

CS 312: Artificial Intelligence Laboratory

Task 8: Reinforcement Learning

Goal: Formulate and solve an MDP problem using policy and value iteration.

Note: This assignment is to be done individually.

- (1) Come up with a sequential decision-making problem and formulate the problem as either a discounted reward or total reward MDP problem.
- (2) Develop code for solving the MDP problem using policy and value iteration.
- (3) Write a report clearly describing the MDP considered and your observations on running the policy and value iteration algorithms on the formulated MDP.
- (4) Further, one should also suggest ways to check whether the algorithm yields optimal policy for the setting considered.

Note:

- (1) The MDPs considered should not be the same across students. There should be a significant difference in the MDP considered.
- (2) Do not take MDPs that have a large number of states/actions. It is not possible to solve them then.
- (3) In order to meet Instruction 1 above, each student should decide the MDP they want to consider as early as possible and fill in this [sheet](#).
- (4) If someone has already taken that MDP problem you are supposed to decide on a new MDP problem.
- (5) Hence, students quickly deciding on their MDP problem will have less chance of changing the MDP problem again.

You should take the consent of the TAs by [filling this sheet](#) to finalize on the MDP problem at the earliest.

Evaluation Criteria:

MDP formulation: 10

Correctness: 20 (10 + 10 for PI and VI)

Report: 15

Code Quality: 5

Deadline: 11:59 PM 6 April 2020

Late Submission Policy: 5% of marks will be deducted per day late.

For Reference :

Reinforcement Learning:

<http://www.cse.iitm.ac.in/~ravi/courses/Reinforcement%20Learning.html>

See lectures 15 -25.

Report Format :

1. [1 mark] MDP Description: Clearly describe (S, A, P, R, N)
2. [5 marks] State-transition Graph for the MDP
3. [1 mark] Optimal Policy: Suggest ways to check whether the algorithm yields optimal policy for the setting considered.
4. [5 marks] Experimental Results: Vary the gamma parameter, show the policy found in each case by both algorithms
5. [2 marks] Comparison of Policy Iteration and Value Iteration
6. [1 marks] Conclusions