



COMP90014

Algorithms for Bioinformatics Week 8B: Assembly in Practice

Assembly In Practice

De Bruijn Graph Simplification

Kmers, coverage, depth Assessing Assemblies Software

Read Errors

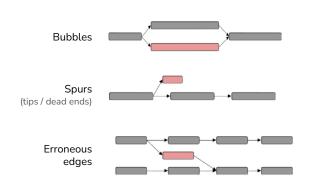
Mainly affects De Bruijn Graphs

OLC: alignment allows mismatches, gaps De Bruijn: use of kmers & exact matching

Middle of read: causes bubbles

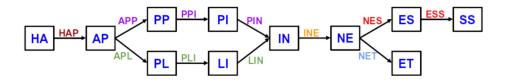
End of read: cause spurs

Use of kmers: causes erroneous edges



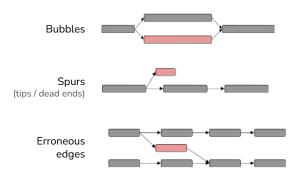
Read Errors





Graphs are very complex before simplification.

- Heterozygosity, read errors, kmer length
- Result in bubbles, spurs, erroneous edges

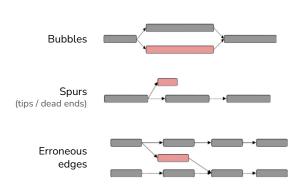


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Naively

- Remove all nodes with low coverage.
- Eg. Expected = 30x
- Remove nodes <= 5x coverage



Graphs are very complex before simplification.

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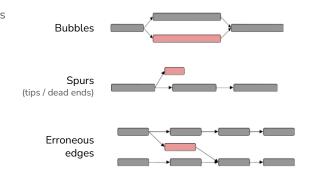
Naively 解决方法 移除覆盖度低的节点

- Remove all nodes with low coverage.
- Eg. Expected = 30x
- Remove nodes <= 5x coverage

Issues

- May remove genuine regions (just had low sequencing depth)
- Doesn't address heterozygosity

杂合性(Heterozygosity):一个位置有不同的碱基 (即存在变异)。 读错误(Read errors):测序过程中产生的错误。 kmer长度:选择的kmer长度可能不适合测序数据的特点



Will explore how Velvet handles graph simplification.

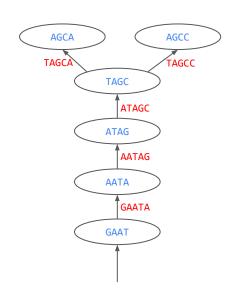
例如,如果一个基因组位置在测序过程中被10个不同的reads覆盖,那么这个位置的覆盖度就是10x。

Step 1: Coalesce non-branching paths

How we built the De Bruijn Graph:

- Nodes: prefix / suffix

- Edges: kmers



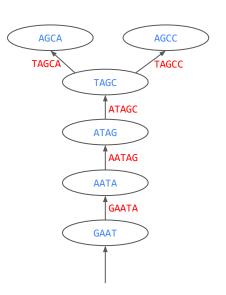
Step 1: Coalesce non-branching paths

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Now we're simplifying. No reason to keep structure as-is.



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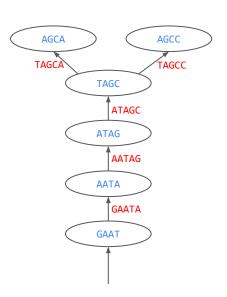
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First task: collapse linear chains into single node.

- Improves space performance (less nodes / edges)
- Improves time performance (traverse less edges)



Step 1: Coalesce non-branching paths

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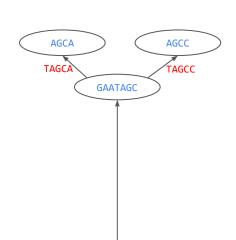
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Now we're simplifying. No reason to keep structure as-is.

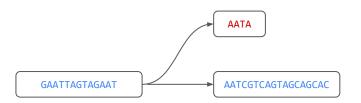
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Step 1: Remove Spurs (tips)

Spur: chain of nodes that is disconnected on one end



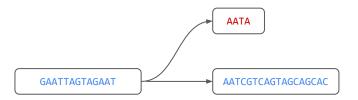
Step 1: Remove Spurs (tips)

Spur: chain of nodes that is disconnected on one end

Similar to OLC spur removal. Straightforward.

Can't be heavy-handed!

