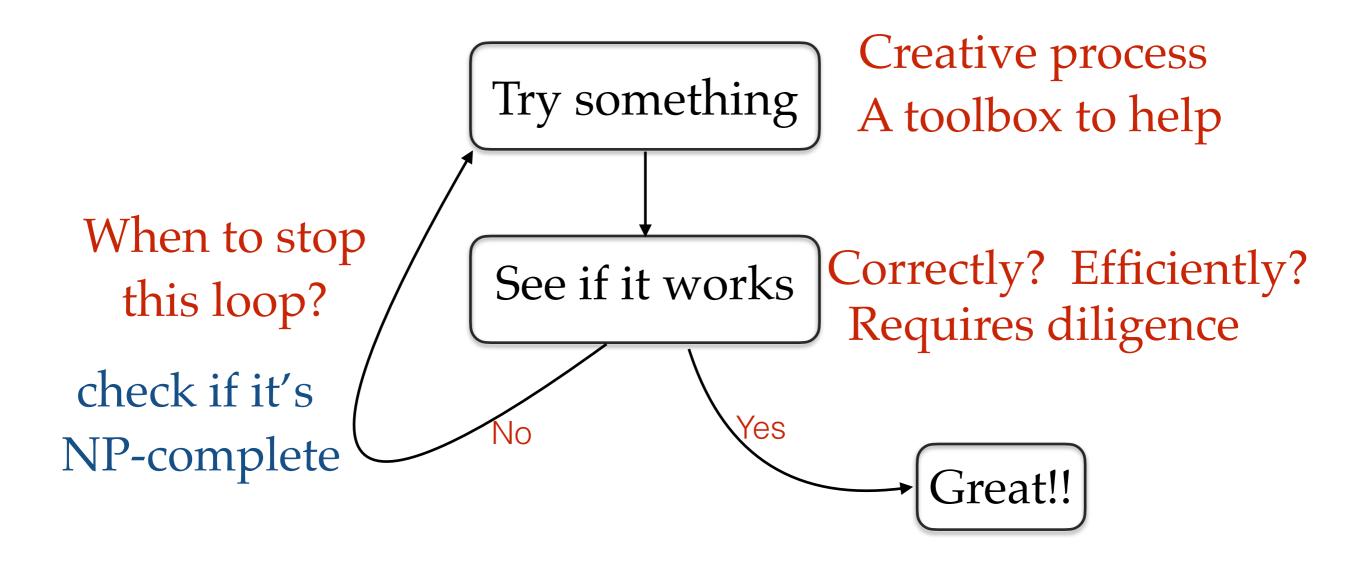
NP, NP-completeness and a Million Dollar Question

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Objectives

How to design algorithms.



History of P

- Notion of Efficient Algorithms there since ancient times
- Addition, Multiplication, GCD, Repeated squaring (Pingala), Astronomical calculations.
- [1950s] Dynamic Programming, Shortest Path, Simplex algorithm, Minimum spanning tree
- [1960s] FFT, Scheduling, Network flow, bipartite matching, and related combinatorial problems
- Doing better than brute force search

P (Polynomial time solvable)

- Edmonds [1965] proposed polynomial time as a characterization of efficient computation
- "It is by no means obvious whether or not there exists an algorithm whose difficulty increases only algebraically with the size of the graph"
- "For practical purposes the difference between algebraic and exponential is more crucial than between finite and non-finite."

P (Polynomial time solvable)

- Why polynomial time?
 - if a procedure is considered efficient, running it n times might also be considered efficient.
 - Polynomial time remains independent of computation model.
 - Another perspective: if you double the input size, the running time gets multiplied by a constant.
 - If the running time is n¹⁰⁰, is it still efficient?
 - In reality, we never get such running times.

Not in P

- [1960s] For many problems, people could not find better than exponential time algorithms.
- There was no clear explanation why some problems are in P, while others are not.
- Try to guess, whether a polynomial time algorithm is known or not.

