CS747 - Assignment 2 Report

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Directory Structure

- encoder.py File to encode maze as transition probabilities and rewards
- decoder.py Decode the policy into directions
- planner.py Main script to call to different algorithms
- vi.py Code to implement value-iteration algorithm
- hpi.py Implementation of Howard's policy iteration algorithm
- *lp.py* Linear programming implementation
- helper.py A file for additional tasks like reading and parsing instances

Implementation Details

Array Dimensions

Transitions(T) and rewards(R) are number arrays of dimensions (S, A, S), Q is defined to be an array of the dimensions (S, A) and V is of dimensions (1, 1, S) where S is the number of states and A is the number of actions

Value Iteration

While $||V - V_{prev}||_{\infty} > \epsilon$ where $\epsilon = 10^{-10}$, update:

$$V_{prev} = V \tag{1}$$

$$V = \max_{s \in A} \sum_{s' \in S} T * (R + \gamma * V_{prev})$$
 (2)

Howard's Policy Iteration

While $P! = P_{prev}$ where P = policy, update:

$$P_{prev} = P \tag{3}$$

$$P = Policy(V_P) \tag{4}$$

Initialize the policy to take action 0 at all the states.

Compute the value function using the policy.

Update the policy as per the computed value function.

Helper

Assumption: The MDP files have multiple space in the discount line. The helper file, however, requires it to be a single space.

File to parse MDP file to generate transition and reward arrays, generating transitions using grid files and reading value-policy file.

Conclusions

Performance comparision:

Value iteration > Howards policy iteration > Linear programming

References

- [1] Python documentation available at https://docs.python.org/3/
- [2] Numpy documentation available at https://numpy.org/doc/
- [3] PuLP documentation available at https://coin-or.github.io/pulp/