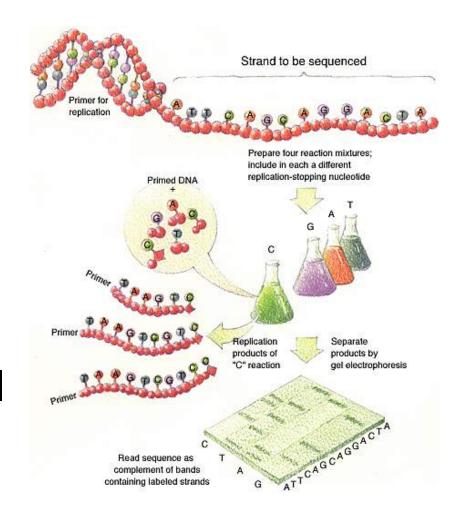
### Eulerian graph

#### **DNA** Sequencing

- Shear DNA into millions of small fragments
- Read 500 700
   nucleotides at a
   time from the small
   fragments (Sanger
   method)



#### Fragment Assembly

- Computational Challenge: assemble individual short fragments (reads) into a single genomic sequence ("superstring")
- Until late 1990s the shotgun fragment assembly of human genome was viewed as intractable problem

#### Shortest superstring problem

- Problem: Given a set of strings, find a shortest string that contains all of them
- Input: Strings  $s_1, s_2, \ldots, s_n$
- Output: A string s that contains all strings
   s<sub>1</sub>, s<sub>2</sub>,..., s<sub>n</sub> as substrings, such that the
   length of s is minimized
- Complexity: NP—hard
- Note: this formulation does not take into account sequencing errors

#### Shortest superstring problem: example

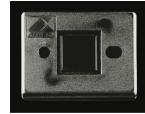
```
The Shortest Superstring problem
Set of strings: {000, 001, 010, 011, 100, 101, 110, 111}
Concatenation
              000 001 010 011 100 101 110 111
Superstring
                        010
                    110
                 011
Shortest
             0001110100
superstring
               001
                   111
                       101
                          100
```

# Sequencing by Hybridization (SBH): History

• **1988:** SBH is suggested as an Financial an alternative sequencing method.

First microarray prototype (1989)

 1991: Light directed polymer synthesis developed by Steve Fodor and colleagues. First commercial DNA microarray prototype w/16,000 features (1994)



500,000 features per chip **(2002)** 

 1994: Affymetrix develops first 64-kb DNA microarray



#### How SBH works

- Attach all possible DNA probes of length / to a flat surface, each probe at a distinct and known location. This set of probes is called the DNA array.
- Apply a solution containing fluorescently labeled DNA fragment to the array.
- The DNA fragment hybridizes with those probes that are complementary to substrings of length / of the fragment.

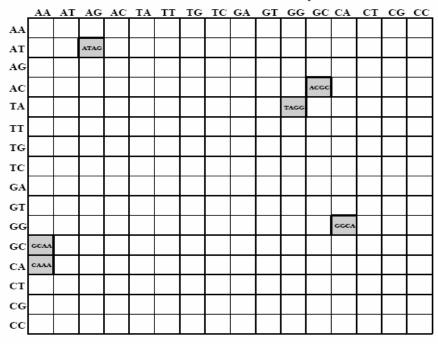
#### How SBH works

 Using a spectroscopic detector, determine which probes hybridize to the DNA fragment to obtain the *I*—mer composition of the target DNA fragment.

 Apply the combinatorial algorithm (below) to reconstruct the sequence of the target DNA fragment from the *I* – mer composition.

