CSE 549: Computational Biology

Computer Science for Biologists Biology





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* ...

What isn't Computer Science?

Don't install operating systems (may develop them)

Don't set up the office network (may study / design network protocols)

Not about Hacking together a program or learning a web-framework — programming \neq CS (may study formal languages and develop new programming languages)

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

Started as a branch of Mathematics — early computing machines Charles Babbage (1791-1871) Ada Lovelace (1791-1871)





Difference engine → Analytical engine* Commonly considered the first "programmer"; developed an algorithm for the analytical engine to compute the Bernoulli numbers

*Analytical engine (would have been the first Turing-complete, general purpose computer) was never completed

en.wikipedia.org

What is "computable"? Early 20th century & birth of "modern" CS









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Kurt Gödel







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Alan Turing





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John von Neumann

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.



Reference genome



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

Next 5 slides courtesy of Ben Langmead

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

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

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Not well-specified.

What makes one genome more likely than another? What constraints do we place on the space of solutions?

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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

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

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

> > AAA AAB ABB BBB BBA BAA ABA BAA

Slide courtesy of Ben Langmead

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

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



Slide courtesy of Ben Langmead

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>) 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 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 strings.

If we give up on finding the *shortest* possible superstring G, 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 — proof for factor 3.5

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

Shortest common superstring: greedy

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.

Slide courtesy of Ben Langmead

Shortest common superstring: greedy



Rounds of merging, one merge per line.

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

Slide courtesy of Ben Langmead

Shortest common superstring: greedy

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 ——

Results from quiz 1 & review

