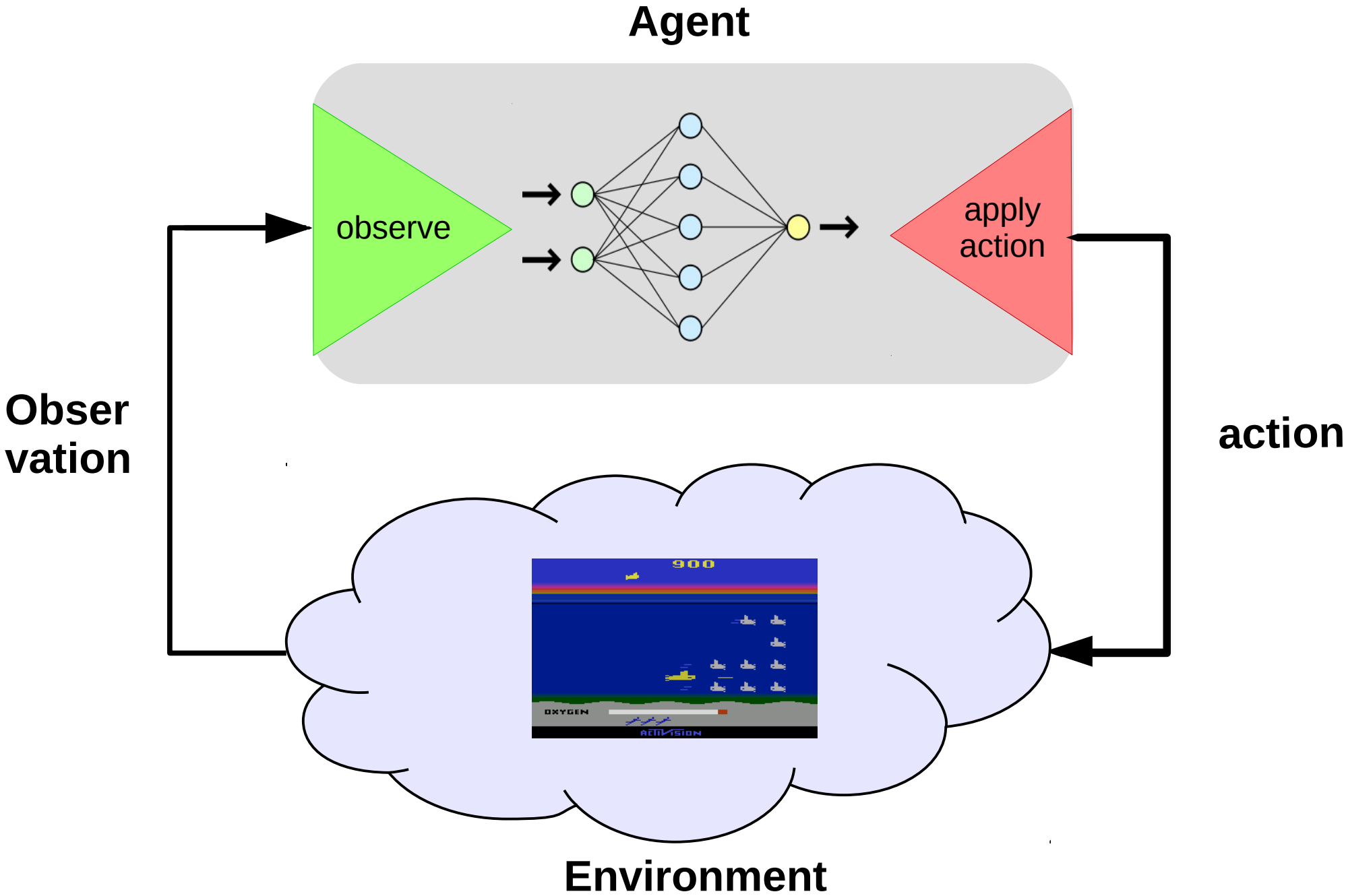
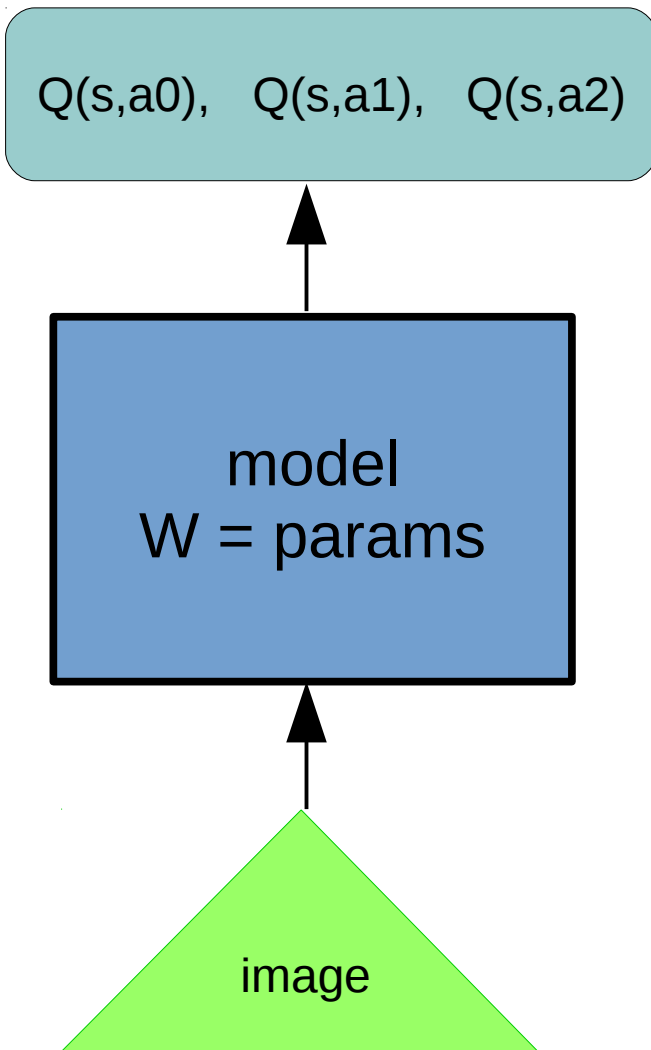


