

# P = NP?

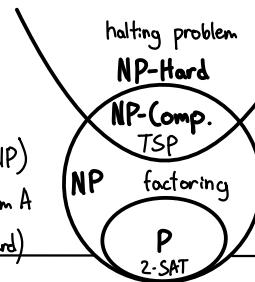
P: solvable in polynomial time

NP: solutions can be verified in polynomial time ('yes solutions')

NP-Hard: at least as hard as the hardest problem in NP (may be outside NP)

- if Problem A is in NP-Hard, every problem in NP can be reduced to Problem A

NP-Complete: problem in NP that all problems can reduce to (NP & NP-Hard)



## Reductions

Using a problem that we know how to solve as a subroutine for a different problem

$A \rightarrow B$  implies...

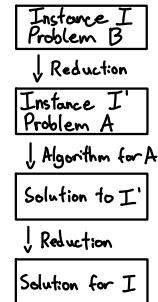
- B is at least as hard as A
- if there is an efficient solver for B, there is an efficient solver for A
- if there is NO efficient solver for A, there is NO efficient solver for B
- a solution to an instance of A exists iff a solution to the corresponding instance of B exists

To prove NP-Complete:

- Show that the problem is in NP (straight forward)
- Reduce from the most similar NP-C problem: (1)  $\text{YES} \Rightarrow \text{YES}$ , (2)  $\text{NO} \Rightarrow \text{NO}$

To prove NP-Hard:

- Reduce from an NP-Hard problem



## Coping with NP Completeness

Approximate Algorithms: for an instance I of a minimization problem, approximation ratio  $\alpha_A = \max_I \frac{A(I)}{\text{OPT}(I)}$

Vertex Cover: Find a maximal matching (disjoint edges) M, return all endpoints in M.  $\text{OPT} \geq \# \text{ of edges in } M \Rightarrow \alpha_A = 2$

Metric TSP: Find an MST T, run DFS on T, skip all repeated vertices in traversal.  $\text{OPT} \geq \text{MST} \Rightarrow 2 \text{OPT} \geq A \Rightarrow \alpha_A = 2$

## Randomized Algorithms

Behavior of algorithm changes based on a random string r

Las Vegas:

- Correctness is guaranteed
  - Runtime is random
  - Goal: Analyze expected worst-case runtime:  $T(n) = \max_r [E_r(x, r)]$ ; show that runtime is low with high probability
- Amplification:  $\Pr(S) = p$ ,  $\Pr(F) = 1-p = q$ ,  $\Pr[\text{all } F] = q^k$   
we want  $q^k \leq \delta \Rightarrow k \ln q \leq \ln \delta \Rightarrow k \geq \frac{\ln(1/\delta)}{\ln(1/q)}$   
use  $1-x \leq e^{-x} \Rightarrow k \geq \frac{1}{p} \ln \frac{1}{\delta}$ ,  $\delta = 0.001 \Rightarrow k \geq 6.91/p$

Union Bound:  $\Pr\left[\bigcup_{i=1}^n A_i\right] \leq \sum_{i=1}^n A_i$

$$\sum_{k=0}^{\infty} \alpha^k = \frac{\alpha(1-\alpha)}{1-\alpha}$$

Monte Carlo:

- Correctness is random
  - Runtime is always bounded
  - Goal: Output a number that is close to true value with high probability
- $* \text{expected # of events} \\ = \text{sum of probabilities} \\ \text{of each event}$

$$** (1-p)^{(1/p)} \leq e^{(-1)} \leq 0.001$$

Linearity of Indicators and Indicators: (1)  $E[X + Y] = E[X] + E[Y]$ , (2)  $E[cX] = cE[X]$ , (3)  $E[\sum X_i] = \sum \Pr[E_i]$

QuickSelect (Las Vegas): Unsorted List  $\rightarrow$  Median, pick random pivot,  $T(n) \leq cn + \frac{1}{2} T(3n/4) + \frac{1}{2} T(n-1) \rightarrow E[T(n)] = O(n)$

One-Sided Error: YES  $\Rightarrow$  YES, NO  $\Rightarrow$  NO with probability p, to get  $\Pr(\text{Success}) \geq 0.999$ , for i in range( $10/p$ ) {if ALG == NO, ret NO} ret YES \*\*

Two-Sided Error: ALG outputs correct with probability  $1/2 + \epsilon$ , 0.999  $\rightarrow$  for i in range( $1/\epsilon^2$ ) {record ALG output} ret majority

Primality Testing (Monte Carlo): Pick a random  $x$  from  $1, \dots, N-1$ , check if  $x^{N-1} \equiv 1 \pmod{N}$ , if so ret PRIME,  $C+K \rightarrow p \geq \frac{1}{2}$

Karger's Algorithm (Monte Carlo): Pick a random edge, contract, until 2 vertices left;  $e_{\text{left}} \geq [(n-i+1) + |c|]/2$ ,  $\Pr(C) = 2/(n-i+1)$

$$\Pr(\text{Success}) \geq \prod_{i=1}^{n-2} \left(1 - \frac{2}{n-i+1}\right) = 1/\binom{n}{2}$$

