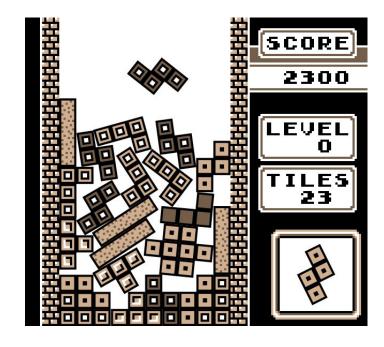
Count of Monte Carlo

Applying Monte Carlo Tree Search to Tetris



Samir Khays Kevin Holdcroft Yoshua Nava Viktor Tuul



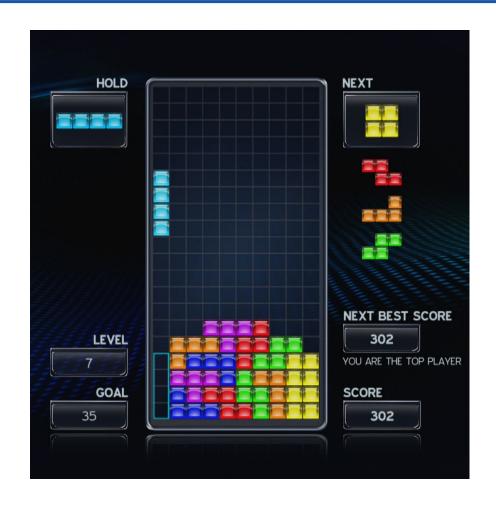
Contents

- Problem Introduction
- Methods
 - Search tree representation
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Tetris

- Classic video game
- Guide 7 types of pieces onto a grid
- Goal to fill rows completely
 - •Full rows disappear
- Know only current and next piece
 - Depending on variant
- Stochastic, non-deterministic





Work done

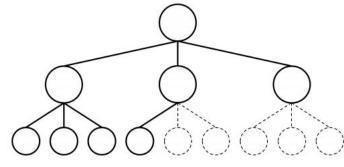
- •Implemented two different AI capable of playing Tetris
 - Monte Carlo Tree Search (MCTS)
 - Breadth-First Search (BFS)
- Originally implemented MCTS
 - Project goal
- •BFS used to tune heuristics
 - Ended up performing excellently



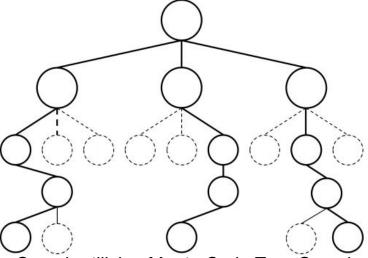
Search tree representation

- Search tree algorithm utilization
 - Represent the game in a search tree
 - Nodes being the current and possible future game states

- State transition function
 - Needed in order to produce a search tree
 - •Was not included in the open source game



Search utilizing Breadth First Search

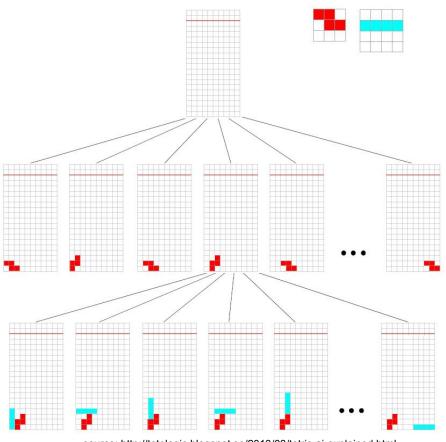


Search utilizing Monte Carlo Tree Search



State transition function

- State transition function
 - •Given any state (board + falling brick) → get every possible child state
 - •In practice: copies of the game and simulations for every brick orientation/place combination
 - •Assignment of randomly chosen bricks \rightarrow MCTS

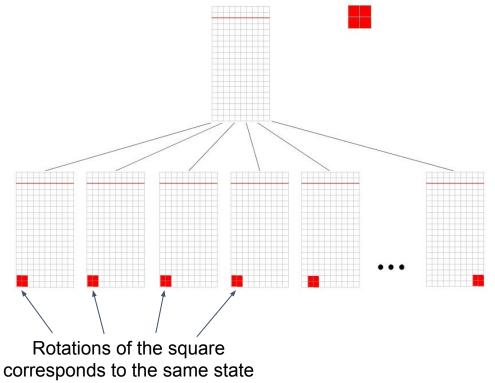


source: http://totologic.blogspot.se/2013/03/tetris-ai-explained.htm



Avoidance of duplicate states

- Local hash tables
 - Efficiency when expanding child nodes (avoidance of duplicate states)
 - Example: square block rotations





MCTS

MCTS

- Start -> Sampling -> Expansion -> Simulation -> Backprop -> Start
- Sampling criteria: UCB1.

