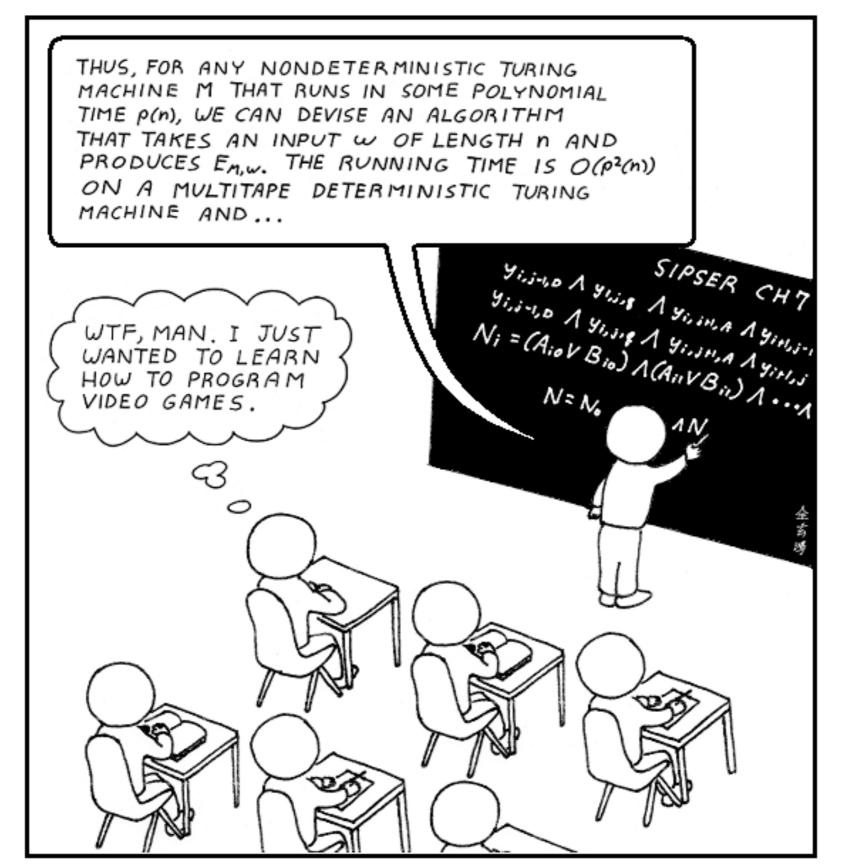
CSE 549: Computational Biology

Computer Science for Biologists Biology





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)

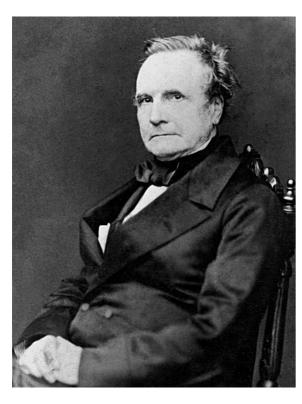
Not about Hacking together a program or learning a web-framework — programming ≠ CS (may study formal languages and develop new programming languages)

