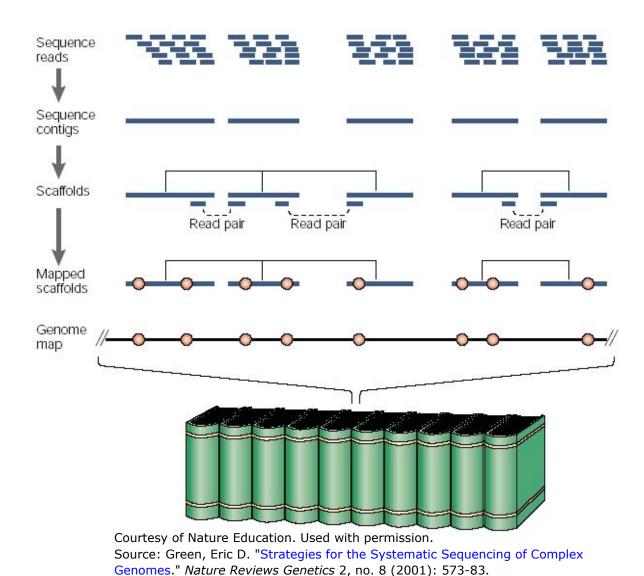
Lecture 6 Genome Assembly

Foundations of Computational Systems Biology
David K. Gifford

1

de novo whole-genome shotgun assembly



Adams, J. (2008) Complex genomes: Shotgun sequencing. Nature Education 1(1)

Whole-genome "shotgun" sequencing starts by copying and fragmenting the DNA

("Shotgun" refers to the random fragmentation of the whole genome; like it was fired from a shotgun)

Input: GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT

Copy: GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT
GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT
GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT
GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT

Fragment: GGCGTCTA TATCTCGG CTCTAGGCCCTC ATTTTTT
GGC GTCTATAT CTCGGCTCTAGGCCCTCA TTTTTT
GGCGTC TATATCT CGGCTCTAGGCCCT CATTTTTT
GGCGTCTAT ATCTCGGCTCTAG GCCCTCA TTTTTT

Assume sequencing produces such a large # fragments that almost all genome positions are *covered* by many fragments...

Reconstruct this

CTAGGCCCTCAATTTTT
CTCTAGGCCCTCAATTTTT
GGCTCTAGGCCCCTCATTTTTT
CTCGGCTCTAGCCCCTCATTTTT
TATCTCGACTCTAGGCCCTCA
TATCTCGACTCTAGGCC
TCTATATCTCGGCTCTAGG
GGCGTCTATATCTCG
GGCGTCGATATCT
GGCGTCTATATCT

From these

→ GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT

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...but we don't know what came from where

Reconstruct this CTAGGCCCTCAATTTTT
GGCGTCTATATCT
CTCTAGGCCCCTCAATTTTT
TCTATATCTCGGCTCTAGG
GGCTCTAGGCCCTCATTTTT
CTCGGCTCTAGCCCCTCATTTTT
TATCTCGACTCTAGGCCCTCA
GGCGTCGATATCT
TATCTCGACTCTAGGCC
GGCGTCTATATCTCG

From these

→ GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT

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Key term: coverage. Usually it's short for average coverage: the average number of reads covering a position in the genome.

CTAGGCCCTCAATTTTT

CTCTAGGCCCTCAATTTTT

GGCTCTAGGCCCTCATTTTTT

CTCGGCTCTAGCCCCTCATTTT

TATCTCGACTCTAGGCCCTCA

TATCTCGACTCTAGGCC

TCTATATCTCGGCTCTAGG

GGCGTCTATATCTCG

GGCGTCGATATCT

GGCGTCTATATCT

GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT

35 nucleotides

177 nucleotides

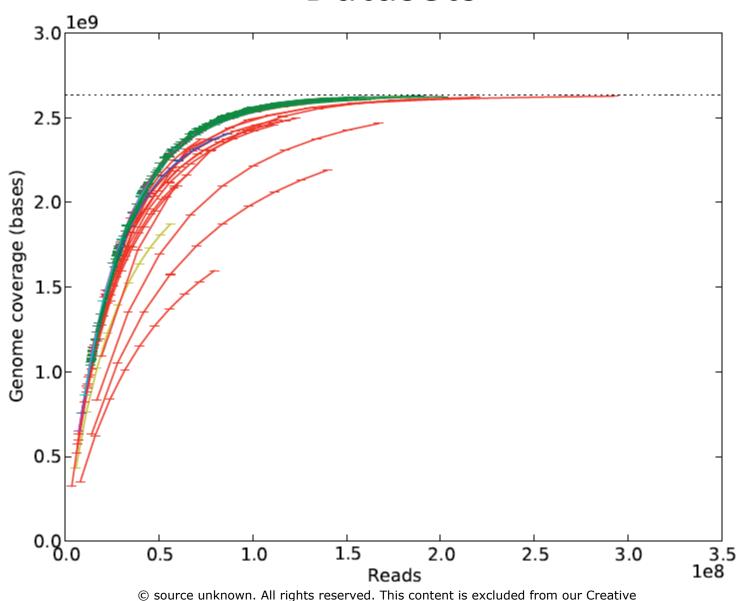
Average coverage = $177 / 35 \approx 7x$

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Estimating Uncovered Bases (Lander Waterman)

- G genome size
- N # of reads
- L Length of read
- NL/G = reads/base = λ (coverage)
 - -Poisson(0, λ) = $e^{-\lambda}$ =~ probability a base is not covered
 - -# of uncovered bases = $^{\sim}$ Ge^{- λ}
 - -# of gaps= \sim Ne^{- λ}

Reads vs. coverage for 1000 Genomes Datasets

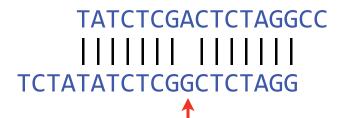


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Coverage could also refer to the number of reads covering a particular position in the genome:

```
CTAGGCCCTCAATTTTT
                 CTCTAGGCCCTCAATTTTT
               GGGTCTAGGCCCTCATTTTTT
           CTCGGCTCTAGCCCCTCATTTT
        TATCTCGACTCTAGGCCCTCA
        TATCTCGACTCTAGGCC
    TCTATATCTCGGCTCTAGG
GGCGTCTATATCTCG
GGCGTCGATATCT
GGCGTCTATATCT
GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT
        Coverage at this position = 6
```

Say two reads truly originate from overlapping stretches of the genome. Why might there be differences?



- 1. Sequencing error
- 2. Difference between inhereted *copies* of a chromosome E.g. humans are diploid; we have two copies of each chromosome, one from mother, one from father. The copies can differ:

Read from Mother: TATCTCGACTCTAGGCC

Read from Father: TCTATATCTCGGCTCTAGG

We'll mostly ignore ploidy, but real tools must consider it

Sequence from Mother: TCTATATCTCGACTCTAGGCC Sequence from Father: TCTATATCTCGGCTCTAGGCC

Courtesy of Ben Langmead. Used with permission.

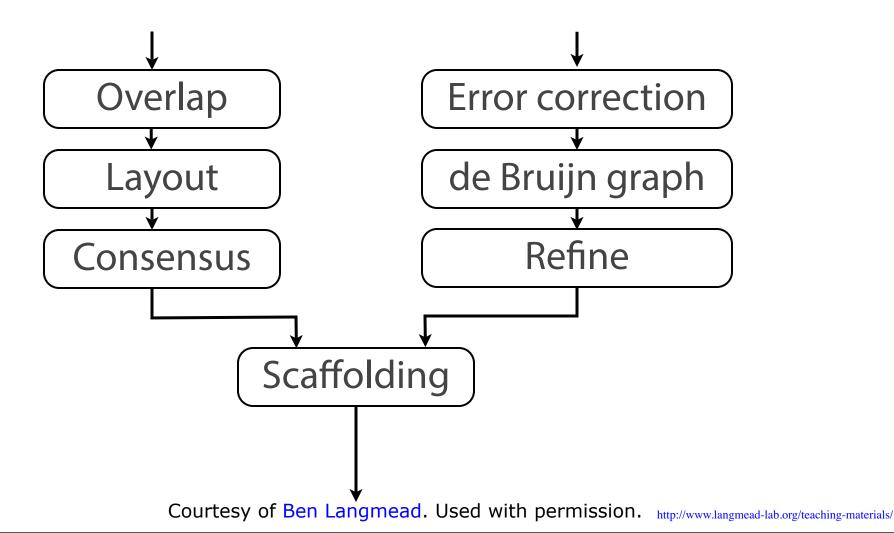
Two approaches to short read assembly

- Overlap Layout Consensus String Graph Assemblers
 - Construct overlap graph directly from reads, eliminating redundant reads; trace path for assembly
 - –Examples: SGA, Fermi
- de Bruijn graph-based assemblers
 - Construct k-mer graph from reads; original reads are discarded
 - Trace path in graph for assembly

