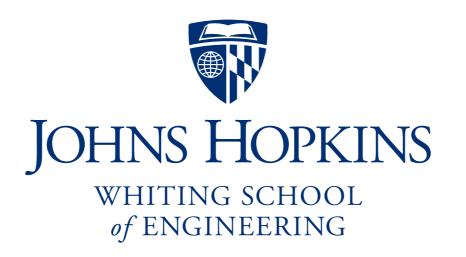
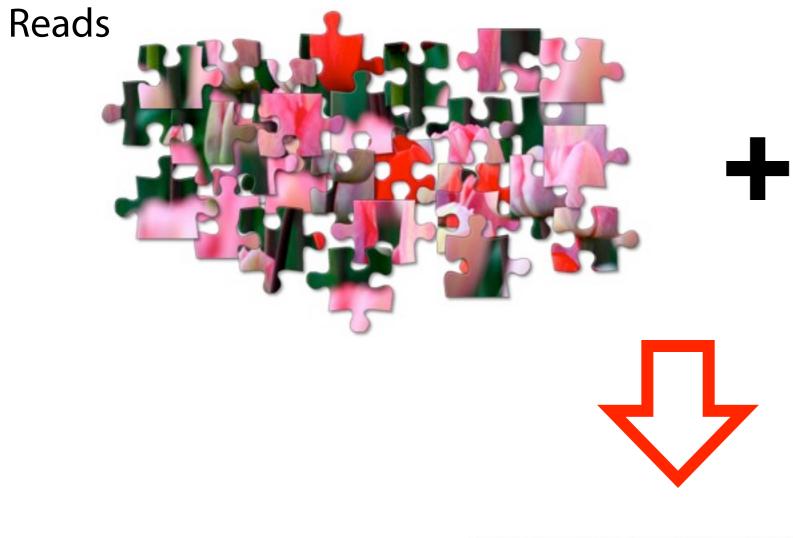
## Assembly & shortest common superstring

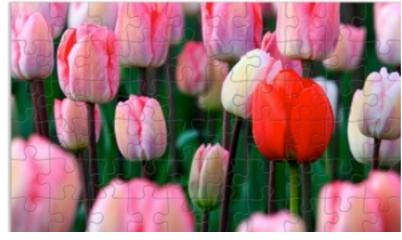
**Ben Langmead** 



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Input DNA



Reference genome



How to assemble puzzle without the benefit of knowing what the finished product looks like?

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

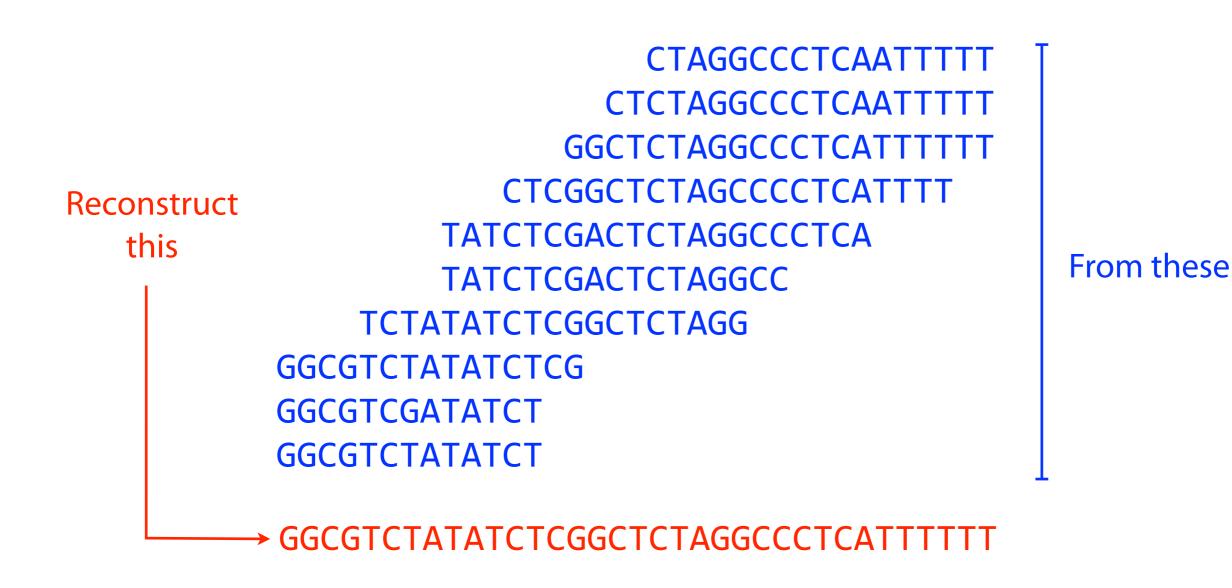
("Shotgun" refers to the random fragmentation of the whole genome; like it was fired from a shotgun)

Input: GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT

Copy: GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT

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...