^{*}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











Kurt Gödel









Kurt Gödel



Alan Turing







Kurt Gödel



Alan Turing



Alonzo Church





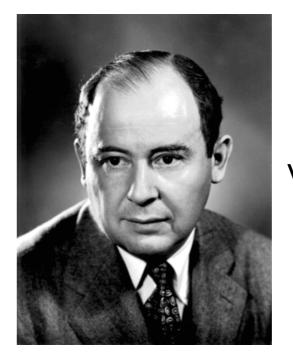
Kurt Gödel



Alan Turing



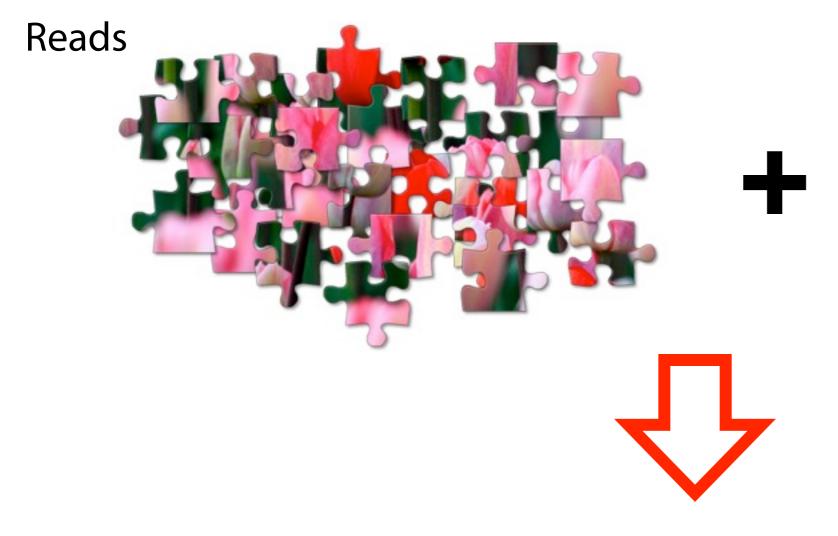
Alonzo Church



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.









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: GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT
GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT
GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT
GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT

Fragment: GGCGTCTA TATCTCGG CTCTAGGCCCCTC ATTTTTT
GGC GTCTATAT CTCGGCTCTAGGCCCCTCA TTTTTT
GGCGTC TATATCT CGGCTCTAGGCCCCT 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
CTCGGCTCTAGGCCCCTCATTTTT
TATCTCGACTCTAGGCCCCTCA
TATCTCGACTCTAGGCC
TCTATATCTCGGCTCTAGG
GGCGTCTATATCTCG
GGCGTCTATATCTCG
GGCGTCTATATCT

From these

→ GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTT1

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

Reconstruct this

CTAGGCCCTCAATTTTT
GGCGTCTATATCT
CTCTAGGCCCTCAATTTTT
TCTATATCTCGGCTCTAGG
GGCTCTAGGCCCTCATTTTT
CTCGGCTCTAGCCCCTCATTTTT
TATCTCGACTCTAGGCCCTCA
GGCGTCGATATCT
TATCTCGACTCTAGGCC
GGCGTCTATATCTCG

From these

→ GGCGTCTATATCTCGGCTCTAGGCCCTCATTTTTT

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

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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)

