

# Programmieren 1

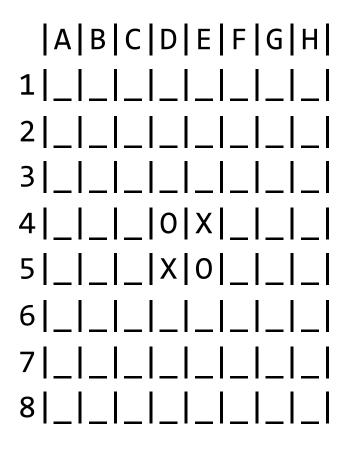
Die Reversi-Challenge



Prof. Dr. Michael Rohs michael.rohs@hci.uni-hannover.de



## Programmieren I Winter Challenge

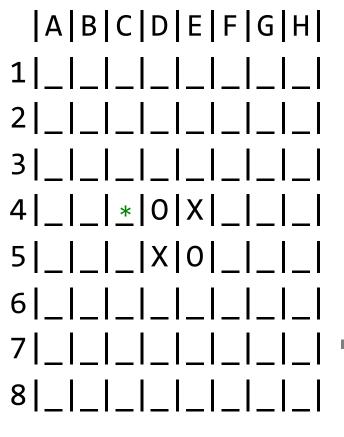


- Programmieren Sie einen Computer-Reversi-Spieler
- Reversi Turnier
  - alle eingereichten Spiele treten gegeneinander an
- Teilnahme freiwillig, zählt nicht als reguläres Übungsblatt
- zusätzliche Übungspunkte
  - 1 Punkt, beste 50%
  - 2 Punkte, beste 10%



#### Reversi Rules

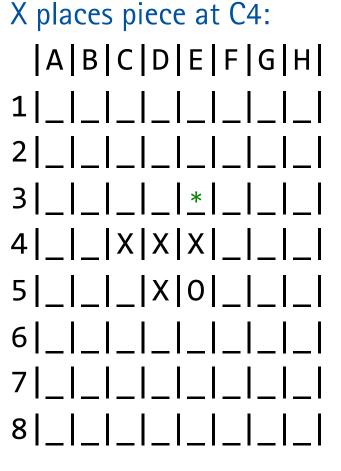
#### Initial board:



- Rules
  - X moves first
  - Pieces are reversed if surrounded (on two sides) by opponent pieces
  - A valid move is one in which at least one piece is reversed
  - If one player cannot make a move, play passes back to the other player
  - When neither player can move, the game ends, the player with the most pieces wins
- Details
  - https://en.wikipedia.org/wiki/Reversi



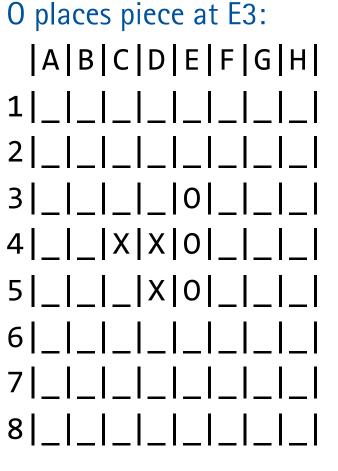
#### Reversi Rules



- Rules
  - X moves first
  - Pieces are reversed if surrounded (on two sides) by opponent pieces
  - A valid move is one in which at least one piece is reversed
  - If one player cannot make a move, play passes back to the other player
  - When neither player can move, the game ends, the player with the most pieces wins
- Details
  - https://en.wikipedia.org/wiki/Reversi



#### Reversi Rules



- Rules
  - X moves first
  - Pieces are reversed if surrounded (on two sides) by opponent pieces
  - A valid move is one in which at least one piece is reversed
  - If one player cannot make a move, play passes back to the other player
  - When neither player can move, the game ends, the player with the most pieces wins
- Details
  - https://en.wikipedia.org/wiki/Reversi



## Programmieren I Winter Challenge

- Aufgabe
  - möglichst guten Reversi-Spieler programmieren
  - 1 sec pro Zug (sonst disqualifiziert, siehe Template)
  - C-Template & Details: WinterChallenge.{zip|pdf} auf Stud.IP
- Abgabe bis 15.1.
  - Quelltext (eine C-Datei)
- Preise
  - Urkunde und Buchpreis für den besten Computer-Reversi-Spieler
- Auswahl
  - Vorrunde: zufällige Paarung, spielen gegeneinander, sammeln Punkte
  - Top 16 (Achtelfinale) spielen gegeneinander beim Reversi-Turnier
  - Termin: letzte Semesterwoche (23.–27.1.), wird noch bekannt gegeben

Das Reversi-Programm darf nicht (z.B. mit einem Server) kommunizieren und darf nur selbst entwickelte Komponenten enthalten (außer der prog1lib und der C-Standardbibliothek).

