

Reproducibility Checklist

[Based on Joelle Pineau's ML Reproducibility checklist]

Descriptions (in reports, theses & papers)
 A clear description of the mathematical setting, algorithm, and/or model. A clear explanation of any assumptions. (If appropriate) An analysis of the complexity (time, space, sample size) of any algorithm.
Code Quality
 Make training code available Make evaluation code available Well documented (e.g., DocStrings) and readable code Unit test your code README.md with precise instructions (commands) for installation and running the code Dependencies and requirements (requirements.txt)
Experimental Reproducibility and Generalization
 Several training repetitions with different random seeds [block seeds across settings be compared to one another] [RL] Seeding of environments to control non-determinism of environments (at least 1 fixed seed in training and several for evaluation) [RL] Several evaluation runs on the same environment [RL] Evaluation on several environments or variations of the same environment Run and report ablation studies to check the impact of different design decisions Optimally, use the same amount (and technique) of hyperparameter optimization for all competitors
Reporting
 All details regarding the experimental setting, incl. software versions and [RL] which env inc reward function, gamma etc pp. A description of results with central tendency (e.g. mean) & variation (e.g. error bars). (If possible and appropriate) Statistical hypothesis tests to show a significant difference in performance (beware of significant vs. substantial) – report alpha, type of test and test assumptions
 ☐ The exact number of training and evaluation runs ☐ Used random seeds (training, agents, environment) ☐ All hyperparameters decisions (settings, ranges, optimization technique and resources) ☐ Compute infrastructure (CPUs, GPUs, TPUs, RAM, OS)
The average runtime for each result, and (if possible) estimated energy cost.