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

Reconstruct this CTAGGCCCTCAATTTTT GGCGTCTATATCT CTCTAGGCCCTCAATTTTT TCTATATCTCGGCTCTAGG GGCTCTAGGCCCTCATTTTTT CTCGGCTCTAGCCCCTCATTTTT TATCTCGACTCTAGGCCCTCA GGCGTCGATATCT TATCTCGACTCTAGGCC GGCGTCTATATCTCG

From these

→ GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT

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

CTAGGCCCTCAATTTTT CTCTAGGCCCTCAATTTTT GGCTCTAGGCCCTCATTTTTT CTCGGCTCTAGGCCCTCATTTTT TATCTCGACTCTAGGCCCTCA TATCTCGACTCTAGGCC TCTATATCTCGGCTCTAGG GGCGTCTATATCTCG GGCGTCTATATCTCG GGCGTCTATATCT GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT 35 nucleotides

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

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

CTAGGCCCTCAATTTT **CTAGGCCCTCAATTTT GGCTCTAGGCCCTCATTTTT** CTCGGCTCTAGCCCCTCATTTT TATCTCGACTCTAGGCCCTCA TATCTCGACTCTAGGCC **TCTATATCTCGGCTCTAGG** GGCGTCTATATCTCG GGCGTCGATATCT **GGCGTCTATATCT** GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT Coverage at this position = 6

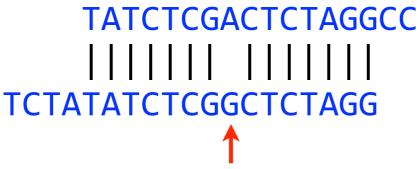
Basic principle: the more similarity there is between the end of one read and the beginning of another...

#### TATCTCGACTCTAGGCC |||||||||| TCTATATCTCGGCTCTAGG

...the more likely they are to have originated from overlapping stretches of the genome:

TATCTCGACTCTAGGCC TCTATATCTCGGCTCTAGG GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT

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:

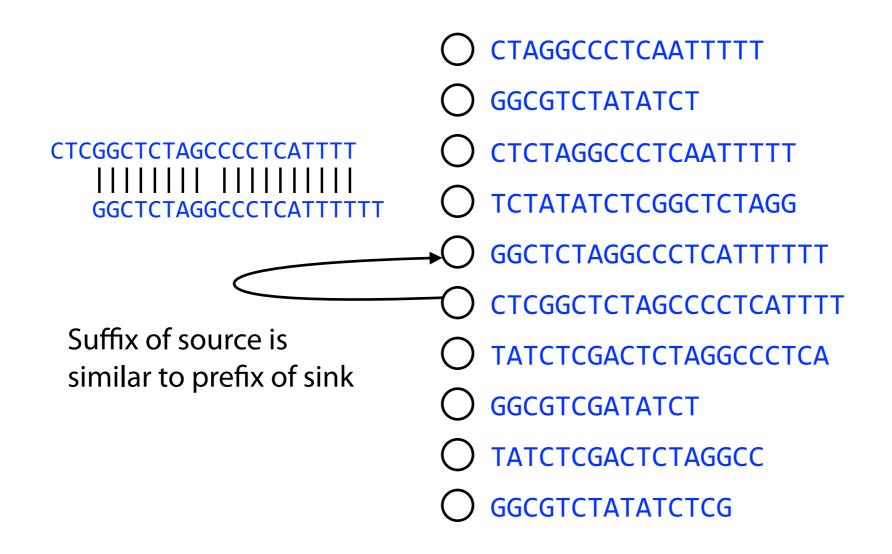
TATCTCGACTCTAGGCC

Read from Father: TCTATATCTCGGCTCTAGG

Sequence from Mother: TCTATATCTCGACTCTAGGCC Sequence from Father: TCTATATCTCGGCTCTAGGCC We'll mostly ignore ploidy, but real tools must consider it

### Overlaps

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



### 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*.

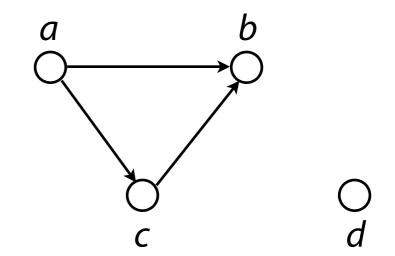
Vertex is drawn as a circle

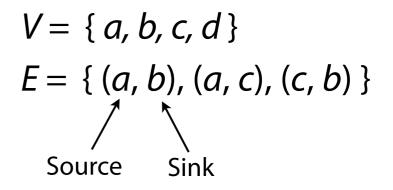
Edge is drawn as a line with an arrow connecting two circles

