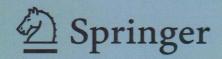


Computational Intelligence

A Methodological Introduction

Third Edition



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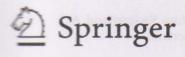
Rudolf Kruse • Sanaz Mostaghim • Christian Borgelt • Christian Braune • Matthias Steinbrecher

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With contributions from Frank Klawonn and Christian Moewes



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