

## Assignment 1 - Bandits

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### 1. Multi-armed Bandits

#### 1.1 Greedy action probability

With the given  $\epsilon$  of 0.5, the greedy action is selected with a probability of

$$1 - \epsilon = 0.5. \quad (1)$$

#### 1.2 On which steps did the bandit explore?

$\epsilon$  definitely occurred at actions:

1.  $A_2 = 2$  - as it has an unknown reward, whereas greedy  $A_2 = 1$  has an expected reward of 1.
2.  $A_5 = 3$  - as it has an unknown reward, whereas  $A_5 = 2$  would be the greedy action.

$\epsilon$  might have occurred on action  $A_1 = 1$ , because all action-values are unknown and it's maximum cannot be chosen.

### 2. Action Selection Strategies

c)

The greedy method always chooses the arm with the highest  $Q$  value. Consequently, the method does not explore if there are better actions.

The  $\epsilon$ -greedy method explores other actions randomly with a chance of  $\epsilon$ . This has the downside of taking non-promising actions and might explain inferior results for the first 200 timesteps, with  $\epsilon = 0.1$ . With the chance of  $1-\epsilon$  greedy action is selected, with the benefit of better  $Q$  values due to exploration.

With more realistic  $Q$  values, the greedy method works better, which might explain the superior performance of the  $\epsilon$ -greedy method for timesteps  $> 200$ .

d)

1. It might be beneficial to explore more in the beginning and be greedy later on. This could be implemented such that  $\epsilon$  is not constant but decays with increasing timesteps.
2. It might be beneficial not to explore randomly, but explore the arm which was selected fewest.

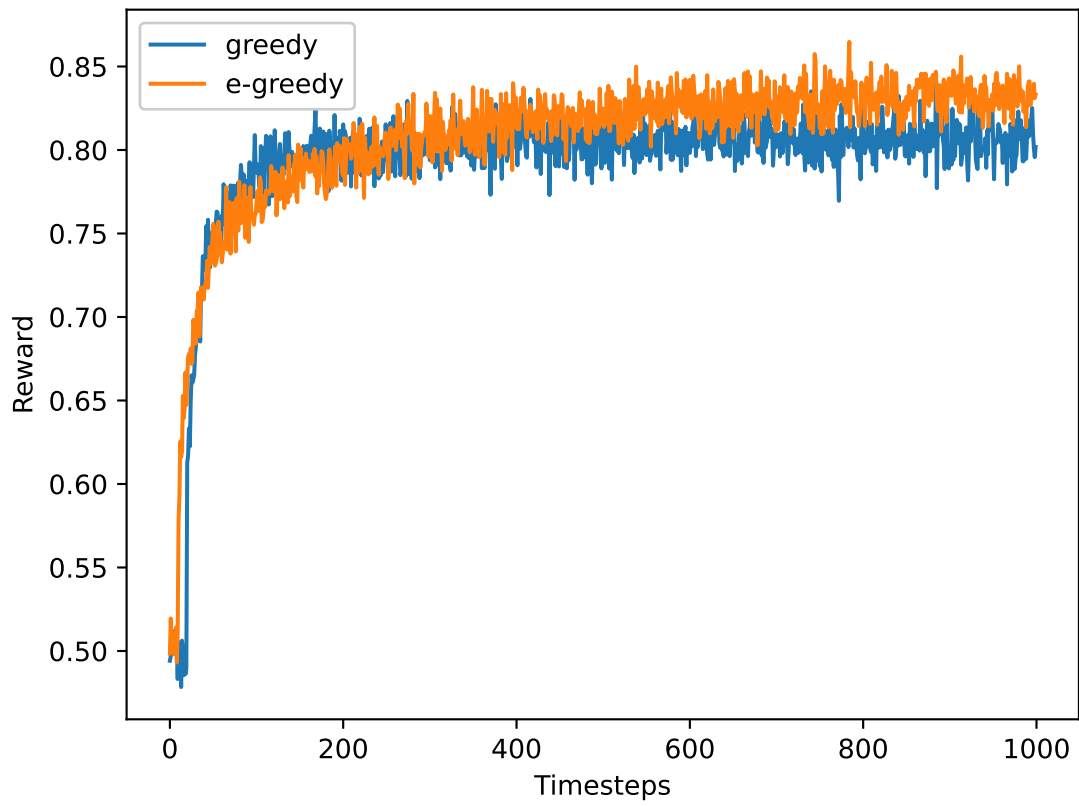


Figure 1: Bandit Strategies