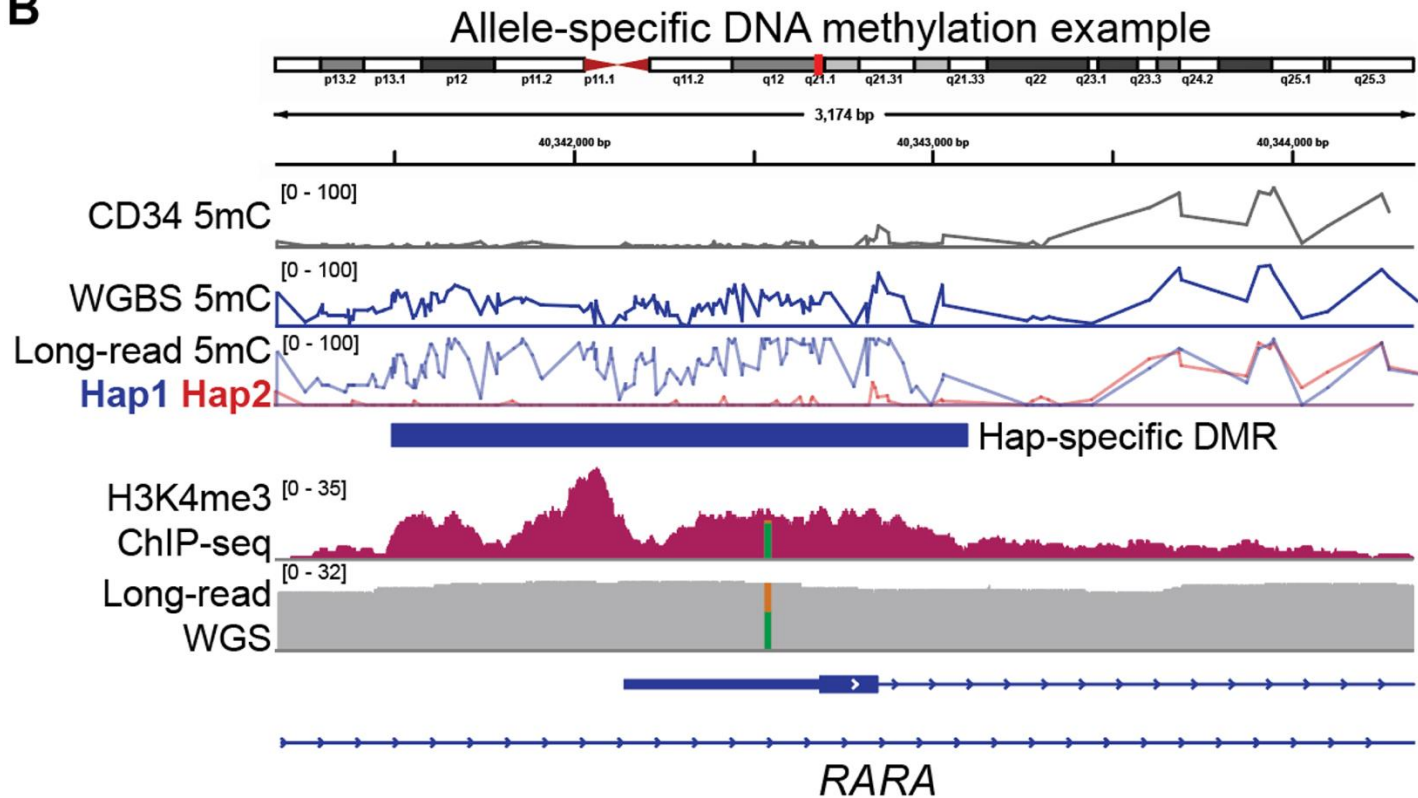
Base modification detection



Can be used for 5mC as well as m6A in direct RNAseq

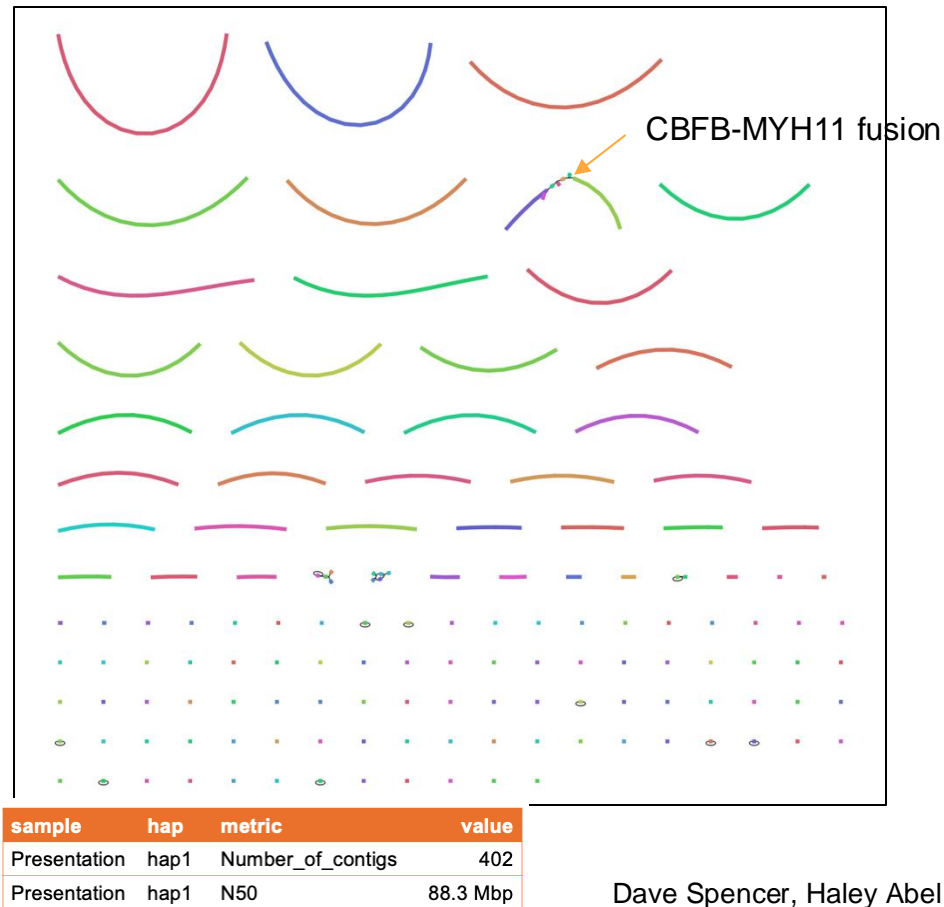
Phasing of reads/modifications

B



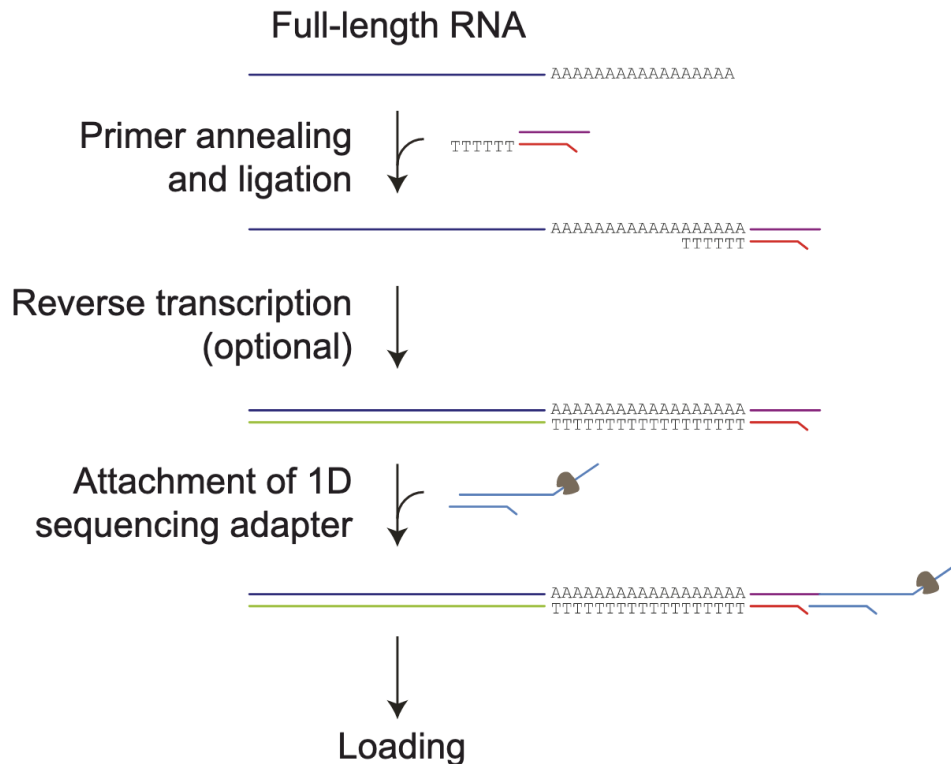
Genome assembly

- Assembly of personal genomes



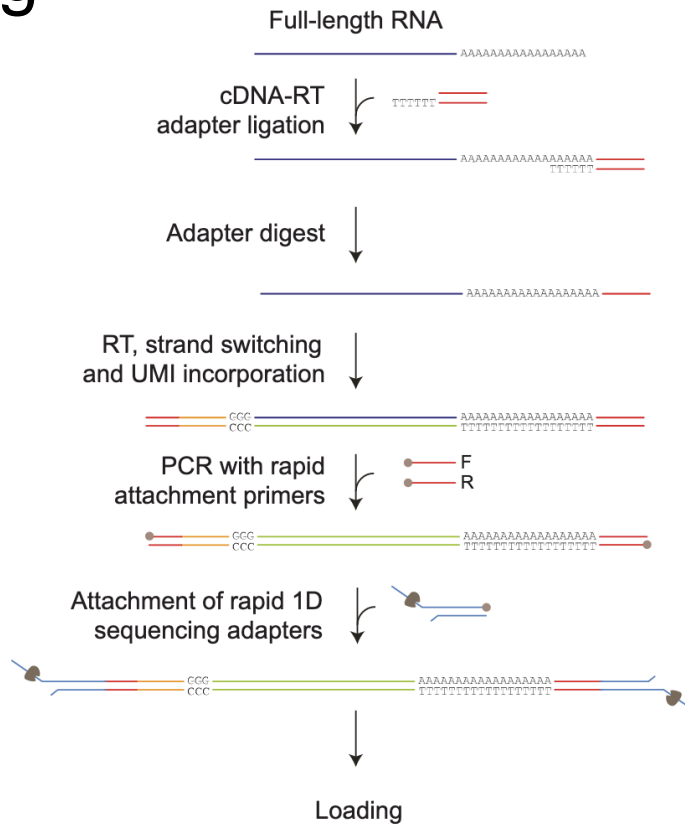
Long-read RNA sequencing

- Direct RNA
- No amplification, less bias
- Preserves base modifications (m6a, etc)

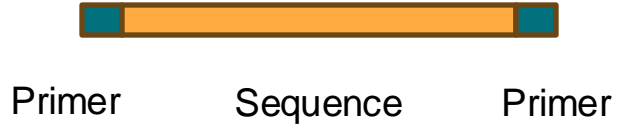


