

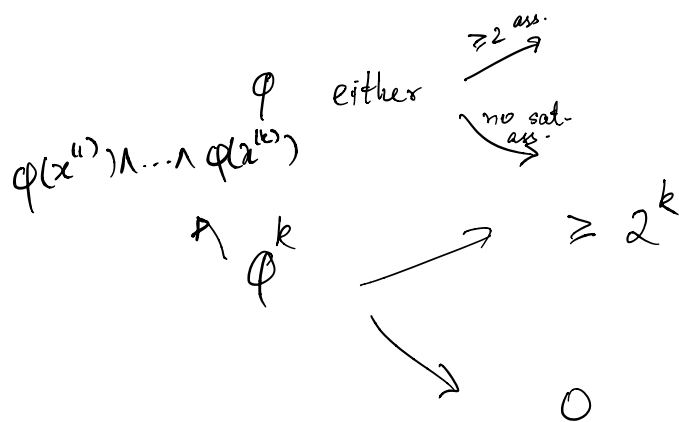
Computational Complexity: Lecture 20.

- Recap:
- #P - counting witnesses.
 - #SAT, Perm are #P-complete
 - PH $\subseteq P^{\#P}$ - VV++ & some modular magic.

Agenda: - Approximate counting

Qns: Exactly computing #satisfying assignments is hard.

But can we approximate the number of SAT assignments?



Computing #SAT approximately seems at least as hard as SAT itself.

Thm: For any ϵ, δ , there is a BPP^{NP} algorithm. A^{SAT} which, on input ϕ , satisfies

$$P_{\delta} \left[A^{SAT}(\phi) \in \#SAT(\phi) \cdot (1 \pm \epsilon) \right] \geq 1 - \delta.$$

with running time $\text{poly}(|\phi|, \frac{1}{\epsilon}, \log \frac{1}{\delta})$.

What can we do with an NP oracle?

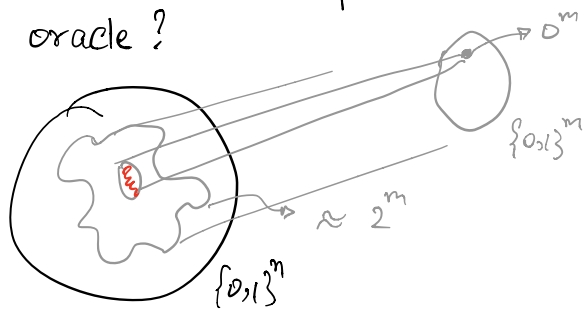
- checking if ϕ is SAT/TAUT
- checking if ϕ has exactly 1 SAT assignment.
- Checking if ϕ has ≥ 42 SAT assignments.

◦◦ If $\#SAT(\phi)$ is "small" then we can compute this exactly using an NP oracle.

A simpler promise problem:

$$\text{approx-count}(\phi, k) = \begin{cases} \text{Yes} & \text{if } \#SAT(\phi) \geq 2^{k+1} \\ \text{No} & \text{if } \#SAT(\phi) \leq 2^k. \end{cases}$$

How can we hope to solve this even with an NP oracle?



$$S = \{a : \phi(a) = 1\}.$$

$|h^{-1}(0^m) \cap S|$
Should roughly
tell us the size
of S .

Lemma: (Leftover Hash Lemma) Let $\mathcal{H} = \{h : \{0,1\}^m \rightarrow \{0,1\}^m\}_h$ be a family of pairwise independent hash functions and let $\epsilon > 0$. For any $S \subseteq \{0,1\}^m$ s.t. $|S| \geq 4 \cdot 2^m / \epsilon^2$, we have.

$$\Pr_{h \in \mathcal{H}} \left[|\{a \in S : h(a) = 0^m\}| \in \frac{|S|}{2^m} \cdot (1 \pm \epsilon) \right] \geq 3/4.$$

$$\forall a \neq y, a, b$$

$$\Pr[h(a) = a \ \& \ h(y) = b] = \Pr[h(a) = a] \cdot \Pr[h(y) = b]$$

$$\Pr[h(a) = a] = 1/2^m.$$

Let us just assume this for now and proceed.

