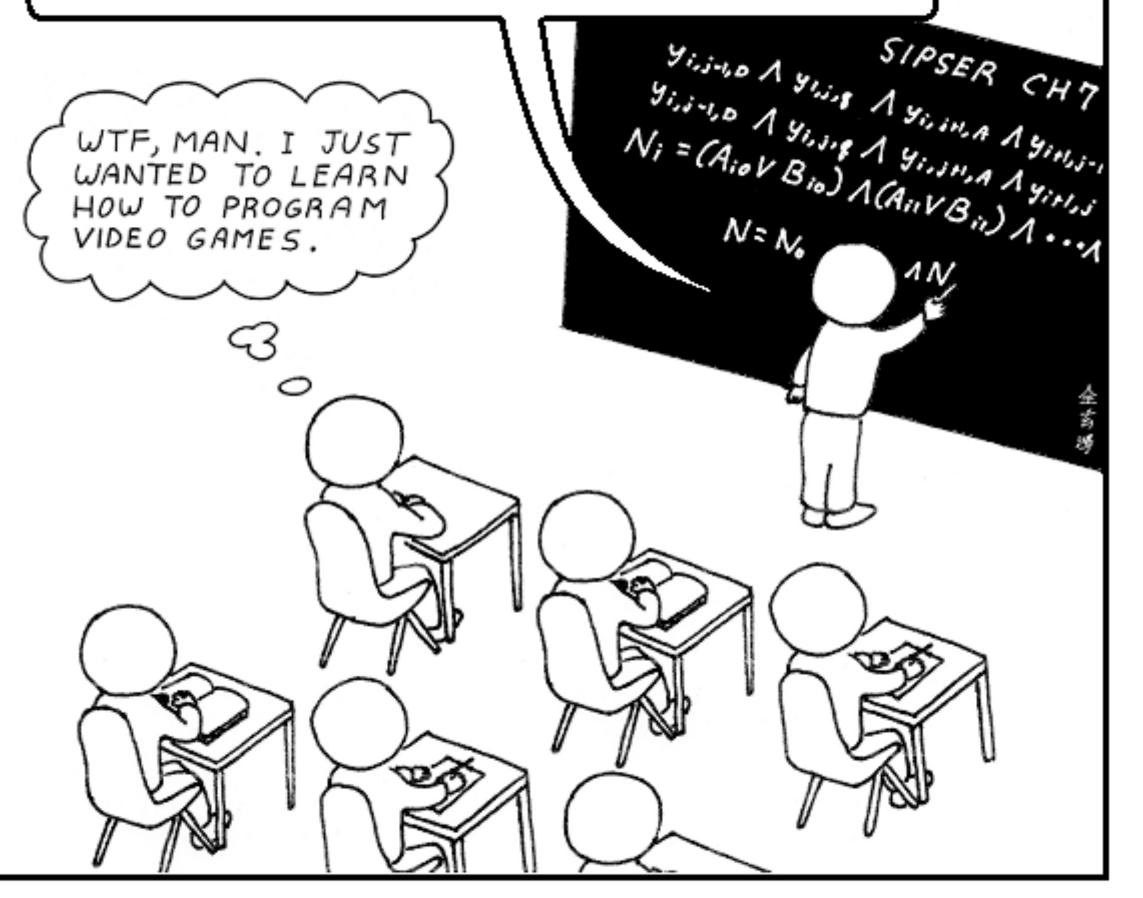
Shortest Common Superstring & Lander-Waterman Statistics



Computational Approaches to Biological Challenges (algorithmic primer)

THUS, FOR ANY NONDETERMINISTIC TURING MACHINE M THAT RUNS IN SOME POLYNOMIAL TIME p(n), WE CAN DEVISE AN ALGORITHM THAT TAKES AN INPUT ω OF LENGTH n AND PRODUCES $E_{n,\omega}$. THE RUNNING TIME IS $O(p^2(n))$ ON A MULTITAPE DETERMINISTIC TURING MACHINE AND...



http://people.cs.pitt.edu/~kirk/cs2110/computer_science_major.PNG

Not actually simple to define constructively

Still debate whether certain areas constitute CS

Computer science is the scientific and practical approach to computation and its applications. It is the systematic study of the feasibility, structure, expression, and mechanization of the methodical procedures (or algorithms) that underlie the acquisition, representation, processing, storage, communication of, and access to information^{*} ...

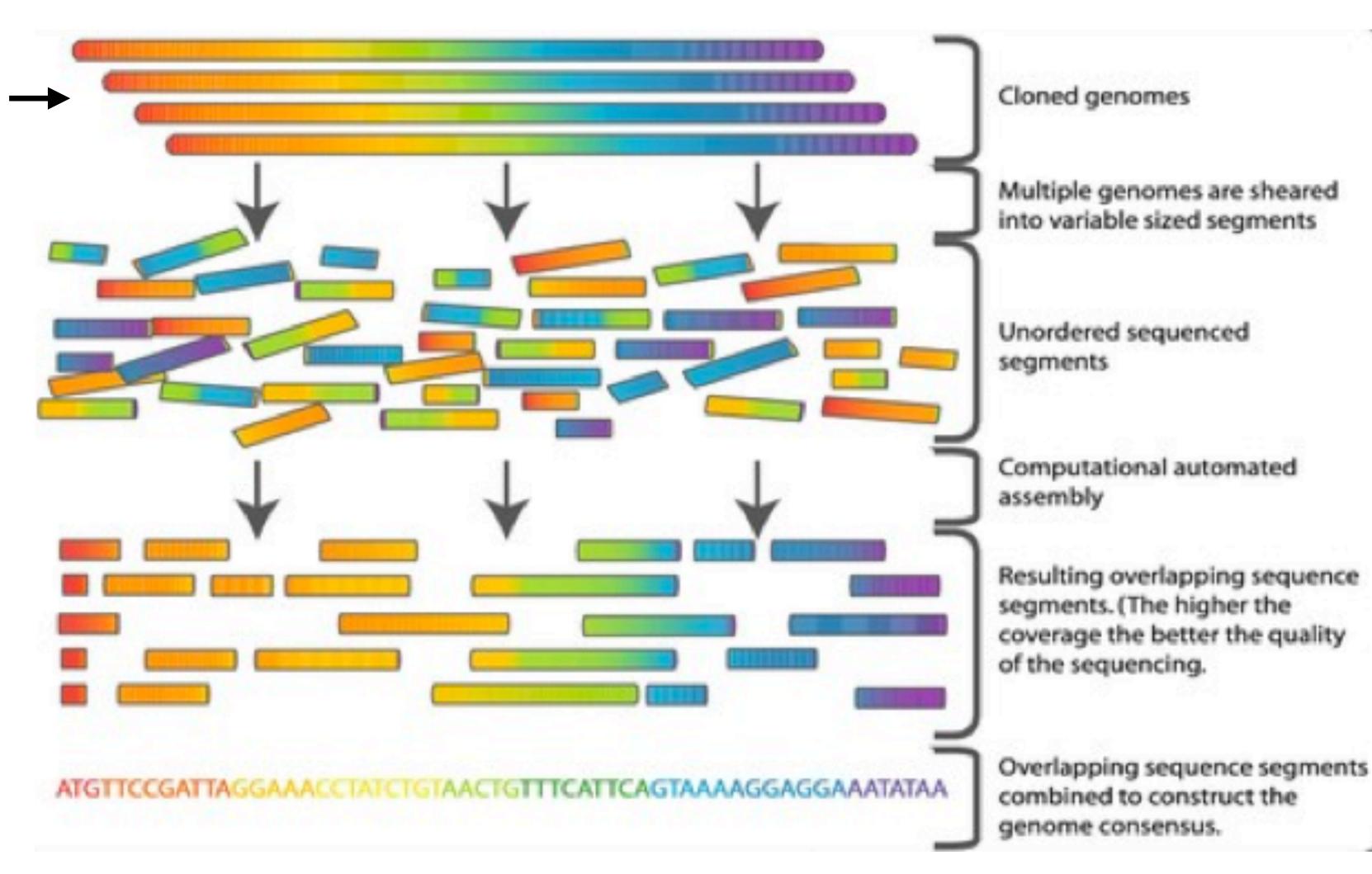
*http://www.cs.bu.edu/AboutCS/WhatIsCS.pdf

Concerned with the development of provably **correct** and efficient computational procedures (algorithms & data structures) to answer **well-specified** problems.

