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978-0-521-86092-5 - Bayesian Inference for Gene Expression and Proteomics

Edited by Kim-Anh Do, Peter Müller and Marina Vannucci

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