

1 The demise of the synapse as the locus of 2 memory: A looming paradigm shift?

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5 Abstract

6 Synaptic plasticity is widely considered to be the neurobiological
7 basis of learning and memory by neuroscientists and researchers in
8 adjacent fields, though diverging opinions are increasingly being recog-
9 nised. From the perspective of what we might call “classical cognitive
10 science” it has always been understood that the mind/brain is to be
11 considered a computational-representational system. Proponents of the
12 information-processing approach to cognitive science have long been
13 critical of connectionist or network approaches to (neuro-)cognitive
14 architecture, pointing to the shortcomings of the associative psychology
15 that underlies Hebbian learning as well as to the fact that synapses
16 are practically unfit to implement symbols. Recent work on memory
17 has been adding fuel to the fire and current findings in neuroscience
18 now provide first tentative neurobiological evidence for the cognitive
19 scientists’ doubts about the synapse as the (sole) locus of memory
20 in the brain. This paper briefly considers the history and appeal of
21 synaptic plasticity as a memory mechanism, followed by a summary of
22 the cognitive scientists’ objections regarding these assertions. Next, a
23 variety of tentative neuroscientific evidence that appears to substan-
24 tiate questioning the idea of the synapse as the locus of memory is
25 presented. On this basis, a novel way of thinking about the role of
26 synaptic plasticity in learning and memory is proposed.

27 1 Introduction

28 Synaptic plasticity is widely considered to provide the neurobiological basis
29 of learning and memory by neuroscientists and researchers in adjacent fields.

30 However, diverging opinions are increasingly being recognised (e.g., Dudai
31 et al., 2015; Poo et al., 2016).

32 Within what we might call “classical cognitive science” (Piattelli-Palmarini,
33 2001) it has always been understood that the mind/brain is to be considered
34 a computational-representational system. Yet, not all cognitive scientists
35 have ever (fully) agreed with this assessment (e.g., Rumelhart et al., 1986).
36 Actually, as of today, large parts of the field have concluded, primarily drawing
37 on work in neuroscience, that neither symbolism nor computationalism are
38 tenable and, as a consequence, have turned elsewhere. In contrast, classical
39 cognitive scientists have always been critical of connectionist or network
40 approaches to cognitive architecture (e.g., Fodor and Pylyshyn, 1988), and
41 recent work on memory (e.g., Gallistel and King, 2009; Gallistel and Matzel,
42 2013; Gallistel and Balsam, 2014) has been adding fuel to the fire. Recent work
43 in neuroscience (Johansson et al., 2014; Chen et al., 2014; Ryan et al., 2015)
44 has now provided first tentative neurobiological evidence for the cognitive
45 scientists’ doubts about the synapse as the locus of memory in the brain.

46 This paper briefly considers the history and appeal of synaptic plasticity
47 as a memory mechanism, followed by a summary of the cognitive scientists’
48 objections to this idea. Next, a variety of tentative neuroscientific evidence
49 that appears to substantiates questioning the idea of the synapse as the locus
50 of memory is considered. On this basis, a novel way of thinking about the role
51 of synaptic plasticity in learning and memory—mentioned only in passing in
52 a recent commentary (Trettenbrein, 2015)—is proposed.

53 **2 A tentative idea with an intuitive appeal**

54 It was Ramón y Cajal who first concluded from his studies of bird brains
55 that neurons touch one another yet remain separate entities; Sherrington
56 later coined the term “synapse” to refer to the microscopic gap between
57 individual nerve cells (Glickstein, 2014). Subsequently, the psychologist
58 Donald Hebb made the at first merely theoretical proposal that changes
59 in synaptic connectivity and strength might constitute the fundamental
60 mechanism for information storage in the brain. It is interesting to note that
61 Ramón y Cajal had, in a way, anticipated this conceptual move when he noted
62 that “[...] interneuronal connectivity [...] is susceptible to being influenced
63 and modified during youthful years by education and habits” (Delgado-García,
64 2015, p. 6). As of today, the general idea that learning is essentially the
65 modification of synapses in an ever-changing plastic brain (a problematic
66 notion; see Delgado-García and Gruart, 2004; Delgado-García, 2015) has
67 become one of the dogmas of modern neuroscience and is usually presented in

popular science as well as in the scientific literature proper as an established fact and “generally accepted” (Brüel-Jungerman et al., 2007).

