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, is there a path from s to t with even length?
 - 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)

- 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 $(x \lor \gamma)$ (x $\lor y$)
- 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 $A \le C$

Reductions (decision problems)

- A problem is said to be a decision problem, if for every input the answer is 'yes' or 'no'.
- Karp-reduction: Problem A reduces to Problem B:
 - If we have a polynomial time algorithm which takes an input ϕ_A for problem A and outputs an input ϕ_B for B such that
 - answer for ϕ_A is 'yes' if and only if answer for ϕ_B is 'yes'.
- $A \le B$ and $B \le C$ implies $A \le C$
- Every computational problem has a natural decision version, hence we will be focusing on decision problems.

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



image source: [1]

Easily Verifiable Proofs

- Subset Sum: given numbers a₁, a₂, a₃, ..., 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?



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) \leq 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.

• 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 $max flow \ge k$

- Given two Polynomial expressions, are they equal?
 - e.g., $(a^2 + n b^2) (c^2 + n d^2) = (ac n bd)^2 + n(ad + bc)^2$
- 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.

- Given a graph *G* and and another graph *H*, is *H* a subgraph of *G*?
- It is NP
- Proof: A subset of vertices of *G*, together with a mapping from vertices of *H*.
- Verifier: checks whether any pair has and edge in *H* if and only if the corresponding pair in *G* has an edge.

- Given a graph *G* and a number *k*, the maximum size of any clique in *G* is *k*?
- Not clear if it is in NP
- Given a graph *G* and a number *k*, the maximum size of any clique in *G* is not *k*?
- Not clear if it is in NP

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
 - 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 \dot{NP}



The tree of Reductions



FIGURE 1 - Complete Problems

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.
- $P = NP \iff$

SAT (and every other problem in NP) has a polynomial time algorithm

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)
 - Given a partially made movie, can it be completed to make a great movie?
 (given a movie, 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 ⁽²⁾"
- Amazingly, most problems turn out to be either in P or NP-complete.
 - Exceptions: Graph Isomorphism, Minimum circuit Size, Factoring

Babai 2015, quasi-poly

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 v₁
 - 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
- (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 (\Rightarrow)

- 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 ¬*x_i*, and hence will not be picked.

Proof of correctness (\Leftarrow)

- 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 ¬*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] <u>https://math.stackexchange.com/questions/</u> <u>3141500/are-these-two-graphs-isomorphic-why-</u> <u>why-not</u>
- [2] <u>http://electronics-course.com/logic-gates</u>