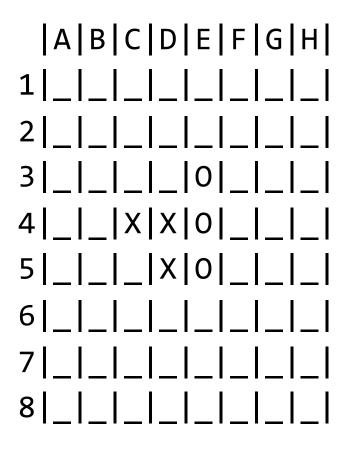


# Two-Player Finite Zero-Sum Strategy Games with Perfect Information

- Two-player: No coalitions possible
- Finite: Finite number of steps till the game is over
- Zero-sum: One player's gain equals another player's loss
- Strategy games: No element of chance beyond the players' moves
- Perfect information: Each player knows the full state of the game
- Basic assumption: Players act in a rational way and aim to maximize their win
- Other examples: Tic-Tac-Toe, Chess, Go



## What should a computer player do?



- What is the best move given the current game state and the list of possible moves?
- Evaluate quality of game states
- Find good moves: precompute many moves
  - If I do this move, what will my opponent's do?
  - How will can I react?
  - What will the opponent do then?



## Estimating the Score (Payoff Value)

- Estimate the quality of the game state
  - Heuristics
- Different possible attributes
  - Number of my pieces minus number of opponent pieces?
    - Score from X's perspective: 4 1 = 3
    - Score from 0's perspective: 1 4 = -3
  - Number of "safe" pieces of a color?
    - A piece in A1 cannot be reversed
  - Evaluate positions on the board, e.g., corners, edges, center
  - Mobility: Number of possible moves for player A or B
  - Number of flipped pieces in a move

|   | Α          | В          | С          | D | E | F | G | Н        |
|---|------------|------------|------------|---|---|---|---|----------|
| 1 | <b> </b> _ | <b> </b> _ | <b> </b> _ | _ | _ | _ | _ | _        |
| 2 | _          | <b> </b> _ | <b> </b> _ | _ | _ |   |   | <u> </u> |
|   | <b> </b> _ |            |            |   |   |   |   |          |
| 1 | _          | <b> </b> _ | X          | X | X |   |   | _        |
| 5 | <b> </b> _ | <b> </b> _ | <b> </b> _ | X | 0 | _ |   | _        |
| 5 | _          | _          | _          | _ | _ |   |   | _        |
|   | _          |            |            |   |   |   |   |          |
| 3 |            |            |            |   |   |   |   |          |



### **Precomputing Moves**

- Board games are simple, small worlds with clear rules
- Consequences of game moves can be precomputed
  - Precompute many moves and evaluate each one, choose "best" one
  - Little "intelligence", much computing power
  - Search space typically too large to compute every state

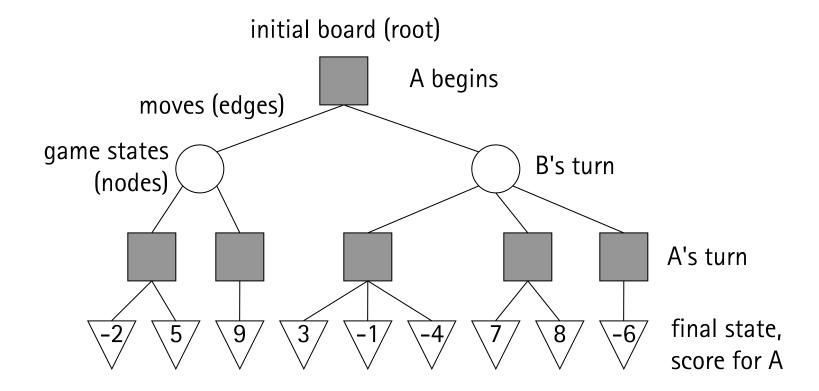
#### Procedure

- Given the current game state (e.g., pieces on the board)
- Determine the set of possible moves for a player A
- Estimate the benefits of each move (precompute, estimate)
- Do the move with the best estimate for player A
   (= greatest loss for player B)



