

What is the computational  
cost of automating  
brilliance or serendipity?  
(P vs NP question and related musings)

COS 116: 4/12/2006

Adam Finkelstein

# Combination lock

Why is it secure?  
(Assume it cannot be picked)



Ans: Combination has 3 numbers 0-35...  
thief must try  $36^3 = 46,656$  combinations



# Exponential running time

$2^n$  time to solve instances of “size”  $n$

Increase  $n$  by 1  $\rightarrow$  running time doubles!

Main fact to remember:

*For  $n = 300$ ,*

*$2^n >$  number of atoms in the visible universe.*



# Boolean satisfiability

$$(A + B + C) \cdot (\bar{D} + F + G) \cdot (\bar{A} + G + K) \cdot (\bar{B} + P + Z) \cdot (C + \bar{U} + \bar{X})$$

- Does it have a satisfying assignment?
- What if instead we had 100 variables?
- 1000 variables?
- How long will it take to determine the assignment?

# Discussion

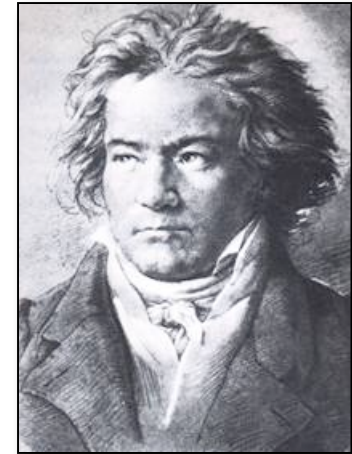
Is there an inherent difference between

being creative / brilliant

and

being able to appreciate creativity / brilliance?

What is a computational analogue of this phenomenon?



# A Proposal

Brilliance = Ability to find  
“needle in a haystack”

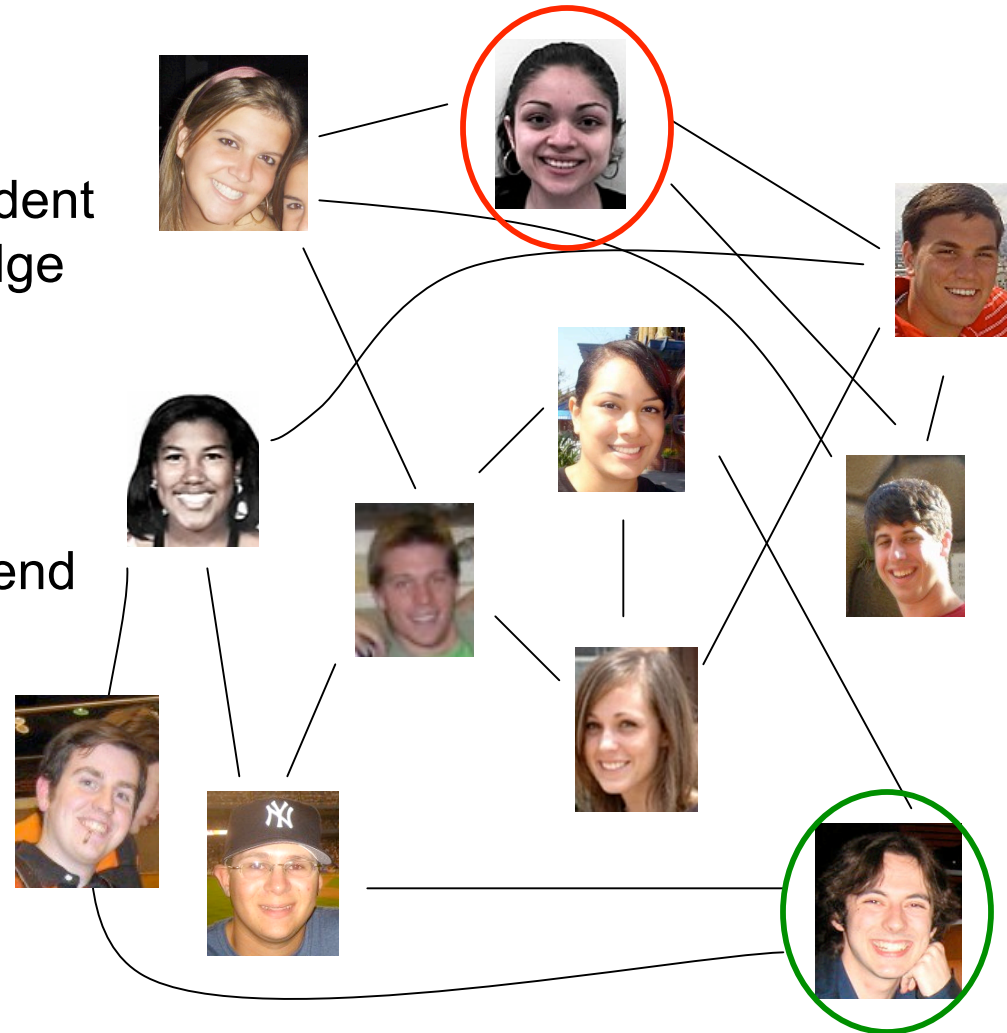
Beethoven found  
“satisfying assignments”  
to our neural circuits  
for music appreciation

Comments??



