## E-HBA:

Using Action Policies for Expert Advice and Agent Typification

# **Appendix**

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This document is an appendix to [1]. It contains a listing of all parameter settings used in the study as well as plots for each experiment.

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### 1 Parameter Settings

#### 1.1 E-HBA

- 5 unique experts (i.e.  $|\Phi_1| = 5$ )
- 5 unique types (i.e.  $|\Theta_2^*| = 5$ )
- Depth of planning horizon: h = 5

### 1.2 Leader-Follower-Trigger Agents

- Maximum number of joint actions in target solutions: 2
- Target solution admissible if average payoff ≥ maximin value (for each player)

#### 1.3 Co-Evolved Decision Trees / Neural Networks

- Number of populations: 2 (one for each player)
- Individuals per population: 50 (first population randomly generated)
- Fitness = average payoff after 20 rounds ( $\in [1,4]$ ) average similarity ( $\in [0,1]$ )
- Each individual evaluated against random 40% of other population
- Resampling method: linear ranking
- Decision Trees:
  - Tree depth: 3 (up to three previous actions of other player)
  - Similarity: percentage of nodes with same action choice
  - Evolutions: 300 (evolution with highest average fitness used)
  - Random mutation of single node (flipping action): 5% of population
  - Random crossing of sub-trees (preserving tree depth): 30% of population

#### • Neural Networks:

- Input layer: 4 nodes (up to two previous joint actions)
- Hidden layer: 5 nodes
- Output layer: 1 node (probability of action 1)
- Each node fully-connected with nodes of next layer
- Standard sigmoidal threshold function
- Similarity: 1 average difference of output for each input
- Evolutions: 1000 (evolution with highest average fitness used)
- Random mutation of single input weight (standard normal shift): 20% of population
- Random crossing of nodes (preserving network structure): 10\% of population

## 2 Plots of Results

### 2.1 Legend



Figure 1: **Legend for bar plots.** The bar plots show the average payoffs of player 1 (E-HBA) over 5,000 rounds. The minimum and maximum achievable payoffs per round were 0 and 1, respectively.

#### 2.2 Results

We use the following abbreviations:

- LFT Leader-Follower-Trigger Agents
- CDT Co-Evolved Decision Trees
- CNN Co-Evolved Neural Networks

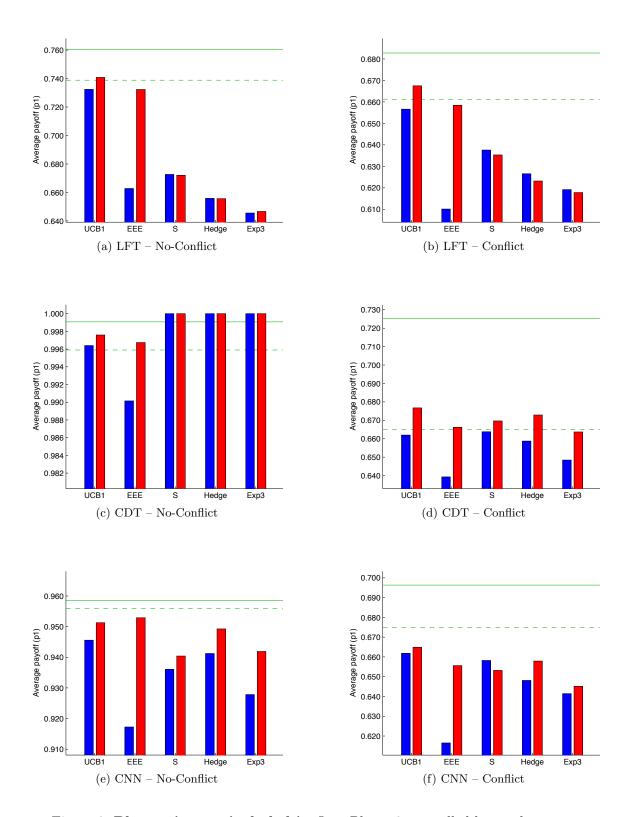


Figure 2: Player 2's type included in  $\Theta_{2}*$ . Player 2 controlled by random type.

