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# I Can't Get No (Boolean) Satisfaction: A Reply to Barrett et al. (2015)

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Sometimes history can be philosophically interesting. Barrett (2011) and colleagues (e.g., Barrett et al., 2015) are to be congratulated on widening the scope of our understanding of animal cognition to include its ecological elements. However, in their eagerness to overturn a narrow model of computation, she and her colleagues have glossed over some rather interesting and salient historical facts. This is poignant, as these facts strengthen their case, and sharpen the focus on the more complete picture of ethologically valid cognition that they are drawing.

The key figure missing from the usual historical narrative is George Boole, whose bi-centenary has just passed and (it just so happens) is the luminary whose soon-to-be-restored home is visible from the office where I type this, in the University he led, and on the machine that his insights made possible.

Barrett (2011) wants to draw a distinction between computation—in a narrow sense—abstracted from any particular setting, and the highly embodied—especially ecologically rooted—cognition that she sees in the animals she studies.

In support of this distinction, she cites Searle's (1990) claim that, as a matter of history, humans tend to use their most impressive piece of technology as a metaphor. As exemplars, the ancient Greeks used models of torque-powered siege devices. Mettrie's (1748) *L'Homme Machine* used images of clockwork brains, Freud's libidinous mind was powered by hydraulic instincts, and so on (see Dungan, 2001 for a more extended discussion).

But, as an important historical fact the order of technology-then-metaphor is the other way round in respect of the computational model. Thinking about thinking—specifically Boole's thinking about thinking—came long before the technology did. The technology grew out of it. Thus, it's less true to say that computers are a metaphor for thinking, than that thinking is a metaphor for computation.

One important difference that modern computers have from the “technology as metaphor” pattern is that in none of the other cases have advances been made in the technology as a result of the comparison. Fountains, hydraulics, and clockwork did not become more sophisticated by reflecting on their mind-like properties. On the other hand, artificial intelligence has advanced considerably—to the point where it might be said, without hyperbole, that AI is in many cases the proof that psychology as a science is advancing. When we can formalize an information processing subsystem we can mechanize it. The fact is that we now live in a world where cars drive themselves, airplanes land themselves, and face recognition software finally works.

Deep Mind is living (!) proof that that the Rescorla and Wagner (1972) model of conditional learning works and this is not a unique example (Van Hasselt et al., 2015). The human mind isn't a computer (Searle is right about this) but it does have thousands of computable functions and we are making progress in understanding them. Will there be anything left over when we have solved all these so-called easy problems of Chalmers (1996)? It is too early to say. However, one thing that won't be left over is the ecology. Barrett et al. (2015) have seen to that, by drawing attention to the fact that said functions will be incomplete unless put in ecological (e.g., locally adaptive) contexts. And that's progress, but it is still functionalist progress. Indeed—it's a justly celebrated advance on the Gibsonian programme of embodied functional analysis of cognition. But—it is not

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115 less functionalist for all that. It turns out that the details of being  
 116 an adapted organism (functioning in its ecology) cannot be fully  
 117 abstracted into discrete disembodied modules fully specifiable in  
 118 terms of brains alone. This might lead some to prematurely think  
 119 that functionalism has met its nadir, but this would be a mistake.  
 120 Before I get to why this is I need to say a few things about the  
 121 Boolean programme that underlies the functionalist revolution  
 122 in cognitive science.

123 For an exhaustive exegesis of Boole's work here the authority  
 124 is (Corcoran, 2003), but the key ideas are quite accessible.  
 125 Boole's basic insight receives its fullest expression in *The Laws*  
 126 of Thought (Boole, 1854) and this is an attempt to draw in  
 127 all human cognition (it was never about just mathematics)  
 128 together in terms of the deep underlying logical structure in  
 129 the most abstract form possible, while still being recognizable  
 130 at a syntactic level—this level being instantiated (in computers)  
 131 in terms of logic gates. Formalizing cognition was itself the  
 132 process which allowed physical computers to be eventually  
 133 possible.

134 The major later figures in this development are well known.  
 135 They include (but are not limited to) Claude Shannon, whose  
 136 1947 master's thesis ushered in modern information theory,  
 137 through Alan Turing whose 1950 paper offered a principled way  
 138 to instantiate a machine that could compute any computable  
 139 function (Turing, 1950). John von Neumann's complex proof of  
 140 how any machine is really a representation of a function (and  
 141 might thereby replicate itself) was also an important landmark,  
 142 in 1966. Although all of these papers had important practical  
 143 outcomes and were (non-accidentally) made by people with  
 144 engineering connections, they were not "how to build" papers.  
 145 They were concerned with the formal ways to represent cognition  
 146 at the most basic level appreciable by human beings. Note that  
 147 this is not the same as saying that this is the only level they exist  
 148 at. Those formalizations resulted in physical objects—such as the  
 149 one I am typing this on—but the causal arrow was not from object  
 150 to concept. Computers (such as the ones used to crack the Enigma  
 151 codes) existed by the time of Shannon, Turing, and others but the  
 152 foundational functionalist work had been done a century before  
 153 by Boole. Thus, it is strictly illegitimate to say that functionalism,  
 154 as a strategy for decomposing thought, relies on the computer  
 155 metaphor. The functionalism came first.

