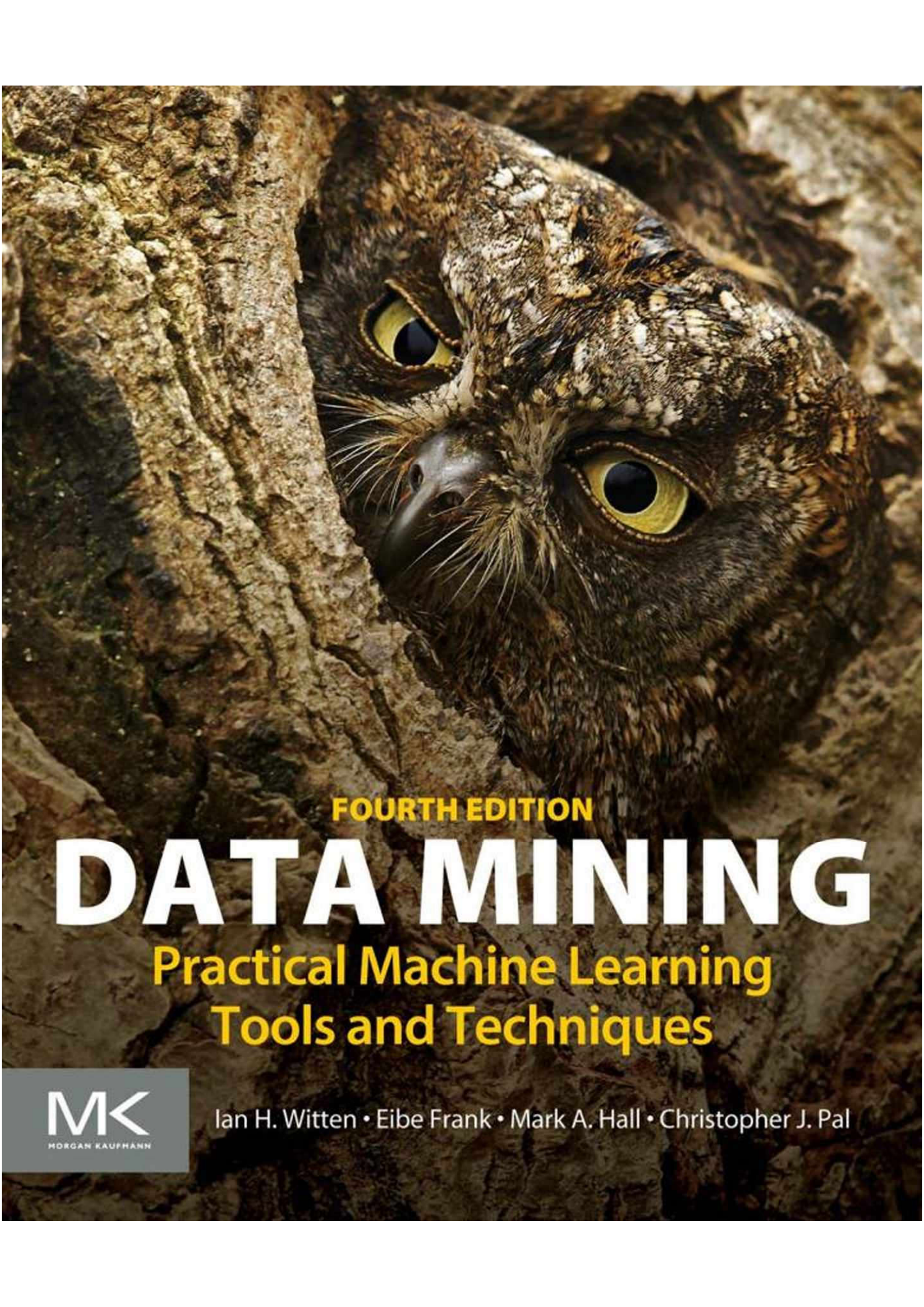




FOURTH EDITION

DATA MINING

**Practical Machine Learning
Tools and Techniques**

MK
MORGAN KAUFMANN

Ian H. Witten • Eibe Frank • Mark A. Hall • Christopher J. Pal

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Fourth Edition

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