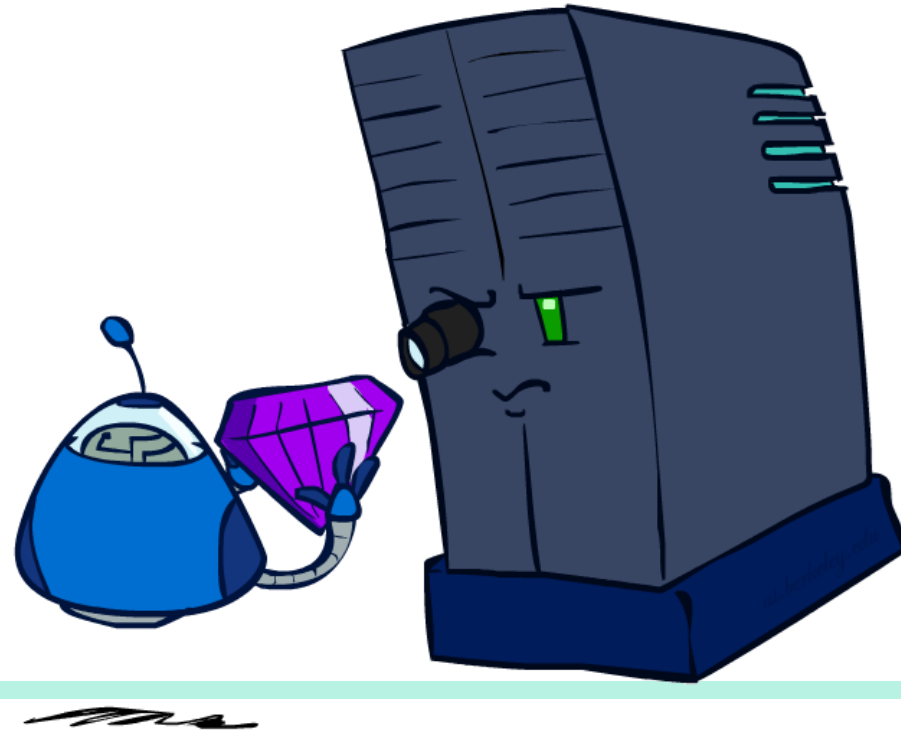


Advanced Topics in AI

Evaluation functions



Instructor: Prof. Dr. techn. Wolfgang Nejdl
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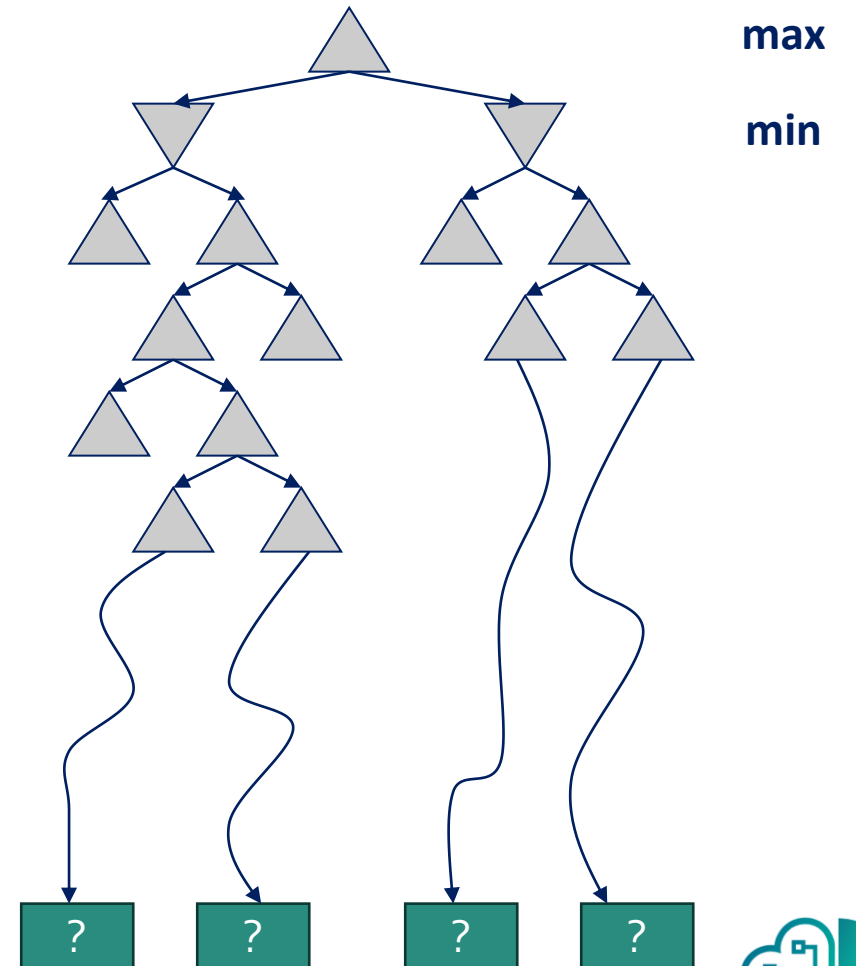
[These slides were created by Dan Klein and Pieter Abbeel for CS188 Intro to AI at UC Berkeley. All materials are available at <http://ai.berkeley.edu>.]

