

# Decision Trees

Lecture 23  
To left or to right

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- Interested in lower-bounds
  - So even allow unbounded computational power
  - Simpler combinatorial structure (need not understand P vs. NP etc.)



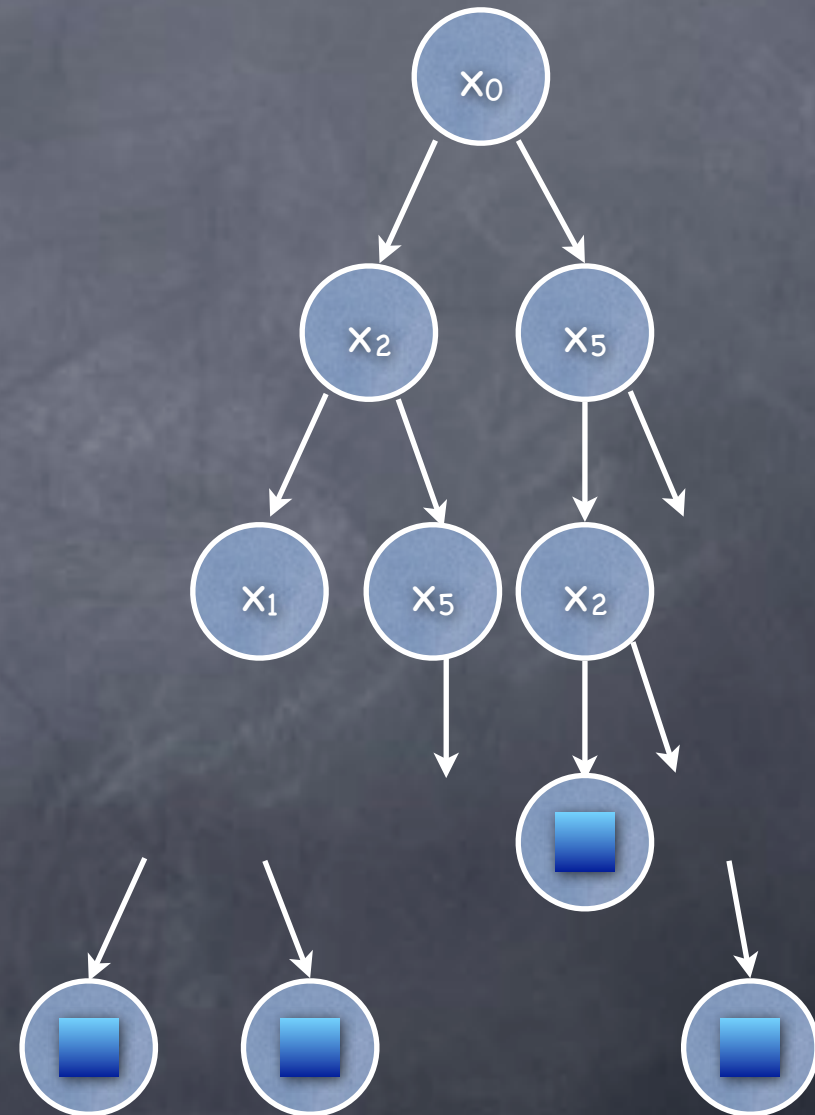
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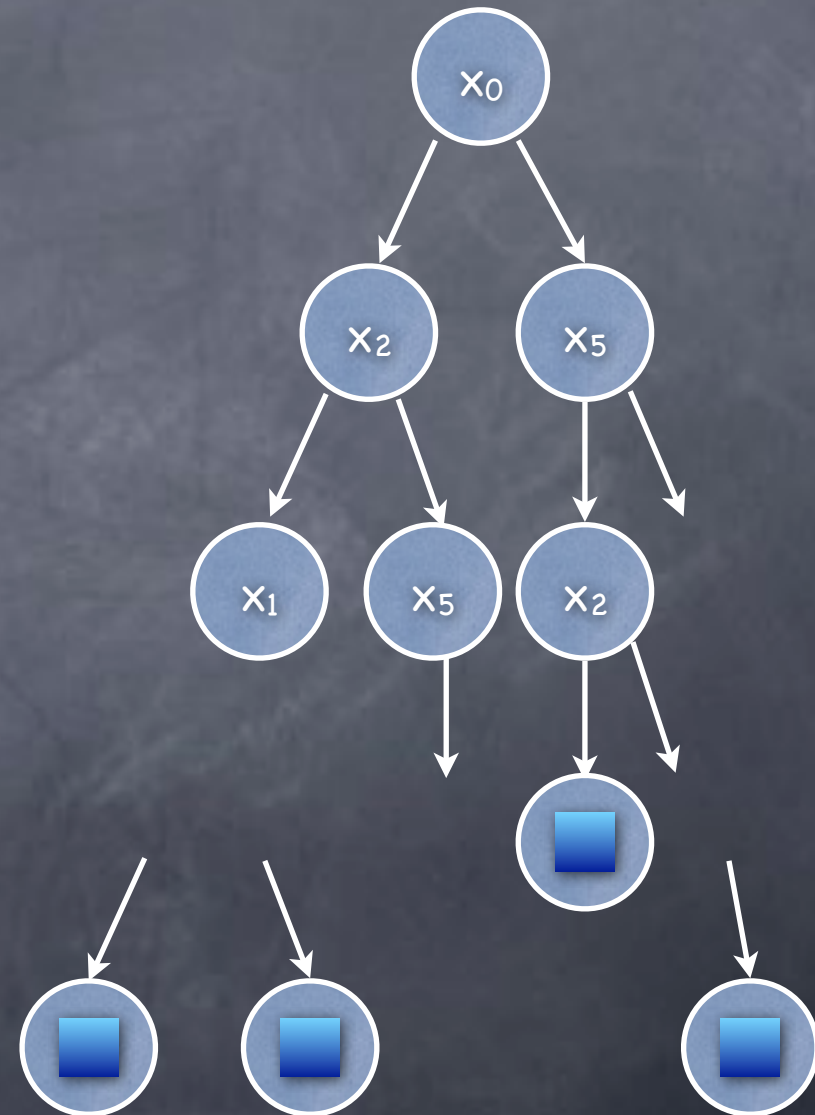
