

Session 2.3 – Ensamblado

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Programa Formación Continua, ISCI III

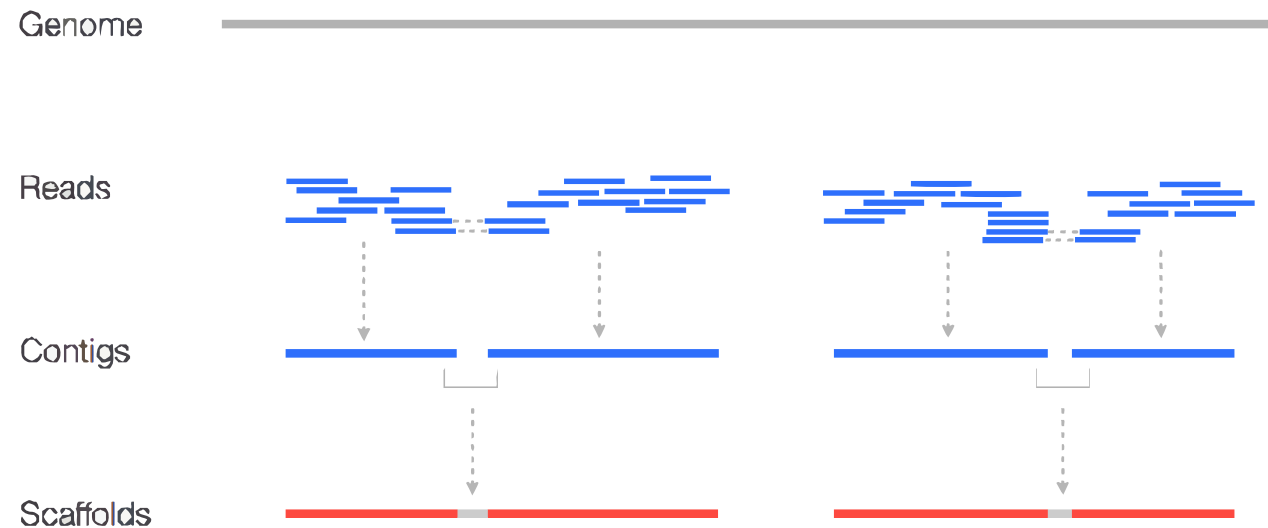
Assembly

Reconstruct the sequence of the original DNA from shorter DNA sequences or small fragments known as reads

- ***De novo***: with no previous knowledge of the genome to be assembled. It overlap the end of the end of each read in order to create a longer sequence.
- ***Assembly with reference***: A similar but not identical genome guides the assembly process. Map reads over supplied genome.

Assembly: contig y scaffold

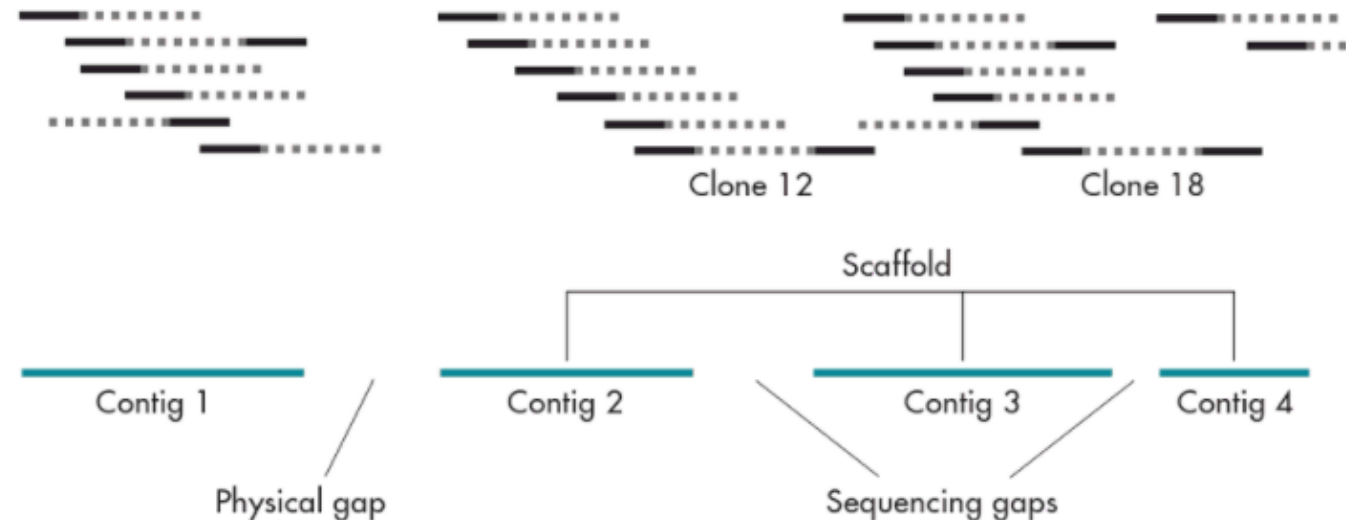
- **Contig:** continuous sequence made up of overlapping shorter sequences
- **Scaffold:** two or more contigs located and rearranged according to spatial information (pair-end, mate pair, reference)



<https://www.biostars.org/p/253222/>

Assembly: gaps

- **Sequencing gaps:** Position and orientation known by spatial information
- **Physical gaps:** No information about adjacent contigs



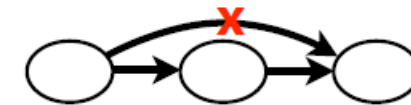
Gene Cloning, Lodge *et al.*

Assembly: Algorithms

- **Overlap, Layout, Consensus (OLC - overlap graph):**
 - O - first overlaps among all the reads are found
 - L - then it carries out a layout of all the reads and overlaps information on a graph
 - Removes redundant and low quality overlaps
 - C - and finally the consensus sequence is inferred

Ex. Newbler, Mira, Celera Assembler, CAP3, PCAP, Phrap, Phusion.