#### **Game Trees**

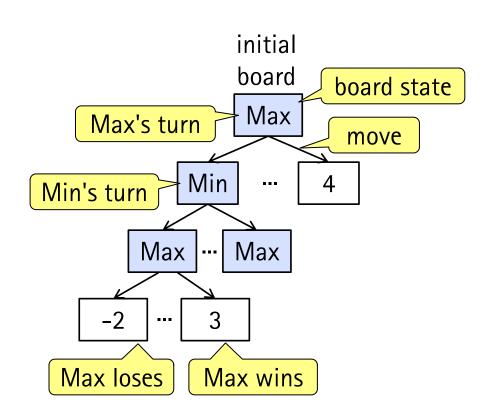
- Description of possible game states and moves
- For the analysis of games





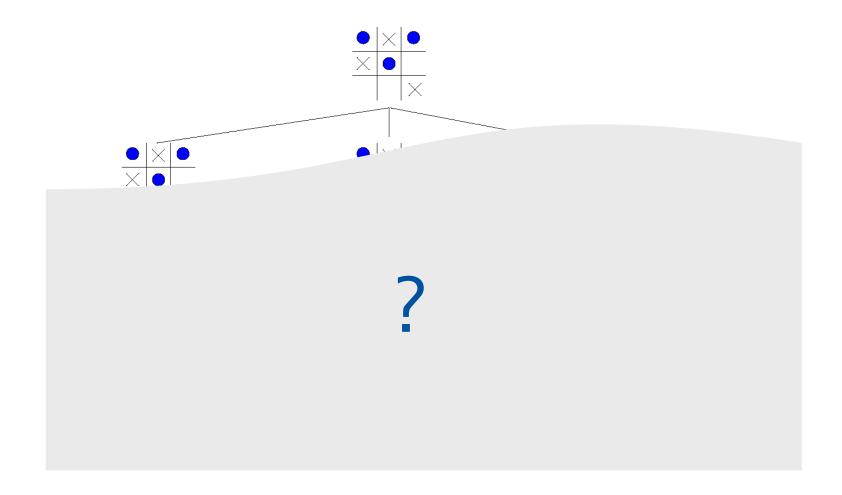
#### **Game Trees**

- Nodes represent possible states of the game board
- Edges represent possible moves
- Player "Max" and "Min" take turns
  - Convention: Max begins
- Leaves denote end situations
  - Contain score for Max
  - Max tries to reach a leaf with a high value
  - Min tries to reach a leaf with a low value
- Real game trees are very large
  - Chess:  $\sim 10^{120}$  nodes, Go:  $\sim 10^{761}$  nodes



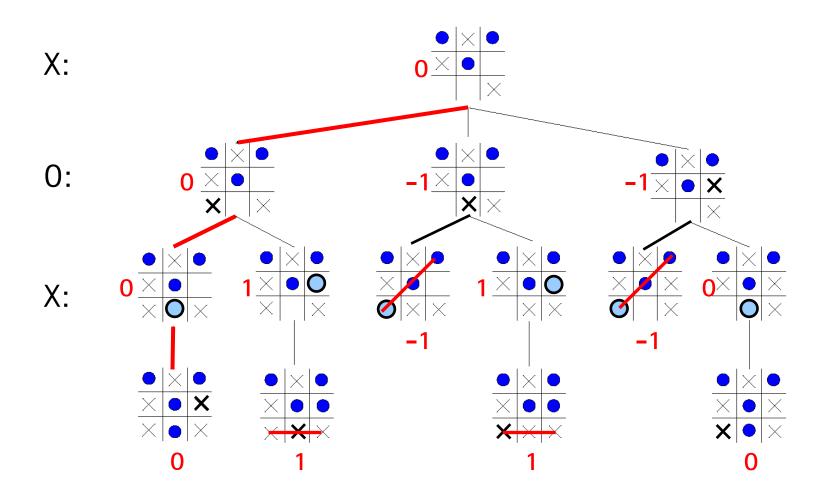






## Game Tree for Tic-Tac-Toe





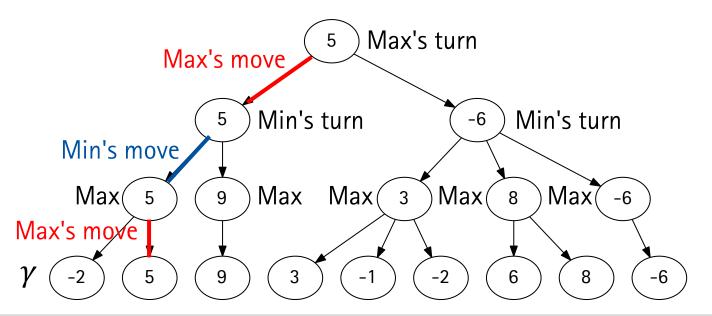


## Minimax Algorithm

- Let  $\gamma$ (leaf) be Max's payoff function (scores for leaves)
- The minimax value v(n) of a node n is defined as:

$$v(n) = \begin{cases} \gamma(n), & \text{if n is a leaf} \\ \max \{v(m) \mid m \text{ is child of n}\}, & \text{if n is an inner Max node} \\ \min \{v(m) \mid m \text{ is child of n}\}, & \text{if n is an inner Min node} \end{cases}$$

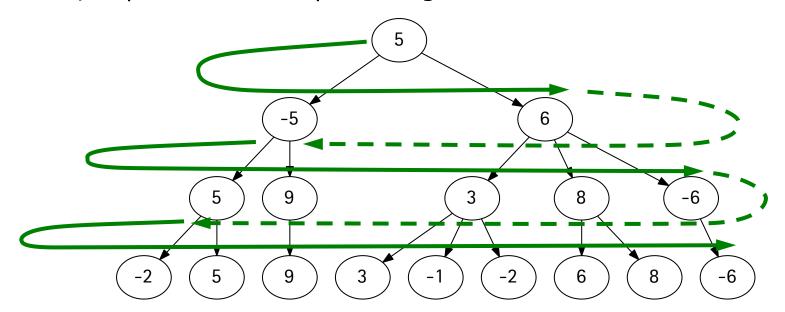
■ Recursive definition→ recursive algorithm





## **Building and Evaluating Game Trees**

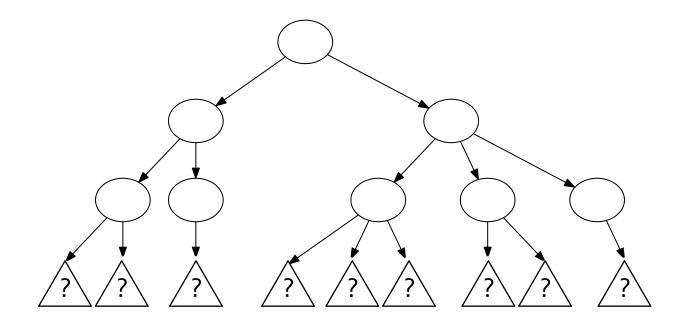
- Problem: Game trees are too large to build them completely
- Solution: Build them to level d, then evaluate
  - Maximum level d may depend on available time
  - Problem: Memory limitations for storing the tree levels
- Alternative: Only expand the most promising nodes





## **Building and Evaluating Game Trees**

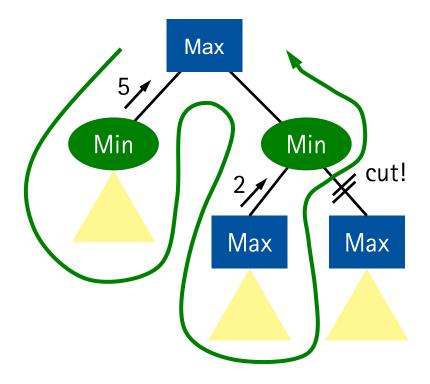
- Problem: Real scores are not known (if stopping at level d)
- Workaround: Estimate score from board state
- Estimation function: "Good" (close to real scores) and efficient





### **Tree Cuts**

- May cut branches if they are worse than what has already been found
  - Cut means: Do not explore further





## $\alpha$ - $\beta$ Pruning

- Expand nodes only if they can possibly change the result
  - Yields the same moves as the original Minimax algorithm
  - Does not explore irrelevant branches of the game tree
- $\alpha$ -bound: The minimum (worst) score that Max can get, given the current exploration state of the game tree
  - $\alpha$ -bound (initially  $-\infty$ ) increases during search of the game tree
- $\beta$ -bound: The maximum (best) score that Max can get, given the current exploration state of the game tree
  - $\beta$ -bound (initially  $\infty$ ) decreases during search of the game tree
- Perform cut if  $\alpha \ge \beta$