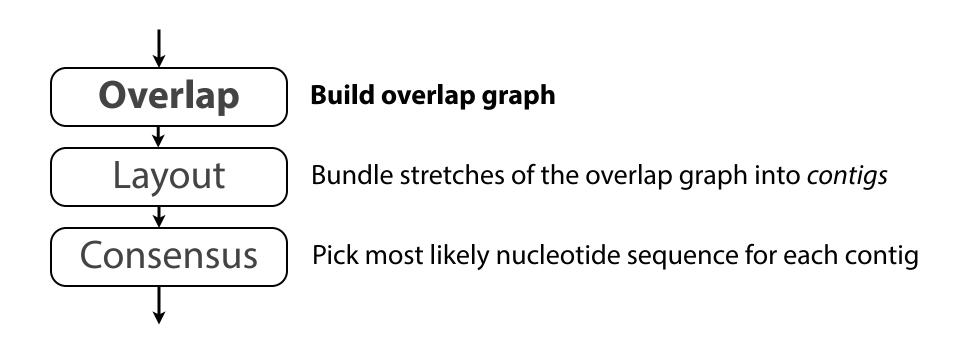
Assembly alternatives

Alternative 1: Overlap-Layout-Consensus (OLC) assembly

Alternative 2: de Bruijn graph (DBG) assembly

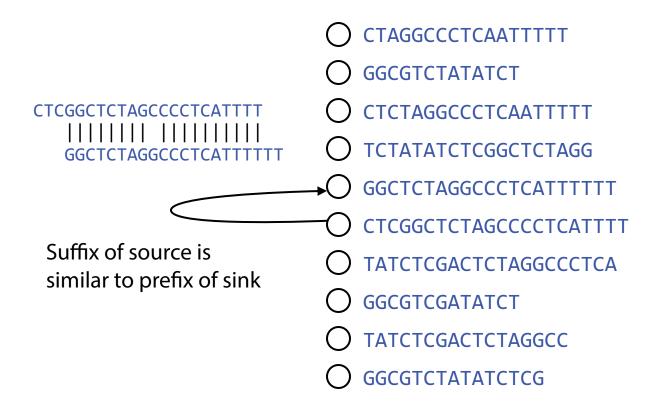


Overlap Layout Consensus



Overlaps

Finding all overlaps is like building a *directed graph* where directed edges connect overlapping nodes (reads)



Courtesy of Ben Langmead. Used with permission.

Directed graph review

Directed graph G(V, E) consists of set of vertices, V and set of directed edges, E

Directed edge is an *ordered pair* of vertices. First is the *source*, second is the *sink*.

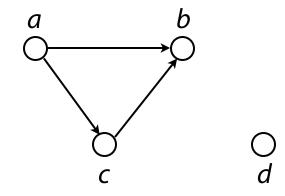
Vertex is drawn as a circle

Edge is drawn as a line with an arrow connecting two circles



Edge also called arc or line

Directed graph also called digraph



$$V = \{a, b, c, d\}$$

 $E = \{(a, b), (a, c), (c, b)\}$
Source Sink

Overlap graph

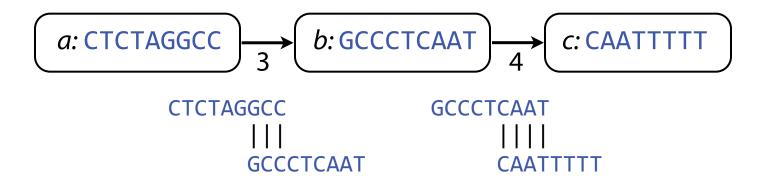
Below: overlap graph, where an overlap is a suffix/prefix match of at least 3 characters

A vertex is a **read**, a directed edge is an overlap between suffix of source and prefix of sink

```
Vertices (reads): { a: CTCTAGGCC, b: GCCCTCAAT, c: CAATTTTT }
```

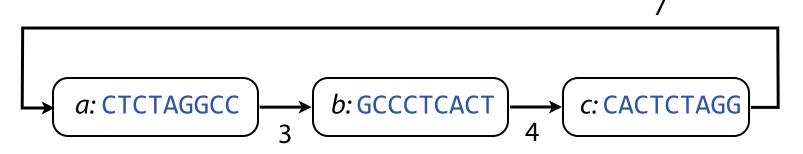
Edges (overlaps): { (*a*, *b*), (*b*, *c*) }

To keep our presentation uncluttered we will not show the treatment of read reverse complements



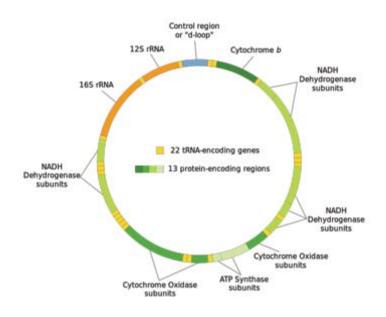
Overlap graph

Overlap graph could contain *cycles*. A cycle is a path beginning and ending at the same vertex.



These happen when the DNA string itself is circular. E.g. bacterial genomes are often circular; mitochondrial DNA is circular.

Cycles could also be due to *repetitive* DNA, as we'll see



Finding overlaps



How do we build the overlap graph?

Assume for now an "overlap" is when a suffix of X of length $\geq l$ exactly matches a prefix of Y, and k is the length of reads

Index: CTCTAGGCC\$GCCCTCAAT\$CAATTTTT\$\$

A merged FM index of all reads allows us to match read prefixes and suffixes to discover overlaps. See SGA paper -

Efficient de novo assembly of large genomes using compressed data structures Jared T Simpson and Richard Durbin Genome Res. 2012. 22: 549–556

SGA algorithm excludes redundant (transitive) edges A -> B -> C excludes A -> C

Finding overlaps

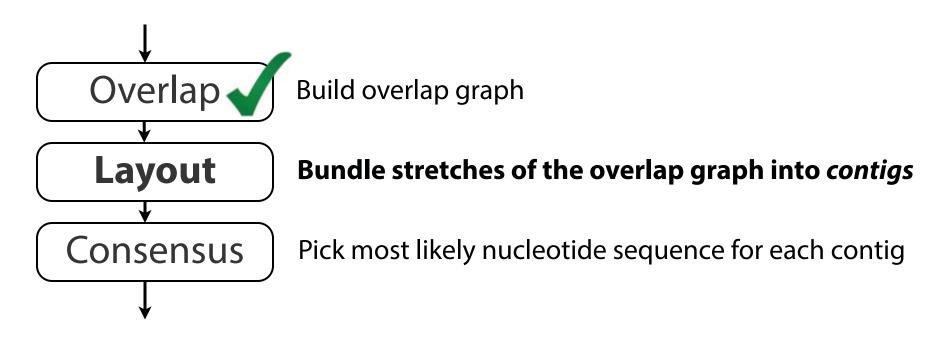
Example overlap graph with l=3 overlap length k=7ACGGCGC 3 GCCGTAC 5 GTACGGC

ACGGCGC 3 GCCGCTA GCCCGCTA

ATTATAT 5 GATTATTG 4 ATTGCC 3 CGCCGCTA

Original string: GCATTATATTTGCGCGTACGGCGCCGCTACA

Overlap Layout Consensus



Formulating the assembly problem

Finding overlaps is important, and we'll return to it, but our ultimate goal is to recreate (assemble) the genome

How do we formulate this problem?

First attempt: the shortest common superstring (SCS) problem

Given a collection of strings *S*, find *SCS*(*S*): the shortest string that contains all strings in *S* as substrings

Without requirement of "shortest," it's easy: just concatenate them

Example: S: BAA AAB BBA ABA ABB BBB AAA BAB

Concatenation: BAAAABBBBAABBBBBAAABAB

______24 _______

SCS(S): AAABBBABAA

10 ──

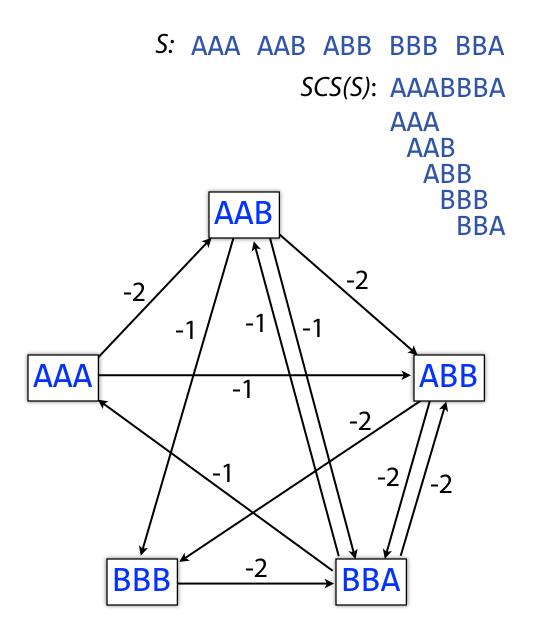
AAA
AAB
ABB
BBB
BBA
BAB
ABA
RAA

Can we solve it?