X: CTCGGCCCTAGG
Y: GGCTCTAGGCC



TAGATTACACAGATTACTGA	TTGATGGCGTAA	CTA	Take reads that make up a contig and line them up		
TAGATTACACAGATTACTGACTTGATGGCGTAA	CTA				
TAG	TTACACAGATTATGACTTCATGGCGTAA	CTA			
TAGATTACACAGATTACTGACTTGATGGCGTAA	CTA				
TAGATTACACAGATTACTGACTTGATGGCGTAA	CTA				
↓	↓	↓	↓	↓	
TAGATTACACAGATTACTGACTTGATGGCGTAA CTA					Take consensus, i.e. majority vote

https://pt.slideshare.net/anton_alexandrov/combining-de-bruijn-graph-overlap-graph-and-microassembly/12?smtNoRedir=1

Assembly: Algorithms

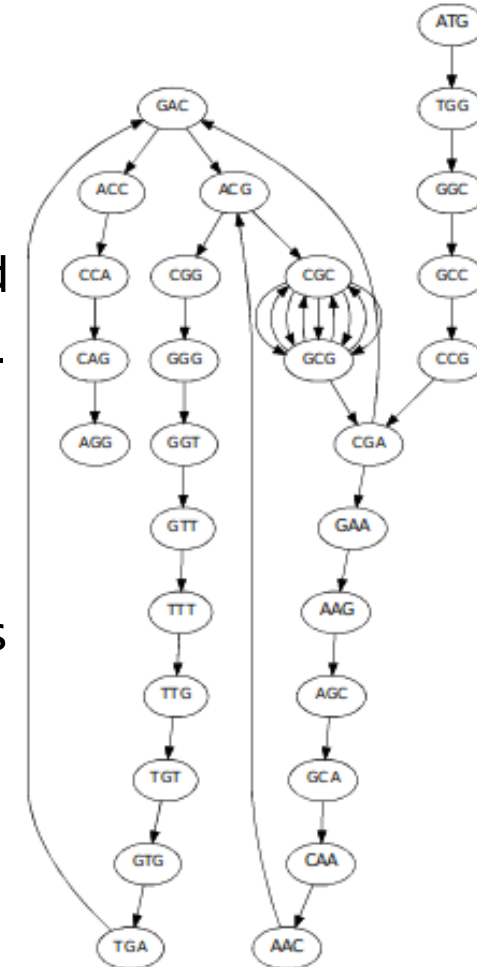
- **De Bruijn Graph (DBG: k-mer graph)**

Chopping reads into much shorter k-mers (fixed length fragments) and then using all the k-mers to form a DBG and infer the contigs.

- Nodes in the graph are k-mers
- Edges represent consecutive k-mers (which overlap by k-n symbols)

Ex. SPAdes, ABySS, Velvet, AllPaths, Soap...

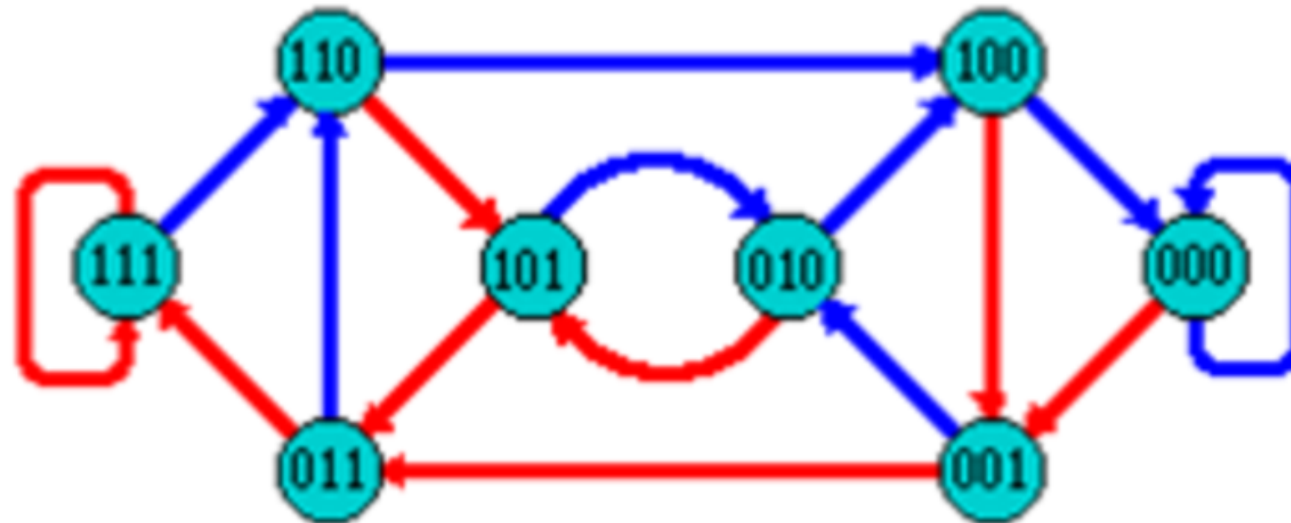
https://medium.com/@han_chen



de Bruijn Graphs

- A directed graph of sequences of symbols
- Nodes in the graph are k-mers
- Edges represent consecutive k-mers (which overlap by k-1 symbols)

