The Evolution of Cognitive Search

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ABSTRACT

18 Search can be defined as an attempt to arrive at a goal at an unknown location in the 19 physical environment, time, memory or any other space. Search is necessary because the 20 quantity and quality of resources essential to survival and reproduction vary in space and 21 time. In addition to exploration through actual body movement in their environment, 22 animals also search their external information space through selective allocation of 23 attention and their internal information space in order to retrieve relevant items from 24 memory. Here we integrate data on search in three distinct domains, physical movement, 25 attention to external information, and locating items in memory, in order to highlight the 26 remarkable similarities among these three domains. First, resources in all three domains 27 are typically distributed in patches. Second, in each of the three domains, animals 28 typically keep searching in patches where they have recently found resources and leave 29 areas where they have either not found or depleted resources. Third, the neurobiological 30 mechanisms modulating the exploration for and exploitation of resources in all three 31 domains involve dopamine and, in many vertebrates, regions of the prefrontal cortex and 32 basal ganglia. We suggest that, throughout evolution, animals co-opted existing strategies 33 and mechanisms used to search their physical space for exploring and exploiting internal 34 and external information spaces. The cross disciplinary integration of theory and data 35 about search can help us guide future research on the mechanisms underlying cognitive 36 search.

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INTRODUCTION

41 Search is one of the most fundamental of all organismal behaviors. Bacteria seek out 42 essential nutrients and steer clear of noxious compounds (Koshland 1980; Eisenbach & 43 Lengeler 2004), plant roots search for water and essential nutrients (Hutchings & de 44 Kroon 1994; McNickle et al. 2009), and the protozoan *Paramecium* shows in addition to 45 chemotaxis also thermotaxis, geotaxis and thigmotaxis (movement in response to touch) 46 (Jennings 1906; Saimi & Kung 1987). In general, organisms that move can search for 47 optimal abiotic settings such as temperature, humidity and sunlight, and the best places 48 for finding essential nutrients, avoiding danger and securing sexual partners.

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50 In addition to physically moving through the environment, animals may search within the 51 information space for cues indicating relevant resources. The information space may be 52 external, for example, requiring the direction of attention in pursuit of cues that signal 53 prey, or the information space may be internal, for instance, requiring the directed 54 activation of memory. Whether involving physical movement or not, search involves 55 navigating some space in pursuit of resources. That is, an individual has to decide 56 whether to move (its body or its attention) or stay where it is, and, if it moves, where it 57 should move to. In the domain of physical space, such search problems have been studied 58 extensively in behavioral ecology (Stephens et al., 2007; Stephens & Krebs, 1986). 59 Research on information search, in both external and internal environments, is 60 developing rapidly (Fu & Gray, 2006; Hills & Hertwig, 2010; Pirolli, 2007; Wilke et al., 61 2009).

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63 In this chapter, we are interested in drawing attention to the potential evolutionary 64 parallels between search across different domains, both external and internal. Why might 65 search in external and internal domains be related in an evolutionary sense? There are 66 three potential forms of evidence. The first is that the neurobiological mechanisms that 67 guide search in different animals may be functionally homologous, deriving from a 68 common ancestral function that was also used to solve search-related problems. The 69 second is that different environments may pose similar kinds of problems for search, 70 generally involving navigating heterogeneous resource distributions to find locations 71 containing resources that maximize fitness. The third form of evidence is that the 72 underlying search strategies share similar characteristics across different environments 73 and domains.

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75 In what follows, we first provide a definition of search, and then we briefly review the 76 three characteristics of environmental structure, search strategies, and neural mechanisms 77 involved in search tasks in external and internal domains. Our review will begin with the 78 physical movement of individuals in space, followed by allocation of attention to external 79 cues, and conclude with search in memory.

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82 What do We Mean by Search?

83 Search can be defined as an attempt to arrive at a goal at an unknown location in the
84 physical environment, time, memory or any other space. Finding a resource typically
85 involves at least two components: an exploration phase that investigates possible

locations where the resource might be located, and an exploitation phase that involves
resource acquisition. Often, the exploration and exploitation phases are not mutually
exclusive, as animals may sample and exploit during exploration and continue exploring
while exploiting.

