

WILEY SERIES IN COMPUTATIONAL STATISTICS



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DATA MINING AND STATISTICS FOR DECISION MAKING

 **WILEY**

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