Consider the 2 symbol alphabet (0 & 1) de Bruijn Graph for $k = 3$



<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

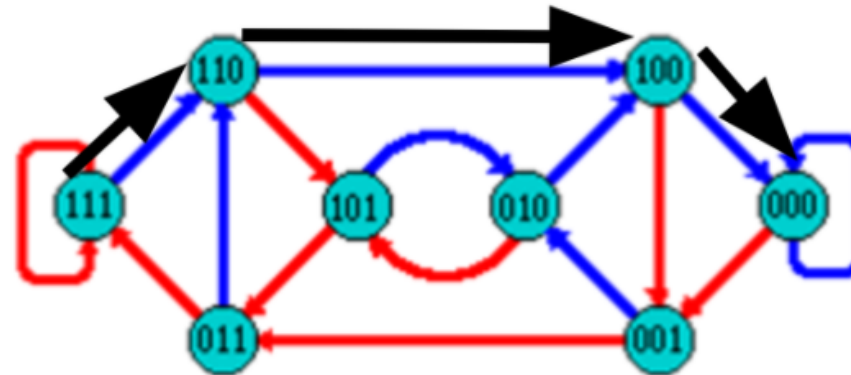
Producing sequences

- Sequences of symbols are produced by moving through the graph

e.g. 111000 = 111 -> 110 -> 100 -> 000

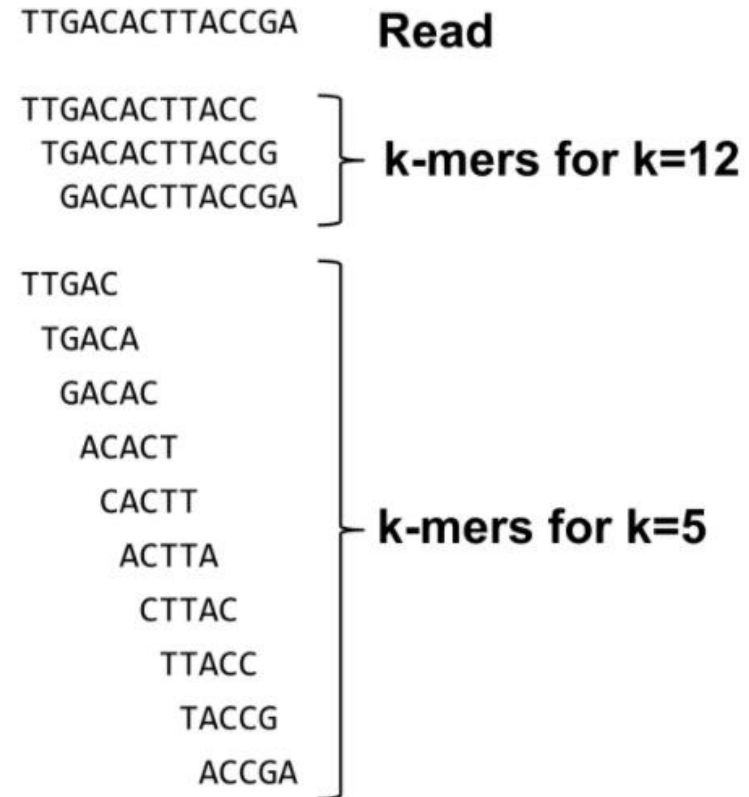
```

1 1 1
  1 1 0
    1 0 0
      0 0 0
-----
1 1 1 0 0 0
  
```



<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

What are K-mers?



<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

K-mers de Bruijn graph



Example #1:

HAPPI PINE INESS APPIN

<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

K-mers de Bruijn graph



Example #1:

HAPPI PINE INESS APPIN

All 4-mers:

HAPP PINE INES **APPI**
APPI NESS PPIN

Unique 4-mers:

HAPP **APPI** PINE PPIN INES NESS

<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

K-mers de Bruijn graph



Example #1:

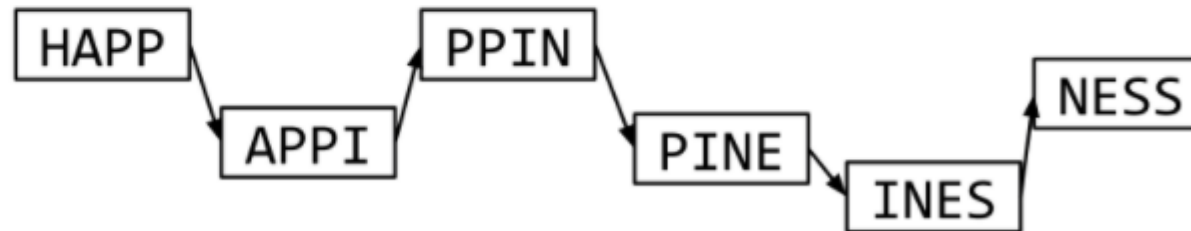
HAPPI PINE INESS APPIN

k = 4 k-mers:

HAPP APPI

PINE PPIN

INES NESS



K-mers de Bruijn graph



Example #1:

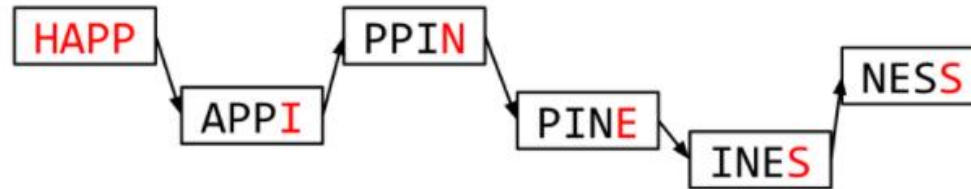
HAPPI PINE INESS APPIN

k = 4 k-mers:

HAPP APPI

PINE PPIN

INES NESS



HAPPINESS

Easy!

<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

The problem of repeats



Example #2:

MISSIS SSISSI SSIPPI

<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

The problem of repeats



Example #2:

MISSIS SSISSI SSIPPI

All 4-mers (9):

MISS	SSIS	SSIP
ISSI	SISS	SIPP
SSIS	ISSI	IPPI

Unique 4-mers (7):

MISS SSIS SSIP ISSI SISS SIPP IPPI

The problem of repeats

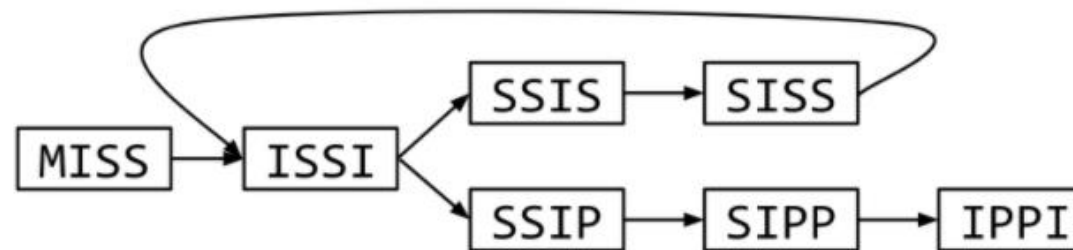


Example #2:

MISSIS SSISSI SSIPPI

All 4-mers:

MISS ISSI SSIS SISS SSIP SIPP IPPI



The problem of repeats

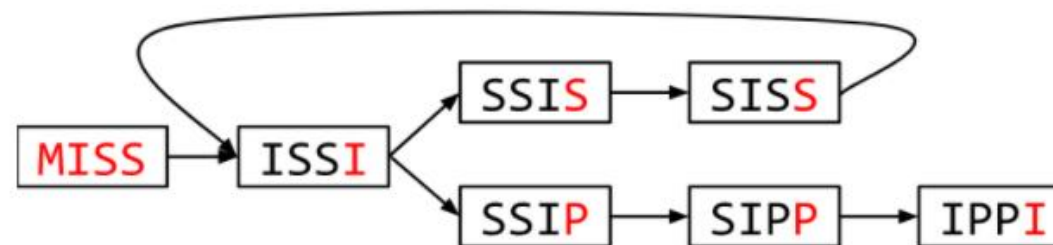


Example #2:

MISSIS SSISSI SSIPPI

All 4-mers:

MISS ISSI SSIS SISS SSIP SIPP IPPI



MISSISSIPPI or MISSISSISSISSIPPI or ...