# MDP & POMDP



# Approximate Q-learning

nice and simple



**Q-values:**

$$\hat{Q}(s_t, a_t) = r + \gamma \cdot \operatorname{argmax}_{a'} \hat{Q}(s_{t+1}, a')$$

**Objective:**

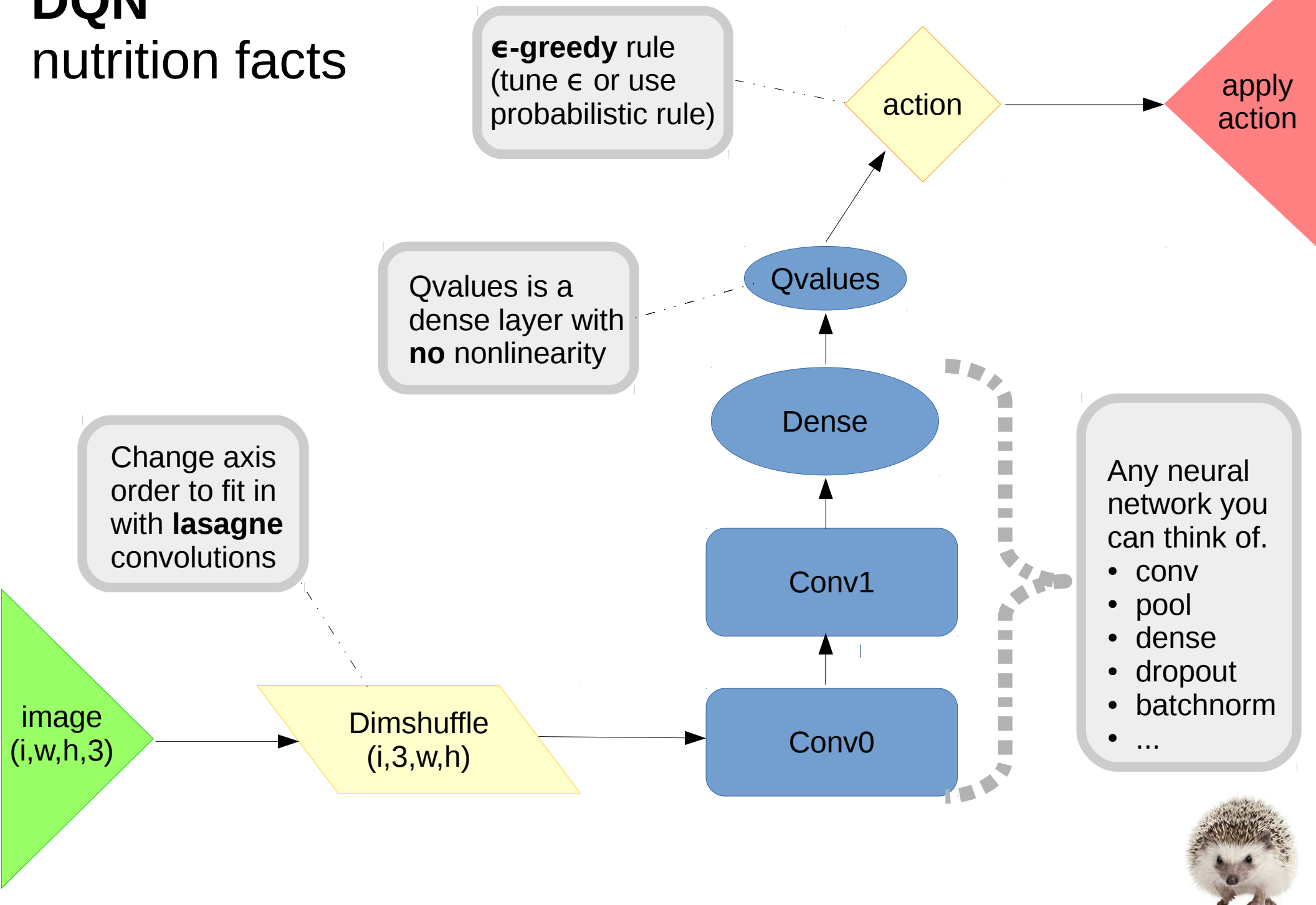
$$L = \left( Q(s_t, a_t) - r + \gamma \cdot \operatorname{argmax}_{a'} Q(s_{t+1}, a') \right)^2$$

