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Robot Path Planning and Cooperation - Foundations, Algorithms and Experimentations

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Abstract

This book presents extensive research on two main problems in robotics: the path planning problem and the multi-robot task allocation problem. It is the first book to provide a comprehensive solution for using these techniques in large-scale environments containing randomly scattered obstacles. The research conducted resulted in tangible results both in theory and in practice. For path planning, new algorithms for large-scale problems are devised and implemented and integrated into the Robot Operating System (ROS). The book also discusses the parallelism advantage of cloud computing techniques to solve the path planning problem, and, for multi-robot task allocation, it addresses the task assignment problem and the multiple traveling salesman problem for mobile robots applications. In addition, four new algorithms have been devised to investigate the cooperation issues with extensive simulations and comparative performance evaluation. The algorithms are implemented and simulated in MATLAB and Webots.

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 Springer

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Preface

The objective of the book is to provide the reader with a comprehensive coverage of two important research problems in mobile robots, namely global path planning and cooperative multi-robots applications with a focus on multi-robot task allocation (MRTA) problem. As such, this book is organized in two major parts: Global Path Planning, and Multi-Robot Task Allocation. The objective of the first part of the book is to respond to a research question that we have been investigating along the two-year period of the iroboapp project: considering the vast array of AI techniques used to solve the robot path planning problem ranging from evolutionary computation techniques (e.g. GA, ACO) to meta-heuristic methods (e.g. A*), which technique is the best? In this part, we first revisit the foundations and present a background of the global path planning problem, and the underlying intelligent techniques used to solve it. Then, we present our new intelligent algorithms to solve these problems, based on common artificial intelligence approaches, and we analyze their complexities. Different simulation models using C++, MATLAB and others have been devised. An extensive comparative performance evaluation study between the path planning algorithms is presented. In addition, we validate our results through real-world implementation of these algorithms on real robots using the Robot Operation System (ROS). The second part of the book deals with cooperative mobile robots. We focus on the multi-robot task allocation (MRTA) problem and we present a comprehensive overview on this problem. Then, we present a distributed market-based mechanism for solving the multiple depot, multiple travel salesman problem which is a typical problem for several robotics applications. A major contribution of this book is that it bridges the gap between theory and practice as it shows how to integrate the global path planning algorithms in the ROS environment and it proves their efficiency in real scenarios. We believe that this handbook will provide the readers with a comprehensive reference on the

global path planning and MRTA problems starting from foundations and modeling, going through simulations and real-world deployments. Links to videos and demonstrations will be included in the book.

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Acronyms

2PPLS	Two-phase Pareto local search
A*	The Astar algorithm
ABC	Artificial bee colony
ACO	Ant Colony Optimization
AD*	Anytime Dynamic A*
AM	Application Master
ANA*	Anytime Nonparametric A*
APF	Artificial Potential Field
ARA*	Anytime Repairing A*
BLE	Broadcast of Local Eligibility
BSP	Bulk Synchronous Parallel
CACO	Conventional ACO
CFor	A set of forbidden configuration
CFree	A set of free configuration
CPD	Compressed path databases technique
CYX	Cycle crossover operator
DWA	Dynamic Window Approach
E*	The E Star algorithm
EDA	Estimation of distribution algorithm
FCE	Free configuration eigen-spaces
FIS	Fuzzy Inference System
FMM	Fast marching method
FOD	Front obstacle distance
GA	Genetic Algorithm
GGA	Grouping genetic algorithms
GGA-SS	Steady-state grouping genetic algorithm
GRASP	Greedy Randomized Adaptive Search Procedure
HACO	Heterogeneous ACO
HDFS	Hadoop Distributed File System
IDPGA	Improved dual-population GA

ILS	Iterated Local Search
IWO	Invasive weed optimization
JPS	Jump point search
LOD	Left obstacle distance
MACO	Modified ACO
MD-MTSP	Multiple Depots MTSP
MLP	Multi-Layer Perceptron
MOKPs	Multi-objective Knapsack problems
MOP	Multi-Objective Optimization
MPCNN	Modified pulsecoupled neural network
MRS	Multi-Robot System
MRTA	Multi-Robot Task Allocation
MTD	Maximum Traveled Distance
MT	Maximum tour
MTSP	Multiple Traveling Salesmen Problem
NM	Node Manager
NN	Neural Networks
ORX	Ordered crossover operator
PFM	The Artificial potential field approach
PFM	Potential field method
PMX	Partially-matched crossover operator
PPaaS	Path Planning as a Service
PRM	The probabilistic roadmap method
PSO	Particle Swarm Optimization
QHS	Quad Harmony Search
RA*	Relaxed AStar
RM	Global Resource Manager
ROD	Right obstacle distance
ROS	Robot Operating System
RRT	Rapidly-exploring random tree
RTMA	Robot and Task Mean Allocation Algorithm
S+T	Services and Tasks
SA	Simulated Annealing
SOM	Self Organizing Maps
SP-CNN	Shortest path cellular neural network
SSSP	Single source shortest path algorithm
TSP	Traveling Salesmen Problem
TS	Tabu Search
TTD	Total Traveled Distance
TWD*	Two Way D*
UAV	Unmanned Air Vehicle
VNS	Variable Neighborhood Search
VRP	Vehicle routing problem
YARN	Yet Another Resource Negotiator

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