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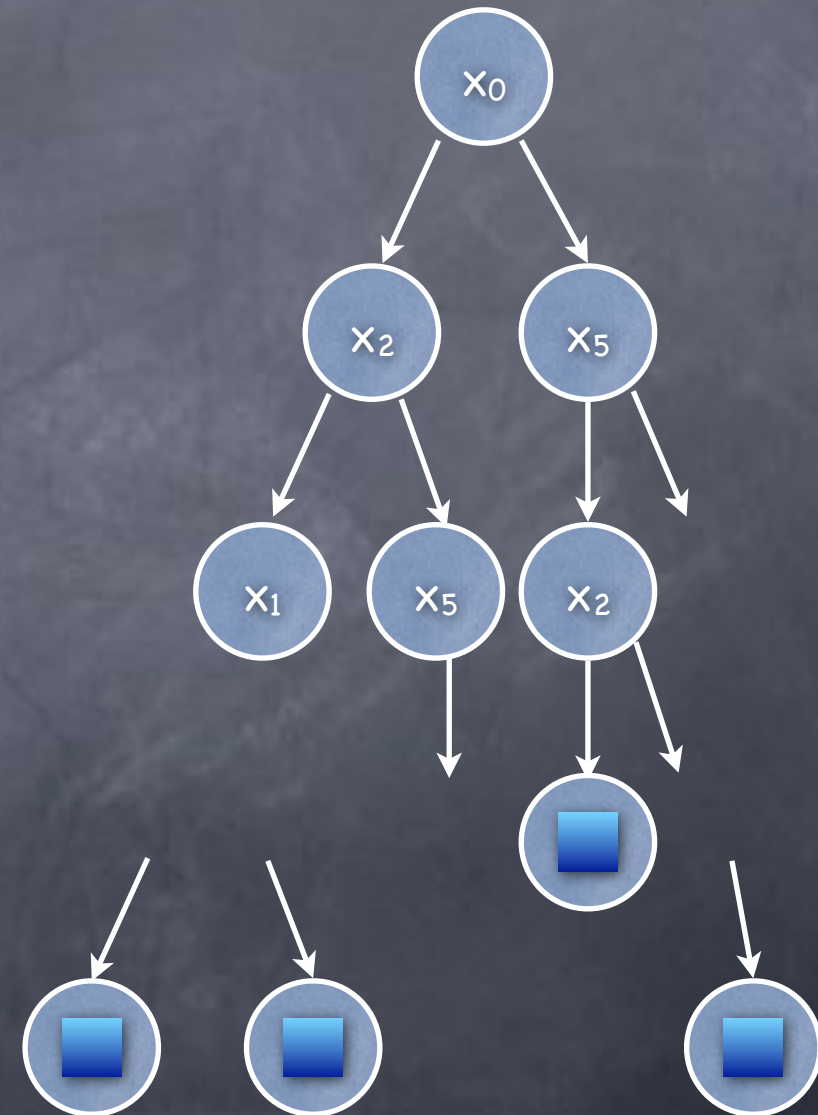
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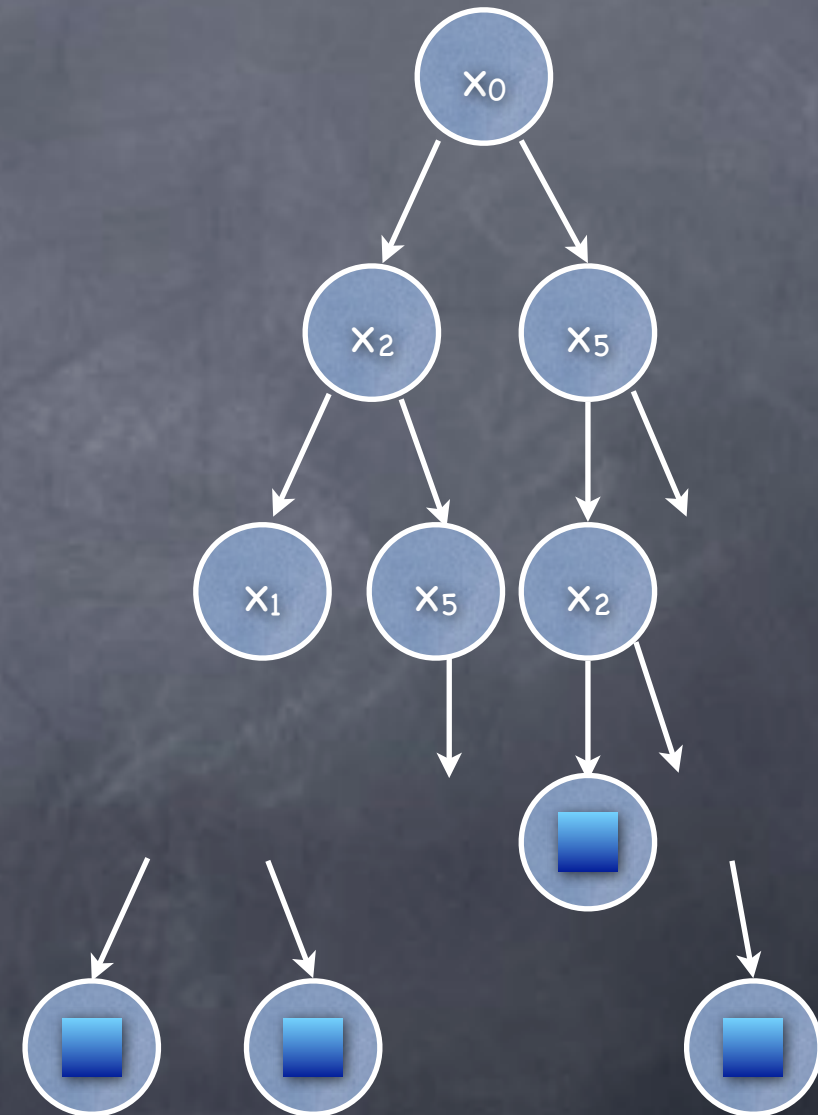
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- For  $n$ -bit input, depth at most  $n$
- Some paths may be shorter
- $\text{DTree}(L) = \min_{\text{alg } A} \max_{\text{input } x} T_{A,x}$   
where  $T_{A,x}$  is the number of bits of  $x$  read by  $A$