**Gradient step:**

$$w_{t+1} = w_t - \alpha \cdot \frac{\delta L}{\delta w}$$

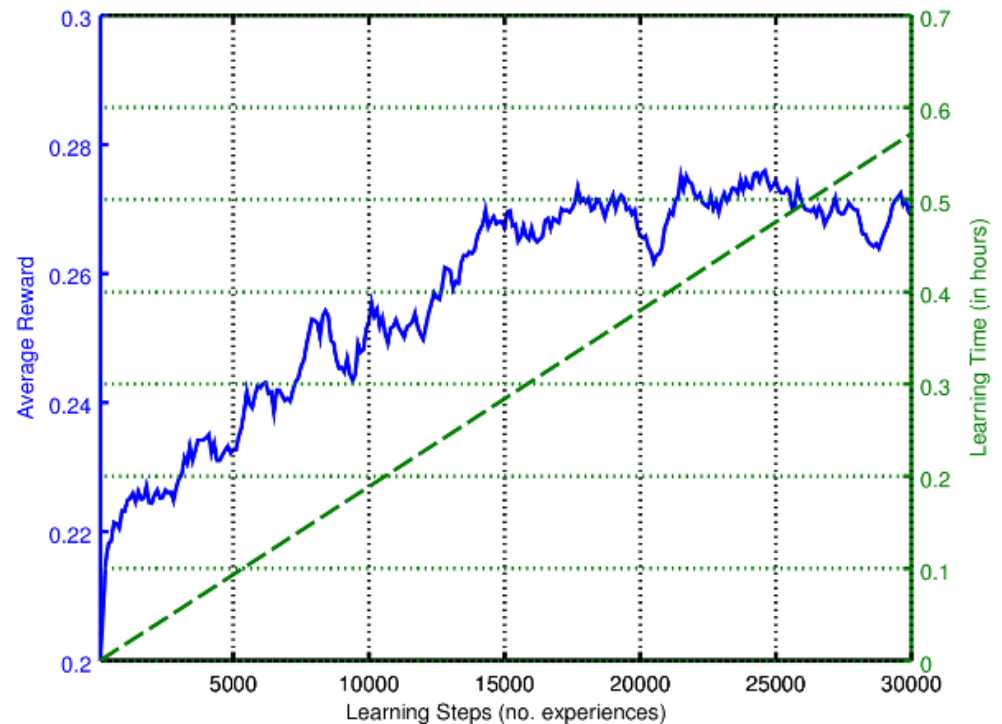
# DQN

nutrition facts



# Approximate Q-learning problems

- Training samples are **not** “i.i.d”,
- Model forgets parts of environment it haven't visited for some time,
- Fallbacks on the learning curve
- Any ideas?



# Deep Q-learning

## Multiple agent trick

**Idea:** Throw in several agents with shared  $W$ .

- Chances are, they will be exploring different parts of the environment,
- More stable training,
- Requires a lot of interaction,
- Alternative to experience replay.