Don't want to remove genuine sequence.



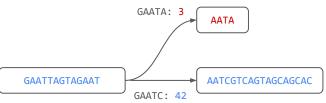
不能过干 粗暴(不 能heavyhanded

Step 1: Remove Spurs (tips)

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Similar to OLC spur removal. Straightforward.

Can't be heavy-handed! Don't want to remove genuine sequence.



筛选tip的时候 看length和mc

Identify tips via length and minority count

Length: Path length along tip < 2k.

Minority count:

- The branch leading to the tip is an inferior route.
- Edge (kmer) to tip branch has lower occurrences than other branches

长度:沿着尖刺的路径长度小干 2k,这里的k是k-mer的长度。例 如果k=20,那么尖刺的长度 计数:通向尖刺的分支是次 优路径,意味着这条路径比其他 路径覆盖度低或出现次数少。

Step 2: Pop bubbles

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Conditions for merging (popping bubbles)

Scenario 1: sequencing error / heterozygosity



Step 2: Pop bubbles

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Conditions for merging (popping bubbles)

Must have same start and end node

Step 2: Pop bubbles

Scenario 1: sequencing error / heterozygosity

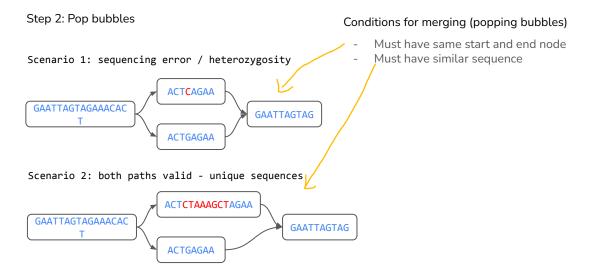
GAATTAGTAGAAACAC
T
ACTCAGAA
GAATTAGTAG
GAATTAGTAG

Scenario 2: both paths valid - unique sequences



Conditions for merging (popping bubbles)

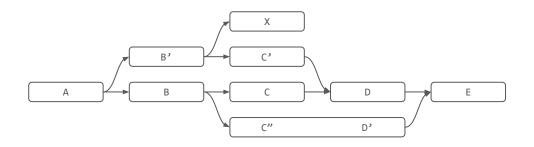
- Must have same start and end node



Step 2: Pop bubbles

Scenario 3: complex (in this figure - nodes labelled, not showing seq)

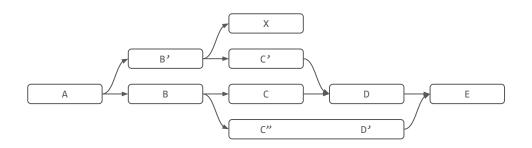
- Must have same start and end node
- Must have similar sequence



Step 2: Pop bubbles

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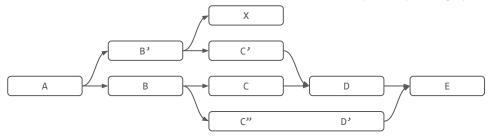
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- Only merge two paths at a time



Step 2: Pop bubbles

Scenario 3: complex (in this figure - nodes labelled, not showing seq)

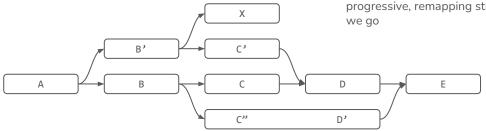
- Must have same start and end node
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- Only merge two paths at a time
- Cannot just delete nodes because this separates part of graph



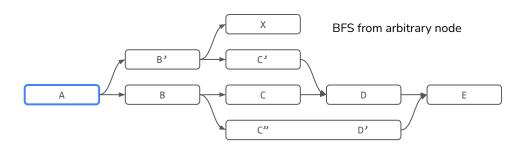
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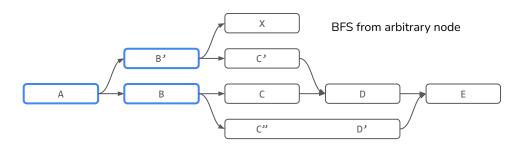
- Must have same start and end node
- Must have similar sequence
- Only merge two paths at a time
- Cannot just delete nodes because this separates part of graph. Must be progressive, remapping structures as we go