Overcoming Resource Limits



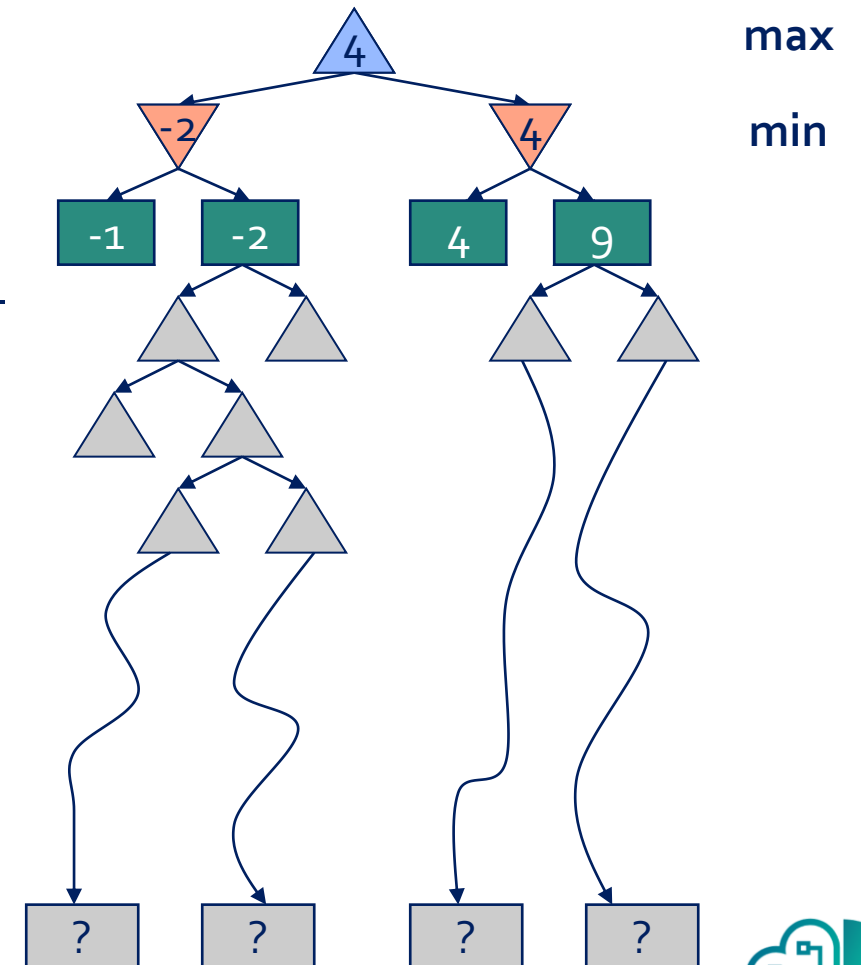
Resource Limits

- Problem: In realistic games, cannot search to leaves!
- Solution: Depth-limited search
 - Instead, search only to a limited depth in the tree
 - Replace terminal utilities with **an evaluation function** for non-terminal positions