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91 Because exploration typically takes time away from exploitation, the modulation between 92 the two represents an optimal control problem in which organisms attempt to minimize 93 the time spent exploring for resources but still acquire sufficient information to maximize 94 resource exploitation. When the search task involves a distinct target, an individual only 95 has to choose the optimal movement strategy that would minimize the time needed to 96 find that target. Typically, however, biologically important resources show large 97 variation in quality and they vary over time and space. Thus an adaptive search usually 98 involves a fitness maximizing decision about the optimal balance between exploration 99 and exploitation. More exploration can lead to finding better resources, but less time 100 available for exploiting those resources. This trade-off between exploration and 101 exploitation is common to both external and internal search problems.

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EXTERNAL SEARCH: MOVEMENT

105 The Structure of the External Resource Environment

All organisms encounter variation in the quantity and quality of resources. In terrestrial 106 107 systems, physical factors including the topography, soil types, winds, solar radiation and 108 precipitation shape the spatial structure of temperature and availability of minerals and 109 water. These in turn generate a variable spatial distribution of plant species and of the 110 organisms associated with them. Such distribution may be either continuous or broken, 111 with the latter implying distinct patches varying in the quality and quantity of a given 112 resource each surrounded with regions lacking that resource. Further diurnal and seasonal 113 variation in abiotic factors adds temporal variation in organismal activity and 114 productivity. This combination of spatial and temporal variation in essential abiotic and 115 biotic resources means that an individual's exact location in time and space can 116 dramatically affect its fitness. Hence individuals attempt to optimize their spatial position 117 at any given time.

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119 Search Strategies in External Space

120 Confined to the question of physical movement, the central issue concerning search in 121 space is whether or not an organism should stay where it is or move elsewhere. 122 Organisms should make this decision in response to heterogeneity in the density of 123 resources in the surrounding environment. Such problems have been studied extensively 124 (Stephens & Krebs 1986; Stephens et al., 2007). One approach for examining adaptive 125 'non-random' foraging behavior involves testing for area-restricted search, which refers 126 to an individual's ability to restrict search to the local area where it has recently found 127 resources before transitioning to more wide-ranging, global exploration (Kareiva & 128 Odell, 1987). Area-restricted search is related to patch-based models of foraging, like the 129 marginal value theorem (Charnov, 1976), but is often employed when patch boundaries 130 are difficult to detect or otherwise 'fuzzy' (Benhamou, 1992; Adler & Kotar, 1999).

132 One of the most primitive forms of search transitions between local and global foraging is 133 the run-and-tumble behavior of bacteria such as *Escheria coli*. E. coli show a change in 134 behavior upon detecting increasing or decreasing food-concentration gradients (Eisenbach & Lengeler, 2004; Koshland, 1980). When such resources are increasing, E. 135 136 *coli* engage in directed 'runs' of swimming behavior using their flaggelar motor. When 137 resources are decreasing, the direction of the flaggelar motor changes, causing the 138 bacteria to tumble randomly, before engaging in another directed swim. This behavior is 139 mediated by phosphorylation cascades within the bacteria, and appears to serve as a 140 method for moving towards high concentration gradients and away from low 141 concentration gradients. Thus, bacteria show evidence of area-restricted search by 142 attempting to stay in areas with higher resource density, but move away from areas with 143 lower resource density.

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145 Figure 1 shows patterns of area-restricted search observed for several classes of animal 146 species: nematodes (Caenorhabditis elegans), house flies (Musca domestica), bumble 147 bees (Bombus bimaculatus), and humans (Homo sapiens). In each case, the central result 148 is that the animal responds to low resource densities by traveling away from them and to 149 high resource densities by staying near them. In the nematode (C. elegans) the animal 150 engages in high angled turns (or pirouettes) following recent encounters with resources, 151 but reduces its number of pirouettes as the time since the last encounter increases (Hills et 152 al., 2004). Similar patterns of increased turning in response to resource encounters are 153 observed in flies (White et al., 1984), bumblebees (Dukas & Real, 1993), and humans 154 (Kalff et al., 2010). This pattern of density contingent foraging in space is ubiquitous 155 across metazoans (Bell, 1990; Hills, 2006).