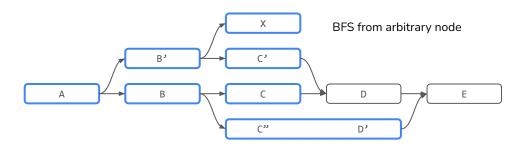
Step 2: Pop bubbles - Tour Bus



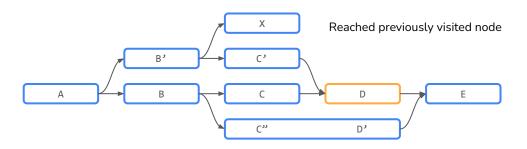
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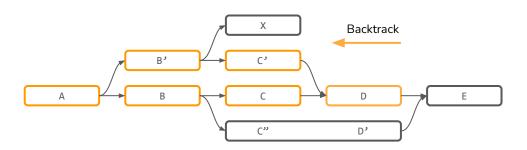
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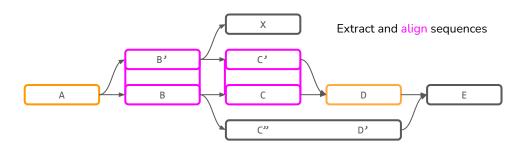
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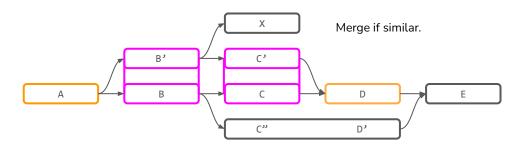
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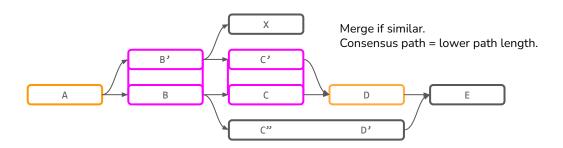
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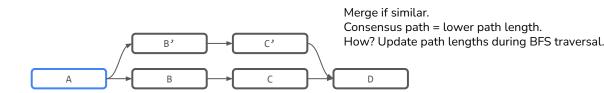
Step 2: Pop bubbles - Tour Bus



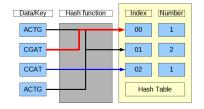
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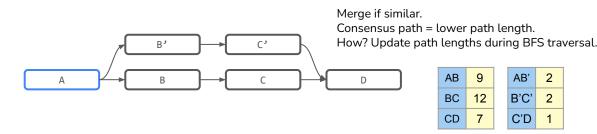


Step 2: Pop bubbles - Tour Bus

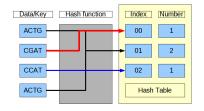


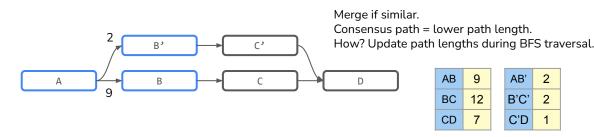
Step 2: Pop bubbles - Tour Bus



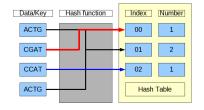


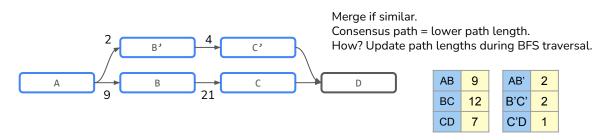
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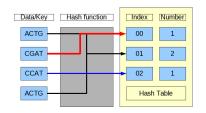


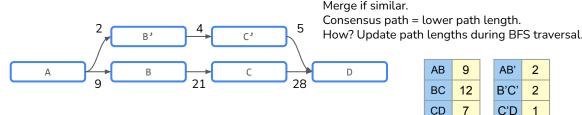
Step 2: Pop bubbles - Tour Bus





Step 2: Pop bubbles - Tour Bus kmer的出现次数



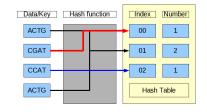


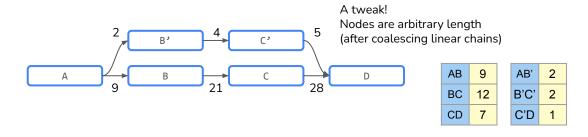
路径权重代表了这条路径上序列的频率或可信度。

这里 权重可能代表特定序列在读序数据中出现 的次数

Step 2: Pop bubbles - Tour Bus

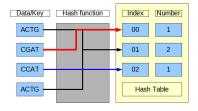
图中B到C和B' 到C' 形成了两个泡泡,算法将识别这些泡泡,并根据一定的规则(如路径长度、权重或覆盖度)决定哪些路径保留,哪些路径移除。

