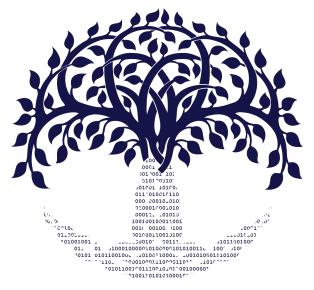
# 



# DTU Health Technology Bioinformatics

### de novo assembly

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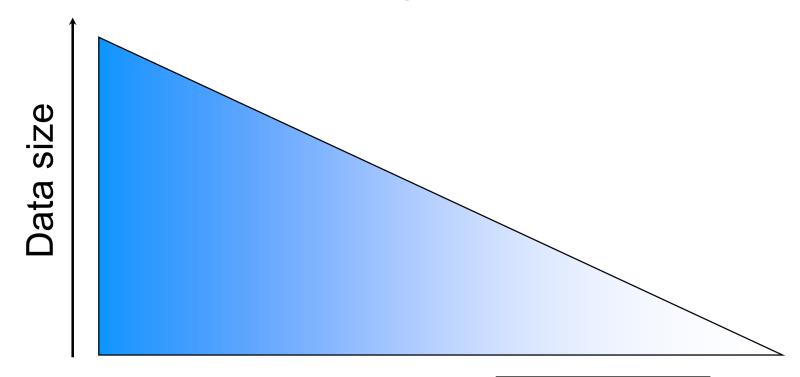


### Menu

- What do I mean by assembly?
- Assembly approaches
- Assembly graphs
- Graph postprocessing filtering
- The woes of repetition
- Benchmarking your assembly



# Generalized NGS analysis



Question

Raw reads

Preprocessing Assembly: Alignment / de novo Application specific: Variant calling, count matrix, ...

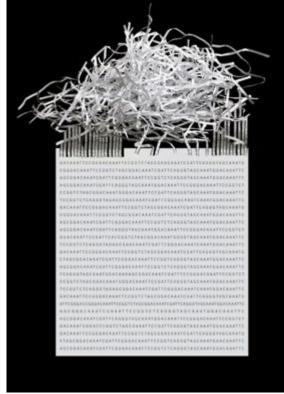
Compare samples / methods

Answer?



# What is de novo assembly?

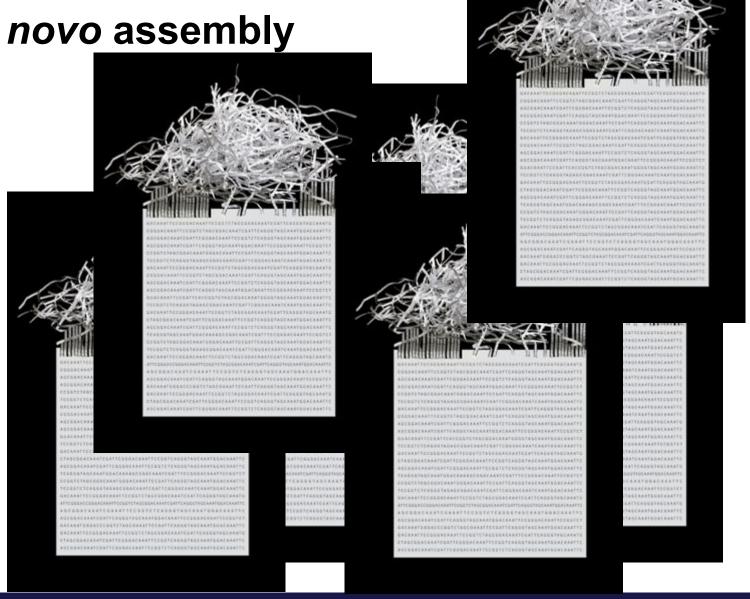
Merge small DNA fragments together so they form a previously unknown sequence





Metagenomic de novo assembly

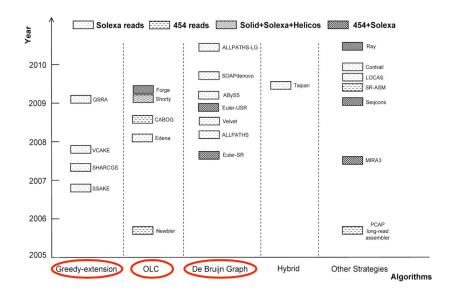
- Thousands of unique pages...
- Written with four letters
- All shredded to small pieces





