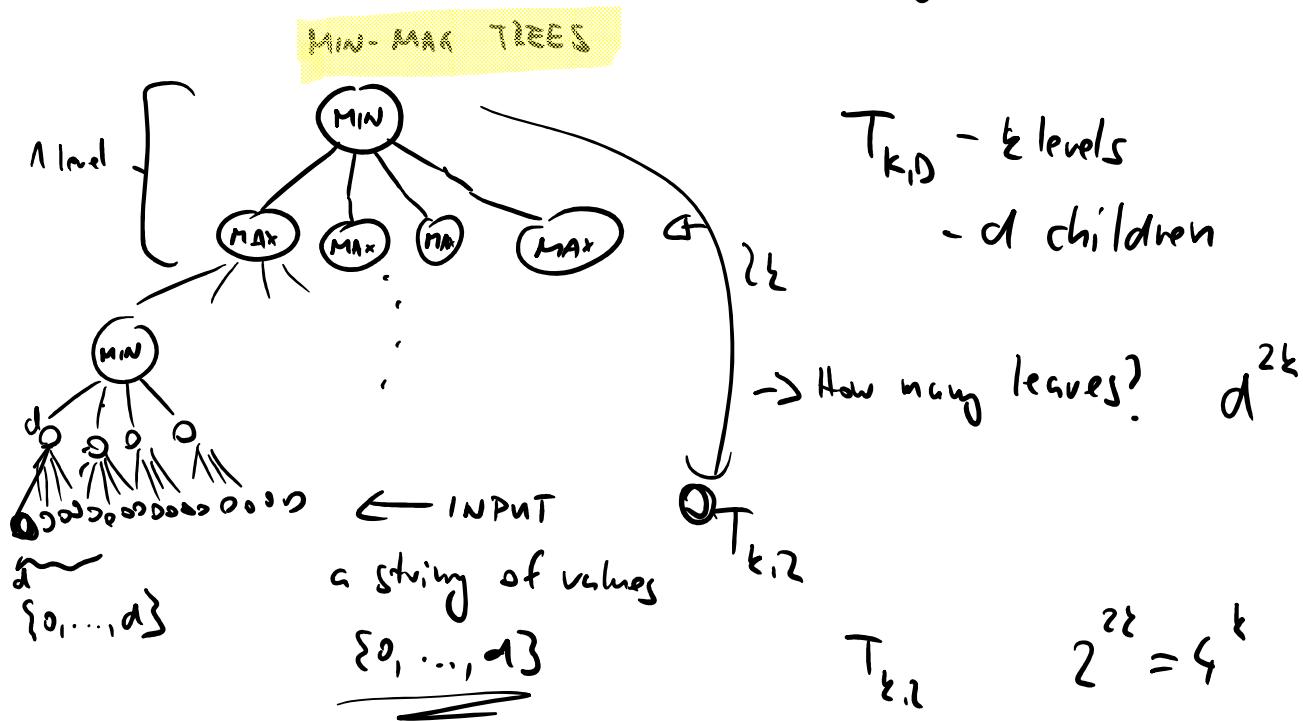


## GAME THEORETIC TECHNIQUES

↳ Game tree evaluation

↳ Yao's min-max technique for proving lower bounds



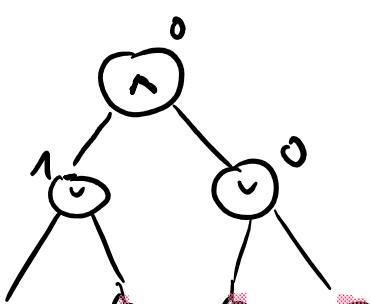
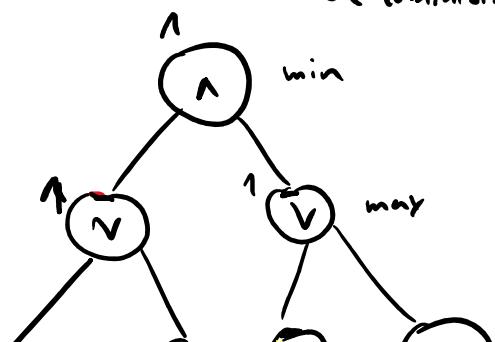
Each leaf contains a value.