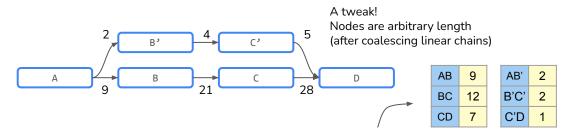




意味着在这一步骤中,具有线性关系的节点(没有分叉的节点链)可以合并成一个具有任意长度的节点。这有助于进一步简化图的结构

Step 2: Pop bubbles - Tour Bus

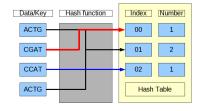




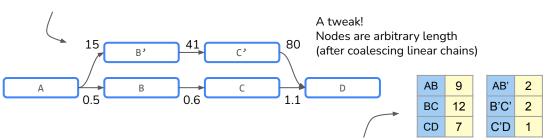
Multiplicity was updated during coalescing.

Path length (AB) = len(B) / multiplicity

Step 2: Pop bubbles - Tour Bus



Actual path lengths using definition

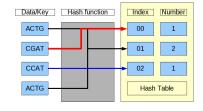


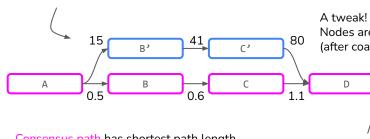
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Step 2: Pop bubbles - Tour Bus

Actual path lengths using definition





Nodes are arbitrary length (after coalescing linear chains)

> AB' 12 B'C' C'D

Consensus path has shortest path length Essentially majority voting

Multiplicity was updated during coalescing.

AB

BC

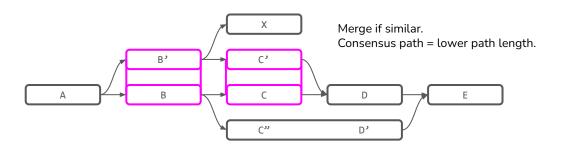
CD

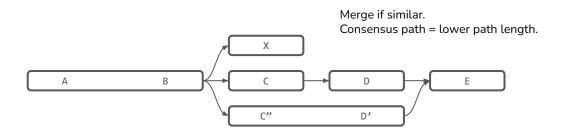
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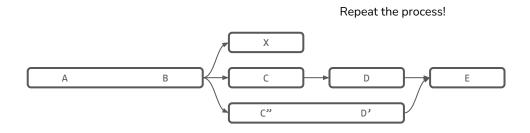
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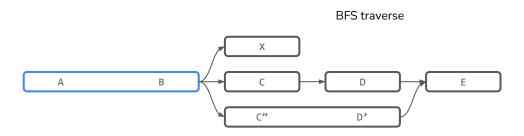
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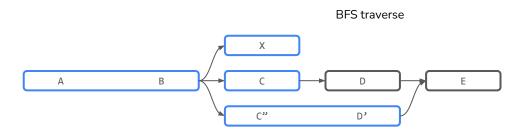


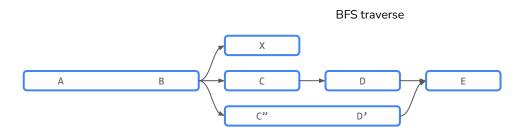


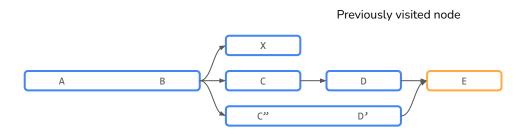


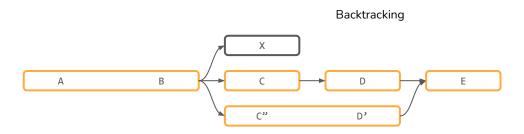


Step 2: Pop bubbles - Tour Bus



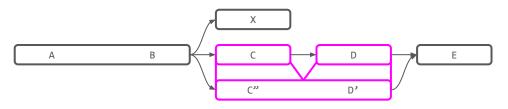


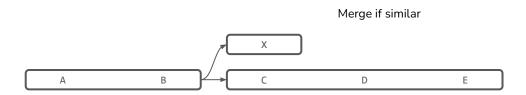


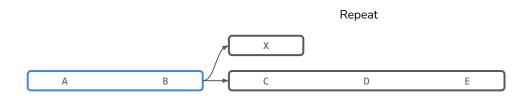


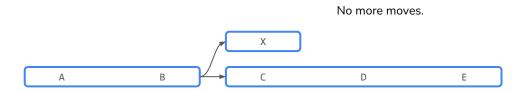
Step 2: Pop bubbles - Tour Bus

Extracting and aligning sequences









Step 2: Pop bubbles - Tour Bus

Paths are redundant if start and end at same nodes. Also required to have similar sequence.

