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Preface

Human language is not the starting point for knowledge representation. Our utterances or our symbols are not the basis for what we desire to convey; they are only representations. Knowledge, the actionable side of information, is rooted in something more fundamental than language. What that something may be is what this book is about.

Competing factions have claimed truth since at least the beginning of communication. Who knows, maybe bees, whales, dingos, and apes also have communities believing different things as true, perhaps even leading to conflict. As humans, we know from wars, missed opportunities, and personal misunderstandings the tragedy that different premises of truth may bring. We have to admit if we want to represent human knowledge to computers that we humans have not done such a hot job representing knowledge to ourselves. Since we are starting out on a journey here to explore knowledge representation (KR) for knowledge management, artificial intelligence, and other purposes, more than a bit of humility seems in order.

Information, by no means a uniformly understood concept, arises from a broader context than gestures, symbols or sounds. For some, information is energy or when missing is entropy, the nuts-and-bits of messages. For some, information is meaning. That we continue to use 'information' in these senses and more, in fact, tells us these senses are properly within the boundaries of the concept. Still, even if we can clear the hurdle of grokking information, we have the next obstacle of deciphering what is knowledge, that which next lies directly on our path. Further, of course, we then need to record somehow and convey all of this if we are to represent the knowledge we have gained to others. Like I say, if we have a hard time communicating all of this to other humans, what can we say about our ability to do so to machines and AI?

But maybe I overthink this. Any tasks us humans do using information that we can automate with acceptable performance may lead to more efficiency and perhaps more job satisfaction for the workers involved. Maybe even more wealth. Conversely, maybe this automation leads to loss of jobs for the workers. I do know, however, if we are ever to rely upon machines to work on our behalf, requiring little or no oversight, then we need to figure out what this knowledge is and how to represent it to the machine. Such is the task of KR. What I try to provide in this book is a way to think and a practical guidebook of sorts for how to approach the questions of computers and knowledge.

The world is real. It exists independent of us or how we may think about it, though our thoughts are also part of our reality. Human history fills but a small thimble yet through the application of reason and truth-testing, including, since the Enlightenment, the scientific method, we humans have increasingly unveiled the truths of Nature, in the process creating wealth and comfort never before seen. Artificial intelligence (AI) will undoubtedly accelerate this trend. How fast that acceleration occurs is, in part, a function of how good we get at representing our knowledge. These representations are the encodings by which intelligent machines will work on our behalf. My quest in this treatise is to help promote this trend. I believe this quest to be noble and, in any case, inevitable. I believe there is something in our nature that compels us to pursue the path of useful information leading to knowledge.

The past decade was a golden one in advances in AI. We can now voice commands and requests to our phones and devices acting as virtual assistants. We are on the verge of self-driving vehicles and automation of routine knowledge worker tasks. Still, the deep learning that underlies many of these advances is an opaque, black box of indecipherable inferences. We don't know why some of this magic works or what the representations are upon which machines draw these inferences. For further advances to occur, for general AI or cognition to arise *in silico*, I believe we will need better ways to represent knowledge, reflective of the nature of information and its integral role in the real world.

I have had a passion for the nature and role of information throughout my professional life. I originally trained as an evolutionary biologist and population geneticist. Since my graduate days, I have replaced my focus on biological information with one based on digital information and computers. My passion has been on the role of information — biological or cultural — to confer adaptive advantage to deal with an uncertain future and as a means of generating economic wealth. My intuition — really, my underlying belief — is that there are commonalities between biological and cultural information. I have been seeking insights into this intuition for decades.

One of my first forays into information technology was a data warehousing venture, where the idea was to find ways to connect structured databases that, in native form, were standalone and unconnected. This venture coincided with the explosive growth of the initial Internet. To support the exploding content we observed that large content suppliers were populating their Web sites with searchable, dynamic databases, hidden from the search engines of that time (before Google's inception). We named this phenomenon the 'deep Web' and did much to define its huge extent and figure out ways to mine it. We saw that, in aggregate, the Web was becoming a giant, global data warehouse, though largely populated by text content and less-so by structured data. We shifted our venture emphasis to text and discovery. This shift raised the perplexing question of how to place information in text on to a common, equal basis to the information in a database, such as a structured record. (Yeah, I know, kind of a weird question.)