BPP^{NP} algo for approx-count(φ, k):

▷ If $k \leq 5$ we can check if $\varphi \geq 2^{k+1}$ sat. assignments using an NP oracle.

▷ If $k \geq 6$. Set $m = k - 5$ and pick a random $h: \{0,1\}^m \rightarrow \{0,1\}^m$ from a p.i.h.f.

Return yes if the formula

" $\varphi(x) = 1 \wedge h(x) = 0$ " has ≥ 48 sat. assignments

& "No" otherwise.

Correctness:

Let $S = \{a: \varphi(a) = 1\}$.

"Yes" case: $|S| \geq 2^{k+1} = 2^{m+6} = 4 \cdot 2^m / \epsilon^2$ for $\epsilon = 1/4$.

LHL says $\Pr \left[\left| \{a \in S: h(a) = 0\} \right| \geq (1-\epsilon) \cdot \frac{|S|}{2^m} \right] \geq \frac{3}{4}$
 $\frac{3}{4} \cdot 64 = 48$

"No" case: $|S| \leq 2^k$. $S \subseteq S'$ $|S'| = 2^k = 2^{m+5}$ $\epsilon = 1/2$

$\Pr \left[\left| \{a \in S': h(a) = 0\} \right| > (1+\epsilon) \cdot \frac{|S'|}{2^m} \right] \leq 1/4$
 $\frac{1}{48}$

□.

Remark: We can push the error down by repeating + majority.

Again the prob. of error can be pushed to any δ by repeating etc

Essentially finishes this



Thm: For any ϵ, δ , there is a BPP^{NP} algorithm A^{SAT} which, on input ϕ , satisfies $\Pr[A^{SAT}(\phi) \in \#SAT(\phi) \cdot (1 \pm \epsilon)] \geq 1 - \delta$. with running time $\text{poly}(|\phi|, \frac{1}{\epsilon}, \log \frac{1}{\delta})$.

modulo the LHL.

Proof of the Leftover Hash Lemma:

Lemma (Leftover Hash Lemma) Let $\mathcal{H} = \{h: \{0,1\}^n \rightarrow \{0,1\}^m\}_n$ be a family of pairwise independent hash functions and let $\epsilon > 0$. For any $S \subseteq \{0,1\}^n$ s.t. $|S| \geq 4 \cdot 2^m / \epsilon^2$, we have $\Pr_{h \in \mathcal{H}} [|\{x \in S: h(x) = 0^m\}| \in \frac{|S|}{2^m} \cdot (1 \pm \epsilon)] \geq 3/4$.

Say $S = \{a_1, \dots, a_r\}$. $X_i = \mathbb{1}(h(a_i) = 0^m)$

$$\mathbb{E}[X] = |S|/2^m = r/2^m = \mu \quad \sum X_i = X$$

Interested in $\Pr[|X - \mu| \geq \epsilon \mu]$

If X_i 's were indep. then Chernoff would have worked.

But X_i 's need not be indep... but they are pairwise indep.

Obs: For any $i \neq j$ $\Pr[X_i = a, X_j = b] = \Pr[X_i = a] \Pr[X_j = b]$

Since x_i 's are pairwise indep,

$$\text{Var}(x) = E[(x-\mu)^2]$$

pairwise indep.

$$\rightarrow = \sum \text{Var}(x_i)$$

$$= \sum_{i=1}^r (E[x_i] - E[x_i]^2)$$

$$= \sum_{i=1}^r \left(\frac{1}{2^m} - \frac{1}{2^{2m}} \right) \leq \frac{r}{2^m}$$

$$\Pr[|x-\mu| \geq \epsilon\mu] = \Pr[(x-\mu)^2 \geq \epsilon^2\mu^2] \leq \frac{E[(x-\mu)^2]}{\epsilon^2\mu^2}$$

$$\leq \frac{r/2^m}{\epsilon^2 \cdot (r/2^m)^2} = \frac{2^m}{\epsilon^2 \cdot |S|} \leq 1/4$$

if $|S|$ is large.

Chebyshev's Ineq: $\Pr[|x-\mu| \geq \epsilon\mu] \leq \frac{\text{Var}(x)}{\epsilon^2\mu^2}$

So what have we learnt?

□

▷ Perm & #SAT are #P-complete.

