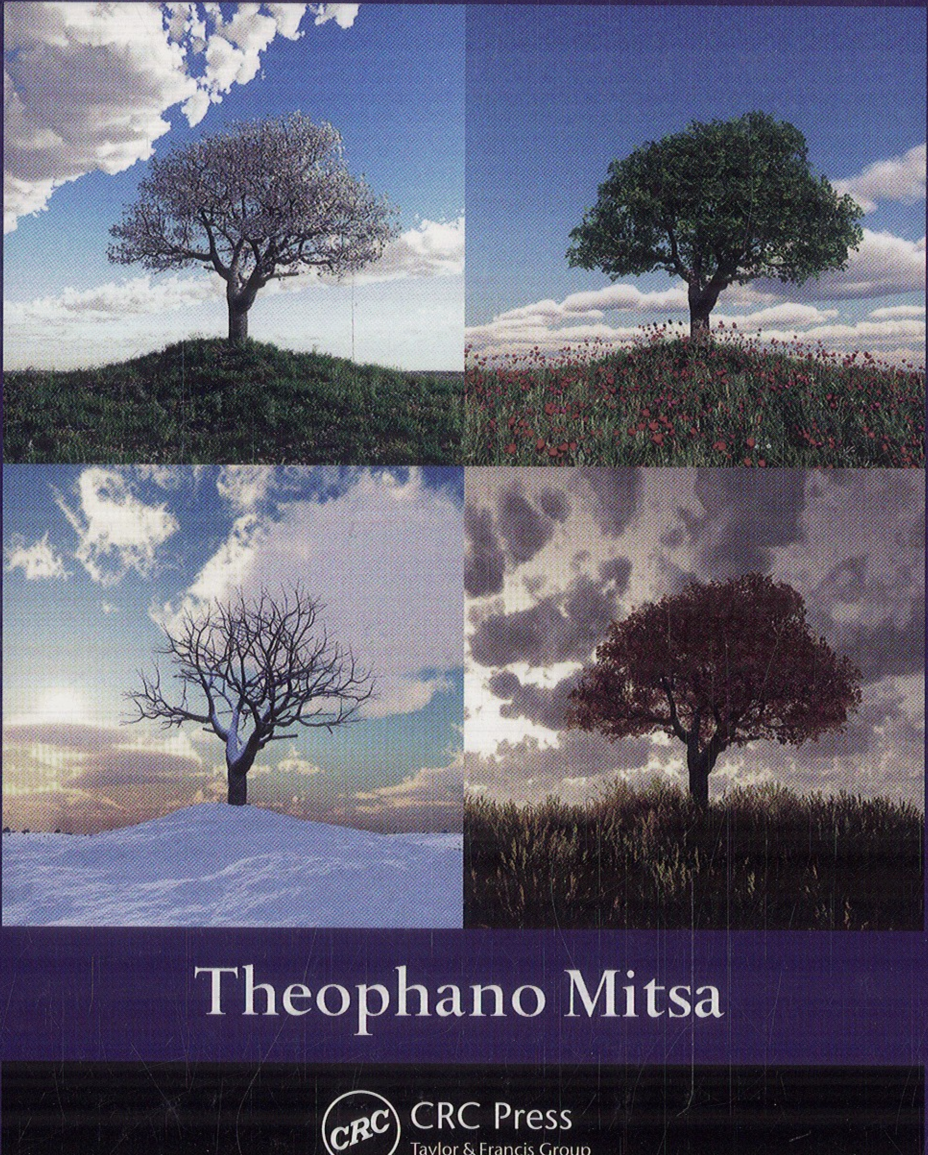
Chapman & Hall/CRC
Data Mining and Knowledge Discovery Series

Temporal Data Mining



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