Modeling Episodic Memory in Cognitive Architectures

A Comparative Study of Soar and Xapagy

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	Background		Methodology		Results		
Definition	Most cognitive architectures include memory systems for storing intermediate results, supporting learning and adaptation in dynamic	Literature Search	Systematic literature collection to identify and select CAs with EM mechanisms documented in detail, for in-depth analysis.	Dimension	Soar	Xapagy	
	environments [1]. Among the various types of memory discussed in cognitive architecture	Comparison Study	Analysis based on comparison framework	Representation & Structure	Symbolic snapshots of working memory; Graph-structured with identifiers and attributes; Temporally ordered episodes	Raw episodic recording: Conceptual overlays; Verb-instance graphs; Each episode is made of atomic VIs connected temporally and contextually	
	literature, episodic memory (EM) has received attention due to its role in enabling agents to recall and learn from past experiences.		Dimension Description	Encoding	Automatic at intervals; Captures changes only in top level of working memory	Instances and VIs are encoded by staying in the focus, gaining salience over time, influenced by marking rate and activity type	
	EM refers to memory for events (episodes), often expressed as the what, the where, and the when [2].		Representation & Structure How are episodes represented within the system? What elements does an episode contain? Encoding How are episodes formed or encoded during system operation?	Storage	Indexed by temporal order; Stores only changes; Does not store substates or retrievals	Weighted sets; Non-indexed; Passive memory; Shadowing; Retrieval occurs via automatic	
Research Gap	While prior work [3] has reviewed episodic memory across a range of cognitive architectures in a broad and high-level manner, a detailed ,		Storage How are episodic memories stored within the architecture?	Retrieval	Cue-based; Explicit queries with features; Returns best-matching episode by recency or similarity	matching between current focus (working memory) and episodic memories; forming predictions or inferences	
	structured comparison among architectures remains lacking. This gap is significant, as understanding how different cognitive		Retrieval How are episodes retrieved and what triggers retrieval? Memory Updates Are there mechanisms for forgetting, updating, or managing memory?	Memory Updates	Static after encoding; No automatic forgetting or generalization	Exponential decay; Self-shadowing and drift; Memories lose salience over time; repeated recalls can distort memory content through drift	
	models conceptualize and implement episodic memory can offer insights into their design principles, cognitive plausibility, and practical utility.		Biological Plausibility How does the architecture resemble human episodic memory? Limitations Are there any known issues, limitations, or missing components in the EM	Biological Plausibility	Inspired by human episodic memory concepts but structurally tied to symbolic reasoning;	Emergent memory dynamics; Models forgetting, interference, and recall bias similar to human episodic memory	
Primary Question	How do different approaches to episodic memory modeling in CAs reflect assumptions about the role of episodic memory inn	¥	model?	Limitations	Retrieval depends on explicit cues; Recency bias; No automatic integration with procedural knowledge; Computational cost increases with memory size	No abstraction; No procedural memory; Episodic-only reasoning; Lacks generalization; Memory cannot be queried directly	
	cognition, and what design trade-offs do they reveal ?	Synthesis	Identify patterns, design trade-offs, and implications.				
	Soar vs Xapagy		Identification of studies via databases		Key Insights		
Purpose	general-purposed VS narrative reasoning only	E Decembri		1) Disti	nctive cognitive goals lead to fundame	ntally different design choices.	
Memory System	multiple long-term memory models VS EM as the only model	Scop	Jentified from: (n=347) Records removed before screening:	2) Cont	rasting assumptions about the role of e	pisodic memory in cognition:	
			polore (n = 108) Records from other fields removed (n = 266) Df Science (n = 60) Duplicate records removed (n = 184)	1)	Soar: supportive to rule-based problem	i-solving	
Symbolic Long-Term Memo	ories		Digital Library (n = 37) ce Direct(n=50)	2)	Xapagy: cognitive behavior emerging s	olely from accumulative experience	
				3) High level conceptualization guides implementation of EM in sub-systems			
Reinforcement Chanking Semantic Learning Semantic	Restormed Cheming Examine Examine Examine Examine Examine Symbolic Working Momory Symbolic Working Momory Figure 1 Figure 2 Figure 2 Machine Symbolic Working Momory Figure 2 Figure 2 Figure 2			1) Soar: structural organization, efficiency, accuracy			
			Records screened by abstract with inclusion criteria (n = 37)		2) Xapagy: flexibility, association, dynamic		
Regular Regular <t< td=""><td>Scr</td><td colspan="3">4) Different development dynamics:</td></t<>		Scr			4) Different development dynamics:		
					1) Soar: extension to an existing system, constrained by legacy systems		
Fig.1 EM (in SOAR) and its s	subsystems [4] Fig.2 EM as the only memory model in Xapagy [5]			2)	Xapagy: design from ground-up, memo	pry-centric, simple	
	Research Question	<u>n = 26)</u>	Records for chosen models (n = 10)				
Main Question	How do different approaches to episodic memory modeling in Soar and Xapagy reflect assumptions about the role of episodic memory in cognition, and what design trade offer do they reveal for cognitive architecture?	Soar, X	apagy, ICARUS, LIDA Soar vs Xapagy		Conclusion		
Sub Questions	and what design trade-offs do they reveal for cognitive architectures? How is episodic memory represented and structured in Soar and 		References	Conclusion Different approaches to EM modeling in Soar and Xapagy reflect distinctive assumptions about role of EM in cognition, and various design trade-offs, between generality and specialization, between symbolic precision and experiential flexibility, aligned			
	Xapagy?	[1] Julija Kotsaruba an	d John K. Tsotsos. A review of 40 years of cognitive architecture research: Core cognitive	Future Work 1) Apply similar comparison framework to additional architectures, to generalize findings and identify broader design trends and patterns.			
	2) What mechanisms are employed by each architecture for encoding ,		a John K. <u>Isotsos</u> . A review of 40 years of cognitive architecture research: Core cognitive applications. arXiv preprint arXiv:1610.08602, 2016.				
	storing, retrieving, and updates of episodic memories?		odic and semantic memory. In <u>Endel Tulving</u> and Wayne Donaldson, editors, Organization I-403. Academic Press, New York, 1972.				
	3) What are the $\ensuremath{\text{key}}$ differences and $\ensuremath{\text{limitations}}$ in the implementation		[3] Luis Martin, Karina Jaime, Felix Ramos, and Francisco Robles. Bio-inspired cognitive architecture of episodic		2) Empirical evaluation through task-based benchmarks		
	of episodic memory between the two architectures?		memory. Cognitive Systems Research, 76:26–45, 2022. [4] John E. Laird. The Soar Cognitive Architecture, chapter Introduction, pages 1–25. MIT Press, 2012.		3) Explore hybrid approaches to model EM in CA		
	4) what implications do these differences have for the development of [5] La		lau <u>Bölöni</u> . An investigation into the utility of episodic memory for cognitive architectures. AAAI Fall ium - Technical Report, 01 2011.		 Other forms of episodic memory – such as imagery – can be incorporated into existing architectures 		