For simulation:

- First, tried out random (light) playouts.
- Then heavy playouts with our BFS.
- Performance limitations for deeper search.
- Winning condition in Tetris? Design decision.

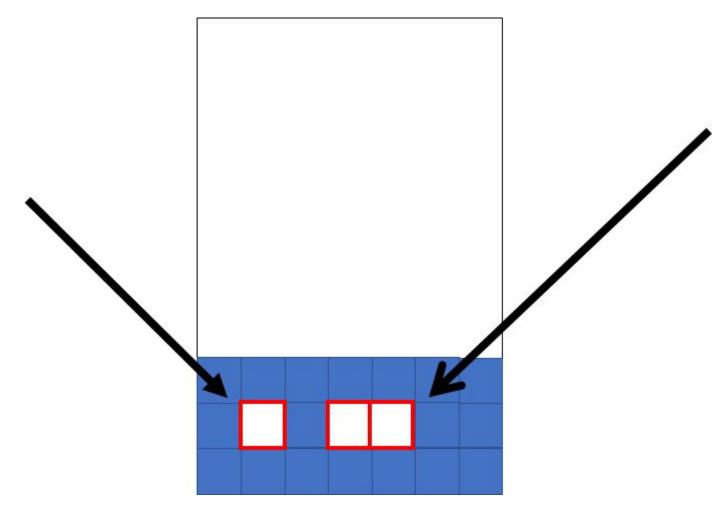


Heuristics

- Number of holes
- Aggregate height
- Cleared lines



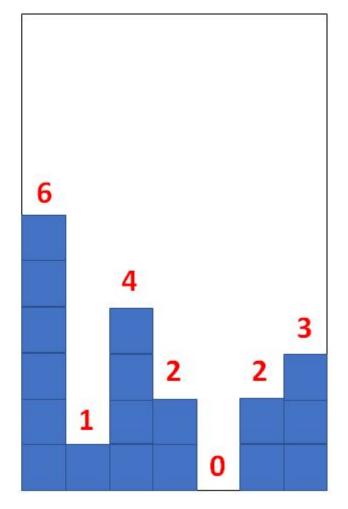
The Number of Holes heuristic





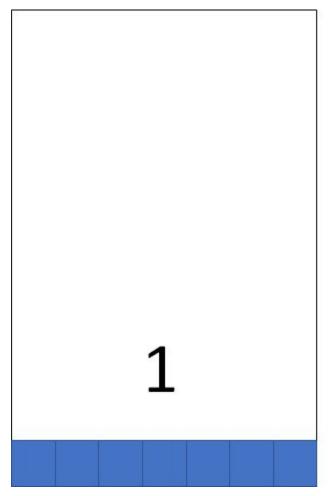
The Aggregate Height heuristic

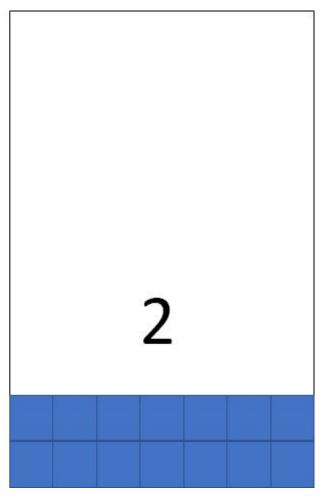
$$\sqrt{\sum_{i=0}^{9} (column_height_i)^2}$$





The Cleared Lines heuristic





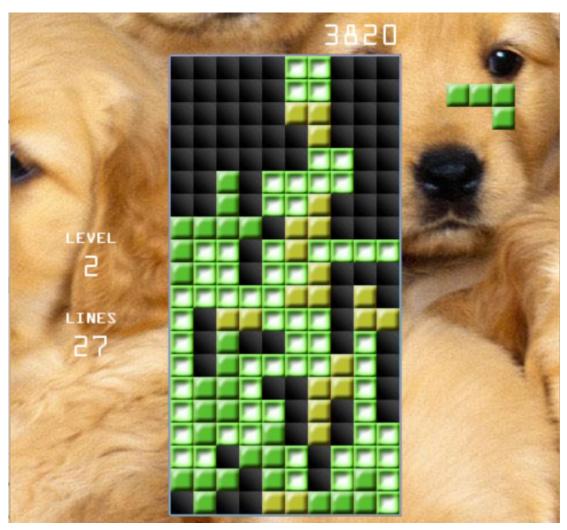


Results

Method	Lines Cleared	Score
BFS (depth 2)	93	11820
BFS (depth 1)	40	5770
MCTS	27	3820