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157 Neural Mechanisms of Search in External Space

What are the neural modulators of spatial search? Despite the abundance of evidence that animals can respond to changing resource densities in space, the neural mechanisms that control this ability are not well understood. Here we focus primarily on dopamine, because other neuromodulators (e.g., norepinephrine and serotonin) are less well understood from a comparative perspective, though they are potentially critical to search and other reward-seeking behaviors (Barron et al., 2010; Cools, this volume).

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165 In nematodes (C. elegans), the modulation between local perseveration and wider-166 ranging exploration is governed, at least in part, by a relationship between pre-synaptic 167 dopaminergic neurons modulating downstream glutamatergic locomotory interneurons. 168 Higher levels of dopamine increase turning angles, while lower levels reduce turning 169 angles. Selectively killing dopaminergic neurons or applying a dopaminergic antagonist 170 (raclopride) removes the capacity for area-restricted search (Hills et al., 2004). 171 Dopaminergic mechanisms also facilitate the increased turning that fruit flies (Drosophila 172 melanogaster) show under the influence of cocaine (Bainton et al., 2000), and 173 conditioned place-preference learning in response to methamphetamine in the flatworm, 174 Dugesia japonica (Kusayama & Watanabe, 2000). In rats (Rattus norvegicus), turning 175 increases in response to agonists for dopaminergic receptors (Robertson & Robertson, 176 1986), and modulation between explorative and exploitative behaviors is mediated by 177 midbrain dopaminergic neurons (Fink & Smith, 1980). In random foraging experiments, injection of a specific antagonist for the dopaminergic receptor subtype D1 into the
nucleus accumbens of rats significantly impaired performance, measured by an increase
in wrong entries into maze arms (Floresco & Phillips, 1999).

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182 Perseveration in response to resources is known to involve a significant dopaminergic 183 component across animal phyla (for a recent review see Barron et al., 2010). In part, this 184 may be due to the relationship between dopaminergic processing and reward sensitivity. 185 Numerous observations of dopaminergic activity in response to rewards, novel and 186 aversive stimuli have been made, and have given rise to terms like 'reward detector' and 187 'novelty detector' (Salamone et al., 1997). Critically, dopaminergic neurons adjust their 188 firing rates in response to unpredicted stimuli that are associated with fitness, such as 189 appetitive and aversive stimuli (Salamone et al., 1997). Dopaminergic neurons are also 190 involved in learning to predict outcomes associated with conditioned stimuli (Ljungberg 191 et al., 1992). In vertebrates, the dopaminergic neurons most often associated with goal-192 directed behaviors are located in the thalamus, striatum, and frontal cortex. These appear 193 to work together to control goal-directed movement in physical space and the focus of 194 attention, which we discuss next.

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ATTENTIONAL SEARCH FOR EXTERNAL INFORMATION

197 The Structure of the External Information Space

As noted above, animals encounter non-random distributions of abiotic and biotic resources as they move through their physical environment. This means that the cues indicating the availability and quality of relevant resources including food, predation, potential mates and competitors also show non-random distribution in time and space. Hence individuals can rely on the spatial and temporal structure of certain information for locating resources.

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205 It is obvious that, in many species, searching involves movement in physical space but 206 the issue of search within the external information space is less apparent. Intuitively, one 207 might argue that individuals should just process all incoming relevant information. It is 208 indeed possible that some organisms with very limited perceptual ability can adopt such 209 an inclusive strategy. In animals with extensive perceptual ability, it is clearly optimal to 210 tune out all irrelevant information. Often however, the flow of relevant information 211 exceeds the information processing rate of both the sensory organs and the brain (Dukas, 212 2002; 2009) For example, in humans, only the fovea, which occupies about 0.01% of the 213 retina and 1.7° of the visual field, transmits high quality visual information. In primates in 214 general, the optic nerve transmits only approximately 2% of the information captured by 215 the retinas and only about 1% of that information is processed by the visual cortex (Van 216 Essen & Anderson, 1995). In short, an individual's sensory organs can capture only a 217 small proportion of the incoming information flow, and the rate of information capture by 218 the sensory organs far exceeds the brain's rate of information processing. This 219 necessitates a strategy for allocating attention to the most relevant cues in the information 220 space at any given time.