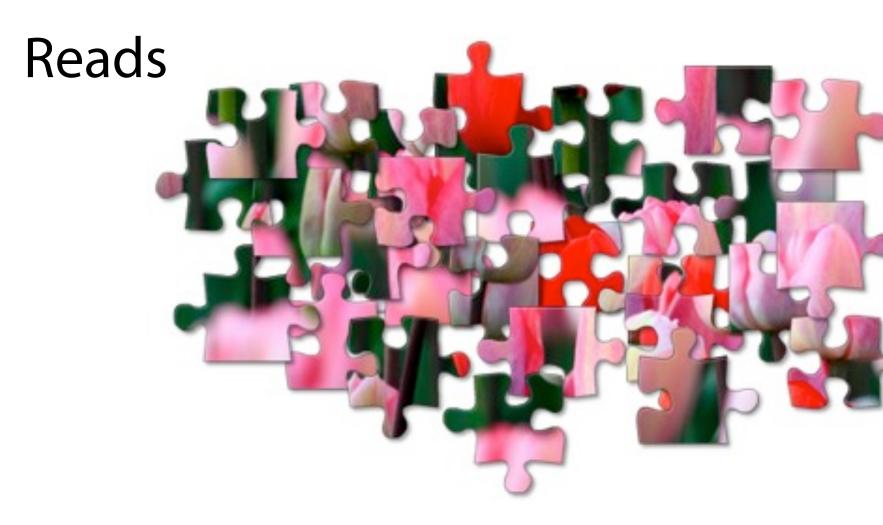
To answer a computational question, we first need a well-formulated problem.

It turns out that a **major challenge** in bioinformatics will simply be determining how to frame the *computational problem* corresponding to a biological question in a well-posed and meaningful way!

What is genome assembly: intuitively?



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



Input DNA



╋

Next 5 slides courtesy of Ben Langmead

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: GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT

Copy: GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT

Fragment: GGCGTCTA TATCTCGG CTCTAGGCCCTC ATTTTTT GGC GTCTATAT CTCGGCTCTAGGCCCTCA GGCGTC TATATCT CGGCTCTAGGCCCT CATTTTT GGCGTCTAT ATCTCGGCTCTAG GCCCTCA TTTTT

GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT

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

CTAGGCCCTCAATTTTT CTCTAGGCCCTCAATTTTT GGCTCTAGGCCCTCATTTTT GGCTCTAGGCCCTCATTTTT this TATCTCGACTCTAGGCCCTCA TATCTCGACTCTAGGCC TCTATATCTCGGCTCTAGG GGCGTCTATATCTCG GGCGTCGATATCT GGCGTCTATATCTC

From these

→ GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT

Reconstruct

this

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

CTAGGCCCTCAATTTTT GGCGTCTATATCT CTCTAGGCCCTCAATTTTT TCTATATCTCGGCTCTAGG GGCTCTAGGCCCTCATTTTT CTCGGCTCTAGCCCCTCATTTT TATCTCGACTCTAGGCCCTCA GGCGTCGATATCT TATCTCGACTCTAGGCC GGCGTCTATATCTCG

→ GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT

From these



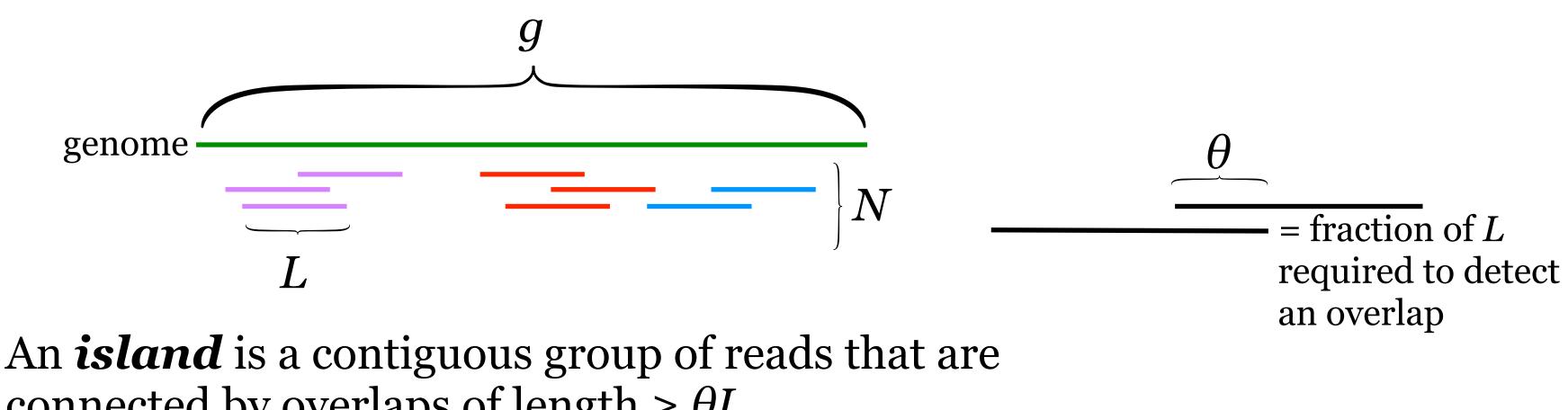
In general: we don't even know exactly how *long* the original string was!

CTAGGCCCTCAATTTTT CTCTAGGCCCTCAATTTTT TCTATATCTCGGCTCTAGG From GGCTCTAGGCCCTCATTTTT these CTCGGCTCTAGCCCCTCATTTT TATCTCGACTCTAGGCCCTCA TATCTCGACTCTAGGCC GGCGTCTATATCTCG

Aside: How Much Coverage is Enough? Lander-Waterman Statistics

Lander ES, Waterman MS (1988). "Genomic mapping by fingerprinting random clones: a mathematical analysis". Genomics 2 (3): 231–239

How many reads to we need to be sure we cover the whole genome?



connected by overlaps of length $\geq \theta L$. (Various colors above)

Want: Expression for expected # of islands given N, g, L, θ .

* Slide from Carl Kingsford

Expected # of Islands

 $\lambda := N/g =$ probability a read starts at a given position (assuming random sampling)

Pr(k reads start in an interval of length x) x trials, want k "successes", small probability λ of success Expected # of successes = λx Poisson approximation to binomial distribution:

 $\Pr(k \text{ reads in leng})$

Expected # of islands = $N \times Pr(read is at rightmost end of island)$ (1- θ)L θ L = $N \times Pr(0 \text{ reads start in } (1-\theta)L)$

 $= N e^{-\lambda(1-\theta)L}$

$$= Ne^{-(1-)}$$

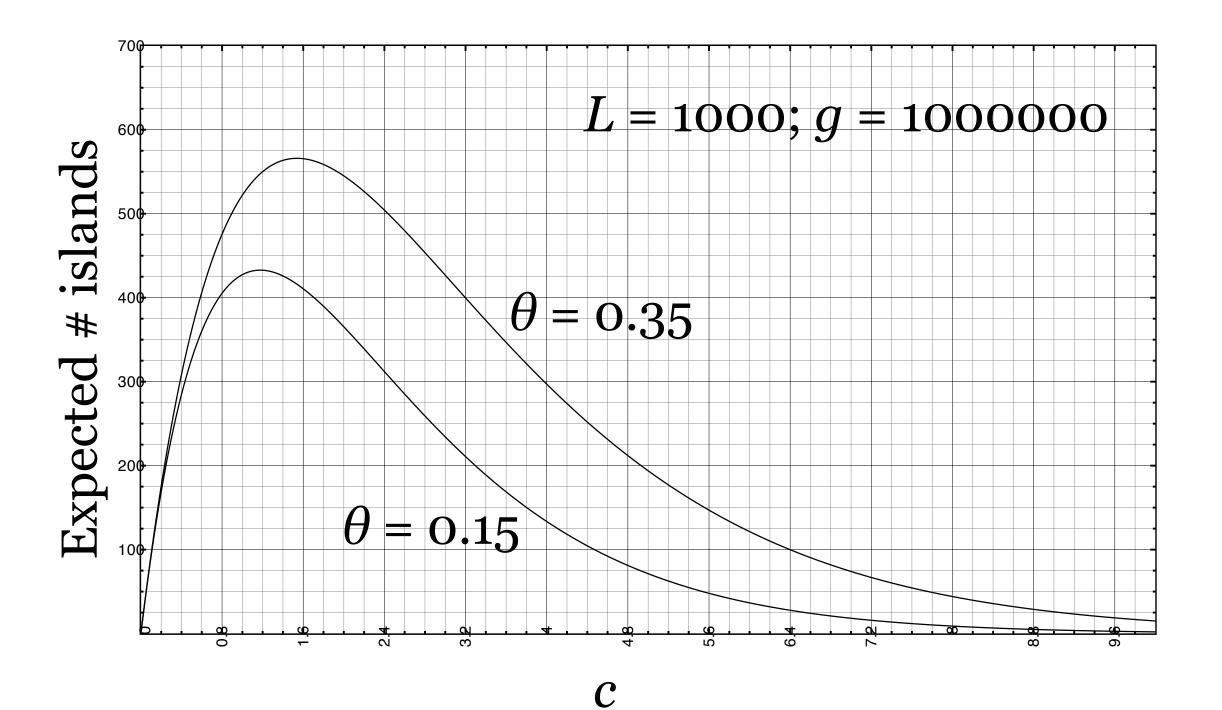
$$\operatorname{gth} x) = e^{-\lambda x} \frac{(\lambda x)^k}{k!}$$

- $= N e^{-\lambda(1-\theta)L} \frac{\lambda^0}{0!}$ (from above)
 - $-\theta$) $LN/g \leftarrow LN/g$ is called the **coverage** c.

