

# SEARCH OF ASSOCIATIVE MEMORY (SAM)

Authored by  
**mohammad looti**

October 22, 2025

## RECOMMENDED CITATION

mohammad looti (2025). *SEARCH OF ASSOCIATIVE MEMORY (SAM)*. PSYCHOLOGICAL SCALES. Retrieved from <https://scales.arabpsychology.com/?p=54483>

## SEARCH OF ASSOCIATIVE MEMORY (SAM)

**Primary Disciplinary Field(s):** Cognitive Psychology, Mathematical Psychology, Memory Modeling

**Proponents:** Richard M. Shiffrin, Jeroen G. W. Raaijmakers

### 1. Core Principles

The Search of Associative Memory (SAM) model is a seminal mathematical framework developed to account for human performance in episodic memory tasks, particularly free recall and recognition. Postulated primarily by Richard M. Shiffrin and Jeroen G. W. Raaijmakers in the late 1970s and early 1980s, SAM provides a quantitative description of how items are encoded, stored, and retrieved from **long-term memory** (LTM). Unlike earlier, less formal models, SAM formalized the dual-store concept--the interaction between a limited capacity **short-term memory** (STM) and a vast LTM--and introduced the critical role of associative strength in guiding memory search processes. The model posits that retrieval is an active, reconstructive process involving sequential searches based on cues and the current context.

A fundamental principle of SAM is that items are stored in LTM not in isolation, but as interconnected nodes, establishing associations between items themselves (item-item associations) and between items and the context in which they were learned (item-context associations). When a person attempts to retrieve a memory, the current state or context acts as a retrieval cue, initiating a probabilistic search. This search process is governed by the associative strengths built up during encoding. A stronger association between the cue and a stored item increases the probability that the item will be sampled from LTM and placed into the limited capacity retrieval buffer (analogous to STM) for subsequent output or evaluation.

The success of SAM lies in its capacity to generate precise, testable predictions regarding a wide range of laboratory memory phenomena. These include the effects of list length, serial position effects (primacy and recency), the organization of recall (clustering), and phenomena related to recognition judgments. By translating psychological principles--such as the nature of memory storage and the dynamics of retrieval failure--into quantifiable parameters, SAM provided a powerful tool for researchers to fit experimental data and compare competing theoretical interpretations of memory function.

### 2. Proponents and Historical Context

The development of SAM cannot be separated from the legacy of the **dual-store models** that dominated cognitive psychology in the 1960s. Specifically, the model builds directly upon the architecture proposed by Richard C. Atkinson and Richard M. Shiffrin in 1968, known as the

Atkinson-Shiffrin Model. This earlier work established the distinction between separate short-term and long-term memory systems, proposing that STM served as a gateway and rehearsal space for information destined for LTM. However, the Atkinson-Shiffrin model primarily focused on the control processes governing memory transfer rather than detailing the mechanics of LTM retrieval itself.

The critical refinement came with the collaboration between Shiffrin and Raaijmakers, leading to the formalized SAM model in the early 1980s. Their goal was to provide a mathematically rigorous account of how the stored contents of LTM are accessed during recall. Raaijmakers and Shiffrin introduced crucial additions, including a sophisticated mechanism for competitive sampling from the associative matrix and a detailed description of context drift--the idea that the mental context surrounding an event changes over time, thus affecting the fidelity of retrieval cues. This transition marked a move from simply describing memory structure to actively modeling the processes involved in the search and retrieval phases.

SAM emerged during a period of intense interest in quantitative modeling within cognitive science, offering an alternative to purely verbal theories of memory. Its success cemented the importance of mathematical models in capturing the complexity of human cognition, providing a framework that was both theoretically rich and empirically verifiable. The structure and mechanisms introduced in SAM have profoundly influenced subsequent research, setting the stage for later computational models of memory, such as the **Retrieving Effectively From Memory (REM)** model and the Context Maintenance and Retrieval (CMR) model.