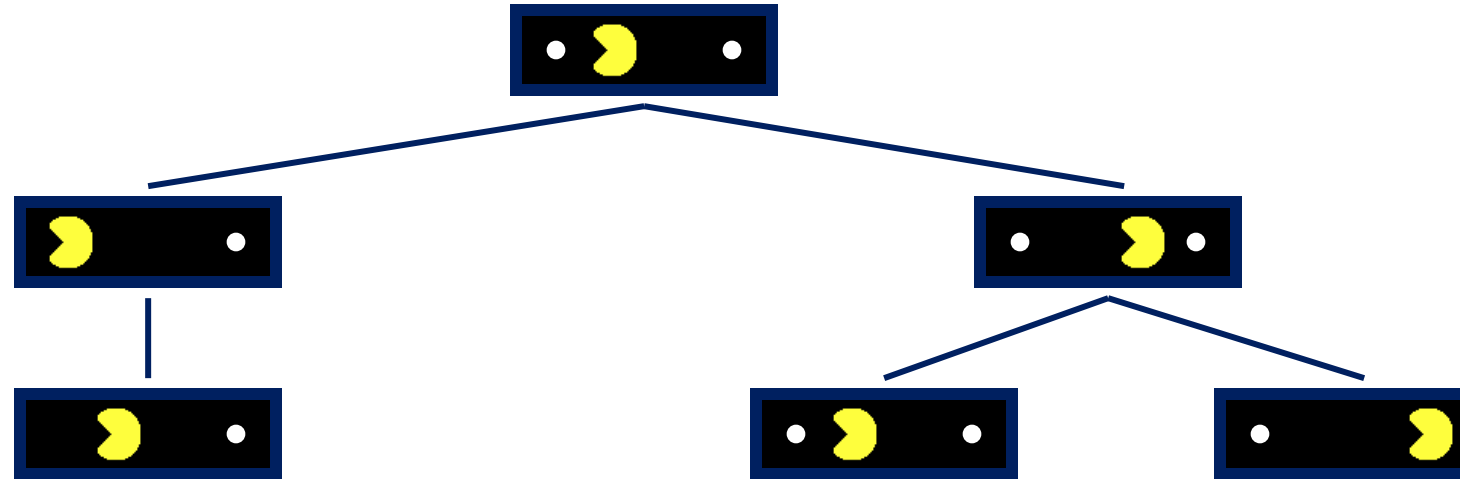


Resource Limits

- Problem: In realistic games, cannot search to leaves!
- Solution: Depth-limited search
 - Instead, search only to a limited depth in the tree
 - Replace terminal utilities with **an evaluation function** for non-terminal positions
- Example:
 - Suppose we have 100 seconds, can explore 10K nodes / sec
 - So can check 1M nodes per move
 - α - β reaches about depth 8 – decent chess program
- Guarantee of optimal play is gone
- More plies makes a BIG difference
- Use iterative deepening for an anytime algorithm



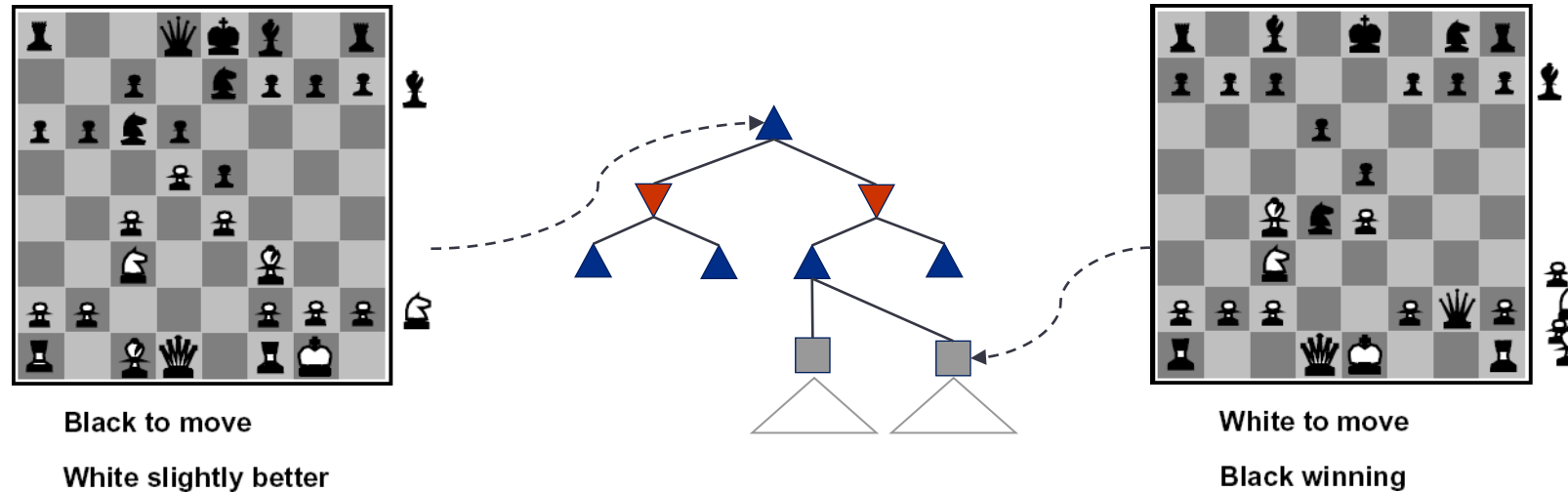
Why Pacman Starves



- A danger of replanning agents!
 - He knows his score will go up by eating the dot now
 - (west, east)
 - He knows his score will go up just as much by eating the dot later
 - (east, west)
 - There are no point-scoring opportunities after eating the dot
 - (within the horizon, two here)
 - Therefore, waiting seems just as good as eating:
 - he may go east, then back west in the next round of replanning!