#### Hybridization on DNA Array

#### Universal DNA Array



DNA target TATCCGTTT (complement of ATAGGCAAA) hybridizes to the array of all 4-mers:

ATAGGCAAA ATAG TAGG AGGC GGCA GCAAA

#### *l*-mer composition

- Spectrum (s, l) unordered multiset of all possible (n l + 1) l-mers in a string s of length n
- The order of individual elements in Spectrum (s, I) does not matter
- For s = TATGGTGC all of the following are equivalent representations of Spectrum (s, 3):
   {TAT, ATG, TGG, GGT, GTG, TGC}
   {ATG, GGT, GTG, TAT, TGC, TGG}
   {TGG, TGC, TAT, GTG, GGT, ATG}

#### I-mer composition

- Spectrum (s, l) unordered multiset of all possible (n l + 1) l-mers in a string s of length n
- The order of individual elements in Spectrum (s, l) does not matter
- For s = TATGGTGC all of the following are equivalent representations of Spectrum (s, 3):
   {TAT, ATG, TGG, GGT, GTG, TGC}
   {ATG, GGT, GTG, TAT, TGC, TGG}
   {TGG, TGC, TAT, GTG, GGT, ATG}
- We usually choose the lexicographically maximal representation as the canonical one.

# Observations: different sequences may have the same spectrum

Different sequences may have the same spectrum:

```
Spectrum(GTATCT,2)=
Spectrum(GTCTAT,2)=
{AT, CT, GT, TA, TC}
```

#### The SBH problem

Goal: Reconstruct a string from its *I*-mer composition

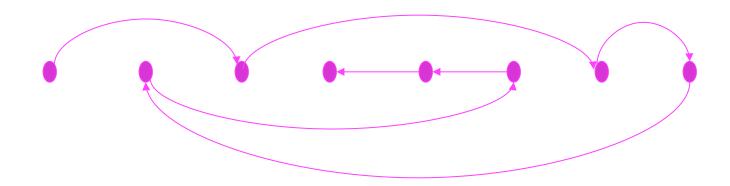
 Input: A set S, representing all I-mers from an (unknown) string s

Output: String s such that Spectrum (s,l) =

# SBH: Hamiltonian path approach

S = { ATG AGG TGC TCC GTC GGT GCA CAG }

ATG AGG TGC TCC GTC GGT GCA CAG



**ATG CAGGTCC** 

Path visited every VERTEX once

#### SBH: Hamiltonian path approach

A more complicated graph:

```
S = { ATG TGG TGC GTG GGC GCA GCG CGT }
```

#### SBH: Hamiltonian path approach

 $S = \{ ATG TGG TGC GTG GGC GCA GCG CGT \}$ 

Path 1:

**ATGCGTGGCA** 

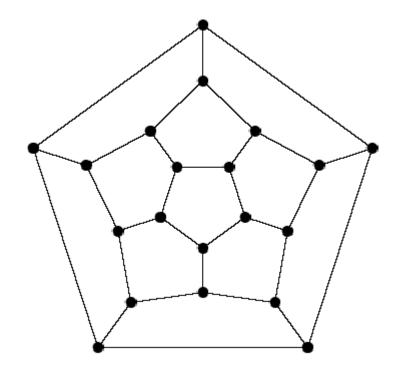
Path 2:

**ATGGCGTGCA** 

#### Hamiltonian cycle problem

 Find a cycle that visits every *vertex* exactly once

NP-complete

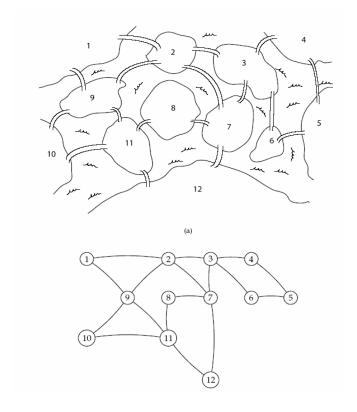


Game invented by Sir William Hamilton in 1857

#### Eulerian cycle problem

 Find a cycle that visits every edge exactly once

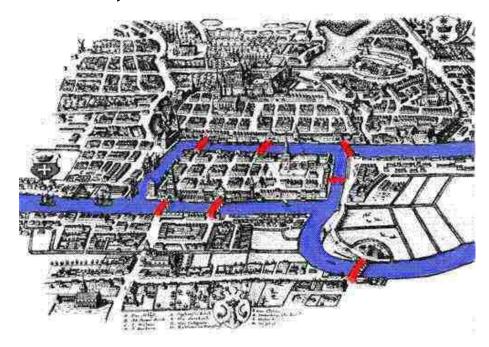
Linear time



More complicated Königsberg

#### The Bridge Obsession Problem

Find a tour crossing every bridge just once *Leonhard Euler*, 1735



Bridges of Königsberg

#### **Euler Theorem**

 A graph is balanced if for every vertex the number of incoming edges equals to the number of outgoing edges:

$$in(v) = out(v)$$

• Theorem: A connected graph is Eulerian if and only if each of its vertices is balanced.

