

Reinforcement Learning

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1 Lectures

1. Introduction to Reinforcement Learning
2. Markov Decision Processes
3. Dynamic Programming
4. Monte Carlo Methods
5. Temporal-Difference Learning
6. n-Step Bootstrapping
7. Planning and Learning with Tabular Methods
8. Reinforcement Learning in Finite State and Action Spaces
9. Function Approximation with Supervised Learning
10. On-Policy Prediction with Function Approximation
11. Value-Based Control with Function Approximation
12. Eligibility Traces
13. Policy Gradient Methods
14. Further Contemporary RL Algorithms
15. Course Completion and Outlook

2 Tutorials

1. Python Tutorial
2. Markov Decision Processes
3. Dynamic Programming
4. Monte Carlo Methods
5. Temporal-Difference Learning

6. [n-Step Methods](#)
 7. [Learning and Planning](#)
 8. [Function Approximation with Supervised Learning](#)
 9. [Function Approximators in Value Prediction](#)
 10. [Function Approximators in Control](#)
 11. [Eligibility Traces](#)
 12. [Policy Gradients](#)
 13. [DDPG & PPO](#)
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