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222 Search Strategies for External Information

223 External information can be envisioned as a multidimensional space generated by the 224 information flow from all sense organs. At any given time, animals must choose what 225 information to attend to. This is analogous to the location choices animals make in their 226 physical space (see above). In the information space, animals should attend to the portion 227 of information flow that would have the greatest effect on fitness (Dukas & Ellner, 1993). 228 For example, when human subjects were more likely to find targets at certain angles of 229 the visual field, they devoted more attention to and had higher detection rates at these 230 angles than subjects searching for randomly distributed targets (Shaw & Shaw, 1977). 231 Similarly, human subjects tend to focus their visual attention in the vicinity of a recently 232 detected target but switch their attention to other spatial locations if no target is found at 233 this area within a short giving-up time. This behavior, which is reminiscent of area-234 restricted search, is called inhibition of return (Klein, 2000; Posner & Cohen, 1984). In 235 general, animals foraging in natural settings should focus their attention on the sensory 236 cues associated with the most profitable food and most likely danger (Dukas, 2002). 237 Whereas much of the research on attention has been done in the visual domain, auditory 238 and olfactory studies have revealed similar patterns of animals focussing on the most 239 relevant cues at any given time (e.g. Skals et al., 2005, Frits and Elhilali 2007, Cross & 240 Jackson 2010).

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242 Animals searching for resources in the physical environment often have to choose the 243 search rate (distance moved per unit time) that would maximize their rate of finding 244 resources (Dukas, 2002; Gendron & Staddon, 1983). Similarly, animals have to choose 245 their range of information processing, which should be negatively related to the difficulty 246 of processing certain information (Dukas & Ellner, 1993). That is, animals can distribute 247 attention broadly (e.g., devote little attention per unit area) when handling easy 248 information but must adopt a narrow focus of attention when handling difficult 249 information. For example, blue jays (Cyanocitta cristata) were trained to search for two 250 prey types, a caterpillar, which could appear in the center of the visual field at a 251 probability of 0.5, and a moth, which could appear in either right or left peripheries of the 252 visual field at a probability of 0.25 per side. The jays were three times more likely to 253 detect the peripheral moth targets when the central caterpillar was conspicuous, i.e., easy 254 to detect, than when it was cryptic and hence difficult to detect. This result is consistent 255 with the prediction that the jays would process information from the whole visual field 256 when the primary task is easy, but would narrow down their focus of attention to the 257 center field when the primary task is difficult (Dukas & Kamil, 2000). That is, the jays 258 modulated their focus of attention, reducing the area from which they processed 259 information when the task became more difficult.

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261 Neural Mechanisms Controlling Attention to External Information

Exactly as dopamine is a key neuromodulator of search in physical space, it plays an important role in search within the external information space. In general, dopamine is involved in subjects' ability to focus and sustain attention on relevant cues. For example, mice (*Mus musculus*) that were genetically manipulated to selectively eliminate phasic firing of dopaminergic neurons showed selective impairment in using relevant cues for learning. This suggests that the phasic firing of dopaminergic neurons modulates selective attention to relevant information (Caron & Wightman, 2009; Zweifel et al., 269 2009). In humans, subjects with a subtype of the dopamine transporter gene associated
270 with higher dopamine levels in the striatum (a region of the brain associated with
271 attention) show a different pattern of inhibition of return than control subjects (Colzato,
272 Pratt, & Hommel, 2010). This suggests involvement of dopamine in the spatial allocation
273 of attention over time.

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275 Dopamine deficit is currently the leading theory for explaining attention-deficit / 276 hyperactivity disorder (ADHD), a mental disorder characterized by a reduced ability to 277 focus and sustain attention and by an excessive level of activity. First, brain imaging 278 studies indicate smaller sizes and lesser activation of brain regions related to dopamine in 279 subjects with ADHD. Second, allelic variation in two genes, the dopamine receptor D4 280 and the dopamine transporter, has been linked to ADHD. Finally, the principal drug for 281 treating ADHD, methylphenidate (Ritalin) increases synaptically released dopamine (Iversen & Iversen, 2007; Swanson et al., 2007). Together, these examples provide 282 283 strong evidence that dopamine modulates the focus of attention to external information 284 similar to the way it modulates perseverative local foraging in external space.