1S

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

SCS(S): AAABBBABAA

─10 **─**

AAA
ABB
ABB
BBA
BAB
ABA
BAA

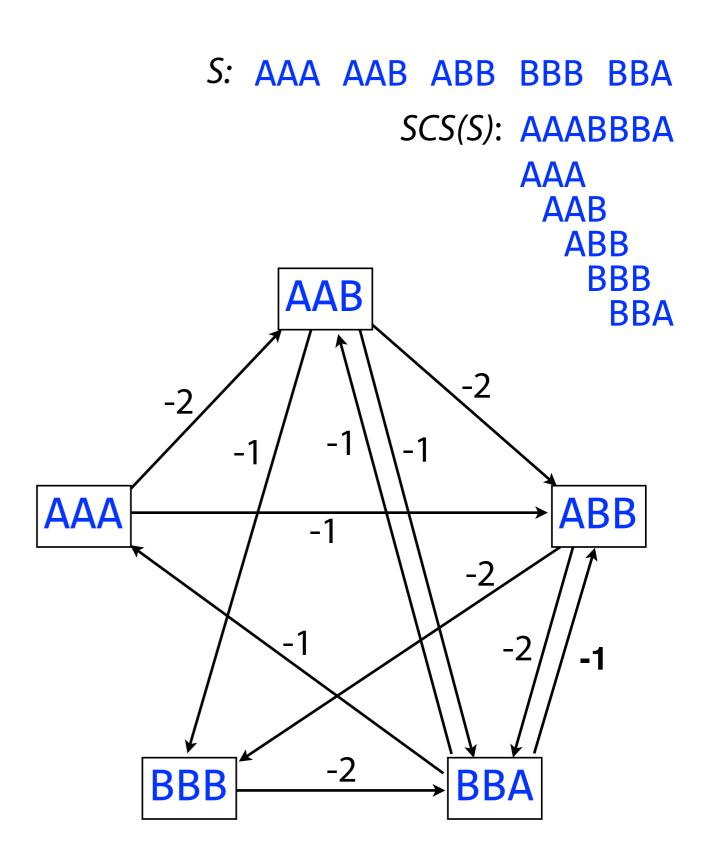
Shortest common superstring

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!

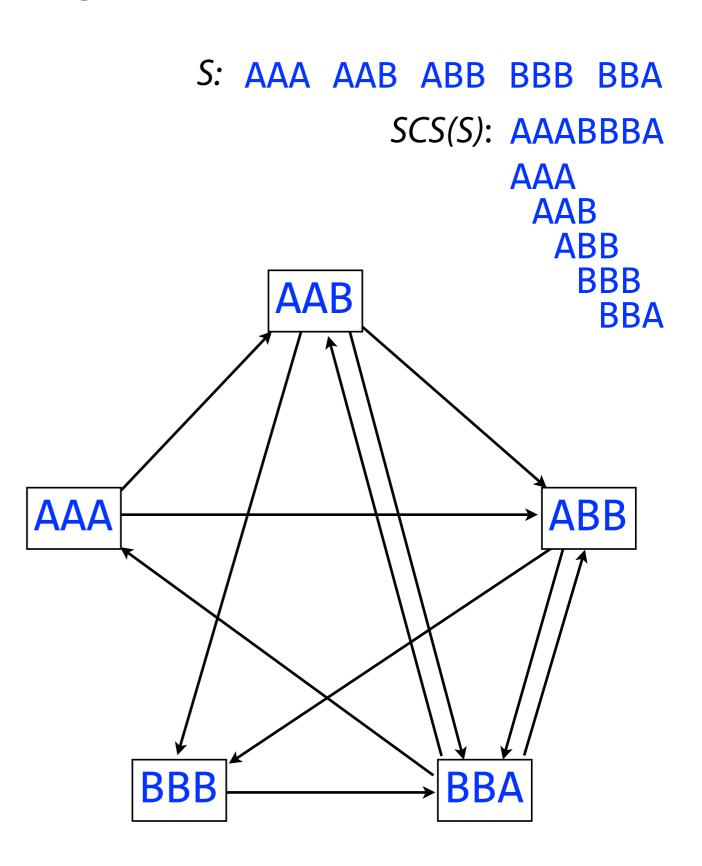


Shortest common superstring

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



Shortest common superstring & friends

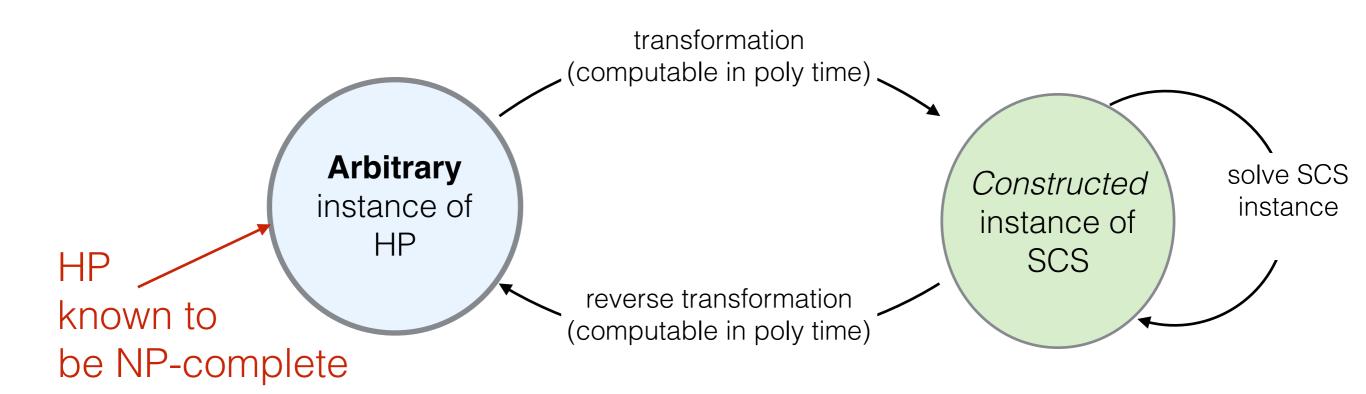
Traveling Salesman, Hamiltonian Path, and Shortest Common Superstring are all NP-hard