Vertex also called node or point

Edge also called *arc* or *line* 

Directed graph also called *digraph* 



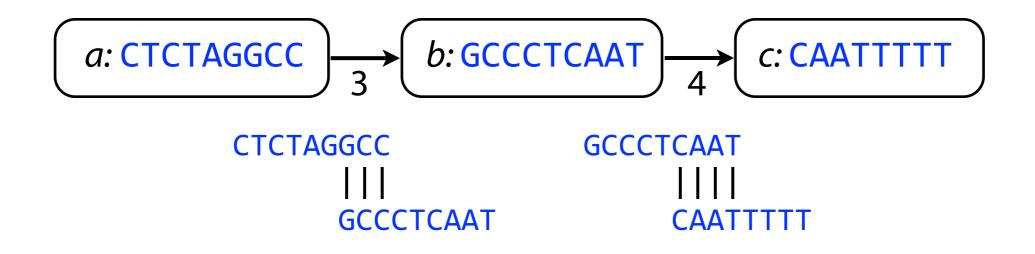


### Overlap graph

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

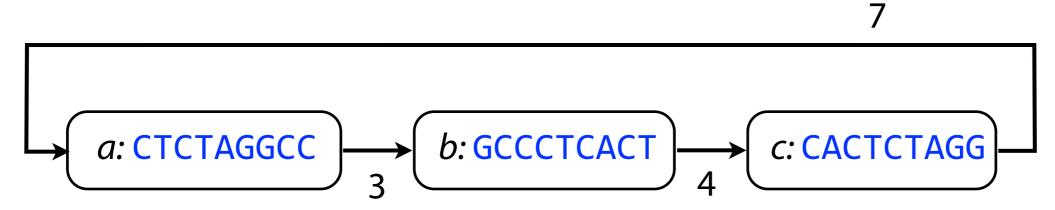
A vertex is a read, a directed edge is an overlap between suffix of source and prefix of sink

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



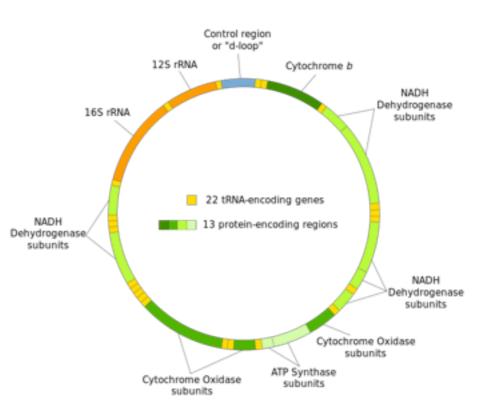
### Overlap graph

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

$$a: \mathsf{CTCTAGGCC} \longrightarrow b: \mathsf{GCCCTCAAT} \longrightarrow c: \mathsf{CAATTTTT}$$

How do we build the overlap graph?

What constitutes an overlap?

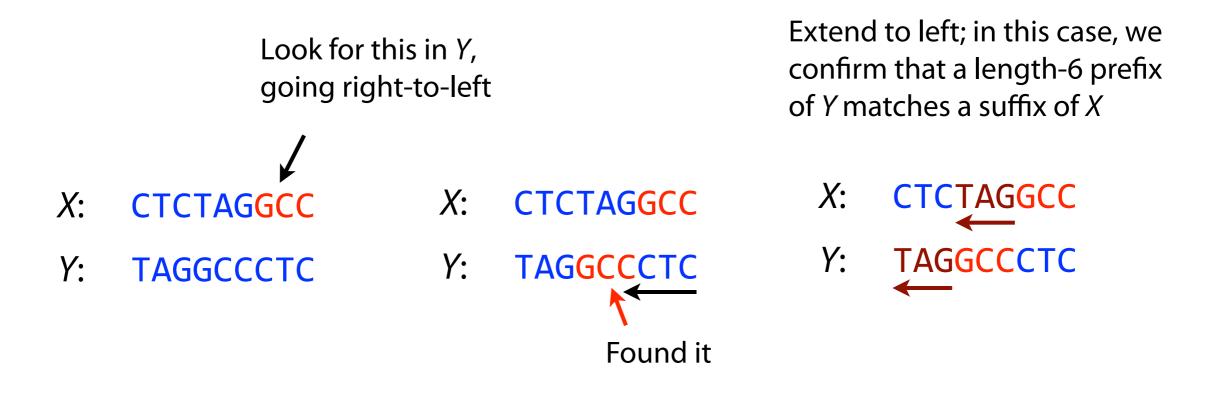
Assume for now an "overlap" is when a suffix of X of length  $\geq l$  exactly matches a prefix of Y, where k is given

## Finding overlaps

Overlap: length-*l* suffix of *X* matches length-*l* prefix of *Y*, where *l* is given

Simple idea: look in *Y* for occurrences of length-*l* suffix of *X*. Extend matches to the left to confirm whether entire prefix of *Y* matches.