Expected # of Islands, 2

We can rewrite this expression to can control: c and θ

Expected # of islands $= Ne^{-(1-\theta)LN/g}$



We can rewrite this expression to depend more directly on the things we

$$= N e^{-(1-\theta)c}$$

$$= \frac{L/g}{L/g} N e^{-(1-\theta)c}$$
$$= \frac{g}{L} c e^{-(1-\theta)c}$$

* Slide from Carl Kingsford

Concerned with the development of provably **correct** and **efficient** computational procedures (algorithms & data structures) to answer **well-specified** problems.

To answer a computational question, we first need a well-formulated problem.

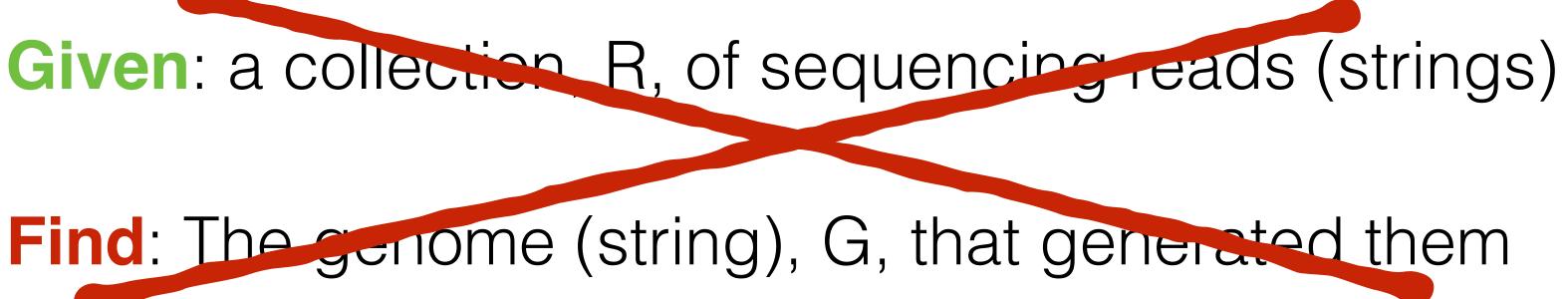
Given: a collection, R, of sequencing reads (strings)

Find: The genome (string), G, that generated them

Concerned with the development of provably **correct** and efficient computational procedures (algorithms & data structures) to answer **well-specified** problems.

To answer a computational question, we first need a well-formulated problem.

Not well-specified. What makes one genome more likely than another? What constraints do we place on the space of solutions?



Concerned with the development of provably **correct** and efficient computational procedures (algorithms & data structures) to answer **well-specified** problems.

To answer a computational question, we first need a well-formulated problem.

Given: a collection, R, of sequencing reads (strings) **Find**: The shortest genome (string), G, that contains

all of them



Shortest Common Superstring

- reads (strings)
- **Find***: The shortest possible genome (string), G, such that s_1, s_2, \ldots, s_k are all substrings of G

How, might we go about solving this problem?

*for reasons we'll explore later, this isn't actually a great formulation for genome assembly.

Given: a collection, $S = \{s_1, s_2, \ldots, s_k\}$, of sequencing

Shortest common superstring

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: **BAAAABBBBAABBBBBBBBBAABBB** ______24 ______ SCS(S): AAABBBABAA ·-----10 -------AAA AAB ABB BBB BBA BAB ABA BAA

order 1: AAA AAB ABA ABB BAA BAB BBA BBB AAA

Slide courtesy of Ben Langmead

order 1: AAA AAB ABA ABB BAA BAB BBA BBB AAAB

Slide courtesy of Ben Langmead

AAABA

Slide courtesy of Ben Langmead

order 1: AAA AAB ABA ABB BAA BAB BBA BBB

AAABABB

Slide courtesy of Ben Langmead

order 1: AAA AAB ABA ABB BAA BAB BBA BBB

Slide courtesy of Ben Langmead

Idea: pick order for strings in *S and* construct superstring

- order 1: AAA AAB ABA ABB BAA BAB BBA BBB
 - AAABABBAABABBBBB ← superstring 1

Slide courtesy of Ben Langmead

- Idea: pick order for strings in *S and* construct superstring
 - order 1: AAA AAB ABA ABB BAA BAB BBA BBB
 - **AAABABBAABABBBBB** ← superstring 1
 - order 2: AAA AAB ABA BAB ABB BBB BAA BBA
 - **AAABABBBBAABBA** ← superstring 2
- Try all possible orderings and pick shortest superstring
- If S contains *n* strings, *n*! (*n* factorial) orderings possible

- order 1: AAA AAB ABA ABB BAA BAB BBA BBB
 - **AAABABBAABABBABBB** ← superstring 1
- order 2: AAA AAB ABA BAB ABB BBB BAA BBA
 - **AAABABBBBAABBA** ← superstring 2
- If S contains *n* strings, *n*! (*n* factorial) orderings possible

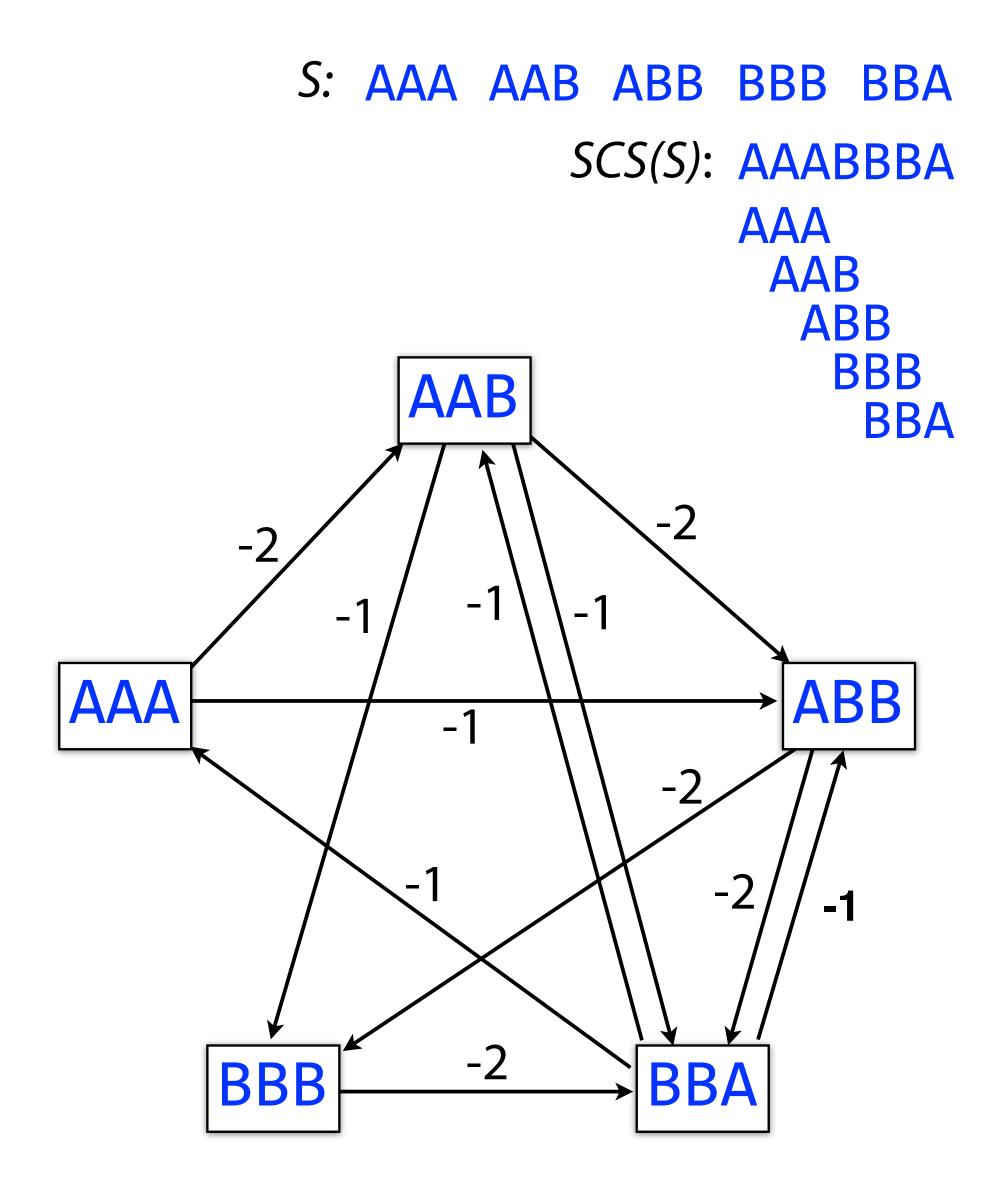
Shortest common superstring

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!



