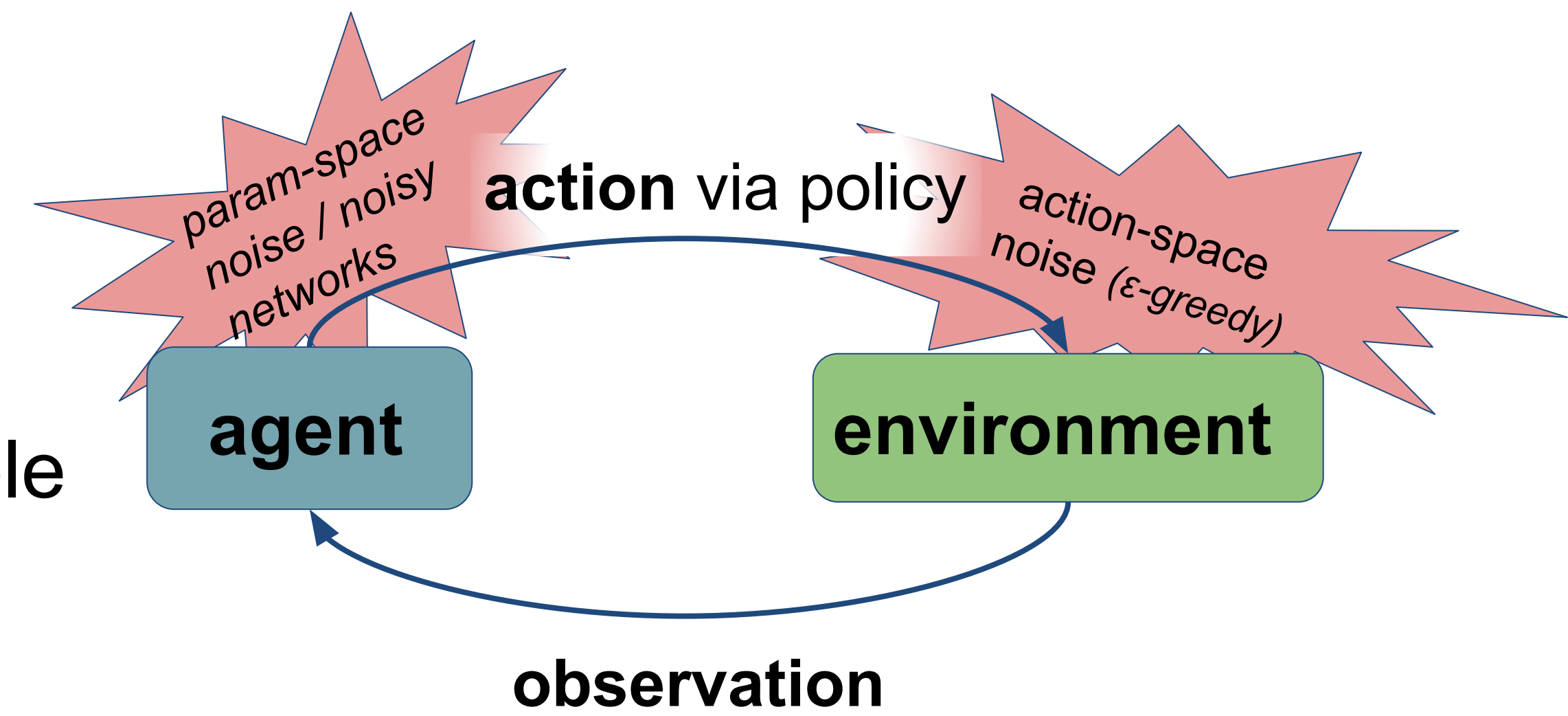


# Network Noise for Exploration

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## Project goal

- Extend OpenAI baselines by including Noisy Networks
- Compare training with other exploration strategies using Deep Double Q-Learning Networks (DDQN)



## Exploration

- Exploration is needed for the agent to move to new states in its environment
- This aids in exploiting the environment by taking the best actions in every state, to reach its goal
- Three different strategies considered

### 1) $\epsilon$ -greedy

- Classic exploration strategy (simple, established)
- Deterministic action choice which maximizes expected reward (Q)
- With probability of  $\epsilon$ : choose random action instead
- Decrease  $\epsilon$  over time

### 2) Parameter Space Noise

- Noise added to weights and biases of action-value-model (MLP)
- Adapt noise to keep a distance  $\delta$  between perturbed and non-perturbed policy
- Reduce  $\delta$  over time

### 3) Noisy Networks

- Noise added to weights and biases of action-value-model (MLP)
- Parameters  $(\mu, \sigma)$  of noise learned with gradient descent along with network weights and biases

## Experiments

- Run on Mountain Car environment (Figure 1)
- Only  $\epsilon$ -greedy solved environment perfectly after long training (approx. 350,000 steps)
- *Parameter Space Noise* learned slower and performed worse (Figure 2)
- *Noisy Networks* learned slower than  $\epsilon$ -greedy
- *Noisy Networks* reached similar reward than  $\epsilon$ -greedy for short training
- Noisy Networks don't need a manually defined  $\epsilon$ - or  $\delta$ -decay function

Figure 1: The underpowered car has to reach the flag as fast as possible.