### 3. Model Architecture: Short-Term and Long-Term Stores

SAM operates on the assumption of a functionally separated, though interacting, memory system architecture. The **Short-Term Store (STS)**, often referred to as the rehearsal buffer, is characterized by a limited capacity and rapid forgetting. Information entering the system, such as a word presented in a list, resides in the STS. While in the STS, the item can be rehearsed, maintaining its presence in conscious awareness, or it can be encoded into the Long-Term Store. The size of the STS is a key parameter in SAM, influencing how many items can be actively maintained and subsequently transferred to LTM.

The **Long-Term Store (LTS)** is conceived as a vast, permanent repository where memory traces are stored in the form of association strengths. When an item is encoded, associations are established between the item and the current context, and between the item and any other items simultaneously present in the STS. These associations are represented in the model by an associative matrix ( $M$ ), where entries reflect the strength of the connection. For example, if items A and B are rehearsed together, the associative strength between A and B, and between A and the encoding context C, increases.

The dynamic interaction between these two stores is crucial during the retrieval phase. When retrieval begins, cues--whether internal (the context) or external (a prompt)--activate the associative matrix in the LTS. The search process samples items probabilistically based on their associative strength relative to the cue. Once an item is successfully sampled from LTM, it temporarily enters the STS. This retrieved item can then serve as a new cue for subsequent retrieval attempts, a process known as **chaining**. The temporary residence in the STS also explains the classic **recency effect** in free recall, as the most recently presented items are often still present in the STS at the time of test, making them immediately available for output.

#### 4. Associative Structure and Context Encoding

The power of SAM derives largely from its sophisticated handling of associative structure. The memory system is formalized as a matrix ( $M$ ) of associative strengths. There are two primary types of associations modeled: **item-context associations** ( $M_{\{ic\}}$ ) and **item-item associations** ( $M_{\{ij\}}$ ). Item-context associations link a specific memory trace ( $i$ ) to the general experimental or environmental context ( $c$ ) present during encoding. These links are critical because the context is almost always the initial cue used to initiate retrieval in free recall tasks.

Item-item associations link one item ( $i$ ) to another ( $j$ ). These associations are strengthened when items are rehearsed together or processed closely in time. In a free recall task, once an item is retrieved and outputted, its identity can be used as a cue to search for the next item, relying heavily on the pre-established item-item associations. For instance, if a participant successfully recalls "Dog," the model dictates that "Dog" becomes a secondary cue, biasing the subsequent search toward items strongly associated with it, such as "Cat" or "Leash."

A particularly innovative aspect of SAM is the concept of **contextual fluctuation** or context drift. The model assumes that the internal context cue, which initiates the search, is not static but changes slightly over the course of an experiment or across different retrieval attempts. This context drift provides a mechanism for explaining why retrieval success decreases over time and why the context cue might become less effective if the subject takes too long to respond. The ability of SAM to mathematically model the dynamic nature of both specific item associations and the background contextual state provides a robust framework for understanding the conditions under which memory retrieval succeeds or fails.

#### 5. Retrieval Mechanisms: Recall and Recognition

SAM provides distinct, yet interconnected, mechanisms for explaining both **free recall** and **recognition memory**, which are the two primary modes of episodic retrieval studied in the laboratory. In a free recall task (e.g., "list all the words you remember"), the primary cue is the overall list context. The model calculates the probability of sampling each stored item

( $P_{\text{sample}}$ ) based on its associative strength to the context cue relative to the strength of all other stored items. Retrieval is competitive; the stronger the association between the context and a particular item, the higher its probability of being sampled.

Once an item is sampled, it enters a retrieval buffer. The participant must then determine if the sampled item is indeed a target item (i.e., belongs to the list they are trying to recall) and avoid outputting items that were not presented (extralist intrusions). If the item is accepted, it is outputted, and the successful item can then be used as a new cue for the next sampling attempt, propagating the recall sequence. If the item is rejected or deemed inadequate, the search process continues until the participant either exhausts their search attempts or the context cue has drifted too far from the initial encoding context.

