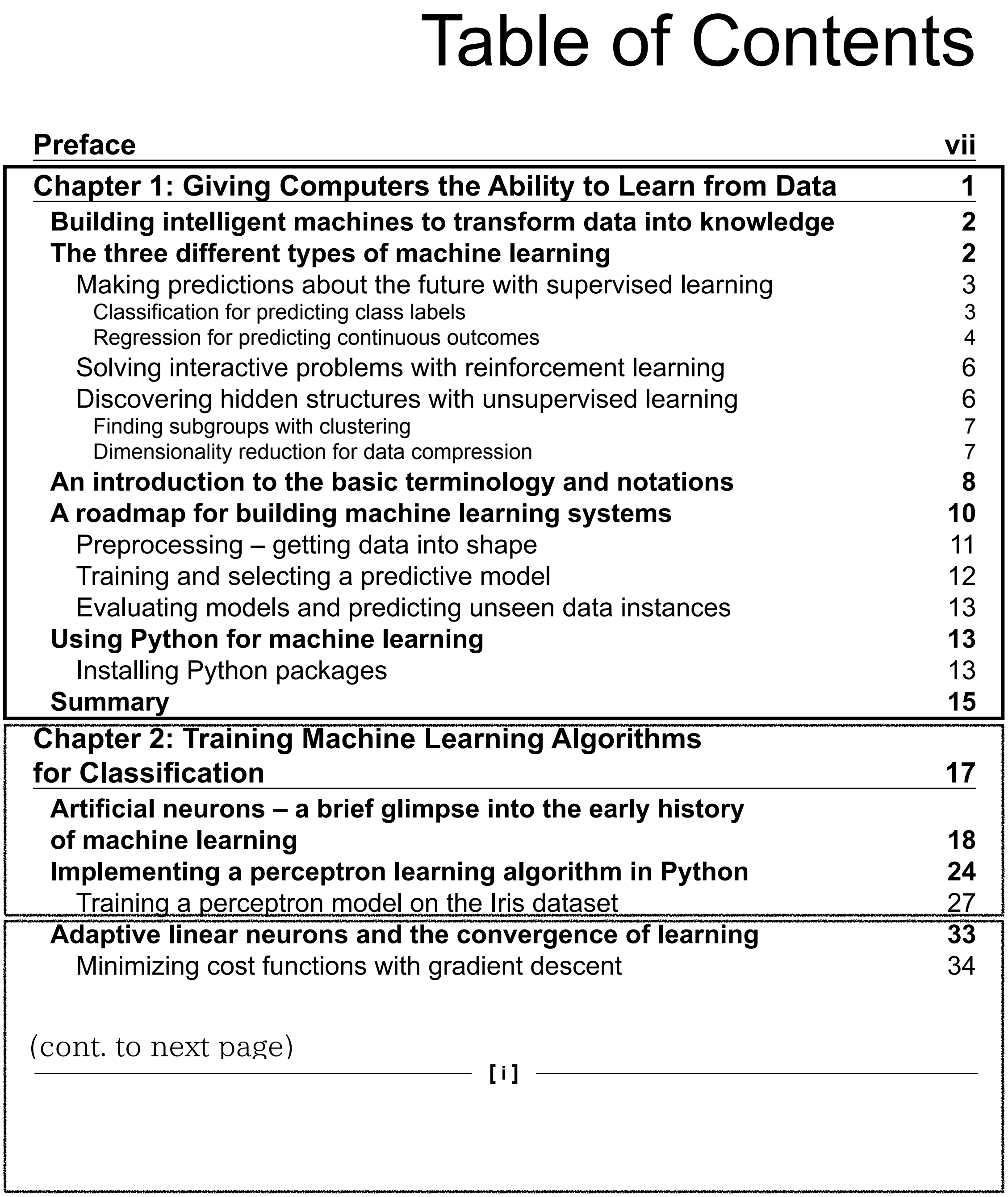
“Python Machine Learning”
12 assignments: from chapter 1 to 6.



