



The Healthcare Intelligence Graph

Designing Adaptive and Interconnected Clinical Platforms

Triveni Kolla



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DeepScience

Published, marketed, and distributed by:

Deep Science Publishing, 2026
USA | UK | India | Turkey
Reg. No. MH-33-0658412
www.deepscienceresearch.com
editor@deepscienceresearch.com
WhatsApp: +91 7977171947

ISBN: 978-93-7185-331-6

E-ISBN: 978-93-7185-978-3

<https://doi.org/10.70593/978-93-7185-978-3>

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Citation: Kolla, T. (2026). *The Healthcare Intelligence Graph: Designing Adaptive and Interconnected Clinical Platforms*. Deep Science Publishing. <https://doi.org/10.70593/978-93-7185-978-3>

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Preface

Healthcare has always been a domain of relationships between clinician and patient, between diagnosis and treatment, between institutions and populations. Yet for decades, the digital systems built to support those relationships have done the opposite: fragmenting data into silos, isolating departments behind incompatible protocols, and forcing care teams to reconstruct the full picture of a patient's journey from disconnected fragments.

This book emerges from a conviction that the next decisive shift in clinical technology will not be a faster database or a smarter algorithm in isolation, it will be the *graph*: a living, interconnected representation of clinical knowledge, patient context, institutional workflow, and real-time signals, all woven into a single adaptive fabric.

The healthcare intelligence graph is not a product or a standard. It is an architectural philosophy one that treats every patient encounter, every diagnostic event, every care pathway as a node in a broader network of meaning. When those nodes are connected intelligently, patterns emerge that no single system could reveal alone: early warning signals hidden across time, care gaps invisible in point-of-care views, populations at risk obscured behind aggregate statistics.

The chapters that follow explore how to design, build, and govern such platforms from the data models and integration layers that form their foundation, to the AI and reasoning engines that make them adaptive, to the governance and interoperability frameworks that make them trustworthy. The engineering challenges are real. So are the ethical stakes. A system that connects clinical intelligence must be designed with as much care for privacy, equity, and human oversight as for performance.

This book is written for architects, clinicians, and technologists who believe that better-connected knowledge systems can measurably improve care — and who are willing to do the disciplined work of building them.

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