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)

Important note: The fact that we modeled SCS as NP-hard 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.

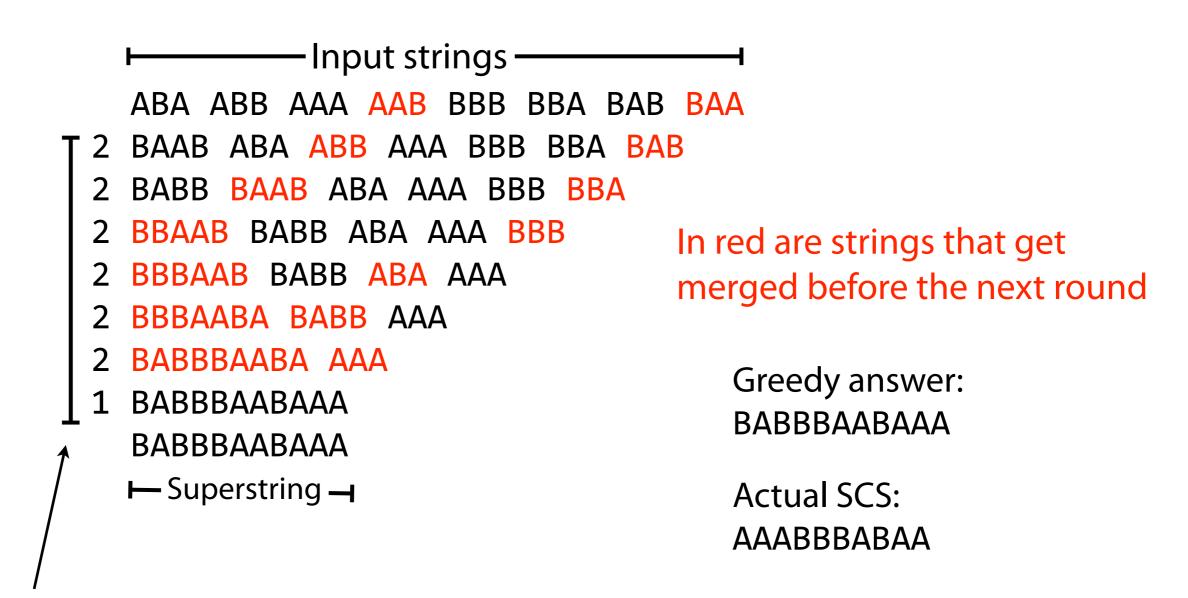
Greedy conjecture factor of 2-OPT *is* the worst case

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):

