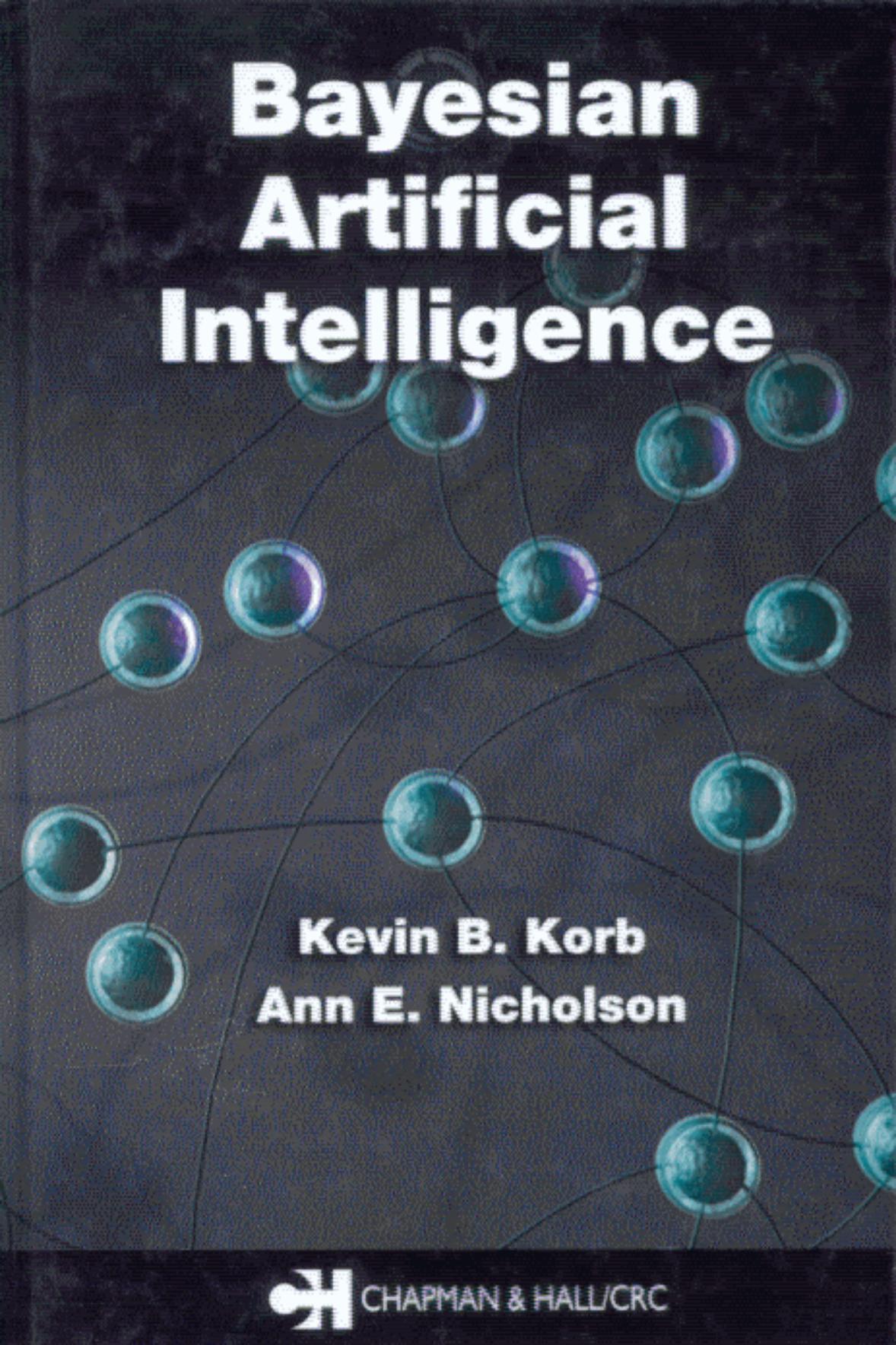


Bayesian Artificial Intelligence

A network graph with nodes represented by blue spheres connected by black lines. The nodes are arranged in a roughly circular pattern, with some lines connecting nodes in the same ring and others crossing between rings.

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CHAPMAN & HALL/CRC

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