<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

Different k

Example #2a:

MISSIS SSISSI SSIPPI

Different k

Example #2a:

MISSIS SSISSI SSIPPI

All 5-mers (6):

MISSI SSISS SSIPP
ISSIS SISSI SIPPI

Unique 5-mers (6, no duplicates):

MISSI ISSIS SSISS SISSI SSIPP SIPPI

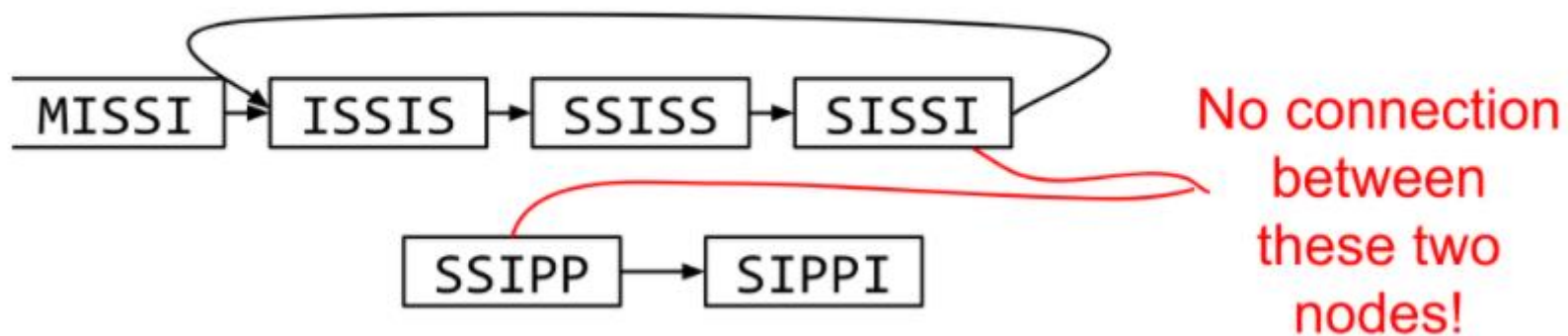
Different k

Example #2a:

MISSIS SSISSI SSIPPI

This time $k = 5$ k-mers:

MISSI ISSIS SSISS SISSI SSIPP SIPPI



<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

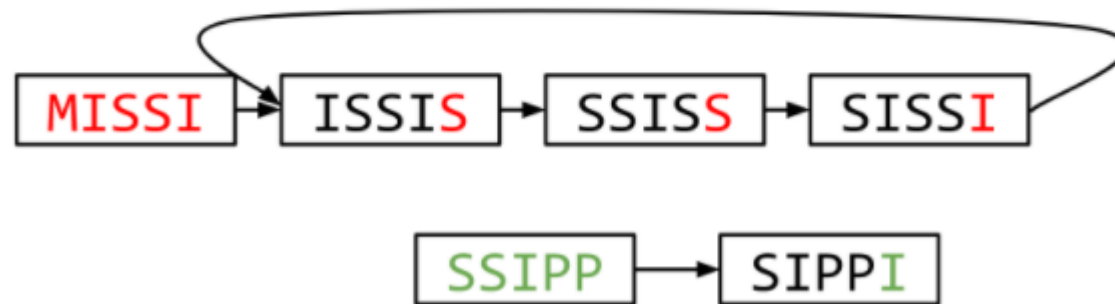
Different k

Example #2a:

MISSIS SSISSI SSIPPI

This time $k = 5$ k-mers:

MISSI ISSIS SSISS SISSI SSIPP SIPPI



MISSISSIS

SSIPPI

<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

Choose k wisely

- Lower k
 - More connections
 - Less chance of resolving small repeats
 - Higher k -mer coverage
- Higher k
 - Less connections
 - More chance of resolving small repeats
 - Lower k -mer coverage

Optimum value for k will balance these effects.

<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

Read errors

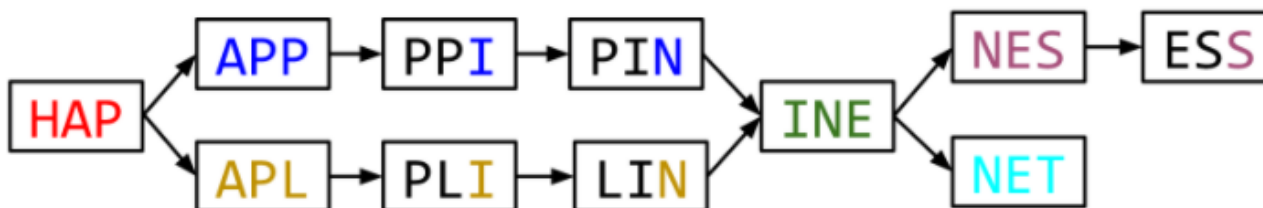


Example #3:

HAPPI INESS APLIN PINET

k = 3 k-mers:

HAP APP PPI INE NES ESS APL PLI LIN PIN NET



6 contigs: HAP APPIN APLIN INE NESS NET

More coverage



- Errors won't be duplicated in every read
- Most reads will be error free
- We can count the frequency of each k-mer
- Annotate the graph with the frequencies
- Use the frequency data to clean the de Bruijn graph

More coverage depth will help overcome errors!

<https://galaxyproject.github.io/training-material/topics/assembly/tutorials/debruijn-graph-assembly/slides.html#23>

SPAdes



- de Bruijn graph assembler by Pavel Pevzner's group out of St. Petersburg
- Uses multiple k-mers to build the graph
 - Graph has connectivity **and** specificity
 - Usually use a low, medium and high k-mer size together.
- Performs error correction on the reads first
- Maps reads back to the contigs and scaffolds as a check
- Under active development
- Much slower than Velvet
- Should be used in preference to Velvet now.