Imagine a modified overlap graph where each edge has cost = - (length of overlap)

SCS corresponds to a path that visits every node once, minimizing total cost along path

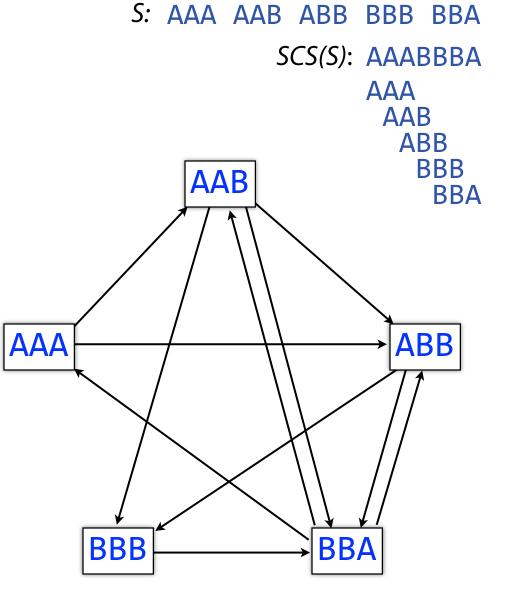
That's the *Traveling Salesman Problem (TSP)*, which is NP-hard!



Say we disregard edge weights and just look for a path that visits all the nodes exactly once

That's the *Hamiltonian Path* problem: NP-complete

Indeed, it's well established that SCS is NP-hard



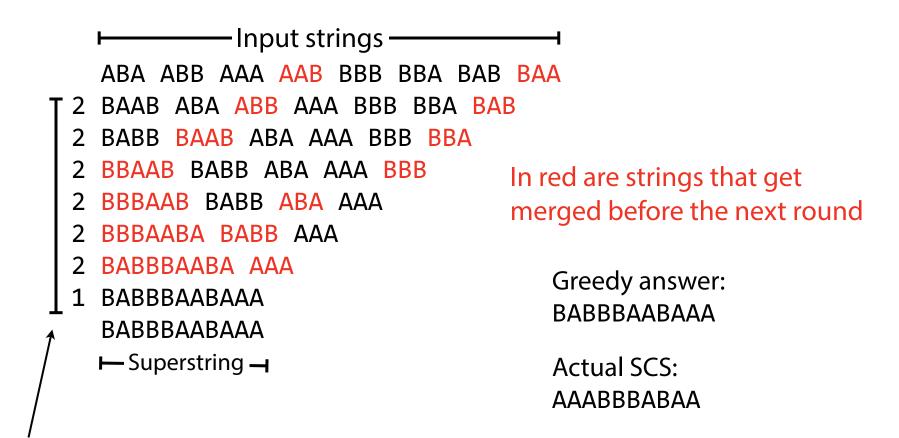
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Let's take the hint give up on finding the shortest possible superstring

Non-optimal superstrings can be found with a *greedy* algorithm

At each step, the greedy algorithm "greedily" chooses longest remaining overlap, merges its source and sink

Greedy-SCS algorithm in action (l = 1):



Rounds of merging, one merge per line.

Number in first column = length of overlap merged before that round.

Greedy algorithm is *not* guaranteed to choose overlaps yielding SCS

But greedy algorithm is a good *approximation*; i.e. the superstring yielded by the greedy algorithm won't be more than ~2.5 times longer than true SCS (see Gusfield, Algorithms on Strings, Trees and Sequences: Computer Science and Computational Biology, 16.17.1)

Greedy-SCS algorithm in action again (l = 3):

———Input strings ——

ATTATAT CGCGTAC ATTGCGC GCATTAT ACGGCGC TATATTG GTACGGC GCGTACG ATATTGC

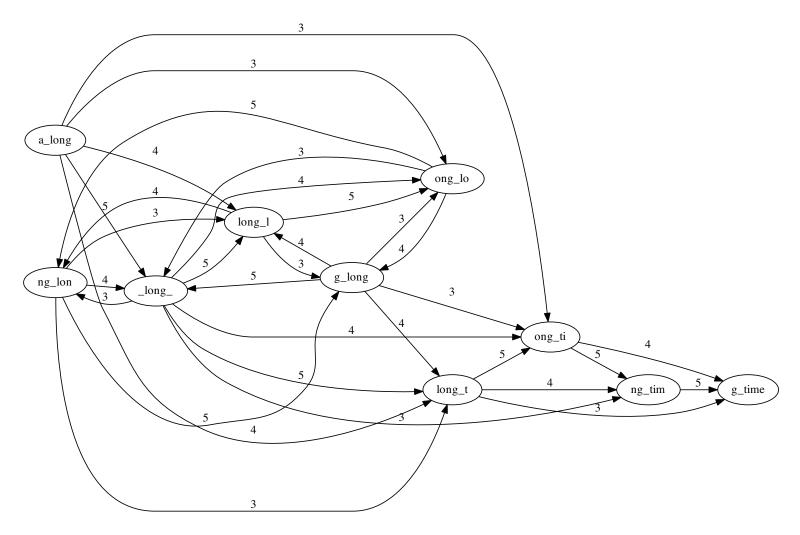
- 6 TATATTGC ATTATAT CGCGTAC ATTGCGC GCATTAT ACGGCGC GTACGGC GCGTACG
- 6 CGCGTACG TATATTGC ATTATAT ATTGCGC GCATTAT ACGGCGC GTACGGC
- 5 CGCGTACG TATATTGCGC ATTATAT GCATTAT ACGGCGC GTACGGC
- 5 CGCGTACGGC TATATTGCGC ATTATAT GCATTAT ACGGCGC
- 5 CGCGTACGGCGC TATATTGCGC ATTATAT GCATTAT
- 5 CGCGTACGGCGC GCATTATAT TATATTGCGC
- 5 CGCGTACGGCGC GCATTATATTGCGC
- 3 GCATTATATTGCGCGTACGGCGC GCATTATATTGCGCGTACGGCGC

——Superstring ——

Another setup for Greedy-SCS: assemble all substrings of length 6 from string a long long long time. l = 3.

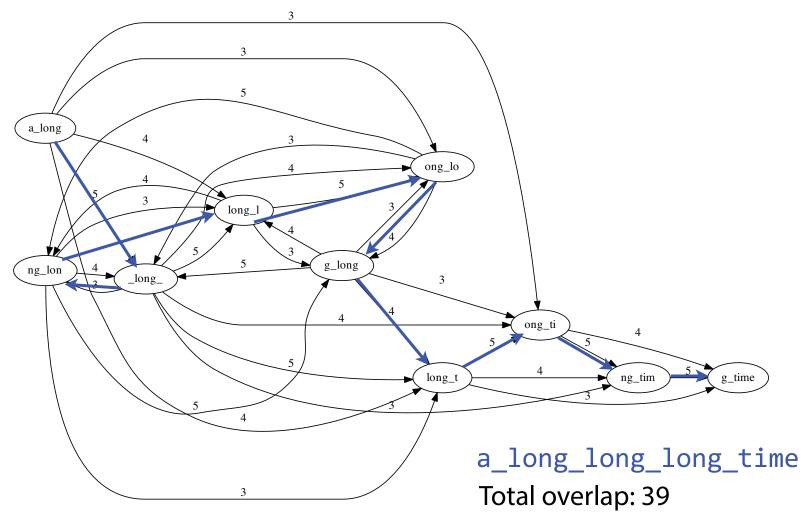
```
ng lon long a long long long ti ong lo long t g long g time ng tim
5 ng_time ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long
5 ng_time g_long_ ng_lon a_long long_l ong_ti ong_lo long_t
5 ng time long ti g long ng lon a long long long lo
5 ng_time ong_lon long_ti g_long_ a_long long_l
5 ong lon long time g long a long long l
5 long lon long time g long a long
5 long lon g long time a long
5 long long time a long
4 a_long_long_time
 a long long time
   We only got back: a_long_long_time (missing a _long)
   What happened?
```

The overlap graph for that scenario (l = 3):



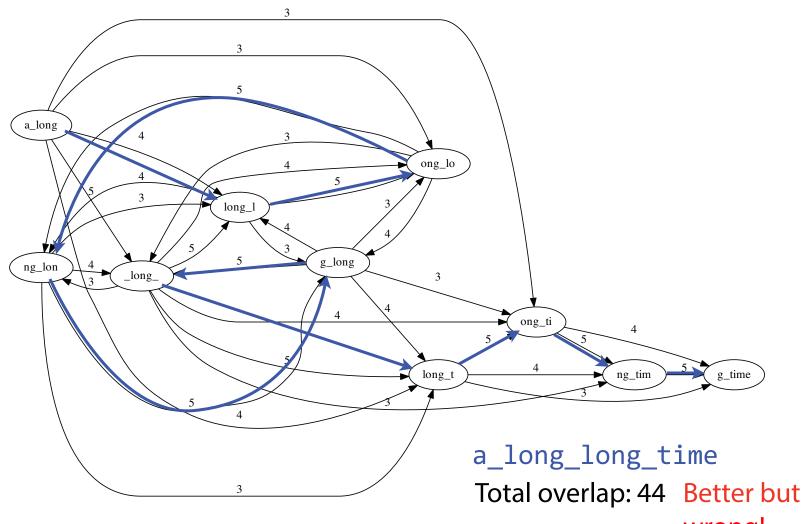
Courtesy of Ben Langmead. Used with permission.