Long-read RNA sequencing

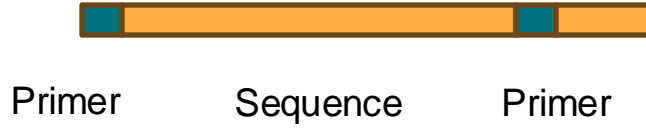
- cDNA sequencing
- much higher yields



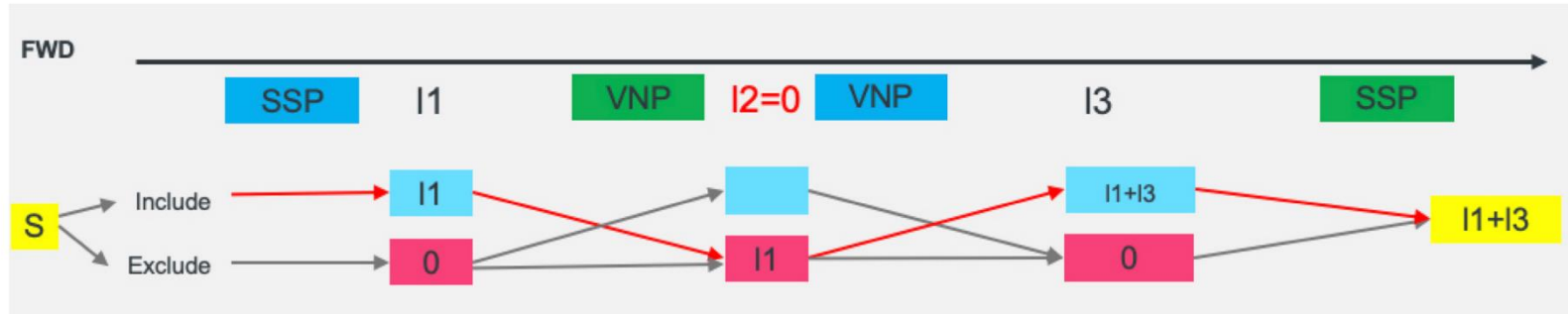
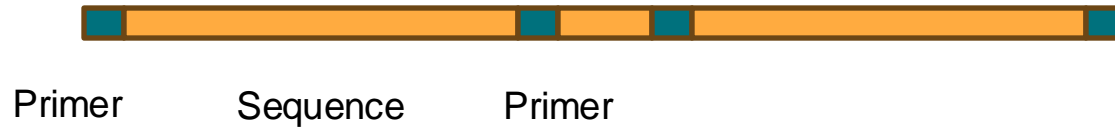
Pychopper



Pychopper

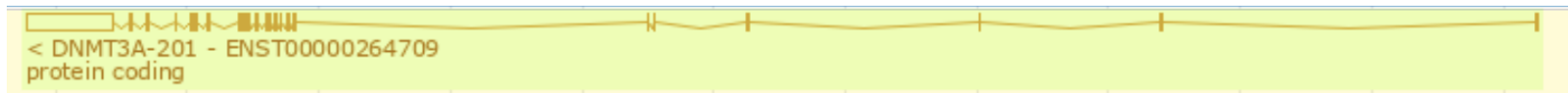


Pychopper

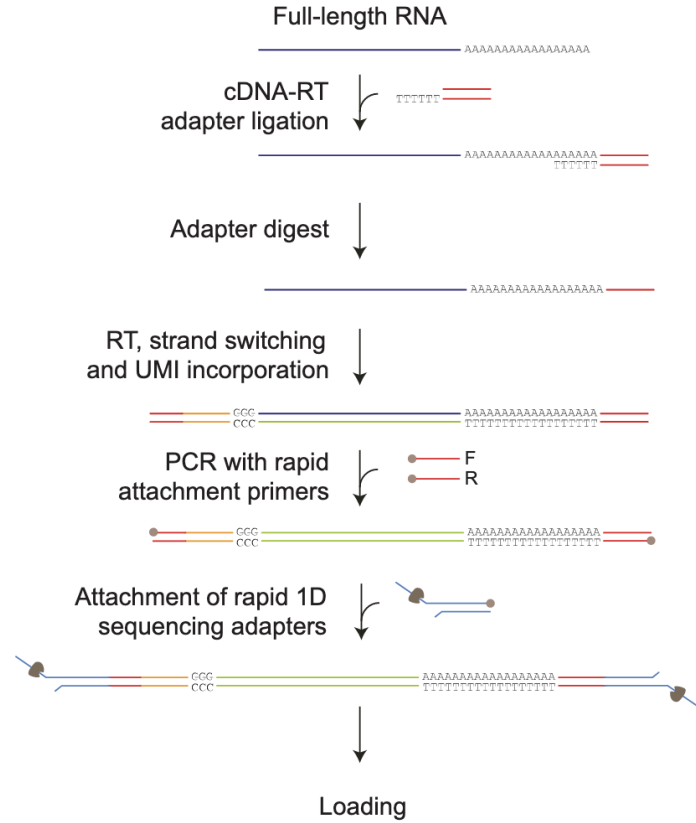


How to estimate duplication rates

- In short read data, reads at the same position are assumed to be duplicates
- How do we know if we're saturating our libraries?