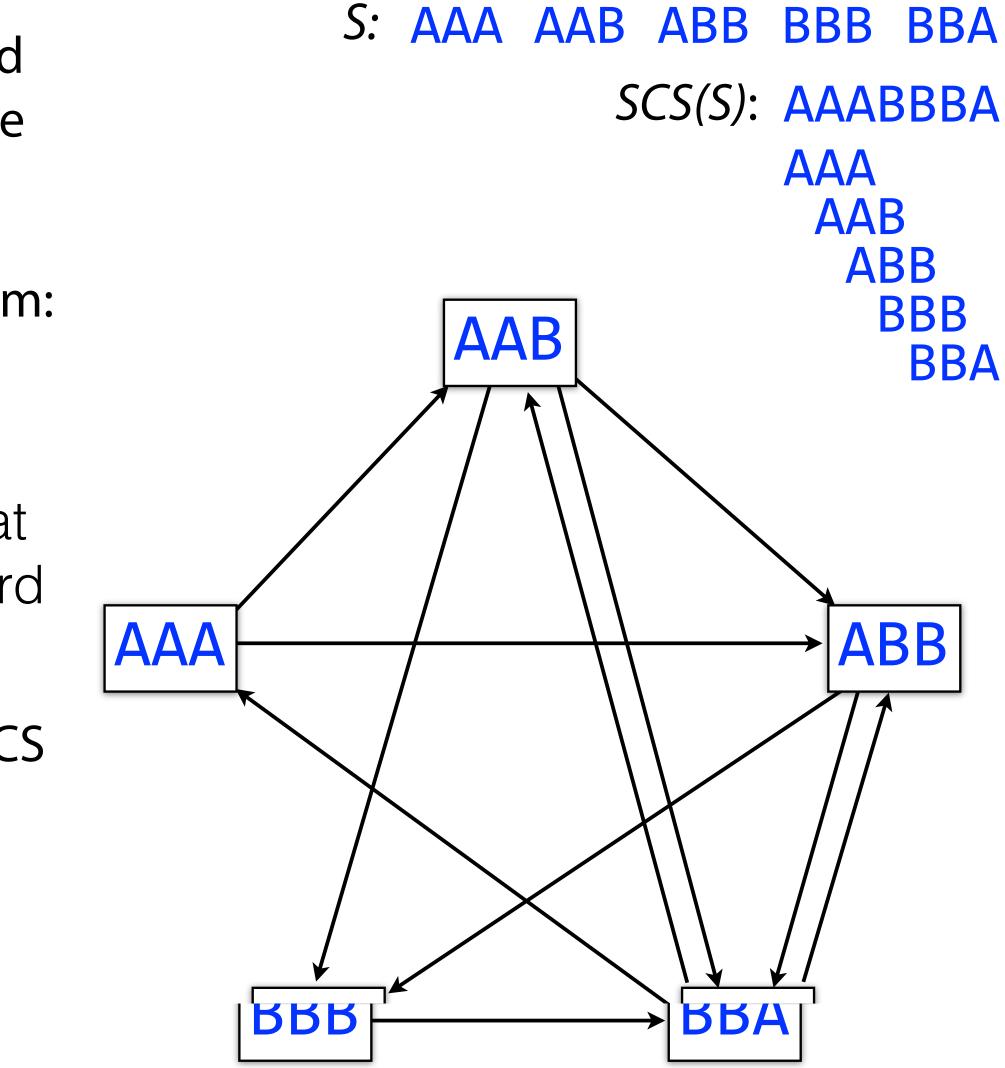
Shortest common superstring

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

So, it's not even the weights that make visiting all nodes once hard

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



Shortest common superstring & friends

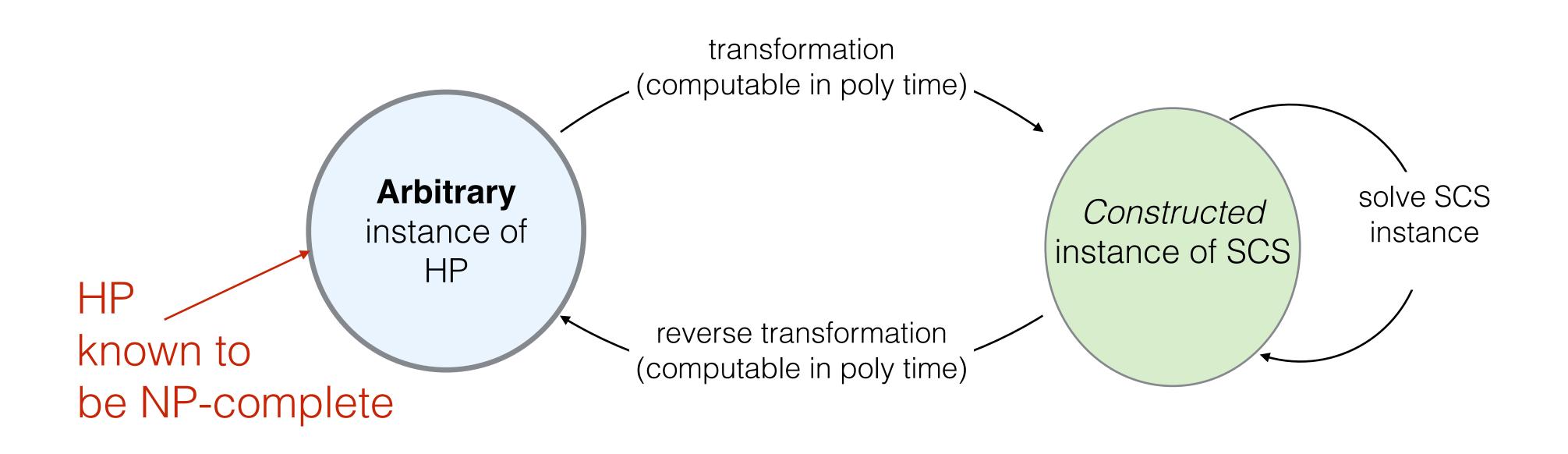
Traveling Salesman, Hamiltonian Path, and Shortest Common Superstring are all NP-hard

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: http://www.cs.berkeley.edu/~vazirani/algorithms)

Who remembers reductions from 351?

Important note: The fact that we modeled SCS as NPhard problems (TSP and HP) **does not** prove that (the decision version of) SCS is NP-complete. To do that, we must **reduce** a known NP-complete problem to **SCS**.

Given an instance I of a known hard problem, generate an instance I' of SCS such that if we can solve I' in polynomial time, then we can solve I in polynomial time. This *implies* that SCS is *at least* as hard as the hard problem.



This can be done e.g. with HAMILTONIAN PATH

Shortest Common Superstring

The fact that (the decision version of) SCS is **NP-complete** means that it is unlikely that there exists *any* algorithm that can solve a general instance of this problem in time polynomial in n — the number of input strings (i.e. reads in the case of genome assembly).

If we give up on finding a *shortest* possible superstring G, and instead look for one that's "near-shortest", how does the situation change?

Shortest Common Superstring

There's a "greedy" *heuristic* that turns out to be an *approximation algorithm* (provides a solution within a constant factor of the the optimum)

At each step, chose the pair of strings with the maximum overlap, merge them, and return the merged string to the collection.

Greedy conjecture factor of 2-OPT *is* the worst case

Open conjecture! We can prove 3.5, but many believe the factor is actually 2.

Different approx. (**not all greedy**)

| ratio | authors | year |
|------------------|--|------|
| | approximating SCS | |
| 3 | Blum, Jiang, Li, Tromp and Yannakakis [4] | 1991 |
| $2\frac{8}{9}$ | Teng, Yao [23] | 1993 |
| $2\frac{5}{6}$ | Czumaj, Gasieniec, Piotrow, Rytter [8] | 1994 |
| $2\frac{50}{63}$ | Kosaraju, Park, Stein [15] | 1994 |
| $2\frac{3}{4}$ | Armen, Stein [1] | 1994 |
| $2\frac{50}{69}$ | Armen, Stein [2] | 1995 |
| $2\frac{2}{3}$ | Armen, Stein [3] | 1996 |
| $2\frac{25}{42}$ | Breslauer, Jiang, Jiang [5] | 1997 |
| $2\frac{1}{2}$ | Sweedyk [21] | 1999 |
| $2\frac{1}{2}$ | Kaplan, Lewenstein, Shafrir, Sviridenko [12] | 2005 |
| $2\frac{1}{2}$ | Paluch, Elbassioni, van Zuylen [18] | 2012 |
| $2\frac{11}{23}$ | Mucha [16] | 2013 |

Golovnev, Kulikov, & Mihajlin. "Approximating Shortest Superstring Problem Using de Bruijn Graphs." Combinatorial Pattern Matching. Springer Berlin Heidelberg, 2013.

Stop when no more overlaps exist. Concatenate resulting strings. l =minimum overlap.

Algorithm in action (l = 1):

- Greedy-SCS: in each round, merge pair of strings with maximal overlap.

minimum overlap.