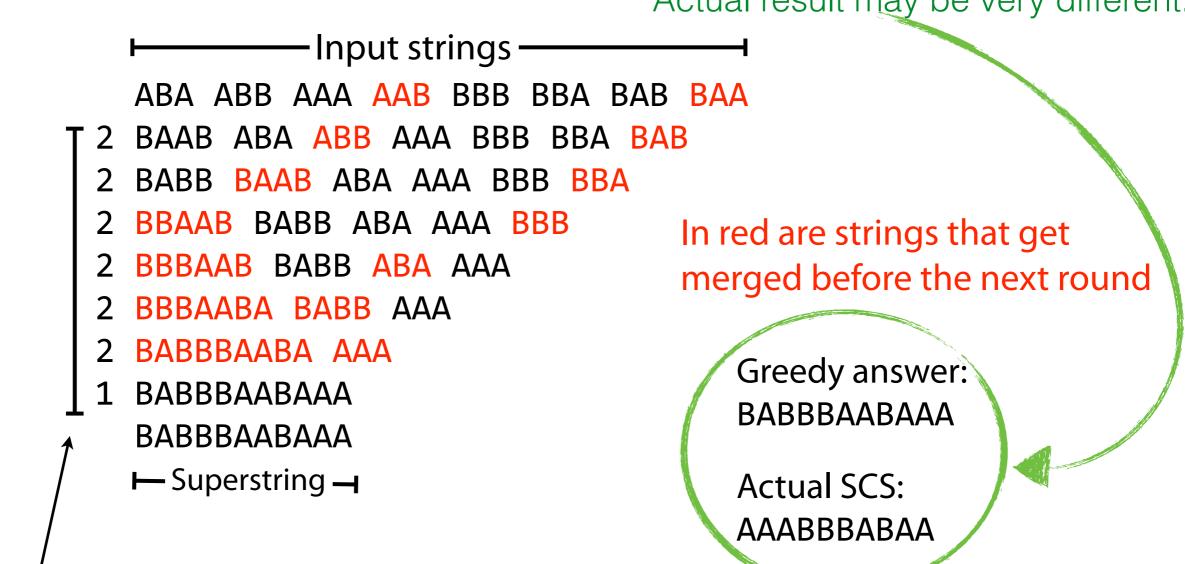


Rounds of merging, one merge per line. Number in first column = length of overlap merged before that round.

Shortest common superstring: greedy

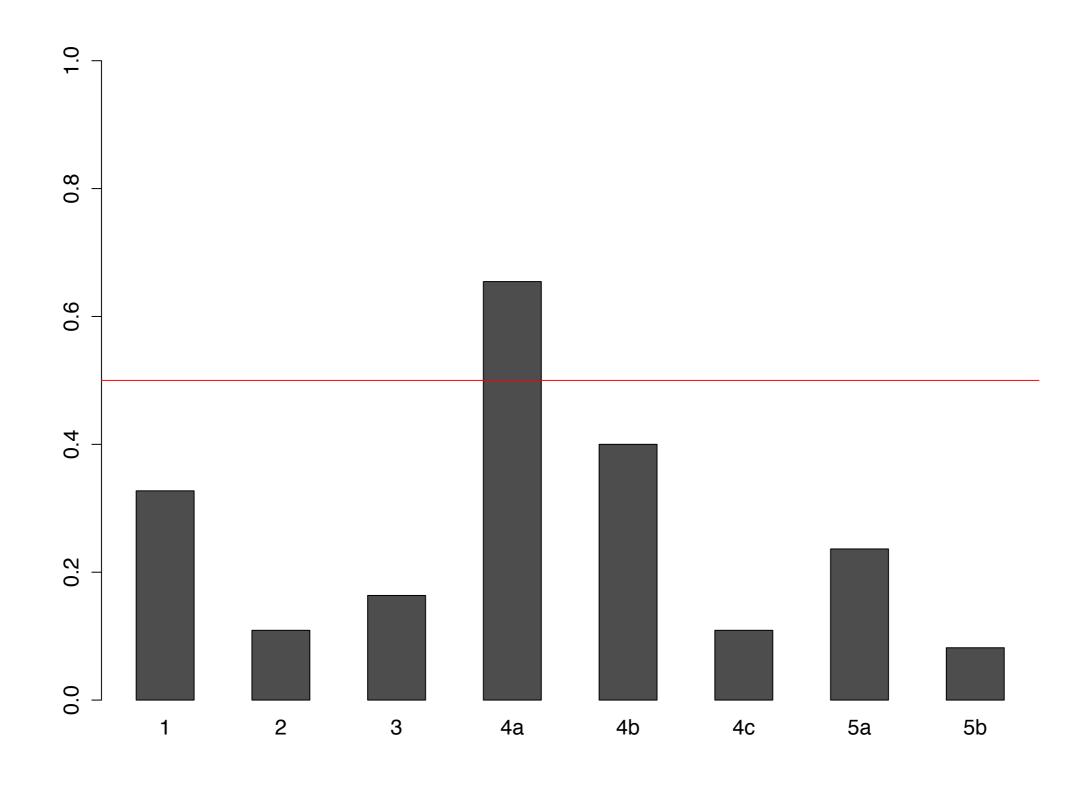
Greedy-SCS algorithm in action (l = 1):