UMIs



UMIs

- UMI at the 3' end of the read

TTT **EGGC**TT **GGAA**TT **GGCC**TT **GGCA**TTT

UMIs

- UMI at the 3' end of the read

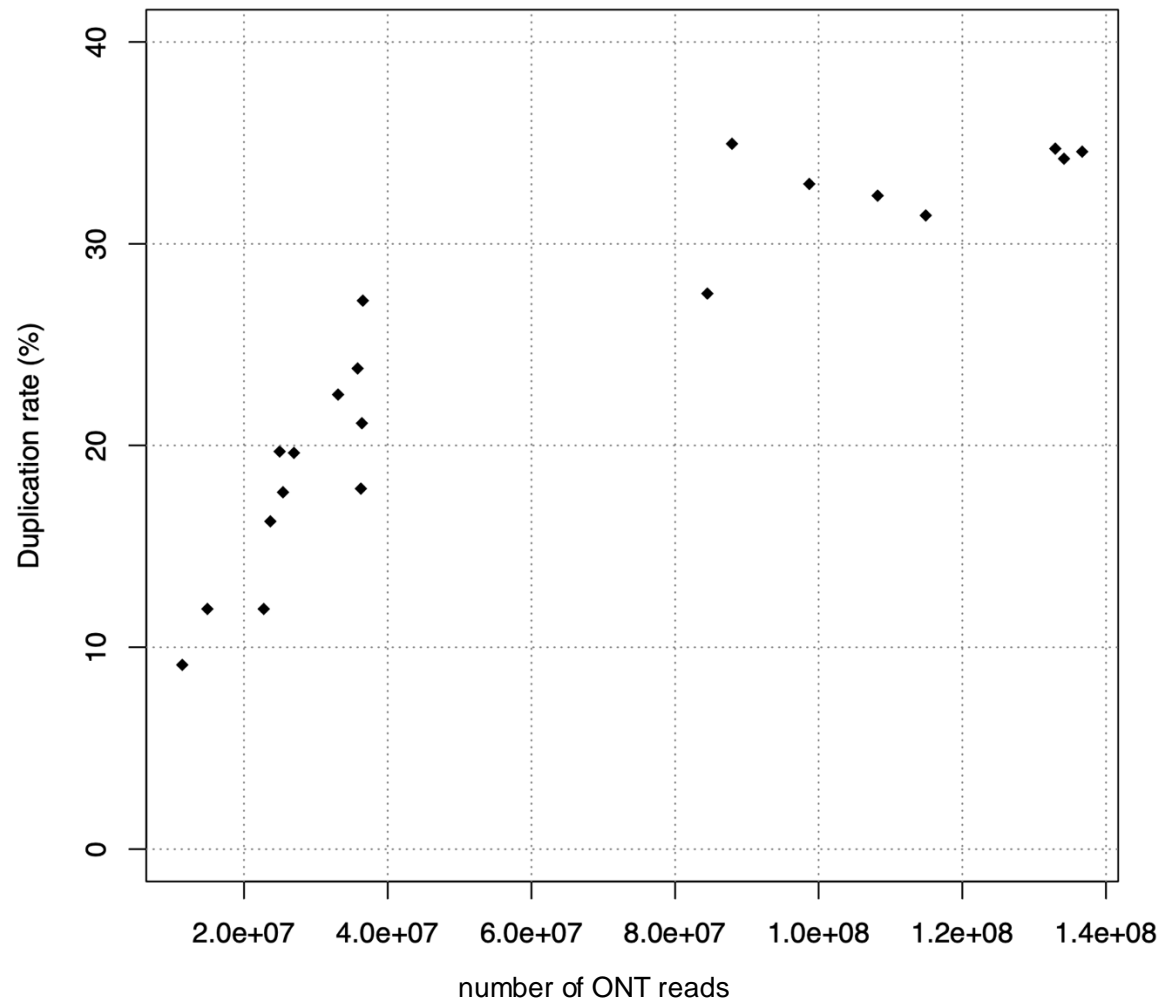
```
30 TTTCACCCTCCACTTCCCGTCTCAGAATT
29 TTTGAAACAGCTTCACCTTGAACTTT
29 TTTCCAATAAAAAAATTACAATTT
29 TTTCAGCAAAATAAAATTCCGGTTT
27 TTTGGAGTTGGGGTTGCGCTTGGGGTTT
27 TTTGAGGTTGGAGTTGGGGTTGGCGTTT
24 TTTGGAGTTGGCGTTGCGGTTGGGGTTT
23 TTTGGGTTGGAAATTGGCGTTGGCAATTT
23 TTTGGGATTAAGATTGGCATTGCGGTTT
23 TTTAGGGTTCGCGTTGGGGTTGCAGTTT
23 TTTAGGGTTAGCGTTGGAGTTGGGGTTT
22 TTTGGCGTTGGGGTTGGCGTTGGCGTTT
22 TTTGGCGTTGGAGTTCAGCTTACGGTTT
22 TTTGCGGTTGGAGTTGGGCTTGGCGTTT
22 TTTACACTTGTGCTCTCCTTAGCCTTT
21 TTTGGGGTTGGAGTTGGCGTTGGCATTT
21 TTTGGCGTTGGCATTGGCGTTGGGGTTT
21 TTTGGCGTTCGGGTTGGAATTCGCGTTT
```

UMIs

- UMI at the 3' end of the read
- Different lengths indicative of high error rate
- only 47% of reads have fully intact UMI
- 7% have no UMI at all
- Even using some error correction with Levenshtein distance, it's ugly

```
30 TTTCACCCTCCACTTCCCGTCTCAGAATT
29 TTTGAAACAGCTTCACCTTGAACTTT
29 TTTCCAATAAAAAAATTACAATTT
29 TTTCAGCAAAATAAAATTCCGGTTT
27 TTTGGAGTTGGGGTTGCGCTTGGGGTTT
27 TTTGAGGTTGGAGTTGGGGTTGGCGTTT
24 TTTGGAGTTGGCGTTGCGGTTGGGGTTT
23 TTTGGGTTGGAAATTGGCGTTGGCAATTT
23 TTTGGGATTAAGATTGGCATTGCGGTTT
23 TTTAGGGTTCGCGTTGGGGTTGCAGTTT
23 TTTAGGGTTAGCGTTGGAGTTGGGGTTT
22 TTTGGCGTTGGGGTTGGCGTTGGCGTTT
22 TTTGGCGTTGGAGTTCAGCTTACGGTTT
22 TTTGCGGTTGGAGTTGGGCTTGGCGTTT
22 TTTACACTTGTGCTCTCCTTAGCCTTT
21 TTTGGGGTTGGAGTTGGCGTTGGCATTT
21 TTTGGCGTTGGCATTGGCGTTGGGGTTT
21 TTTGGCGTTCGGGTTGGAATTCGCGTTT
```

$$0.98^{28} = 0.56$$



What does the data look like?

Legend

Basecalled Estimated

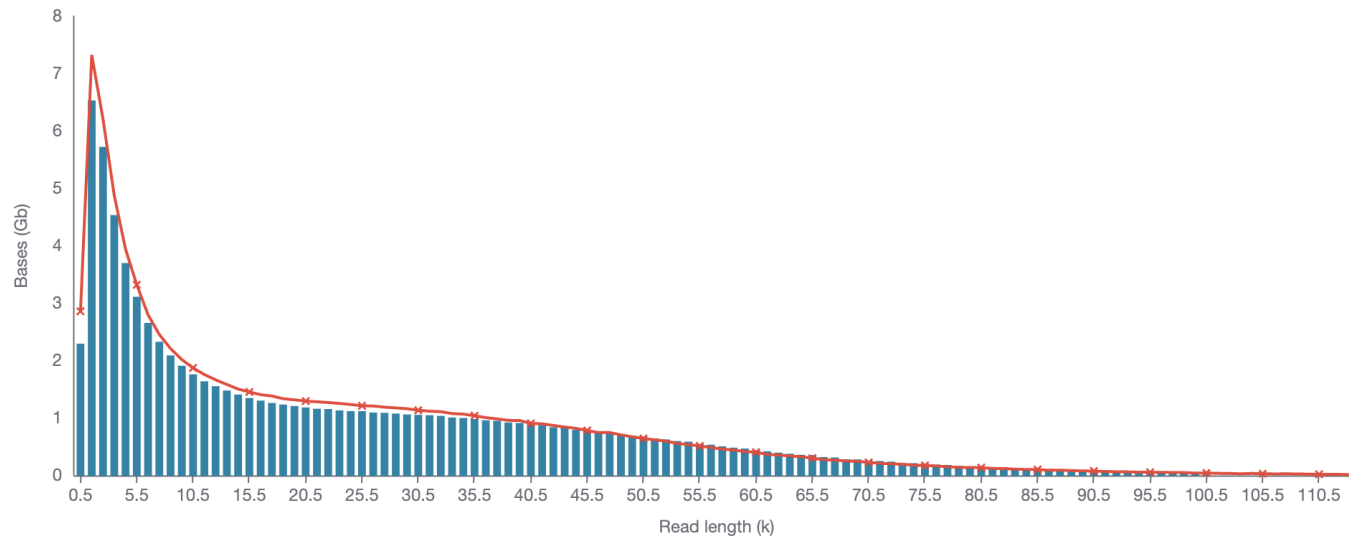
Estimated N50

17.75 kb

% Basecalled

100%

their relative amounts.