Tim Berners-Lee, inventor of the World Wide Web, and colleagues put forward a

vision of the semantic Web in a *Scientific American* article in 2000.* The article painted a picture of globally interconnected data leveraged by agents or bots designed to make our lives easier and more automated. The late <u>Douglas Adams</u>, of *Doctor Who* and *A Hitchhiker's Guide to the Galaxy* fame, had presciently produced a fascinating and entertaining TV program on the same topic for BBC2 about ten years earlier. Called *Hyperland*, you can see this self-labeled 'fantasy documentary' from 1990 in its entirety on YouTube. The 50-min presentation, written by and starring Adams as the protagonist having a fantasy dream, features Tom, the semantic simulacrum (actually, <u>Tom Baker</u> from *Doctor Who*). Tom is the "obsequious, and fully customizable" personal software agent who introduces, anticipates and guides Adams through what is a semantic Web of interconnected information. Laptops (actually an early Apple), pointing devices, icons, and avatars sprinkle this *tour de force* in an uncanny glimpse into the (now) future.

One of the premises of the semantic Web is to place what we now call unstructured, semi-structured and structured information on to a common footing. The approach uses the RDF (Resource Description Framework) data model. RDF provided an answer to my question of how to combine data with text. I am sure there were other data models out there at the time that could have perhaps given me the way forward, but I did not discover them. It took RDF and its basic *subject-predicate-object* (*s-p-o*) 'triple' assertion to show me the way ahead. It was not only a light going on once I understood but the opening of a door to a whole new world of thinking about knowledge representation.

The usefulness of ideas behind the semantic Web and the semantic technologies supporting it lured me to switch emphasis again. I founded a new company with Frédérick Giasson, and we proceeded to provide semantic technology solutions to enterprises over the next ten years. The Web today is almost unrecognizable from the Web of 15 years ago. If one assumes that Web technologies tend to have a five year or so period of turnover, we have gone through three to four generations of change on the Web since the initial vision for the semantic Web.

Many of our engagements were proprietary, though we did provide three notable open source projects. We developed a general semantic platform for ontology (knowledge graph) and data management, the still-active <u>Open Semantic Framework</u> project. To help information interoperate, we created <u>UMBEL</u>, a subset of Cyc and a contributor to our current efforts, as a set of reference concepts that users can share across different Web datasets. Based on that experience, we designed a successor reference knowledge structure, <u>KBpedia</u>, a combination of upper knowledge graph and leading public knowledge bases. We talk much about KBpedia throughout since it is this book's reference knowledge structure.

The marrying of electronic Web knowledge bases — such as <u>Wikipedia</u> or internal ones like the Google search index or its <u>Knowledge Graph</u> — with improvements in <u>machine-learning algorithms</u> is systematically mowing down what used to be called the <u>Grand Challenges</u> of computing, such as machine translation or language understanding. Sensors are also now entering the picture, from our phones to our homes

^{*} Berners-Lee, T., Lassila, O., and Hendler, J., "The Semantic Web," Scientific American Magazine, 2001.

and our cars, that exposes the higher-order requirement for data integration combined with semantics. Natural language processing (<u>NLP</u>) kits have improved in accuracy and execution speed; many semantic tasks such as tagging or categorizing or questioning already perform at acceptable levels for most projects. We naturally call the marriage of these knowledge sources with AI 'knowledge-based artificial intelligence.' KBAI is one of the potential payoffs that would arise from better ways to represent knowledge, and thus is a common theme throughout the book.

Combining information goes beyond the technical challenges of matching forms and formats. We need to tackle the question of meaning, inextricably entwined with context and perspective. Cinemaphiles will readily recognize <u>Akira Kurosawa's</u> <u>Rashomon</u> film of 1951. In the 1960s, one of the most popular book series was <u>Lawrence Durrell's</u> <u>The Alexandria Quartet</u>. Both, each in its way, tried to get at the question of what is the truth by telling the same story from the perspective of different protagonists. Whether you saw Kurosawa's movie or read Durrell's books, you know the punchline: truth is very different depending on the point of view and experience — including self-interest and delusion — of each protagonist.

All of us recognize this phenomenon of the blind men's view of the elephant. The problem we are trying to solve is how to connect information meaningfully. For that, we need to somehow capture the ideas of perspective and context, as well as the usual vagaries of imprecise semantics. Root cause analysis for what it takes to achieve meaningful, interoperable information suggests one pivotal factor is to describe source content adequately in context to its use. Capturing and reflecting context is essential if we are to get information sources to work together, a capability we give the fancy label of 'interoperability.' We also need to assemble and represent this information such that we can reason over it and test new knowledge against it, a structural form we call a 'knowledge graph.' All of this requires a logical and coherent theory – a grounding – for how to represent knowledge.