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- We are interested in showing DTree lower-bounds for these problems

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  - Before  $n$  nodes, set of inputs contain  $0^n$  and another input, no matter what bits were queried at the nodes

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  - Until then, graph can be connected or disconnected: by setting all unqueried edges to Yes or all to No

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  - $\text{Maj}(x) = 1$  iff #1s in  $x > \text{\#0s}$  (assume  $|x|$  odd)
    - Adversary strategy: alternately answer 0 and 1

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  - Exercise

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  - Clearly correct. Number of bits read?



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  - In each iteration at most  $\text{OCert}(L)$  bits queried

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  - 0-certificate: enough variables so that can evaluate just one input wire for AND gates, and all input wires for OR gates
  - 1-certificate: enough variables so that can evaluate just one input wire for OR gates, and all input wires for AND gates
  - If regular AND-OR tree,  $\text{OCert}(L) \times \text{1Cert}(L) = \text{number of leaves} = \text{DTree}(L)$



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  - “**Sensitivity**” is a lower-bound on  $\text{DTree}(L)$
- Will explore some in **exercises**



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    - i.e., randomly choose a (deterministic) decision tree

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- Question: How to prove lower-bounds against randomization?

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- Both have the same expected cost!! (not obvious: follows from LP duality)

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- Useful: Can show lower-bound for randomized algorithms via lower-bound on **distributional complexity for deterministic algorithms**