Assembly: Scaffolding

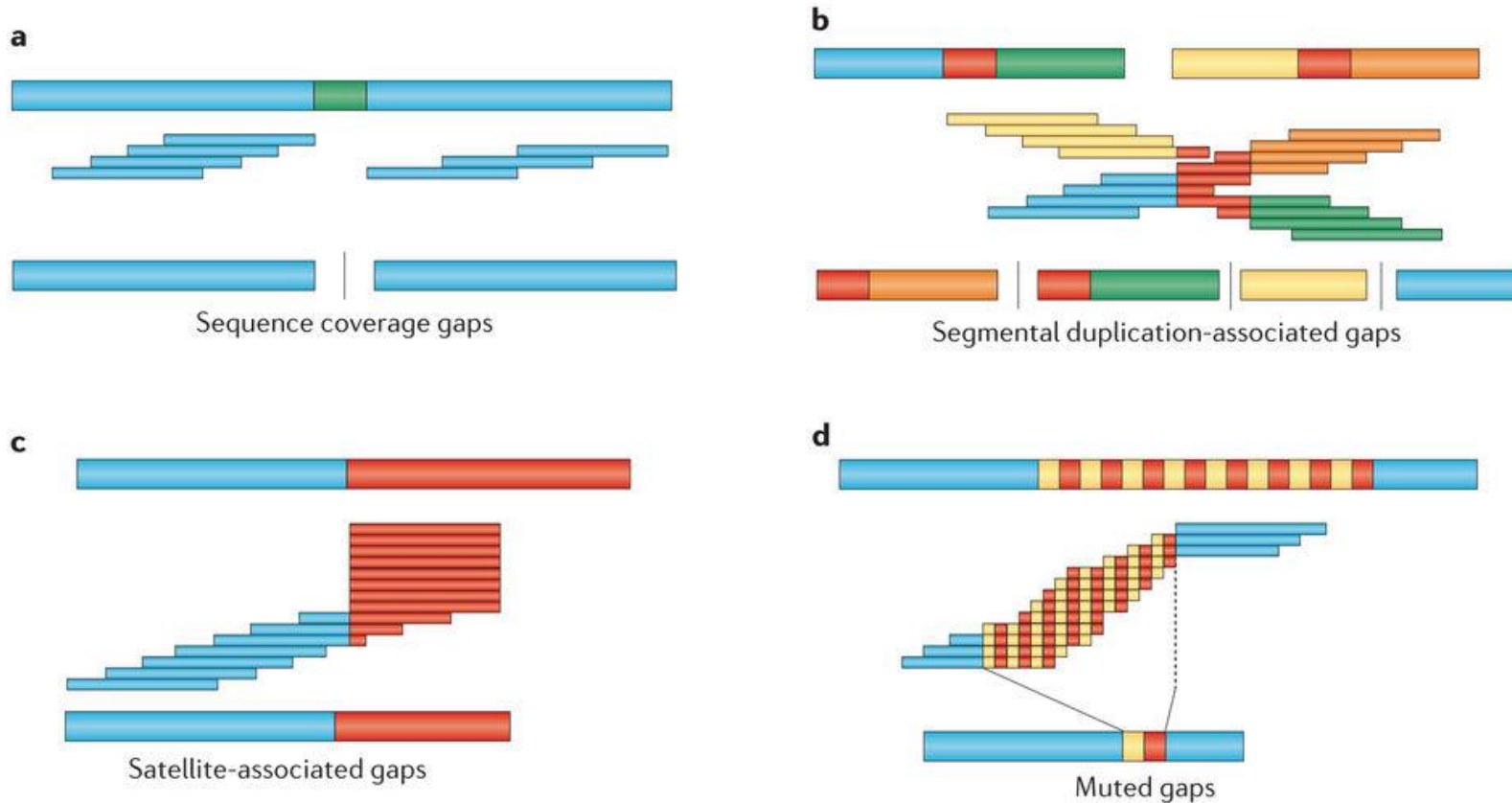
- **From draft:**
 - Order contigs** (Nucmer, if there is reference it can be used to align and guide)
 - Fill the GAPS** (GapFiller, fill sequencing gap (not physical gap))
 - Solve repeated** sequence ambiguities (Expander)
 - Resequence** with different library:
 - Longer fragments and/or distance
- **Tools for assembly improvement**
 - SSPACE (Scaffolding) REAPR (evaluate scaffolding, breaking incorrect scaffolds)
- **Assembly visualizing**
 - Artemis, ACT (compare two or more sequences), Icarus (Quast)

A move back to OLC

- New long read technologies
 - PacBio and MinIon
- Assemblers: HGap, CANU
 - Use overlap, layout consensus approach
- CANU can perform hybrid assemblies with long and short reads



Ensamblado: Errores



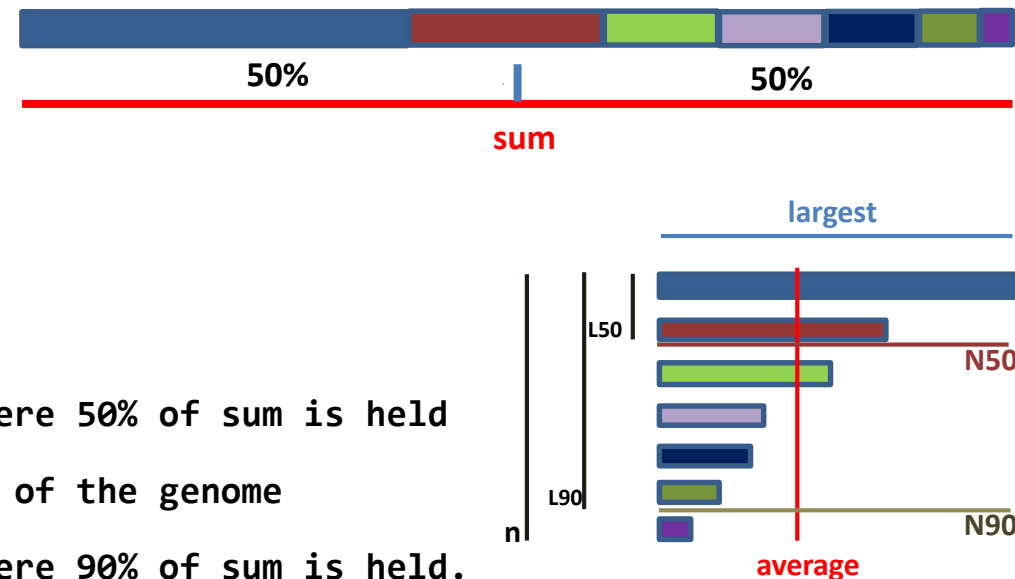
- **A. Gaps – región del genoma sin secuenciar**
- **B. Duplicaciones de gran tamaño**
 - Quimeras
- **Regiones repetidas colapsadas**
 - **C. Terminales**
 - **D. Intersticiales**