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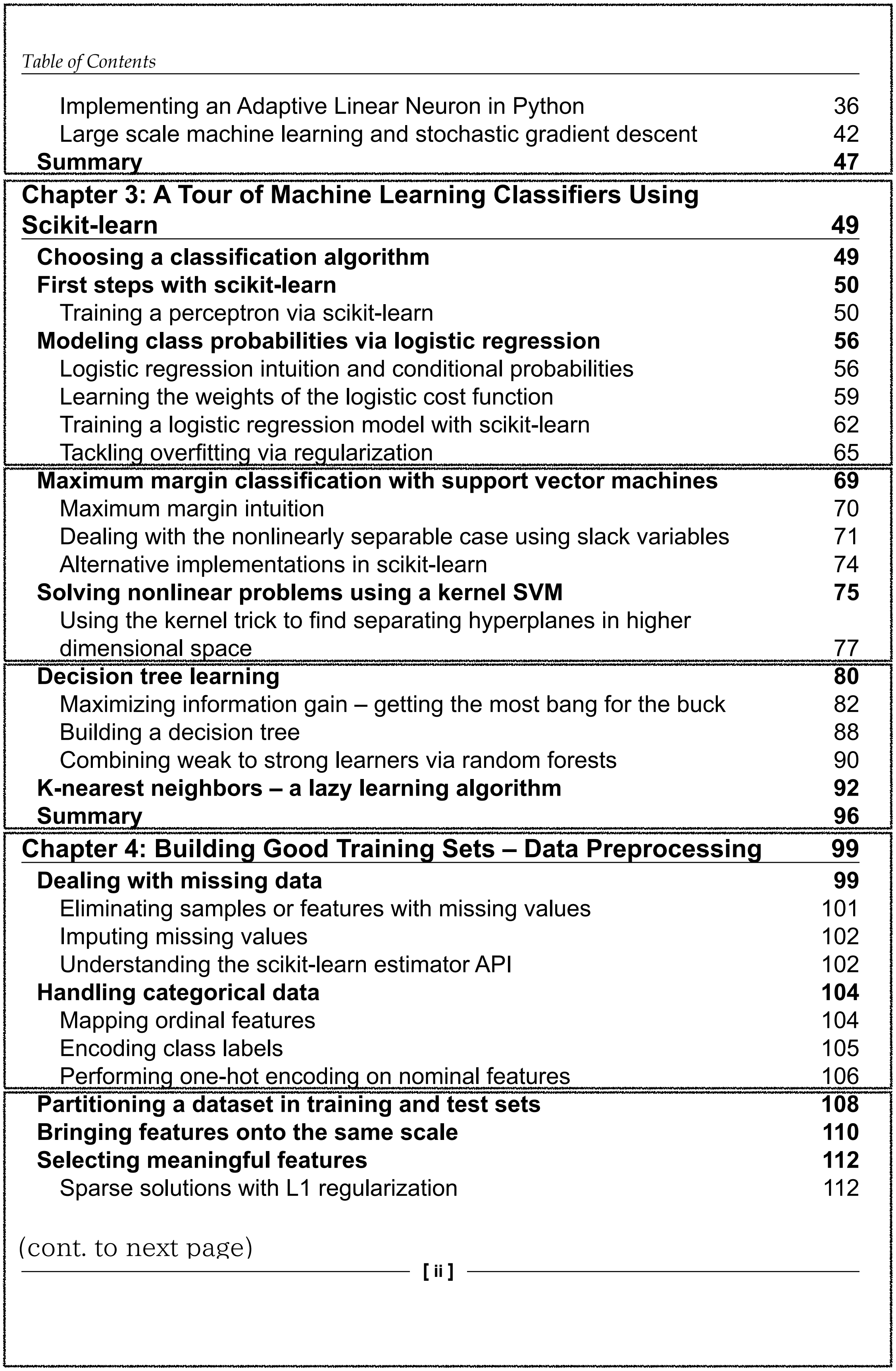

