Advanced Topics in Reinforcement Learning Lecture 22: Evaluation and Reproducibility

Announcements

- Next week: Multi-agent RL
- No class Thursday!



Variation in RL





Variation in RL

- High stochasticity in deep RL algorithms:
 - MDP is stochastic.
 - Exploration policy is usually stochastic.
 - Mini-batch gradient descent with a replay buffer is stochastic.
 - Network weight initialization is stochastic.
 - GPUs have some non-determinism.

• Any single run of an RL algorithm may mislead you about expected performance.



Experimental Protocol

- For each algorithm considered, run multiple trials.
- Report aggregate statistics for each algorithm such as mean, median, and mode.
- Report confidence measures such as standard error, %-confidence interval, or interguartile range.
- How many trials?
 - Depends...
 - number of trials.
 - Otherwise, use power analysis to decide if sample size is sufficient. \bullet

• If computational requirements of experiments are light, can just re-run with a larger



Common Random Seeds

- You plan to compare algorithms A and B in simulated environment X.
- Since data in RL is stochastic, what if algorithm A gets unlucky with observed data?
- To prevent this, fix a set of common random seeds. Ensure that each algorithm is ran once with each seed.

```
1 import random
3 seeds = [random.randint(le6) for _ in range(n_trials)]
5 for seed in seeds:
      # Run algorithm A with seed
       . . .
      # Run algorithm B with seed
9
10
       . . .
```





Yunfu's Presentation

• <u>Slides</u>



Hyper-parameters Matter





Hyper-parameter Selection

- Report how hyper-parameters selected. Why?
 - more important than the final hyper-parameter values.
- How to select:
 - Graduate student descent
 - Random search
 - Grid search
 - neural architecture search.

Best hyper-parameters are problem dependent and so the process is

More advanced: Bayesian optimization, population-based training,



Community Standards

- - Reproducibility challenges
 - Reproducibility check-lists
- Culture challenges:
 - Emphasis on positive results and novelty.
 - \bullet questions.

What should conference and journal standards for reproducibility be?

Low benefit for reproducing work of others and considering incremental



Sharing Code

- Is code sharing sufficient?
 - Not necessarily, may require missing compute and data resources.
- Why do different code-bases give different performances?
 - Deep RL implementations often have hidden tricks that go beyond the base algorithm.
 - Example: observation normalization, i.e., divide observed state variables by a running average of their standard deviation.



Summary

- algorithms a challenge.
- these values is necessary for replicability.

Deep RL experiments are very stochastic. Makes evaluation of deep RL

• Deep RL algorithms can be sensitive to their hyper-parameters. Reporting

Community standards and values matter for what type of science is done.



Action Items

- Multi-agent RL reading for next week.
- Continue making progress on your final project.