Genetic variation and the de novo assembly of human genomes
Chaisson *et al.* 2015

Nature Reviews | Genetics

Assembly: Metrics

- `sum` = total bases number
- `n` = contigs number
- `average` = average contig length
- `largest` = largest contig
- `N50` = length of the shortest contig where 50% of `sum` is held
- `L50` = number of contigs which have 50% of the genome
- `N90` = length of the shortest contig where 90% of `sum` is held.
- `L90` = number of contigs which have 90% of the genome



Assembly: Evaluation

- Software that evaluate differets algorithms & parameters
iMetAMOS, *Koren et al.*, *BMCBioinformatics* 2014, 15:126
GAGE-B, *Magoc et al.*, *Bioinformatics* 2013,29(14):1718-25
- **Graph evaluation:** Bandage, Wick R.R., Schultz M.B., Zobel J. & Holt K.E. (2015)
- **Assembly evaluation:** Quast, *Gurevich et al.*, *Bioinformatics* 2013, 29:8
- **Metrics for a good assembly:**
Large N50
Sum closest to expected
Low n
Low L50

Assembly: Evaluation - Quast

- Assembly evaluation: Quast, *Gurevich et al.*, *Bioinformatics* 2013, 29:8

Worst Median Best ☒ Show heatmap

	RA_L2073_paired_assembly	RA_L2391_paired_assembly	RA_L2677_paired_assembly	RA_L2978_paired_assembly	RA_L2281_paired_assembly	RA_L2450_paired_assembly	RA_L2701_paired_assembly
Genome statistics							
Genome fraction (%)	81.079	88.828	84.92	90.172	85.733	88.172	92.463
Duplication ratio	1	1	1.001	1.001	1.001	1	1
# genomic features	1736 + 824 part	2113 + 600 part	1881 + 768 part	2157 + 611 part	1992 + 637 part	2073 + 643 part	2368 + 412 part
Largest alignment	16612	33033	21336	25068	29638	30305	40471
Total aligned length	2 405 510	2 635 297	2 519 300	2 675 166	2 543 440	2 615 874	2 743 222
NGA50	3176	6162	4234	5948	5104	5358	9519
LGA50	267	151	219	153	166	166	96
Misassemblies							
# misassemblies	23	1	14	2	17	12	4
Misassembled contigs length	84193	9611	45868	6390	111 490	72 879	37 962
Mismatches							
# mismatches per 100 kbp	17	18.78	15	16.71	341.39	15.75	13.49
# indels per 100 kbp	1.21	1.25	1.87	1.94	7.27	1.45	0.87
# N's per 100 kbp	0	0	0	0	0	0	0
Statistics without reference							
# contigs	748	546	684	569	569	584	392
Largest contig	16612	33033	21336	25068	30915	30305	40471
Total length	2 440 656	2 676 227	2 562 578	2 714 287	2 629 607	2 618 624	2 787 129
Total length (>= 1000 bp)	2 439 127	2 676 227	2 559 569	2 714 287	2 628 029	2 615 105	2 785 415
Total length (>= 10000 bp)	257 236	739 181	320 638	811 392	700 516	658 319	1 419 641
Total length (>= 50000 bp)	0	0	0	0	0	0	0

[Extended report](#)

Assembly: Evaluation - Quast

- Assembly evaluation: Quast, *Gurevich et al.*, *Bioinformatics* 2013, 29:8



Assembly: Assemblers

Name	Type	Technologies	Author	Presented /Last updated	Licence*	Homepage
DNASTAR Lasergene Genomics Suite	(large) genomes, exomes, transcriptomes, metagenomes, ESTs	Illumina, ABI SOLiD, Roche 454, Ion Torrent, Solexa, Sanger	DNASTAR	2007 / 2016	C	link
Newbler	genomes, ESTs	454, Sanger	454/Roche	2004/2012	C	link
Canu	Small and large, haploid/diploid genomes	PacBio/Oxford Nanopore reads	Koren et al. ^[8]	2001 / 2018	OS	link
SPAdes	(small) genomes, single-cell	Illumina, Solexa, Sanger, 454, Ion Torrent, PacBio, Oxford Nanopore	Bankevich, A et al.	2012 / 2017	OS	link
Velvet	(small) genomes	Sanger, 454, Solexa, SOLiD	Zerbino, D. et al.	2007 / 2011	OS	link
*Licences: OS = Open Source; C = Commercial; C / NC-A = Commercial but free for non-commercial and academics						