Read length (kb)	Aggregated reads (Mb)
------------------	-----------------------

100 - 164	886.98
-----------	--------

164 - 228	36.06
-----------	-------

228 - 292	4.02
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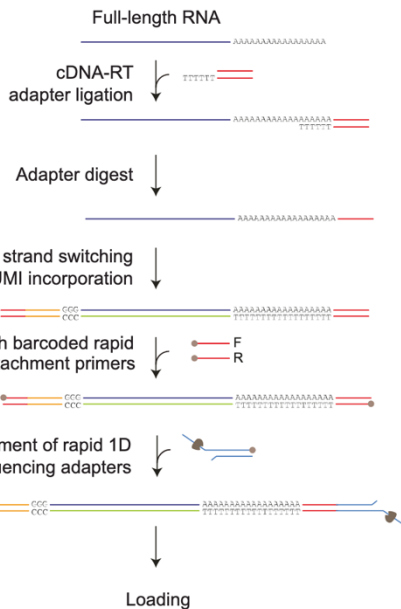
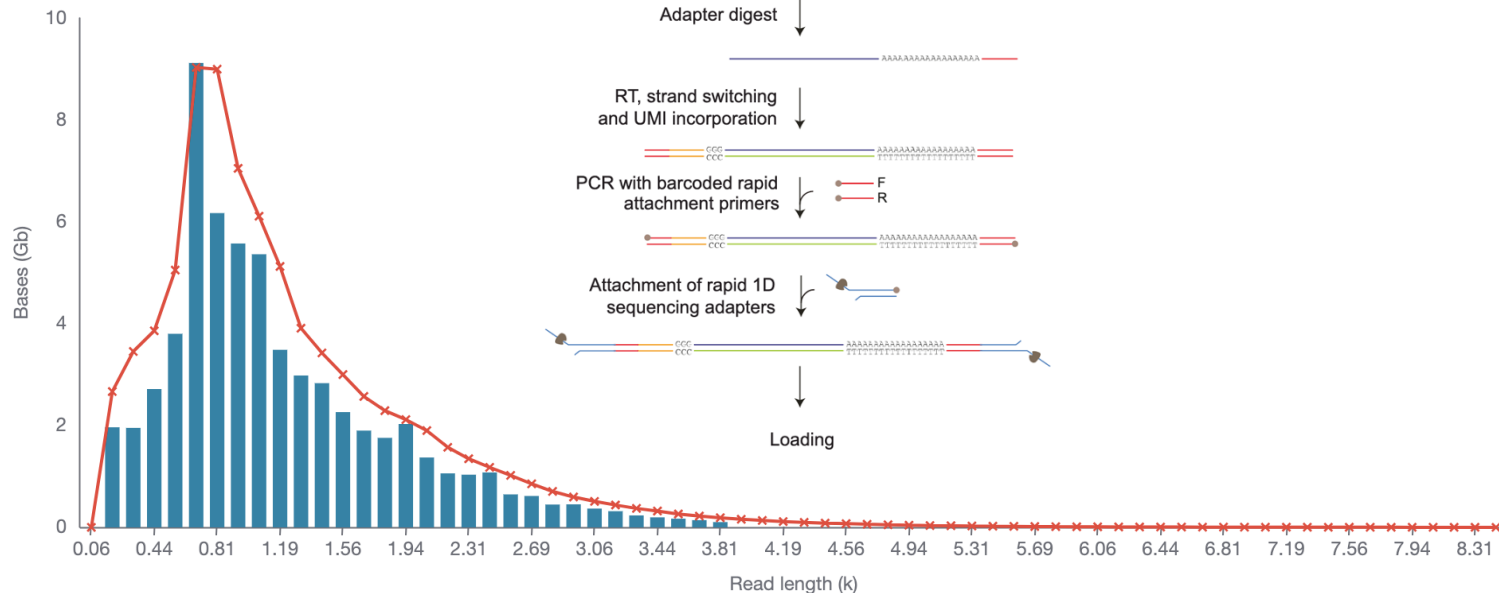
292 - 344	0.35
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Genomic DNA – standard prep

What does the data look like?