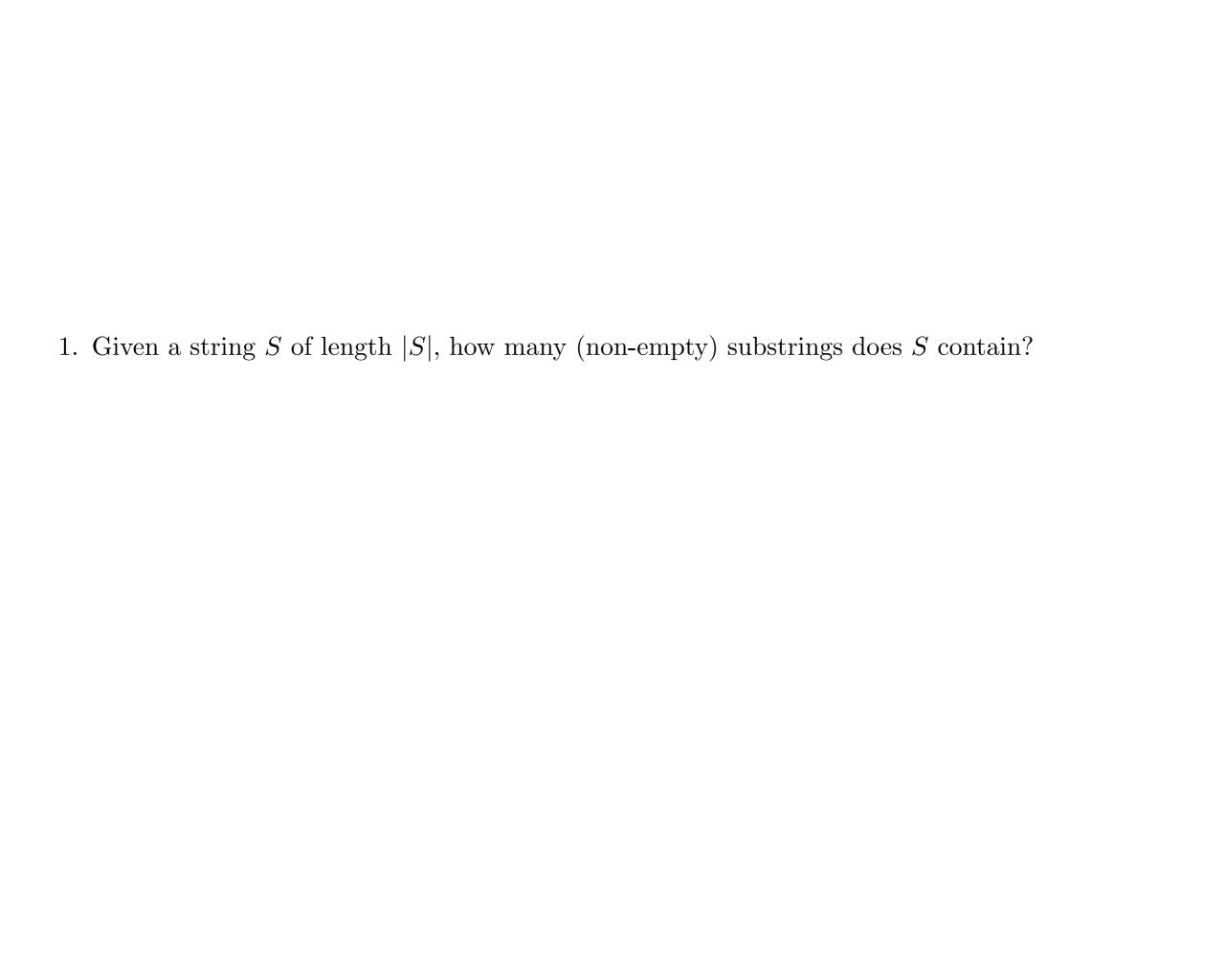
Note: approx. guarantee is on length of the superstring Actual result may be very different.