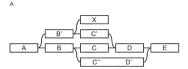
Tour Bus algorithm

Dijkstra-like breadth-first search

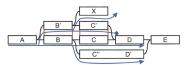
Since we have coalesced non-branching paths already, start at any arbitrary node. All have 2+ in- or out-edges.

Visit nodes in BFS manner

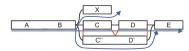
When reach visited node (join), <u>backtrack</u> to determine best path.



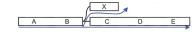
В



C



D



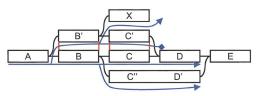
Step 2: Pop bubbles - Tour Bus

For given node, walk BFS to the right. Update path lengths as we go.

 $A \rightarrow B$: length(B) % multiplicity of edge to B

This is essentially priority voting.

Prioritises higher confidence branches.



When reach visited node (D):

Traceback visited edges

Find closest common ancestor (A)

Extract sequences from traceback paths

Align each sequence & merge if similar (red)

Merging occurs consecutively, left to right

Consensus sequence determined by path length

Zerbino & Birney (2008): Velvet

Tour Bus的谨慎性:Tour Bus算法非常谨慎,它保留了大部分独特的区域

剩余错误:剩下的错误可能是由持续的错误、嵌合体(chimeras,指的是在生物学中由两种不同物种的遗传物质混合而成的组织),或是重复序列导致的。

Step 3: Remove Erroneous Connections

Occurs after Tour Bus.

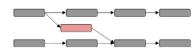
Tour bus is cautious, preserves most unique regions

Remaining are erroneous due to persisting errors, chimeras, repeats

难点:错误连接通常不会形成可识别的循环或结构,使得它们难以识别和移除。

Tricky!

Don't form recognizable loop / structure



Only solution - coverage cutoff

Remove any node with less then set threshold for coverage.

Repetitive sequence will remain, as it's coverage will be high. Can't do much about this.

It is not possible to resolve a repeat of length ${f N}$ with reads less than length ${f N}$

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Both OLC and De Bruijn approaches handle repeats by essentially leaving them out.

After simplification, nodes with 2+ in-edges / out-edges are unresolvable

At these branch points we break the assembly into fragments (contigs)

It is not possible to resolve a repeat of length ${\bf N}$ with reads less than length ${\bf N}$

解决长度为N的重复序列,如果读取长度小于N,是不可能的。

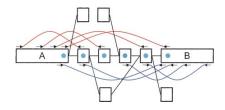
Both OLC and De Bruijn approaches handle repeats by essentially leaving them out.

After simplification, nodes with 2+ in-edges / out-edges are unresolvable

At these branch points we break the assembly into fragments (contigs)

*If paired-end short reads (+ De Bruijn):

- Can use mate pairs to help resolve
- See Velvet Breadcrumb algorithm https://doi.org/10.1101/gr.074492.107



Assembly In Practice

De Bruijn Graph Simplification

Kmers, coverage, depth

Assessing Assemblies

Software

Coverage vs. depth

基因组测序中覆盖度(Coverage)和深度(Depth)的概念。

表示至少被一个测序读段(read)覆盖的基因组分数。换句话说,它衡量了测序读段覆盖基因组的广度。例如,如果覆盖度是90%,这意味着基因组中有90%的区域至少被一个测序读段覆盖到。

Coverage (or breadth)

Fraction of the genome sequenced by at least one read

表示平均每个给定区域被多少个测序读段覆盖。深度可以告诉我们某个特定区域的测序数据有多少重叠,是对测序重复性的一种衡量。如果一个区域的深度是10,这意味着这个区域在平均意义上被10个测序读段所覆盖