#### Euler Theorem: proof

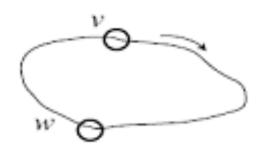
Eulerian → balanced
 for every edge entering v (incoming edge)
 there exists an edge leaving v (outgoing edge). Therefore

$$in(v) = out(v)$$

Balanced → Eulerian
 ???

#### Algorithm for Constructing an Eulerian Cycle

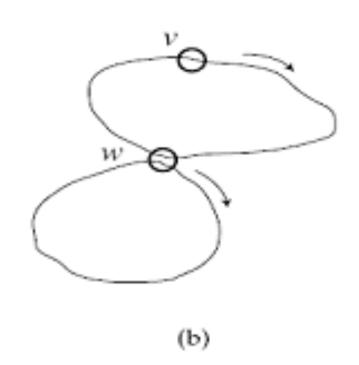
Start with an arbitrary vertex v and form an arbitrary cycle with unused edges until a dead end is reached. Since the graph is Eulerian this dead end is necessarily the starting point, i.e., vertex v.



(a)

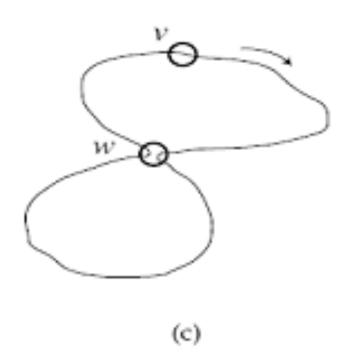
### Algorithm for Constructing an Eulerian Cycle (cont'd)

b. If cycle from (a) above is not an Eulerian cycle, it must contain a vertex w, which has untraversed edges. Perform step (a) again, using vertex w as the starting point. Once again, we will end up in the starting vertex w.



#### Algorithm for Constructing an Eulerian Cycle

c. Combine the cycles from (a) and (b) into a single cycle and iterate step (b).



#### Euler Theorem: extension

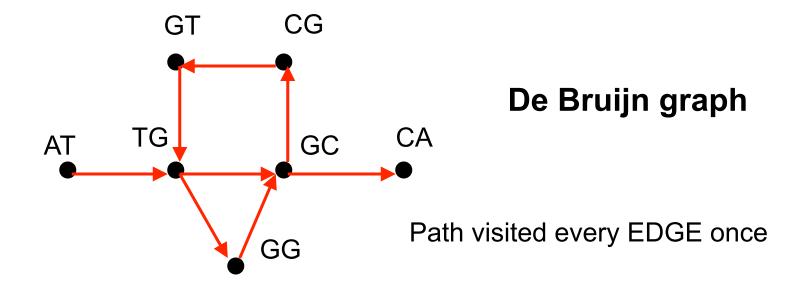
 Theorem: A connected graph has an Eulerian path if and only if it contains at most two semi-balanced vertices and all other vertices are balanced.

