# CSE 421: Introduction to Algorithms I: Overview

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University of Washington Computer Science & Engineering

#### CSE 421, Su '12: Introduction to Algorithms

#### CSE Home

Administrative	Lecture:	LOW 206 (schematic)	MW 10:50-12:20		
Schedule & Reading	Instructor TA:	: <u>Larry Ruzzo</u> , ruzzo <sup>ę</sup> cs Tamara Bonaci, tbonaci <sup>ę</sup> uw	Office Hours Lo W 1:00- 2:00 V.edu M	6298	
Lecture Notes 1: Overview & Example 2: Analysis	Course En	NW.CS.Wash	uu gout probably sh بر	eneral interest student/staff Q&A ould <u>change their default subscrip</u>	
Course Email/BBoard Subscription Options Class List Archive E-mail Course Staff GoPost BBoard Lecture Notes 1: Overview & Example 2: Analysis Mttp:///////////////////////////////////					
Prerequisites: either <u>CSE 312</u> or <u>CSE 322</u> ; either <u>CSE 326</u> or <u>CSE 332</u> .					
	Credits: 3				
	Grading: Homework, Final. Homework will be a mix of paper & pencil exercises and programing. O Late Policy: Unless otherwise announced, weekly homeworks will be due by 4:00PM on Thursdays electronically, including scanned versions of handwritten papers, via the Catalyst drop box link at left				
Extra Credit: Assignments may include "extra credit" sections. These will enrich your understanding glory, not the points, and don't start extra credit until the basics are complete.					
Textbook: <u>Algorithm Design</u> by Jon Kleinberg and Eva Tardos. Addison Wesley, 2006. (Availa					

### What you have to do

#### Homework (~60% of grade)

Programming?

perhaps some small projects

Written homework assignments

English exposition and pseudo-code Analysis and argument as well as design

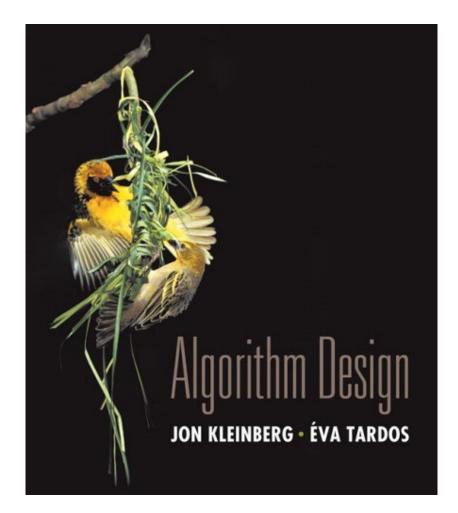
Final Exam



Late Policy:

Papers and/or electronic turnins generally due Thursdays by 4:00 pm; minus 20% per day thereafter

#### Textbook



<u>Algorithm Design</u> by Jon Kleinberg and <u>Eva Tardos</u>. Addison Wesley, 2006.

#### What the course is about

Design of Algorithms design methods common or important types of problems analysis of algorithms - efficiency correctness proofs

### What the course is about

Complexity, NP-completeness and intractability solving problems in principle is not enough algorithms must be efficient some problems have no efficient solution NP-complete problems important & useful class of problems whose solutions (seemingly) cannot be found efficiently, but *can* be

checked easily

# Very Rough Division of Time

#### Algorithms (6-7 weeks)

Analysis

Techniques: greedy, divide&conquer, dynamic programming Applications: graph algorithms, flows & matchings Complexity & NP-completeness (2-3 weeks)

Check online schedule page for (evolving) details



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CSE 417, Wi '06: Approximate Schedule

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		Due	Lecture Topic	Reading
Week 1	М		Holiday	
1/2-1/6	W		Intro, Examples & Complexity	Ch. 1; Ch. 2
	F		Intro, Examples & Complexity	
Week 2 1/9-1/13	м		Intro, Examples & Complexity	
	W		Graph Algorithms	Ch. 3
	F		Graph Algorithms	1

## Complexity Example

Cryptography (e.g., RSA, SSL in browsers)