Depth

Average number of reads that cover any given region

Intuitively: more reads should increase coverage and depth

直观上,更多的测序读段会增加基因组的覆盖度和深度

Coverage
$$= \frac{3}{8}$$

 $Depth = \frac{3}{8}$

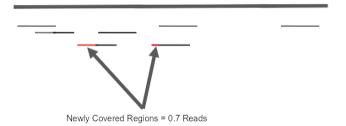
Coverage
$$=\frac{4.5}{8}$$

$$\text{Depth} = \tfrac{5}{8}$$



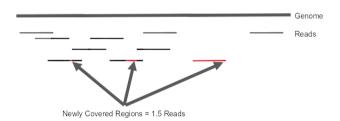
Coverage
$$=\frac{5.2}{8}$$

$$\text{Depth} = \tfrac{7}{8}$$



Coverage
$$=\frac{6.7}{8}$$

$$\text{Depth} = \tfrac{10}{8}$$



Calculating depth

depth: The number of reads covering each position in the genome.

Depends on number of reads and read length (L).

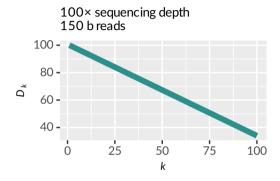
k-mer depth depends on read length:

$$D_k = D \times \frac{L - k + 1}{L}$$

k-mer depth: The number of observedk-mers that match a givenk-length segment of the genome.

- i.e. the number of occurrences of an identical k-mer.
- ø depth and coverage are related but not the same.
- number of sequences at a single position vs. fraction of genome sequenced
- don't worry about the terminology

Choosing *k* in practice



$$D_k = D \times \frac{L-k+1}{L}$$

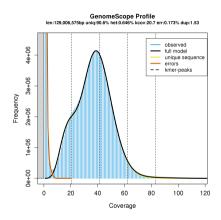
Lower k

- more connections → more loops
- lower chance of resolving small repeats
- higher k-mer coverage

Higher k

- $\bullet \ \ \text{fewer connections} \to \text{more} \\ \ \ \text{components} \\$
- higher chance of resolving small repeats
- lower k-mer coverage
- Choose k (empirically) to balance these effects

k-mer based genome size estimation



Here, "Coverage" means k-mer depth

for k-mer depth,

$$D_k = D \times \frac{L - k + 1}{L}$$

we know read depth is the yield over the genome size,

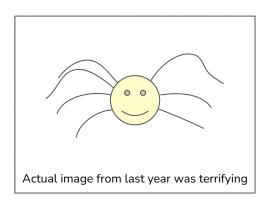
$$D = N \times \frac{L}{G}$$

just plug in values to get genome size,

$$G = N \times \frac{L - k + 1}{D_k}$$

- try to work out the genome size here (pretend you can't see the answer):
 - the graph was made using 31-mers
 - the peak is roughly k = 40
 - we had 100 million 150 b reads

Example: Coverage of the Tarantula genome



The size estimate of the tarantula genome based on k-mer analysis is 6 Gb and we sequenced at 40× depth from a single female A. geniculata.

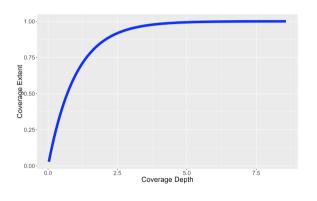
Depth =
$$N \times \frac{L}{G}$$

$$40 = N \times \frac{100}{6 \times 10^9}$$

$$N = 40 \times \frac{6 \times 10^9}{100}$$

$$N = 2.4 \times 10^9$$

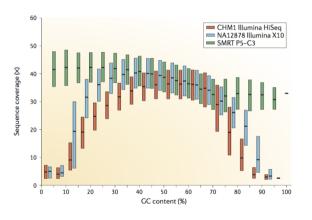
Coverage and depth are related



Approximately, assuming random reads:

$$\mathsf{Coverage} = \mathbf{1} - e^{-\mathsf{Depth}}$$

These days: just buy more sequence



- Sequencing is not random
 - GC and AT rich regions are under represented
 - Other chemistry quirks
- More depth needed for:
 - sequencing errors
 - polymorphisms

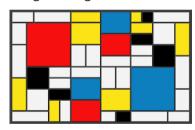
Assembly In Practice

De Bruijn Graph Simplification Kmers, coverage, depth <u>Assessing Assemblies</u> Software

Contiguity Completeness Correctness

Contiguity

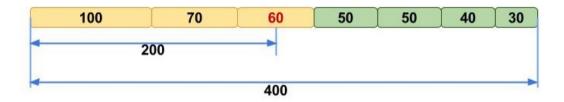
- Aim:
 - Fewer contigs
 - Longer contigs