Say k = 3

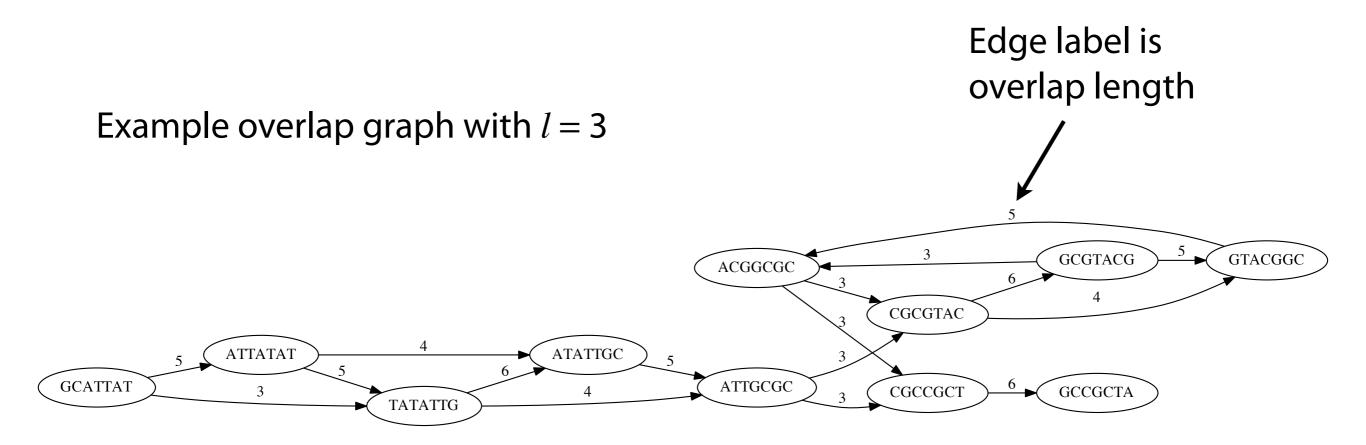


### Finding overlaps: implementation

```
def suffixPrefixMatch(x, y, k):
    ''' Return length of longest suffix of x of length at least k that
        matches a prefix of y. Return 0 if there no suffix/prefix
        match has length at least k. '''
    if len(x) < k or len(y) < k:
        return 0
    idx = len(y) # start at the right end of y
    # Search right-to-left in y for length-k suffix of x
    while True:
        hit = string.rfind(y, x[-k:], 0, idx)
        if hit == -1: # not found
            return 0
        ln = hit + k
        # See if match can be extended to include entire prefix of y
        if x[-ln:] == y[:ln]:
            return ln # return length of prefix
        idx = hit + k - 1 # keep searching to left in Y
    return -1
```