The fundamental principle of Hebb’s ideas on learning and memory was later poignantly summarised by Shatz as “[...] cells that fire together wire together” (1992). Hebb himself more elaborately suggested that

[w]hen an axon of cell A is near enough to excite cell B and repeatedly or persistently takes part in firing it, some growth process or metabolic change takes place in one or both cells such that A’s efficiency, as one of the cells firing B, is increased. (Hebb as cited in Glickstein, 2014, p. 253)

While these ideas about synaptic plasticity originally were purely theoretical in nature they have long-since been confirmed experimentally with the discovery of long-term potentiation (LTP) and long-term depression (LTD) as complementary neurobiological mechanisms.

In this context, it is important to point out the crucial role that (cognitive) psychology and its philosophical predecessors have played in this overall development. Not only was Hebb himself an early neuropsychologist trained by Karl Lashley, his idea that learning occurs whenever two cells fire together is clearly reminiscent of Lockean associative psychology which, in turn, can be traced back in the history of ideas all the way to Aristotle. This associative aspect of Hebbian learning has an intuitive appeal, as Gallistel and King (2009) recount tongue-in-cheek: How else could you explain that if you hear “salt” you will probably also think “pepper?” Associationism has come in different flavors since the days of Skinner, but they all share the fundamental aversion towards internally adding structure to contingencies in the world (Gallistel and Matzel, 2013). Crucially, it is only against this background of association learning that LTP and LTD seem to provide a neurobiologically as well as psychologically plausible mechanism for learning and memory.

3 Why not (only) the synapse?

If learning actually is the association of ideas in the Lockean sense, then everything is fine and LTP and LTD provide mechanisms that enable brains to carry forward information in time. However, as the proponents of a computational-representational view of the mind/brain have been arguing almost since the inception of cognitive science against the backdrop of behaviorism in the 1950s, learning is not association. In similar fashion, Gallistel and collaborators (2009; 2013) have in recent years been pointing to this conceptual flaw with regard to how learning and memory are understood in

neuroscience and other cognitive sciences, convincingly arguing that indeed not even the most fundamental properties of associative learning can be accounted for by LTP and LTD.

Pavlov himself already knew that the associative strength between two stimuli is determined by a multitude of experimental aspects in combination, meaning that all the different variables are encoded in a single association. In order to read information from memory it would be necessary to know about the mapping rules employed when writing that information to memory. However, in such a scenario where a variety of different variables have been mixed and encoded in a single association it is mathematically impossible to determine the value of any of the variables that entered into the original calculation: Mathematically speaking, association is a many-one function so that Pavlov's goal of discovering general "laws of association" cannot be met. Instead, in order to regain useful information from associative strengths, it is necessary to assume that there are different mapping rules (i.e. neurobiological processes) for every synapse. It stands to reason that this is an unpleasant assumption which runs counter to Pavlov's intent.

Furthermore, Hebbian learning rests on the idea that the neuron is the basic unit of information processing in the brain only by virtue of its connectivity profile, meaning that strength and patterns of synaptic connections constitute the way in which information is represented and stored in the brain. However, there is no *a priori* reason why this should be the case and while the vast majority of work in cognitive neuroscience of course relies on imaging or (indirectly) recording from entire populations of neurons if not functional modules or systems, we need not and should not assume that information processing in the brain stops at the cellular level.