- Roommate Allocation:
 - n students, some like each other, some don't.
 - Allocate rooms s.t. roommates like each other.
 - Polynomial time algorithm known or not?
 - Yes [Edmonds 1965]

- Triple Roommate Allocation:
 - n students, some like each other, some don't.
 - Allocate rooms s.t. all 3 roommates like each other.
 - Polynomial time algorithm known or not?
 - No

- Given a graph and a number k, is there a path of length k?
 - Not known to be in P
- Given a graph with s and t vertices, are there two edge disjoint paths from s to t?
 - In P
- Given a graph and four vertices s_1 , s_2 , t_1 , t_2 , are there disjoint paths $s_1 \rightsquigarrow t_1$ and $s_2 \rightsquigarrow t_2$?
 - In P
- Same problem in directed graphs?
 - Not known to be in P

- Given a graph with edges colored red or blue, is there an s-t path with alternating red and blue edges?
 - In P
- Same problem in directed graphs?
 - Not known to be in P

- Given a number (in binary), is it factorizable?
 - In P (only in 2002)
- Given a number (in binary), find its factors?
 - Not known to be in P

- Given a set of intervals, largest subset of disjoint intervals
 - In P
- Given a graph, find the largest independent set (vertices sharing no edges).
 - Not known to be in P
- Given a graph, find the largest set of edges not sharing any vertex
 - In P
- Given a graph, find the largest set of triangles not sharing any vertex
 - Not known to be in P

- Set of trains arriving/departing at a station, can we schedule using k platforms?
 - In P
- Given a list of courses, and pairs which should avoid a clash, can we schedule using k time slots?
 - Not known to be in P
 - Also known as graph coloring (easy for 2 colors)
- n Jobs, m processors, not every processor can handle every job. Processors can work in parallel. Can we finish in k units of time?
 - In P (via Network Flow)

- Given a set of integers, is there a subset with sum equal to zero?
 - Not known to be in P
- Given a set of integers (loads), distribute them among m machines, so that maximum total load (makespan) is minimized.
 - Not known to be in P
 - known as load balancing
- Given a set of integers, partition them into two groups with equal sum
 - Not known to be in P
 - Known as Partitioning

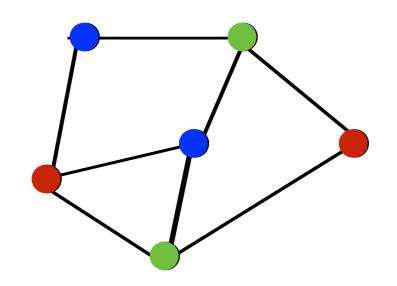
- Minimum weight spanning tree
 - In P
- Steiner Tree: Given a subset of vertices (terminals), find the minimum weight tree that connects the terminals
 - Not known to be in P
- Traveling Salesperson problem: given a list of cities, you have to visit every city and come back with minimum cost
 - Not known to be in P

- Satisfiability: given a Boolean formula, is it satisfiable?
 - Not known to be in P
- Minimum circuit size: Given a Boolean function, is there a circuit for it with at most k Boolean operations?
 - Not known to be in P

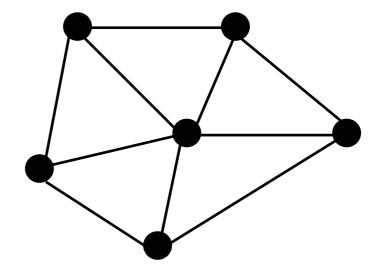
Towards NP

- [1960s] For many problems, people could not find better than exponential time algorithms.
 - Subset sum, Load balancing, Traveling Salesperson, Graphs Isomorphism, Primality,
 Linear programming, Minimum Circuit Size, Satisfiability,
 3-colorability
- People observed some of these can be reduced to others.
- For example, 3-colorability ≤ SAT
- In fact, all of these problems reduce to SAT.
- The key common property of these problems was having "easily verifiable proofs" for the 'yes' instances.

