## Assignment #3

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## Problem 1: Q-Learning with Continuous State

Consider a system with a single continuous state variable x and actions  $a_1$  and  $a_2$ . An agent can observe the value of the state variable as well as the reward in the observed state. Assume a discount factor  $\gamma = 0.9$ .

- (a) Assume that function approximation is used with  $Q(x, a_1) = w_{0,1} + w_{1,1}x + w_{2,1}x^2$  and  $Q(x, a_2) = w_{0,2} + w_{1,2}x + w_{2,2}x^2$ . Give the Q-learning update equations.  $Q(x, a) \leftarrow Q(x, a) + \alpha[R(x) + 0.9 \max_{a'} Q(x', a') Q(x, a)],$  where  $a' \in a_1, a_2, \alpha$  indicates learning rate and R is the reward.
- (b) Assume that  $w_{i,j} = 1$  for all i, j. The following transition is observed: x = 0.5, observed reward r = 10, action  $a_1$ , next state x = 1. What are the updated values of the parameters assuming a learning rate of 0.5?

$$\begin{split} w_{i,1} \leftarrow w_{i,1} + \alpha [r + 0.9 \max_{a' \in a_1, a_2} Q(x', a') - Q(x, a_1)] \frac{\partial Q(x, a_1)}{\partial w_{i,1}}, \text{ where } i = 0, 1, 2; \\ w_{i,1} \leftarrow 1 + 0.5 [10 + 0.9(1 + 1 + 1) - (1 + 0.5 + 0.25)] \frac{\partial w_{0,1} + 0.5 w_{1,1} + 0.25 w_{2,1}}{\partial w_{i,1}}, \text{ where } i = 0, 1, 2; \\ w_{0,1} \leftarrow 1 + 5.475 * 1 = 6.475; \\ w_{1,1} \leftarrow 1 + 5.475 * 0.5 = 3.7375; \\ w_{2,1} \leftarrow 1 + 5.475 * 0.25 = 2.36875; \end{split}$$

## Problem 2: Policy Gradient with Continuous State

Assume that Q-function with function approximation is used together with the softmax function to form a policy  $\pi_{\theta}(s, a) = e^{Q_{\theta}(s, a)} / \sum_{a'} e^{Q_{\theta}(s, a')}$ . Assume that there are two actions with  $Q(x, a_1) = w_{0,1} + w_{1,1}x + w_{2,1}x^2$  and  $Q(x, a_2) = w_{0,2} + w_{1,2}x + w_{2,2}x^2$  for a real valued variable x.

(a) Give the update equations for the *REINFORCE* algorithm. Assume that the treturn at the current step is G and the action taken is  $a_1$ .

$$w_{i,j} \leftarrow w_{i,j} + \alpha * G\nabla_{w_{i,j}} \ln \pi_{w_{i,j}}(x, a_1),$$
  
where  $i = 0, 1, 2, j = 1, 2$ , and  $\alpha$  indicates learning rate.

(b) Assume that  $w_{i,j} = 1$  for all i, j and return G = 5 is received. What are the updated values of the parameters assuming x = 0.5 and a learning rate of 0.5?

$$\begin{split} w_{i,j} \leftarrow w_{i,j} + \alpha * G \nabla_{w_{i,j}} \ln \pi_{w_{i,j}}(x, a_1), \text{ where } i = 0, 1, 2 \text{ and } j = 1, 2; \\ w_{0,1} \leftarrow 1 + 0.5 * 5 \nabla_{w_{0,1}} \ln \frac{e^{w_{0,1} + 0.5 + 0.25}}{e^{w_{0,1} + 0.5 + 0.25} + e^{1 + 0.5 + 0.25}} = 1 + 2.5 * \frac{e}{e + e} = 1 + 1.25 = 2.25 \\ w_{1,1} \leftarrow 1 + 0.5 * 5 \nabla_{w_{1,1}} \ln \frac{e^{1 + 0.5w_{1,1} + 0.25}}{e^{1 + 0.5w_{1,1} + 0.25} + e^{1 + 0.5 + 0.25}} = 1 + 2.5 * \frac{0.5e^{0.5}}{e^{0.5} + e^{0.5}} = 1 + 0.625 = 1.625 \end{split}$$

$$\begin{split} w_{2,1} \leftarrow 1 + 0.5 * 5 \nabla_{w_{2,1}} \ln \frac{e^{1 + 0.5 + 0.25 w_{2,1}}}{e^{1 + 0.5 + 0.25 w_{2,1}} + e^{1 + 0.5 + 0.25}} &= 1 + 2.5 * \frac{0.25 e^{0.25}}{e^{0.25} + e^{0.25}} = 1.3125 \\ w_{0,2} \leftarrow 1 + 0.5 * 5 \nabla_{w_{0,2}} \ln \frac{e^{1 + 0.5 + 0.25}}{e^{1 + 0.5 + 0.25} + e^{w_{0,2} + 0.5 + 0.25}} &= 1 + 2.5 * \frac{-e}{e + e} = 1 - 1.25 = 0.25 \\ w_{1,2} \leftarrow 1 + 0.5 * 5 \nabla_{w_{1,2}} \ln \frac{e^{1 + 0.5 + 0.25}}{e^{1 + 0.5 + 0.25} + e^{1 + 0.5 w_{1,2} + 0.25}} &= 1 + 2.5 * \frac{-0.5 e^{0.5}}{e^{0.5} + e^{0.5}} &= 1 - 0.625 = 0.325 \\ w_{2,2} \leftarrow 1 + 0.5 * 5 \nabla_{w_{2,2}} \ln \frac{e^{1 + 0.5 + 0.25}}{e^{1 + 0.5 + 0.25} + e^{1 + 0.5 + 0.25 w_{2,2}}} &= 1 + 2.5 * \frac{-0.25 e^{0.25}}{e^{0.25} + e^{0.25}} &= 0.6875 \end{split}$$