

Def. A polynomial randomized approx. scheme (PRAS) for a counting problem  $\Pi$  is a randomized alg.  $A$  that takes an input  $I$ , real number  $\varepsilon > 0$ , and in  $\text{poly}(n)$  time produces output  $A(I)$  s.t.

$$P \left[ \underbrace{(1-\varepsilon) \#(I)}_{\substack{\text{number} \\ \text{of sols} \\ \text{to } I}} \leq A(I) \leq (1+\varepsilon) \#(I) \right] \geq \underbrace{\frac{3}{4}}_8$$

Def. FPRAS polytime in  $\left(n, \frac{1}{\varepsilon}, \log \frac{1}{\delta}\right)$ .

counting DNF

$f: U \rightarrow \{0,1\}$  boolean function over  $U$   
 $G = \{u \in U : f(u) = 1\}$ .

Assume computing  $f(u)$  takes unit time.

$F(x_1, \dots, x_n) :=$  Boolean formula in DNF form.

let  $U = \{0,1\}^n$  set of all  $2^n$  assignments

Classical Monte-Carlo method:

-  $N$  independent samples from  $U$  say  
 $u_1, u_2, \dots, u_N$ .

$$Y_i = \begin{cases} 1 & \text{if } f(u_i) = 1 \\ 0 & \text{o.w.} \end{cases}$$

estimator:

$$Z = |U| \cdot \sum_{i=1}^N \frac{Y_i}{N}$$

Thm (Estimator theorem): (straight forward from Chernoff bound)

Let  $p = \frac{|G|}{|U|}$ . The Monte-Carlo method  
yields an  $(\epsilon, \delta)$ -approximation provided:

$$N \geq \frac{4}{\epsilon^2 p} \ln \frac{2}{\delta}.$$

problem:  $\frac{1}{p}$  could be exponentially large.

idea: importance sampling.

skewed sampling that concentrates the  
probability on the sub-problem of interest.

## The coverage algorithm:

- Let  $V$  be a universe.
- $H_1, H_2, \dots, H_m \subseteq V$  s.t.
  - 1)  $\forall i, |H_i|$  can be computed in poly-time.
  - 2) It's possible to sample  $u$  from any  $H_i$
  - 3)  $\forall u \in V$ , one can determine whether  $u \in H_i$

Goal: Estimate  $|H| = |H_1 \cup H_2 \cup \dots \cup H_m|$

- Consider a DNF formula  $F(x_1, \dots, x_n)$
- let the  $i$ th clause  $C_i$  be a conjunction of  $r_i$  literals.

-  $V = 2^n$  assignments

-  $H_i =$  assignments that satisfy  $C_i$

- We have:

$$|H_i| = 2^{n-r_i}$$

- To sample from  $H_i$ , set all literals in  $C_i$  to true and the rest are set randomly.

- Testing  $u \in H_i$  can be done in linear time.

- Def:

$$U = H_1 \uplus H_2 \uplus \dots \uplus H_m$$

multiset union.

- convention: the elms of  $U$  are ordered pairs  $(v, i)$  corresponding to  $v \in H_i$

$$U = \{(v, i) : v \in H_i\}$$

- Observe that:

$$|U| = \sum_{i=1}^m |H_i| \geq |H|$$

- Def: the coverage set of  $v \in V$

$$\text{cov}(v) = \{(v, i) : (v, i) \in U\}$$

the size of  $\text{cov}(v)$  is the multiplicity of  $v$  in  $U$ .

- The following are immediate:

1) The number of cover sets is exactly  $|H|$ . These coverage sets are easy to compute.

2) The coverage sets partition  $U$

$$U = \bigcup_v \text{cov}(v).$$

3)  $|U|$  is easily computed as

$$U = \sum_{v \in H} |\text{cov}(v)|$$

4) For all  $v \in H$ ,  $\text{cov}(v) \leq m$ .

Def. The function  $f: U \rightarrow \{0,1\}$  is defined as:

$$f((v, i)) = \begin{cases} 1 & \text{if } i = \min \{j : v \in H_j\} \\ 0 & \text{o.w.} \end{cases}$$

The set  $G$  can be rewritten as:

$$G = \left\{ (v, j) \in U : f((v, i)) = 1 \right\}.$$

(assign  $v$  to the lowest index  $C_j$  where  $v$  satisfies  $C_j$ ).

Def. canonical elm. for  $\text{cov}(v)$ . as the elm that corresponds to the occurrence of  $v$  in the lowest-index  $H_j \ni v$ .

(crucial observation:  $|G| = |H|$ ).

Goal: Estimate the size

$$G \subseteq U \quad \text{s.t.}$$

$$G = f^{-1}(1)$$

Lemma:  $\rho := \frac{|G|}{|U|} \geq \frac{1}{m}$ .

$$\text{pf: } |U| = \sum_{v \in H} \text{cov}(v) \\ \leq m \cdot |H| = m \cdot |G|.$$

$$\Rightarrow \frac{|U|}{|G|} \leq m.$$

Thm: The monte-carlo method yields an  $(\epsilon, \delta)$ -approximation w/

$$N \geq \frac{4m}{\epsilon^2} \lg \frac{2}{\delta} \text{ samples. is poly-time.}$$

pf: To sample from  $U$ :

$$P[\text{pick } i] = \frac{|H_i|}{|U|} = \frac{2^{n-r_i}}{\sum_j |H_j|}$$

Then sample  $v \in H_i$  u.a.r

To check if  $v \in G$ :  
check if  $\forall j < i$ ,  $v$  does not satisfy  $C_j$

Intuition: sampling from  $V$  fails b/c  $V$  can be exponentially larger than  $H$ .  
In the new approach each satisfying assignment is sampled proportional to the # clauses it satisfies.