Python example: http://nbviewer.ipython.org/7089885

### Finding overlaps



Original string: GCATTATATTGCGCGTACGGCGCCGCTACA

### 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

> *SCS(S)*: **AAABBBABAA** → 10 →

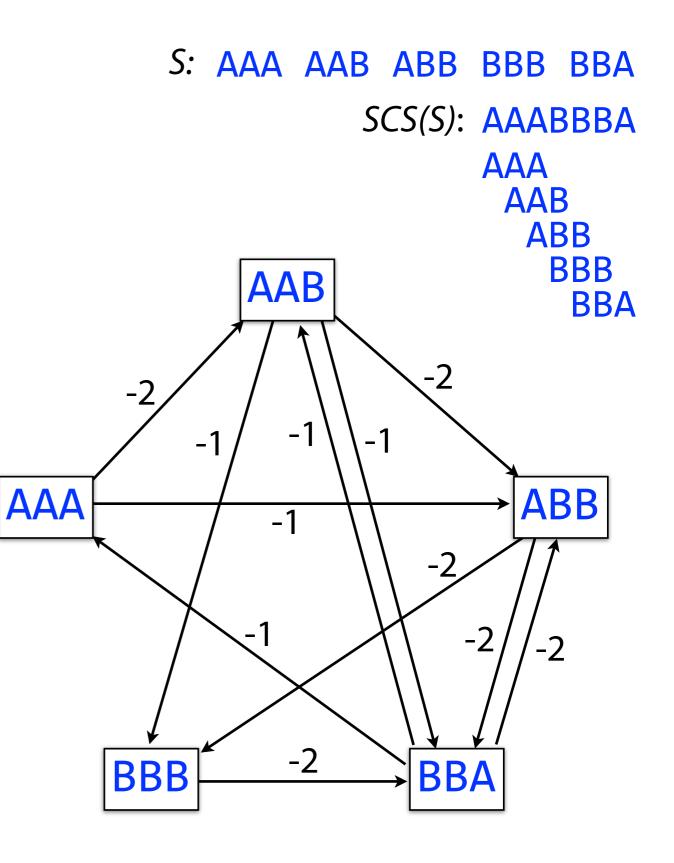
> > AAA AAB ABB BBB BBA BAA ABA BAA

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

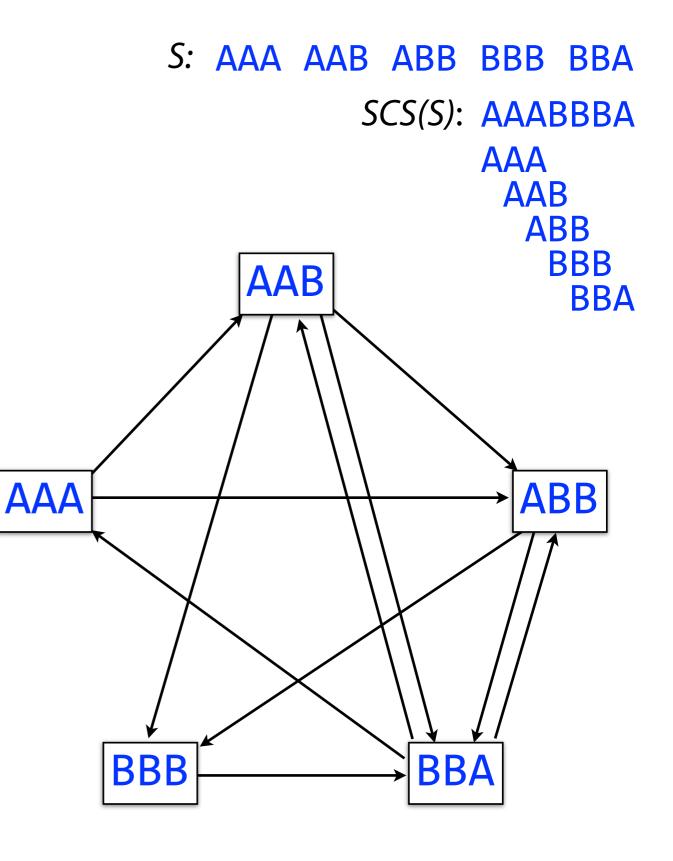
That's the *Traveling Salesman Problem* (*TSP*), which is NP-hard!



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



### Shortest common superstring & friends

Traveling Salesman, Hamiltonian Path, and Shortest Common Superstring are all NP-hard

For refreshers on Traveling Salesman, Hamiltonian Path, NP-hardness and NP-completeness, see Chapters 34 and 35 of "Introduction to Algorithms" by Cormen, Leiserson, Rivest and Stein, or Chapters 8 and 9 of "Algorithms" by Dasgupta, Papadimitriou and Vazirani (free online: <u>http://www.cs.berkeley.edu/~vazirani/algorithms</u>)

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):

–Input strings — ABA ABB AAA AAB BBB BBA BAB BAA 2 BAAB ABA ABB AAA BBB BBA BAB 2 BABB BAAB ABA AAA BBB BBA 2 BBAAB BABB ABA AAA BBB In red are strings that get 2 BBBAAB BABB ABA AAA merged before the next round 2 BBBAABA BABB AAA 2 BABBBAABA AAA Greedy answer: BABBBAABAAA BABBBAABAAA BABBBAABAAA ⊢ Superstring – **Actual SCS:** AAABBBABAA

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 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