- Metrics
 - Number of contigs
 - Average contig length
 - Median contig length
 - Maximum contig length
 - N₅₀

contigs "是指在基因组组装过程中生成的连续的DNA序列

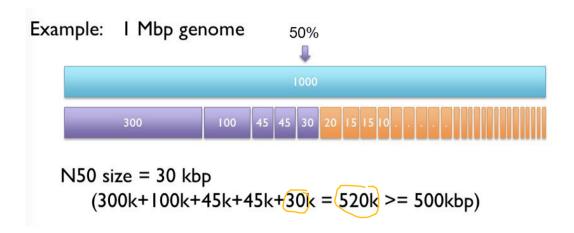
Contiguity: N₅₀ length



The sequence length of the shortest contig at 50% of the total genome length

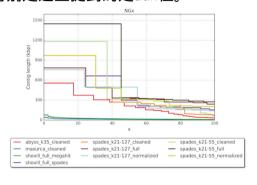
N50长度是基因组组装中一个关键的质量指标。它代表的是长度在中位数以上的contigs(序列片段)的最小长度,也就是说,当你将所有contigs按长度从长到短排序,并且累加长度时,N50是那个长度,累加到这个长度的所有contigs包含了至少50%的基因组总长度。

Contiguity: *N*₅₀ length



Area under the N_x curve?

基因组组装质量评估的另一个指标——Nx 曲线下面积(Area under the Nx curve) ,特别是这里提到的是auN值。



$$auN = \frac{\sum_{i} L_{i}^{2}}{\sum_{j} L_{j}}$$

- sum of the scaffold lengths squared, divided by genome size
- takes the whole distribution into account
- quicker to calculate
- nobody uses it

图表显示的是不同基因组组装软件(如ABySS, SPAdes, MaSuRCA等)输出的Nx曲线。Nx曲线是一个用于评估基因组组装连续性的图表,其中x轴代表的是组装得到的scaffold(基因组较大片段的组装结果)或contig所覆盖的基因组百分比,y轴代表的是长度至少为x%基因组大小的最短scaffold或contig的长度。

完整性等于 组装基因组的大小 估计的基因组大小

Completeness

完整性也可以通过计算组装基因 组中的核心基因数量与数据库中的核心基因数量的比例来评估。

Total size:

$$Completeness = \frac{Assembled genome size}{Estimated genome size}$$

- Completeness =
- Proportion of the original genome represented by the assembly
 - between 0 and 1
 - estimates are not perfect

Core genes:

$$Completeness = \frac{\text{Number of core genes in assembly}}{\text{Number of core genes in database}}$$

- Proportion of expected core genes found in other organisms that are present in the assembly
- Assumes that core genes and other genes are assembled at the same rate



Correctness

- Proportion of the assembly that is free from mistakes
- Errors include:
 - 1. Mis-joins
 - 2. Repeat compressions
 - 3. Unnecessary duplications
 - 4. Indels / SNPs caused by assembler

这张幻灯片讲述了基因组组装的"正确性" (Correctness),指的是组装基因组中没有错误的比例。这里提到的错误包括:

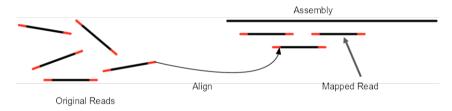
Mis-joins - 错误连接,指的是在组装过程中,本不该连接在一起的序列被错误地拼接起来。Repeat compressions - 重复压缩,发生在组装算法错误地将重复序列缩短。Unnecessary duplications - 不必要的重复,指的是在基因组中错误地复制了某些片段。Indels / SNPs caused by assembler - 组装器引起的插入/缺失(Indels)或单核苷酸多态性(SNPs),即组装过程中产生的错误插入或删除一个或几个核苷酸,或者错误地改变了单个核苷酸。

Correctness: check for self consistency

将所有的读段(reads)重新对齐到拼接片段(contigs)上 - 这一步是为了验证组装拼接的准确性。读段是原始的基因组序列数据,而拼接片段是将这些读段通过计算方法拼接起来的较长的序列片段。

- Align all the reads back to the contigs
- Look for inconsistencies

寻找不一致之处 - 对齐后,需要检查读段与拼接片段之间的匹配度。任何不匹配或者对齐不当的地方都可能表示组装过程中产生了错误。



幻灯片下方的插图显示了读段与组装后的基因组之间的对齐。黑色线表示组装后的基因组序列(Assembly),黑色带红色端点的线段表示原始读段(Original Reads),灰色箭头显示了如何将这些读段对齐到组装序列上(Align),对齐后的读段称为映射读段(Mapped Read)。

Assembly In Practice

De Bruijn Graph Simplification Kmers, coverage, depth Assessing Assemblies <u>Software</u>

de Bruijn graph assemblers

Velvet.