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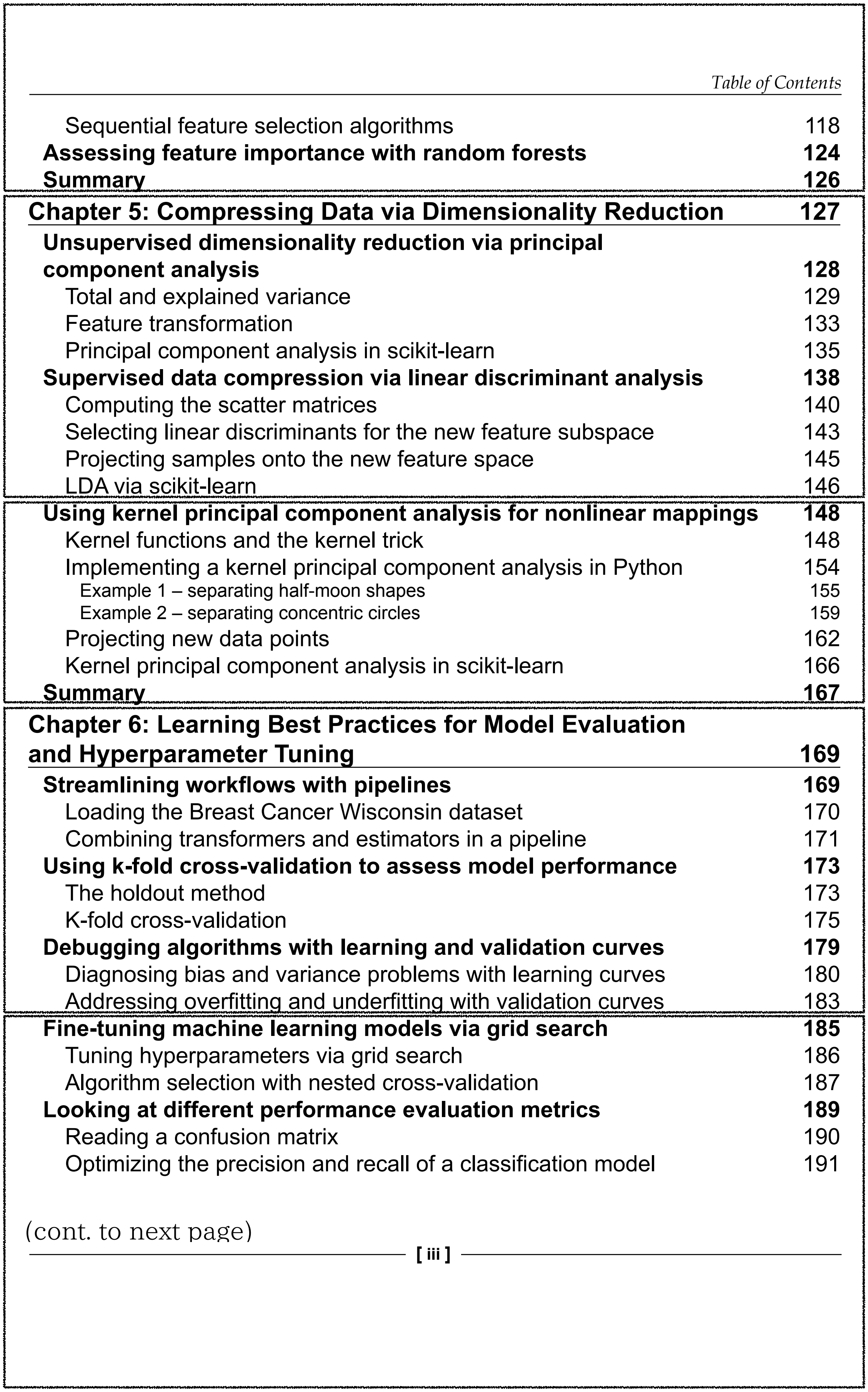

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