In contrast, recognition memory (e.g., "was this word on the list?") is typically modeled using a global familiarity process within the SAM framework, though later versions, like the SAM-based models developed by Gillund and Shiffrin (1984), introduced an association-based retrieval component. When presented with a test probe ( $X$ ), the model evaluates the sum of the associative strengths between the probe  $X$  and all relevant cues (the study context and potentially other items in LTM). This calculation yields a measure of **familiarity** or retrieval strength ( $SS$ ). If  $SS$  exceeds a predefined threshold, the item is recognized as having been studied. The key difference is that recall involves an active, cue-driven search and sampling process, whereas recognition often relies on a passive, quantitative assessment of overall associative strength.

## 6. Empirical Validation and Applications

The SAM model has achieved remarkable empirical success by accurately fitting data across numerous standard memory paradigms. One of its most significant accomplishments is modeling the **serial position curve** in free recall. SAM accounts for the **primacy effect** (better recall of early list items) by proposing that early items benefit from more rehearsal time in the STS, leading to stronger item-context associations in LTM. It accounts for the **recency effect** (better recall of late list items) because the most recent items are still highly active in the STS or, in the LTM component, their context association is strongest because the retrieval context is most similar to the encoding context.

Beyond the serial position curve, SAM has been crucial in explaining the phenomenon of **semantic clustering**. When participants are asked to recall a randomly mixed list of words from different categories (e.g., animals, professions), they tend to cluster their responses by category. SAM explains this by assuming that strong pre-existing semantic associations between category members increase the item-item associative strength ( $M_{ij}$ ). Once one item from a category is retrieved, it serves as a powerful cue to retrieve other items from the same category before the search shifts to a different category, providing a powerful explanation for the organization observed

in memory output.

Furthermore, the mathematical rigor of SAM allows for the estimation of underlying psychological parameters, such as the probability of encoding, the rate of context drift, and the capacity of the STS, from empirical data. This ability to extract meaningful parameters has allowed researchers to compare memory processes across different populations (e.g., young vs. old adults) or different experimental conditions (e.g., intentional vs. incidental learning), solidifying SAM's role as a vital tool for quantitative memory analysis in cognitive psychology.

## 7. Criticisms and Successors

Despite its comprehensive nature and empirical successes, the SAM model, particularly in its original formulation, faced several theoretical and practical criticisms. A key limitation often cited is its strict adherence to the **dual-store dichotomy**. As evidence mounted suggesting that forgetting in STM might be due to interference rather than simple decay, and that LTM and STM may share more common processes than initially assumed, the rigid separation of the two stores became problematic for some researchers.

Another major point of debate centers on the handling of context. While SAM introduced the crucial idea of context encoding and drift, the context representation itself is often treated abstractly, sometimes limiting its ability to explain how specific environmental cues or complex cognitive states influence retrieval. Furthermore, the original model struggles to account for phenomena like **source memory** (remembering where or when an item was encountered) without significant modification, as its primary focus is on item recognition and recall rather than the detailed retrieval of episodic details.

These limitations paved the way for the development of successor models that retained SAM's fundamental associative principles but integrated advances in representation and processing. The **Context Maintenance and Retrieval (CMR)** model, developed by Michael Kahana and others, is a prominent extension. CMR maintains the SAM structure but introduces a more dynamic and continuous representation of context that is updated not only by external cues but also by the items being retrieved, allowing it to better account for temporal contiguity effects and chaining in recall. Thus, while SAM is a historical cornerstone, its principles continue to form the foundation of most modern computational theories of episodic memory retrieval.

## Further Reading

[Search of Associative Memory \(SAM\)](#)

[Atkinson-Shiffrin Memory Model](#)

[Richard M. Shiffrin](#)