Based on the observation that every neuron performs a rather stereotypical computational operation on its input (Kandel et al., 2013), it seems plausible that much more information processing is actually going on inside the cell. After all, the neuron in itself is an incredibly complex and morphologically diverse structure, of which, despite all progress, we still have not reached a satisfactory understanding. Incidentally, once we adopt such a perspective we also no longer face the additional problem that changes in synaptic conductance (i.e. modification of synaptic weights) are actually not accessible to computational processes being carried out inside the cell.

3.1 The view from classical cognitive science

Despite the fact that the information-processing approach to the study of the mind/brain has gone somewhat out of fashion within the cognitive sciences in the past decades, Gallistel and King (2009) have presented a convincing

144 argument for an understanding of the mind/brain as being Turing complete.
145 (In this context, see table 1 for working definitions of key terms.) Needless to
146 say, Gallistel and King’s original argument is by far more detailed than the
147 sketch that I can provide here, yet the main points should surface and will
148 hopefully suffice for further treatment.

149 Once we reconsider the classical cognitive scientist’s conception of the
150 mind/brain as a computational-representational system it is evident that the
151 brain must adhere to the abstract architectural properties of a universal Turing
152 machine, meaning that it is capable of universal computation. This, of course,
153 is not to say that the brain must resemble von Neumann’s implementation of
154 a Turing machine, but that it nevertheless seems to adhere to the abstract
155 properties of a Turing machine. Crucially, as David Marr put it, “[v]iewing
156 our brains as information-processing devices is not demeaning [...]” (2010,
157 p. 361). It might be the case that the brain is capable of carrying out
158 computations that a Turing machine cannot compute, but we do not know
159 whether this is the case nor how the brain might achieve this feat.

160 Provided that the cognitive functions exhibited by human brains require
161 the capabilities of Turing machines, one could quickly be led to a *Scala*
162 *Naturae* interpretation of the evolution of computational abilities of nervous
163 systems. Humans seem to be generalists, whereas animal learning is usually
164 seen as highly domain-specific (e.g., Gallistel, 1999). Thus, one might be led
165 to conclude that only human brains are capable of universal computation.
166 In fact, some have made this exact claim with regard to human language
167 capacities (e.g., Steedman, 2014). However, this would clearly be a mistake,
168 as it is now understood that even insect navigation already requires the
169 capabilities of a Turing machine (Gallistel, 1998; Gallistel and King, 2009). In
170 other words, “[...] ants have already climbed all the way up Nature’s ladder.”
171 (Berwick and Chomsky, 2016, p. 132)

172 The crucial feature of a Turing machine is its memory component: the
173 (hypothetical) machine must possess a read/write memory in order to be
174 vastly more capable than a machine that remembers the past only by changing
175 the state of the processor, as does, for example, a finite-state machine without
176 read/write memory. Thus, there must be an efficient way of storing symbols
177 in memory (i.e. writing), locating symbols in memory (i.e. addressing), and
178 transporting symbols to the computational machinery (i.e. reading). It is
179 exactly this problem, argue Gallistel and King (2009), that has by and large
180 been overlooked or ignored by neuroscientists.

181 Now, when we are looking for a mechanism that implements a read/write
182 memory in the nervous system, looking at synaptic strength and connectivity
183 patterns might be misleading for many reasons. Most pressingly, as Gallistel
184 and King point out, synapses might already be too complex in terms of

185 implementing such a very basic function:

186 In the final analysis, however, our skepticism rests most strongly
187 on the fact that the synapse is a circuit-level structure, a structure
188 that it takes two different neurons and a great many molecules
189 to realize. It seems to us likely for a variety of reasons that the
190 elementary unit in the memory mechanism will prove to be a
191 molecular or sub-molecular structural unit. (2009, p. 282)

192 Hence, they suggest turning to DNA and RNA, which already implement the
193 functionality of a read/write memory at the sub-molecular level. Interestingly,
194 in discussing recent work on memory, Poo et al. reach a similar conclusion
195 when they remark that “[...] some other mechanisms, potentially involving
196 epigenomic modifications in engram neurons, appear to be necessary for
197 memory trace storage” (2016, p. 8).