Algorithm in action (l = 1):

—Input strings —— AAA AAB ABB BBB BBA

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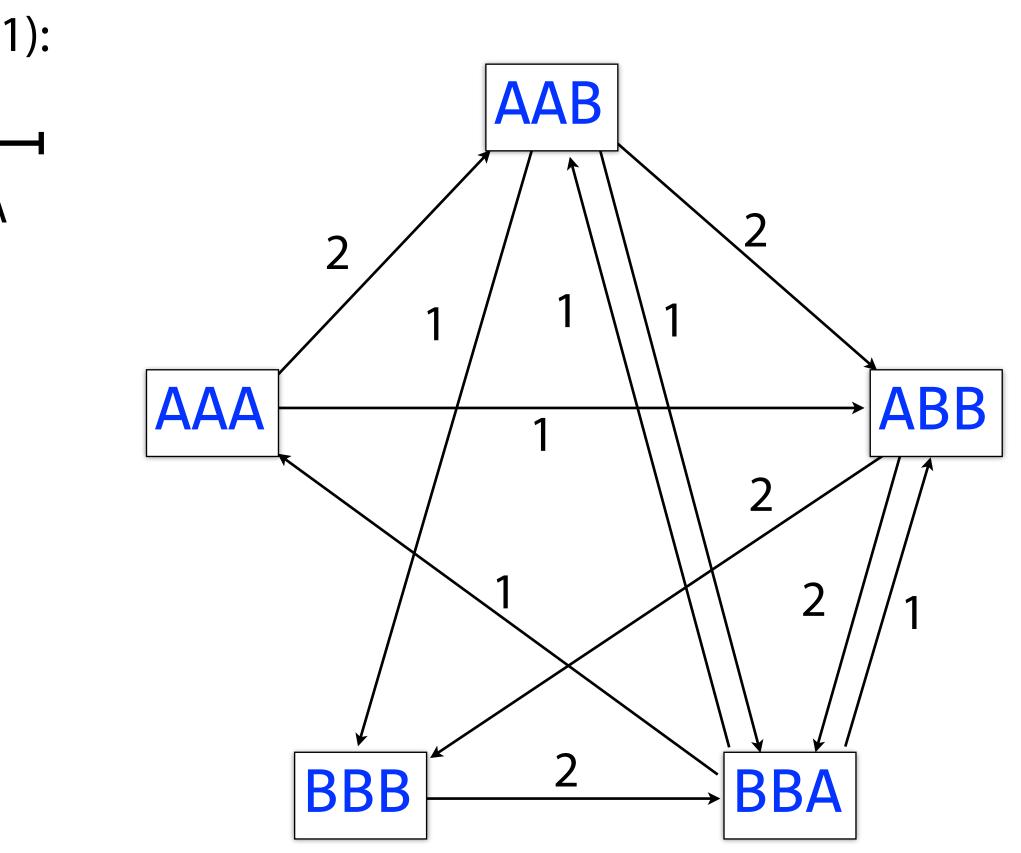
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Slide courtesy of Ben Langmead

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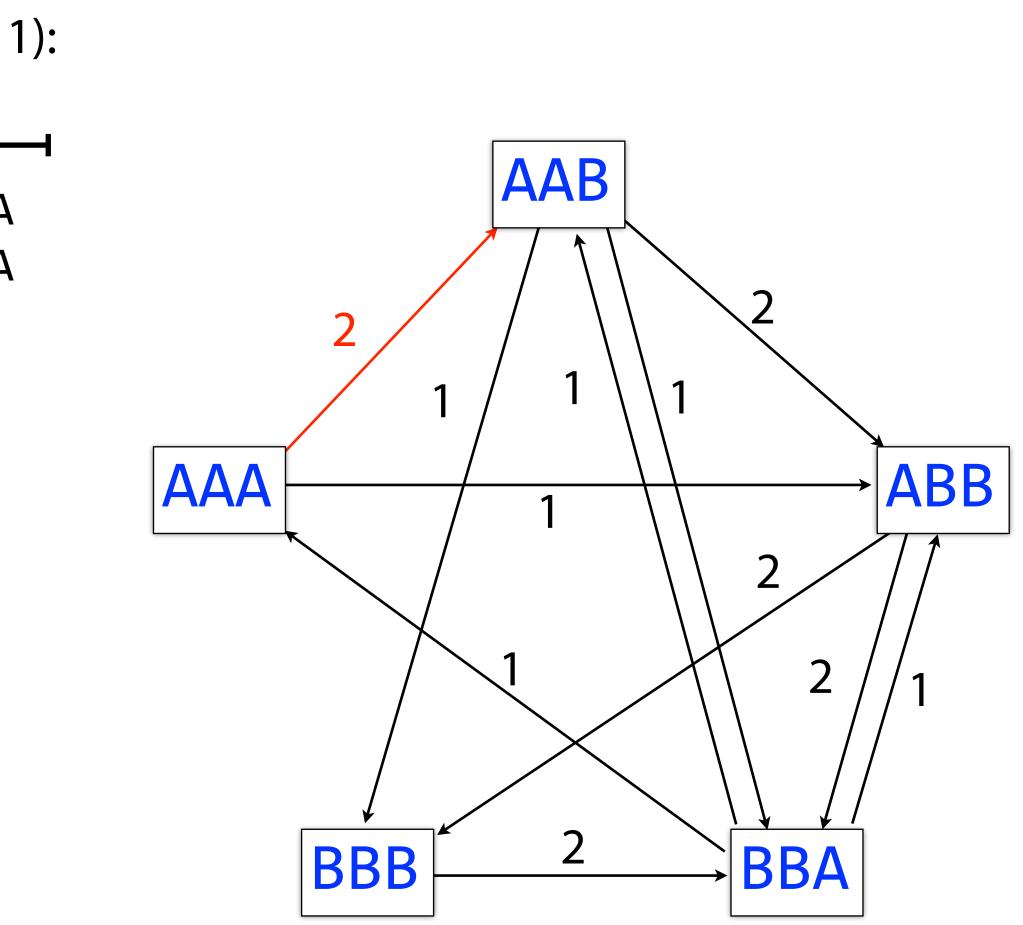
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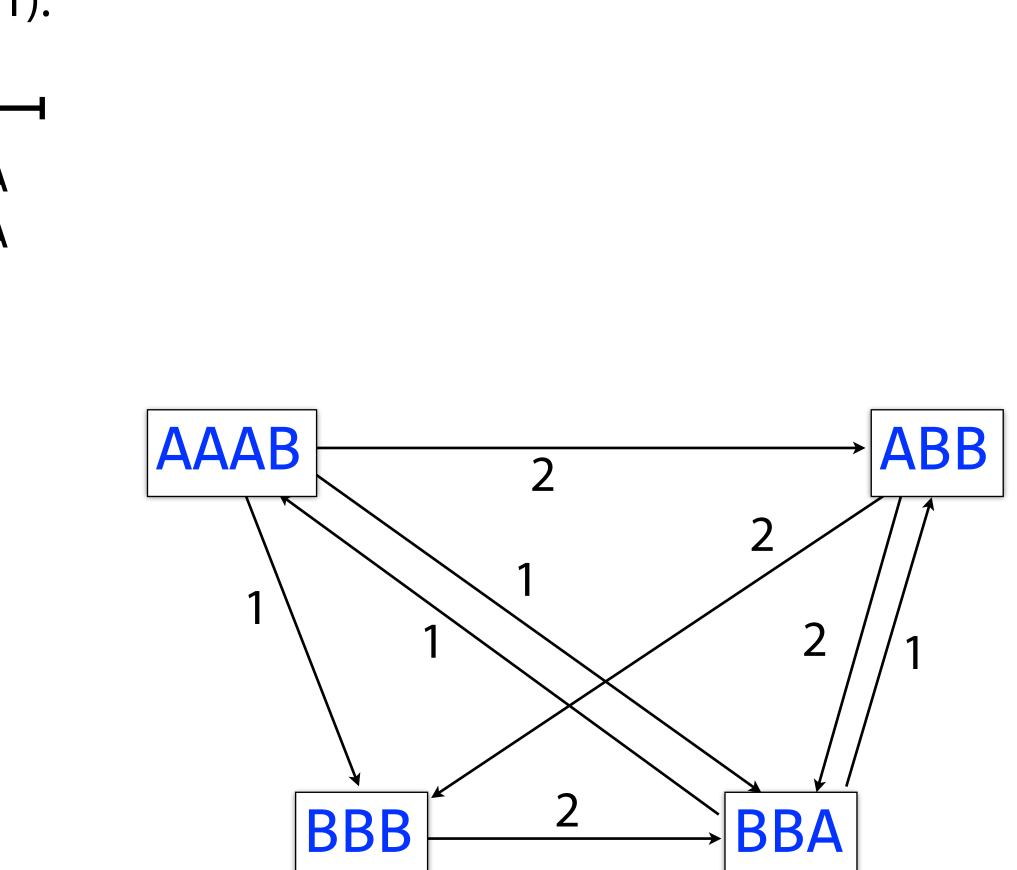


