Noname manuscript No.

(will be inserted by the editor)

Natural Actor Critic Do you have a subtitle? If so, write it here

First Author \cdot Second Author

Received: date / Accepted: date

F. Author first address

 $\begin{tabular}{l} Tel.: $+123$-45-$678910 \\ Fax: $+123$-45-$678910 \\ E-mail: fauthor@example.com \end{tabular}$

S. Author second address

1 Paper

Natural Actor Critic:

- Main Version [6].
- 2nd Version: Natural Actor-Critic in Neurocomputing [4].
- 3rd version: RL of motor skills with policy gradients in NN [5].

Recommended by Jan:

- Policy Evaluation with TD [3].
- Incremental NAC algorithms [1].
- Jan said that a paper form C. Dann is very important. Did he mean Policy Evaluation with TD by Dann or did he mean a second paper?

Research:

- Comparison of four natural gradient algorithms (co-author Sutton) [2].

2 Meetings & Notes

Meetings:

- 12.12.18: Notes from Jan can be found in ".\Notes Jan 12.12.18"

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Table 1 Please write your table caption here

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Abstract Insert your abstract here. Include keywords, PACS and mathematical subject classification numbers as needed.

Keywords First keyword \cdot Second keyword \cdot More

3 Introduction

- Steepest ascent direction of performance object with respect to any metric $M(\theta)$: $M(\theta)^{-1}\nabla_{\theta}J(\mu_{\theta})$
- The natural gradient ist the steepest ascent direction with respect to the Fisher information metric $M_{\pi}(\theta) = E_{s \sim \rho^{\pi}, a \sim \pi_{\theta}} [\nabla_{\theta} log \pi_{\theta}(a|s)^{T} \nabla_{\theta} log \pi_{\theta}(a|s)]$
- For deterministic policies: $M_{\mu}(\theta) = E_{s \sim \rho^{\mu}} [\nabla_{\theta} \mu_{\theta}(s) \nabla_{\theta} \mu_{\theta}(s)^{T} w]$
 - Limiting case of the Fisher information metric: policy variance reduced to zero
- Combining DPG theorem with compatible function approximation gives $\nabla_{\theta} J(\mu_{\theta}) = E_{s \sim \rho^{\mu}} [\nabla_{\theta} \mu_{\theta}(s) \nabla_{\theta} \mu_{\theta}(s)^{T} w]$ so steepest ascent direction reduces to $M_{\mu}(\theta)^{-1} \nabla_{\theta} J_{\beta}(\mu_{\theta}) = w$

Your text comes here. Separate text sections with

4 Section title

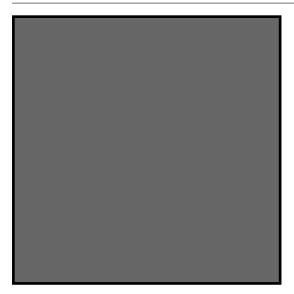
Text with citations [?] and [?].

4.1 Subsection title

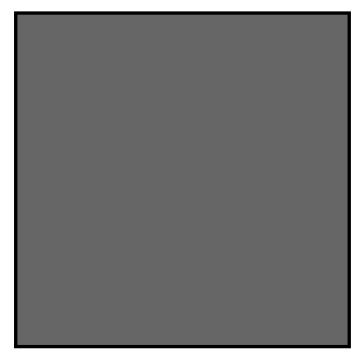
as required. Don't forget to give each section and subsection a unique label (see Sect. 4).

Paragraph headings Use paragraph headings as needed.

$$a^2 + b^2 = c^2 (1)$$



 ${\bf Fig.~1}~{\rm Please~write~your~figure~caption~here}$



 ${\bf Fig.~2}~{\rm Please~write~your~figure~caption~here}$

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References

1. Bhatnagar S, Ghavamzadeh M, Lee M, Sutton RS (2008) Incremental natural actor-critic algorithms. In: Advances in neural information processing systems, pp 105-112

- 2. Bhatnagar S, Sutton RS, Ghavamzadeh M, Lee M (2009) Natural actorcritic algorithms. Automatica 45(11):2471-2482
- 3. Dann C, Neumann G, Peters J (2014) Policy evaluation with temporal differences: A survey and comparison. The Journal of Machine Learning Research 15(1):809–883
- 4. Peters J, Schaal S (2008) Natural actor-critic. Neurocomputing 71(7-9):1180–1190
- 5. Peters J, Schaal S (2008) Reinforcement learning of motor skills with policy gradients. Neural networks 21(4):682-697
- 6. Peters J, Vijayakumar S, Schaal S (2005) Natural actor-critic. In: European Conference on Machine Learning, Springer, pp 280–291