198 A mechanism as essential as memory has to be efficient in all respects, be
199 it implementational complexity or energy efficiency. Another part of Gallistel
200 and collaborators’ argument for the point of view they put forward is the
201 observation that neural computation is demonstrably incredibly fast, therefore
202 making it much more likely that the memory mechanism is (sub-)molecular
203 in nature so that computational machinery and memory can be located in
204 close physical proximity in order to minimize the distance over which a signal
205 has to be transmitted (a process which evidently is “slow” in the nervous
206 system in comparison to, for example, conventional computers).

207 **3.2 Some tentative evidence**

208 To this day, tentative evidence for the (classical) cognitive scientists’ reserva-
209 tions towards the synapse as the locus of memory in the brain has accumulated.
210 A lot of groundbreaking work concerning the way in which the brain carries
211 forward information in time was actually performed on comparatively simple
212 model organisms such as *Aplysia* and has then been extrapolated to speculate
213 about what might be going on in human mind/brains (e.g., Kandel and
214 Siegelbaum, 2013). Interestingly, it is recent work in this exact domain which
215 has indicated that the idea of synaptic conductance as the basic memory
216 mechanism is insufficient and incomplete at best.

217 In Kandel and collaborators’ by now classic work with *Aplysia*, changes
218 in synaptic conductivity were shown to alter how the animal reflexively
219 responds to its environment. But not even in *Aplysia* all synapses are equally
220 susceptible to change, many appeared not to be very plastic (Kandel and
221 Siegelbaum, 2013). Recent work with cultured *Aplysia* motor and sensory
222 neurons by Chen et al. (2014) has revealed that long-term memories appear

to persist covertly in cell bodies and can be restored after synapses have been eliminated. Long-term memory persisted after pharmacological elimination of synapses that had been produced only after learning had occurred, calling the role of synapses as the presumed locus of memory into serious doubt.

Similarly and possibly even more convincing, in a groundbreaking study building on earlier work (Hesslow et al., 2013) that already pointed to the mismatch between LTD in Purkinje cells and cerebellar learning, Johansson et al. (2014) investigated how the response of Purkinje cells changes during learning. Studying eyeblink conditioning, they showed that the cells could learn the temporal relationship between paired stimuli during conditioning. Strikingly and in stark contrast to widespread belief, the timing of responses exhibited by conditioned Purkinje cells after conditioning did not depend on a temporally patterned input. Consequently, Johansson et al. conclude that both, timing mechanism and memory trace, are located within the Purkinje cell itself. As they put it, “[...] the data strongly suggest that the main timing mechanism is within the Purkinje cell and that its nature is cellular rather than a network property” (Johansson et al., 2014, p. 14933).

Lastly, in a recent study supposed to demonstrate the increase in synaptic strength and density of dendritic spines during memory consolidation, Ryan et al. (2015), to their own surprise, showed that changes in synaptic strength are not directly related to storage of new information in memory. In accordance with the literature on memory consolidation, Ryan et al. found that injection of protein synthesis inhibitors induced retrograde amnesia, meaning that the memory could not be retrieved. However, when optogenetically activating the neurons previously tagged during the conditioning process, memories could nevertheless be retrieved despite chemical blocking, indicating that the formation of synapses or strengthening of synaptic weights is not critical to memory formation as such.

3.3 Rate of synaptic turnover

The synaptic trace theory of memory requires synaptic conductance and connectivity to change during learning, that is when new information is being memorised. Studies by Xu et al. (2009) and Yang et al. (2009) both used two-photon microscopy, a feat of contemporary technology that makes it possible to trace individual synaptic spines over prolonged periods of time (i.e. weeks to months), in order to investigate the predicted changes in motor cortex during acquisition of a new motor skill.