3-colorability



A 3-colorable graph



A graph not 3-colorable

Satisfiability

- $(x \lor y \lor z)$ AND $(\neg x \lor y)$ AND $(x \lor y)$ AND $(x \lor \neg y)$
- Satisfiable
- $(\neg x \lor y)$ AND $(\neg y \lor z)$ AND $(\neg z \lor \neg x)$ AND (x)
- Unsatisfiable

3-colorability reduces to SAT

- Given a graph, can we color vertices with 3 colors?
 - create Boolean variables to represent the coloring
 - 3 Boolean variables for each vertex x_i , y_i , z_i
 - encode the verification procedure as Boolean constraints
 - each vertex has a color $(x_i \lor y_i \lor z_i)$ for each i
 - adjacent vertices have different colors
 - for every edge (i,j): $\neg(x_i \land x_j)$, $\neg(y_i \land y_j)$, $\neg(z_i \land z_j)$
 - Boolean formula = AND of all the constraints.
 - Graph is 3-colorable if and only if there is an satisfying assignment for the above Boolean formula
 - An algorithm for SAT will give an algorithm for 3-colorability
- [Cook, Levin 1971] All of the problems mentioned reduce to SAT

Reductions

- Problem A reduces to problem B $(A \le B)$
 - if A can be solved in polynomial time using a given subroutine that solves B.
 - task of solving A reduces to task of solving B
- Example: Taxi scheduling reduces to bipartite matching
- Example: Multiplication reduces to squaring
 - Multiplication is as easy as squaring
 - Squaring is as hard as Multiplication

Reductions

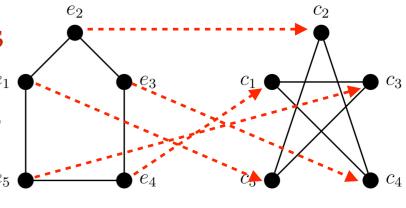
- Problem A reduces to Problem B:
 - (1) convert input φ_A for A to input φ_B for B (or set of inputs φ_{B1} , φ_{B2} , φ_{B3})
 - (2) Solution(φ_B) should be converted to solution(φ_A).
- Conclusion:
 - A is as easy as B.
 - B is as hard as A.
- $A \le B$ and $B \le C$ implies $A \le C$

NP or Easily Verifiable Proofs

- Many problem which seemed hard have easily verifiable proofs for 'yes' inputs.
- Load Balancing: is there a load allocation with makespan at most k?
 - Proof: an allocation with makespan $k' \le k$
 - Verifier: check if the proposed allocation is valid and its makespan
- Factorize Numbers: is a given number factorizable?
 - Proof: two factors
 - Verifier: multiply the proposed two numbers and check if you get the input number.
- Not clear if 'no' inputs have easily verifiable proofs.

Easily Verifiable Proofs

- SAT: given a Boolean formula (CNF AND of ORs), is there an assignment of variables, which makes it true? Example. $(\neg x \lor y) \land (\neg y \lor x)$
 - Proof: an satisfying assignment to the variables (example: True, False)
 - Verifier: check if the proposed assignment makes the formula true.
- Graph Isomorphism: given two graphs, are they isomorphic?
 - Proof: a mapping between two sets of vertices
 - Verifier: check if the given mapping preserves edges and non-edges



Easily Verifiable Proofs

- Subset Sum: given numbers a_1 , a_2 , a_3 , ..., a_n and a number b, is there a subset of a_i 's that sum up to b?
 - Proof: a subset of numbers
 - Verifier: check if the proposed subset has sum equal to b
- Circuit Size: given a Boolean function f truth table, is there a circuit with at most s gates that computes f?
 - Proof: a circuit
 - Verifier: verify if the circuit output matches with the truth table for every input

The Class NP

- Definition: a 'yes or no' decision problem is in NP if there is an easily verifiable proof for each 'yes' input.
- a 'yes or no' decision problem is in NP if there is a polynomial time Algorithm V (verifier), such that for any input *x*,
 - if x is a 'yes' input then there exists c s.t. $size(c) \le poly(size(x))$ and V(x, c) = True.
 - if x is a 'no' input then for any c, V(x, c) = False

The Class NP

P - can find the solution in polynomial time
 NP - can verify a proposed solution in polynomial time

$$P \subseteq NP$$

- if a problem is in P, it is also in NP.
- A problem falling into NP is a positive thing.
- NP does not mean Non-Polynomial time.
- NP stands for Non-deterministic polynomial time.

Problems in NP?

