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Sommario/riassunto

This book focuses on Explainable AI (XAI) concepts, tools, frameworks, techniques and applications. It introduces knowledge graphs (KG) to support the need for trust and transparency into the functioning of AI systems, and shows how Intelligent applications can be used to greater effect in finance and healthcare.
