### Week 2 Notes

# I. Frewalds Algorithm

Inputs: Matrices A, B, C & Znxn

Problem - Verify whether AB = C.

Straightforward approach is to compute product  $P = A \cdot B$  and and check if P = C entry-wise. Can we do better?

Idea: Choose random re{0,13" and check 4 A (Bn) = Cn.

Note: Choosing n= (n,,-,nn) uniformly from 10,13" is equivalent to choosing each Ti independently and uniformly from 60, 13.

Lemma: Pr[ABr + Cn] 2 1/2.

Pf: Let D=AB-C nonzero. W.l.o.g., suppose D, +O.  $Pr[(Dn)_i = 0] = Pr[\sum_{j=1}^n D_{ij} n_j = 0]$  $= P_{\pi} \left[ \pi_{1} = -\frac{\sum_{j=2}^{h} D_{ij} \pi_{j}}{D_{11}} \right]$ 

$$\Pr_{n_1,\dots,n_n}\left[n_1=-\frac{1}{D_{ii}}\sum_{j=2}^n D_{ij}n_j\right]$$

$$= \frac{\sum_{\chi_{2,\dots,\chi_{n}}} P_{\chi_{1}} \left[ \chi_{1} = -\frac{1}{D_{11}} \sum_{j=2}^{n} D_{1j} \chi_{j} \left[ \chi_{2} = \chi_{2} \right] \right] P_{\chi_{2},\dots,\chi_{n}} \left[ \chi_{2} = \chi_{2} \right]}{\chi_{n} = \chi_{n}} P_{\chi_{n}} \left[ \chi_{1} = \chi_{2} \right]$$

$$\in \{0,1\}$$

$$= \sum_{\chi_{2,...,\chi_{n}} \in \{0,1\}} P_{\chi_{1}} \left[ \chi_{1} = -\frac{1}{D_{11}} \sum_{j=2}^{n} D_{ij} \chi_{j} \right] \cdot \frac{1}{2^{n-1}}$$

$$\underbrace{\sum_{\substack{\chi_{2,\ldots,\chi_{n}} \in \{0,1\}}} \frac{1}{2} \cdot \frac{1}{2^{n-1}} = \frac{1}{2} \cdot 2^{n-1} \cdot \frac{1}{2^{n-1}} = \frac{1}{2}$$

where the last inequality is because  $Pr[r_1=3] \le \frac{1}{2}$  for any 3 (the probability is exactly  $\frac{1}{2}$  if  $3 \in \{0,1\}$  and 0 otherwise).

Final algorithm: For i=1,..., t, choose  $r^{(i)} \in \{0,1\}^3$  independently and reject 'y  $A(Br^{(i)}) = Cr^{(i)}$  for any i. Otherwise, accept.

 $P_{T}$  [AB  $\neq$  C but also accepts]  $\leq \left(\frac{1}{2}\right)^{t}$ 

If we set t = [log 1/8], error probability < 8.

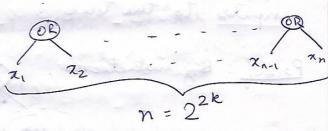
#### IT. AND/OR tree

 $T_{k}: \{0,13^{2k} \Rightarrow \{0,13\}$ 

Fact: Any deterministic algorithm, needs S2(n) time to

compute Tk.

(See optional exercises!)



## Randomized algorithm:

- Start at the root
- If (AND) node,
  - Randomly choose between left and right
  - Evaluate chosen subtree recursively
  - If O, return O
  - It I, evaluate other subtree recursively & return that value.

- If OR node,
  - Randomly choose between left and right
  - Evaluate chosen suttree neursebely.
  - If I, return 1.
  - If O, evaluate other subtree recursionally & return that value.

Define: Xk = max [# of steps to evaluate Te(x)]

Theorem:  $\mathbb{E}[X_k] \leq 3 \cdot \mathbb{E}[X_{k-1}] + O(1)$  $\Rightarrow \mathbb{E}[X_k] = O(3^k) = O(n^{0.793})$ 

Note: The only thing that's random here is the algorithm's behavior! Theorem holds for all inputs to the tree.

Proof: Regardations (AND)

Let's first look of the this OR this or the triangular triang

Time to compute the OR gate depends on the values of its subtree. Let's define:

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\left( x_{1}, y_{2} \right) & = & \sum_{x_{1}, y_{2}} \sum_{x_{2}} \left( x_{2}, y_{2} \right) \\
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Lemma: (i)  $\mathbb{E} \left[ OR(0,0) \right] \le 2 \cdot \mathbb{E} \left[ X_{k-1} \right] + O(1)$ (ii)  $\mathbb{E} \left[ OR(1,0) \right] \le 1.5 \quad \mathbb{E} \left[ X_{k-1} \right] + O(1)$   $\mathbb{E} \left[ OR(0,1) \right] \le 1.5 \quad \mathbb{E} \left[ X_{k-1} \right] + O(1)$ (iii)  $\mathbb{E} \left[ OR(1,1) \right] \le \mathbb{E} \left[ X_{k-1} \right] + O(1)$ 

- (1) When both subtrees evaluate to O, both must be evaluated no matter which is done first. So, E[OR(a,b)] < 2. E[Xe-1] + O(1). Here, we are implicitly using the fact that the algorithm's choice at the top DR node doesn't affect its choices in evaluating the subtrees.
- (ii) If a=1, b=0, the algorithm evaluates one subtree if left is chosen but evaluates Two subtrees if right is chosen.