• Given an integer *n*, are there integers *x*, *y*, *z* such that

$$x^3 + y^3 + z^3 = n$$
?

e.g. for n = 39: $134476^3 - 159380^3 + 117367^3 = 39$.

- Not clear, because *x*, *y*, *z* can be much larger than n. Efficient verification may not be possible.
- Given a flow network, is there a flow with $f^{out}(s) = k$?
 - Yes.
 - Proof: for every edge a flow value.
 - Verifier checks flow conservation, capacity constraints, and computes
 fout(s)
 - Proof: 0
 - Verifier runs the max flow algorithm and checks whether $max flow \ge k$

Problems in NP?

- Given two Polynomial expressions, are they equal?
 - e.g., $(a^2 + nb^2)(c^2 + nd^2) = (ac nbd)^2 + n(ad + bc)^2$
- Not clear if it is in NP
- Given two Polynomial expressions, are they different?
 - It is in NP
 - Proof: a substitution of variables with numbers
 - Verifier: evaluates both the expressions on this substitution and verifies if evaluations are different.

NP-completeness

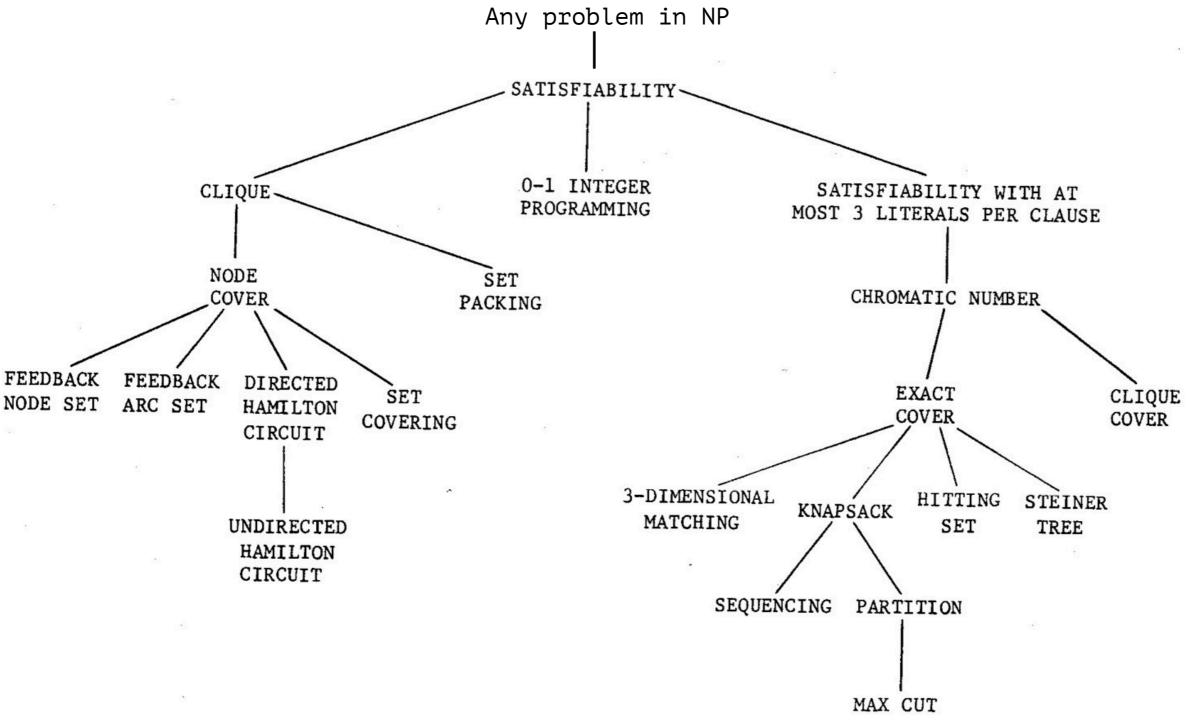
• Cook-Levin [1971]: If we have a subroutine for SAT problem, we can design a polynomial time algorithm for every problem in NP

NP-Complete

NP

- for any problem *A* in NP, $A \leq SAT$
- SAT is 'NP-complete'.
- Karp [1972]: 21 other problems are NP-complete.
 - TSP, Subset Sum, Integer Programming,
 Graph Coloring, Job Sequencing, Independent Set,
 3D-matching etc.
 - They are all equivalent and are hardest problems in NI

The tree of Reductions



P vs NP

- Thousands of problems have been shown to be NP-complete.
- If you solve any of them, all of them get solved.
- People have not been able to give an efficient algorithm in last 50 years, for any of these.
- One can say, there is just one NP-complete problem.