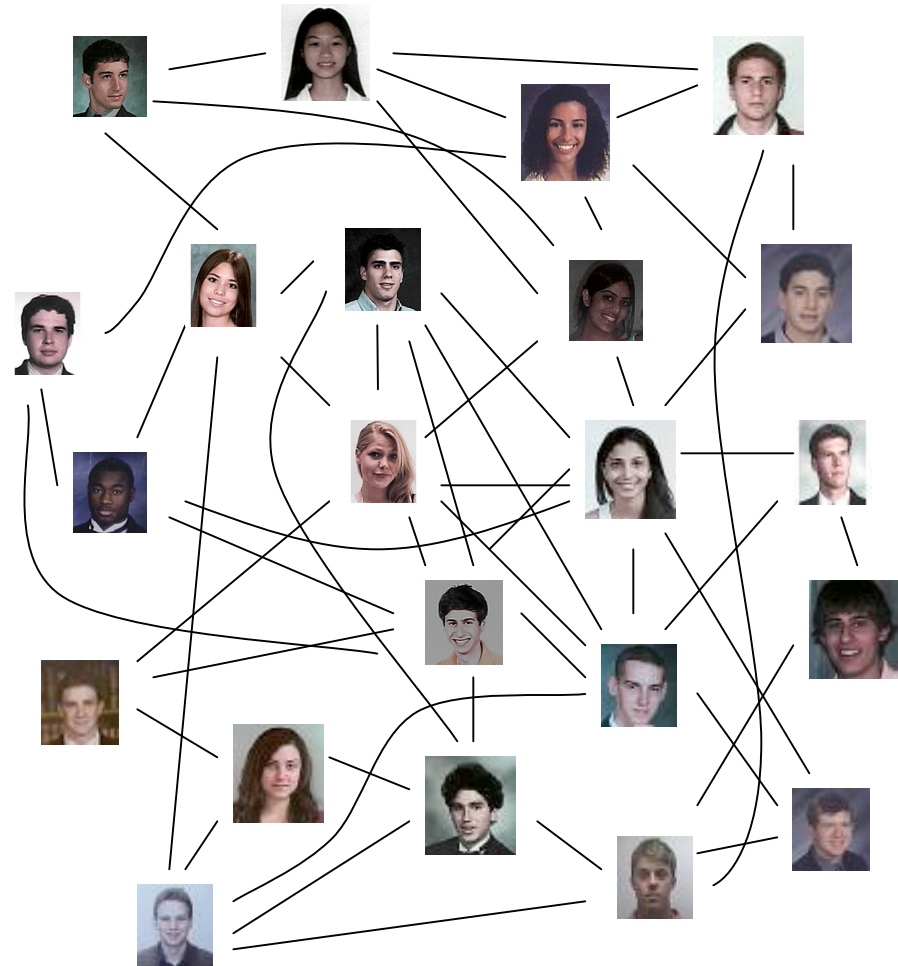
# Rumor mill problem

- Social network for COS 116
- Each node represents a student
- Two nodes connected by edge if those students are friends
- Johanna starts a rumor
- Will it reach Kieran?
- Suggest an algorithm
- How does running time depend on network size?
- Internet servers solve this problem all the time (“traceroute” in Lab 9).



# CLIQUE Problem

- In this social network, is there a CLIQUE with 5 or more students?
- CLIQUE: Group of students, every pair of whom are friends
- What is a good algorithm for detecting cliques?
- How does efficiency depend on network size and desired clique size?



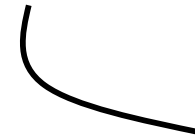


# Harmonious Dorm Floor


Given: Social network involving  $n$  students.

Edges correspond to pairs of students who don't get along.

Decide if there is a set of  $k$  students who would make a harmonious group (everybody gets along).



Just the Clique problem in disguise!



# Exhaustive Search / Combinatorial Explosion

Naïve algorithms for many “needle in a haystack” tasks involve checking all possible answers → exponential running time.

- Ubiquitous in the computational universe
- Can we design smarter algorithms?

# Traveling Salesman Problem (aka UPS Truck problem)

- Input:  $n$  points and all pairwise inter-point distances, and a distance  $k$
- Decide: is there a path that visits all the points (“salesman tour”) whose total length is at most  $k$ ?



# Finals scheduling



- Input:  $n$  students,  $k$  classes, enrollment lists,  $m$  time slots in which to schedule finals
- Define “conflict”: a student is in two classes that have finals in the same time slot
- Decide: if schedule with at most  $C$  conflicts exists?

# The P vs NP Question



- P: problems for which solutions can be found in polynomial time ( $n^c$  where  $c$  is a fixed integer and  $n$  is “input size”). Example: Rumor Mill
- NP: problems where a *good solution* can be checked in  $n^c$  time. Examples: Boolean Satisfiability, Traveling Salesman, Clique
- Question: **Is P = NP?**  
“Can we automate brilliance?”  
  
(Note: Choice of computational model --- Turing machine, pseudocode, etc. --- irrelevant.)