Evaluation Functions

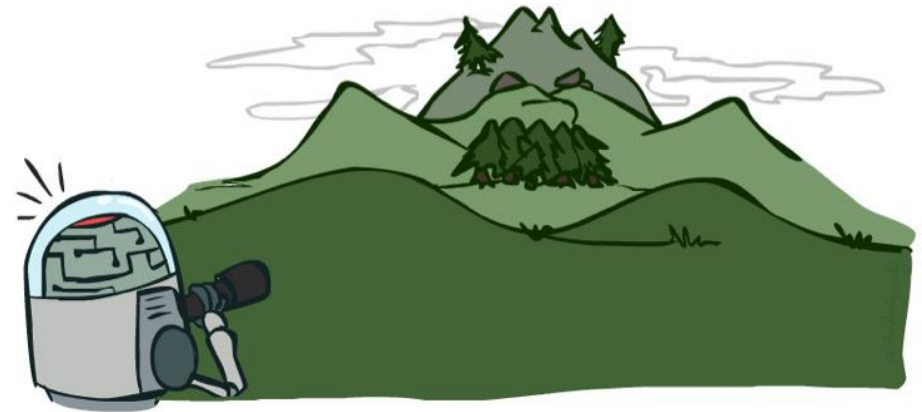
- Evaluation functions score non-terminals in depth-limited search



- **Ideal function:** returns the actual minimax value of the position
- **In practice:** typically weighted linear sum of features:
 - $Eval(s) = w_1f_1(s) + w_2f_2(s) + \dots + w_nf_n(s)$
 - e.g. $f_1(s) = (\text{num white queens} - \text{num black queens})$, etc.

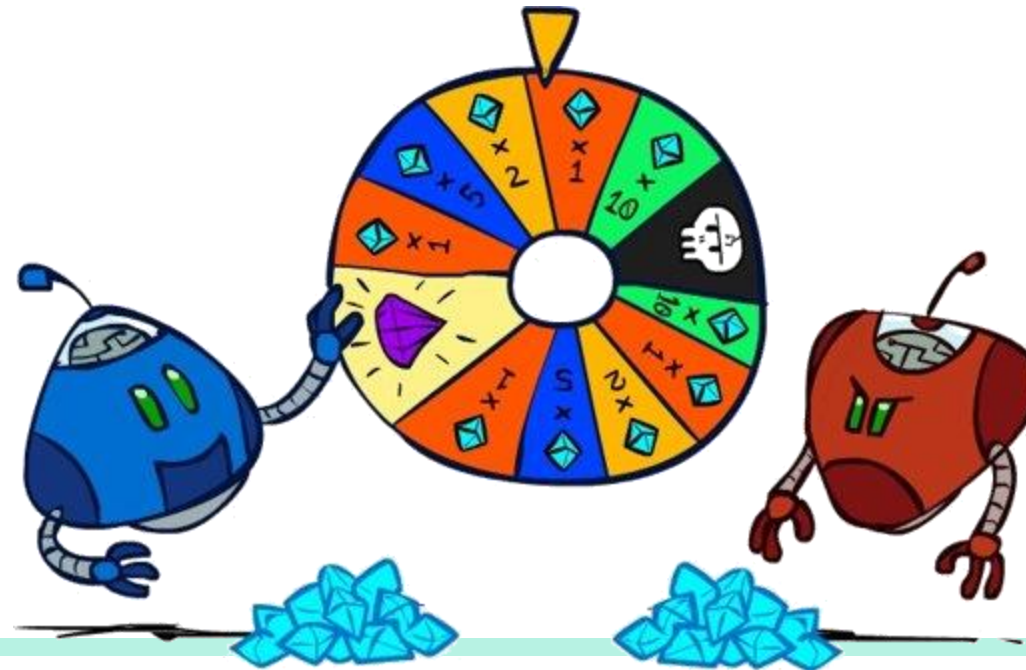
Depth Matters

- Evaluation functions are always imperfect
- The deeper in the tree the evaluation function is buried, the less the quality of the evaluation function matters
- An important example of the tradeoff between complexity of features and complexity of computation



Advanced Topics in AI

Next: Uncertainty & Expectimax



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