# An Investigation of AI Tree Search Methods and Their Effectiveness

# at Playing the Card Game Gwent

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# Objectives

- L. Simulate the Game of Gwent
- 2. Build Al Agents for Gwent
- 3. Empirically Compare Al agents
- 4. If Possible Optimize Agents
  Using Parallel Programming

### Motivation

### Artificial Intelligence (AI):

- Not all games are the same [1]
- Games can be **difficult** for Al to solve in reasonable time [1]
- The use of data structures and sampling techniques to model games.

### The Domain (Gwent):

- New domain
- Interesting rules
- Hidden information (unknown hands)[2]
- It's a fun game

Edition). Pearson, 2009.

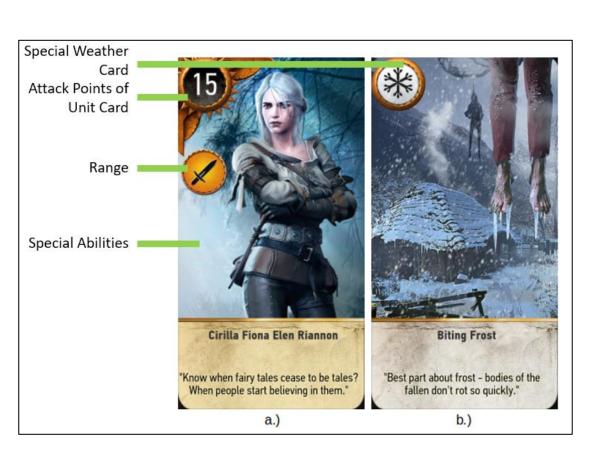


Example game board [2]

### Background

#### Gwent

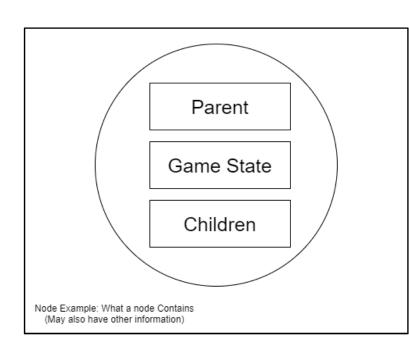
- Rules (Excluding deck building) [2]:
  - 2 Player turn-based
  - Best of **three** rounds
  - Players can pass a round
  - Only draw 10 random cards before 1<sup>st</sup> round
  - Units are placed in specific rows
  - Round wins calculated from total attack points

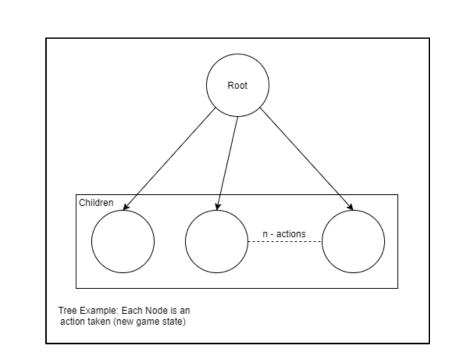


Special Cards	Abilities
Weather - <b>Reduce attack points</b> to 1 (for each card in a specific row)	Spy – add to opponent side but <b>draw</b> 3 <b>new cards</b> into hand
Commander's Horn – doubles attack points of row	Medic – <b>revive</b> discarded unit cards
Hero – a special unit which is not <b>affected</b> by other special cards	Tight bond – <b>multiplies</b> sibling card <b>attack points</b> by number of cards on board

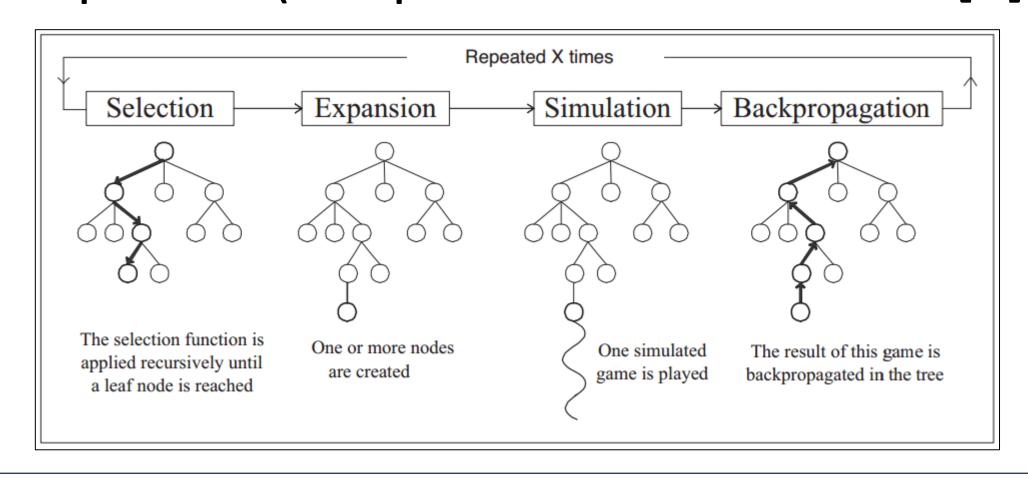
#### Monte-Carlo Tree Search (MCTS) [3]