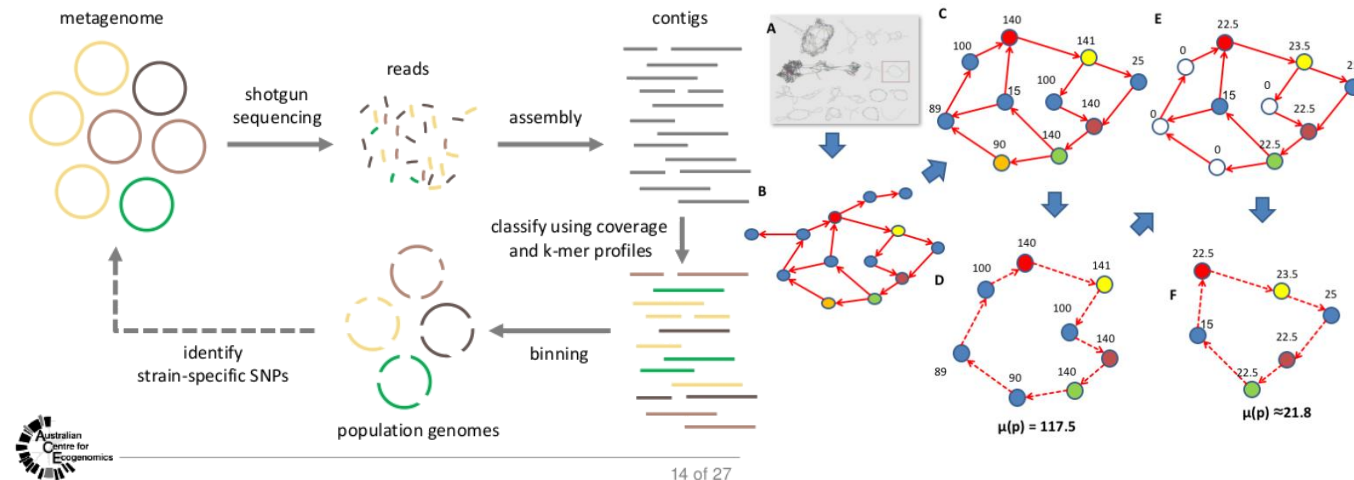
Assembly: Specials assemblers

- **Diploid genomes** recovering genomes from metagenomic data

- **Metagenomics**

- **Plasmids**

- **Transcriptome**



Hybrid genome assembly – short and long reads

https://en.wikipedia.org/wiki/Hybrid_genome_assembly

1

Short reads: ~150 nucleotides long (from 2nd gen technology)

Long reads: 100s-1000s nucleotides long (from 3rd gen technology)

2

Ambiguity in sequence assembly

Ambiguity in sequence assembly

3

Hybrid assembly helps resolve ambiguities with higher coverage and differing read lengths

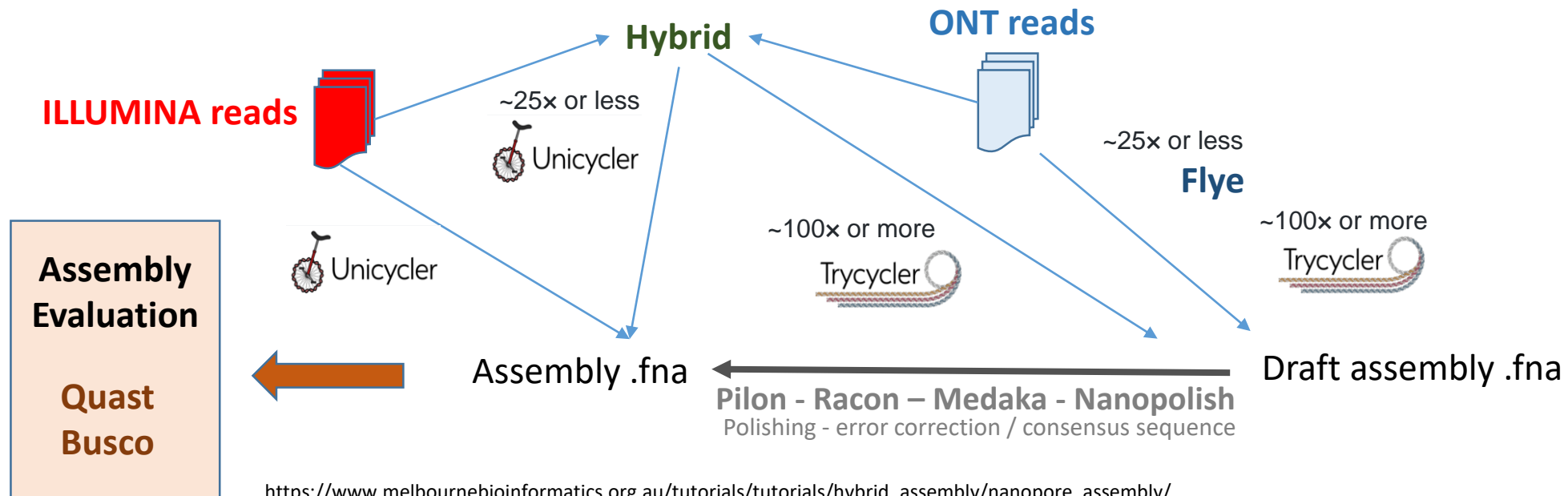
Hybrid genome assembly - nanopore and illumina

Short reads (ILLUMINA) + Long reads (ONT) → deNovo assembly (De novo assembly is the process of assembling a genome from scratch using only the sequenced reads as input - no reference genome is used.) → **high-quality assembly**