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INTERNAL INFORMATION SEARCH

The previous two sections have focused on search via physical movement in the
environment and through selective tuning to external information. This section concerns
search either for information in memory or for solutions to problems that require internal
manipulation of information.

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292 The Structure of Internal Information

As demonstrated above, external stimuli often present themselves in a non-random,
spatially auto-correlated fashion—with rewards associated with a specific location likely
to signal rewards close to that location in the near future. Does memory and the structure
of the relationships between items in memory implicate an auto-correlated structure, and
do we see evidence of this structure in recall from memory?

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299 Studies of written language using network analysis find evidence for a strongly clustered 300 environment. With nodes representing words and links representating relations between 301 words, these networks often show small-world structure, indicating that in text, words are 302 much more likely to appear together in small clusters of related items than one would 303 expect by chance (Cancho & Solé, 2001). Similar small-world structure has also been 304 identified in internal search when people are asked to say the first word that comes to 305 mind after hearing another word (i.e., free association) (Steyvers & Tenenbaum, 2005). 306 Moreover, this structure of language and free association networks is well correlated with 307 the order in which children learn about language (Hills et al., 2010a). This indicates that 308 the patchy internal structure of memory may be tightly linked with the patchy external 309 structure of information.

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311 Search Strategies for Internal Information

Research on free recall from natural categories and list learning consistently find that
groups of semantically similar words are produced together (Bousfield, 1953; Romney et
al., 1993). This clustering in output is often considered to be the result of a dynamic

315 search process that modulates between local and global search policies. One of the most 316 prominent and successful memory search models, the search of associative memory 317 model, employs this dynamic local-to-global search policy (Raaijmakers & Shiffrin, 318 1981). Local search is assumed to occur via item level similarity, e.g., with recently 319 recalled items in memory activating other related items in memory. Global search activates all items in relation to the overarching category and context (e.g., "animals" in 320 321 "Say all the animals you can think of"). Transitions from local to global search occur 322 when local resources become depleted, e.g., when the participants recover too many 323 items that they have already recovered. Interestingly, this model of memory search was 324 developed in cognitive psychology independent of models in behavioral ecology, but it 325 shares the signature behavioral pattern associated with area-restricted search in physical 326 space: modulating between exploration and exploitation in response to recent experience 327 with the resource environment.

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329 Similar evidence for local perseveration due to memory activation has been found in 330 experiments based on word priming. In these experiments, a person is first shown a 331 word-prime (e.g., BIRD) and then asked to determine whether or not a second word-332 target was a true word or a nonword (e.g., ROBIN or ROLIN, respectively). Relative to 333 an uninformative word-prime, Neely (1977) demonstrated both facilitation (faster 334 response times) and inhibition (slower response times) in people's ability to determine 335 the identity of the word-target by manipulating whether the word-target was expected or 336 unexpected following the word-prime. This elegantly demonstrates that expectations 337 create local activation in memory following the presentation of a prime, and that these 338 can reduce the time it takes to recognize objects associated with those memories, but also 339 increase the time it takes to recognize objects that are not associated with those 340 memories.

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342 Research on sequential solutions in problem-solving tasks also demonstrates that people 343 show local perseveration in internal search environments. For example, people tend to 344 produce solutions that are more clustered together (i.e., similar) than one would expect by 345 random generation, for example, in math search tasks (Hills, 2010), and anagram search 346 tasks (Hills et al., 2010b). In one case, Hills et al. (2010b) had participants search within 347 scrambled sets of letters for multiple words. Participants would see a letter set like 348 BLNTAO, and they could find "BOAT", "BOLT", etc. An analysis of the string 349 similarity (e.g., bigram similarity comparing the number of shared letter pairs: 'BO', 350 'OA', etc.) between subsequent solutions determined that participants tended to produce 351 solutions that were most similar to their last solution. This was true even though previous 352 solutions were not visible. The results indicate that participants were searching locally 353 around previous solutions, before transitioning to a global search strategy (Figure 2).