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AAA AAB ABB BBB BBA
AAA AAB ABB BBB BBB BBA
AAAB ABB BBB BBB BBA

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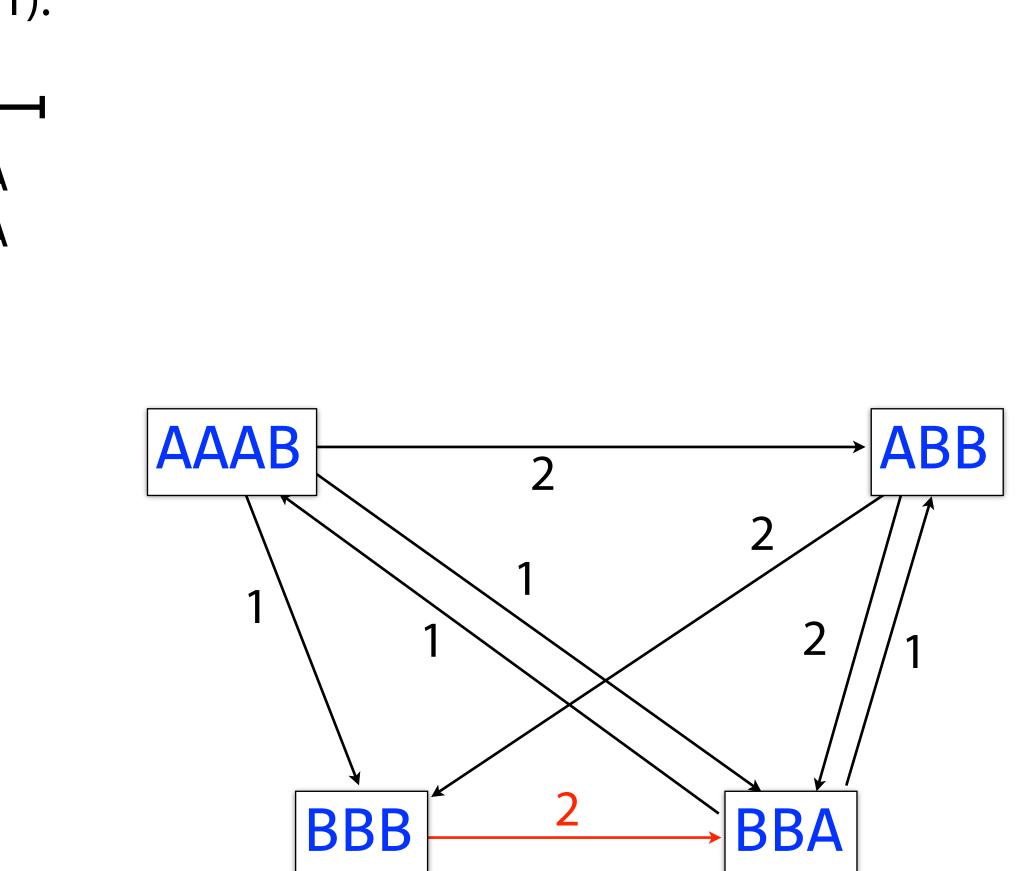


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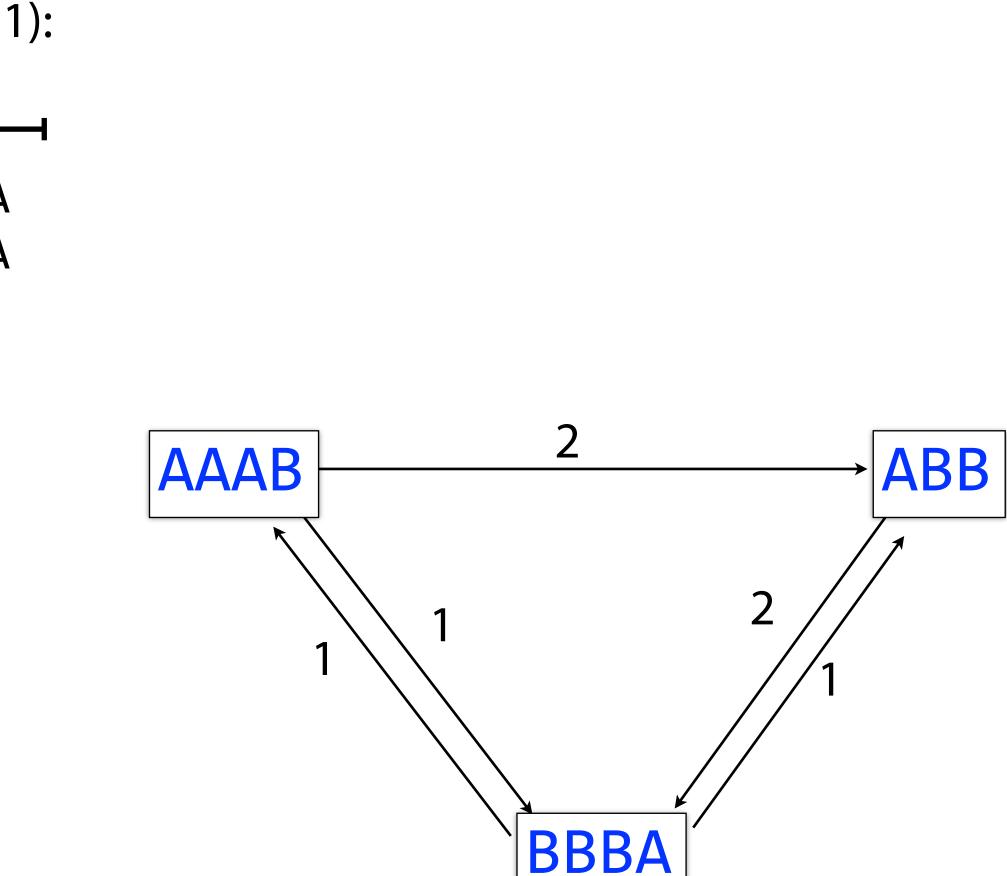


minimum overlap.

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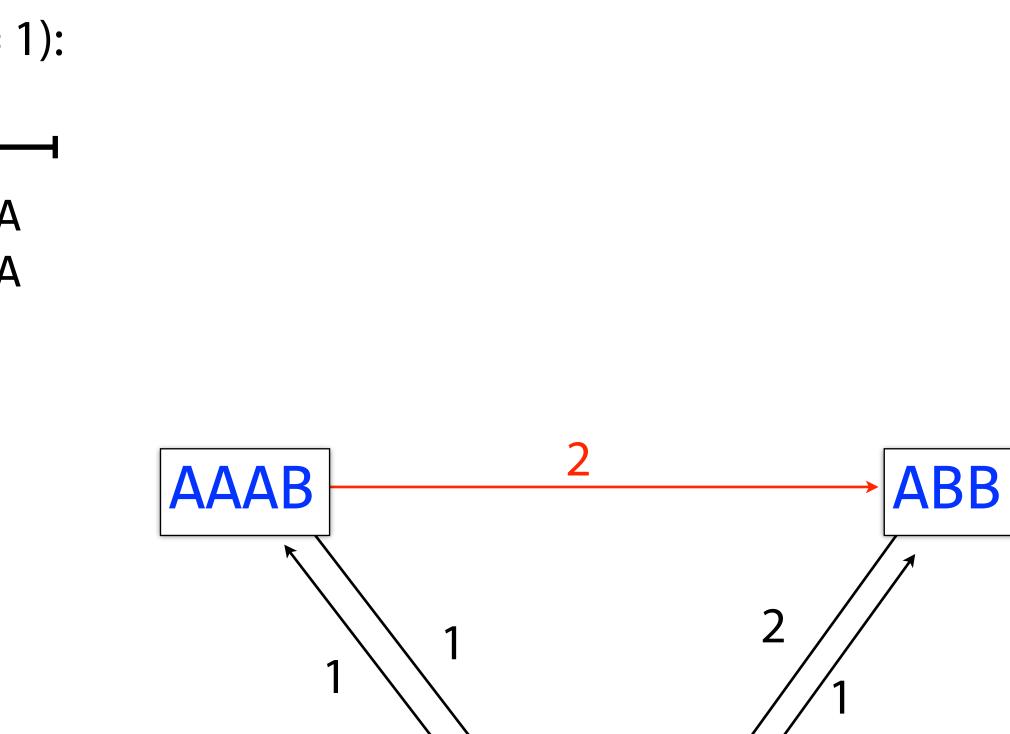
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Algorithm in action (l = 1):

Input strings
 AAA AAB ABB BBB BBA
 AAA AAB ABB BBB BBA
 AAAB ABB BBB BBA
 AAAB BBBA ABB

BBBA

- Greedy-SCS: in each round, merge pair of strings with maximal overlap.
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minimum overlap.