Secret: p,q prime, say 512 bits each

Public: n which equals  $p \ge q$ , 1024 bits

In principle

```
there is an algorithm that given n will find p and q:
try all 2^{512} > 1.3 \times 10^{154} possible p's: kinda slow...
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In practice

no fast algorithm known for this problem (on non-quantum computers) security of RSA depends on this fact

("quantum computing": strongly driven by possibility of changing this)

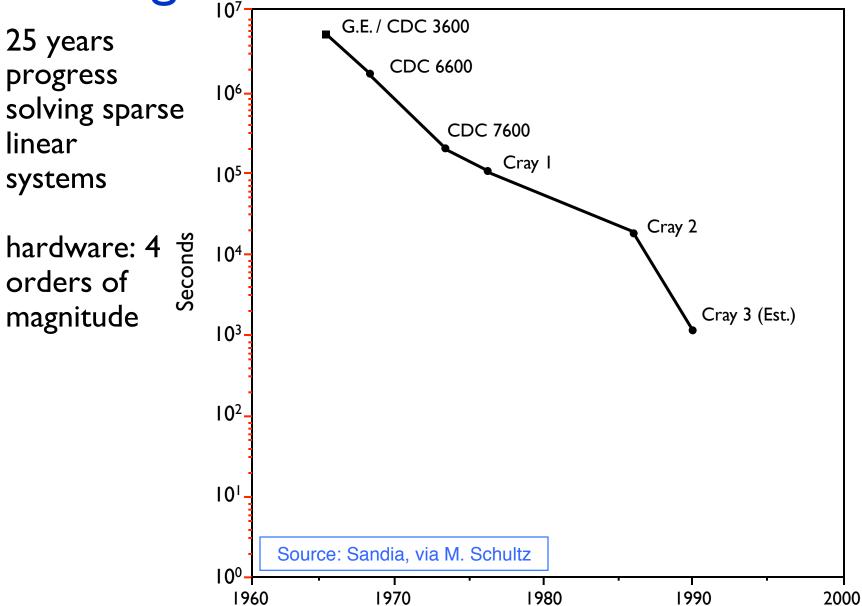
### Algorithms versus Machines

We all know about Moore's Law and the exponential improvements in hardware...

Ex: sparse linear equations over 25 years

10 orders of magnitude improvement!

# Algorithms or Hardware?



#### Algorithms or Hardware? $10^{7}$ G.E. / CDC 3600 25 years CDC 6600 progress 106 solving sparse CDC 7600 linear Cray I 105 systems Cray 2 hardware: 4 Seconds 104orders of magnitude Cray 3 (Est.) 103 Sparse G.E. software: 6 Gauss-Seidel 102 orders of magnitude

Source: Sandia, via M. Schultz

1970

SOR

1980

CG

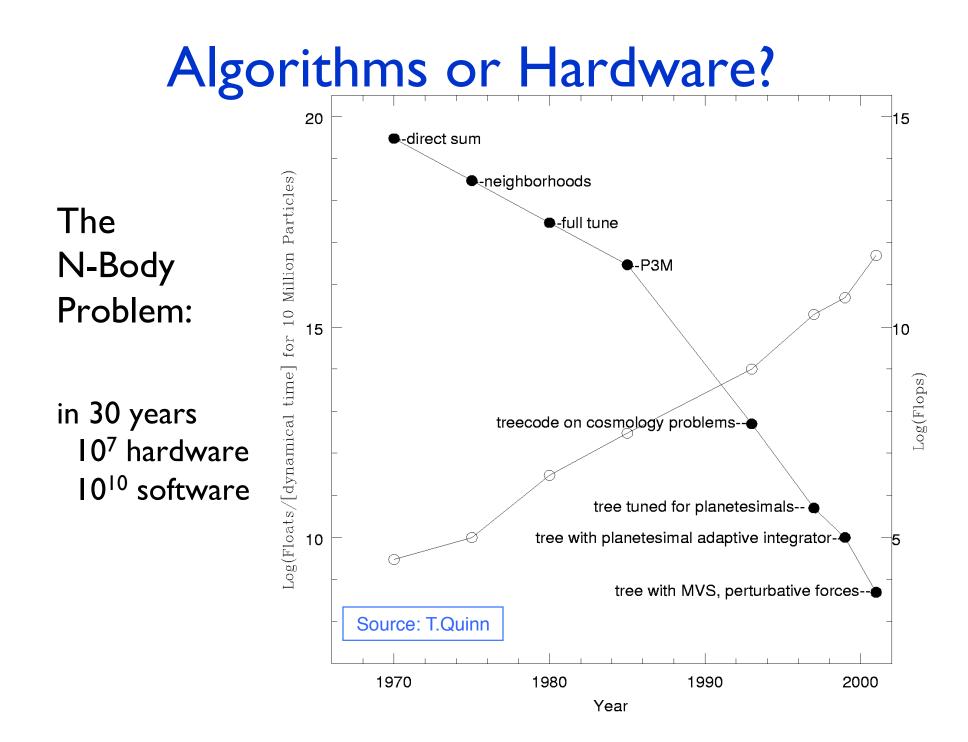
1990

101

100

1960

2000



#### Algorithm: definition

Procedure to accomplish a task or solve a well-specified problem

Well-specified: know what all possible inputs look like and what output looks like given them "accomplish" via simple, well-defined steps Ex: sorting names (via comparison) Ex: checking for primality (via +, -, \*, /,  $\leq$ )