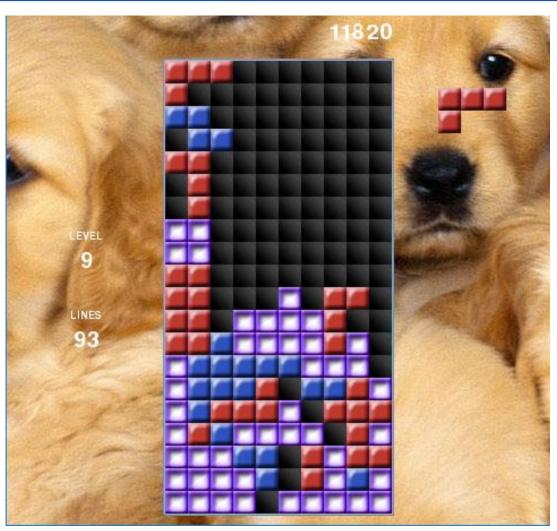


Results MCTS



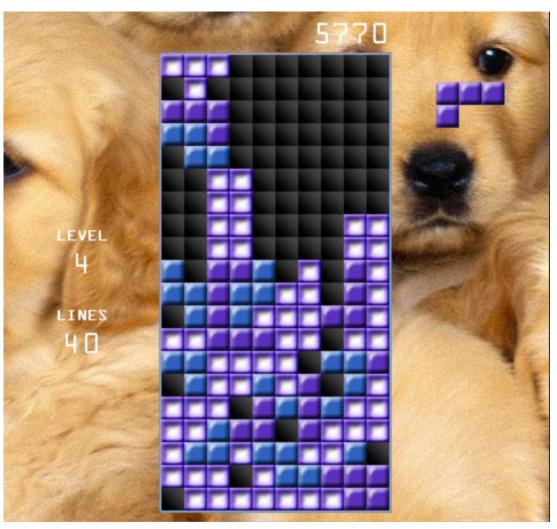


Results BFS (depth 2)





Results BFS (depth 1)

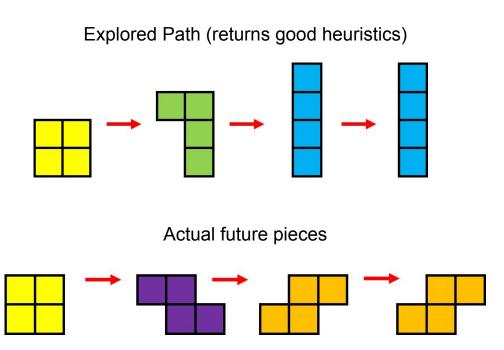


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Discussion

- Believe MCTS biased by having random paths with ideal piece drops
 - •Will move based on anticipation of future pieces
 - •e.g. two straight pieces in a row
 - Tetris is stochastic
 - •When pieces never come, past move is in a bad spot
 - "Counting chickens before they hatch"
 - Performance limitations





Discussion

- •BFS works since it tries to optimize the map based on current piece
 - Current heuristics try to make the best possible map
 - Only works with what it already has
 - Similar to how actual people play
- •MCTS robust to pieces, rather than to map state
 - Attempts to place piece in a way to optimize random child states
 - Theoretically valid, but biased toward piece choice



Future work

- Play in Real-time
- Could Improve MCTS
 - More Heuristics
 - Genetic Algorithms for Heuristics
 - •Including MCTS 'C' value
 - Translate to cpp instead, or use pipes
 - Python too slow
- •Bug in the later stages of the game
 - Causes pieces to stack in a column to one side



Final Comments

- •MCTS catered towards scenarios when there is a clear advantage towards searching deep
 - Situations with few actions
 - Scenarios where actions can be undone based on new information
 - •MCTS more suitable for games like Pac-Man
- •Thank-you to Professor Tumova and Professor Jensfelt for this class and for your insight
 - You TAs were great too



Questions?