The overlap graph for that scenario (l = 3):



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The overlap graph for that scenario (l = 3):



Same example, but increased the substring length from 6 to 8

Got the whole thing: a_long_long_time

Why are substrings of length 8 long enough for Greedy-SCS to figure out there are 3 copies of long?

One length-8 substring spans all three longs

Repeats

Repeats often foil assembly. They certainly foil SCS, with its "shortest" criterion!

Reads might be too short to "resolve" repetitive sequences. This is why sequencing vendors try to increase read length.

Algorithms that don't pay attention to repeats (like our greedy SCS algorithm) might *collapse* them

The human genome is ~ 50% repetitive!

Repeats

Basic principle: repeats foil assembly

Another example using Greedy-SCS:

```
Input: it_was_the_best_of_times_it_was_the_worst_of_times
```

Extract every substring of length k, then run Greedy-SCS. Do this for various l (min overlap length) and k.

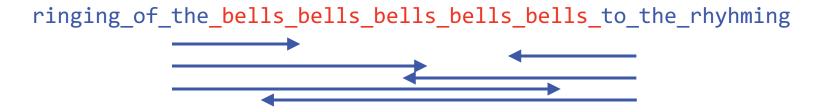
```
l,k
output
3,5 the_worst_of_times_it_was_the_best_o
3,7 s_the_worst_of_times_it_was_the_best_of_t
3,10 _was_the_best_of_times_it_was_the_worst_of_tim
3,13 it_was_the_best_of_times_it_was_the_worst_of_times
```

Repeats

Basic principle: repeats foil assembly

Longer and longer substrings allow us to "anchor" more of the repeat to its non-repetitive context:

Often we can "walk in" from both sides. When we meet in the middle, the repeat is resolved:



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Repeats

Basic principle: repeats foil assembly

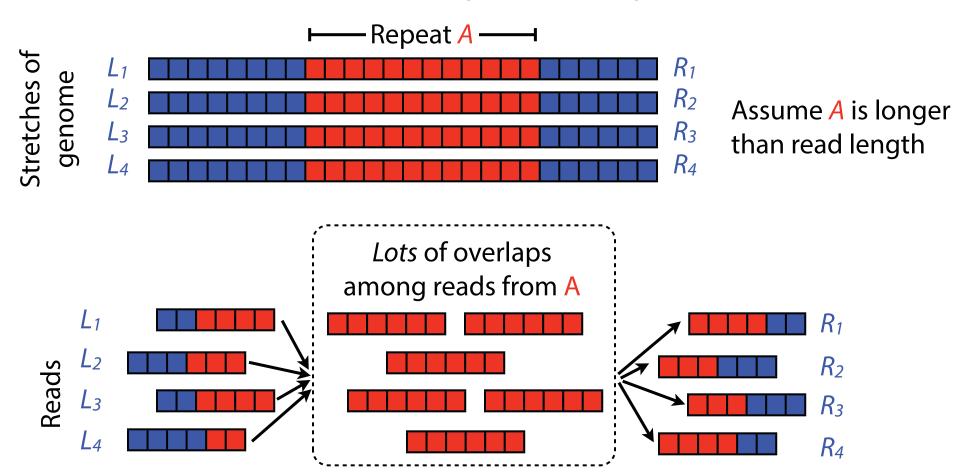
Yet another example using Greedy-SCS:

```
Input: swinging_and_the_ringing_of_the_bells_bells_bells_bells_bells
```

longer and longer substrings allow us to "reach" further into the repeat

Repeats

Picture the portion of the overlap graph involving repeat A



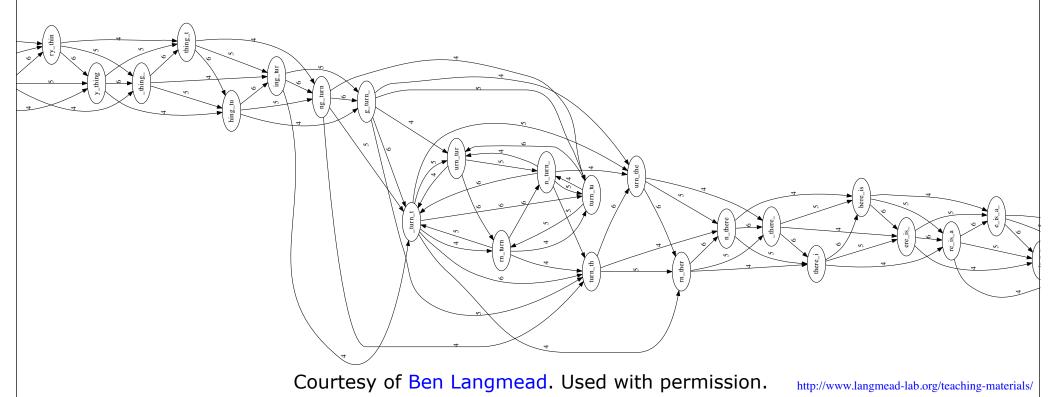
Even if we avoid collapsing copies of A, we can't know which paths in correspond to which paths out

Courtesy of Ben Langmead. Used with permission.

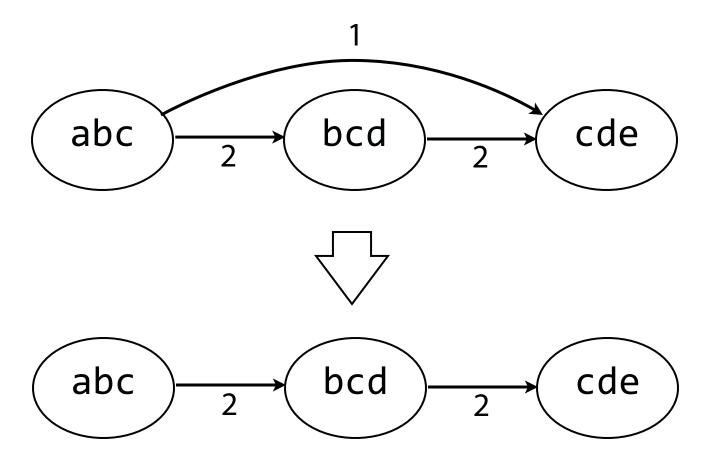
The overlap graph is big and messy. Contigs don't "pop out" at us.

Below: part of the overlap graph for

to_every_thing_turn_turn_turn_there_is_a_season
$$l=4, k=7$$

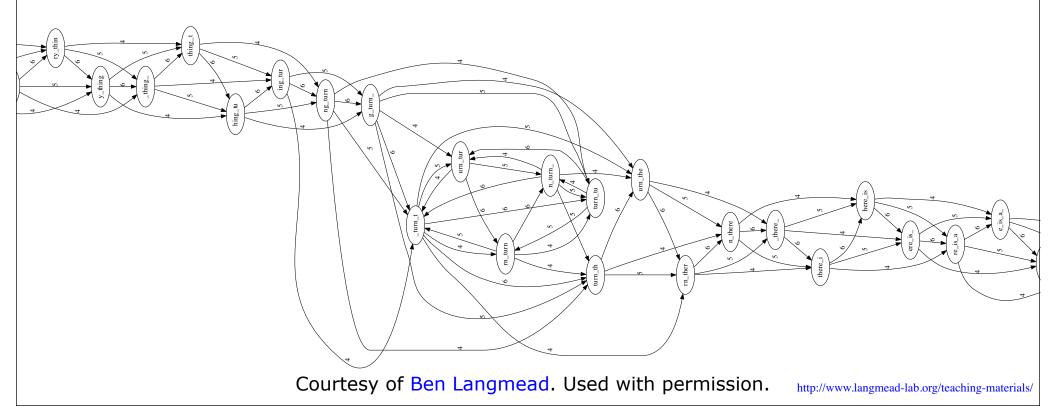


Picture gets clearer after removing some transitively-inferrible edges



Remove transitively-inferrible edges, starting with edges that skip one node:

