

RL-Course

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Table of contents

Preface	4
Outline	5
1 The Dynamic Programming Algorithm	6
1.1 Introduction	6
1.2 The Basic Problem	6
1.3 The Dynamic Programming Algorithm	6
1.4 State Augmentation and Other Reformulations	6
1.5 Some Mathematical Issues	6
1.6 Dynamic Programming and Minimax Control	6
1.7 Notes, Sources, and Exercises	6
2 Finite Markov Decision Processes	7
2.1 The Agent–Environment Interface	7
2.2 Goals and Rewards	7
2.3 Returns and Episodes	7
2.4 Unified Notation for Episodic and Continuing Tasks	7
2.5 Policies and Value Functions	7
2.6 Optimal Policies and Optimal Value Functions	7
2.7 Optimality and Approximation	7
2.8 Summary	7
3 Dynamic Programming	8
3.1 Policy Evaluation (Prediction)	8
3.2 Policy Improvement	8
3.3 Policy Iteration	8
3.4 Value Iteration	8
3.5 Asynchronous Dynamic Programming	8
3.6 Generalized Policy Iteration	8
3.7 Efficiency of Dynamic Programming	8
3.8 Summary	8
4 Applications	9
4.1 Recycling Robot	9
4.2 A robot with randomly moves in a grid world.	9

Preface

This notes are based in the course from Berstekas for the MIT see all lectures and other resources for complete the understanding.

Outline

The textbook for chapter one is the Bertsekas' book (Bertsekas 2005). Chapter 2 and 3 are adapted from Suttons' book (Ch.3,Ch.4 Sutton and Barto 2018). For application and broad conection with more Matchine Learning Applications we refer to (Brunton and Kutz 2019). Also we recomend for a handbook of algorithms (Szepesvári [2022] \copyright 2022). For applictions wiht implemented code we follows the books (Bilgin 2020).

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4.2 A robot with randomly moves in a grid world.

References

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