7_avg_exp.py

1. For all bots which are currently at a node & idle

Update:

- true idleness of all nodes in graph (true) = $+\Delta t$
- Store all the true idleness values at this time stamp
- Expected idleness of nodes at which bots are currently present is calculated as the avg of true idleness while going through a particular edge (now, expected idleness is function of edge not node)

<u>Calculate</u>: here, learning rate (α) = 0.1, discount factor (γ) = 0.95

• Value function all edges where bots are present (*Q*) =

$$Q^{new}(s_t, a_t) \leftarrow \underbrace{Q(s_t, a_t)}_{\text{old value}} + \underbrace{\alpha}_{\text{learning rate}} \cdot \underbrace{\left(\underbrace{r_t}_{\text{reward}} + \underbrace{\gamma}_{\text{discount factor}} \cdot \underbrace{\max_{a} Q(s_{t+1}, a)}_{\text{estimate of optimal future value}} - \underbrace{Q(s_t, a_t)}_{\text{old value}}\right)}_{\text{new value (temporal difference target)}}$$

- Reward $(r_t) = log(|expect true|)$ and $r_t = 0$ if [expect = true]
- Softmax of Value function = value_exp = $\frac{e^{Q_i} *expect [i]}{\sum_{j=i}^{k} e^{Q_{j*}} expect [j]}$

(summation over all edges)

Set:

• True idleness of nodes where bots are present = 0

OBSERVATION model: bot will calculate the expected idleness as an average of all the past true idleness it has seen when it last visited the node while travelling **along that particular edge**.

The name 7_avg_exp indicates => avg = averaging true idleness to get expected idleness (OBSERVATION model)

exp = including the expected idleness into the softmax calculation

2. For a bot deciding the next node to visit

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• True idleness of the node where the bot is present = 0

Decision Making: here, we chose ϵ =0.1

- With (1ε) probability, check all neighbours and visit the one with highest value of = $[value_exp]$
- With ϵ probability, go to a random node

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DRAWBACK:

• Even when |expect-true| = 1, reward = 0.