

→ **Game theory**

→ **Lower bounds** on efficiency of Randomized Algorithms

→ Example: Tree evaluation

		Bob		
		R	P	S
Alice	→ R	0	-1	1
	→ P	1	0	-1
	→ S	-1	1	0

→ Game evaluation matrix

→ Alice is trying to **maximize** the outcome

→ Bob is trying to **minimize** the outcome

Generally this GEM is $[M_{ij}]$ of real numbers

If Alice chooses strategy i , in the worst case she gets $\min_j [M_{ij}]$

What is Alice's best strategy? $\max_i \min_j M_{ij} = O_A$

What is Bob's best strategy? $\min_j \max_i M_{ij} = O_B$
guaranteed outcome of strategy j

There are games for which $O_A = O_B$

→ **equilibrium**

		Bob	
		R	P
Alice	→ R	0	-1
	→ P	1	0
	→ S	2	1

MIXED STRATEGIES

Alice's strategy = probability distribution over rows P
 Bob's strategy = probability distribution over columns q column vectors of probabilities

$$P^T M q = \sum_{ij} P_i q_j M_{ij} = \text{Expected value of game } M \text{ with strategies } P \text{ and } q$$

For fixed strategy of Alice P she is guaranteed to achieve value of at least $\min q P^T M q$

For fixed strategy of Alice p she is guaranteed to achieve value of at least $\min_q p^T M q$

Alice's best strategy $\max_p \min_q p^T M q = O_A$

Bob's best strategy $\min_q \max_p p^T M q = O_B$

Von Neumann's theorem (Von Neumann's equilibrium)

$$\forall M \max_p \min_q p^T M q = \min_q \max_p p^T M q$$

Loewy's theorem

$$\max_p \min_k p^T M e_k = \min_q \max_j e_j^T M q$$

$e_j = (0, \dots, 1, \dots, 0)$
 j^{th} position

$$a^T = p^T M$$

Proof sketch

For fixed p : $p^T M q = a^T q = a_1 q_1 + \dots + a_n q_n$

find the smallest a_i
 and choose $q_i = 1$ and others = 0

$a^T q$ is linear in q and $\sum_i q_i = 1$.

Recall that randomized algorithms can be seen as mixtures of deterministic algorithms.

	Bob					
	A_1	A_2	A_3	...	A_n	
Alice	l_1	$C(I_1, A_1)$	$C(I_1, A_2)$	-	-	-
	l_2
	l_3

	l_n

} = M

deterministic algorithms
of which randomized algorithm chooses from

$E(C(I_p, A_q))$ = expected running time for input distribution p and rand. algorithm characterized by probability q .

= $p^T M q$.

by probability q .

$$= P^T M q.$$

VN's thm:

$$\max_P \min_q E(C(I_P, A_q)) = \min_q \max_P E(C(I_P, A_q))$$

Loomis thm:

$$\max_P \min_{A_i \in \mathcal{I}} E(C(I_P, A_i)) = \min_q \max_{i \in \mathcal{I}} E(C(I_i, A_q))$$

$\forall P, q$

$$\min_{A_i \in \mathcal{I}} E(C(I_P, A_i)) \leq \max_{i \in \mathcal{I}} E(C(I_i, A_q))$$

for a chosen input distribution find the best deterministic algorithm
for given algorithm (randomized) find the worst input

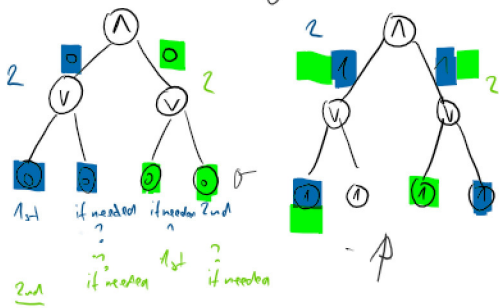
a lower bound!
we are interested in this!

Tree evaluation example

Example input distribution

all 0's w.p. $1/2 \sim$ (all nodes evaluate to 0)
 all 1's w.p. $1/2 \sim$ (all nodes evaluate to 1)

Deterministic alg

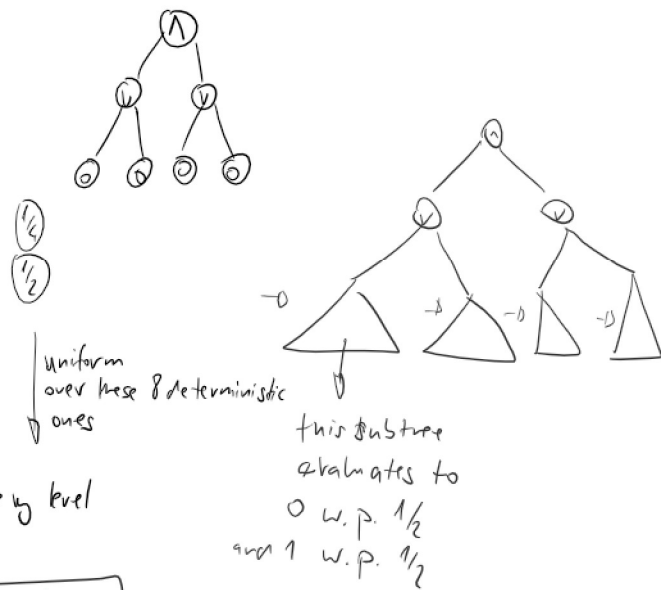


first node to evaluate \sim 4 choices
 choice of hex node \sim 2 choices

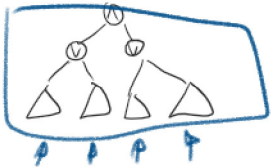
In total 8 deterministic algorithms for every level

Any randomized algorithm needs at least 2 evaluations for each level $\Rightarrow 2^k$ complexity with k levels,

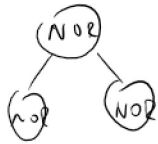
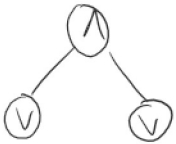
Example distribution? :



Example distribution?