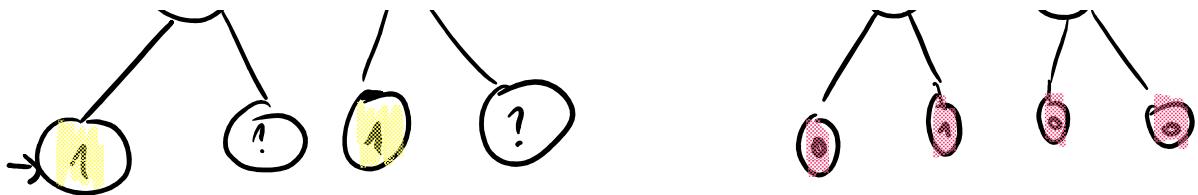
Each **min** node contains the smallest value of all it's children

Each **MAX** node contains the largest value of all it's children

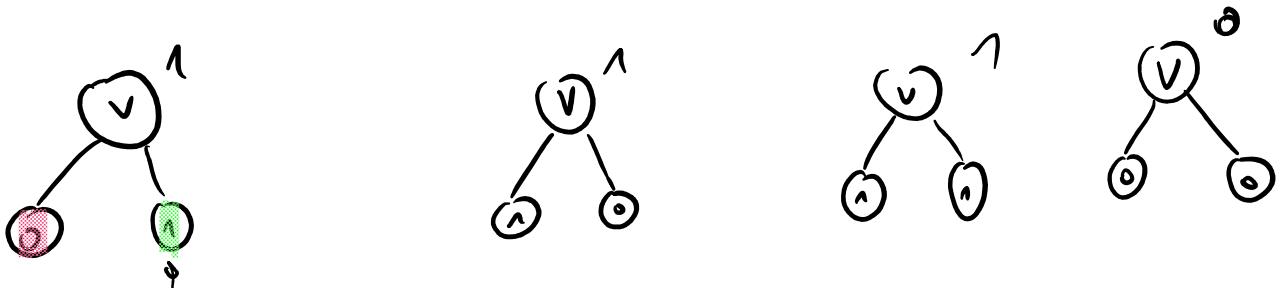
Goal is to find the value of the root,

$T_{k,2} \rightsquigarrow$  binary trees with input  $\times$  a string  $\{0,1\}^k$   
 Deterministic algorithm: Depth-first left-right





Randomized algorithm.  
Order of children evaluation is randomized.