Algorithm in action (l = 1):

Input strings AAA AAB ABB BBB BBA AAA AAB ABB BBB BBA AAAB ABB BBB BBA AAAB BBBA ABB AAABB BBBA

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AAABB BBBA

minimum overlap.

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minimum overlap.

Algorithm in action (l = 1):

—Input strings —— AAA AAB ABB BBB BBA AAA AAB ABB BBB BBA AAAB ABB BBB BBA AAAB BBBA ABB AAABB BBBA AAABBBA

That's the SCS

- Greedy-SCS: in each round, merge pair of strings with maximal overlap.
- Stop when no more overlaps exist. Concatenate resulting strings. l =



AAA AAB ABB BBA BBB \downarrow AAAB ABB BBA BBB

Slide courtesy of Ben Langmead

AAA AAB ABB BBA BBB \downarrow AAAB ABB BBA BBB \downarrow AAAB ABBA BBB

Slide courtesy of Ben Langmead

AAA AAB ABB BBA BBB \downarrow AAAB ABB BBA BBB \downarrow AAAB ABBA BBB \downarrow AAABBA BBB

Slide courtesy of Ben Langmead

AAA AAB ABB BBA BBB \downarrow AAAB ABB BBA BBB \downarrow AAAB ABBA BBB \downarrow AAABBA BBB AAABBABBBB - superstring, length=9

Slide courtesy of Ben Langmead

AAA AAB ABB BBA BBB \downarrow AAAB ABB BBA BBB AAAB ABBA BBB AAABBA BBB AAABBABBBB — superstring, length=9

AAABBBA ← superstring, length=7

Greedy answer isn't necessarily optimal

Slide courtesy of Ben Langmead

Why else might it not be a good model for assembly?

Greedy-SCS assembling all substrings of length 6 from: a_long_long_time. l = 3.

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ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long g_time ng_tim

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Greedy-SCS assembling all substrings of length 6 from: a long long long time. l = 3.

ng_time ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long

```
ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long g_time ng_tim
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ng_time ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long ng_time g_long_ ng_lon a_long long_l ong_ti ong_lo long_t

```
ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long g_time ng_tim
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ng_time ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long ng time g long ng lon a long long l ong ti ong lo long t ng_time long_ti g_long_ ng_lon a_long long_l ong_lo

```
ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long g_time ng_tim
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ng_time ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long ng time g long ng lon a long long l ong ti ong lo long t ng_time long_ti g_long_ ng_lon a_long long_l ong_lo ng time ong lon long ti g long a long long l

```
ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long g_time ng tim
```

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ng time ng lon long a long long long ti ong lo long t g long ng_time g_long_ ng_lon a_long long_l ong_ti ong_lo long_t ng time long ti g_long_ ng_lon a_long long_l ong_lo ng_time ong_lon long_ti g_long_ a_long long_l ong lon long time g long a long long l

```
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ng_time ng_lon <u>long</u> a_long long_l ong_ti ong_lo long_t g_long ng_time g_long_ ng_lon a_long long_l ong_ti ong_lo long_t ng time long ti g_long_ ng_lon a_long long_l ong_lo ng_time ong_lon long_ti g_long_ a_long long_l ong_lon long_time g_long_ a_long long_l long_lon long_time g_long_ a_long

```
ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long g_time ng_tim
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ng time ng lon long a long long long ti ong lo long t g long ng_time g_long_ ng_lon a_long long_l ong_ti ong_lo long_t ng_time long_ti g_long_ ng_lon a_long long_l ong_lo ng_time ong_lon long_ti g_long_ a_long long_l ong_lon long_time g_long_ a_long long_l long_lon long_time g_long_ a_long long_lon g_long_time a_long

```
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ng time ng lon long a long long long ti ong lo long t g long ng_time g_long_ ng_lon a_long long_l ong_ti ong_lo long_t ng_time long_ti g_long_ ng_lon a_long long_l ong_lo ng_time ong_lon long_ti g_long_ a_long long_l ong_lon long_time g_long_ a_long long_l long_lon long_time g_long_ a_long long_lon g_long_time a_long long long time a long

```
ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long g_time ng_tim
```

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ng time ng lon long a long long long ti ong lo long t g long ng_time g_long_ ng_lon a_long long_l ong_ti ong_lo long_t ng time long ti g_long_ ng_lon a_long long_l ong_lo ng time ong lon long ti g long a long long l ong_lon long_time g_long_ a_long long_l long_lon long_time g_long_ a_long long_lon g_long_time a_long long long time a long a_long_long_time

```
ng_lon _long_ a_long long_l ong_ti ong_lo long_t g_long g_time ng_tim
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ng_time ng_lon <u>long</u> a_long long_l ong_ti ong_lo long_t g_long ng_time g_long_ ng_lon a_long long_l ong_ti ong_lo long_t ng time long ti g_long_ ng_lon a_long long_l ong_lo ng_time ong_lon long_ti g_long_ a_long long_l ong_lon long_time g_long_ a_long long_l long_lon long_time g_long_ a_long long_lon g_long_time a_long long_long_time a_long a_long_long_time

Foiled by repeat!

```
ng lon long a long long l ong ti ong lo long t g long g time ng tim
```

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

_long_time long_lon ng_long__long_lo g_long_t ong_long g_long_l a_long_l _long_time a_long_lo long_lon ng_long_ g_long_t ong_long g_long_l _long_time ong_long_ a_long_lo long_lon g_long_t g_long_l g_long_time ong_long_ a_long_lo long_lon g_long_l g_long_time ong_long_ a_long_lon g_long_l g_long_time ong_long_l a_long_lon g_long_time a_long_long_l a_long_long_time a_long_long_long_time

Got the whole thing: a long long long time