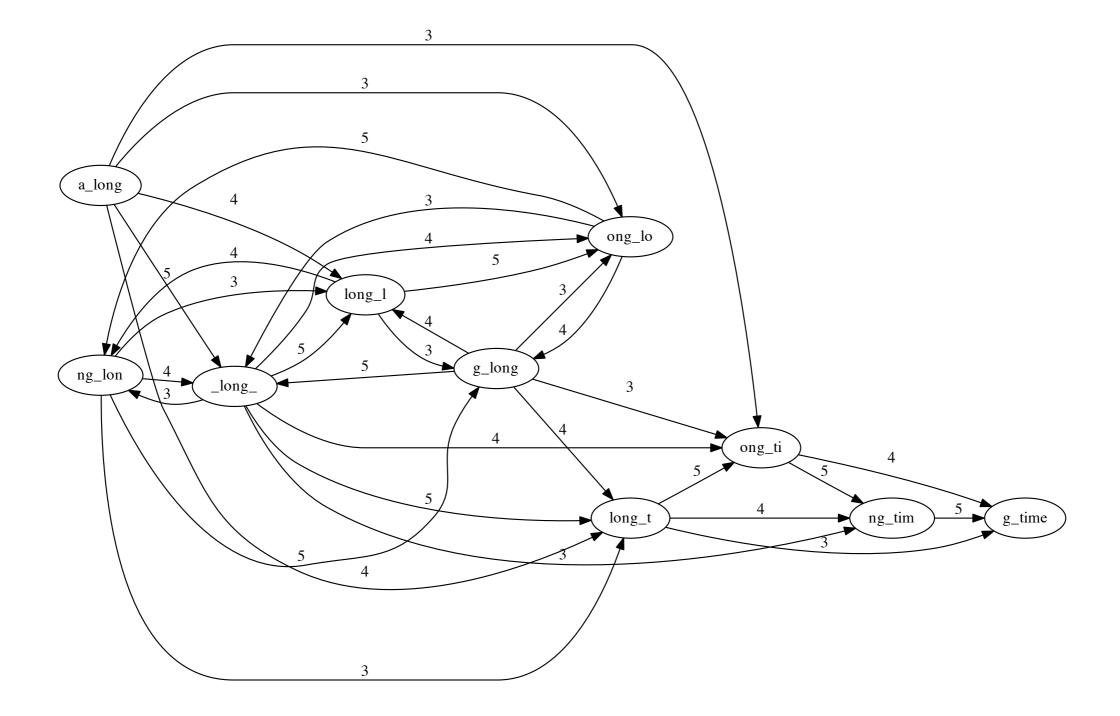
——Superstring ——

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

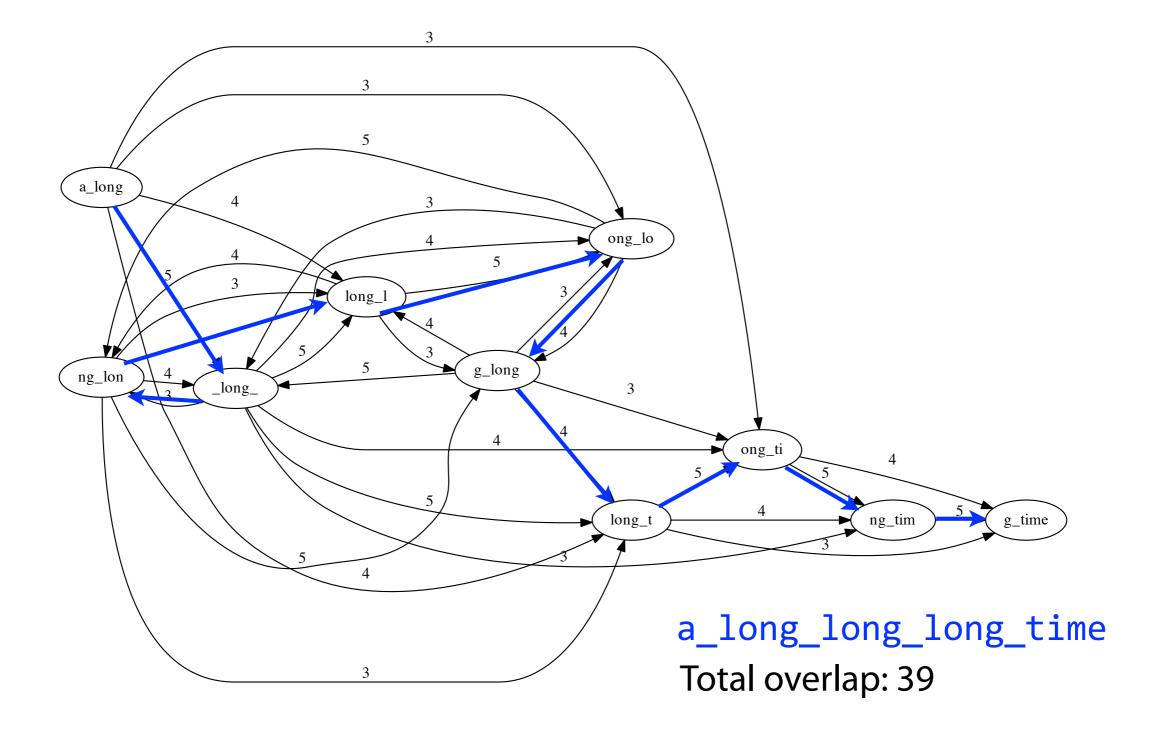
```
ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long g_time ng_tim
5 ng_time ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long
5 ng_time g_long_ ng_lon a_long long_l ong_lo long_t
5 ng_time long_ti g_long_ ng_lon a_long long_l
5 ng_time ong_lon long_ti g_long_ a_long long_l
5 ong_lon long_time g_long_ a_long
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
```

I only got back: a\_long\_long\_time (missing a \_long) What happened?