As the researchers had anticipated, they found that learning of the new motor skill indeed was accompanied by the formation of new synaptic connections. Yet, the more puzzling finding of their studies is that synaptic spines

were found to be still turning over at a rather high rate in absence of learning. As a matter of fact, the rate of synaptic turnover in absence of learning is actually so high that the newly formed connections (which supposedly encode the new memory) will have vanished in due time. It is worth noticing that these findings actually are to be expected when considering that synapses are made of proteins which are generally known to have a short lifetime.

Nevertheless, the observation that synapses are turning over at a high rate even in absence of learning, of course, is paradoxical. Interestingly, this was already noticed by A. von K  lliker, a contemporary of Ram  n y Cajal (Delgado-Garc  a, 2015). Today, Bizzi and Ajemian observe as an aside in a review of the current state of research on voluntary movement:

If we believe that memories are made of patterns of synaptic connections sculpted by experience, and if we know, behaviorally, that motor memories last a lifetime, then how can we explain the fact that individual synaptic spines are constantly turning over and that aggregate synaptic strengths are constantly fluctuating? (2015, p. 91)

Just as Bizzi and Ajemian go on to describe, this finding is amongst those that are the most challenging to the idea that the synapse is the locus of memory in the brain.

Synapses have been found to be constantly turning over in all parts of cortex that have been examined using two-photon microscopy so far (see papers cited in Yang et al., 2009), meaning that (motor) memories by far outlive their supposed constituent parts. It seems that there are two possible ways of resolving this puzzle: We can either assume that memories are perpetually being retrieved from memory and re-encoded during this constant turnover, or we might conclude that the widely presumed relation between synaptic conductance and connectivity and memory is not as direct as conventional wisdom would have it. Provided that there is some merit to the idea that brain’s memory mechanism might be localized to neurons’ somata, a separation between learning and memory seems indicated.

4 The need to separate learning and memory

That learning and memory might be dissociated has been implicitly acknowledged in the neuroscience literature (e.g., Bannerman et al., 1995; Saucier and Cain, 1995). The word has not (yet) spread to other disciplines, presumably because experimental results have, to a certain extent, been somewhat ambiguous (Martin and Morris, 2002). Consequently, these reported tentative

findings are usually not readily interpreted as evidence countering the idea of associative LTP/LTD as the putative basis for learning and memory.

A good example is spatial learning which is crucially dependent on hippocampus (Bannerman et al., 1995; Martin and Morris, 2002). Without going into great detail here, it can be said that the production of LTP (though not its maintenance) has been shown to crucially depend on *N*-methyl-D-aspartate (NMDA) receptors located in the dendritic spine of the postsynaptic neuron (Brüel-Jungferman et al., 2007; Siegelbaum et al., 2013). It thus follows that if these receptors are chemically blocked, learning should be impaired or rendered impossible. Indeed, animals with blocked LTP exhibit impairment of their ability for spatial learning. But this general statement requires some qualifications as Bannerman et al. (1995) as well as Saucier and Cain (1995) have shown that pretraining can actually “compensate” for pharmacological blocking of NMDA receptors so that animals perform (close to) normal.

Otherwise put, when animals were pretrained in navigating in one water maze they could readily learn to navigate in a second one despite the chemical blocking of LTP. This might be interpreted as to indicate that NMDA receptors do play a role in initial learning of a new skill (i.e. navigating a maze) but do not appear to play any role in altering the specifics, or maybe better ‘contents,’ for example, when a new map is added to memory, respectively when the already existing representation is being updated. Provided that blocking of NMDA receptors did not prevent the acquisition of new information it seems reasonable to purport that a memory mechanism other than LTP was at work here, though the nature of this mechanism remains unknown. All in all, we might take this as an indication for a dissociation of (spatial) learning and the memory mechanism(s) as such, an interpretation that has abundant representational implications (see also Gallistel and Matzel, 2013).