Philosophically...

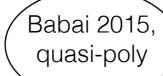
- Problems intuitively/philosophically in class NP
 - is a given mathematical statement (provably) true? (a proposed proof can be verified)
 - is there a cure for a mentioned disease?
 (a proposed cure can be verified)
 - given the public key, can you find the private key? (a private-public key pair can be verified)
 - is there a great film?
 (given a film, you can critic it)

Philosophically...

- P vs NP = Mechanical vs Creativity
- P = NP would mean
 - all diseases can be cured,
 - all mathematical conjectures can be resolved,
 - crypto systems can broken
 - All film critics can make great films

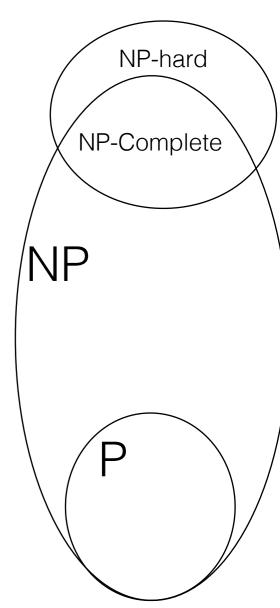
How is it useful?

- Widely believed P ≠ NP, but no proof for it.
 Million Dollars for a proof either way.
- You encounter a new problem X and can't find a Polynomial time algorithm for it try to prove that it is NP-hard.
 - Choose a suitable NP-complete/NP-hard problem *H* and reduce *H* to your problem *X*.
 - I.e., *H* can be solved using a subroutine for *X*.
 - "I am not able to design an algorithm for it, but nobody could in last 50 years "
- Amazingly, most problems turn out to be either in P or NP-complete.
 - Exceptions: Graph Isomorphism, Minimum circuit Size, Factoring



NP-complete and NP-hard

- Problem *X* is said to be NP-complete if
 - 1. *X* is in NP
 - 2. Every problem in NP reduces to *X*
- Problem Y is said to be NP-hard if
 - Every problem in NP reduces to Y



Any problem in NP reduces to SAT [Section 8.4 in Kleinberg Tardos]

- There is a verifier algorithm V such that for any input x,
 - if x is a 'yes' input then there exists y s.t. V(x,y) = True.
 - if x is a 'no' input then for all y, V(x,y) = False
- Reduction: Given x, output a boolean formula f(x) such that
 - if x is a 'yes' input then f(x) has a satisfying assignment
 - if x in a 'no' input then f(x) does not have a satisfying assignment
- Proof y encoded as Boolean variables.
- Each step of algorithm *V* will be converted to a Boolean constraint.

Any problem Q in NP reduces to SAT

- Algorithm *V*: Input $(1,0,1,0,0,...,y_1,y_2,...,y_m)$
- Say it uses *p* bits memory and time *T*.
- Create another pT Boolean variables.
- At time t, an instruction will apply AND/OR/NOT on some memory locations and store it in another location
- $z_{t+1,5} = z_{t,3} \vee z_{t,9}$
- f(x) = AND of all such Boolean constraints.
- f(x) has a satisfying assignment (y,z) if and only if algorithm V outputs True on input $(x, y_1, y_2, ..., y_m)$ if and only if x is a yes input.

IND-SET decision

- IND-SET (OPT): given a graph *G*, find the largest independent set of vertices.
- IND-SET (decision): given a graph *G*, and a number *k*, is there an independent set of size *k*?
- Reduction from IND-SET (OPT) to IND-SET (decision)
 - first find the largest size of an independent set
 - start with $k \leftarrow n$
 - keep decreasing *k* till *G* has no independent set of size *k*-1

IND-SET optimization and decision

- Reduction from IND-SET (OPT) to IND-SET (decision)
 - first find the largest size of an independent set, say it is *k*
 - check whether $G v_1$ has an independent set of size k
 - if yes, recursively find an independent set of size k in $G v_1$
 - if no, then recursively find an independent set of size k-1 in (G neighbors(v_1)) and include v_1 with it.

