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2 **Editorial**

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5 **Human reasoning is heterogeneous—as Jon Barwise informed us**

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15 It is sad but true that most people in AI and related fields, upon hearing the word
16 ‘reasoning’, imagine a sequence of purely linguistic expressions which follow
17 standard rules of deductive inference for elementary two-valued logic. (Other
18 similarly one-dimensional schemes may come to mind at the mention of this word.
19 For example, so-called ‘Bayesian reasoning’ is probabilistic, but relative to the issue
20 at hand, this is of no help because, compared with the human case, probabilistic
21 formalisms are also thoroughly one-dimensional; they make no use, for example,
22 of diagrams or other pictographic representations, or of semantic models.)
23 Human reasoners greatly exceed such rigid inflexible modes of reasoning.

24 The present issue is devoted to taking seriously the brute fact that human
25 reasoning is ‘heterogeneous’; it involves not just declarative formulae of the classical
26 sort, processed in the classical way, but also diagrams, images, models, underlying
27 semantic relationships between propositions (e.g. intuitive similarity), etc., and
28 non-deductive procedures (e.g. abduction) for processing such things. In addition,
29 when (untrained) human reasoning involves linguistic information, it often departs
30 radically from the canon of what is normatively correct reasoning over such standard
31 information, and the departure is sometimes very effective for the particular task
32 at hand.
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34 Johnson-Laird has long held that human reasoning extends well beyond standard
35 logic, and he stands as a seminal figure in the history of heterogeneous reasoning, as
36 it is uncovered and studied via empirical techniques, and rendered at least to some
37 degree in computational form. According to his mental models theory (which by
38 now is supported by a large amount of empirical data), logically untrained people
39 predominantly reason not over formulae or their relatives (e.g. declarative sentences
40 in some natural language), but rather over ‘mental models’. His paper explains the
41 mental models theory in connection with spatial reasoning, and shows that this
42 theory predicts something that some other contributors to the volume have
43 presupposed—diagrams facilitate human reasoning.

44 Although mental models theory appeared on the scene long ago, another
45 scheme (minus experiments in psychology that support mental models) predates
46 Johnson-Laird’s theory by many years: Peirce’s alpha, beta, and gamma systems
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49 for reasoning in ‘visual’ form. Van Heuveln offers an interesting argument for the
50 view that Peirce’s approach, although devised long ago, captures a blending of
51 two currently competing schools of thought in the psychology of reasoning:
52 the aforementioned mental models theory, and so-called mental logic theory, which
53 holds that untrained human reasoners abide by rules of inference over formulae—
54 just not all the rules of inference that are in standard elementary extensional
55 logical systems.

56 Our own contribution presents a method, originally devised by Yang and fine-
57 tuned a bit by Bringsjord, for cracking the kind of reasoning problems seen on the
58 Graduate Record Examination (GRE) in a manner which, like van Heuveln’s
59 account of Peircean reasoning, integrates mental models and mental logic. This
60 method makes use of the semantics of modal logic, is decidedly heterogenous, and in
61 general flows from a new theory of human reasoning known as ‘mental meta-logic’.
62 If such methods are in fact followed by humans taking the GRE (and there is some
63 empirical evidence that that is the case), then even when tackling tightly defined
64 reasoning problems on ‘high-stakes’ standardized tests, they far exceed the old-style
65 form of reasoning seen in elementary logic.

66 Three of the papers in this issue (by Sun, by Wang and Hofstadter, and by
67 Wang, Johnson, and Zhang) are linked by seeking to model heterogeneous human
68 reasoning in working computer models.

69 Sun explores the difference between rule-based reasoning (which corresponds to
70 standard deductive inference in elementary logic) and similarity-based reasoning, a
71 phenomenon (in which the underlying semantic categories referenced by the relevant
72 arguments affect judgments regarding whether these arguments are valid) which is
73 well supported by empirical data. He explains how the impressive CLARION
74 cognitive architecture can encompass these and other forms of (as he puts it)
75 ‘everyday’ reasoning in humans.

76 Wang and Hofstadter describe a fascinating reasoning system known as NARS,
77 designed to model categorization. NARS is heterogeneous across both inference type
78 (syllogistic inference and an interesting form of parallel inference), and representa-
79 tion. Although NARS is said to be a ‘reasoning system’, it exceeds the reach of
80 the narrow interpretation of that phrase, and lies closer to a full-blown cognitive
81 architecture.

82 Wang, Johnson, and Zhang explore the order effect in belief revision. Belief
83 revision is a process in which one revises one’s current belief in the light of new data,
84 and is commonly thought to be an essential component of human abductive
85 reasoning. The order effect is seen when the final belief is significantly affected by
86 the temporal order of information presentation. The authors explore the effect
87 in traditional experimental fashion and via the UECHO cognitive architecture.
88 This two-pronged approach appears to show that, at least under certain conditions,
89 the order effect results from one’s coherently and dynamically adaptive expectations
90 of the statistical properties of the environment.

91 Stenning and Gresalfi show, through examination of data obtained in connection
92 with students tackling certain classroom-level tasks involving mathematics and
93 biology, that heterogenous representation and reasoning is irrepressible, even in
94 contexts that at least seem to be based on only homogeneous linguistic processing.
95 Stenning and Gresalfi not only mention the Hyperproof system, in good part a
96 reflection of Jon Barwise’s conception of heterogeneous reasoning, but also provide

97 a fascinating argument for the view that there is no fundamental ‘interlingua’ into
98 and out of which the diverse representation and reasoning systems used by humans
99 come and go. This view is one that Barwise explicitly affirmed.

100 Note that this ‘heterogeneous reasoning’ issue is dedicated to none other than
101 K. Jon Barwise, arguably (to this point, at any rate) *the* seminal figure in the study of
102 formal heterogeneous reasoning. While much of Barwise’s contributions to logic,
103 mathematics, and the rigorous study of information will probably never be forgotten
104 among relevant experts, and while we stand in awe of much of this work, we point
105 out that Jon brought heterogeneous reasoning to life in the form of working computer
106 programs that could be profitably used even by students in introductory logic
107 courses. One of us (SB) began his study of infinitary logic (in connection with the
108 foundations of AI) with his reading of Barwise, but it was that first proof built
109 in Hyperproof that was truly unforgettable. Here was a system in which proofs
110 incorporated visual reasoning in a real way, *and* a natural way. The present issue
111 contains a fascinating vision, expressed by John Etchemendy and Dave Barker-
112 Plummer, of how Hyperproof can be generalized. John and Dave were collaborators
113 with Jon Barwise and others in the development of Hyperproof and the textual
114 material associated with it, and their generalization will no doubt carry his work
115 forward into the future.

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