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355 Neural Mechanisms in Internal Information Search

Several studies have found that the trajectories taken through long-term memory are
related to working memory span (Rosen & Engle, 1997), which is well known to be
tightly connected with dopaminergic processing (Cools & D'Esposito, 2009). Rosen and
Engle (1997) found that participants with higher working memory spans tend to produce
longer sequences of clustered items in a category fluency task than individuals with lower

361 working memory spans. Hills & Pachur (In prep.) used a social fluency task ("say all the 362 people that you know") and had participants reconstruct the social network over which 363 they were searching. Using semantic memory models, they found that participants with 364 lower working memory spans transitioned more frequently between global and local cues 365 in memory than individuals with higher working memory spans. This transitioning is 366 similar to the transition between exploratory and exploitative behavior described above 367 for spatial and attentional foraging.

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369 Cools and D'Esposito (2009) have suggested that a proper balance between prefrontal 370 and striatal dopamine levels is the key modulator of cognitive stability and cognitive 371 flexibility. Kane and Engle (2002) have suggested that it is the cognitive control of 372 attention, i.e., the ability to focus on one subgoal to the exclusion of other, distracting 373 stimuli, that is the underlying factor determining working memory span. They have 374 further suggested that this ability is mediated by prefrontal cortex modulation of activity 375 in other areas of the brain. In other words, individuals with higher working memory 376 spans are better at exploiting local information in internal search, while individuals with 377 lower working memory spans tend to leave patches of local information more readily.

PROSPECTS

380 The data we have presented above indicate three central points about external and 381 internal search. First, the environments in which organisms search both externally and 382 internally share similar structural properties—with resources tending to be patchily 383 distibuted. Second, various search strategies often rely on this patchiness to focus search 384 around areas where resources have been found recently, and thus to facilitate resource 385 acquisition based on their non-random distribution. Finally, the neural mechanisms 386 controlling search, especially those involving dopamine, the prefrontal cortex, and the 387 basal ganglia, are often shared across species and search environments.

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While the data above help us integrate information about the structure, strategies and
mechanisms of search in external and internal environments, we still lack substantial
knowledge about the cognitive ecology of search. Below we highlight three key issues
requiring further research.

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394 Physical Search as an Evolutionary Precursor of Cognitive Search

395 Might the similarity between external physical search and internal information search 396 indicate an origin for goal-directed cognition (i.e., cognitive control) from an 397 evolutionary precursor devoted to spatial foraging and feeding related behaviors? Across 398 metazoans (i.e., vertebrates and invertebrates), we find similar mechanisms modulating 399 physical search for resources (Barron et al., 2010). As outlined above, in vertebrates 400 (especially mammals) we find roughly the same mechanisms modulating search for 401 information. This suggests a potential evolutionary homology between search in physical 402 space and cognitive search, with the derived form broadening the domains of search to 403 information (Hills, 2006). What other evidence would provide support for or against this 404 hypothesis?

406 The comparative evolutionary approach to search also raises several other questions. Are 407 different forms of cognitive search domain specific or domain general? Recent research 408 demonstrated priming in humans from external to internal search (Hills et al., 2008), 409 based on empirical data indicating that prior experience in spatial foraging influenced a 410 subsequent search in an 'internal' problem solving task. In this experiment, participants 411 who first searched in a visuospatial task for clustered or diffuse resources, then searched 412 for word solutions in anagrams as if those solutions were more or less clustered, 413 respectively. This may indicate a domain general search process, consistent with our 414 understanding of executive processing in cognition as a method for navigating 415 hierarchical subgoals (Hills et al., 2010b). Are there other forms of search that might be 416 guided by such a domain general process, or by other domain specific processes (e.g., 417 mate search)?