IND-SET is NP-complete

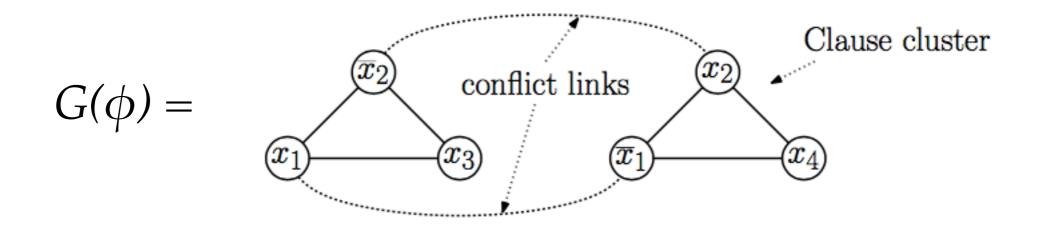
- (1) IND-SET is in NP
 - the proof will be an independent set of size *k*
 - the verifier will check if it is indeed an independent set and has size *k*
- (2) IND-SET is in NP-hard
 - That is, any problem in NP reduces to IND-SET
 - We already know that any problem in NP reduces to SAT
 - We will show that SAT reduces to IND-SET
 - From transitivity of reductions, it will follow that IND-SET is NP-hard
 - Step 1 (homework): SAT reduces to 3-SAT (given formula is 3-CNF)

3-SAT to IND-SET Reduction

- IND-SET: given a graph *G* and a number *k*, is there an independent set of size *k*?
- Reduction: Given a 3-CNF formula ϕ , we want to construct a graph $G(\phi)$ and a number $k(\phi)$ such that
 - if ϕ has a satisfying assignment, then $G(\phi)$ has an independent set of size $k(\phi)$
 - if ϕ does not have a satisfying assignment, then $G(\phi)$ does not have any independent set of size $k(\phi)$
 - Construction of $G(\phi)$:
 - for every clause, introduce a triangle with one vertex corresponding to each literal in the clause
 - add an edge between two vertices across triangles if one corresponds to x_i and the other corresponds to $\neg x_i$
 - $k(\phi)$ = number of clauses = number of triangles

SAT to IND-SET Reduction

- Example:
- $\phi = (x_1 \vee \neg x_2 \vee x_3) \wedge (\neg x_1 \vee x_2 \vee x_4)$



$$-k(\phi)=2$$

Proof of correctness (\Longrightarrow)

- Suppose ϕ is satisfiable, choose one satisfying assignment.
- From each clause, pick an arbitrary literal which is set to true.
- Pick vertices corresponding to the picked literals.
- The number of vertices picked is equal to the number of clauses.
- The picked vertices form an independent set because
 - 1) we pick one vertex from each triangle
 - 2) if vertex corresponding to literal x_i is picked then its neighbors correspond to $\neg x_i$, and hence will not be picked.

Proof of correctness (←)

- Suppose $G(\phi)$ has an independent set of size $k(\phi)$.
- Since $k(\phi)$ = number of triangles, the independent set must have one vertex from each triangle.
- For each vertex in the independent set, set corresponding literal true.
- This is possible because the independent set cannot have vertices corresponding to both x_i and $\neg x_i$
- Set the remaining variables (if any) arbitrarily.
- This is a satisfying assignment ϕ for because for every clause at least one literal is set true.

Homework

- Easy reductions:
 - IND-SET reduces to VERTEX-COVER
 - IND-SET reduces to CLIQUE
 - VERTEX-COVER reduces to SET-COVER
 - SAT reduces to INTEGER LINEAR PROG
- Not so easy:
 - k-CLIQUE reduces to k-COLORING
 - SAT reduces to DIRECTED HAMILTONIAN CYCLE

Thank you

References

- [1] https://math.stackexchange.com/questions/ 3141500/are-these-two-graphs-isomorphic-whywhy-not
- [2] http://electronics-course.com/logic-gates