#### SBH: Eulerian path approach

S = { ATG, TGC, GTG, GGC, GCA, GCG, CGT }

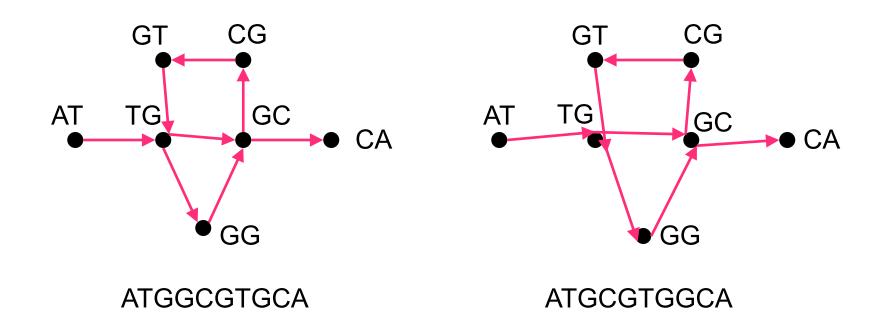
Vertices correspond to (l-1) – mers : { AT, TG, GC, GG, GT, CA, CG }

Edges correspond to I – mers from S



#### SBH: multiple solutions

S = { AT, TG, GC, GG, GT, CA, CG } corresponds to two different paths:



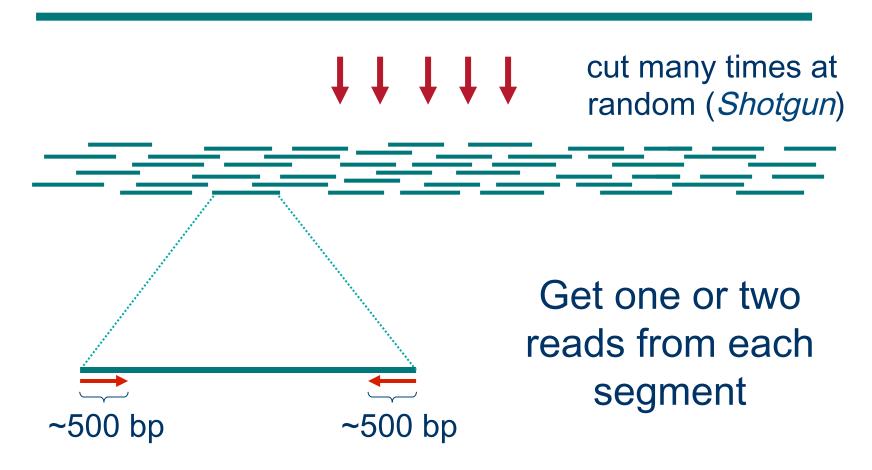
# Representing alternative SBH solutions by Eulerian graphs

- # of solutions ←→ # of Eulerian cycles
  - BEST theorem

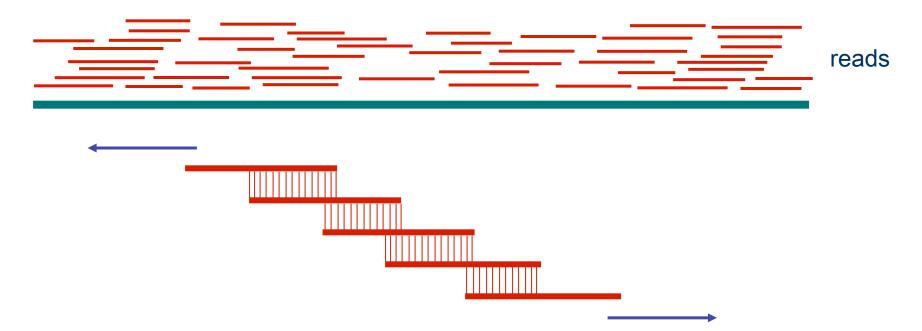
#### Reads assembly

#### Shotgun Sequencing

genomic segment



#### Fragment Assembly



Cover region with ~7-fold redundancy

Overlap reads and extend to reconstruct the original genomic region

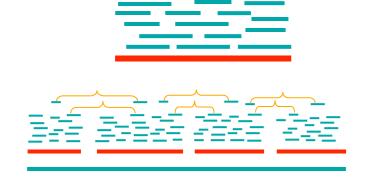
#### Overlap-Layout-Consensus

Assemblers: ARACHNE, PHRAP, CAP, TIGR, CELERA

Overlap: find potentially overlapping reads



Layout: merge reads into contigs and contigs into supercontigs



**Consensus:** derive the DNA sequence and correct read errors

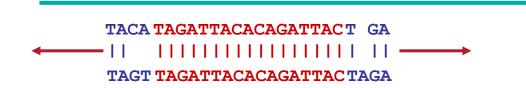
..ACGATTACAATAGGTT..