1 1 0 0
0 0 1 1



$$(a \vee b) \wedge (c \vee d)$$

||

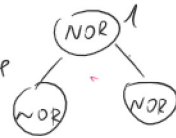
$$(a \text{ NOR } b) \text{ NOR } (c \text{ NOR } d)$$

ab	a NOR b
00	1
01	0
10	0
11	0

$$Pr(\text{leaf}=1) = p$$

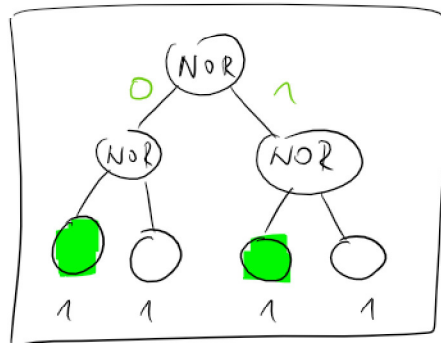
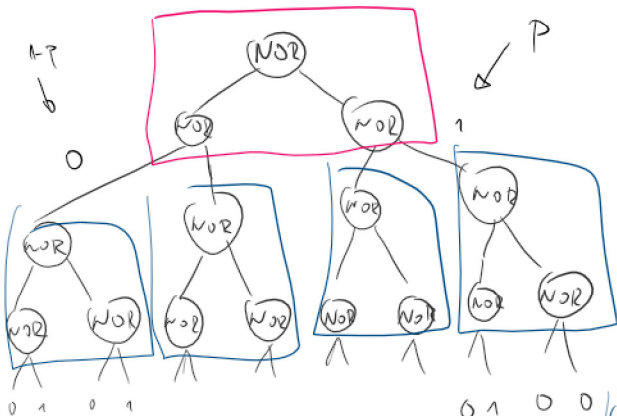
$$Pr(\text{leaf}=0) = (1-p)$$

$p = (1-p)^2$ then the probability of a tree evaluating to 1
 $p = \frac{3-\sqrt{5}}{2}$ is p and to 0 $1-p$.



evaluating to 1

H is recursive!



	2nd	?	1st	?		
$(1-p)^3 p$	0	0	0	1	4	0.09
$(1-p)^2 p$	0	0	1	0	3	0.09
$(1-p)^2 p^2$	0	0	1	1	3	0.05
0	1	0	0	0	2	0.09

$$\sum_{\text{input}} p(\text{input}) \cdot h(\text{input}) \geq 2$$

0	1	0	1	4	0,05
0	1	1	0	3	0,05
0	1	1	1	3	0,03
1	0	0	0	2	0,09
1	0	0	1	3	0,05
1	0	1	0	2	0,05
1	0	1	1	2	0,03
1	1	0	0	2	0,05
1	1	0	1	3	0,05
1	1	1	0	2	0,03
1	1	1	1	2	0,02
0	0	0	0	2	0,14

$p = \frac{2-\sqrt{3}}{2} \approx 0,38 \quad (1-p)^3 = 0,24$
 $p^2 \approx 0,15 \quad (0,038) = (1-p)^1$
 $p^3 \approx 0,06 \quad (1-p) = 0,62$
 $p^4 \approx 0,02$

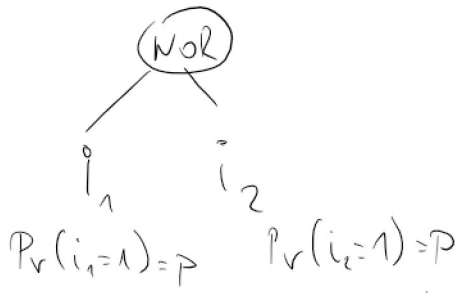
$(1-p)^4 = p^2 < 0,14$
 $p \cdot (1-p)^3 < 0,09$
 $p^2 \cdot (1-p)^2 = p^3 < 0,05$
 $p^3 \cdot (1-p) < 0,03$
 $p^4 < 0,02$

always rounding down.

Expected number of steps: $4 \cdot (0,09 + 0,05) + 3 \cdot (0,09 + 4 \cdot 0,05 + 0,03)$
 $+ 2 \cdot (2 \cdot 0,09 + 2 \cdot 0,05 + 2 \cdot 0,03 + 0,14 + 0,02)$
 \parallel
 $0,56 + 0,96 + 1$
 \parallel
 $2,52 < 3$

\checkmark lower bound with all 0 or all 1 inputs \downarrow upper bound from historical IV

Easier derivation consistent with slides:

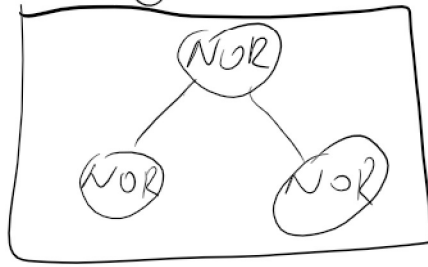


Need to examine only one input

need to examine both inputs

each NOR will take exactly $p + (1-p) \cdot 2 = (2-p) < 1.61$ steps to

evaluate. Therefore



takes roughly

$$p \cdot (2-p) + (1-p) \cdot 2 \cdot (2-p) = (2-p)^2 < 2.61 \text{ steps to evaluate}$$

Finally we have

$$2 < \underline{2.52} < 2.61 < 3$$

↓
this is a bit lower due to probability rounding down