- Velveth
 - makes and counts (hashes) the k-mers in O(N) time
- Velvetg
 - Makes the graph in O(U) time. U = unique k-mers.
 - k-mer depth cutoff to simplify the graph
 - Makes contigs in O(E) time. E = edges in graph
- single k-mer size
- requires multiple runs to optimise parameters

Spades



- uses multiple k-mer sizes
 - low, medium and high k-mer size
 - graph has connectivity AND specificity
- performs error correction on the reads first
- maps reads back to contigs to check consistency
- development still active
- slower than Velvet, but don't have to run multiple times.

Some commonly-used assemblers

Short-read based

de Bruijn graph:

- SPAdes
- Velvet
- AbySS
- DISCOVAR / ALLPATHS
- Meraculous
- SOAPdenovo
- many more: see
 <u>De novo sequence assemblers</u> on
 Wikipedia

OLC algorithm:

wgs-assembler (celera)

Long read (mostly OLC)

- Canu
- Flye
- Shasta

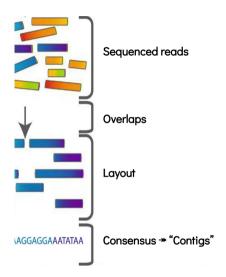
Hybrid

- MaSuRCA
- Unicycler

Special cases

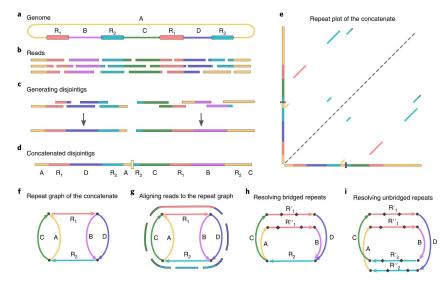
metagenomes, transcriptomes ...

Canu assembler



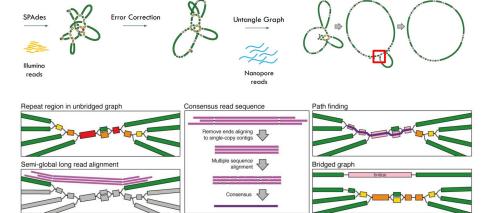
- fork of Celera assembler
- corrects the reads
 - takes longest reads to make $\approx 20 \times$ coverage depth
 - uses left over shorter reads to "correct" long reads
 - trims off the low quality ends
- OLC assembly

Flye: repeat graph assembler



Mikhail Kolmogorov, Jeffrey Yuan, Yu Lin and Pavel A. Pevzner, Nat. Biotech. 2019

Unicycler



Ryan Wick et al. 2017

Transcriptome Assembly

Transcriptome

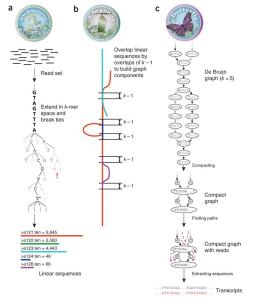
Set of all mRNA transcripts Complex in Eukaryotes due to splice forms

De novo assembly can reconstruct transcriptome

RNAseq reads are input data
Output contigs are the set of transcripts

Examples

Trinity
Velvet
SOAPdenovo-Trans
Trans-ABySS



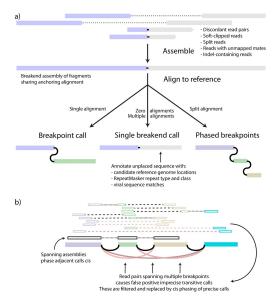
Trinity - de novo transcriptome assembly Grabherr et al. (2011)

Breakpoint Assembly

Assembly even pops up in surprising places!

GRIDSS2

Short-read structural variant (SV) caller Gathers reads possibly involved with a SV Performs assembly to reconstruct variant



GRIDSS2 - Assembly to identify breakpoints

Cameron et al. (2021)





Thank you!

Today: Assembly in Practice

Next time: Dimensionality Reduction