156 So much for history. Are there independent reasons for  
 157 thinking that the functionalist programme is not to be lightly  
 158 set aside? Indeed there are, but here I will only mention a few  
 159 relevant to Barrett et al. (2015) general programme, which I  
 160 should stress, are not things that they necessarily deny.

161 It's commonly asserted that the computational metaphor is  
 162 about the formal manipulation of symbols (Searle, 1990). But  
 163 this is a half-truth. At one level, a level that makes semantic  
 164 sense to a human observer, computers manipulate symbols. But  
 165 mainly what they do is turn logic gates on and off really fast.  
 166 And no human observer would be able to make any sense of that  
 167 at the speeds that it occurs in a modern computer. Of course,  
 168 if you delve deeper still what we have in the computer is bits  
 169 of information, and witnessing that wouldn't convey anything  
 170 much that an unaided human observer could make meaningful.  
 171 Indeed, the (physical) computer is itself the aid. Boole's key

172 insight was to analyse the logic of human cognition at the mid-  
 173 level and realize that this level could be formalized. And once  
 174 something can be formalized it can be mechanized. And the proof  
 175 that he was right is the tasty pudding of modern computing—  
 176 which undeniably works, or you would not be reading this.

177 Does a modern desktop computer (or any computer for that  
 178 matter) replicate human consciousness? Of course it doesn't. But  
 179 the formalization of human cognition is a different matter—the  
 180 computer comes along almost as a by-product of the attempt to  
 181 do that (albeit a by-product that demonstrates that we must be on  
 182 to something).

183 It might be objected that humans do not naturally think in  
 184 terms of logic gates. And this is true, but hardly to the point.  
 185 We are typically unconscious of the underlying computational  
 186 structure of things that come naturally to us. Most of us are  
 187 unconscious of the grammar of our native tongues unless it is  
 188 formally taught to us, and it is entirely unnecessary to learn  
 189 the formal grammar of a language to be able to converse in it.  
 190 Nevertheless, the formal grammar lays bare the structure of that  
 191 language.

192 A follow-up objection might be that, while it is admitted that  
 193 Boole laid bare the formal elements of some aspects of human  
 194 thought, there are others left untouched. This may well be true  
 195 and if it is true then the attempt to build upon his insights with  
 196 formal instantiations of computation into physical systems that  
 197 replicate human thought will be forever doomed. Once again—it  
 198 is too early to tell.

199 One further common mistake is to note that humans  
 200 aren't conscious of these sorts of processes. Cognition is not  
 201 consciousness.  [Vec \(2000\)](#) drew insightful attention to  
 202 precisely this fact. He noted that the tasks that required very  
 203 smart humans to perform (e.g., diagnose disease, fly airplanes,  
 204 play chess) were comparatively trivial to automate (incidentally—  
 205 this doesn't imply that the automated version completely  
 206 captures the path of human cognition to achieving them). At the  
 207 same time, it proved very hard to automate things that to humans  
 208 were trivial, such as climbing stairs and recognizing faces. The  
 209 solution to this paradox is that evolutionarily ancient processes  
 210 do not need to draw on novel conscious elements. But—and this  
 211 is the crucial point—they are nonetheless cognitive functions for  
 212 all that.

213 Computational modeling is rooted in the realization that all  
 214 observations reveal detectable differences. These are information.  
 215 If a set of these can be meaningfully grouped into a system  
 216 then a change is a state change, and any regularities in such  
 217 changes describe a computational—that is a functional-system.  
 218 Thus, computation would exist even if computers didn't—this is  
 219 where critiques like those of Searle's (1990) miss the point. The  
 220 fact that an existing physical computer is, as he puts it, "just a  
 221 hunk of junk" is neither here nor there. Once the system can  
 222 move between states and store them it's a Turing machine, Post  
 223 machine, or Lambda calculus (Church, 1936)—which for these  
 224 purposes don't have any significant differences between them. All  
 225 such functional states are computational states—defined by the  
 226 moving from one state to another. Knowledge—and it doesn't  
 227 matter here if we are talking about humans, other animals, or  
 228 even plants, is therefore the acquiring of usable local regularities.

An ecology, in other words. Evolution has produced systems that predict things about their environments (brains) that sometime hang out together in social groups. But all of these things are computational states—and adding ecology to the complete picture does not change this fact. Indeed, it deepens it by showing how affordances must be part of the complete functional picture. Indeed, as Barrett et al. are showing, the minimalist bet of some branches of cognitive science—e.g., that we could completely capture the functionalist understanding of the organism without seeing the details of the system it lives in, may well turn out to be false. It turns out that we do need to understand how an organism responds to affordances, that the functional details of perceptual organization matter, and so forth.

But, since we are all functionalists, we really have very good reason to all get along. If it really is functionalism all the way down—then there is no radical split to be had between functional models and the ones Barrett et al. (2015) espouse. What she and her colleagues have done is draw attention to the need for (computational) systems to be closely connected to their

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ecologies. Specifically, that perception and cognition indeed need to be closely related (Barrett, 2011, p.22). It might be noted that, in this, she echoes the call of Brooks (1990) whose use of the concept of subsumption layers reminds us that one way to escape the representational issue in artificial systems is to make the system use the real world as its model and in this they offer a much needed route to allow affordances to enter into the modeling. Functionalism isn't just the only game in town. It's the only game in any ecology.

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