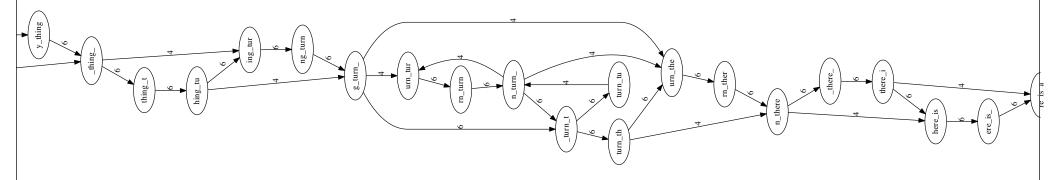
Before:



Remove transitively-inferrible edges, starting with edges that skip one

node:

After:

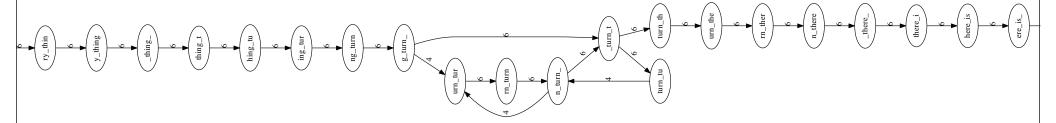


Remove transitively-inferrible edges, starting with edges that skip one

or two nodes:

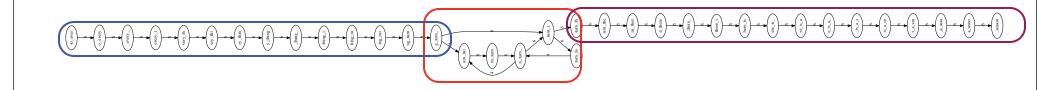


After:



Even simpler

Emit *contigs* corresponding to the non-branching stretches

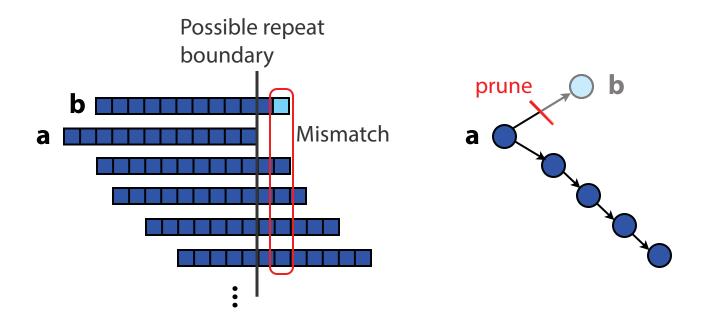


Contig 1

to_every_thing_turn_ turn_there_is_a_season

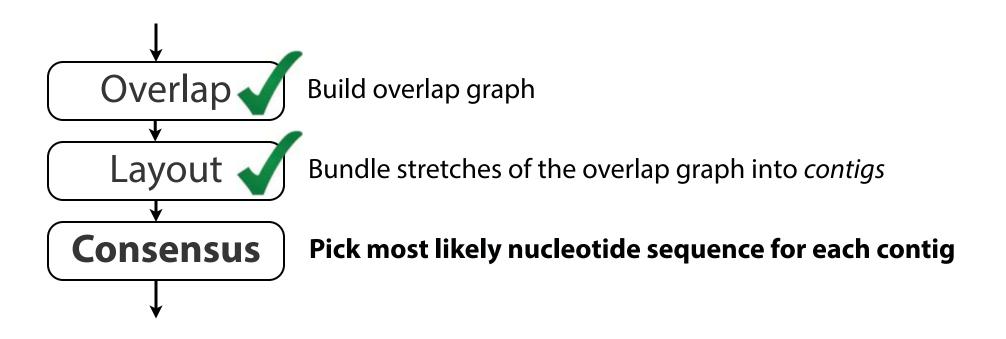
Unresolvable repeat

In practice, layout step also has to deal with spurious subgraphs, e.g. because of sequencing error



Mismatch could be due to sequencing error or repeat. Since the path through **b** ends abruptly we might conclude it's an error and prune **b**.

Overlap Layout Consensus



Consensus

TAGATTACACAGATTACTGA TTGATGGCGTAA CTA
TAGATTACACAGATTACTGACTTGATGGCGTAAACTA
TAG TTACACAGATTATTGACTTCATGGCGTAA CTA
TAGATTACACAGATTACTGACTTGATGGCGTAA CTA
TAGATTACACAGATTACTGACTTGATGGCGTAA CTA

Take reads that make up a contig and line them up

TAGATTACACAGATTACTGACTTGATGGCGTAA CTA

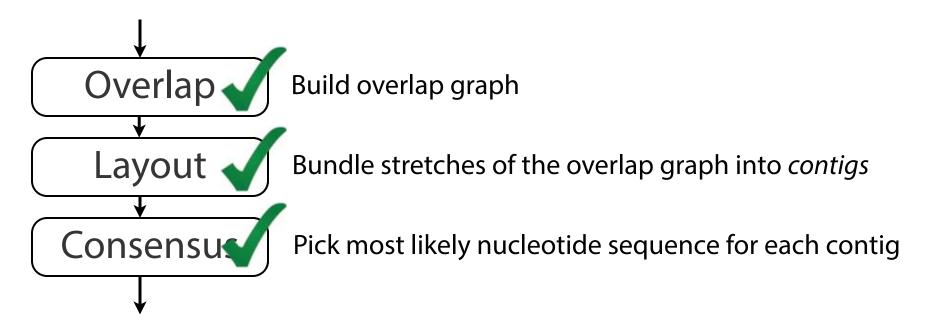
Take *consensus*, i.e. majority vote

At each position, ask: what nucleotide (and/or gap) is here?

Complications: (a) sequencing error, (b) ploidy

Say the true genotype is AG, but we have a high sequencing error rate and only about 6 reads covering the position.

Overlap Layout Consensus



What's the main drawback of OLC?

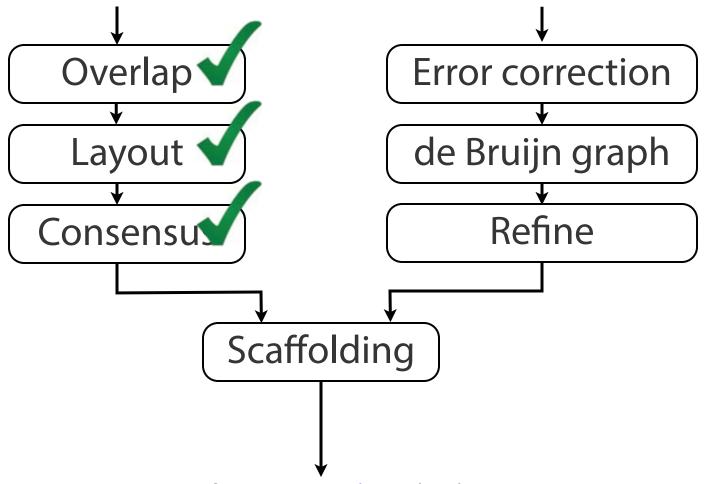
Building overlap graph can be slow.

2nd-generation sequencing datasets are ~ 100s of millions or billions of reads, hundreds of billions of nucleotides total

Assembly alternatives

Alternative 1: Overlap-Layout-Consensus (OLC) assembly

Alternative 2: de Bruijn graph (DBG) assembly



Courtesy of Ben Langmead. Used with permission.



Table removed due to copyright restrictions.

SGA contigs cover 95% of autosomes and chr X (non "N" bases) NA12878 1.2 x 10⁹ reads 40x coverage

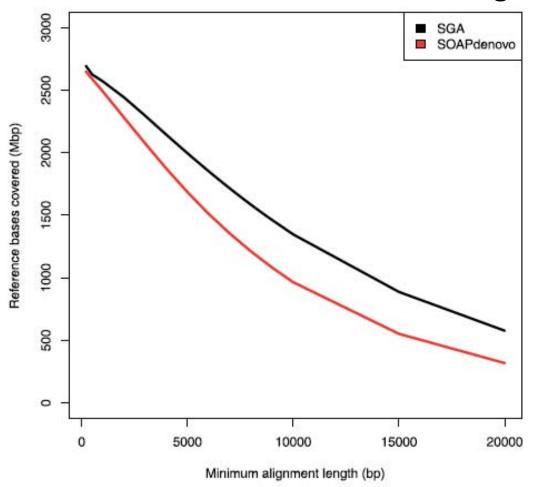


Figure 3. The amount of the human reference genome covered by a contig as a function of the minimum contig alignment length. For each length L on the x-axis, contig alignments less than L bp in length were filtered out and the amount of the reference genome covered by the remaining alignments was calculated.