#### Overlap

- Find the best match between the suffix of one read and the prefix of another
- Due to sequencing errors, need to use dynamic programming to find the optimal overlap alignment
- Apply a filtration method to filter out pairs of fragments that do not share a significantly long common substring

#### Overlapping Reads

- Sort all k-mers in reads  $(k \sim 24)$
- Find pairs of reads sharing a k-mer
- Extend to full alignment throw away if not >95% similar



#### Overlapping Reads and Repeats

- A *k*-mer that appears N times, initiates N<sup>2</sup> comparisons
- For an Alu that appears 10<sup>6</sup> times → 10<sup>12</sup> comparisons too much

#### Solution:

Discard all k-mers that appear more than  $t \times \text{Coverage}$ ,  $(t \sim 10)$ 

#### Finding Overlapping Reads

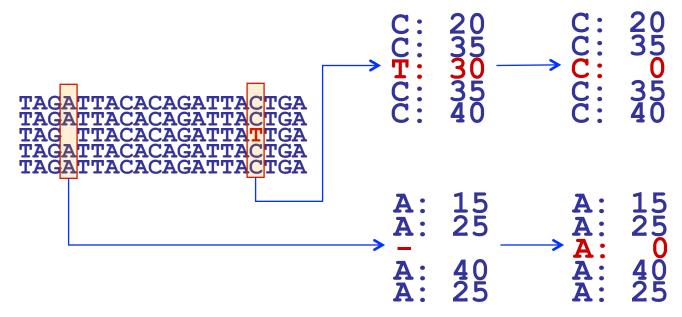
## Create local multiple alignments from the overlapping reads

TAGATTACACAGATTACTGA
TAGATTACACAGATTACTGA
TAG TTACACAGATTACTGA
TAGATTACACAGATTACTGA
TAGATTACACAGATTACTGA
TAGATTACACAGATTACTGA
TAGATTACACAGATTACTGA
TAGATTACACAGATTATTGA
TAGATTACACAGATTACTGA

## Finding Overlapping Reads

(cont'd)

Correct errors using multiple alignment

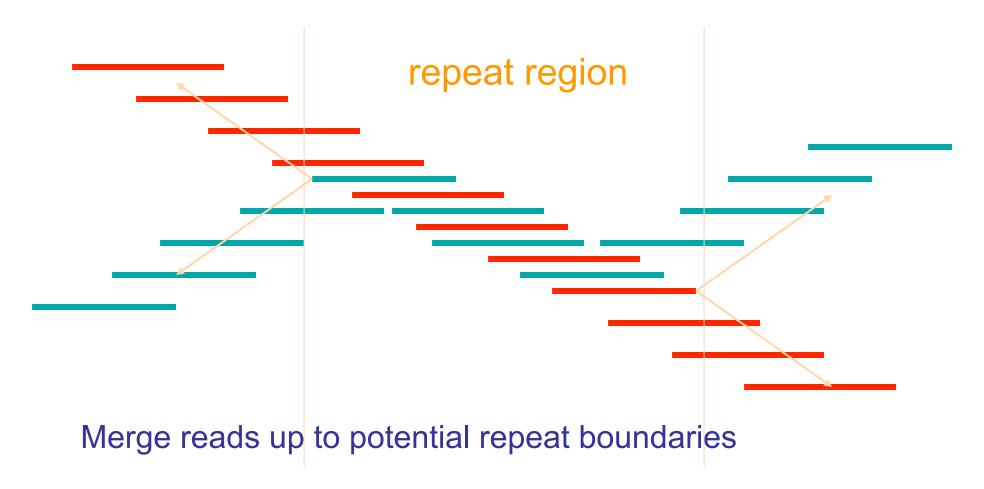


- Score alignments
- Accept alignments with good scores

#### Layout

- Repeats are a major challenge
- Do two aligned fragments really overlap, or are they from two copies of a repeat?
- Solution: repeat masking hide the repeats!!!
- Masking results in high rate of misassembly (up to 20%)
- Misassembly means alot more work at the finishing step

