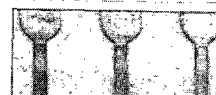




Data Mining

**Practical Machine Learning Tools and Techniques,
Second Edition**

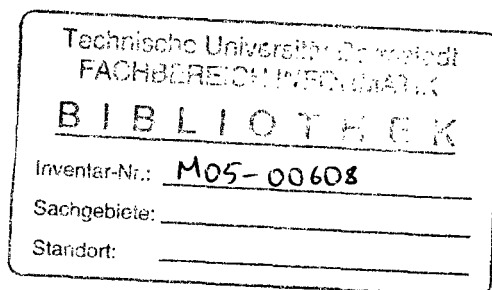


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