# Which approaches?

- Greedy ("Simple" approach)
- Overlap-Layout-Consensus (OLC)
- de Bruijn graphs







### Simple approach - Greedy

- Principle:
  - 1. Pairwise alignment of all reads
  - 2. Identify fragments that have largest overlap
  - 3. Merge these
  - 4. Repeat until all overlaps are used
- Can only resolve repeats smaller than read length
- High computational cost with increasing no. reads



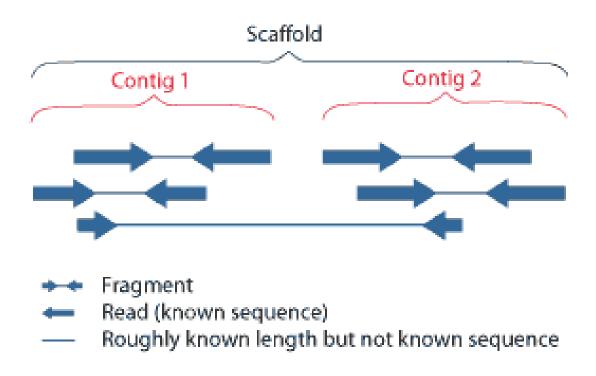
### Reads > Contigs > Scaffolds

- Overlap Layout Consensus and de Bruijn use a similar general approach.
  - 1. Try to correct sequence errors in reads with high coverage
  - 2. Assemble reads to contiguous sequence fragments "contigs"
  - 3. Identify repeat contigs
  - 4. Combine and order contigs to "scaffolds", with gaps representing regions of uncertainty



### Glueing the reads together

- Reads are assembled into contigs
- Contigs can be bridged into scaffolds by
  - Mapping against reference genome
  - Low quality sequence
  - Paired-end read information
  - Other methods such as Hi-C

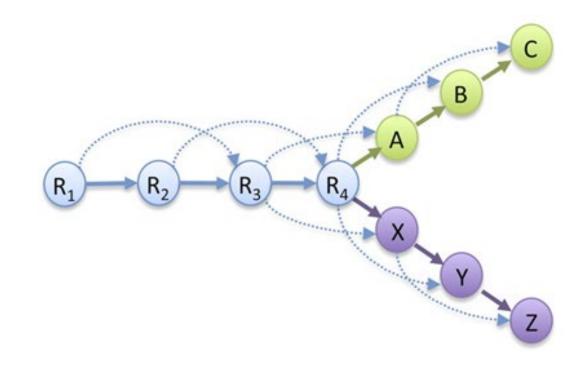




### **Overlap-Layout-Consensus**

- Create overlap graph by all-vs-all alignment (Overlap)
- Build graph where each node is a read, edges are overlaps between reads (Layout)

GACCTACA ACCTACAA R2: R3: CCTACAAG R4: CTACAAGT A: TACAAGTT ACAAGTTA B: C: CAAGTTAG X: TACAAGTC Y: ACAAGTCC Z: CAAGTCCG

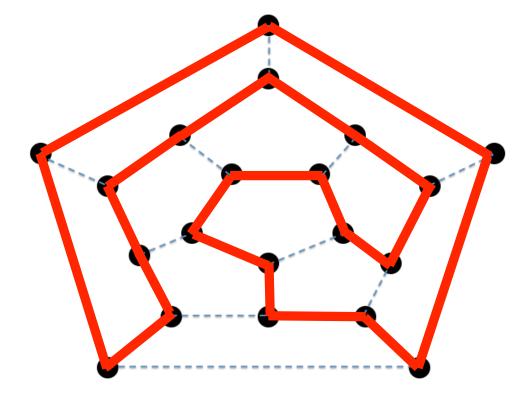


Schatz et al., Genome Res, 2010



### **Overlap-Layout-Consensus**

- Create consensus sequence
- We need to use graph theory to solve the graph
- Walk the Hamiltonian path
- Eg. visit each node exactly once
- We need to do an all-vs-all alignment!



Imagine trying to solve this for a graph of hundred of thousands of nodes (=reads)



### **Overlap-Layout-Consensus**

- Not good with many short reads -> lots of alignment!
- With short read lengths, hard to resolve repeats
- Good for large read lengths:
  - PacBio, Oxford Nanopore, 10X Genomics, 454, Ion Torrent, Sanger
- Example assemblers: Canu, Celera, Newbler