E [OR (a, b)]

= \frac{1}{2} \text{E[OR(a,b)| left first]} + \frac{1}{2} \text{E[OR(a,b)| right finit]}

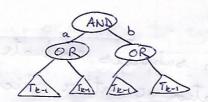
<== (E[X<sub>k-1</sub>]) + 1/2 (2 E[X<sub>k-1</sub>]) +0(1)

= 1.5 E[Xk-1]+0(1).

Same for a = 0, b = 1.

(iii) If both subtrees evaluate to 1, only one will be evaluated no matter which is done first. E[OR(1,1)] < E[XL-1] + O(1).

Now, look at the top AND gate:



Again, cost of evaluating AND depends on its subtrees

Let AND (a, b) = 
$$x_{1...,x_{n}}$$
: OR( $T_{k,1}(\cdot)$ ,  $T_{k,1}(\cdot)$ )=a # of steps  $x_{n}$ : OR( $T_{k,1}(\cdot)$ ,  $T_{k,1}(\cdot)$ )=b to evaluate  $T_{k}(x)$ 

Lemma: (i)  $\mathbb{E}[AND(0,0)] \leqslant 2 \cdot \mathbb{E}[X_{k-1}] + O(1)$ (ii)  $\mathbb{E}[AND(1,0)] \leqslant 2.75 \cdot \mathbb{E}[X_{k-1}] + O(1)$ ,  $\mathbb{E}[AND(0,1)] \leqslant 2.75 \cdot \mathbb{E}[X_{k-1}] + O(1)$ (iii)  $\mathbb{E}[AND(1,1)] \leqslant 3 \cdot \mathbb{E}[X_{k-1}] + O(1)$ .

Pf: (i) When both subtrees conservaluate to 0, only one is evaluated, no matter which is done first.

Each subtree takes OR(0,0) steps in the worst case. So:

EEXED E[AND(0,0)] ≤ E[OR(0,0)] + O(1) ≤ 2 E[X<sub>e-1</sub>] + O(1)

from part (i) of previous lemma.

(ii) Suppose a=1, b=0. The algorithm evaluates both subtrees if left is done first but only evaluates one if right is done first

E[AND(1,0)]=  $\frac{1}{2} \cdot E[AND(1,0) \mid night first] + \frac{1}{2} E[AND(1,0) \mid left first]$   $\leq \frac{1}{2} \cdot E[OR(0,0)] + \frac{1}{2} (max \{E[OR(1,1)] \oplus R(1,0) \} + E[OR(0,1)] + E[OR(0,$ 

The first inequality is because on OR gate can be I for three possible inputs: (1,1),(1,0),(0,1).

Same holds for E[AND(0,1)].

(iii) If both subtrees evaluate to 1, both need to be evaluated no matter which is done first. So:

E[AND(1,1)]

< 2. max { E[OR(1,0)], [E[OR(0,1)]} +0(1)

< 2. 1.5 E[X<sub>e-1</sub>] +0(1) = 3. E[X<sub>e-1</sub>] +0(1).

Finally, since Xx bounds the largest possible cost of evaluating the root:

 $E[X_{k}] \leq \max \{E[AND(0,0)], E[AND(1,0)], E[AND(0,1)], E[AND(1,0)] \}$  $\leq 3 \cdot E[X_{k-1}] + O(1).$ 

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## III. Randomized Quicksort Recall Rand QS from Week 1: pivot is chosen uniformly from subarray at each step. Suppose input array is a,,.., an (distinct). Rename the sorted version of this array as bi, ..., bn. Also, define bij = {bi, bi+1, --, bj-1, bj} for 15i ( ) 5 n. We prove the following: Theorem: If X is the total number of comparison made by Rand QS, E[X] = O (n log n) Proof: First, note that any pair of elements is compared at most once. Because elements are only compared to the pivot and subsequently, that pivot is not used in any recursive call. Define the indicator rendom variable Xij = 1 [bi and bj are compared by Rand QS] Since bi and by are compared at most once, $X = \sum_{i=1}^{n} \sum_{j=i+1}^{n} X_{ij}.$ $\Rightarrow$ $E[X] = \sum_{i=1}^{n} \sum_{j=i+1}^{n} E[X_{ij}]$

When are bi and by compared?

If some element not in bij is chosen as pivot, both bi and by fall on the same side of the partition.

Let x be the first element in big chosen as pivot.

- If b: < x < b; , then X ij = 0 because bi and bj foll on different sides of the partition and are never compared.
- · If z=bi or x=bz, then Xij=1.

So,  $\mathbb{E}[X_{ij}] = \Pr[\text{first element chosen as pivot}]$   $= \frac{2}{j-i+1}$ 

The last line is because each element in bij has equal chance of being selected as the first pivot

$$E[X] = \sum_{i=1}^{n} \sum_{j=i+1}^{n} \frac{2}{j-i+1}$$

$$= \sum_{i=1}^{n} \sum_{k=2}^{n-i+1} \frac{2}{k}$$

$$\leq \sum_{i=1}^{n} \sum_{k=1}^{n} \frac{2}{k} \leq 2 \sum_{i=1}^{n} 2 \ln n = O(n \log n)^{n}$$

Where we used that  $H_n := \sum_{k=1}^n \frac{1}{k} = \ln n + \Theta(1)$ .