## COMP3130 – Group Project in Computer Science 10×10 Othello Learning Agent

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## 1. Abstract

An agent to play the board game *Othello* is created, with the ability to learn through reinforcement. The minimax algorithm is used for game playing, and a static evaluation function for the leaf nodes is learnt by self play. The agent learns the insignificance of the number of stones, and the significance of stone positioning. This agent is played against itself, and against other developed agents, and it's performance is analysed.

2. Problem overview

3. Solution overview

4. Optimisations

5. Static evaluation function

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## 6. Learning

## 6.1. TD-λ

The TD- $\lambda$  algorithm was used to learn the weighting of the static evaluation function's various features. The implementation of this algorithm is based on the one implemented by the Knightcap agent (http://citeseerx.ist.psu.edu/viewdoc/summary?doi=10.1.1.140.2003) At the end of each game played, the agent adjusts it's weights according to the following formula

$$w := w + \alpha \sum_{t=1}^{N-1} \Delta \tilde{J}(x_t, w) \times \left[ \sum_{j=t}^{N-1} \lambda^{j-t} d_t \right]$$
 
$$w \quad \text{The vector of weights} \quad x_t \quad \text{The } t^{th} \text{ board in the game}$$
 
$$\alpha \quad \text{The learning rate} \quad \lambda \quad \text{The discount factor}$$
 
$$N \quad \text{The number of states in the game}$$
 
$$\Delta \tilde{J} \quad \text{The derivative of the } \tilde{J} \text{ function}$$

In the above formula, the  $\tilde{J}$  function estimates the probability of winning from a given state, given a set of weights for features and our board.

- 6.2. ELO arena
- 6.3. Comparison of TD- $\lambda$  and ELO arena
- 7. Performance evaluation
- 8. Improvements