```
long_lon ng_long__long_lo g_long_t ong_long g_long_l ong_time a_long_l _long_ti long_tim
long_time long_lon ng_long__long_lo g_long_t ong_long g_long_l a_long_l _long_ti
```

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

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_long_time

g_long_l

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

a_long_long_long_time

g_long_l

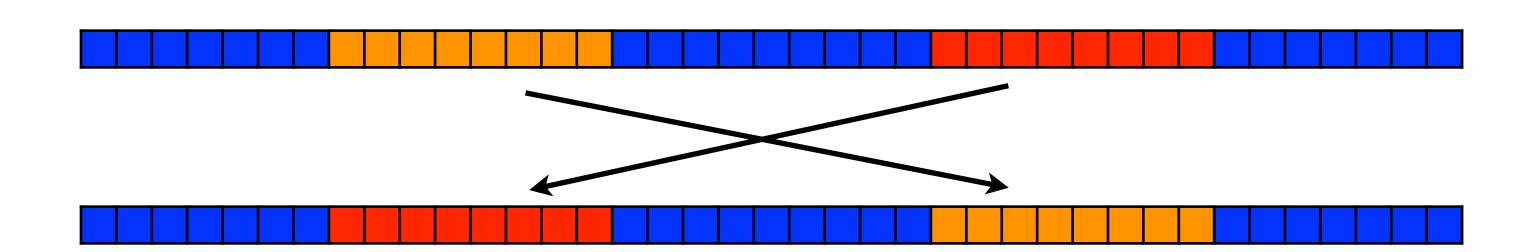
One length-8 substring spans all three longs

Third law of assembly

without mistakes depends on length of reads and repetitive patterns in genome

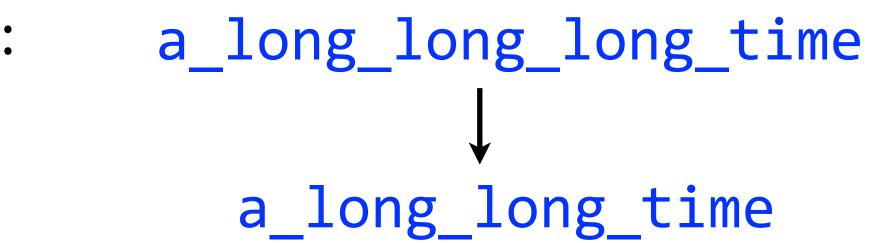
Collapsing a tandem repeat:

Spurious rearrangement:



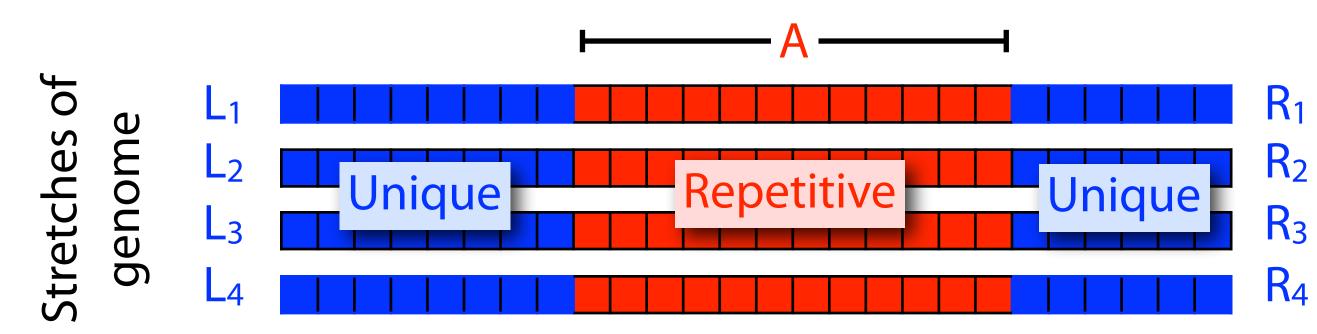
Slide courtesy of Ben Langmead

Repeats make assembly difficult; whether we can assemble



Repeats foil assembly

Portion of overlap graph involving repeat family A

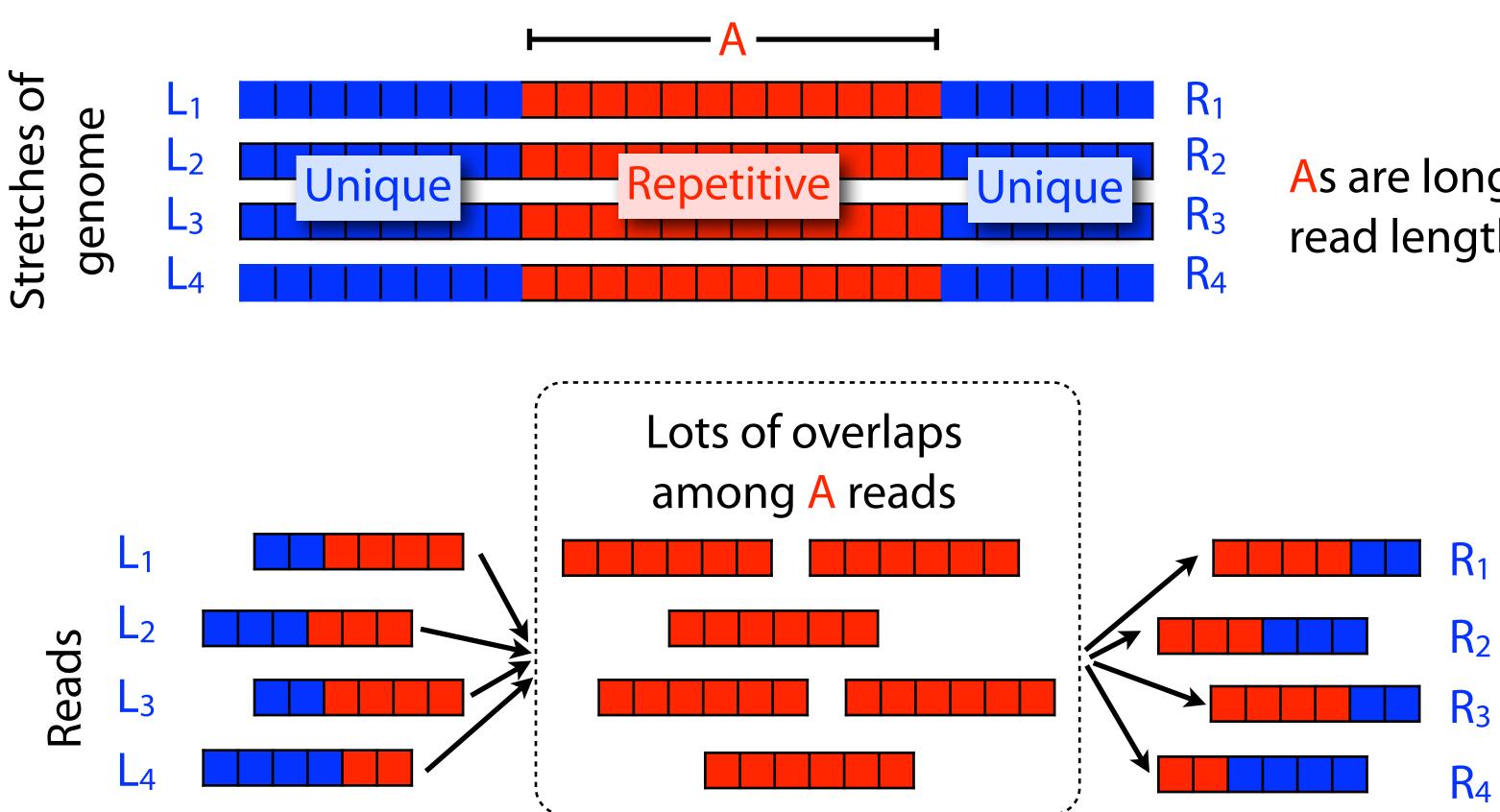


Slide courtesy of Ben Langmead

As are longer than read length

Repeats foil assembly

Portion of overlap graph involving repeat family A

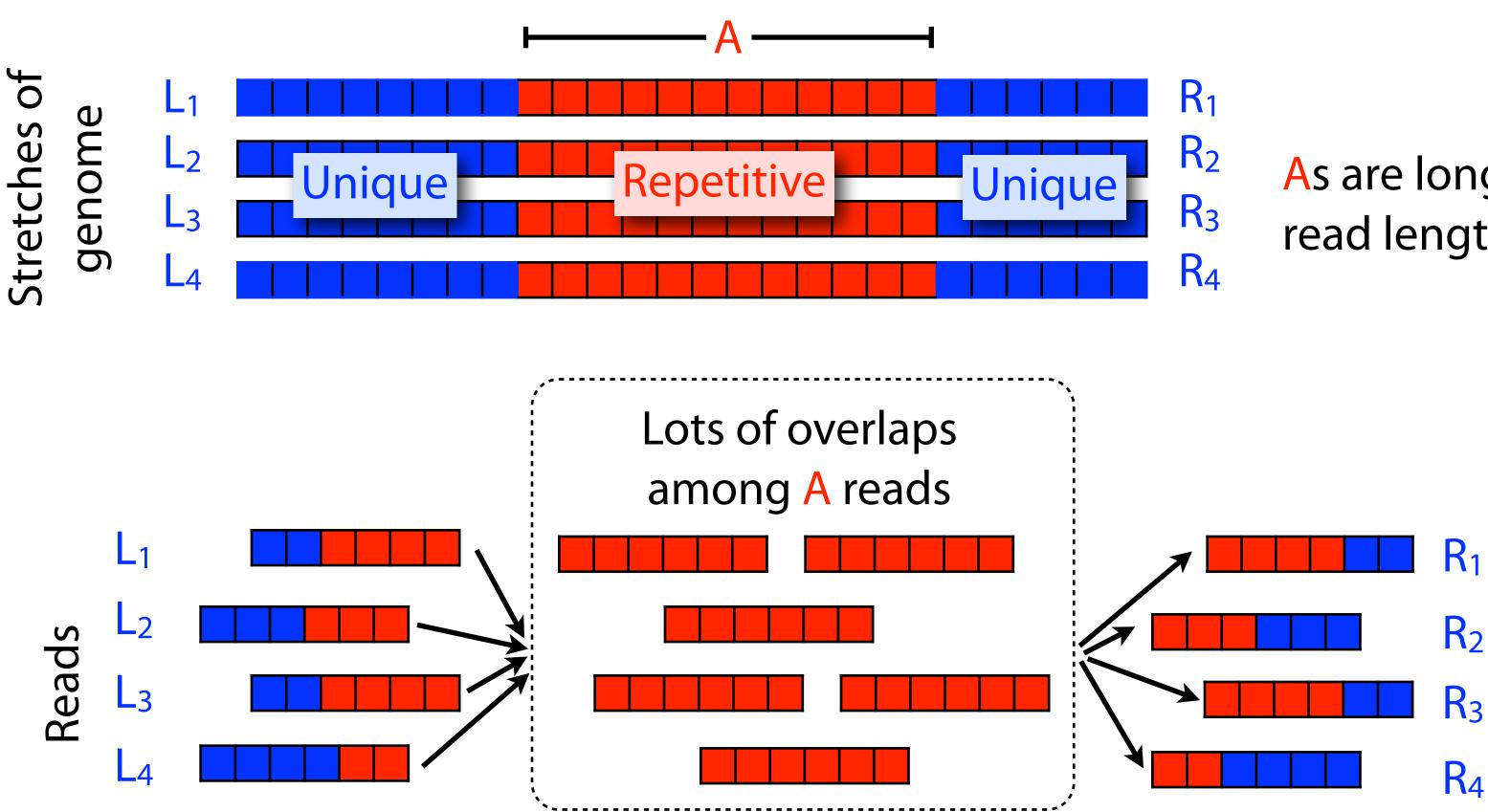


Slide courtesy of Ben Langmead

As are longer than read length

Repeats foil assembly

Portion of overlap graph involving repeat family A



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

Slide courtesy of Ben Langmead

As are longer than read length

Take-home message:

We are interested in *correct and efficient algorithms* for solving *well-specified* problems.

We must be careful about how we *pose* the problems.

Actually, shortest common superstring is a rather poor model for sequence assembly, due to repeats and errors.