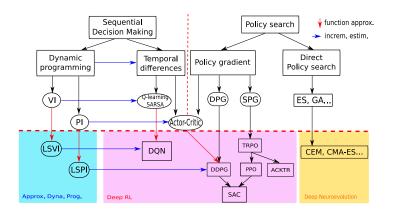
From Policy Gradient to Actor-Critic methods Introduction: the 4 routes to deep RL

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The Big Picture

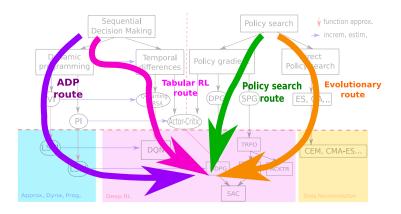


► A very partial view of the whole Deep RL literature



Press.

The four routes

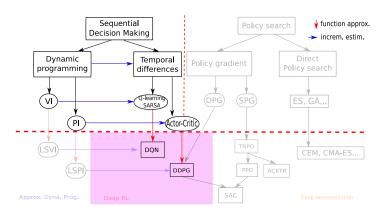


► Four different ways to come to Deep RL



(B) (B) (B) (B) (B)

The Tabular RL route

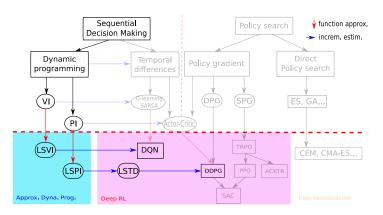


- The favorite route of beginners
- ► Start from Sutton&Barto, present Q-learning, SARSA and Actor-Critic
- Add function approximation, go to DQN, then DDPG





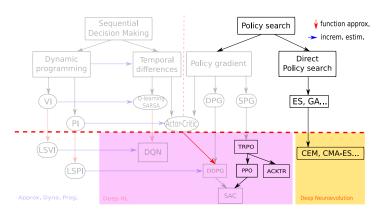
The Approximate Dynamic Programming route



- ▶ The favorite route of mathematicians
- ▶ I never travelled this route



The Evolutionary route

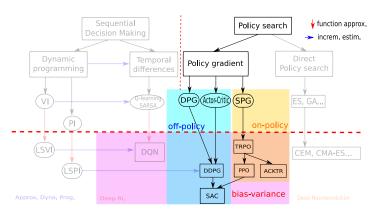


- The favorite route of non-RL people
- ► Much more efficient than RL people think





The Policy Search route



- ► The favorite route of roboticists
- ► The one I'm travelling in these lessons





Outline of lessons content

- 1. The policy search problem
- 2. Policy Gradient derivation
- 3. Understanding the Policy Gradient
- 4. From policy gradient with baseline to actor-critic
- 5. Bias-variance trade-off
- 6. On-policy vs off-policy
- 7. TRPO, ACKTR and PPO
- 8. DDPG and TD3
- 9. SAC
- 10. Wrap-up



Any question?



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