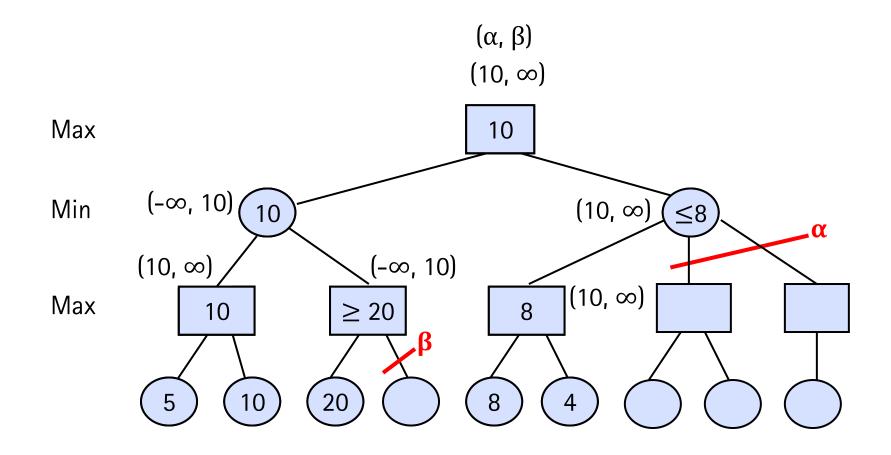


## $\alpha$ - $\beta$ Pruning Algorithm

```
int max_value(Node *g, int \alpha, int \beta) {
     if (limit reached(g)) return eval(g);
     for (Node *n = g->first; n != NULL; n = n->next) {
          \alpha = \max(\alpha, \min \text{ value}(n, \alpha, \beta));
          if (\alpha \ge \beta) return \beta; //\beta cut
     return α;
int min_value(Node *g, int \alpha, int \beta) {
     if (limit reached(g)) return eval(g);
     for (Node *n = g->first; n != NULL; n = n->next) {
          \beta = \min(\beta, \max_{\alpha} (\alpha, \beta));
          if (\beta \le \alpha) return \alpha; // \alpha cut
     return β;
```

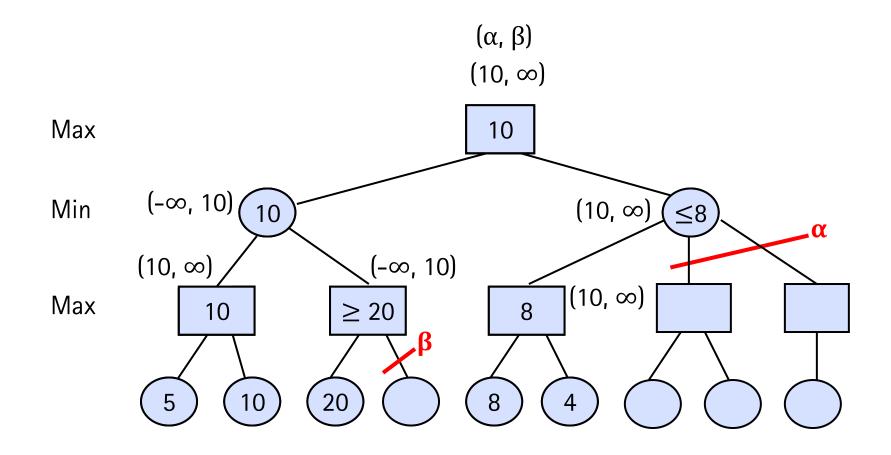


## $\alpha$ - $\beta$ Pruning Example





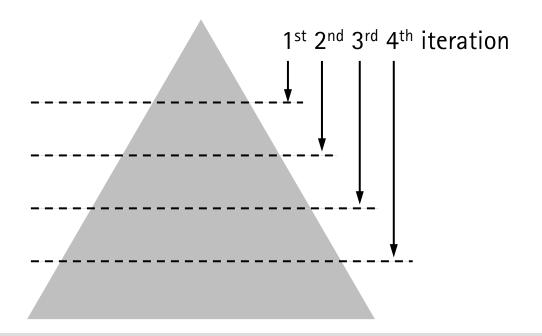
## $\alpha$ - $\beta$ Pruning Example





## **Iterative Deepening**

- Build and evaluate game tree to level d, evaluate
- If still sufficient time, build and evaluate to level d+1, evaluate
- Evaluation to level d is cheap compared to evaluation to level d+1
- Memorize information from earlier iterations
  - E.g., use results of earlier iterations for reordering





## **Other Aspects**

- Time management
  - Best use of the available time per move (and in total)
- Learning games
  - Can a game learn if it plays against itself?