RL Entry Example Learning Note

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2.1 About Learning Model

2.1.1 Which type of RL

Value network, Q network or policy gradient? Answer: Q network. Each sate has 2 actions: left or right. And each action in that state has a Q value.

2.1.2 Update Model

How to make action decision At each state, the decision is made by selection the action with the largest Q value.

How is the Q value calculated Immediately after the decision is made, its corresponding Q value is updated, based on:

The immediate reward R, is dependent on whether it is destination. If it is destination, the value is 100, otherwise it is constant negative value.

The difference between the next Q and current Q

In total, the Q update rule is given by:

$$Q(S, a) = Q(S, a) + \alpha \{R + \gamma \max Q(S_a, :) - Q(S, a)\}$$

2.1.3 Deterministic world or stochastic world

This seems to be a deterministic world because the outcome cause by a certain action is deterministic.

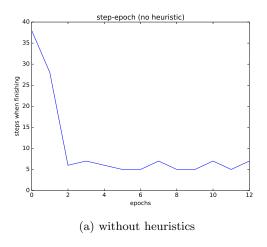
2.2 About Code

2.2.1 What is the format of q table?

Here q_table is a table generated by pd.DataFrame. Thus it is a pd.DataFrame object. The rows and columns of q_table can be retrieved by using pd's API. Each row is corresponding to a state, and each column is corresponding to an action. Here we have 6 states, and each state has 2 actions. This table can be constructed by using "pd.DataFrame".

	left	right
0	0	0
0	0	0
0	0	0
0	0	0
0	0	0
0	0	0

Table 1: The form of q table



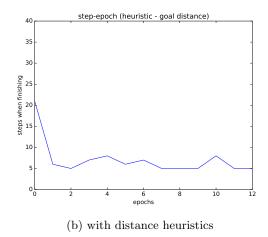


Figure 1: Step comparison

2.2.2 q table API

Here q_table is a table generated by pd.DataFrame. Thus it is a pd.DataFrame object. The rows and columns of q_table can be retrieved by using pd's API.

.iloc 'loaction based indexing'. The table.iloc[x,y], where x and y are numbers. The .loc method is the same as the indexing method as the traditional array indexing.

.ix 'Key indexing method'. The index of table.ix[a,b], where a and b should be the key. For example, in this tutorial, iloc[1,0] is the same as ix[a,']eft'.

2.3 Q Value Update Experiment

In this section we tested scripts with different Q value updates policy. In experiment 1, we use the default update policy, and in experiment 2, we add an heuristic to the update rule. The action which makes the agent closer to the goal destination will get an extra rewards. In both experiments, we record the step-epoch trend and the q-table after training.

The new heuristic is given by:

$$Q(S,a) = Q(S,a) + \alpha \{R + \gamma \text{max} Q(S_a,:) + \frac{0.1}{S_{\text{goal}} - S_a} - Q(S,a)\}$$

According to figure 1, the agent will have to use less steps to reach the goal in the first 2 epochs. However, based on the new heuristic, the q value may not be able to converge because it always gets rewards for getting closer to the goal. This reward will never be 0 even in very late epochs.