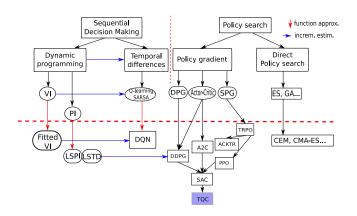
From Policy Gradient to Actor-Critic methods Introduction: the 4 routes to deep RL

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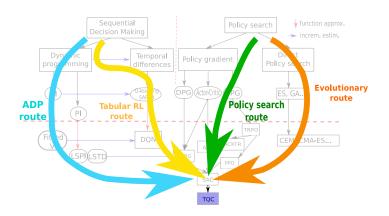
The Big Picture



A very partial view of the whole RL literature



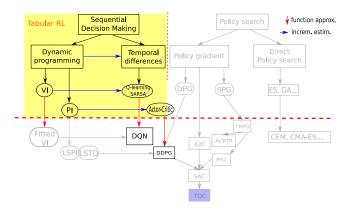
The four routes to deep RL



► Four different ways to come to Deep RL



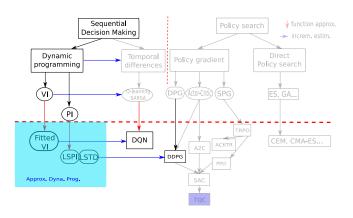
The Tabular RL route



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

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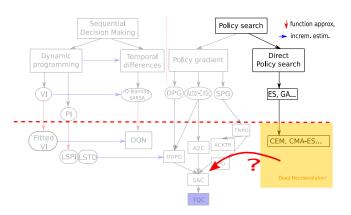
The Approximate Dynamic Programming route



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



The Evolutionary route

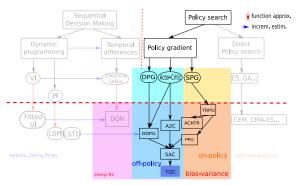


- ▶ The favorite route of black-box optimisation people
- ► Much more efficient than RL people think





The Policy Search route



- ► The favorite route of roboticists
- ► The one I'm travelling in these lessons
- ► Central question: difference between PG with baseline and Actor-Critic



Marc P. Deisenroth, Gerhard Neumann, Jan Peters, et al. A survey on policy search for robotics. Foundations and Trends® in Robotics, 2(1-2):1-142, 2013

Outline

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



Any question?



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References



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