Our client efforts over the past decade were converging on design thoughts about the nature of information and how to signify and communicate it. The bases of an overall philosophy regarding our work emerged around the teachings of <u>Charles</u> <u>Sanders Peirce</u> and <u>Claude Shannon</u>, each explicating one of the boundary senses of information. Shannon emphasized the message and mechanical aspects of information; Peirce emphasized meaning in both breadth and depth. In the combination, we see semantics and groundings as essential to convey accurate messages. Simple forms, so long as they are correct, are always preferred over complex ones because message transmittal is more efficient and less subject to losses (inaccuracies). How we could represent these structures in graphs affirmed the structural correctness of our design approach. The now visible re-awakening of artificial intelligence helps to put the semantic Web in its proper place: a key subpart, but still a subset, of AI.

I first encountered Charles S. Peirce from the writings of John Sowa about a decade ago. Sowa's writings are an excellent starting point for learning about logic and ontologies, especially his articles on Peirce and signs.* Early on it was clear to me that knowledge modeling needed to focus on the inherent meaning of things and

^{*} Use https://www.google.com/search?as_q="peirce"&as_sitesearch=jfsowa.com for a listing.

concepts, not their surface forms and labels. Sowa helped pique my interest that Peirce's <u>theory of semiotics</u> was perhaps the right basis for getting at these ideas.

In the decade since that first encounter, I have based some writings on Peirce's insights. I have developed a fascination with his life and teachings and thoughts on many topics. I have become convinced that Peirce — an American philosopher, logician, scientist, and mathematician — was possibly one of the greatest thinkers ever. While the current renaissance in artificial intelligence can certainly point to the seminal contributions of <u>George Boole</u>, Shannon, <u>Alan Turing</u>, and <u>John von Neumann</u> in computing and information theory (among many others), my view, not alone, is that C.S. Peirce belongs in those ranks from the perspective of <u>knowledge representation</u>, the *meaning* of information, and hewing to reality.

The importance of studying Peirce for me has been to tease out those principles, design bases and mindsets that can apply Peircean thinking to the modern challenge of knowledge representation. This knowledge representation is like Peirce's categorization of science or signs but is broader still in needing to capture the nature of relations and attributes and how they become building blocks to predicates and assertions. In turn, we need to subject these constructs to logical tests to provide a defensible basis for what is knowledge and truth given current information. Then, all of these representations need to be put forward in a manner (symbolic representation) that is machine readable and computable.

In reading and studying Peirce for more than a decade, it has become clear that he had insights and guidance on every single aspect of this broader KR problem. My objective has been to take these piece parts (Peirce parts?) and recombine them into a whole consistent with Peirce's *architectonic*. How can Peirce's thinking be decomposed into its most primitive assumptions to build up a new KR representation? These are the points I argue in the book, while also sharing the experience of how we may integrate these viewpoints into working knowledge management systems.

I have no intent for balance in this exposition. There are wonderful textbooks and handbooks available if you are seeking a neutral presentation on knowledge representation in computer and information science. The lens I use is strictly that of Peirce and his views that contribute to an understanding of knowledge representation, at least how I read and understand those views. Peirce further guides the scope and organization of this book. One of Peirce's signal contributions was the philosophy of pragmatism, according to a specific maxim and a recommended methodology to follow, what the Peirce scholar Kelly Parker calls a '*practionary*.' To my knowledge, this book employs this Peircean methodology for the first time. Given this emphasis, we will by necessity need to tackle many Peircean concepts, some with arcane or jaw-breaking labels. That is a small price to pay to gain entry into his brilliant insights.

I also minimize math and equations in the book. I provide many salient references for exploring topics further. I try to emphasize how to think and organize. I avoid cookbook steps or prescriptive techniques or methods. I do not recommend specific tools. Rather, because of the coherence of Peirce's views, I use how I understand him and his writings, including interpretations by others, hopefully, to bring a consistent approach, logic, and mindset to the question of knowledge representation. By straddling today's two separate worlds of Peirce scholarship and knowledge representation I perhaps risk disappointing both camps. One of my points, though, is that the camps should be separate no longer.

I would first like to thank my colleague and partner, Frédérick Giasson, for his creativity and effort in our commercial ventures over the past decade. He was not only the implementer of the many systems we developed, and a constant fount of ideas and innovation, but a great friend and a calm and cool influence during those engagements. Though I am the recorder of the results in this book, he deserves cobilling for why and how this book came into being.

I want to thank those who have encouraged me over many years to write this book, including from many commenters on my *AI3:::Adaptive Information* blog. I especially thank Fred, Steve Ardire, Alianna Maren, Alan Morrison, Gary Richmond, Amit Sheth, and Peter Yim for their encouragement. I further thank Amit for his kind efforts to help me find and secure a publisher.

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I find it wondrous that the human species has come to learn and master symbols. That mastery, in turn, has broken the shackles of organic evolution and has put into our hands and minds the very means and structure of information itself. The *lingua franca* for doing so is knowledge representation, best done, I believe, following the guidelines of Charles Sanders Peirce.

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