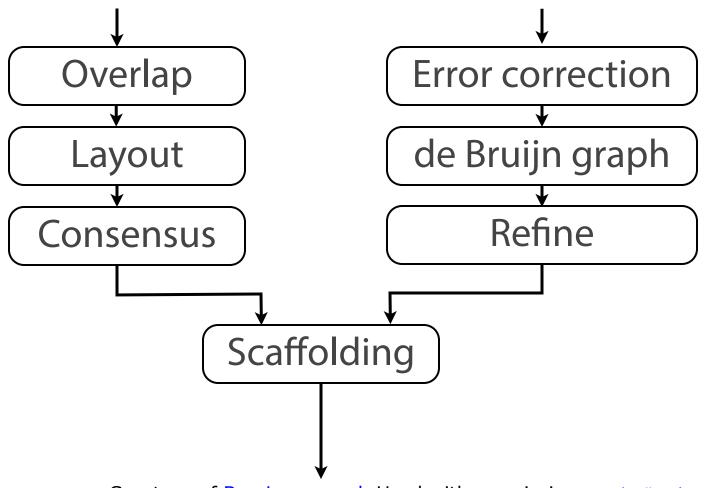
Courtesy of Cold Spring Harbor Laboratory Press. Used with permission.

Source: Simpson, Jared T., and Richard Durbin. "Efficient De Novo Assembly of Large Genomes using Compressed Data Structures." *Genome Research* 22, no. 3 (2012): 549-56.

Assembly alternatives

Alternative 1: Overlap-Layout-Consensus (OLC) assembly

Alternative 2: de Bruijn graph (DBG) assembly



Courtesy of Ben Langmead. Used with permission.

De Bruijn graph assembly

A formulation conceptually similar to overlapping/SCS, but has some potentially helpful properties not shared by SCS.

k-mer

"k-mer" is a substring of length k

S: GGCGATTCATCG

mer: from Greek meaning "part"

A 4-mer of *S*: ATTC

All 3-mers of *S*: GGC

GCG
CGA
GAT
ATT
TTC
TCA
CAT
ATC
TCG

I'll use "k-1-mer" to refer to a substring of length k - 1

As usual, we start with a collection of reads, which are substrings of the reference genome.

AAA, AAB, ABB, BBB, BBA

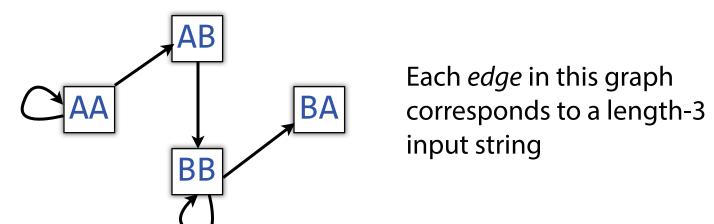
AAB is a k-mer (k = 3). AA is its left k-1-mer, and AB is its right k-1-mer.

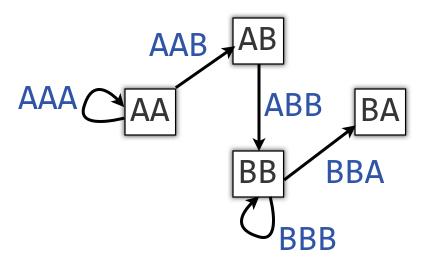
Take each length-3 input string and split it into two overlapping substrings of length 2. Call these the *left* and *right 2-mers*.

AAABBBA

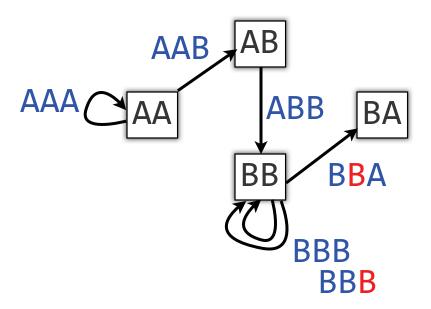
take all 3-mers: AAA, AAB, ABB, BBB, BBA

Let 2-mers be nodes in a new graph. Draw a directed edge from each left 2-mer to corresponding right 2-mer:





An edge corresponds to an overlap (of length k-2) between two k-1 mers. More precisely, it corresponds to a k-mer from the input.



If we add one more B to our input string: AAABBBBA, and rebuild the De Bruijn graph accordingly, we get a *multiedge*.

Eulerian walk definitions and statements

Node is balanced if indegree equals outdegree

Node is semi-balanced if indegree differs from outdegree by 1

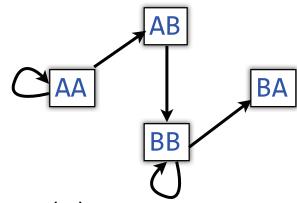
Graph is *connected* if each node can be reached by some other node

Eulerian walk visits each edge exactly once

Not all graphs have Eulerian walks. Graphs that do are *Eulerian*. (For simplicity, we won't distinguish Eulerian from semi-Eulerian.)

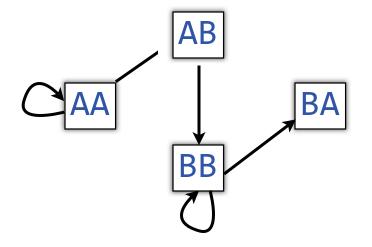
A directed, connected graph is Eulerian if and only if it has at most 2 semi-balanced nodes and all other nodes are balanced

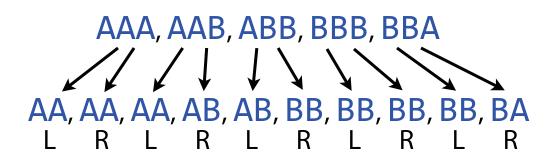
Jones and Pevzner section 8.8



Courtesy of Ben Langmead. Used with permission.

Back to our De Bruijn graph





Is it Eulerian? Yes

Argument 1: $AA \rightarrow AA \rightarrow AB \rightarrow BB \rightarrow BB \rightarrow BA$

Argument 2: AA and BA are semi-balanced, AB and BB are balanced

A procedure for making a De Bruijn graph for a genome

Assume *perfect sequencing* where each length-*k* substring is sequenced exactly once with no errors

Pick a substring length *k*: 5

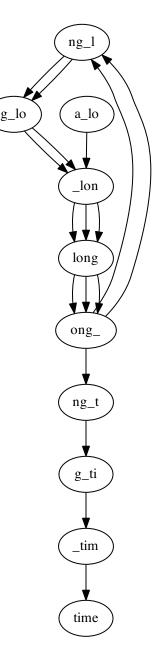
Start with each read:

a_long_long_time

Take each *k* mer and split into left and right *k*-1 mers

long ong_

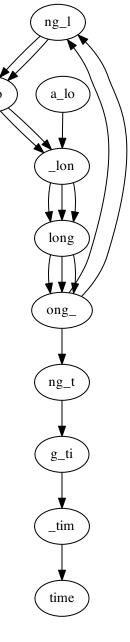
Add k-1 mers as nodes to De Bruijn graph (if not already there), add edge from left k-1 mer to right k-1 mer

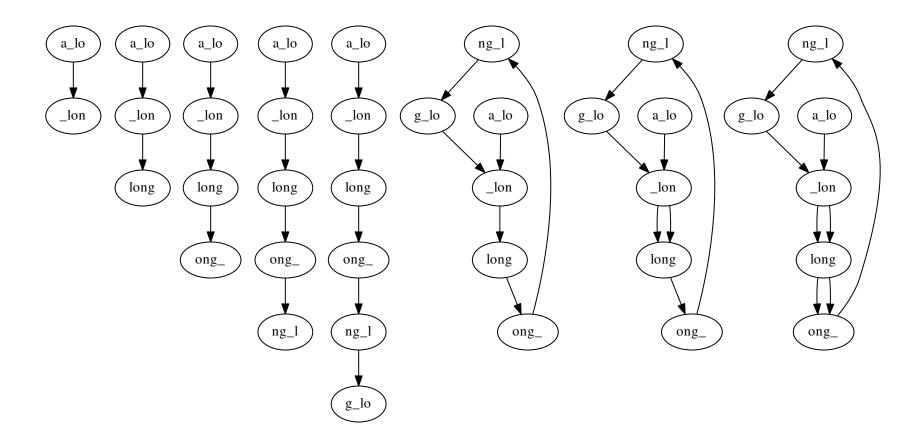


Courtesy of Ben Langmead. Used with permission.