We might now once again turn to (classical) cognitive science and consider these findings against the backdrop that learning is highly domain-specific (e.g., Chomsky, 1975; Gallistel, 1999): An information-processing perspective on the mind/brain necessarily leads to the postulation of domain-specific learning mechanisms. Hence, based on the above-mentioned studies we might postulate that the (spatial) learning mechanism is only partially dependent on synaptic plasticity. Acquisition of the skill relies on synaptic plasticity and thus a process of neural reorganisation, whereas altering the specifics (i.e. acquisition of new information, respectively “updating” of information already stored in memory) does not. It follows that LTP and thus synaptic plasticity cannot provide the brain’s basic memory mechanism.

In the sense of Gallistel and King (2009), learning is the process of extracting information from the environment, whereas memorizing is the processes of storing this information in a manner that is accessible to computation.

340 It is interesting to note that once learning and memory are conceived of as
 341 separate processes, the above-mentioned observation that synaptic spines are
 342 still turning over at a very high rate in absence of learning does no longer
 343 pose such a severe problem. In somewhat similar fashion, we can interpret
 344 the findings of Ryan et al. (2015) against this background, so that we might
 345 say that in their study information was extracted from the environment (i.e.
 346 learning occurred) and stored in memory independently of the process of
 347 memory consolidation, that is alteration of synaptic weights and connectivity.

348 Lastly, all of this is not to say that synaptic plasticity and networks are of
 349 no importance for learning and memory. Fodor and Pylyshyn (1988) already
 350 reviewed the implications of connectionist models, concluding that connection-
 351 ism might best be understood as a neutral “theory of implementation” of the
 352 actual cognitive architecture, provided that one gives up anti-representational
 353 tendencies inherent to the approach. As a consequence, the question no longer
 354 is whether symbolic representations are “real,” but how (i.e. on what level)
 355 they are actually implemented in the brain. The challenge for critics of the
 356 synaptic plasticity hypothesis will therefore be to come up with concrete
 357 suggestions for how memory might be implemented on the sub-cellular level
 358 and how cells then relate to the networks in which they are embedded.

359 5 Rethinking synaptic plasticity

360 The realization that the synapse is probably an ill fit when looking for a
 361 basic memory mechanism in the nervous system does not entail that synaptic
 362 plasticity should be deemed irrelevant. Quite to the contrary, there of course
 363 is ample and convincing evidence that synaptic plasticity is a prerequisite for
 364 many forms of learning (see e.g., Martin and Morris, 2002; Münte et al., 2002;
 365 Bruel-Jungerman et al., 2007; Jäncke, 2009; Dudai et al., 2015; Ryan et al.,
 366 2015; Poo et al., 2016). However, it occurs to me that we should seriously
 367 consider the possibility that the observable changes in synaptic weights and
 368 connectivity might not so much constitute the very basis of learning as they
 369 are the result of learning.

370 This is to say that once we accept the conjecture of Gallistel and collabo-
 371 rators that the study of learning can and should be separated from the study
 372 of memory to a certain extent, we can reinterpret synaptic plasticity as the
 373 brain’s way of ensuring a connectivity and activity pattern that is efficient
 374 and appropriate to environmental and internal requirements within physical
 375 and developmental constraints. Consequently, synaptic plasticity might be
 376 understood as a means of regulating behavior (i.e. activity and connectivity
 377 patterns) only after learning has already occurred. In other words, synaptic

378 weights and connections are altered after relevant information has already
379 been extracted from the environment and stored in memory.

380 Over roughly the last decade, evidence that supports such an interpretation
381 has been piling up, suggesting that the brain is (close to) “optimally wired.”
382 It seems that axons and dendrites are close to the smallest possible length,
383 at least within a cortical column (Chklovskii et al., 2002; Chklovskii, 2004)
384 and possibly also globally (Cherniak et al., 2004; Lewis et al., 2012; Sporns,
385 2012). As noted by Chklovskii et al., this “optimality” has further pressing
386 implications for the idea that the synapse is the locus of memory, after all,

387 [. . .] an increased number of synapses could not be accommodated
388 without degrading performance in some way because the cortex is
389 already optimally wired in the sense that the number of synapses
390 is already maximal (2002, p. 345).