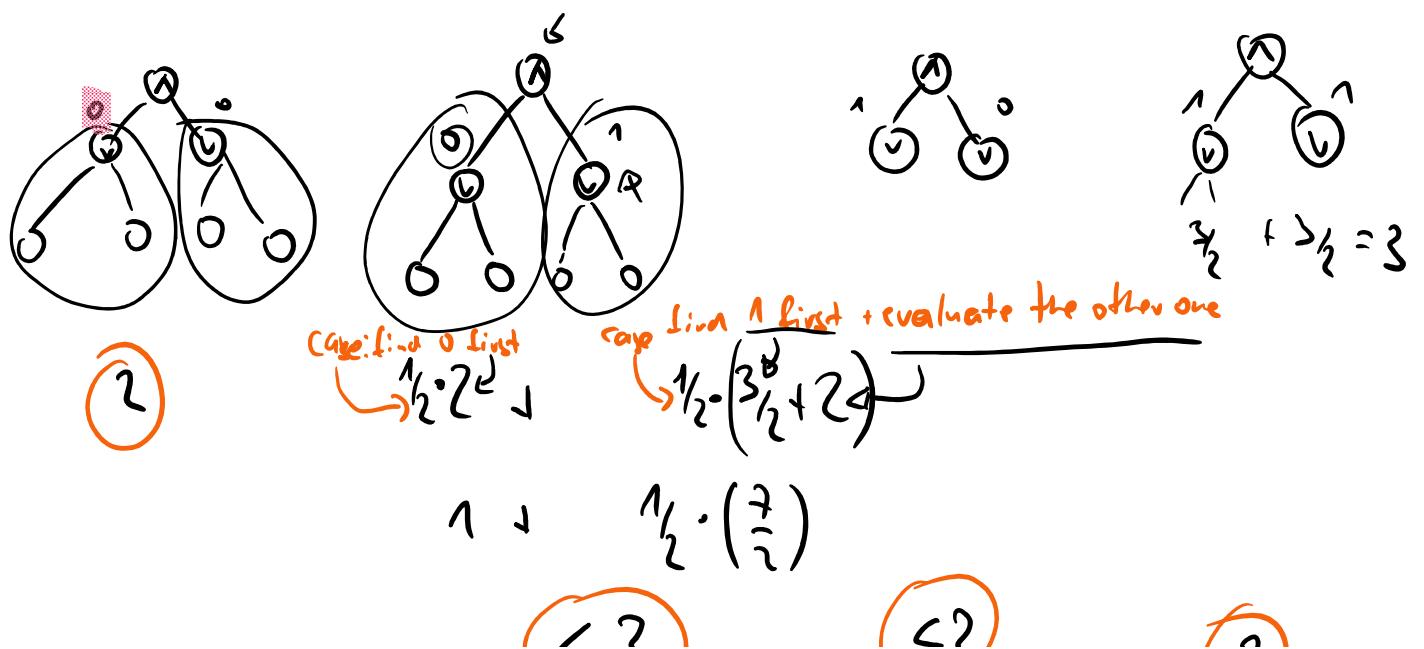
**Red** First = 2 evaluations       $\frac{3}{2}$  evaluations      1 evaluation      2 evaluations

**Green** First = 1 evaluation

Average =  $\frac{3}{2}$  evaluations

act	2	1	1	2
rand	$\frac{3}{2}$	$\frac{3}{2}$	1	2
	—	—	—	—

Induction  $k \geq 1$



$\leftarrow 3$

$\leftarrow 3$

$\leftarrow 3$

$$\mathbb{E}(x) = \frac{1}{2} \cdot \mathbb{E}(\textcircled{1}) + \frac{1}{2} \cdot \mathbb{E}(\textcircled{2})$$

Overall average  $\leftarrow 3$

On average  $T_{\epsilon,2}$  takes less than  $3^{\epsilon}$  leaves to evaluate

I-H.  $T_{\epsilon-1,2}$  takes  $3^{\epsilon-1}$  to evaluate

I.S.  $\nexists T_{\epsilon,2}$  takes  $3^{\epsilon}$ .

$$n = 2^{72} \quad 3^{\epsilon} = n^{0.79..}$$

### Basics of game theory

		Bob			G Strategies
		R	P	S	
Alice	R	0	-1	1	
	P	1	0	-1	
	S	-1	1	0	

$\rightarrow$  Game evaluation matrix

$\rightarrow$  Alice is trying to maximize the outcome

$\rightarrow$  Bob is trying to minimize the outcome

$$\begin{aligned} P \text{ strategies} \quad O_A &= -1 \\ O_B &= 1 \end{aligned}$$

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

if Alice chooses strategy  $i$ , in the worst case

if Alice chooses strategy  $i$ , in the worst case  
she gets  $\min_j M_{ij}$ .

What is Alice's best choice?

$$\max_i \min_j M_{ij} = O_A$$

What is Bob's best choice?

$$\min_j \max_i M_{ij} = O_B$$

There are games, for which  $O_A = O_B$



$$0 -1 -2 \quad O_A = 0$$

$$1 \quad 0 \quad -1 \quad O_B = 0$$

$$\rightarrow 2 \quad 1 \quad \boxed{0}$$

### Mixed strategies

Alice = probability distribution on rows  $P$

Bob = probability distribution on columns  $q$

$P$  and  $q$  are column vectors of probability

$$P^T M q = \sum_i \sum_j P_i q_j M_{ij} = \text{Average value of game with mixed strategies } P \text{ and } q.$$

for fixed strategy of Alice  $P$ . She is guaranteed to get at least  $\min_q P^T M q$  points