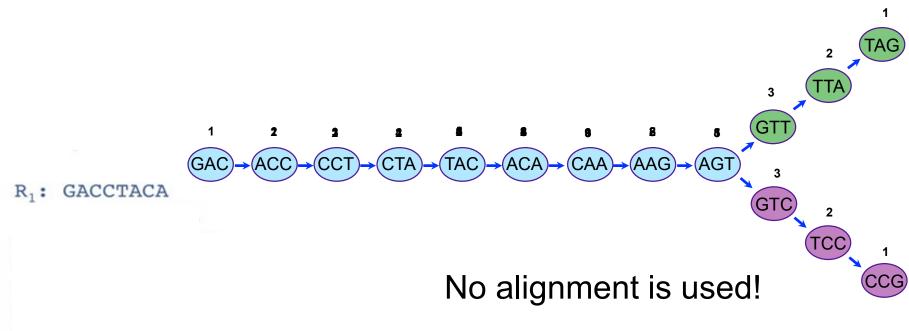
### de Bruijn graph

- Directed graph of overlapping items (here DNA sequences)
- Instead of comparing reads, decompose reads into *k*-mers
  - Graph is created by mapping the k-mers to the graph
  - Each k-mer only exists once in the graph
  - Problem is reduced to walking Eulerian trail (visiting each edge once) this is a solveable problem



### How is the graph constructed?

• Same 10 reads, extract k-mers from reads and map onto graph, k = 3:



Different assemblers uses different modifications of the de Bruijn graphs

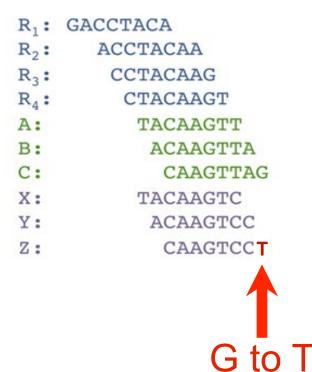


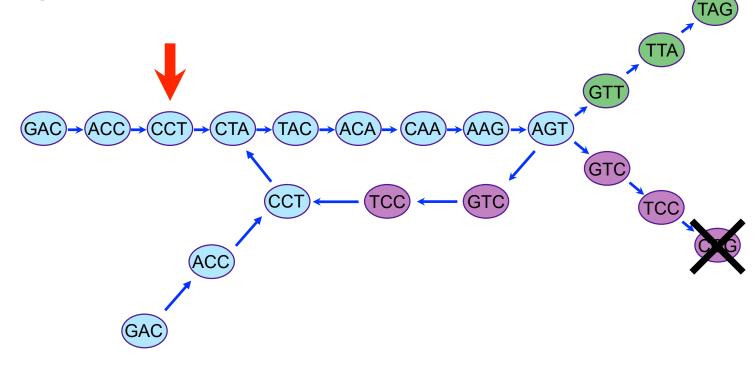
### Drawbacks ...

- Lots of RAM required (1-1000 GB!)
- Optimal *k* can not be identified *a priori*, must be experimentally tested for each dataset
- small k: very complex graph, large k: limited overlap in low coverage areas
- Iterative approach to find best assembly



# **Complicated graphs**





Large genomes with many repeats/errors creates very large graphs

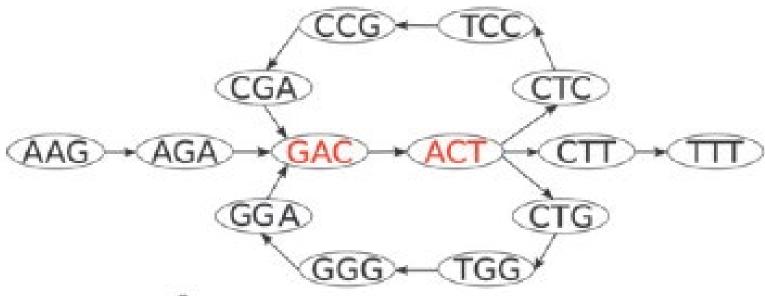


### Create the *de* Bruijn graph of this genome using k=3

### AAGACTCCGACTGGGACTTT



# AAGACTCCGACTGGGACTTT



A de Bruijn graph of a sequence



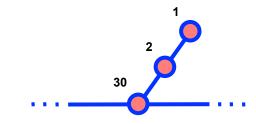
# After building: Simplify

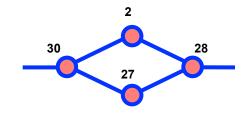
Clip tips

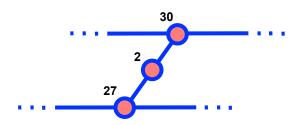
(seq err, end)

Pinch bubbles (seq err, middle, SNP)

Remove low cov. links



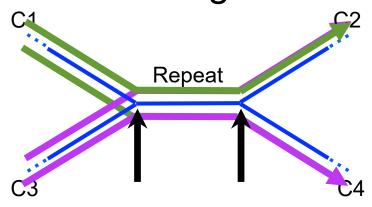






# **Create contigs and scaffolds**

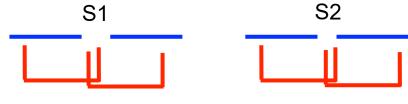
Cut graph at repeat boundaries to create contigs



