Review: Accelerating Multi-agent Reinforcement Learning with Dynamic Co-learning

Steve Homer[1], Fabian Perez[1], Quinten Rosseel[1] and Matthias Humt[1]

[1]{steven.homer, fabian.perez, quinten.rosseel, matthias.humt}@vub.be

Abstract

The article being reviewed introduces a

Introduction

In the introduction a clear specification of the problem is provided, including related work. Look for additional but relevant references which you can cite in this part.

Hello, here is some text without a meaning. This text should show what a printed text will look like at this place. If you read this text, you will get no information. Really? Is there no information? Is there a difference between this text and some nonsense like "Huardest gefburn"? Kjift – not at all! A blind text like this gives you information about the selected font, how the letters are written and an impression of the look. This text should contain all letters of the alphabet and it should be written in of the original language. There is no need for special content, but the length of words should match the language.

Methods

In the Methods section you specify how the model is constructed.

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Results and Discussion

In the Results section you provide a descriptions of the simulations you have done and their results. Also include the information concerning the parameter settings. In the Discussion section you provide a summary and an explanation of your work.

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Conclusion

Related work. All articles cited in the original paper.

(Carroll and Seppi, 2005), (Garant et al., 2015), (Ghavamzadeh et al., 2006), (Gmytrasiewicz and Doshi, 2005), (Guestrin et al., 2002), (Kitano et al., 1999), (Lazaric et al., 2008), (Littman, 2001), (Nair et al., 2005), (Nedic and Ozdaglar, 2009), (Oliehoek et al., 2008), (Price and Boutilier, 2003), (Rényi et al., 1961), (Taylor and Stone, 2009), (Vickrey and Koller, 2002), (Witwicki and Durfee, 2010), (Zhang et al., 2010), (Zhang and Lesser, 2013)

Questions to be answered by reviewing party:

1. Does the introduction explain clearly the content of the paper

2. whether there is sufficient background information to understand the relevance of the work

3. whether the methods are clearly explained (can the results be reproduced?)

4. whether the results answer the questions asked in the paper.

5. whether all questions are answered

6. whether the conclusion is sufficient

7. and whether the overall style is ok and

8. whether you believe things are missing in the discussion.

9. etc.

10. 3 positive points concerning the work, clearly specifying why you think they are well- done or interesting

11. 3 negative points, which may include missing/unclear explanations or suggestions for improvement

12. at least 3 clear and relevant questions on the content or the methods used which can be asked (next to other questions).

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