Uses game-state trees





Has 4 phases (Adapted from references [3])



# Implementation

#### Random Agents:

- 3 Agents using pseudo-random number generators
  - Mersenne Twister
  - Marsaglia's Xorshift
  - Fast Rand

#### Heuristic Agent:

 Pick a card with the maximum reward from the player's hand using this linear combination:

```
reward = 0.2 * (AttackPoints - EnemyAttackPoints)
+ 0.2 * (HandSize - EnemyHandSize)
+0.6 * (RoundWins - EnemyRoundWins)
```

#### Random Rollout:

- Iterate through hand
- simulate n-number of games with each card
- Pick card with **most wins**
- Compute rollouts in parallel

#### MCTS Agent:

- Selection initially samples randomly based on hand
- **Expansion** based on selection
- Simulation uses random rollout
- Back-propagation reward:
  - Win -1.0
  - Draw = 0.5
  - Loss = -1.0

#### MCTS2 Agent:

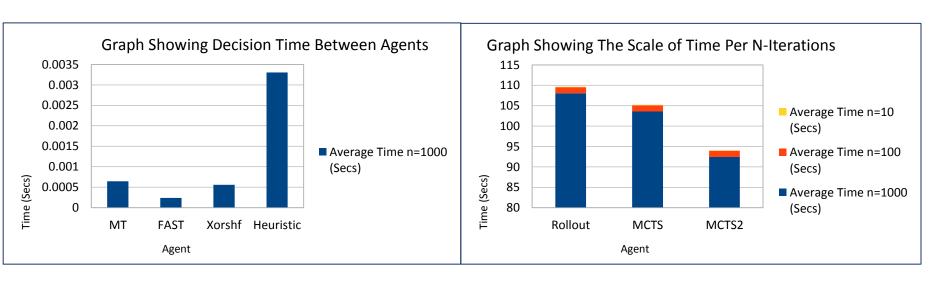
- Prunes away potential paths which are worse than the best
- Found during back-propagation

### Results

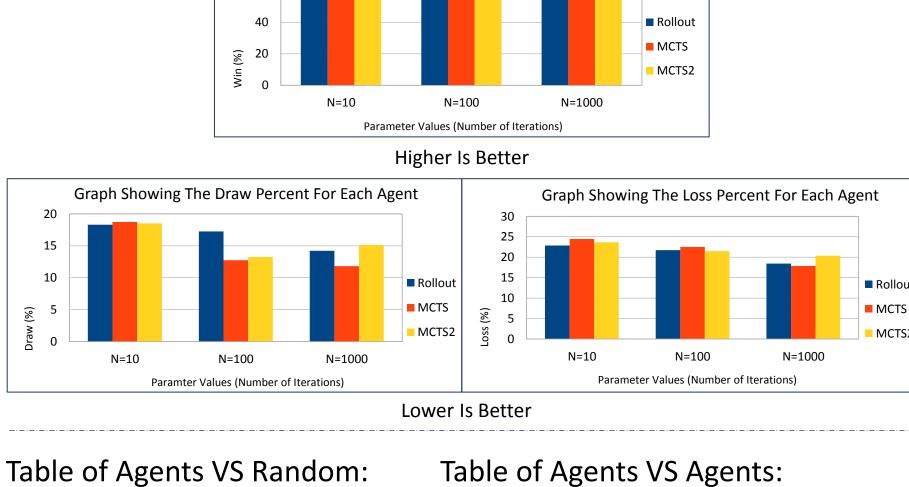
#### Method:

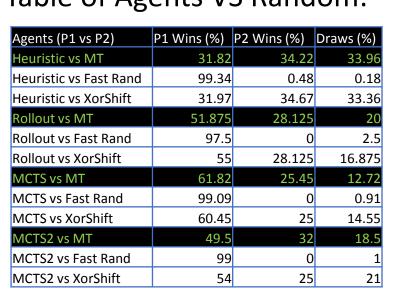
- Als are pitted against each other in a round-robin style tournament
- Each "game" is run **twice** with the players and decks swapped (for x runs)

#### Agent Times:



#### Agents VS Random:







## Future Work

- 1. Logic based agent
- 2. Hybrid MCTS (Mixture with Minimax algorithm)
- 3. More Complex Rules for Gwent
- 4. Machine learning and statistical sampling techniques



[2] CD PROJEKT S.A., CD PROJEKT RED. GWENT Guide. pages 1–4. Poland, 2015.
[3] Chaslot, Guillaume and Bakkes, Sander and Szita, Istvan and Spronck, Pieter.
Monte-Carlo Tree Search: A New Framework for Game AI. In Christian Darken and Michael Mateas, editors, AIIDE. The AAAI Press, 2008.