Correctness often subtle Analysis often subtle Generality, Simplicity, 'Elegance' Efficiency

time, memory, network bandwidth, ...

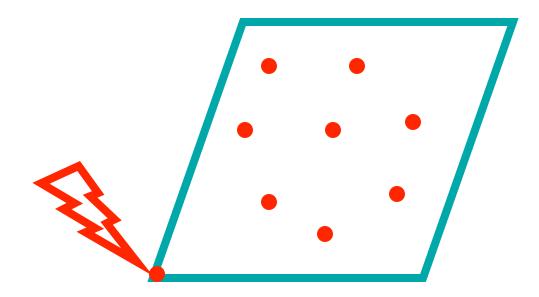
## Algorithms: a sample problem

Printed circuit-board company has a robot arm that solders components to the board

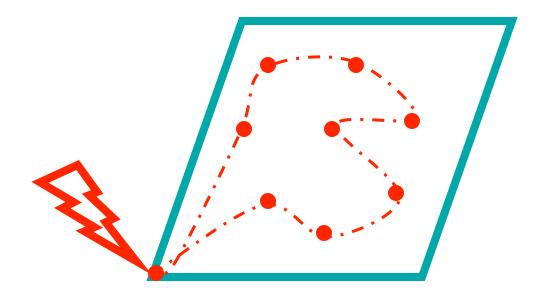
Time: proportional to total distance the arm must move from initial rest position around the board and back to the initial position

For each board design, find best order to do the soldering

#### Printed Circuit Board



#### Printed Circuit Board



#### A Well-defined Problem

Input: Given a set S of *n* points in the plane Output: The shortest cycle tour that visits each point in the set S.

Better known as "TSP"

How might you solve it?

# Nearest Neighbor Heuristic

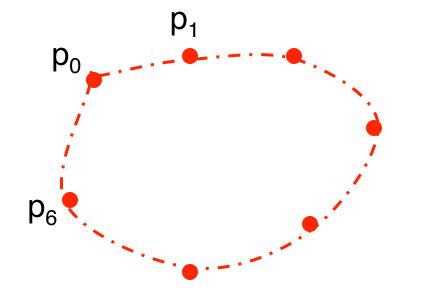
Start at some point  $P_0$ Walk first to its nearest neighbor  $P_1$ 

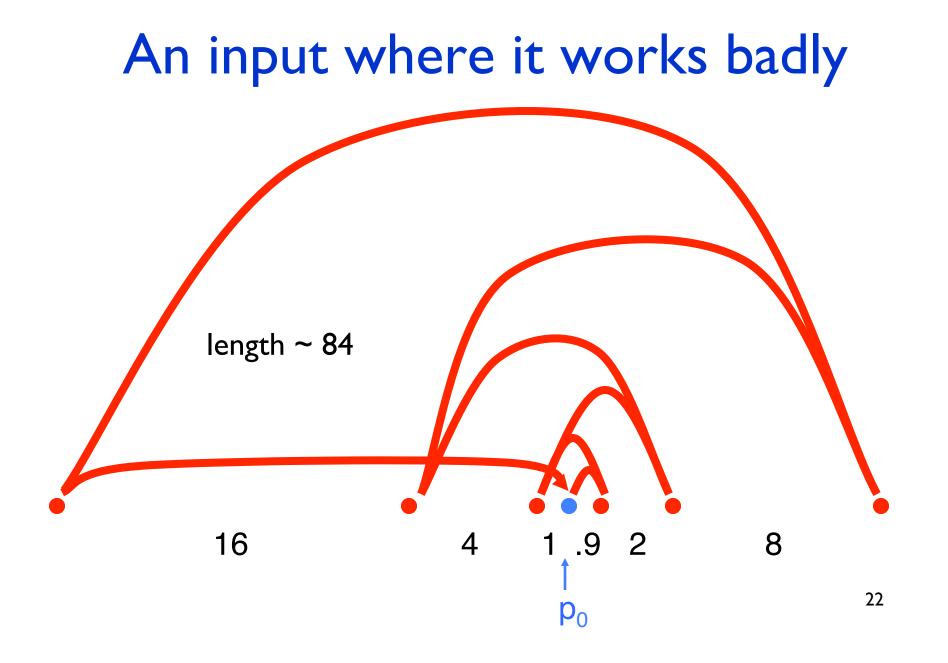
#### heuristic:

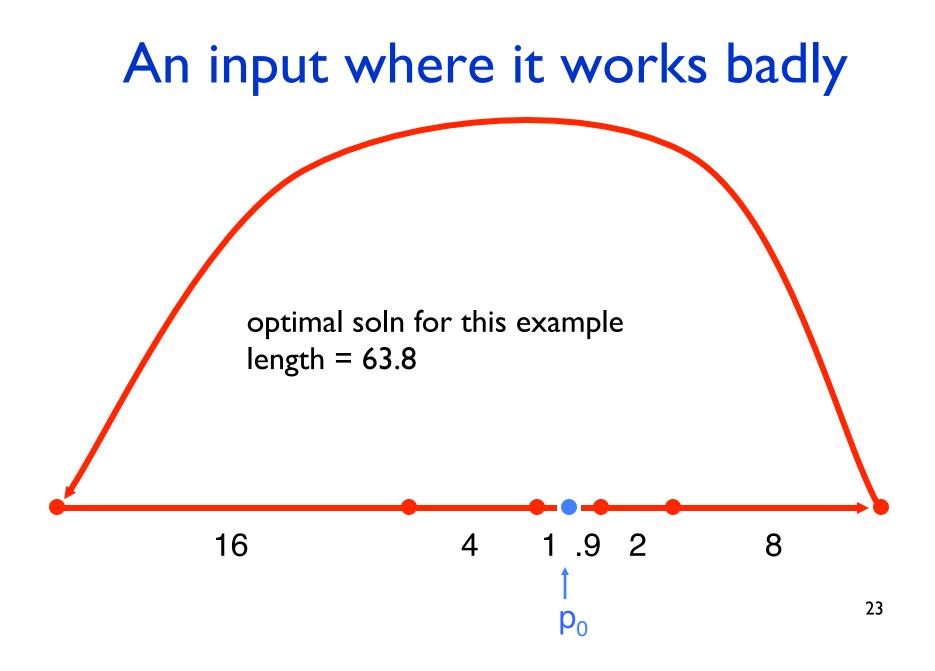
A rule of thumb, simplification, or educated guess that reduces or limits the search for solutions in domains that are difficult and poorly understood. May be good, but usually *not* guaranteed to give the best or fastest solution.

Repeatedly walk to the nearest unvisited neighbor  $p_2$ , then  $p_3$ ,... until all points have been visited Then walk back to  $p_0$ 

#### Nearest Neighbor Heuristic





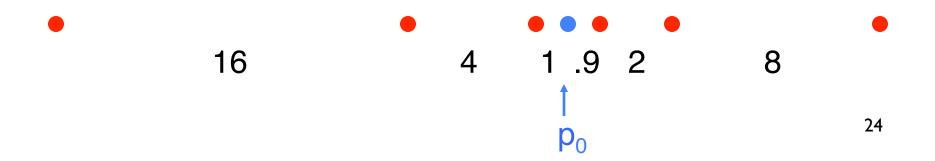


#### Revised idea - Closest pairs first

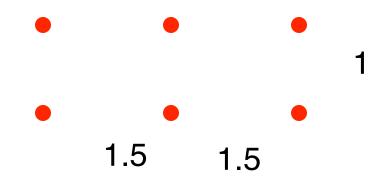