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419 Do flexible cognitive capacities rely on balancing neuromodulation, similar to the 420 cognitive search trade-off between exploration and exploitation outlined above? Many 421 pathologies of goal-directed behavior (e.g., ADHD, Parkinson's, stereotypes in autism, 422 drug addiction, etc) involve dopamine in a way that would be predicted from the neural 423 control of animal foraging behavior, with more (or less) synaptic dopamine leading to 424 higher (or lower) levels of perseveration and attentional focus (Hills, 2006). Cools and 425 Robbins (2004) have argued that it is a balance between too-high and too-low dopamine 426 levels that generate the "optimal state of mind"-patterns of behavior associated with too 427 much or too little dopamine are consistently inflexible, often being too compulsive or 428 impulsive for the demands of the environment. This flexibility is potentially one of the 429 guiding selective forces in the evolution of the brain, as relatively larger brains appear to 430 confer greater flexibility—an observation called the *cognitive-buffer hypothesis* (Sol, 431 2009). Can we better operationalize what flexibility means in terms of searching for 432 information? And what might be the various evolutionary origins of this flexibility?

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How do search strategies evolve to conform to different environments, be they differently
distributed environments in physical space, or higher dimensional environments like
memory or external information? What other selective forces might underlie the
evolution of differences across various search strategies? Finally, what forms of evidence
would provide more insight into the evolutionary origins of cognitive search, be they
spatial or otherwise?

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441 What is the Biological Basis of Cognitive Search?

442 In our analysis of the neural mechanisms underlying search, we focused on the common 443 denominator of neuromodulation by dopamine, which, in vertebrates, is localized 444 principally in the prefrontal cortex and basal ganglia. Whereas this shared characteristic 445 of neuromodulation by dopamine is intriguing and deserves further exploration, a fuller 446 examination must also include more specific details about other brain regions, 447 neuromodulators, and patterns of neuronal firings involved in search within each of the 448 distinct spaces discussed here. Do we find further common mechanisms at this deeper 449 level of analysis? Can existing knowledge about biological mechanisms of search within 450 one domain, say selective attention in external space (Knudsen, 2007; Salamone et al.,

2007), help us understand mechanisms of search in another area, for example, retrievalfrom the internal information space?

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454 The Organization of Internal Information

455 Whereas we focused here on similarities across search environments and search 456 mechanisms, there are also some important differences. Perhaps the most significant 457 distinction between external and internal search environments is that searchers typically 458 cannot control the distribution of targets in the external environment but may affect the 459 way they store their own information. That is, natural selection may have shaped the 460 architecture of internally stored information in order to maximize some utility such as the 461 speed of recall or the numbers of items recalled. Perhaps existing models and data on 462 search in external space can help us understand the selective pressures and constraints 463 operating on the structure of internal search environments.

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466 How are the Algorithms for Search Shared Across Domains?

467 What are the other dimensions along which we can characterize search? Part of the 468 power of search as a paradigm is our ability to use search algorithms in one domain to 469 inform research in other domains. Here we highlighted the trade-off between 470 exploitation and exploration, that is closely aligned with models of patch foraging. 471 Similar search strategies borrowed from behavioral ecology have recently been applied to 472 human information processing, for example, in terms of giving-up rules in problem 473 solving (Wilke et al., 2009) and how the structure of information in the world wide web 474 can facilitate appropriate search policies (Fu & Pirolli, 2007). But there are potentially 475 other ways to implement search policies and many dimensions along which they may be 476 defined. Given that some characterizations of search, e.g., exploitation versus 477 exploration, better lend themselves to comparative analysis-both across organisms and 478 algorithms—understanding how we define the dimensions of search and characterize 479 different search policies may help us to better integrate our understanding of search and 480 cognitive abilities. 481

