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Engineering Artificially Intelligent Systems

A Systems Engineering Approach to Realizing
Synergistic Capabilities

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Preface

Jay Forrester, creator of system dynamics, remarked we “live in a complex world of nested feedback loops,” involving cascading interdependencies across these loops that vary in complexity, space, and time. Many, if not most, of the current AI/machine learning algorithms, data and information fusion processes, and related methods are attempting, in software, to estimate situations in this complex world. Hence, these algorithms and processes must gracefully and efficiently adapt to the technical challenges (e.g., data quality) induced by nested feedback loops and interdependencies.

To realize effective and efficient designs of these computational systems, a systems engineering perspective may provide a framework for identifying the interrelationships and patterns of change between components, rather than static snapshots, as systems engineering methods are about seeing “wholes.” The increasing prevalence of intelligent systems within society will reveal yet more interdependencies, in that AI-enabled intelligent systems have both upstream processes impacting them and downstream processes they affect. The benefits of studying cascading interdependencies through a systems engineering perspective are essential to understanding their behavior and for the adoption of complex system-of-systems in society. Knowledge about the world is needed for multiple applications.

Models of complex situations (e.g., patterns of life) are typically attacked with reductionism. However, in the absence of modeling dynamics and evolving interrelationships, these strategies fall short. If causal models of feedback loops and interrelations can be developed, then system modeling, and the ability to estimate system dynamics, will be as correct as possible. However, real-world situations of Forrester’s “nested feedback loops” involve uncertainty and incompleteness, impeding the calculation of accurate estimates. These topics continue to be researched by the data and information fusion communities, whose processes form snapshots and estimates about component entities and situational states. However, the data and information fusion communities should also seek opportunities (where appropriate) for employing AI/machine learning beyond these component-level estimates, toward modeling synergies and dependencies within and across estimation loops to achieve maximum situational awareness.

The genesis of this book was the scholarly presentations given at the Association for the Advancement of Artificial Intelligence (AAAI) Spring Symposium, held during March 22–24, 2021, titled “Leveraging Systems Engineering to Realize Synergistic AI/Machine Learning Capabilities”. However, we decided to refocus the title of the book to Engineering Artificially Intelligent Systems to emphasize pragmatic aspects. More information about the AAAI 2021 Spring Symposium series can be found online (<https://aaai.org/Symposia/Spring/sss21.php>).

Due to the worldwide COVID-19 pandemic the symposium was held virtually. We invited leading experts from around the world to explore the effects of cascading interdependencies, real-world dynamics, and subsequent opportunities to leverage systems engineering principles to design and develop AI/machine learning and data and information fusion estimation processes that accurately represent complex world states. These

relationships are exacerbated by the unpredictability of human decision-makers, uncertainty in raw and fused data, and the large trade space of algorithm permutations orchestrated to solve given problems. The field of systems engineering brings opportunities to address and model the full range of complex, synergistic feedback loops in modern complex systems, toward the realization of cost-effective designs.

The chapters of this book are extensions of abstracts and presentations from the AAAI 2021 Spring Symposium on Leveraging Systems Engineering to Realize Synergistic AI/Machine Learning Capabilities. The topics included AI, machine learning and reasoning, data and information fusion, systems engineering, interdependence, human systems, human biases and limitations, trust and complex AI systems, and ethics. In particular, the call for papers sought contributions on the following:

- Artificial Intelligence, Machine Learning, and Reasoning
- Data and Information Fusion
- Systems Engineering of Complex Interdependent Subsystems
- Systems Engineering for Artificial Intelligence
- Artificial Intelligence for Systems Engineering
- Systems Engineering and Federated Learning
- Human Systems Integration and Visualizations
- Human Biases, Limitations, and Capabilities with Large Scale AI-based Systems
- Trust and Acceptance of Complex AI-based Systems
- Emergence in Complex System-of-Systems
- Ethics in Deploying Complex AI-based systems
- Societal Consequences of Interacting Complex AI-based systems
- Explainability and Interpretability of Artificial Intelligence
- Uncertainty Propagation in AI-based Complex Systems
- Modeling User Interactions with Large Scale AI-based Systems

More details about the symposium, including the full program, can be found on our webpage (<https://sites.google.com/view/systems-engineering-ai-ml/home>). In addition, we invited additional authors, who did not participate in our symposium, to join us by contributing chapters to this book.

In our deliberations and preparations for the symposium and the LNCS book, we were assisted by our Program Committee, some of whom became presenters at our symposium and contributors to this LNCS book. We thank them, we thank our invited speakers, and we thank our regular speakers.

August 2021

William F. Lawless
James Llinas
Donald A. Sofge
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Addendum

The Association for the Advancement of Artificial Intelligence (AAAI) Symposium in the spring of 2021, originally scheduled to be held at Stanford University, USA, was instead converted into a virtual symposium conducted over Zoom as a result of the COVID-19 pandemic. Its reduced two-day agenda consisted of eight invited speakers who gave hour-long presentations and eight regular speakers who gave half hour presentations, along with the two speakers who gave opening comments on both days. All of the symposium papers were submitted to the lead organizer via EasyChair, who conducted a review of each paper that was accepted for presentation, assisted by the co-organizers (reducing potential conflicts of interest, the lead organizer did not give a talk at the symposium). In a normal year not affected by COVID-19, we would have had more speakers participating on the regularly scheduled third day of the symposium (usually a half-day).

Of the 18 speakers at the AAAI Spring Symposium, 10 papers were expanded into chapters and submitted to the lead editor of this LNCS book. An 11th chapter contributed by the lead organizer gives an introduction, an overview of the science, and a review of the chapters contributed, which was only reviewed by the lead editor of the book. Of the 10 content chapters, four were written by invited speakers at the symposium. Each of these 10 chapters was thus triple-reviewed: for the symposium, following editing for inclusion in this LNCS book, and for review comments as part of the introductions to each chapter.

Finally, having collected only 10 content chapters, a shortfall that we attributed partly to the COVID-19 pandemic, we invited five additional authors to contribute to the LNCS book. These invited chapters were reviewed twice, once by the LNCS lead editor for editorial content, and once by the symposium's lead organizer for the introduction to each chapter. All of the chapter contributors were asked to address all of the editorial comments, which they did.

The editorial process was arduous, but it assured that the contents of the book were uniformly edited, that conflicts of interest were prevented or minimized, and that the state of the science was advanced.

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