Artificial Intelligence for deterministic 2 players games with UCT

Pierre Gueth

2 octobre 2013

UCT and other AI

Tree search algorithm

- Brute force approach
- Exhaustive search of every game possible
- High complexity
- Works only for small game (tic-tac-toe)

MinMax

- Standard AI for chess
- Tree search with limited depth
- Need score function
- Constant time consumption
- Doesn't work for open game like Go

UCT

- Collaboration between INRIA and Taiwan university [LEE2009]
- Used in MOGO, first Go AI to beat professional player in 2008
- Monte Carlo / Tree search hybrid
- No need for score function, only game completion
- · Flexible time consumption and high scalability

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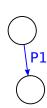
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- Deterministic game states
- No cyclic game states
- Each game should end by a win of either player or a draw

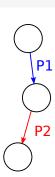


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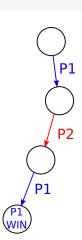


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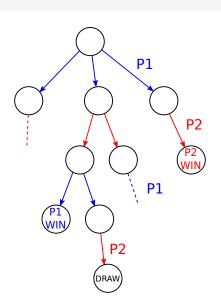
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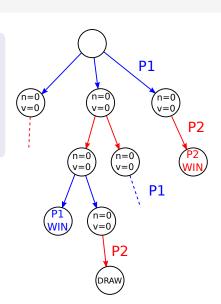
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- Play random moves for both players until game reaches its end
- 2 Update upstream game state value
- Select initials moves and go back to step 1

Each state has two variables:

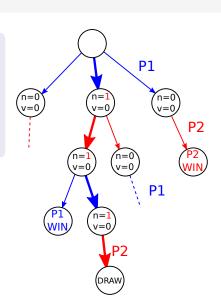
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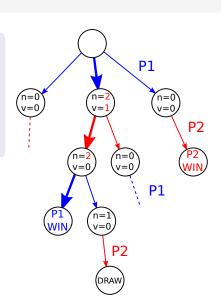
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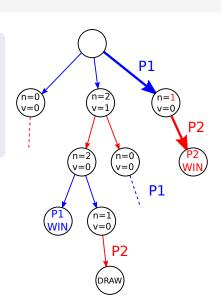
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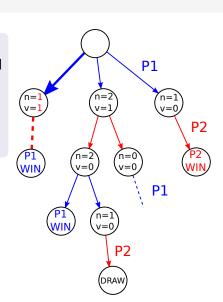
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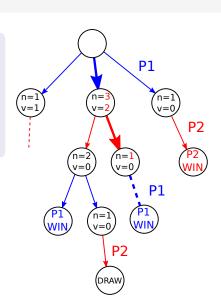
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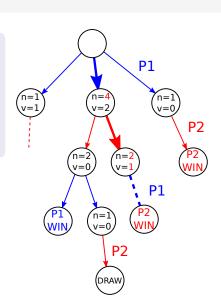
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Simulated games selection

Exploration

Each move has the same probability of being selected. This ensures that every strategies are considered.

Exploitation

The probability of selecting a move is proportional to the winning probability.

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Trade-off

- For small values of n, use the exploration strategy.
- For large values of n, use the exploitation strategy.

C++ implementation

- I implemented UCT in C++
- Use Qt, CMake but mainly based on standard library
- $\bullet \sim$ 3000 C++ lines

https://github.com/elcerdo/uct

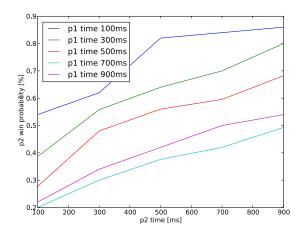
Easy implementation of new games

```
class Board {
public:
    virtual ~Board() =0;

    virtual Board *deepcopy() const =0;
    virtual Move *parse move_string(Token player,const char *string) const =0;
    virtual void print() const =0;
    virtual bool is_move_valid(const Move &move) const =0;
    virtual Moves get_possible_moves(Token player) const =0;
    virtual void play_move(const Move &move) =0;
    virtual void play_random_move(Token player) =0;
    virtual Token check_for_win() const =0;
    virtual Token play_random_game(Token next_player);
};
```

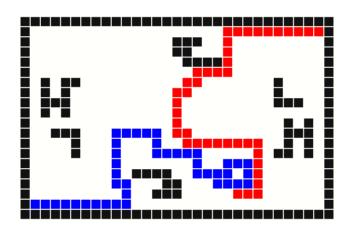
- Derive abstract board class
- Implement a few virtual methods
- That's it!
 UCT handles the hard part of choosing the best move for you.

Al strength vs. computation time: Othello



Al win probability increase linearly with computation time for computation time below 1s.

Google Al Challenge 2010 : Tron



UCT finished 50th over more than 3000 contestants.

Demonstration Can you beat the AI?

Thanks for your attention

Pierre Gueth

Wide technical and academical knowledge

- Classe préparatoire PT
- ENS Cachan aggrégation physique appliquée / EEA
- Thèse au laboratoire CREATIS (Université Lyon I) Imagerie médicale ultrasonore Estimation de mouvement
- Post doc au Centre Léon Bérard (Lyon)
 Simulation Monte-Carlo Protonthérapie (GATE)
 Imagerie γ-prompt

Numerous computer science side project

- Freesiege, Blocks, ...
- Monte-Carlo, UCT, ...
- Autojump, cluster submission tools