391 The role of synaptic plasticity thus changes from providing the funda-
392 mental memory mechanism to providing the brain’s way of ensuring that
393 its wiring diagram enables it to operate efficiently with regard to environ-
394 mental and internal pressures. Viewed against the background that synapses
395 are practically unfit to implement the cognitive scientists’ beloved symbols,
396 it seems that we seriously have to consider that synaptic plasticity might
397 not implement a memory mechanism as such. Instead, changes in synaptic
398 conductance and connectivity might provide a bundle of mechanisms which
399 regulate and ensure that the network and its modules perform and interact
400 efficiently.

401 In this regard, it is vital to note that while cognitive science tells us
402 that learning is domain-specific, these observations unfortunately cannot
403 tell us whether the basic memory mechanism is rather uniform or not. An
404 evolutionary argument could be put forward in favour of a view where the basic
405 memory mechanisms is highly conserved, but such a theory has not yet been
406 confirmed to facts. If memory actually turns out to be sub-cellular in nature,
407 synaptic plasticity would of course not be rendered irrelevant. However, what
408 would change is the function commonly attributed to synapses: For example,
409 one possibility is that synapses could be understood as providing “access points”
410 to information already stored in memory inside the cell (Ryan et al., 2015),
411 instead of a way of carrying forward information in time. Memories stored in
412 cells could thus possibly be considered to be synapse-specific, meaning that
413 activating different synapses will elicit different events in the cell.

6 A tentative outlook

To sum up, it can be said that when it comes to answering the question of how information is carried forward in time in the brain we remain largely clueless. Fittingly, in a recent autobiographical account of his research, scientific career, and personal life, Michael Gazzaniga commented on the current problems of (cognitive) neuroscience, concluding that “[...] neuroscience still has not collected the key data because, to some extent, it is not known what that key data even is” (Gazzaniga, 2015, p. 190).

Apparently, very much as Marr (2010) envisioned, the “classical” cognitive scientist’s analysis of the information-processing problem at stake in the study of memory now has yielded first hints with regard to where neurobiologists should be looking for this key data when studying the brain’s fundamental memory mechanism(s): inside the cell. Tentative evidence from a wide variety of work in neuroscience seems to provide support for the idea that the synapse is an ill fit when looking for the brain’s basic memory mechanism: memory persists despite synapses having been destroyed and synapses are turning over at very high rates even when nothing is being learned. All things considered, the case against synaptic plasticity is convincing, but it should be emphasised that we are currently also still lacking a coherent alternative.

Adolphs (2015) optimistically listed the problem of how learning and memory work among those that he expects to be solved by neuroscientists within the next 50 years. We shall see how this turns out, but, if anything, the evidence and recent findings discussed here seem to indicate to me that we will have to rethink many of the basic propositions in the cognitive sciences and especially neuroscience in order to actually achieve this. Yet, it is not at all implausible that in the years to come we might see the paradigm shift that Gallistel and Balsam (2014) have been calling for.

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Term	Exposition
finite-state machine	An abstract machine that can be in only one state at a time and a finite number of states in total. Its memory is defined by the number states available.
Turing machine	A finite-state machine extended with a so-called tape. The tape is a read/write memory component where symbols can be stored and recovered.
Turing completeness	Refers to the ability of a given set of instructions to simulate a Turing machine.
von-Neumann implementation	Denotes a common schematic circuit concept (and its many offshoots) that actually implement a universal Turing machine.

Table 1: Nomenclature. This table provides brief expositions of terms and concepts from theory of computation that might not be familiar to all readers. Note that these are working definitions for the purpose of this paper, they are not meant to be exhaustive.