1. Record Nr. UNINA9910824178803321 Autore Schwartz Howard M. Titolo Multi-agent machine learning: a reinforcement approach / / Howard M. Schwartz Pubbl/distr/stampa Hoboken, New Jersey:,: John Wiley & Sons, Inc.,, 2014 ©2014 **ISBN** 1-118-88448-5 1-118-88461-2 1-118-88447-7 Edizione [1st edition] Descrizione fisica 1 online resource (458 p.) Classificazione TEC008000 Disciplina 519.3 Soggetti Reinforcement learning Differential games Swarm intelligence Machine learning Lingua di pubblicazione Inglese **Formato** Materiale a stampa Livello bibliografico Monografia Note generali Description based upon print version of record. Nota di bibliografia Includes bibliographical references at the end of each chapters and index. Nota di contenuto Cover; Title Page; Copyright; Preface; References; Chapter 1: A Brief Review of Supervised Learning; 1.1 Least Squares Estimates; 1.2 Recursive Least Squares; 1.3 Least Mean Squares; 1.4 Stochastic Approximation; References; Chapter 2: Single-Agent Reinforcement Learning; 2.1 Introduction; 2.2 n-Armed Bandit Problem; 2.3 The Learning Structure: 2.4 The Value Function: 2.5 The Optimal Value Functions: 2.6 Markov Decision Processes: 2.7 Learning Value Functions; 2.8 Policy Iteration; 2.9 Temporal Difference Learning; 2.10 TD Learning of the State-Action Function; 2.11 Q-Learning 2.12 Eligibility TracesReferences; Chapter 3: Learning in Two-Player Matrix Games; 3.1 Matrix Games; 3.2 Nash Equilibria in Two-Player Matrix Games; 3.3 Linear Programming in Two-Player Zero-Sum Matrix Games: 3.4 The Learning Algorithms: 3.5 Gradient Ascent Algorithm: 3.6 WoLF-IGA Algorithm; 3.7 Policy Hill Climbing (PHC); 3.8 WoLF-PHC Algorithm; 3.9 Decentralized Learning in Matrix Games; 3.10 Learning

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Sommario/riassunto

"Multi-Agent Machine Learning: A Reinforcement Learning Approach is a framework to understanding different methods and approaches in multi-agent machine learning. It also provides cohesive coverage of the latest advances in multi-agent differential games and presents applications in game theory and robotics. Framework for understanding a variety of methods and approaches in multi-agent machine learning. Discusses methods of reinforcement learning such as a number of forms of multi-agent Q-learning Applicable to research professors and graduate students studying electrical and computer engineering, computer science, and mechanical and aerospace engineering"-"Provide an in-depth coverage of multi-player, differential games and Gam theory"--