Legend

Basecalled Estimated



Estimated N50

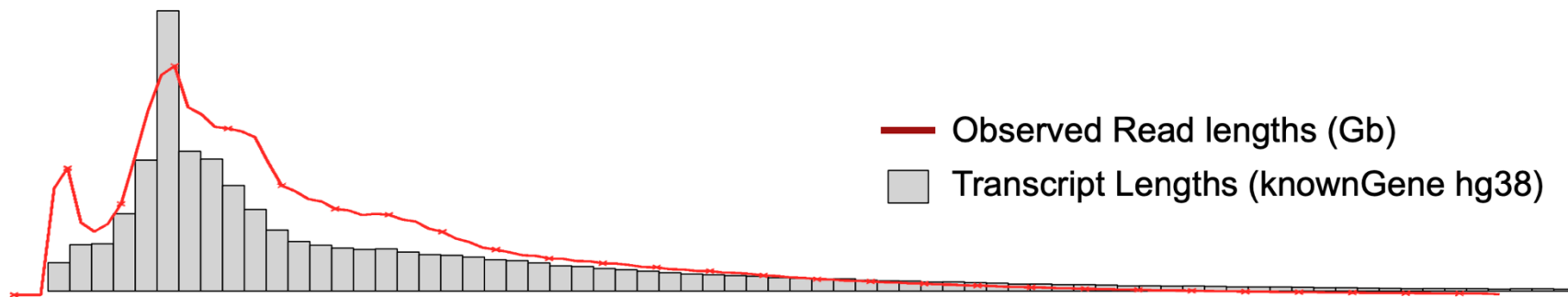
1.02 kb

% Basecalled

100%

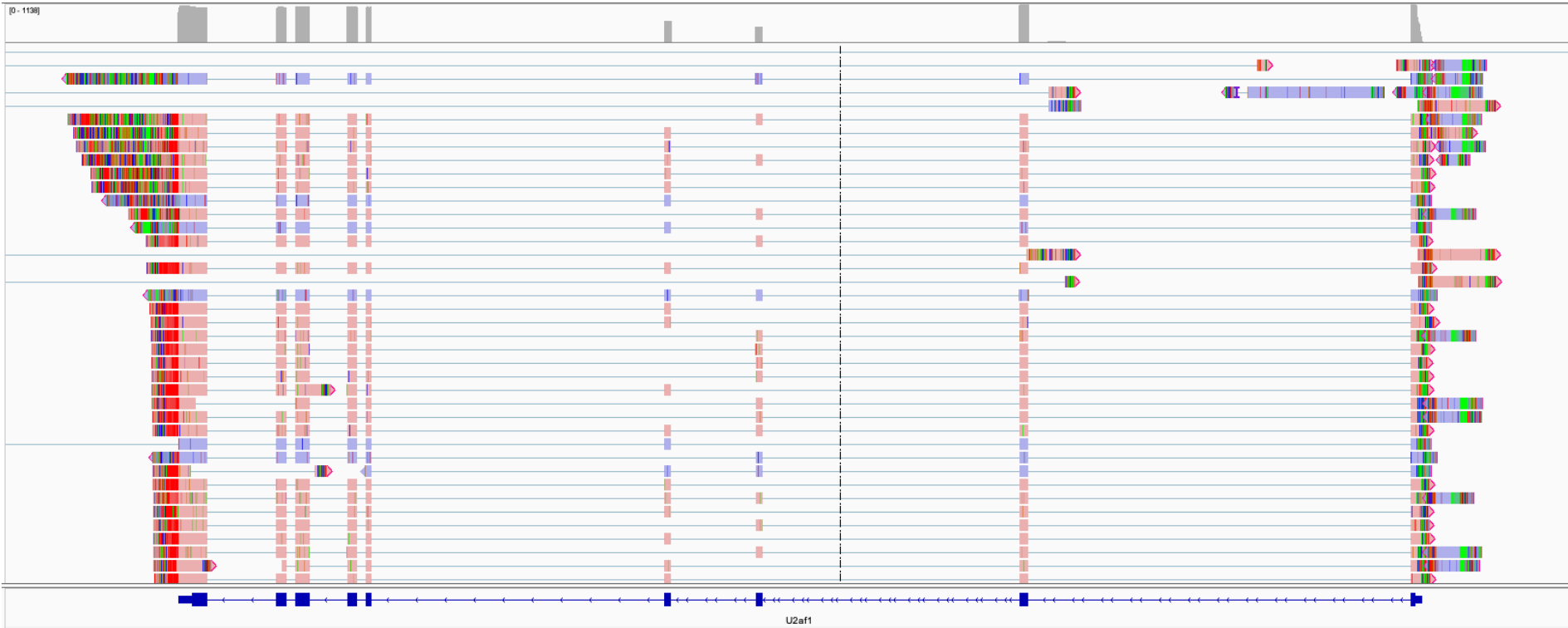
RNA/cDNA – standard prep

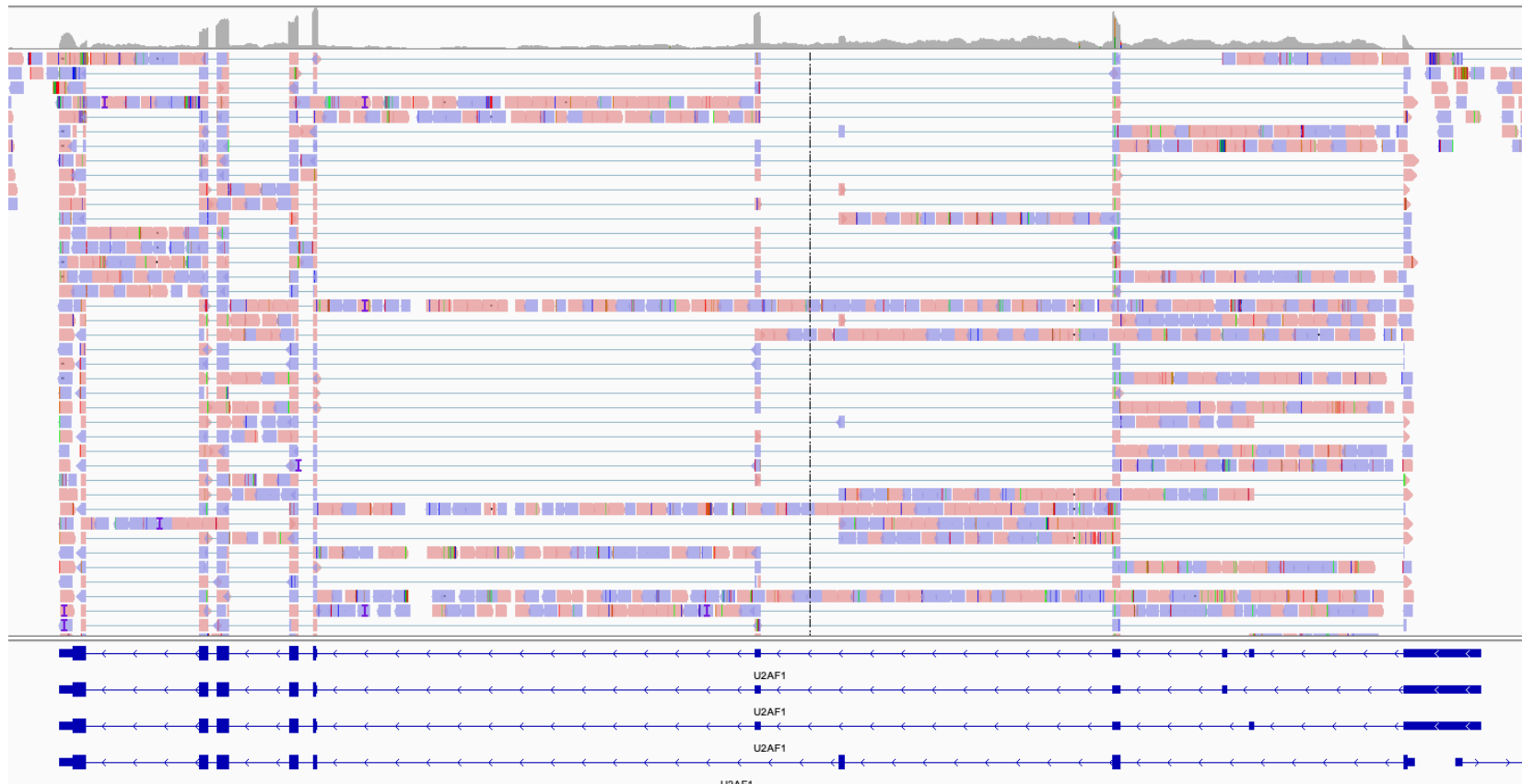
What does the data look like?



cDNA – standard prep

[0 - 1138]





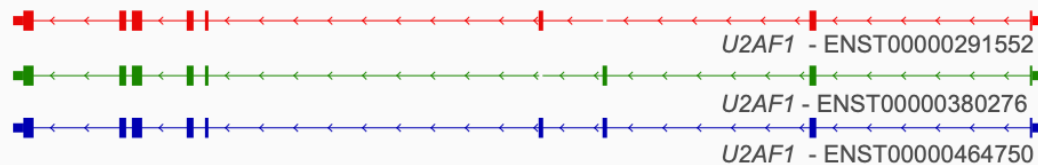
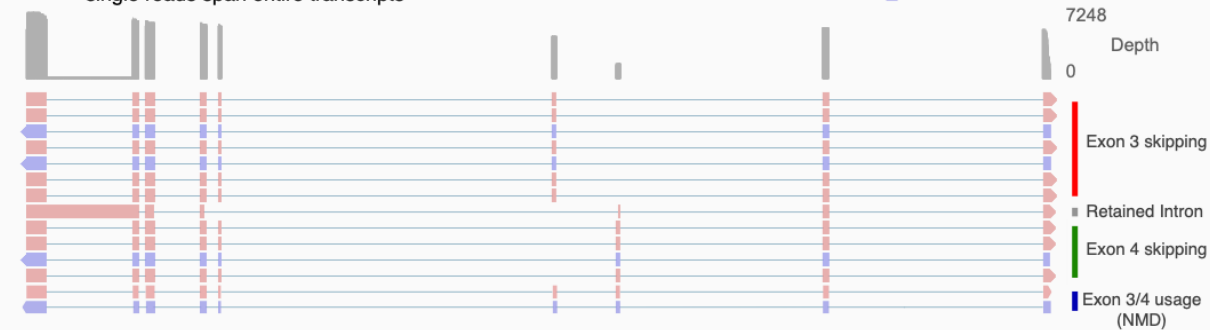
Short-read coverage (65M reads)

single reads span 1-3 exons



Long-read coverage (48M reads)

single reads span entire transcripts



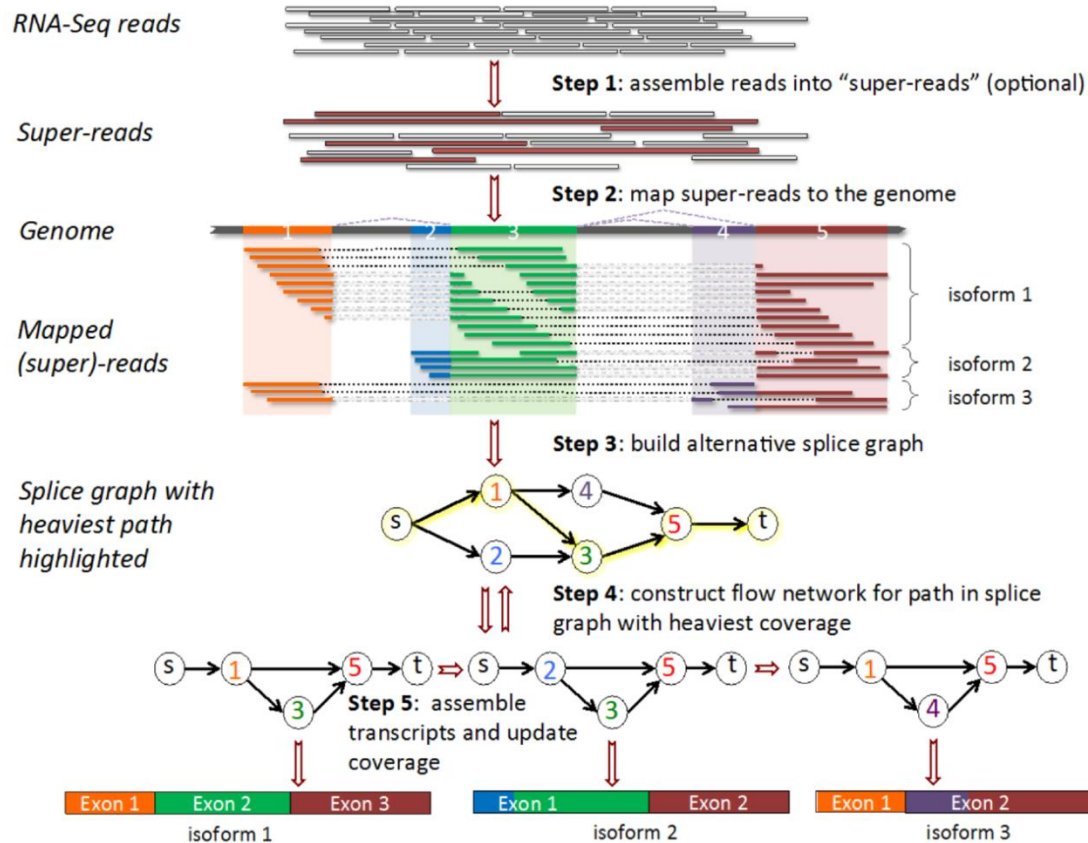
chr21

43,095,000

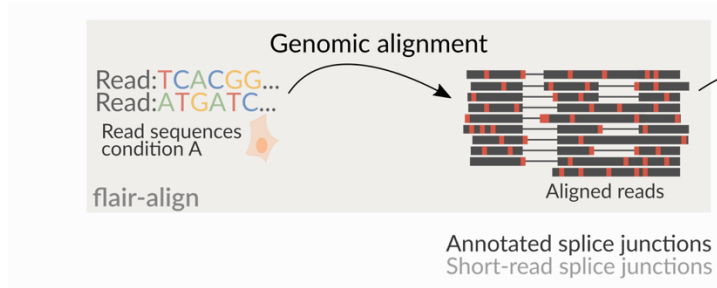
43,100,100

43,105,000

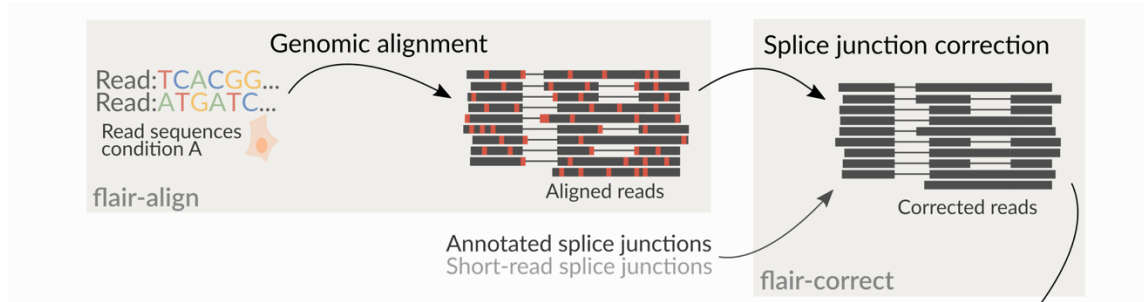
Estimating transcript abundance – short-read, Stringtie