Rounds of merging, one merge per line. Number in first column = length of overlap merged before that round.

Results from survey & review





1. Given a string S of length |S|, how many (non-empty) substrings does S contain? ${|S| \choose 2}$

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```
for each vertex v in V:
    sp[v] = infinity

sp[s] = 0
for each vertex v in topo(V):
    for each w in predecessors(v):
        sp[v] = min(sp[w] + 1, sp[v])
return sp[t]
```

4. I am given two coins, which I will flip simultaneously. Let X = H be the event that the first coin comes up heads and X = T be the event that the first coin comes up tails. Let Y = H and Y = T be defined similarly for the second coin. If you know that $\Pr(X = H, Y = H) = 0.25$ and $\Pr(X = T, Y = H) = 0.5$ and $\Pr(X = H, Y = T) = 0.05$.

(a) What is
$$Pr(X = T, Y = T)$$
?

(b) What is
$$Pr(X = T)$$
?

	Y=H	Y=T
X=H	0.25	0.5
X=T	0.05	?

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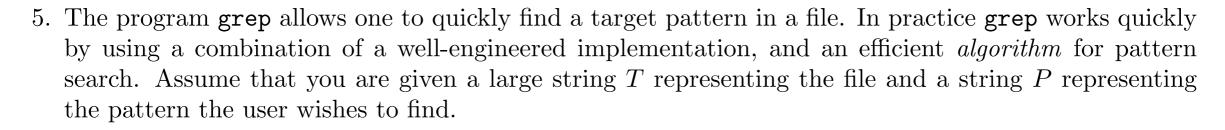
(b) What is Pr(X = T)?

	Y=H	Y=T
X=H	0.25	0.5
X=T	0.05	(1-0.8) = 0.2

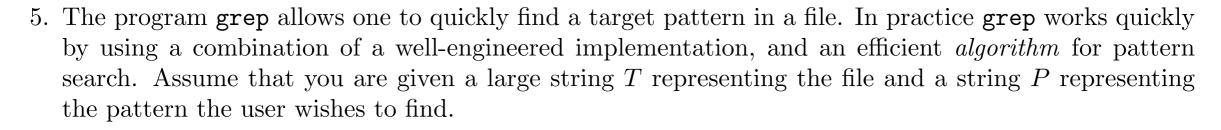
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 - (b) What is Pr(X = T)? Pr(X = T, Y = H) + Pr(X = T, Y = T) = 0.5 + 0.2 = 0.7
 - (c) Are the two coin flips statistically independent (say why or why not)?

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 - (c) Are the two coin flips statistically independent (say why or why not)? No, since e.g. $\Pr(X = T, Y = T) \neq \Pr(X = T) \cdot \Pr(Y = T)$.



- (a) What algorithm might you use to search for the pattern P in T?
- (b) What is the running time of your algorithm in terms of |T| and |P|?



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- 5. The program grep allows one to quickly find a target pattern in a file. In practice grep works quickly by using a combination of a well-engineered implementation, and an efficient algorithm for pattern search. Assume that you are given a large string T representing the file and a string P representing the pattern the user wishes to find.
 - (a) What algorithm might you use to search for the pattern P in T? Knuth-Morris-Pratt
 - (b) What is the running time of your algorithm in terms of |T| and |P|? O(|T| + |P|)