• For genome assembly each k-mer is recorded in "twin" nodes – one node in the forward direction and one node in reverse complement

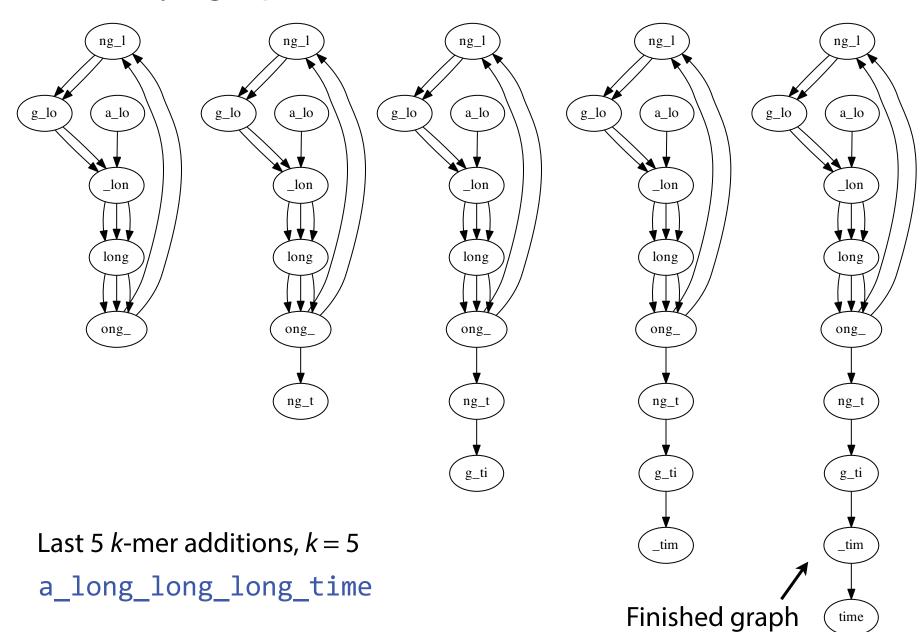
- k is odd so no node can be its own reverse complement
- We will not show reverse complement twin nodes to cut down on clutter





First 8 k-mer additions, k = 5

a_long_long_time



Courtesy of Ben Langmead. Used with permission.

With perfect sequencing, this procedure always yields an Eulerian graph. Why?

Node for *k*-1-mer from left end is semi-balanced with one more outgoing edge than incoming *

Node for *k*-1-mer at right end is semi-balanced with one more incoming than outgoing *

Other nodes are balanced since # times k-1-mer occurs as a left k-1-mer = # times it occurs as a right k-1-mer

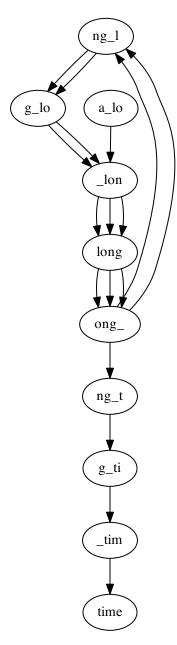
* Unless genome is circular

Courtesy of Ben Langmead. Used with permission.

g_lo a lo lon long ong_ ng_t g_ti tim

Assuming perfect sequencing, procedure yields graph with Eulerian walk that can be found efficiently.

We saw cases where Eulerian walk corresponds to the original superstring. Is this always the case?



No: graph can have multiple Eulerian walks, only one of which corresponds to original superstring

Right: graph for ZABCDABEFABY, k = 3

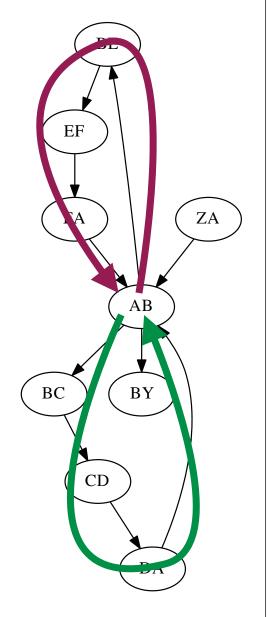
Alternative Eulerian walks:

$$ZA \rightarrow AB \rightarrow BE \rightarrow EF \rightarrow FA \rightarrow AB \rightarrow BC \rightarrow CD \rightarrow DA \rightarrow AB \rightarrow BY$$

$$ZA \rightarrow AB \rightarrow BC \rightarrow CD \rightarrow DA \rightarrow AB \rightarrow BE \rightarrow EF \rightarrow FA \rightarrow AB \rightarrow BY$$

These correspond to two edge-disjoint directed cycles joined by node AB

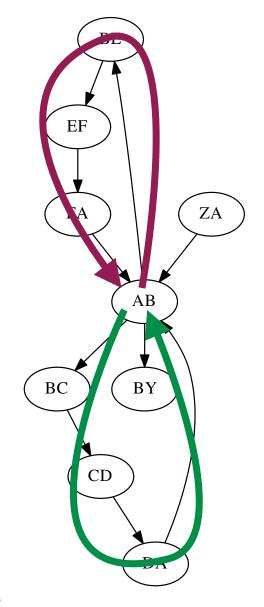
AB is a repeat: ZABCDABEFABY



Courtesy of Ben Langmead. Used with permission.

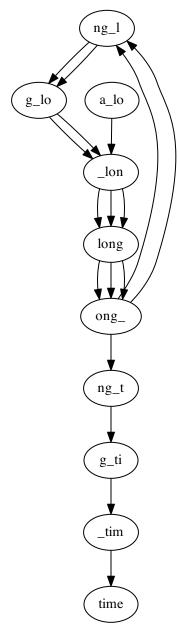
This is the first sign that Eulerian walks can't solve all our problems

Other signs emerge when we think about how actual sequencing differs from our idealized construction



Gaps in coverage can lead to disconnected graph

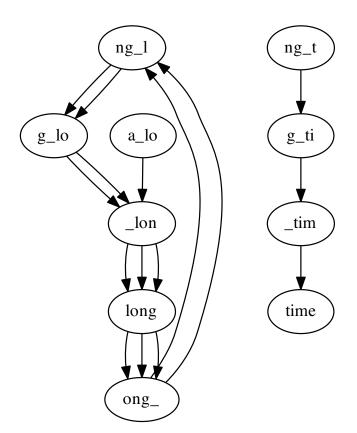
Graph for a_long_long_long_time, k = 5:



Courtesy of Ben Langmead. Used with permission.

Gaps in coverage can lead to disconnected graph

Graph for a_long_long_time, k = 5 but omitting ong_t:

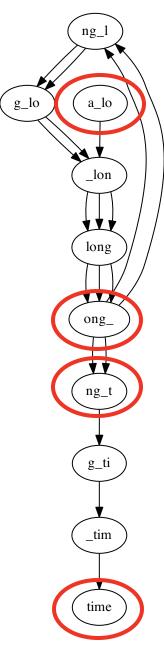


Connected components are individually Eulerian, overall graph is not

Differences in coverage also lead to non-Eulerian graph

Graph for a_long_long_long_time, k = 5 but with extra copy of ong_t:

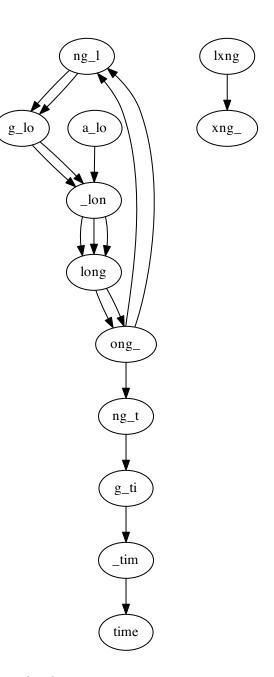
Graph has 4 semi-balanced nodes, isn't Eulerian



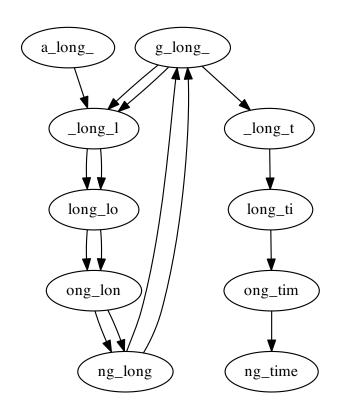
Errors and differences between chromosomes also lead to non-Eulerian graphs

Graph for a_long_long_long_time, k = 5 but with error that turns a copy of long_ into lxng_

Graph is not connected; largest component is not Eulerian



Courtesy of Ben Langmead. Used with permission.



How much work to build graph?

