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FUNDAMENTALS OF

MACHINE LEARNING
FOR **PREDICTIVE DATA ANALYTICS**

ALGORITHMS, WORKED EXAMPLES, AND CASE STUDIES

มหาวิทยาลัยเทคโนโลยีราชมงคลพระนคร

ห้องสมุดสาขาพระนครเหนือ



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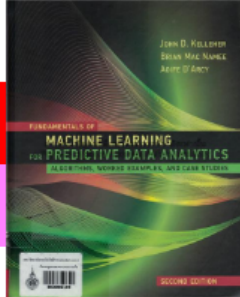
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