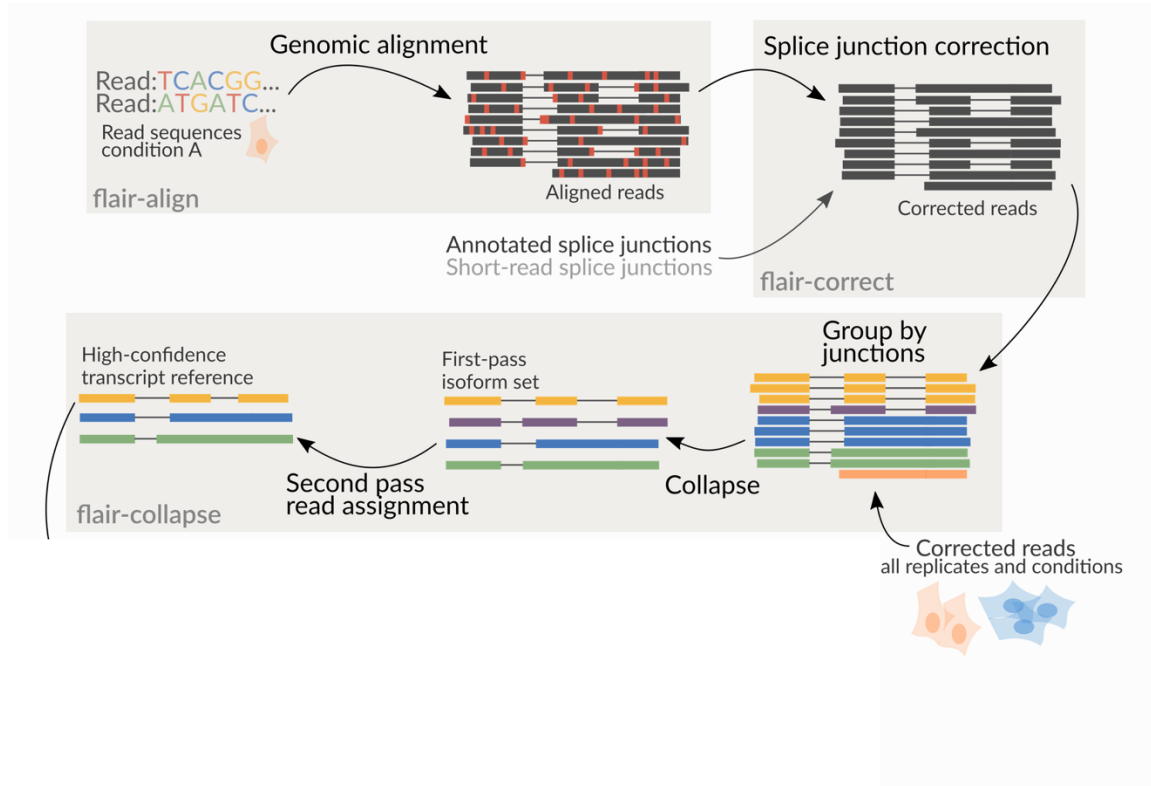
Estimating transcript abundance – long read



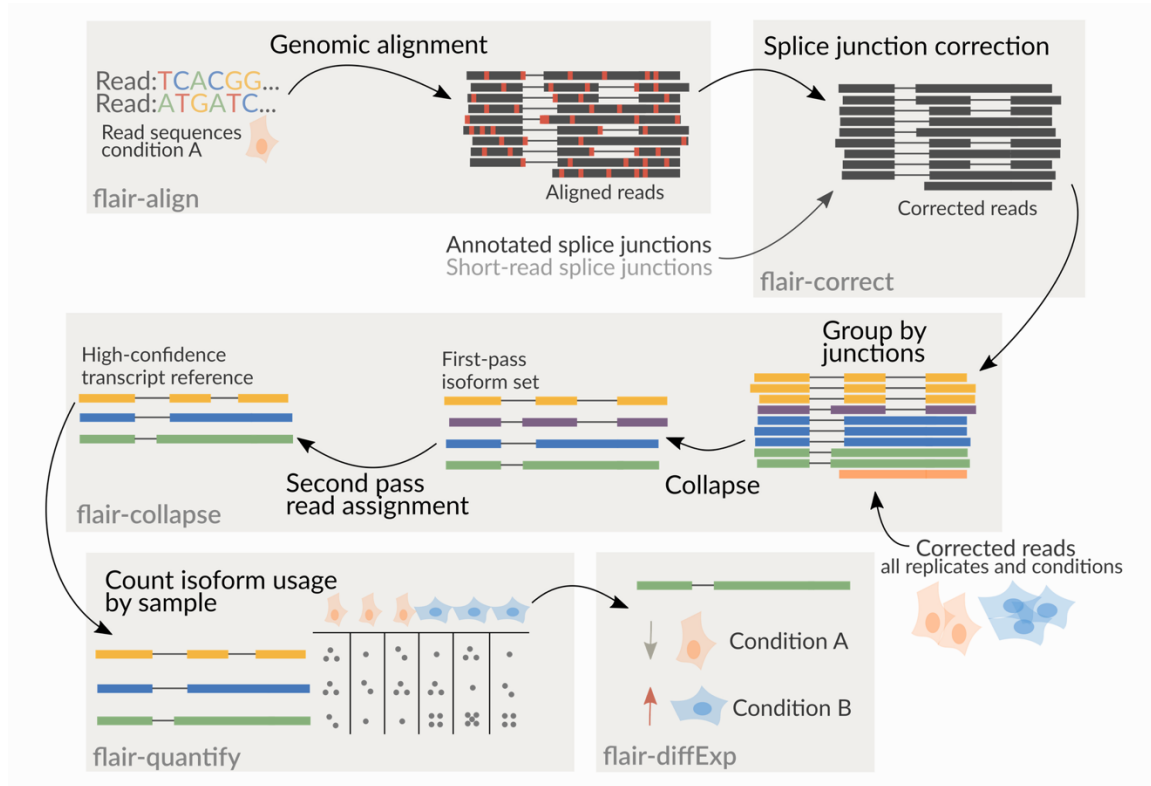
Estimating transcript abundance – long read