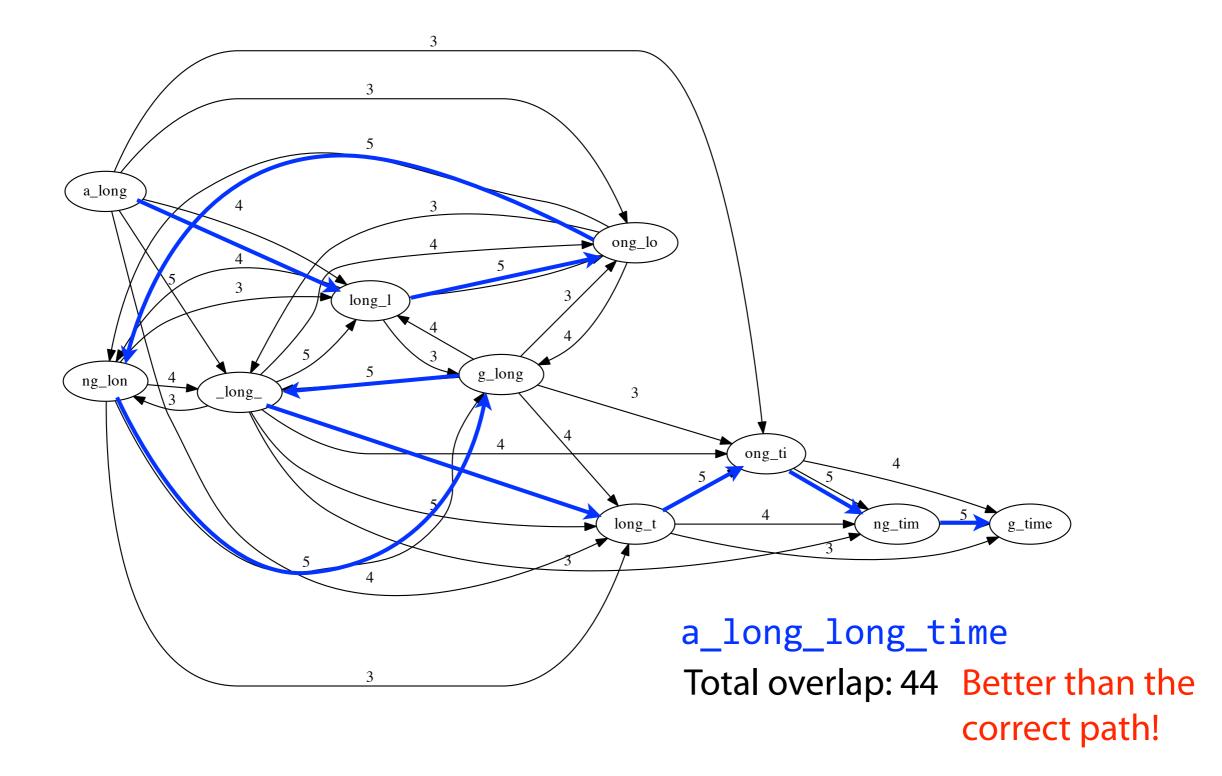
The overlap graph for that scenario (l = 3):



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



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



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

long\_lon ng\_long\_\_long\_lo g\_long\_t ong\_long g\_long\_l ong\_time a\_long\_l \_long\_ti long\_tim 7 long\_time long\_lon ng\_long\_\_long\_lo g\_long\_t ong\_long g\_long\_l a\_long\_l \_long\_ti 7 \_long\_time long\_lon ng\_long\_\_long\_g\_long\_t ong\_long g\_long\_l a\_long\_l 7 \_long\_time ong\_long\_ a\_long\_lo long\_lon g\_long\_t g\_long\_l 7 g\_long\_time ong\_long\_ a\_long\_lo long\_lon g\_long\_t 7 g\_long\_time ong\_long\_ a\_long\_lon g\_long\_l 7 g\_long\_time ong\_long\_ a\_long\_lon g\_long\_l 7 g\_long\_time ong\_long\_l a\_long\_lon g\_long\_l 7 g\_long\_time a\_long\_long\_l 8 a\_long\_long\_long\_time 8 a\_long\_long\_time 8 a\_long\_long\_time

#### Got the whole thing: a\_long\_long\_long\_time

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

```
a_long_long_time
g_long_l
```

One length-8 substring spans all three longs

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!

