CSE6512: Randomization in Computing

Lecture 5, Sep 13 2011.

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Fingerprinting techniques are used to verify equality of two objects. (obj<sub>1</sub>  $\stackrel{?}{=}$  obj<sub>2</sub>)

## **Example1:**

Input: three n×n matrices A,B and C

Output: "yes" if AB=C

"no" otherwise.

**Fact:** This problem can be solved in  $O(n^{2.376})$  time using the best known deterministic matrix multiplication algorithm.

By using fingerprinting technique, we can solve this problem more efficiently.

# A randomized algorithm 1 (Monte Carlo)

Pick a random vector  $r=[r_1,r_2,...,r_n]$  where  $r_i$  is from  $\{0,1\}$   $\{r_1,...,r_n\}$  are picked using a 2-sided fair coin)

Check if A(Br) = Cr.

End of alg.

**Fact:** if AB  $\neq$  C then prob.[A(Br)= Cr]  $\leq \frac{1}{2}$ 

### Proof:

Let D=AB-C

Assume that D≠0. This means that there exists at least one non-zero row in D.

Without loss of generality, let this row be  $[d_1, d_2, ..., d_k, 0, ..., 0]$  where  $k \ge 1$ . Obviously we get

Prob.[Dr=0 | D  $\neq$  0]  $\leq$  prob.[ $\sum_{i=1}^{k} d_i \times r_i = 0 | D \neq 0$ ]

RHS 
$$\leq$$
 prob.  $\left[r_1 = \frac{-\sum_{i=2}^k d_i \times r_i}{d_1} \mid D \neq 0\right]$ 

We can employ the principle of deferred decisions here. In particular, assume that we have already chosen  $r_2$ , ..,  $r_n$ , and now we are fixing the value of  $r_1$ .

Since  $r_1 \in \{0,1\}$ , we can see that the above probability is  $\leq \frac{1}{2}$ .

A better algorithm for this problem would be:

## **Algorithm 2**

For i=1 to  $\alpha$  logn do

Pick a random  $r \in \{0, 1\}^n$ 

If A (Br)  $\neq$  Cr then output "AB  $\neq$  C" and quit

End for

Output: "AB=C"

End of alg.

Probability that the above algorithm gives an incorrect answer is  $\leq \left(\frac{1}{2}\right)^{\alpha \log n} = n^{-\alpha}$ 

 $\Rightarrow$  run time of the algorithm is  $O(n^2 \log n)$ .

# Example 2.

Input: two degree-n polynomials  $p_1(x)$ ,  $p_2(x)$  and one 2n-degree polynomial  $p_3(x)$ 

Output: "yes" if  $p_1(x) \times p_2(x) = p_3(x)$  and "no" otherwise.

**Fact**: we can solve this problem using DFT in  $O(n \log n)$  time.

## A Randomized algorithm

Let S be a subset of the field F s.t |S| > 2n

Let 
$$Q(x)=p_1(x) \times p_2(x)-p_3(x)$$

Pick a random  $r \in S$ 

If Q(r) = 0 output "yes"

Else output "no"

End of alg.

## **Analysis:**

If  $p_1(x) \times p_2(x) = p_3(x)$ , the algorithm will never give an correct answer. On the other hand, if  $p_1(x) \times p_2(x) \neq p_3(x)$ , the algorithm might give an incorrect answer. This algorithm has one-sided error.

Now we show that the probability that the algorithm gives an incorrect answer is very low: In particular, we'll show that Prob.[Q(r)=0 | Q(x)  $\neq 0$ ]  $< n^{-\alpha}$ .

**Note:** Q(x) is a degree-2n polynomial and hence it has at most 2n distinct zeros.

$$\Rightarrow$$
 Prob[Q(r)=0 | Q(x)  $\neq$  0]  $\leq \frac{2n}{|S|}$ . We want this to be low.

$$\Longrightarrow \frac{2n}{|s|} \le n^{-\alpha}$$
. This will happen if

$$|s| > 2n^{\alpha+1}$$
.

## Example 3.

Input: A multivariate polynomial on n variables  $Q(x_1, x_2, ..., x_n)$ 

Output: "yes" if  $Q(x_1, x_2, ..., x_n) \equiv 0$ "no" otherwise

## **Definitions:**

- 1. Degree of a term is the sum of exponents of the variables in the term.
- 2. Total degree of  $Q(x_1, x_2,..., x_n)$  is the maximum degree of its terms.

## **Theorem: (Schwartz & Zippel)**

Let S be a subset of the field F and let  $r_1, r_2, ..., r_n$  be random elements from S. Then, prob. $[Q(r_1, r_2, ..., r_n) = 0 \mid Q(x_1, x_2, ..., x_n) \not\equiv 0] \le \frac{d}{|s|}$  where d is the total degree of  $Q(x_1, x_2, ..., x_n)$ .

**Fact:** Prob.[A]  $\leq$  prob.[A| $\overline{B}$ ]+prob.[B] since prob.[A]=prob.[A|B] prob.[B] + prob.[A| $\overline{B}$ ] prob.[ $\overline{B}$ ]  $\leq$  prob.[A| $\overline{B}$ ]+prob.[B]

### **Proof:**

We use induction on n.

Base case: The theorem holds for n=1 (proof is given in example 2).

Induction step: Assume that the theorem holds for up to n-1 variables. We show that it holds for n variables as well.

Let 
$$Q(x) = \sum_{i=0}^{k} x_1^i Q_i(x_2, x_3, ..., x_n)$$
 where  $k \le d$ .

 $\Longrightarrow Q_k(x_2,x_3,...,x_n)\not\equiv 0$  and the total degree of  $Q_k(x_2,x_3,...,x_n)\leq d$ -k.

By induction hypothesis, we get

Prob.[
$$Q_k(r_2,r_3,...,r_n)=0$$
]  $\leq \frac{d-k}{|s|}$ -----(1)

Let 
$$Q(x_1, r_2, ..., r_n) = q(x_1) = \sum_{i=0}^k x_1^i Q_i(r_2, r_3, ..., r_n)$$

Prob.[q(r<sub>1</sub>)=0| 
$$Q_k(r_2,r_3,...,r_n) \neq 0$$
]  $\leq \frac{k}{|s|}$ -----(2)

Let A be the event

$$Q(r_1,r_2,...,r_n)=0$$

Let B be the event

$$Q_k(r_2,...,r_n)=0$$

Using the fact prob.[A] < prob.[A]  $\overline{B}$ ] + prob.[B]

We get (from Equations 1 and 2)

Prob.[Q(
$$r_1, r_2, ..., r_n$$
)=0 | Q( $x_1, x_2, ..., x_n$ )  $\not\equiv 0$ }  $\leq$ 

$$Prob. \left[ \ Q(r_1, r_2, ..., r_n) = 0 \ | \ Q_k(r_2, ..., r_n) \neq 0 \ \right] + prob. \left[ Q_k(r_2, ..., r_n) = 0 \right] \leq \frac{d-k}{|s|} + \frac{k}{|s|} \leq \frac{d}{|s|}$$