## COM S 6810 Theory of Computing

February 24, 2009

## Lecture 11: Complexity of Counting I

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## 1 Basic Concepts

Given a polynomial-time computable relation R, let  $L_R$  be the language defined by

$$L_R := \{x \mid \exists y, (x, y) \in R\}.$$

Then it holds that  $L_R \in NP$ . Next we consider a 'harder' problem: compute number of y's satisfying  $(x, y) \in R$ .

**Definition 1.** Define the function  $f_R(x): \{0,1\}^* \to \mathbb{N}$  as  $f_R(x) = |\{y \mid (x,y) \in R\}|$ . Let

$$\#R := \{(x,k) \mid f_R(x) \ge k\}.$$

**Proposition 1.** For all polynomial-time computable R, deciding #R and computing  $f_R$  are Turing-reducible to one another.

**Proof.** The reduction from  $f_R(x)$  to #R is obvious. Conversely, if we have the access to an oracle deciding whether  $(x, k) \in \#R$ , then  $f_R(x)$  can be computed by performing a binary search on the value of k, costing poly(x) time.

**Definition 2.** #P is defined by the class of languages L such that L=#R for some polynomial-time computable relation R. L is #P-complete if  $L \in \#P$  and  $R \leq_p L$  for all  $R \in \#P$ .

For every NP language decided by NTM M, there is a 'generalized' problem in #P which computes the number of certificates that make M accepting a given input x. Therefore,

Fact.  $NP \leq \#P \subseteq PSPACE$ .

**Definition 3.** We say f is a parsimonious reduction from #Q to #R, if it is polynomial-time computable and for all x,  $f_Q(x) = f_R(f(x))$ .

**Notation 1.** If #R is parsimoniously reducible from Q, we write  $\#Q \leq_{par} \#R$ .

If f is a parsimonious reduction from #Q to #R, then  $L_Q \leq L_R$ , since  $x \in L_Q$  iff  $f(x) \in L_R$ . Conversely, if  $\#Q \leq_{par} \#R$ , then  $(x,k) \in \#Q \Leftrightarrow (f(x),k) \in \#R$ .

**Theorem 1.** #SAT is #P-complete.

Conjecture. #L is #P-complete implies that L is NP-complete.

Unfortunately, this conjecture is FALSE:

**Theorem 2.** There exists a polynomial-time computable relation R such that #R is #P-complete but  $L_R \in P$ .

**Proof.** Define R as follows:

$$(x, y') \in R \Leftrightarrow y' = 0 \lor (y' = 1y \land (x, y) \in R_{SAT}),$$

where  $(x, y) \in R_{SAT}$  iff the boolean formula  $\phi$  described by x is satisfied by assigning the values described by y to the variables in  $\phi$ .

It is obvious that  $L_R \in \mathsf{P}$  since  $(x,0) \in R$  for all  $x \in \{0,1\}^*$ , namely  $L_R = \{0,1\}^*$ . On the other hand,  $\#\mathsf{SAT} \leq_{par} \#R$ , since  $(x,k) \in \#\mathsf{SAT} \Leftrightarrow (x,k+1) \in \#R$ . Therefore #R is  $\#\mathsf{P}$ -complete.

**Definition 4.** Given an  $n \times n$  matrix A, its permanent is defined by

$$perm(A) = \sum_{\sigma \in S_n} \prod_{i=1}^n A_{i,\sigma(i)}.$$

**Theorem 3** (Valiant). Computing permanent of 0-1 matrices is #P-complete.

As shown in Figure 1, given an  $n \times n$  0-1 matrix A, a bipartite graph G(X,Y,E) can be built as follows:  $X = \{x_1, x_2, \dots, x_n\}, Y = \{y_1, y_2, \dots, y_n\}, (x_i, y_j) \in E \Leftrightarrow A_{i,j} = 1$ .

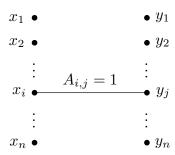


Figure 1: Constructing bipartite graph for 0-1 matrix A

Then it is easy to verify that the permanent of A equals the number of perfect matchings in G. Therefore, counting the number of perfecting matchings in a bipartite graph is also #P—complete .

## 2 Approximate Counting

**Theorem 4.** Given any polynomial p, there exists a PPT A such that

$$Pr\left[\#\mathrm{SAT}(\phi)\cdot\left(1-\frac{1}{p(n)}\right)\leq A^{\mathsf{NP}}(\phi)\leq \#\mathrm{SAT}(\phi)\cdot\left(1+\frac{1}{p(n)}\right)\right]\geq 1-2^{-n}.$$

**Basic idea.** For all  $\phi$ , if we can find a rough approximation  $A'(\phi)$  such that

$$\#SAT(\phi) \cdot 2^{-i} \le A'(\phi) \le \#SAT(\phi) \cdot 2^{i}$$

for some constant i, then we are able to obtain a tighter approximation by:

- (1) construct  $\phi'$  from  $\phi$  such that  $\#SAT(\phi') = \#SAT(\phi)^k$  for some k;
- (2) output  $A'(\phi')^{1/k}$ .

Since

$$\#SAT(\phi)^k \cdot 2^{-i} = \#SAT(\phi') \cdot 2^{-i} < A'(\phi') < \#SAT(\phi') \cdot 2^i = \#SAT(\phi)^k \cdot 2^i$$

it holds that

$$\#\operatorname{SAT}(\phi) \cdot 2^{-i/k} \le A'(\phi')^{1/k} \le \#\operatorname{SAT}(\phi) \cdot 2^{i/k}.$$

For step (1),  $\phi'$  can be constructed by

$$\phi' = \bigwedge_{i=1}^k \phi(\vec{x}_i),$$

where  $\phi(\vec{x}_i)$  is a copy of  $\phi$  with the variables renamed to  $\vec{x}_i$ .

Consider GAP-SAT:

$$\Pi_Y = \{ (\phi, k) \mid \#SAT(\phi) \ge 8k \};$$
  
 $\Pi_N = \{ (\phi, k) \mid \#SAT(\phi) \le k/8 \}.$ 

**Claim.** There exists a polynomial-time TM A such that  $A^O$  approximates #SAT within factor  $8^{1.5}$  where O is an oracle that solves GAP-SAT.

**Proof.** Let  $A(\phi)$  work as follows:

- 1:  $i \leftarrow 0$
- 2: while  $O(\phi, 8^i) = 1 \text{ do}$
- $i \leftarrow i + 1$
- 4: end while
- 5: return  $8^{i-\frac{1}{2}}$

After exiting the while loop, it holds that  $O(\phi, 8^i) \neq 1$  and  $O(\phi, 8^{i-1}) = 1$ , which implies that  $8^{i-2} < \#SAT(\phi) < 8^{i+1}$ . Thus

$$8^{-1.5} < \frac{\#SAT(\phi)}{8^{i-\frac{1}{2}}} < 8^{1.5}.$$

Next lecture we will show how to solve GAP-SAT (with the power of randomness).