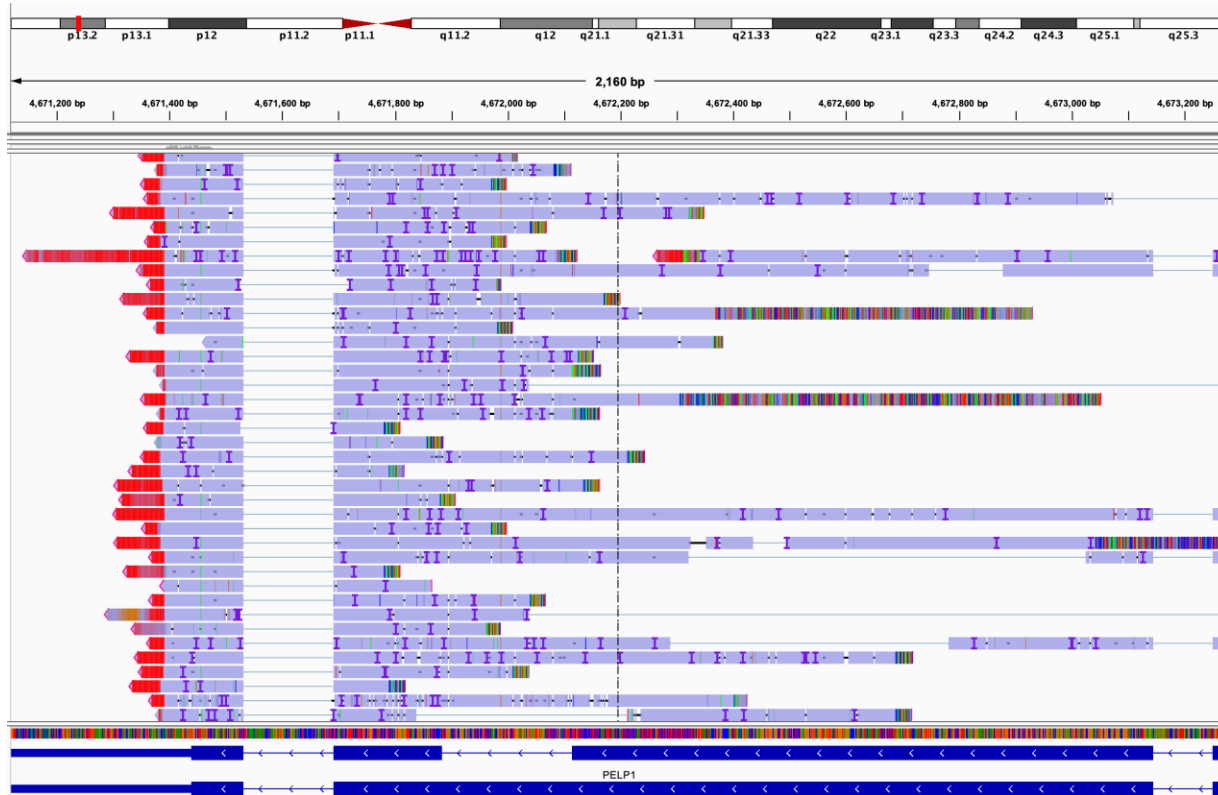
Estimating transcript abundance – long read



Estimating transcript abundance – long read

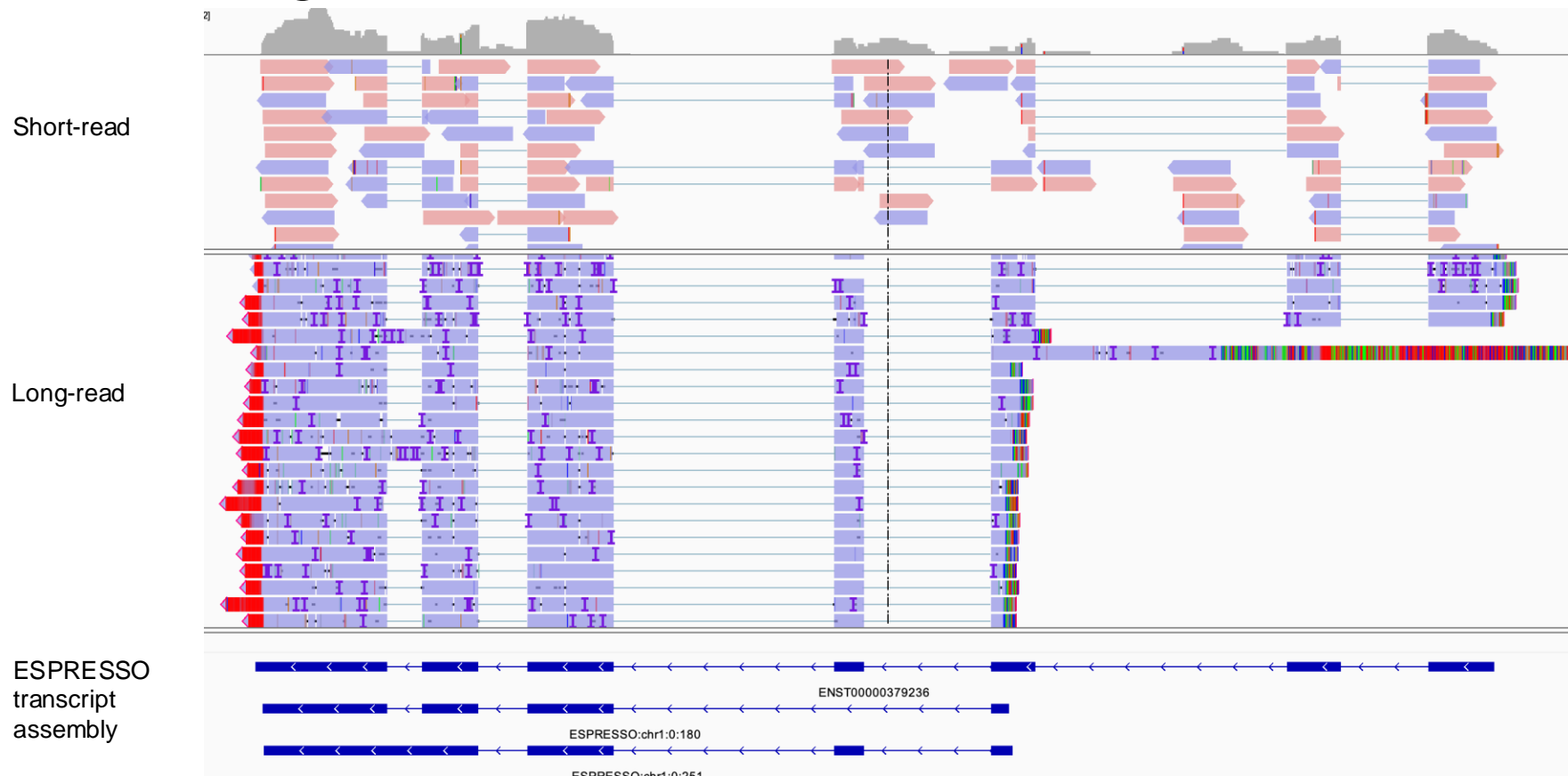


Technical artifacts



Looks like fragmentation of this RNA throughout the long exon. (This is an egregious case – most are more mild)

