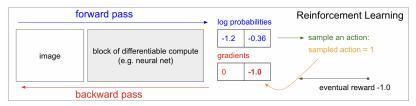
Deep Deterministic Policy Gradient (DDPG) for Atari Galaga CS4100 Project Proposal

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Policy Gradient Methods



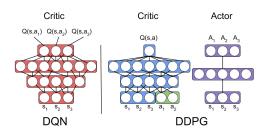
(note that this uses stochastic policy)

Policy Gradient Methods

$$J(\mu_{\theta}) = \int_{\mathcal{S}} \rho^{\mu}(s) r(s, \mu_{\theta}(s)) ds$$
$$= \mathbb{E}_{s \sim \rho^{\mu}} \left[r(s, \mu_{\theta}(s)) \right] \tag{8}$$

$$\nabla_{\theta} J(\mu_{\theta}) = \int_{\mathcal{S}} \rho^{\mu}(s) \nabla_{\theta} \mu_{\theta}(s) \left. \nabla_{a} Q^{\mu}(s, a) \right|_{a = \mu_{\theta}(s)} ds$$
$$= \mathbb{E}_{s \sim \rho^{\mu}} \left[\nabla_{\theta} \mu_{\theta}(s) \left. \nabla_{a} Q^{\mu}(s, a) \right|_{a = \mu_{\theta}(s)} \right]$$
(9)

Actor-Critic



- 2 Networks:
 - critic: (s, a) \rightarrow V(s')
 - actor: $s \rightarrow a$
- Continuous Actions
- Action-Replay Buffer
- Target Network Updates

Frameworks

- OpenAl Gym
- PyTorch/Autograd
- OpenCV
- LibVirt

Approach

- 1 Set up environment
- 2 Implement Deep Stochastic Policy Gradient Algorithm
- 3 Add Replay Buffer, Target Network
- 4 Implement Actor-Critic For Deterministic Policy Gradient Algorithm

References

- DeepMind's 'Continuous Control with Deep Reinforcement Learning' (2016)
- DeepMind's 'Deterministic Policy Gradient Algorithms' (2014)
- DeepMind's 'Playing Atari with Deep Reinforcement Learning' (2015)
- Karpathy's 'Deep Reinforcement Learning: Pong from Pixels' (https://karpathy.github.io/2016/05/31/rl/)