Alice's best strategy is  $\max_P \min_q P^T M q$

Bob's best strategy is  $\min_q \max_P P^T M q$

Von Neumann's thm.

$$\max_P \min_q P^T M q = \min_q \max_P P^T M q$$

Loomis thm

$$\max_P \min_k P^T M e_k = \min_q \max_i e_i^T M q,$$

where  $e_i = (0, \dots, \overset{i^{th} \text{ position}}{1}, \dots, 0)$

Proof.

for fixed  $P$   $\underbrace{P^T M q}$  is a function linear in  $q_1, \dots, q_n$  scalars variables

$$P^T M q = a_1 q_1 + a_2 q_2 + \dots + a_n q_n$$

$$q_1 + q_2 + \dots + q_n = 1$$

find smallest  $a$  ( $\omega \log a_k$ )

and min is obtained by  $q = (0..1..0)$

$q$

$k^{th}$  position

	$A_1 \ A_2 \ \dots \ A_n$
$I_1$	$C(I_1, A_1)$
$I_2$	
$I_3$	
$\vdots$	
$I_n$	

$C(I_j, A_k)$  is the length of computation of  $A_k$  on input  $I_j$ .

$E[C(I_p, A_q)]$  = expected running time for input distribution  $p$  and rand. algorithms specified by  $q$ .

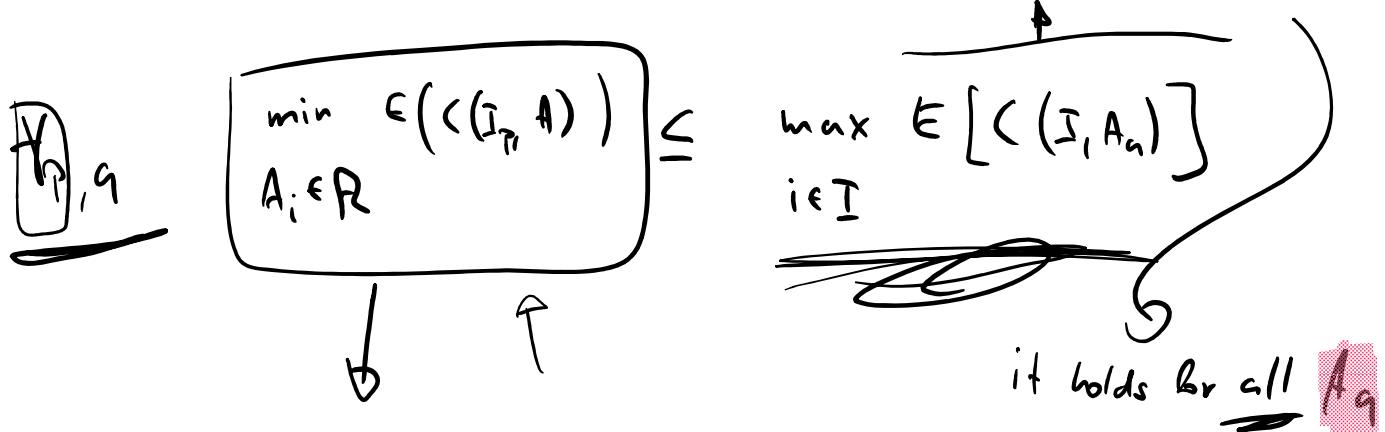
Neuman's thm:

$$\max_p \min_q E(C(I_p, A_q)) = \min_q \max_p E(C(I_p, A_q))$$

Loomis thm:

$$\max_P \min_{A_i \in \mathcal{A}} E(C(I_p, A_i)) = \min_q \max_{i \in I} E(C(I_q, A_i))$$