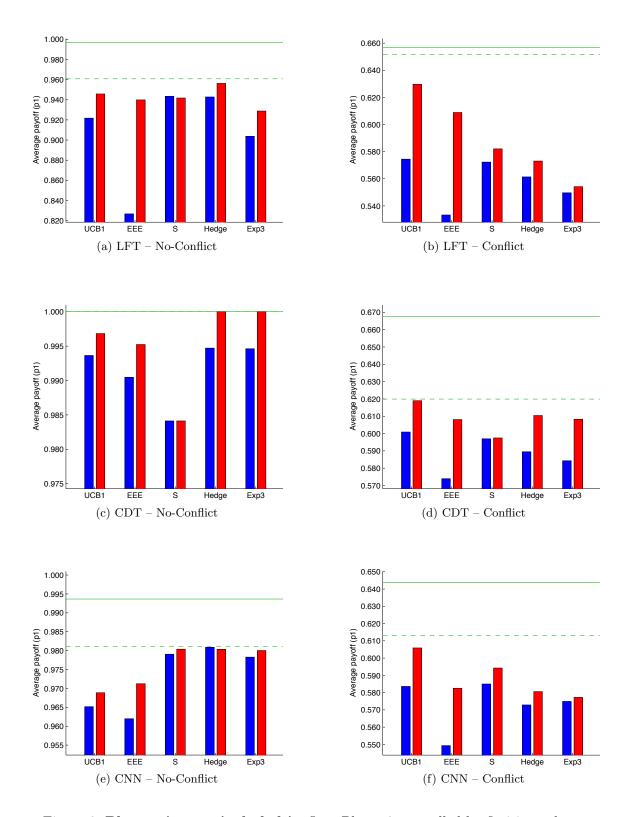


Figure 3: Player 2's type included in  $\Theta_2*$ . Player 2 controlled by fictitious player.

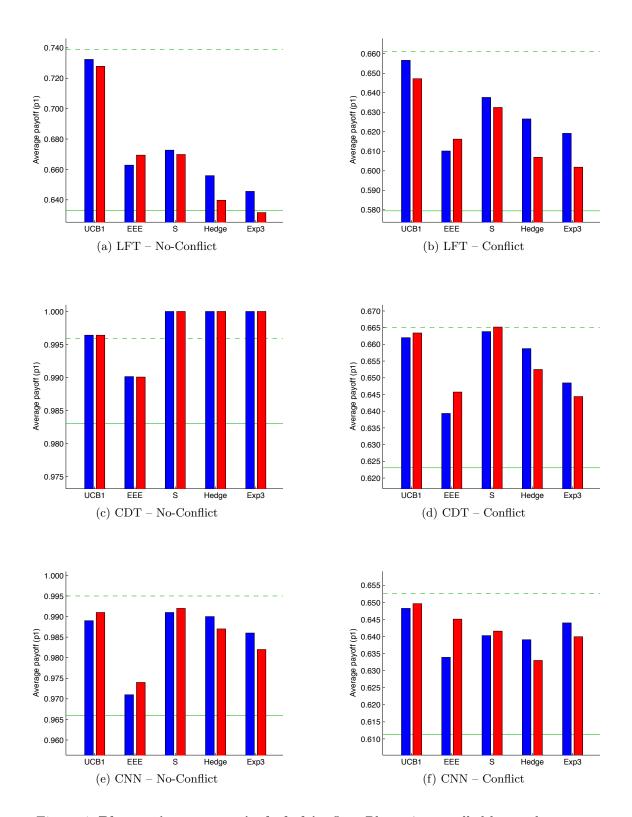


Figure 4: Player 2's type not included in  $\Theta_{2}*$ . Player 2 controlled by random type.

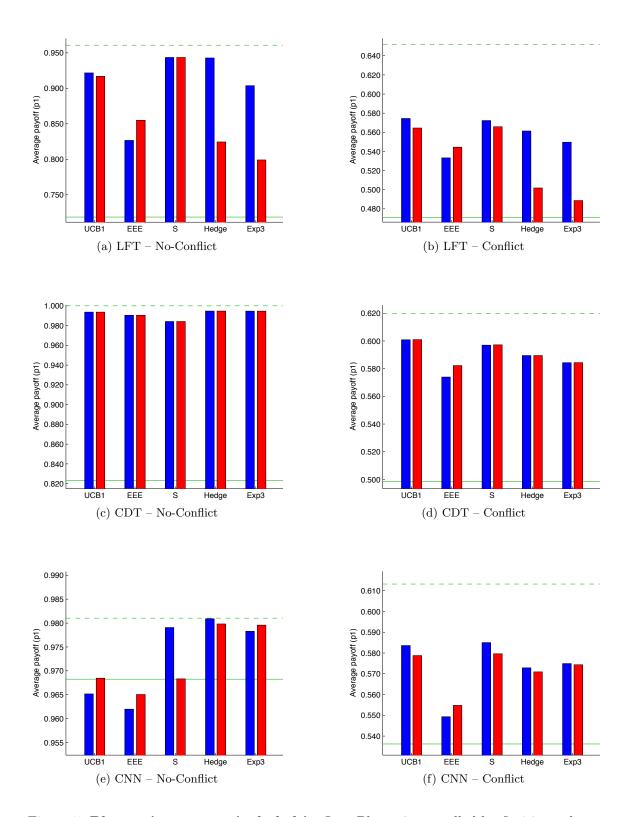


Figure 5: Player 2's type not included in  $\Theta_{2}*$ . Player 2 controlled by fictitious player.

## References

[1] S. V. Albrecht, J. W. Crandall, and S. Ramamoorthy. E-HBA: Using action policies for expert advice and agent typification. In *Proceedings of the Second AAAI-Workshop on Multiagent Interaction without Prior Coordination*, Austin, Texas, USA, January 2015.