ONT: >40.000b, higher error rate – **genome structure**

ILLUMINA: 300b, lower error rate – **high base-level accuracy**

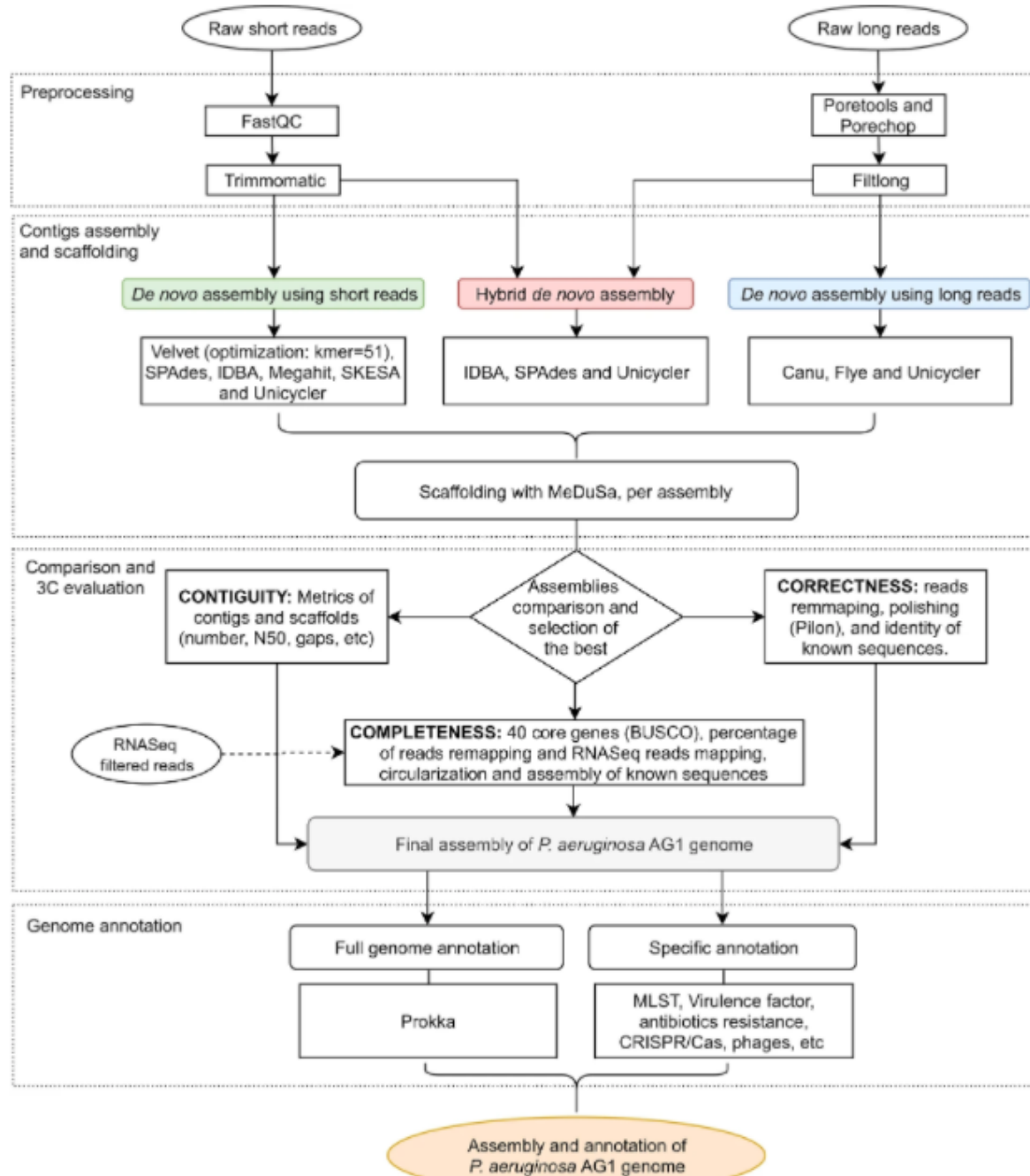
Higher COST



https://www.melbournebioinformatics.org.au/tutorials/tutorials/hybrid_assembly/nanopore_assembly/

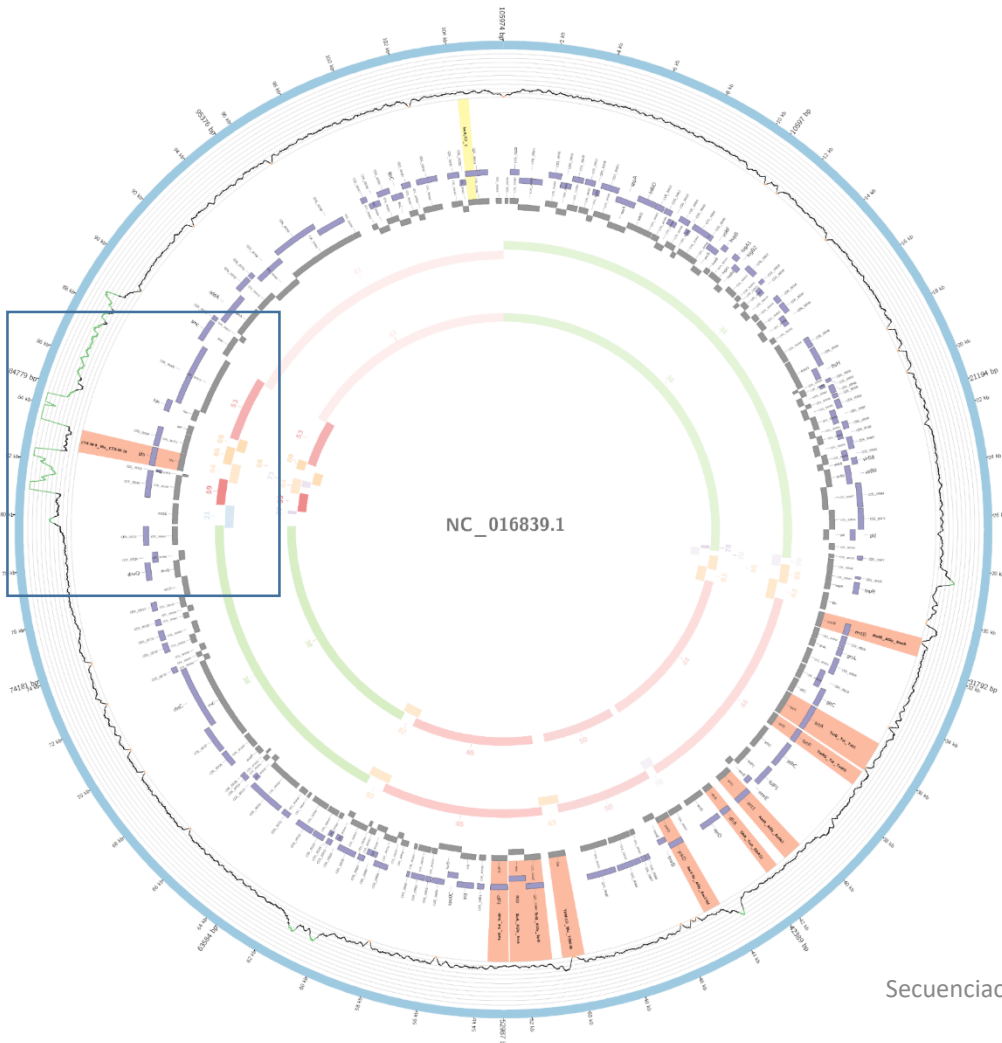
<https://github.com/rrwick/Unicycler>

<https://denbi-nanopore-training-course.readthedocs.io/en/latest/index.html>



Molina-Mora et al.,
Scientific Reports 2020

PlasmidID



Thanks for your attention!

Questions ?