the worst average for a problem

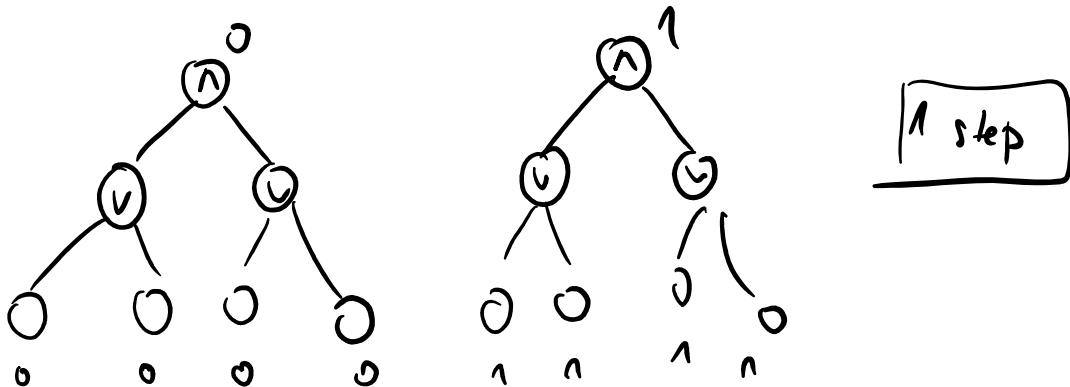


We need to find distribution  
on the inputs which has an optimal  
deterministic algorithm, with complexity as high as possible.

### Example

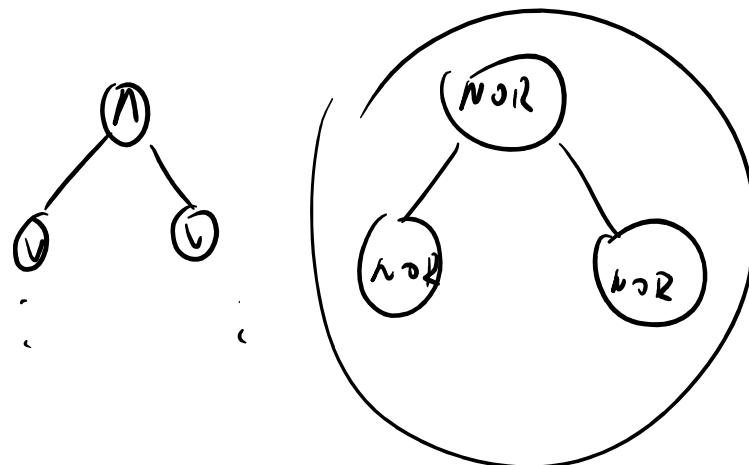
Input for tree evaluation is always all 0  
or always all 1 vector

What is the optimal det. algorithm?



$$\begin{array}{c}
 \left[ \begin{array}{cccc} 1 & 0 & 0 & \rightarrow 0 \\ 0 & 1 & 1 & \rightarrow 0 \end{array} \right] \\
 \left[ \begin{array}{ccc} 0 & 1 & 0 & \rightarrow 1 \\ 1 & 0 & 1 & \rightarrow 1 \end{array} \right]
 \end{array}$$

2 steps



$$(a \vee b)_n (c \vee d)_n = (a \vee b)_m (c \vee d)_m$$

$$\begin{array}{ccccc}
 & 0 & 0 & & \\
 \xrightarrow{\delta} & \boxed{\begin{array}{l} \Pr(\text{leaf} = 1) = p \\ \Pr(\text{leaf} = 0) = 1-p \end{array}} & & & \\
 & \phi & & &
 \end{array}$$

$\nwarrow$ $\downarrow$	$\begin{matrix} 0 & 0 \\ 0 & 1 \\ 0 & 0 \\ 1 & 0 \\ 1 & 1 \end{matrix}$	$\nearrow$ $\uparrow$
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$$\begin{array}{ccc}
 n^{0.6\dots} & \leq & n^{0.79\dots} \\
 n^{0.76\dots} & = & n^{0.79\dots}
 \end{array}$$

$$\begin{aligned}
 (1-p)^2 &= p \\
 p &= \frac{(1-p)}{2}
 \end{aligned}$$