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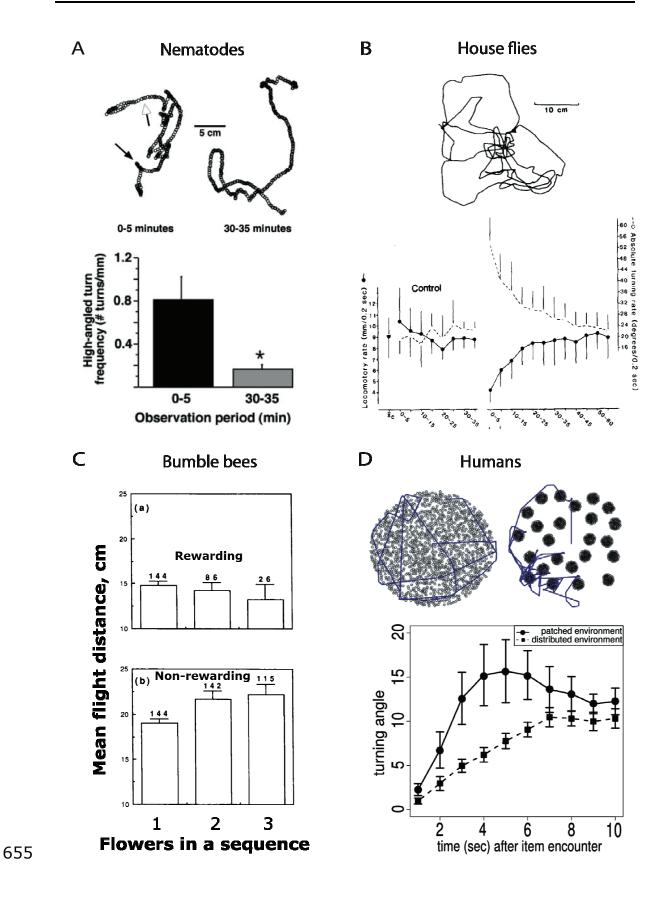
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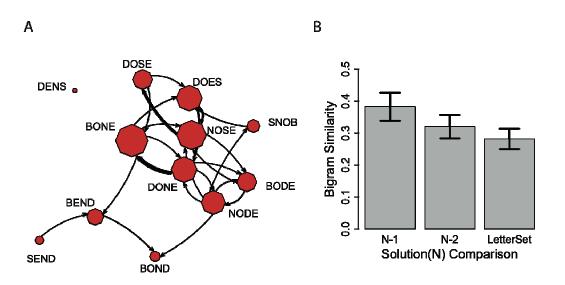
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656 Figure 1. Evidence of resource contingent foraging in A) nematodes (*Caenorhabditis* 657 elegans), B) house flies (Musca domestica), C) Bumble bees (Bombus bimaculatus), and 658 D) humans (*Homo sapiens*). In A) the top panel shows foraging paths for the nematode 659 C. elegans 0-5 minutes after encountering food and 30-35 minutes later. The filled arrow 660 indicates a high angled turn; the unfilled arrow indicates a region of the path with no 661 turning. The lower panel shows that high angled turns are significantly more likely for 662 the interval more recently associated with food (from Hills et al., 2004). In B) the top 663 panel shows a 69 second path for the house fly *M. domestica* immediately after 664 encountering food (at the central dot). The lower panel shows the quantitative 665 comparison of turning angle (open circles) and locomotory rate (closed circles) for 666 control flies (on the left) and flies immediately after encountering food (on right) (from 667 White et al., 1984). In C) the top panel shows a significantly decreasing flight distance to 668 the next flower following sequences of one, two, or three rewarding flowers for the 669 bumblebee B. bimaculatus. The lower panel shows a significantly increasing flight 670 distance after a series of one, two, or three non-rewarding flowers (data from Dukas & 671 Real, 1993). In D) the top panel shows typical paths for humans foraging in a 3-672 dimensional environment with invisible resources arrayed in distributed or clustered 673 arrangements. The lower panel shows that humans show significantly increased turning 674 following encounters with resources in clustered environments than in distributed 675 environments (from Kalff et al., 2010). 676



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Figure 2. Behavior in an anagram search task. A. A visual depiction of the between 680 word transitions produced by all participants in the letter set NSBDOE. Participants 681 looked for words they could make from letters in the letter set (using four or more 682 letters). Nodes represent solutions and links between nodes represent transitions between 683 words, with the arrow showing which word came second. Node size is proportional to 684 the number of participants who provided that solution for this letter set. Link thickness is 685 proportional to the number of participants who made that transition. For visual clarity, 686 only transitions that took place more than twice are represented with a link. B. The 687 bigram similarity of the present solution to previous (N-1) and two-back (N-2) solutions 688 and to the original letter set, showing that solutions tended to have the highest string 689 similarity to solutions produced nearby. Error bars are standard error of the mean. Figure 690 borrowed from Hills, Todd, & Goldstone, 2010. 691

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