Use paired end information to resolve repeats and combine to scaffolds



Fill potential gaps using PE reads



The assembly is done



### **Iterate parameters**

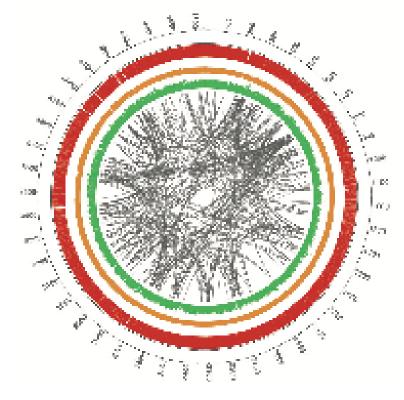
- Re-run with different k-sizes, find optimum
- Run with multiple k-mers at the same time! (eg. SPAdes)
- Compare assembly statistics such as, assembly length, N50, no. contigs

- Assembly refinement
  - Break contigs not supported by PE/MP reads
  - Analyze assembly using REAPR or QUAST



### Successful de novo assembly

- Success is a factor of:
  - Genome size, genomic repeats(!), ploidy
  - High coverage, long read lengths, PE/MP libraries



Repeats in *E. coli* 



### Improving de novo assemblies

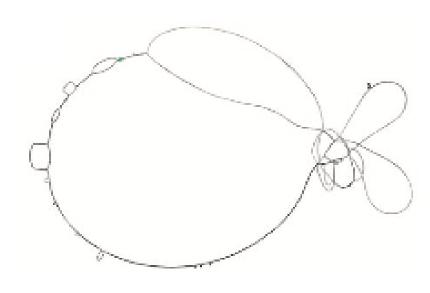
- Paired end & Mate pair for long range continuity
- Hybrid approaches (combine Illumina with PacBio/Oxford Nanopore)
- Synthetic long reads: Illumina Synthetic Reads (Moleculo) or 10X Genomics
- Hi-C contact maps



# Two bacterial genomes de Bruijn graphs

Few repeats

"more" repeats



Flicek & Birney, Nat.Methods 2009

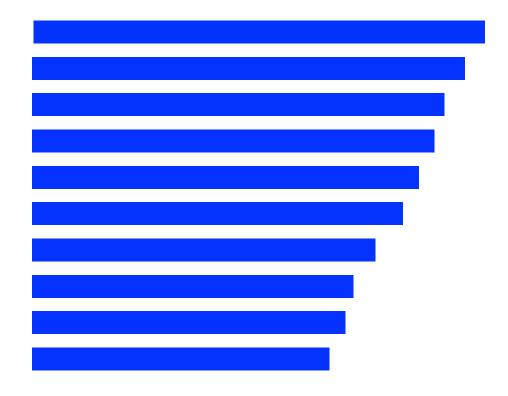
Zerbino, 2009



### N50: Assembly quality

N50: What is the smallest piece in the largest half of the assembly?

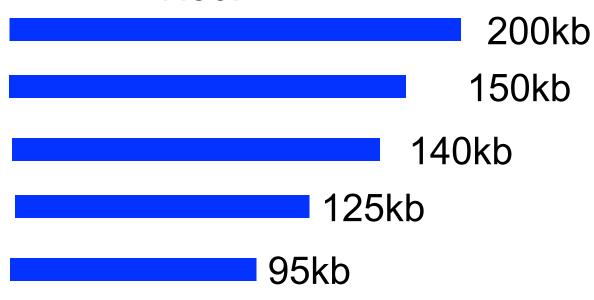
- Calculate sum of assembly
- Order contigs by size
- Sum contigs starting by largest
- When half the sum is reached, N50 is the length of the contig





# N50 example

5 scaffolds, calculate N50:



Sum: 200+150+140+125+95=710kb

Half: 710 / 2 = 355kb

$$200kb + 150kb = 350kb$$

$$350kb + 140kb = 490kb$$



### Some assemblers

- OLC: <u>Canu</u>, <u>Newbler</u>
- de Bruijn: Allpaths-LG, SPAdes (best), Velvet, SOAPdenovo, Megahit (very lean), ...
- other: MIRA, SGA, <u>Flye</u> (very good for 3g NGS)

Used in exercises today



### **Exercise time!**

http://teaching.healthtech.dtu.dk/22126/index.php/Denovo exercise