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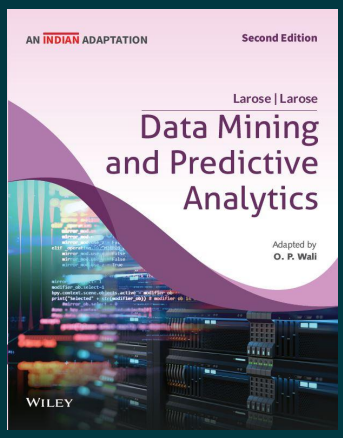

Data Mining and Predictive Analytics, 2ed (An Indian Adaptation)

By Daniel T. Larose,Chantal D. Larose,O.P. Wali

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Description

Data Mining and Predictive Analytics serves as an introduction to data mining methods and models, including association rules, clustering, neural networks, logistic regression, and multivariate analysis. The authors apply a unified “white box” approach to data mining methods and models. This approach is designed to walk readers through the operations and nuances of the various methods, using small data sets, so readers can gain an insight into the inner workings of the method under review.

About the Author

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