Repeatedly join the closest pair of points

(s.t. result can still be part of a single loop in the end. I.e., join endpoints, but not points in middle, of path segments already created.)

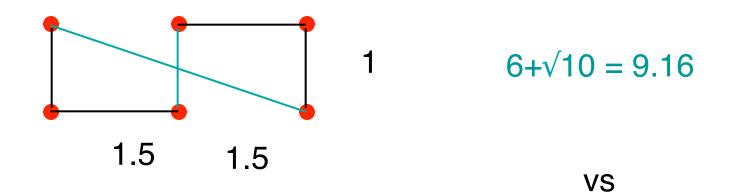
How does this work on our bad example?

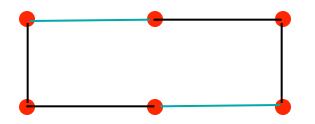


### Another bad example



#### Another bad example





### Something that works

"Brute Force Search":

For each of the n! = n(n-1)(n-2)...I orderings of the points, check the length of the cycle you get

Keep the best one

#### Two Notes

#### The two incorrect algorithms were greedy

- Often very natural & tempting ideas
- They make choices that look great "locally" (and never reconsider them)
- When greed works, the algorithms are typically efficient
- BUT: often does not work you get boxed in
- Our correct alg avoids this, but is incredibly slow
  - 20! is so large that checking one billion orderings per second would take 2.4 billion seconds (around 70 years!) And growing: n! ~  $\sqrt{2 \pi n} \cdot (n/e)^n \sim 2^{O(n \log n)}$

## The Morals of the Story

Algorithms are important Many performance gains outstrip Moore's law Simple problems can be hard Factoring, TSP Simple ideas don't always work Nearest neighbor, closest pair heuristics Simple algorithms can be very slow Brute-force factoring, TSP Changing your objective can be good Guaranteed approximation for TSP And: for some problems, even the best algorithms are slow