## Online Algorithms

Regret: Alg's # of mistakes - OPT (# of mistakes by best expert);  $\sum_{t=1}^T \mathbb{1}(\text{guess}^{(t)} \neq \text{real}^{(t)}) \leq \min \sum_{c=1}^T \mathbb{1}(o_i^{(c)} \neq \text{real}^{(c)})$

Halving Algorithm: for at least one perfect expert,  $T = O(\log n)$ , regret =  $O(\log n)$

Weighted Majority: penalize incorrect experts  $w_i^{(t+1)} \leftarrow w_i^{(t)}(1-\varepsilon)$ , else  $w_i^{(t)}$ ,  $\varepsilon = 0.5 \rightarrow E[r] \leq 1.4 \text{OPT} + 2.4 \log(n)$

Random Weighted Majority: no regret for the right exps, samples expert at random using weighted distribution rather than choosing the expert with the largest weight. experts' behavior is NOT randomized

MW Garuntee:  $E[C_{\text{ALG}}(T)] \leq \min C_i(T) + O(\sqrt{T \log N})$  or  $E[\text{Regret}(T)] = E[C_{\text{ALG}}(T) - \text{OPT}] \leq O(\sqrt{T \log N})$

asymptotics

- $f(n) = O(g(n))$
- $f(n) = \Theta(g(n))$
- $f(n) = \Omega(g(n))$

$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$

$\begin{cases} < \infty \\ = c \\ > 0 \end{cases}$	$\begin{cases} \log_b(M \cdot N) = K \cdot \log_b M \\ \log_b(b^k) = k \\ \alpha^{\log_b n} = n^{\log_b \alpha} \\ \log_b n = \frac{\ln n}{\ln b} \end{cases}$	$\begin{cases} \ln n \rightarrow 1/n \\ a^n \rightarrow a^n \ln n \\ (\ln n)^k \rightarrow k(\ln n)^{k-1}/n \\ n \log n \rightarrow \log n \end{cases}$	$\begin{cases} 1 + t + \dots + n = O(n^2) \\ 1 + \frac{1}{2} + \dots + \frac{1}{n} = O(\log n) \\ a + a^2 + \dots + a^n = \frac{a(1 - r^n)}{1 - r} \\ 1 + p + p^2 + \dots + p^k = \frac{p^{k+1} - 1}{p - 1} \\ \text{inf geometric} \rightarrow \frac{a}{1 - r} \end{cases}$
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exponential  $>$  polynomial  $>$  logarithm  
 $3^n > 2^n > n^5 > n^4 > n^3 > (\log n)^3 > (\log n)^2 > \log \log n$

try substitution! or L'Hopital's!

master's theorem

$T(n) = aT(n/b) + O(n^d)$

$> O(n^d)$   
 $d = \log_b a$   
 $< O(n^{\log_b a})$

tree method

$T(n) = 2T\left(\frac{n}{2}\right) + O\left(\frac{n}{\log n}\right) \quad \tau(1)=1$

$n \sum_{j=0}^{\log n} \frac{1}{2^j \log \frac{n}{2^j}} = n \sum_{j=0}^{\log n} \frac{1}{\log n - \log 2^j}$

$= n \sum_{j=0}^{\log n} \frac{1}{\log n - j}$

$= n \left( \sum_{j=1}^{\log n} \frac{1}{j} \right)$

$= n \log \log n$

maths

- $\log_b(M \cdot N) = \log_b M + \log_b N$
- $\log_b(b^k) = k$
- $\alpha^{\log_b n} = n^{\log_b \alpha}$
- $\log_b n = \frac{\ln n}{\ln b}$
- $\sum_{j=1}^K \frac{1}{j} = \log K$
- $\ln n \rightarrow 1/n$
- $a^n \rightarrow a^n \ln n$
- $(\ln n)^k \rightarrow k(\ln n)^{k-1}/n$
- $n \log n \rightarrow \log n$
- $1 + p + p^2 + \dots + p^k = \frac{p^{k+1} - 1}{p - 1}$
- $\text{inf geometric} \rightarrow \frac{a}{1 - r}$

BFS/DFS:  $O(|V| + |E|)$

Kosaraju's:  $O(|V| + |E|)$

- run DFS
- run BFS on Grev w/ decreasing post-order

Dijkstra's:  $O(|E| \sqrt{|V|} \log |V|)$

- init dist =  $\infty$ , par = None
- dist( $v_0$ ) = 0, add to minH
- update all neighbors of first

Bellman-Ford:  $O(|V|||E|)$

- works w/ neg edges!
- repeat  $|V|-1$  times:
  - for each  $e$  in E:
    - update dist(v)

squeeze + guess & check

- guess U/L bounds

$T(n) = T(3n/5) + T(4n/5)$

guess  $an^b$ ,  $a=1$  since  $\tau(1)=1$

$T(n) = nb: (3n/5)^b + (4n/5)^b$

$\Rightarrow b=2$

$T(n) = T(\sqrt{n}) + 1$

$m = \log n$

$T(2^m) = 2T(2^{m/2}) + O(m)$

$S(m) = T(2^m)$

$S(m) = 2S(m/2) + O(m)$

$\Rightarrow S(m) = m \log m$

$T(2^m) \leq m \log m$

$T(2^{\log n}) = T(n) = \log n \log \log n$

