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Jan Kozak

# Decision Tree and Ensemble Learning Based on Ant Colony Optimization

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*My greatest, heartfelt thanks go to my beloved wife Dorota, who stayed with me in good and bad moments—and to my dear daughters: Asia, who showed me, that there is no such thing as impossible and that I should never give up, and Ola, who taught me how much we can achieve in life by helping others.*

*Thank you...*

# Preface

This book, as its title suggests, is devoted to decision trees and ensemble learning based on ant colony optimization. Accordingly, it not only concerns the aforementioned important topics in the area of machine learning and combinatorial optimization, but combines them into one. It is this combination that was decisive for choosing the material to be included in the book and determining its order of presentation.

Decision trees are a popular method of classification as well as of knowledge representation. As conceptually straightforward, they are very appealing and have proved very useful due to their computational efficiency and generally acceptable effectiveness. At the same time, they are easy to implement as the building blocks of an ensemble of classifiers—in particular, decision forests. It is worth mentioning that decision forests are now considered to be the best available off-the-shelf classifiers. Admittedly, however, the task of constructing a decision tree is a very complex process if one aims at the optimality, or near-optimality, of the tree. Indeed, we usually agree to proceed in a heuristic way, making multiple decisions: one at each stage of the multistage tree construction procedure, but each of them obtained independently of any other. However, construction of a minimal, optimal decision tree is an NP-complete problem. The underlying reason is that local decisions are in fact interdependent and cannot be found in the way suggested earlier.

The good results typically achieved by the ant colony optimization algorithms when dealing with combinatorial optimization problems suggest the possibility of using that approach effectively also for constructing decision trees. The underlying rationale is that both problem classes can be presented as graphs. This fact leads to the possibility of considering a larger spectrum of solutions than those based on the heuristic mentioned earlier. Namely, when searching for a solution, ant colony optimization algorithms allow for using additional knowledge acquired from the pheromone trail values produced by such algorithms. Moreover, ant colony optimization algorithms can be used to advantage when building ensembles of classifiers.

This book is a combination of a research monograph and a textbook. It can be used in graduate courses, but should also be of interest to researchers, both specialists in machine learning and those applying machine learning methods to cope with problems from any field of R&D. The topics included in the book are discussed thoroughly. Each of them is introduced in such a way as to make it accessible to a computer science or engineering graduate student. All methods included are discussed from many angles, in particular regarding their applicability and the ways to choose their parameters. In addition, many unpublished results are presented.

The book is divided into two parts, preceded by an introduction to machine learning and swarm intelligence. The first part discusses decision tree learning based on ant colony optimization. It includes an introduction to evolutionary computing techniques for data mining (Chap. 2). Chapter 3 describes the most popular ant colony algorithms used for learning decision trees. Chapter 4 introduces some modifications of the ant colony decision tree algorithm. Examples of practical applications of the methods discussed earlier are presented in Chap. 5.

The second part of the book discusses ensemble learning based on ant colony optimization. It begins (Chap. 6) with an introduction presenting known solutions of a similar type—more precisely, evolutionary computing techniques in ensemble learning—as well as a literature review. Chapter 7 includes formal definitions and example applications of ensembles based on random forests. Chapter 8 introduces an adaptive approach to building a decision forest with ant colony optimization. The book concludes with some final remarks and suggestions for future research.

Let me end with mentioning the persons whose advice, help and support were indispensable for writing this book. Special thanks are due to Prof. Jacek Koronacki for the possibility of collaboration, his extraordinary commitment, fruitful discussions and all comments. Similar thanks to Prof. Beata Konikowska, who showed unusual meticulousness and gave me many valuable remarks and pieces of advice. I would also like to thank Prof. Mikhail Moshkov for his continuing comments and long discussions. Sincere thanks are due to Dr. Przemysław Juszczak, who has been deeply involved in creating this book.

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