### Merge Reads into Contigs



#### Repeats, Errors, and Contig Lengths

- Repeats shorter than read length are OK
- Repeats with more base pair differencess than sequencing error rate are OK
- To make a smaller portion of the genome appear repetitive, try to:
  - Increase read length
  - Decrease sequencing error rate

#### **Error Correction**

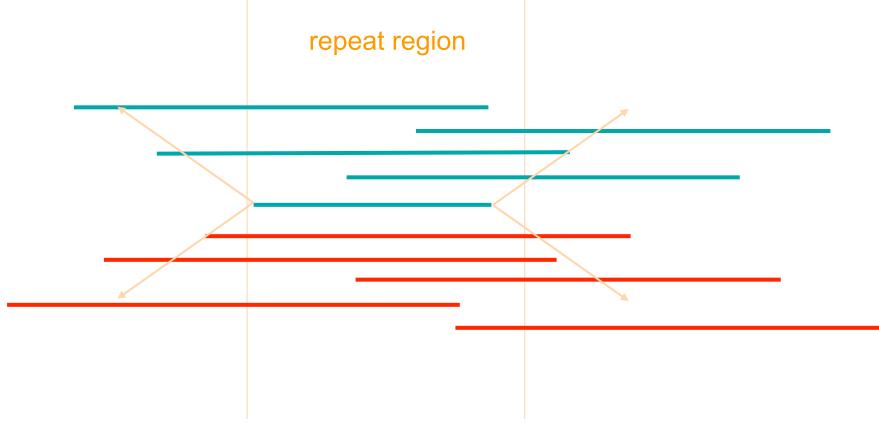
#### Role of error correction:

Discards ~90% of single-letter sequencing errors

decreases error rate

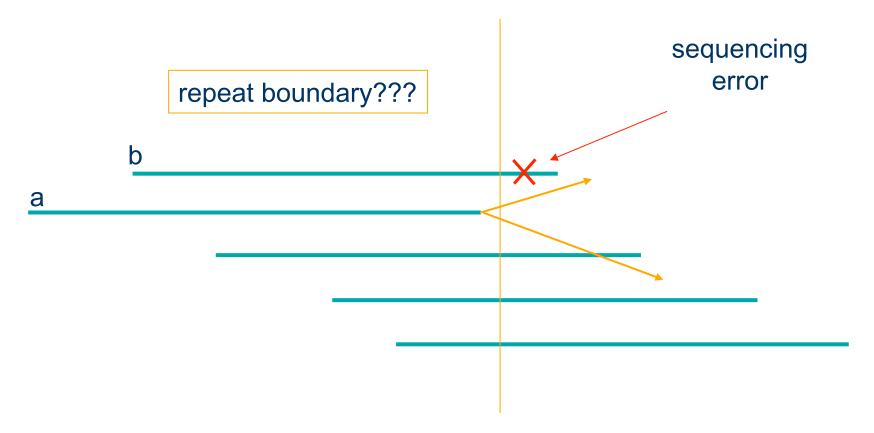
- ⇒ decreases effective repeat content
- ⇒ increases contig length

#### Merge Reads into Contigs (cont'd)



- Ignore non-maximal reads
- Merge only maximal reads into contigs

#### Merge Reads into Contigs (cont'd)



Ignore "hanging" reads, when detecting repeat boundaries

#### Merge Reads into Contigs (cont'd)



Insert non-maximal reads whenever unambiguous

### Link Contigs into Supercontigs



Normal density

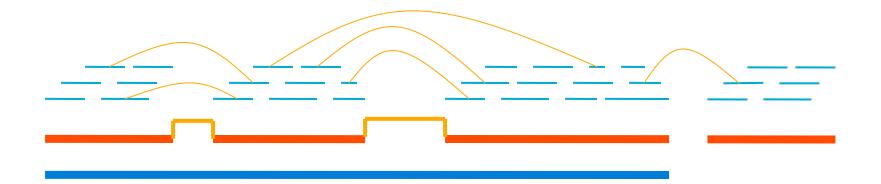
Too dense: Overcollapsed?

