

Homework 5: Paper 2 Review

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1 SUMMARY

1.1 What is the paper about?

The paper addresses the credit assignment problem in cooperative coevolutionary algorithms, particularly when fitness depends heavily on joint actions among agents. It introduces the Global Alignment Local Error (GALE), a loss function that generates better local evaluations by aligning them with the global evaluations, successfully improving cooperative behavior and performance in multiagent systems.

1.2 Main contributions

The contributions of this paper are development of the GALE loss function to provide aligned local evaluations and implementation of the loss function in cooperative coevolution that improves the quality of credit assignment, leading to a significant performance boost in a simulated multiagent systems.

1.3 The paper's strengths

In the paper, the authors address credit alignment issues in cooperative coevolutionary algorithms by pointing out the weaknesses of the previous research, which did not look into the local evaluations. The approach of using Global Alignment Local Error (GALE) for training local evaluations aligns well with the need for better credit assignment strategies. The authors provided reliable data set in the experiment parameters section and mostly showed the results of using GALE and other approaches. They also offered the future research directions that the GALE method did not consider the parts of network extensions and introducing a penalty for discrete local evaluations.

1.4 The paper's weaknesses

The contents of the paper focused heavily on providing the general knowledge such as the meaning of the equations and better credit assignment repeatedly. If the authors elaborated the specific steps why we need to pursuit better credit alignment, it would be better paper. Lastly, some results of the experiment cannot support the claims. In Figure 2., the performance of GALE is not that much superior to other approaches such as FitCritic + MAE and MSE since there are some fluctuations implying that it is hard to determine which one is best, leading to overinterpretation of results.

1.5 What are the conclusions?

The paper concludes that the GALE method notably improves credit assignment in cooperative co-evolutionary environments. This results in better teaming behaviors and a significant improvement in performance compared to previous methods. The authors suggest that future work could explore incorporating sequential modeling mechanisms and improving learnability of the local evaluations using alignment.

2 GENERAL RATING: WEAK

2.1 Presentation, Readability, and Organization

The order of contents are well-organized (introduction, background, method, experiment, results, and conclusion), developing why their method has higher performance than other methods. However, the authors focus heavily on pointing out the weakness of previous research and improving existing methods, not telling the story from high-level perspective. Their methodology itself looks appropriate and nicely handles the problem but they did not mention what the essence of the problem is and how to reach the tentative conclusion to illustrate their work. The paper is easy to read, but I'm not sure that the readers who have no background knowledge, and it is mostly free of grammatical errors and typos.

2.2 Problem definition

In the paper, the authors addresses cooperative coevolutionary algorithms that have limitations of credit assignment and degradation of local approximation using fitness critics. This paper provides the algorithm to overcome these limitations, and it is a little vague to illustrate why it is problem to solve.

2.3 Originality

The work is original since it proposes a novel loss function to align local and global evaluations. Previous researches in this field have weakness of handling the alignment of both local and global evaluations. In the multi domain, it is necessary to improve teaming behavior for global objective and agent's local goals.

2.4 Significance and Usefulness

The problem addressed is crucial for the advancement of cooperative algorithms in complex multiagent environments. Since the agents behavior can affect each other and the overall goal, the balanced or optimal solution is needed to fulfill the better teaming. There are some attempts to optimize agents' policies but these focus heavily on team performance rather than individual performance, meaning that for the team performance, some agents' less effective policies are easily discarded.

2.5 Technical Soundness

The GALE methodology is sound and innovative for addressing the credit assignment problem in cooperative coevolution. However, some results of the experiment cannot support the claims. In Figure 2., the performance of GALE is not that much superior to other approaches such as FitCritic + MAE and MSE since there are some fluctuations implying that it is hard to determine which

one is best. Moreover, in the Figure 6., the authors mentioned the cleanest delineation between the value of the POI the agent needs to move, but the explanation is not enough to catch the meaning of the results.

2.6 Analysis, Impact, and Conclusion

The authors provide analysis of their results, incorporating statistical measures to verify their findings. The use of performance plots allows for a detailed assessment of GALE. They provide partially the implications of the results, focusing on its enhancement of agent collaboration in complex environments. Then, they highlight the impact of using GALE that can learn a quality local approximation of the global evaluation. After that, the authors pointed out that the proposed GALE ignores the intermediate information provided by the states leading up to the final state, and they suggested that incorporating sequential modeling mechanisms can be considered to assign credit in even more complex domains. Furthermore, development of learnability of the local evaluations using alignment can also be potential research direction.

2.7 Background and Reference

The paper provides the proper background knowledge to understand the flow of previous works. There is no related work section, but, in the background section, the authors cites some researches related to the subtopic. The references are useful to understand previous attempts to handle the given topic, and there are 35 resources to make the authors' work reliable.