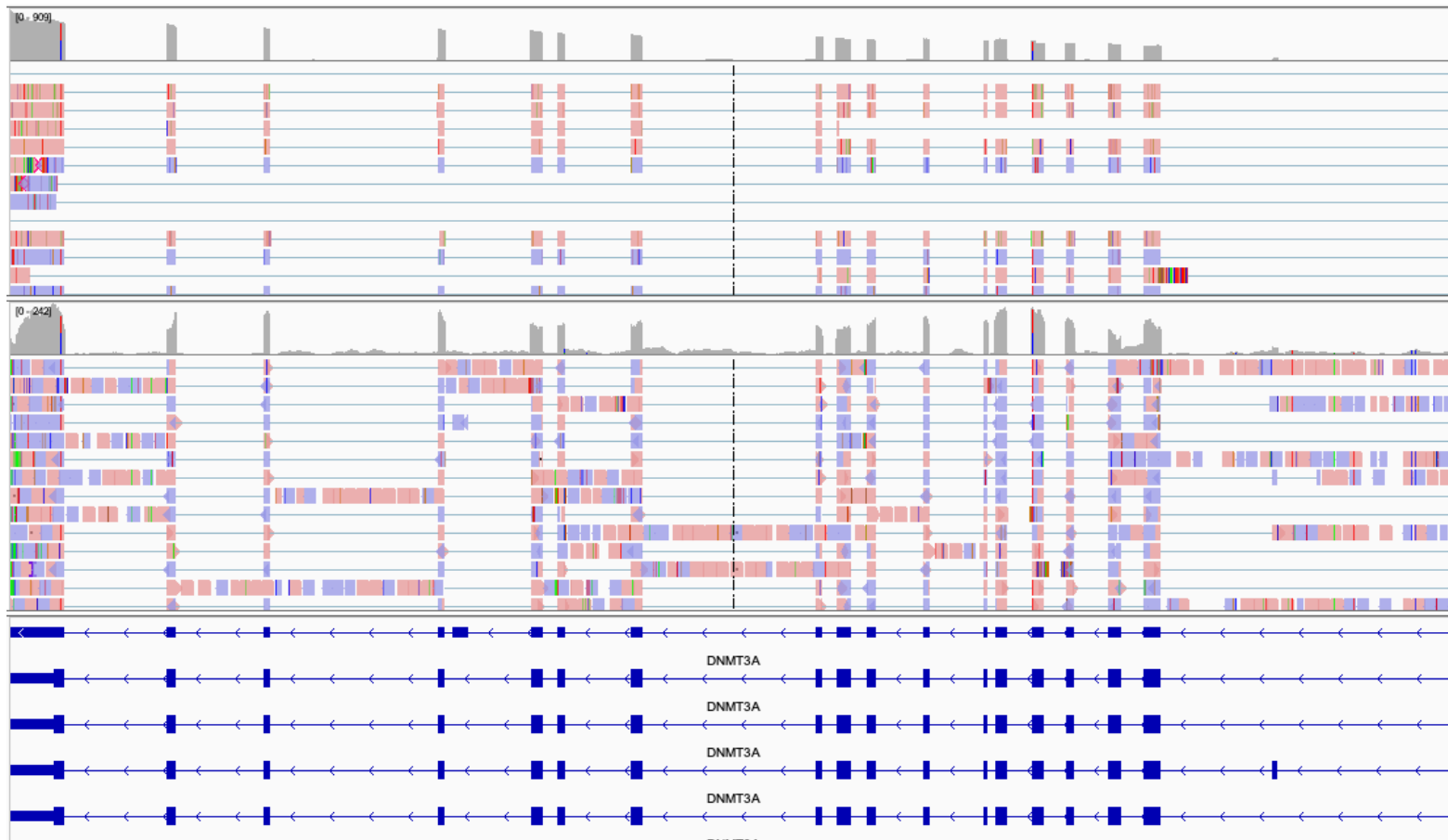
"Full-length" reads



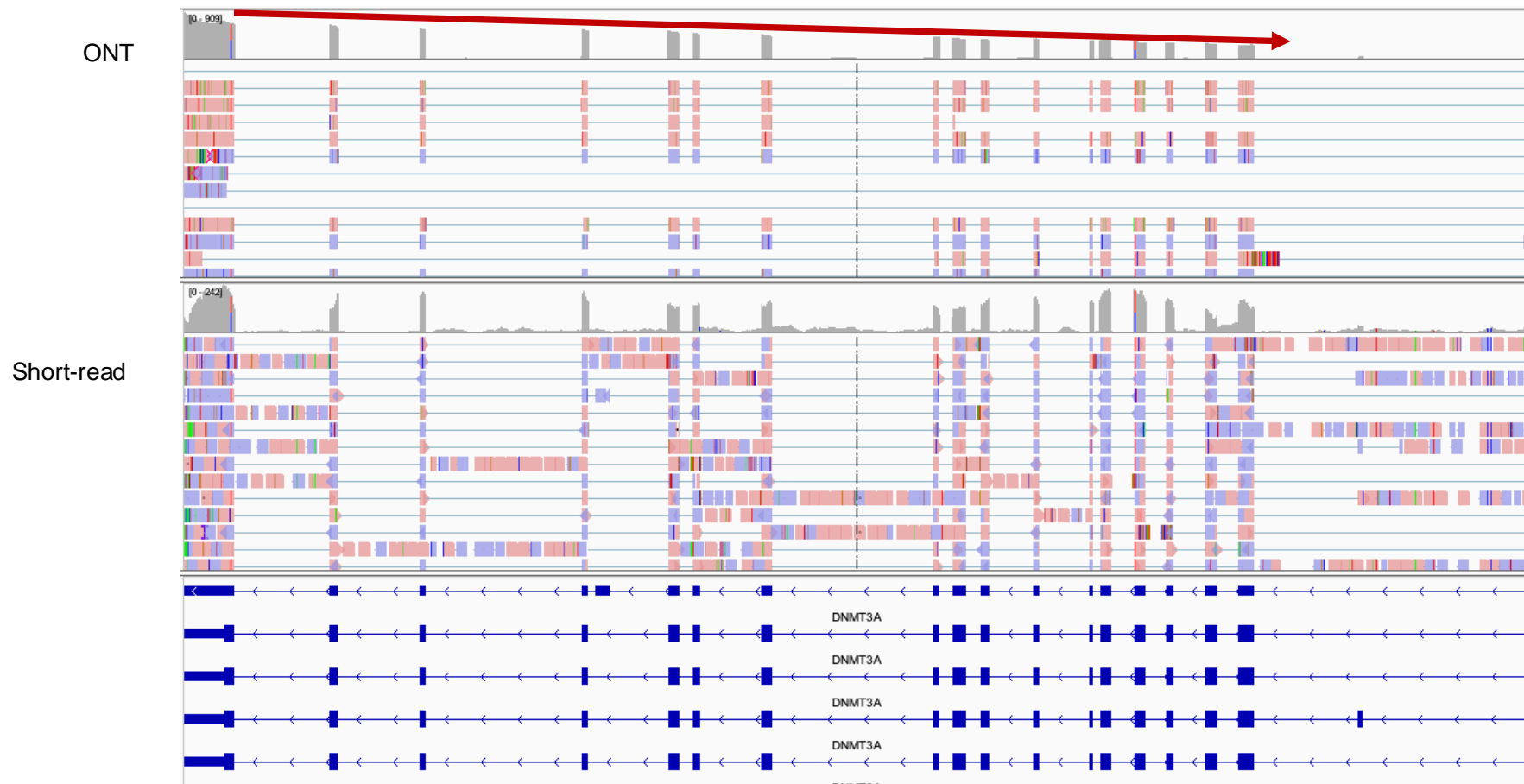
human DNMT3A

ONT

Short-read

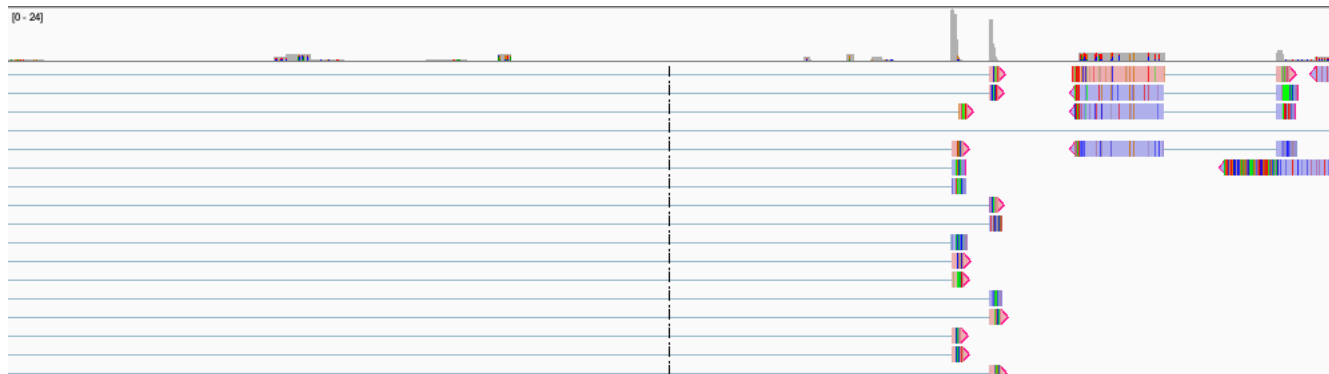


human DNMT3A

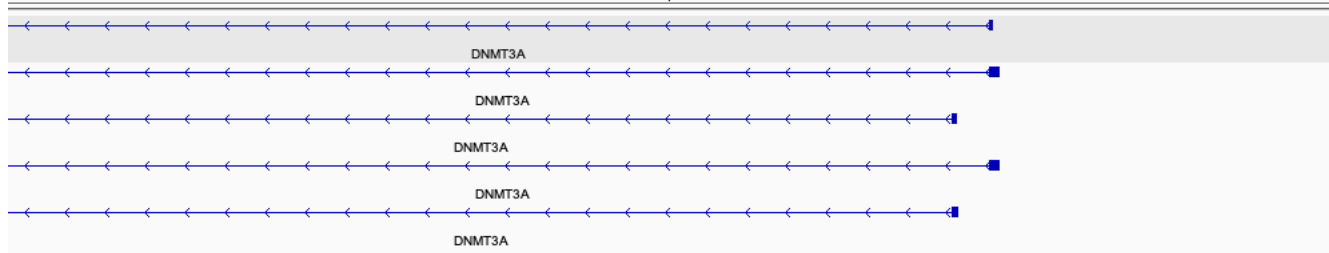
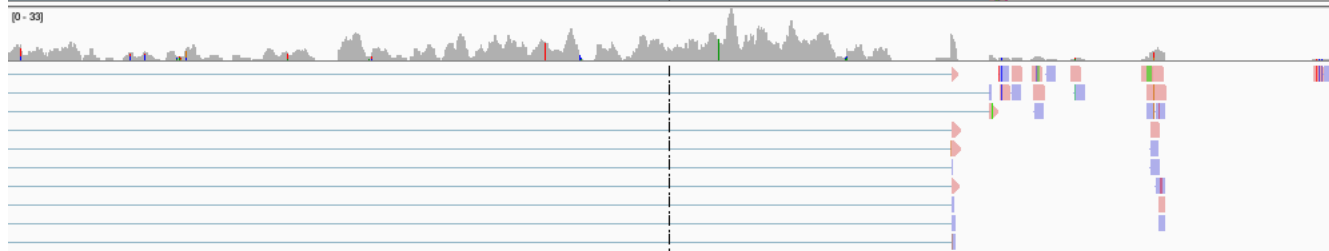


human DNMT3A

ONT



Short-read



Truncated reads

- Appear to be caused by RNA fragmentation
- assessing RIN values of your samples can help – choose clean ones when possible
- When not possible, iteratively assemble transcripts and remove non-full-length reads

Assignment

- Start with some long-read RNAseq data from a cell line
- QC the data, trim adapters
- Align the reads
- Examine a few genes to see how the data looks