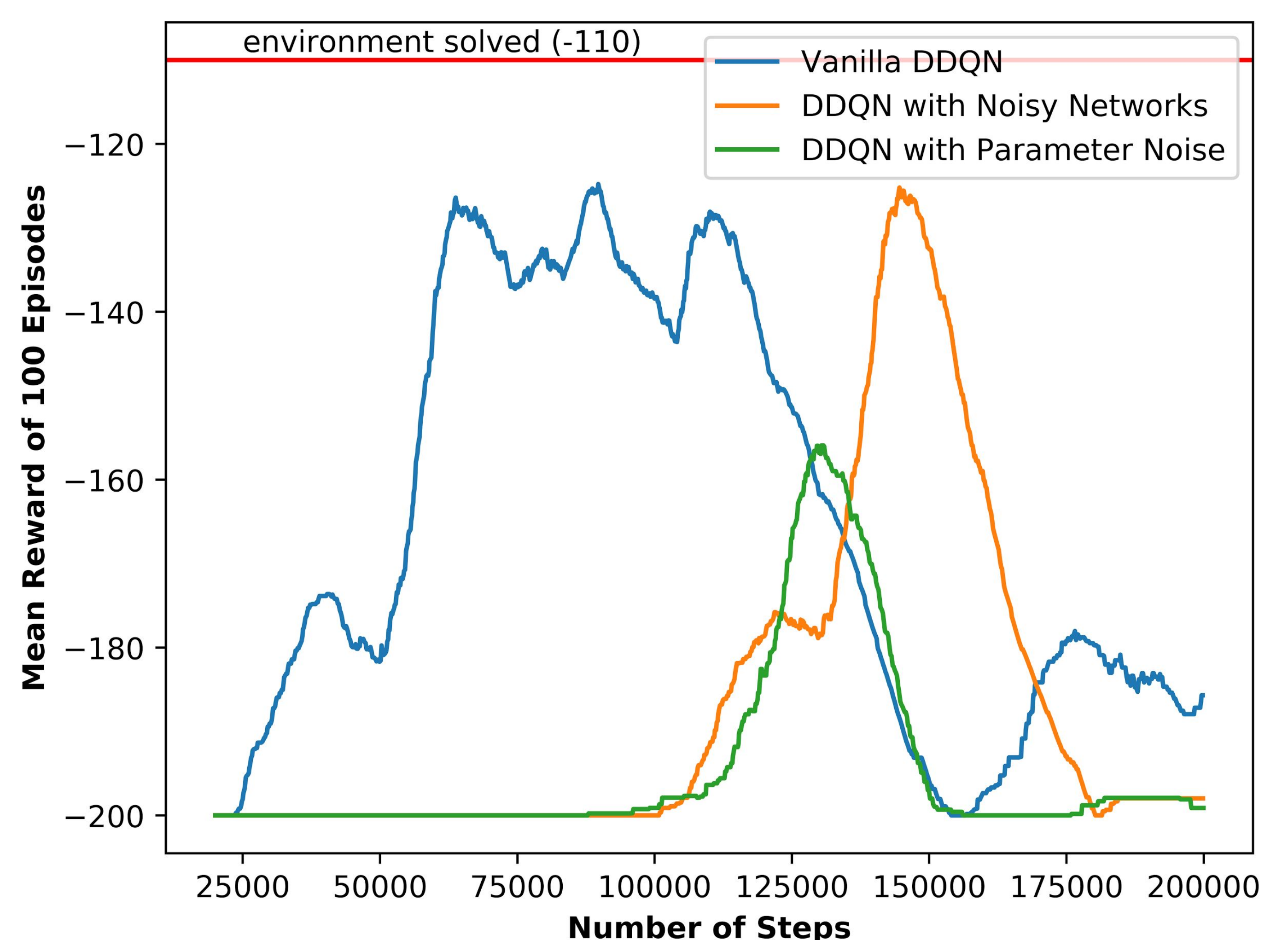
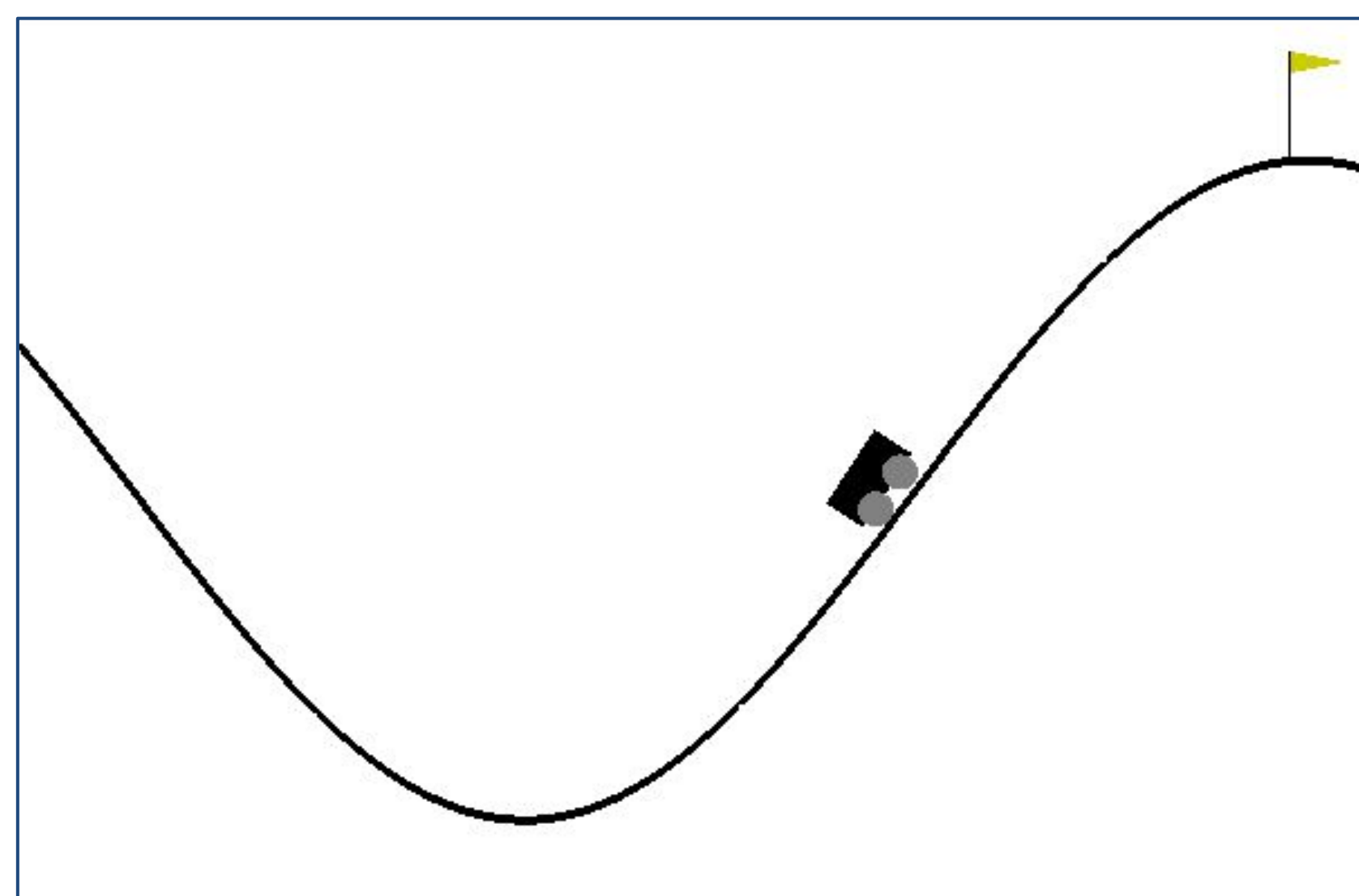


Figure 2: Reward while training DDQN reinforcement learning model in Mountain Car environment with different exploration strategies.

**References:** Fortunato, M. *et al.* Noisy Networks for Exploration. (2017). Plappert, M. *et al.* Parameter Space Noise for Exploration. (2017). Dhariwal, P. *et al.* OpenAI Baselines. (GitHub, 2017);

**Code & Data:** [https://github.com/dekuenstle/dl\\_lab\\_2017/tree/master/final\\_project\\_robotics](https://github.com/dekuenstle/dl_lab_2017/tree/master/final_project_robotics)