

Quiz# 1 : E1 277: Reinforcement Learning

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Department:
Programme:

1. Consider a multi-armed bandit with three arms A , B , and C . The decision maker pulls these arms in the order A, B, C, B, A, C, B and gets the rewards $5, 4, 1, 6, 3, 7, 2$, respectively. Assuming initial Q -values for each arm as zero, find the final values of $Q(A)$, $Q(B)$ and $Q(C)$? (2 marks)

• $N(A) = 1$

$$Q(A) \leftarrow Q(A) + \frac{1}{N(A)} (R - Q(A))$$
$$= 0 + 1(5 - 0) = 5$$

• $N(B) = 1$

$$Q(B) \leftarrow Q(B) + \frac{1}{N(B)} (R - Q(B)) = 4$$

• $N(C) = 1$

$$Q(C) \leftarrow Q(C) + \frac{1}{N(C)} (R - Q(C)) = 1$$

• $N(B) = 2$

$$Q(B) \leftarrow Q(B) + \frac{1}{N(B)} (R - Q(B))$$
$$= 4 + \frac{1}{2}(6 - 4) = 5$$

• $N(A) = 2$

$$Q(A) \leftarrow Q(A) + \frac{1}{N(A)} (R - Q(A))$$
$$= 5 + \frac{1}{2}(3 - 5) = 4$$

• $N(C) = 2$

$$Q(C) \leftarrow Q(C) + \frac{1}{N(C)} (R - Q(C))$$
$$= 1 + \frac{1}{2}(7 - 1) = 4$$

• $N(B) = 3$

$$Q(B) \leftarrow Q(B) + \frac{1}{N(B)} (R - Q(B))$$
$$= 5 + \frac{1}{3}(2 - 5) = 4$$

Final values:

$$Q(A) = Q(B) = Q(C) = 4.$$

2. Consider a finite horizon MDP for which the single stage costs at each of the times $0, 1, \dots, N-1$ are the same, i.e., the functions $g_0 = g_1 = \dots = g_{N-1} \triangleq g$. Let the terminal cost be g_N and it only depends on the terminal state as before. Assume now that $J_{N-1}(x) \geq g_N(x)$, $\forall x \in S$ (where S denotes the state space). Show that this implies $J_k(x) \geq J_{k+1}(x)$, for all $k = 0, 1, \dots, N-1$ and all $x \in S$. (3 marks)

As per DP algorithm,

$$J_N(x) = g_N(x) \quad \forall x.$$

$$J_{N-1}(x) = \min_{a \in A(x)} E_Y [g(x, a, Y) + J_N(Y)]$$

$$\begin{aligned} &\geq \min_{a \in A(x)} E_Y [g(x, a, Y) + J_N(Y)] \\ &= J_N(x), \quad \forall x. \end{aligned}$$

Proceeding similarly, we obtain

$$J_k(x) \geq J_{k+1}(x) \quad \forall k = 0, 1, \dots, N-1$$

and all x .