# NP-complete Problems

Problems in NP that are “the hardest”

- If they are in P then so is **every** NP problem.

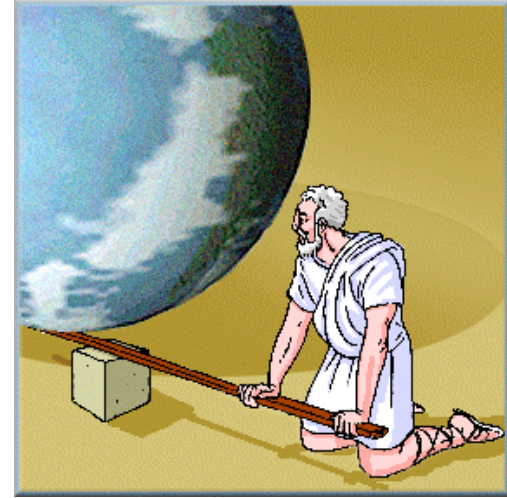
Examples: Boolean Satisfiability, Traveling Salesman, Clique, Finals Scheduling, 1000s of others

How could we possibly prove these problems are “the hardest”?



# “Reduction”

“If you give me a place to stand, I will move the earth.”  
– Archimedes (~ 250BC)



“If you give me a polynomial-time algorithm for Boolean Satisfiability, I will give you a polynomial-time algorithm for every NP problem.” --- Cook, Levin (1971)


“Every NP problem is a satisfiability problem in disguise.”





# Dealing with NP-complete problems

 **Heuristics** (algorithms that produce reasonable solutions in practice)

 **Approximation algorithms** (compute provably near-optimal solutions)





# *Computational Complexity Theory:* Study of Computationally Difficult problems.

A new lens on the world?



- Study matter → look at mass, charge, etc.
- Study processes → look at computational difficulty

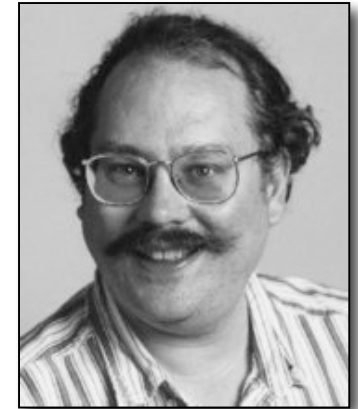
# Example 1: Economics

General equilibrium theory:

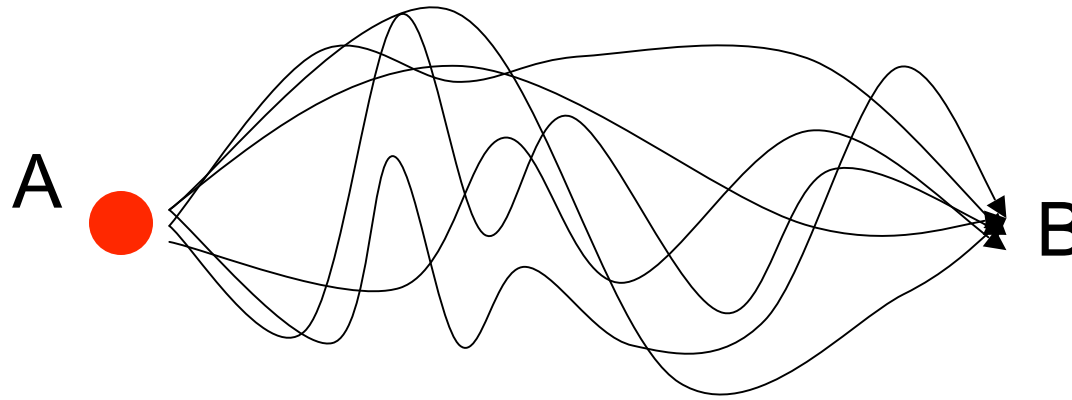
- Input:  $n$  agents, each has some initial endowment (goods, money, etc.) and preference function
- General equilibrium: system of prices such that for every good, demand = supply.
- Equilibrium exists [Arrow-Debreu, 1954]. Economists assume markets find it (“invisible hand”)
- But, no known efficient algorithm to compute it. How does the market compute it?



## Example 2: Quantum Computation



Peter Shor



- Central tenet of quantum mechanics: when a particle goes from A to B, it takes all possible paths all at the same time
- [Shor'97] Can use quantum behavior to efficiently factor integers (and break cryptosystems!)
- Can quantum computers be built, or is quantum mechanics not a correct description of the world?

# Example 3: Artificial Intelligence

What is computational complexity of language recognition?

Chess playing?

Etc. etc.



Potential way to show the brain is not a computer:  
*Show it routinely solves some problem that provably takes exponential time on computers.*



# Why is P vs NP a Million-dollar open problem?

- If  $P = NP$  then Brilliance becomes routine (best schedule, best route, best design, best math proof, etc...)
- If  $P \neq NP$  then we know something *new and fundamental* not just about computers but about the world (akin to “Nothing travels faster than light”).

# Next time: Cryptography (practical use of computational complexity)