Basic principle: repeats foil assembly

Another example using Greedy-SCS:

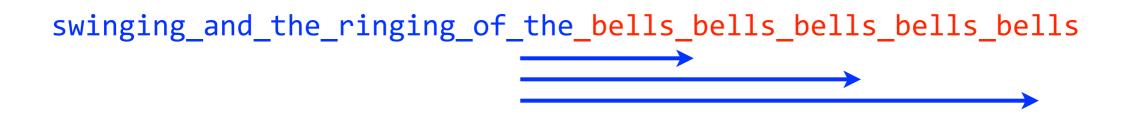
Input: it\_was\_the\_best\_of\_times\_it\_was\_the\_worst\_of\_times

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

- 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

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:

ringing\_of\_the\_bells\_bells\_bells\_bells\_bells\_to\_the\_rhyhming

Basic principle: repeats foil assembly

Yet another example using Greedy-SCS:

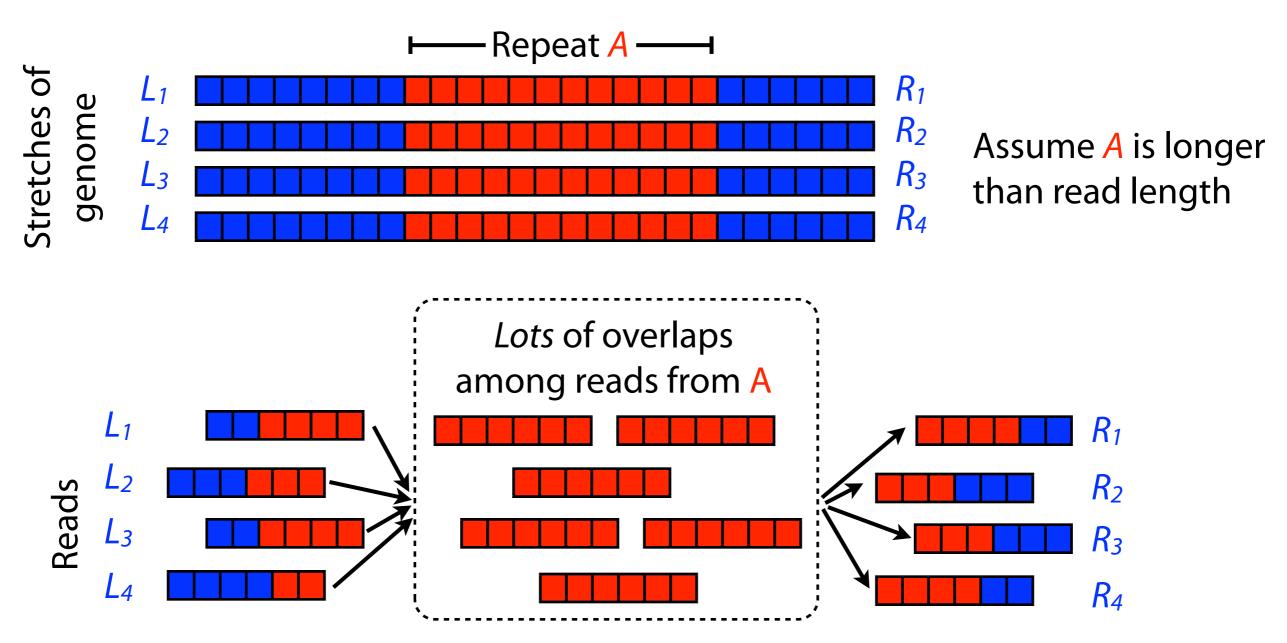
Input: swinging\_and\_the\_ringing\_of\_the\_bells\_bells\_bells\_bells\_

#### *l, k* output

- 3,7 swinging\_and\_the\_ringing\_of\_the\_bells\_bells
- 3,13 swinging\_and\_the\_ringing\_of\_the\_bells\_bells\_bells
- 3,19 swinging\_and\_the\_ringing\_of\_the\_bells\_bells\_bells\_b
- 3,25 swinging\_and\_the\_ringing\_of\_the\_bells\_bells\_bells\_bells\_bells

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

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* 

#### Shortest common superstring: post mortem

SCS is flawed as a way of formulating the assembly problem

No tractable way to find optimal SCS

Had to use Greedy-SCS. Answers might be too long.

SCS spuriously collapses repetitive sequences

Answers might be too short, by a lot!

Need formulations that are (a) tractable, and (b) handle repeats as gracefully as possible

Remember: repeats foil assembly no matter the algorithm. This is a property of read length and repetitiveness of the genome.

### Taxonomy of assembly approaches

Search for most parsimonious explanation of the reads (shortest superstring)

Exact solutions are intractable (e.g. TSP), but a greedy approximation is possible

Any solution will collapse repeats spuriously

Search for "maximum likelihood" explanation of the reads; i.e. force solution to be consistent with uniform coverage

No solutions (that I know of) are tractable

Give up on unresolvable repeats and use a tractable algorithm to assemble the resolvable portions. **This is what real tools do.**