For each k-mer, add 1 edge and up to 2 nodes

Reasonable to say this is O(1) expected work

Assume hash map encodes nodes & edges

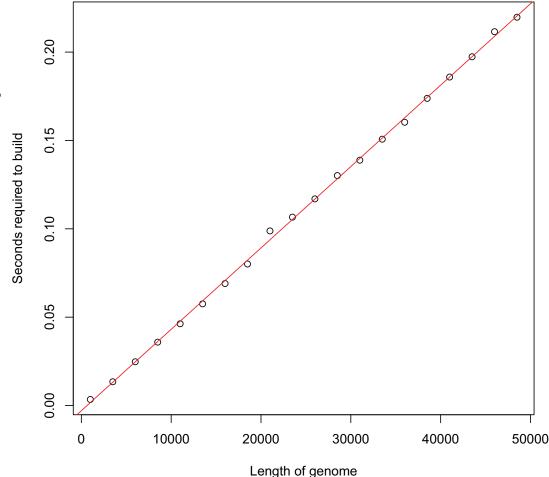
Assume k-1-mers fit in O(1) machine words, and hashing O(1) machine words is O(1) work

Querying / adding a key is O(1) expected work

O(1) expected work for 1 k-mer, O(N) overall

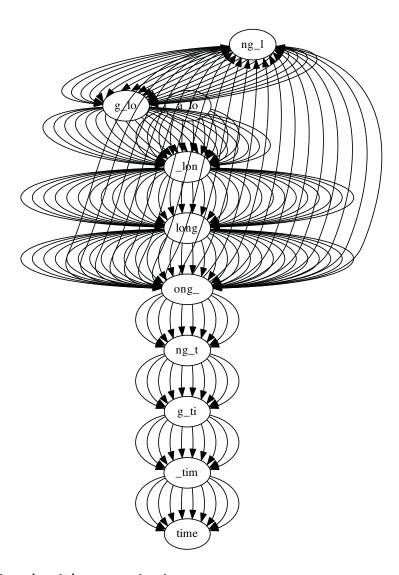
Timed De Bruijn graph construction applied to progressively longer prefixes of lambda phage genome, k = 14

O(N) expectation appears to work in practice, at least for this small example



Courtesy of Ben Langmead. Used with permission.

In typical assembly projects, average coverage is ~ 30 - 50

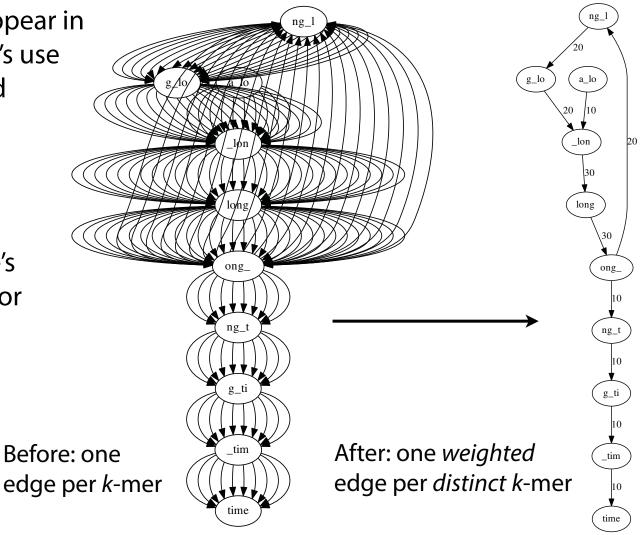


In typical assembly projects, average coverage is ~ 30 - 50

Same edge might appear in dozens of copies; let's use edge *weights* instead

Weight = # times k-mer occurs

Using weights, there's one weighted edge for each distinct k-mer



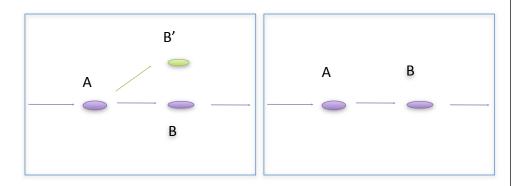
Courtesy of Ben Langmead. Used with permission.

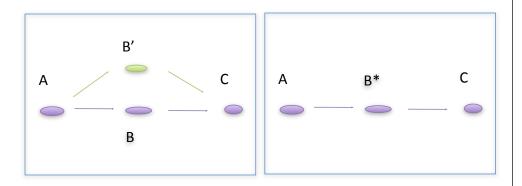
Graph topology based error eorrection

- -Errors at end of read
 - Trim off 'dead-end' tips

- –Errors in middle of read
 - Pop Bubbles

- –Chimeric Edges
 - Clip short, low coverage nodes





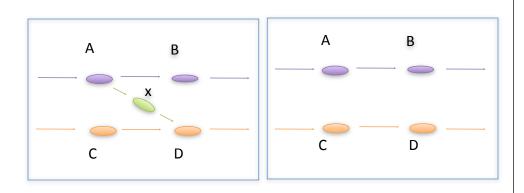


Figure adapted from presentation by Michael Schatz

What are the limitations of De Bruijn graphs?

Reads are immediately split into shorter k-mers; can't resolve repeats as well as overlap graph

Only a very specific type of "overlap" is considered, which makes dealing with errors more complicated.

Read coherence is lost. Some paths through De Bruijn graph are inconsistent with respect to input reads. Need to thread reads though De Bruijn graph to recover information lost when reads are fragmented into k-mers.

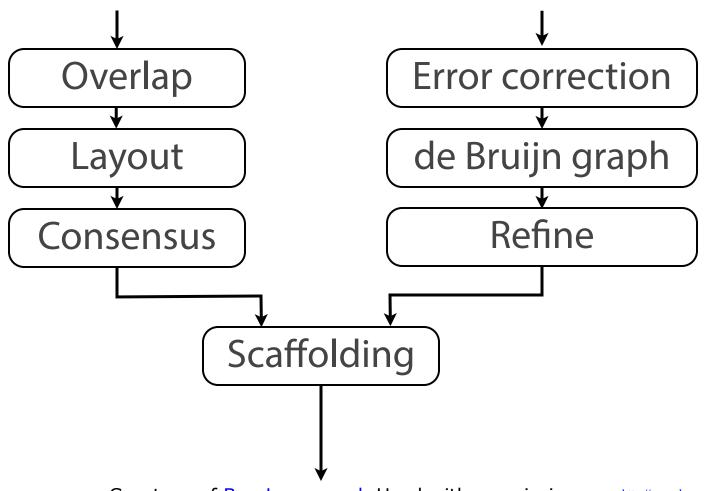
This is the OLC \leftrightarrow DBG tradeoff

Single most important benefit of De Bruijn graph is speed and simplicity.

Assembly alternatives

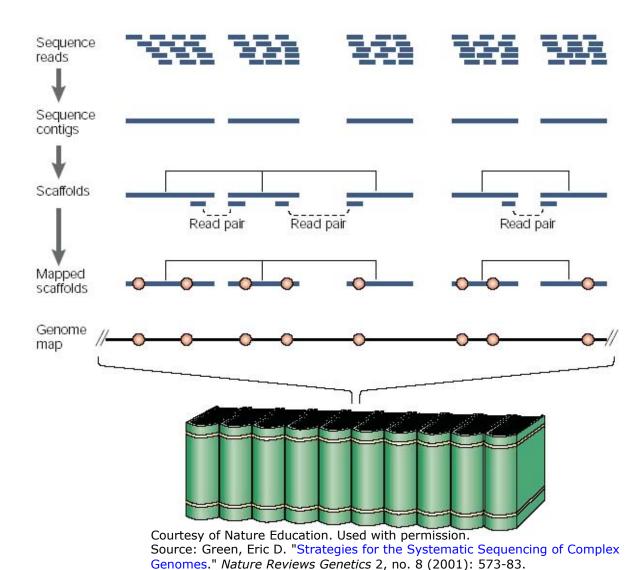
Alternative 1: Overlap-Layout-Consensus (OLC) assembly

Alternative 2: de Bruijn graph (DBG) assembly



Courtesy of Ben Langmead. Used with permission.

de novo whole-genome shotgun assembly



N50 - contig/scaffold length or larger that contains 50% of bases

OLC De Bruijn De Bruijn De Bruijn k = 61 k = 67 k = 59

Table 1. Assembly statistics for C. elegans data set

	SGA	Velvet	ABySS	SOAPdenovo
Scaffold N50 size	26.3 kbp	31.3 kbp	23.8 kbp	31.1 kbp
Aligned contig N50 size	16.8 kbp	13.6 kbp	18.4 kbp	16.0 kbp
Mean aligned contig size	4.9 kbp	5.3 kbp	6.0 kbp	5.6 kbp
Sum aligned contig size	96.8 Mbp	95.2 Mbp	98.3 Mbp	95.4 Mbp
Reference bases covered	96.2 Mbp	94.8 Mbp	95.9 Mbp	95.1 Mbp
Reference bases covered by contigs ≥1 kb	93.0 Mbp	92.1 Mbp	93.9 Mbp	92.3 Mbp
Mismatch rate at all assembled bases	1 per 21,545 bp	1 per 8786 bp	1 per 5577 bp	1 per 26,585 bp
Mismatch rate at bases covered by all assemblies	1 per 82,573 bp	1 per 18,012 bp	1 per 8209 bp	1 per 81,025 bp
Contigs with split/bad alignment (sum size)	458 (4.4 Mbp)	787 (7.2 Mbp)	638 (9.1 Mbp)	483 (4.4 Mbp)
Total CPU time	41 h	2 h	5 h	13 h
Max memory usage	4.5 GB	23.0 GB	14.1 GB	38.8 GB

100 MBase genome, 33.8M read pairs, 100bp reads each end, 250bp insert size

Efficient de novo assembly of large genomes using compressed data structures Jared T Simpson and Richard Durbin Genome Res. 2012. 22: 549-556