Inconsistent links: Overcollapsed?

# Link Contigs into Supercontigs (cont'd)

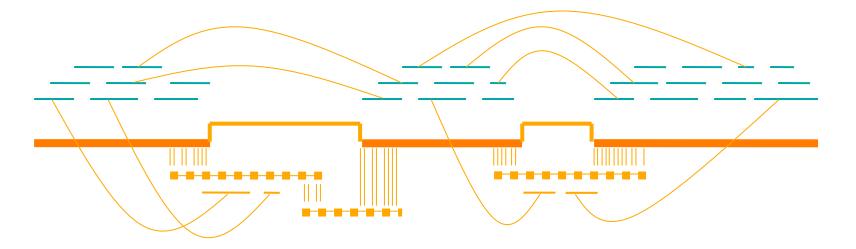
Find all links between unique contigs

Connect contigs incrementally, if ≥ 2 links

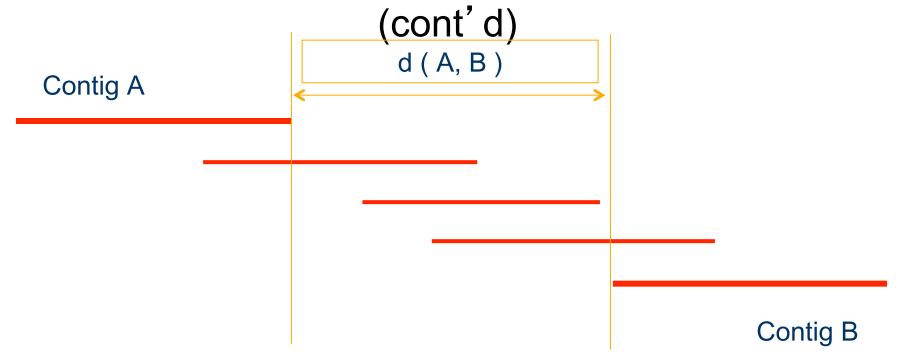


# Link Contigs into Supercontigs (cont'd)

Fill gaps in supercontigs with paths of overcollapsed contigs



### Link Contigs into Supercontigs



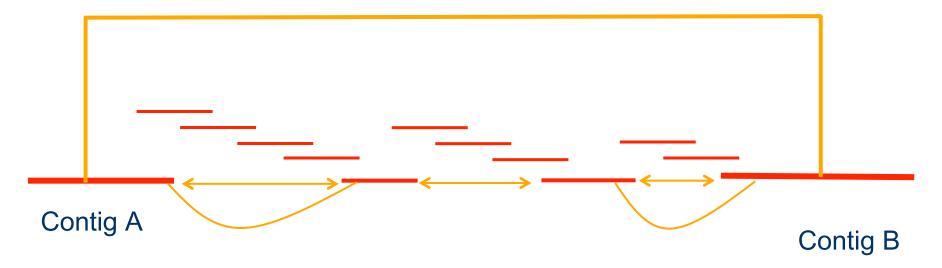
Define G = (V, E)

V := contigs

E := (A, B) such that d(A, B) < C

Reason to do so: Efficiency; full shortest paths cannot be computed

## Link Contigs into Supercontigs (cont'd)



Define T: contigs linked to either A or B

Fill gap between A and B if there is a path in G passing only from contigs in T

#### Consensus

 A consensus sequence is derived from a profile of the assembled fragments

 A sufficient number of reads is required to ensure a statistically significant consensus

Reading errors are corrected

#### Derive Consensus Sequence



Derive multiple alignment from pairwise read alignments

Derive each consensus base by weighted voting

#### Problems with the shotgun approach