▷ #P = FP \Rightarrow P = NP.

But we can do a lot more with a #P oracle than with an NP-oracle.

$$P^{NP} \subseteq \Sigma_2$$

$$P^{\#P} \supseteq PH$$

▷ But if we only wish to approximately compute #SAT (or Perm) we can do that in BPP^{NP} .

(Kuldeep Meel & Moshe Yardi)

▷ Fact: Perm of a non-negative matrix can actually be approximately computed in P! [Jerrum-Sinclair-Vigoda].

Some properties of the Permanent.

▷ Downward self-reducibility:

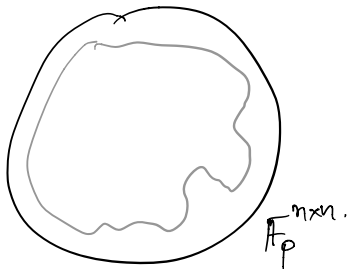
If you know how to solve $n \times n$ permanents, you can solve $(n+1) \times (n+1)$ permanents.

$$\begin{bmatrix} x_{11} & \dots & x_{1,n+1} \\ \vdots & & \vdots \\ x_{n+1,1} & \dots & x_{n+1,n+1} \end{bmatrix} = x_{11} \cdot \boxed{} + x_{12} \cdot \boxed{} + \dots + x_{1,n+1} \cdot \boxed{}$$

▷ Random self-reducibility.

$$\boxed{}_{n \times n}$$

Fix a prime p ($\approx n^2$).
Consider the task of computing $\text{Perm}(X) \pmod p$.



Suppose we have an algorithm A that computes $\text{Perm} \pmod p$ on $\geq 1 - \frac{1}{3(n+1)}$ fraction of inputs.

Claim: Given A , we can compute $\text{Perm} \pmod p$ everywhere with prob $\geq 2/3$.

Pf: Say we are given a matrix $X_{n \times n}$.

Pick a matrix Y uniformly at random.

Since $p \approx n^2$, let $\alpha_0, \alpha_1, \alpha_2, \dots, \alpha_n$ be distinct & non-zero residues mod p .

$$\text{Perm}(X+tY) = f_0 + f_1 t + f_2 t^2 + \dots + f_n t^n.$$

$$\begin{bmatrix} x_{11} + t y_{11} \\ \vdots \\ x_{nn} + t y_{nn} \end{bmatrix} \rightarrow \text{Perm}(X)$$

Let $\beta_i = \text{Perm}(X + \alpha_i Y)$ for $i = 0, 1, \dots, n$

$$\begin{bmatrix} 1 & \alpha_0 & \alpha_0^2 & \dots & \alpha_0^n \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 1 & \alpha_n & \alpha_n^2 & \dots & \alpha_n^n \end{bmatrix} \begin{bmatrix} f_0 \\ f_1 \\ \vdots \\ f_n \end{bmatrix} = \begin{bmatrix} \beta_0 \\ \vdots \\ \beta_n \end{bmatrix}$$

Matrix is invertible.
 $\text{Det} = \prod (\alpha_i - \alpha_j)$

$$\Rightarrow \begin{bmatrix} f_0 \\ \vdots \\ f_n \end{bmatrix} = V^{-1} \begin{bmatrix} \beta_0 \\ \vdots \\ \beta_n \end{bmatrix}$$

$$\Rightarrow f_0 = \sum_{i=0}^n (V^{-1})_{0,i} \cdot \beta_i$$

What are the chances that $\mathcal{A}(X + \alpha_i Y) = \text{Perm}(X + \alpha_i Y)$ for all $i = 0, \dots, n$?

Obs: For any fixed matrix X , if Y is random and $\alpha_i \neq 0$ then $X + \alpha_i Y$ is also random!

$$P_{\mathcal{A}} \left[\exists i : \mathcal{A}(X + \alpha_i Y) \neq \text{Perm}(X + \alpha_i Y) \right] \leq \frac{1}{3^{n+1}} \leq \frac{1}{3} \quad \square$$

Turns out, even if A computes Perm on $1/n$ fraction of inputs, that's enough to get the same conclusion above!

"low-degree polynomials are error-correcting codes".

Ref: "Permanent is hard even on a good day"
by Yuval Filmus.