



# A Normative Account of the Influence of Contextual Familiarity and Novelty on Episodic Memory Policy

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## Summary

- Empirically, a familiar/novel stimulus triggers a lingering retrieval/encoding mode, even if this stimulus is incidental (Duncan et al. 2012, 2016, 2019). However, it is unclear why this is the case from a normative standpoint.
- A neural network model that learns an episodic memory (EM) policy in an autocorrelated environment can capture human data (Duncan et al. 2016, 2019), suggesting that these lingering EM modes are an optimal adaptation to environment statistics (see also Anderson & Schooler, 1991, and Honey et al., 2023), which is consistent with the neuromodulation account (e.g. Hasselmo et al., 1996)

## Familiarity/novelty trigger lingering EM modes

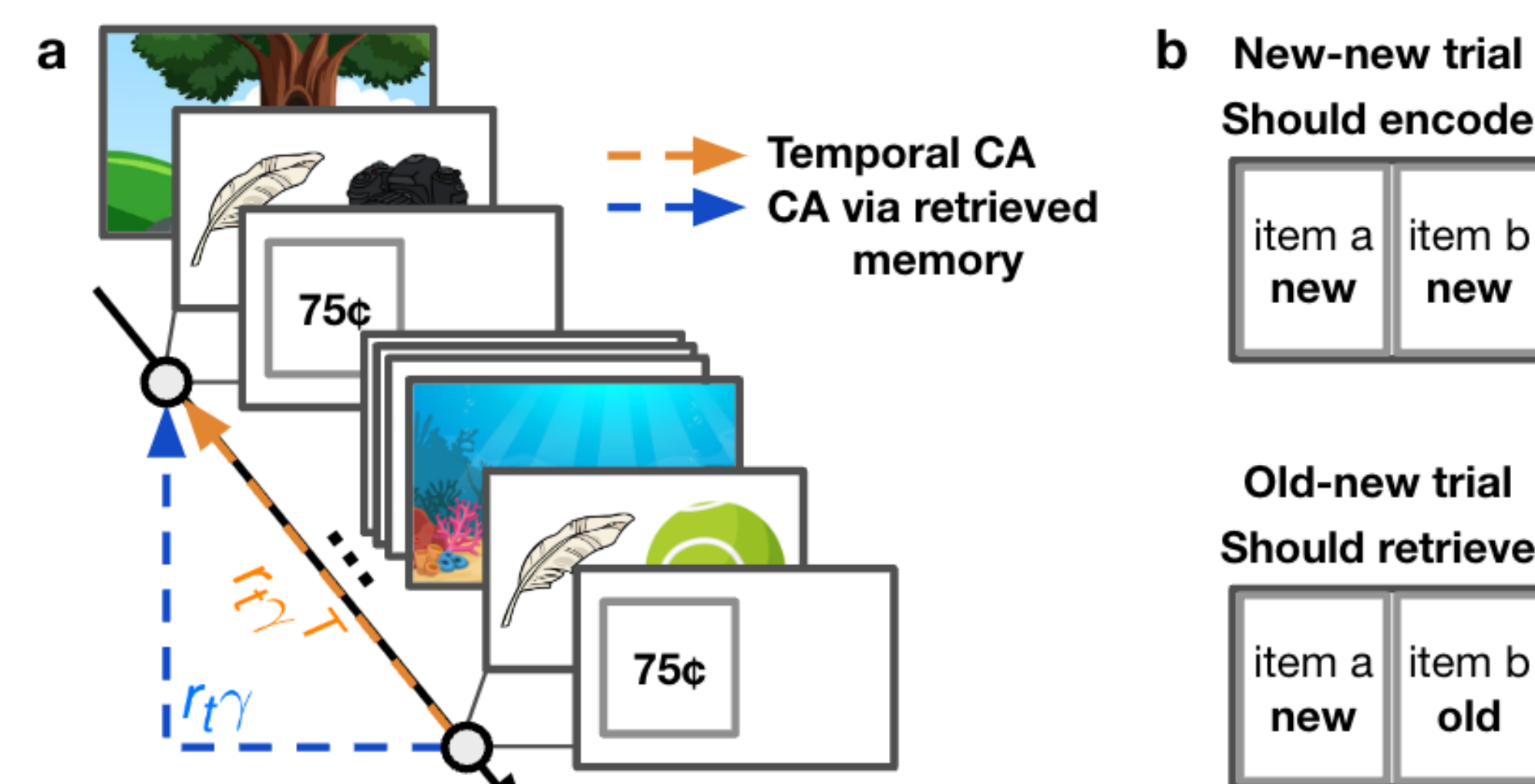


Figure 1: a) The task paradigm of Duncan et al. (2016) and an illustration of credit assignment (CA) via retrieved memory. b) Participants either choose between two new items or a new versus an old item.

- In Duncan et al. (2016; Fig.1a), subjects were instructed to choose the higher value item between two new items or an old versus a new item (Fig.1b; note that items had consistent values over time). It is optimal to encode in new-new trials and recall in old-new trials.
- Before every choice, a context image, either familiar or novel, was presented (Fig.1a).
- The findings show that choices were more influenced by past EMs in a familiar context, and EM encoding was better in a novel context.

## A neural net model of EM policy

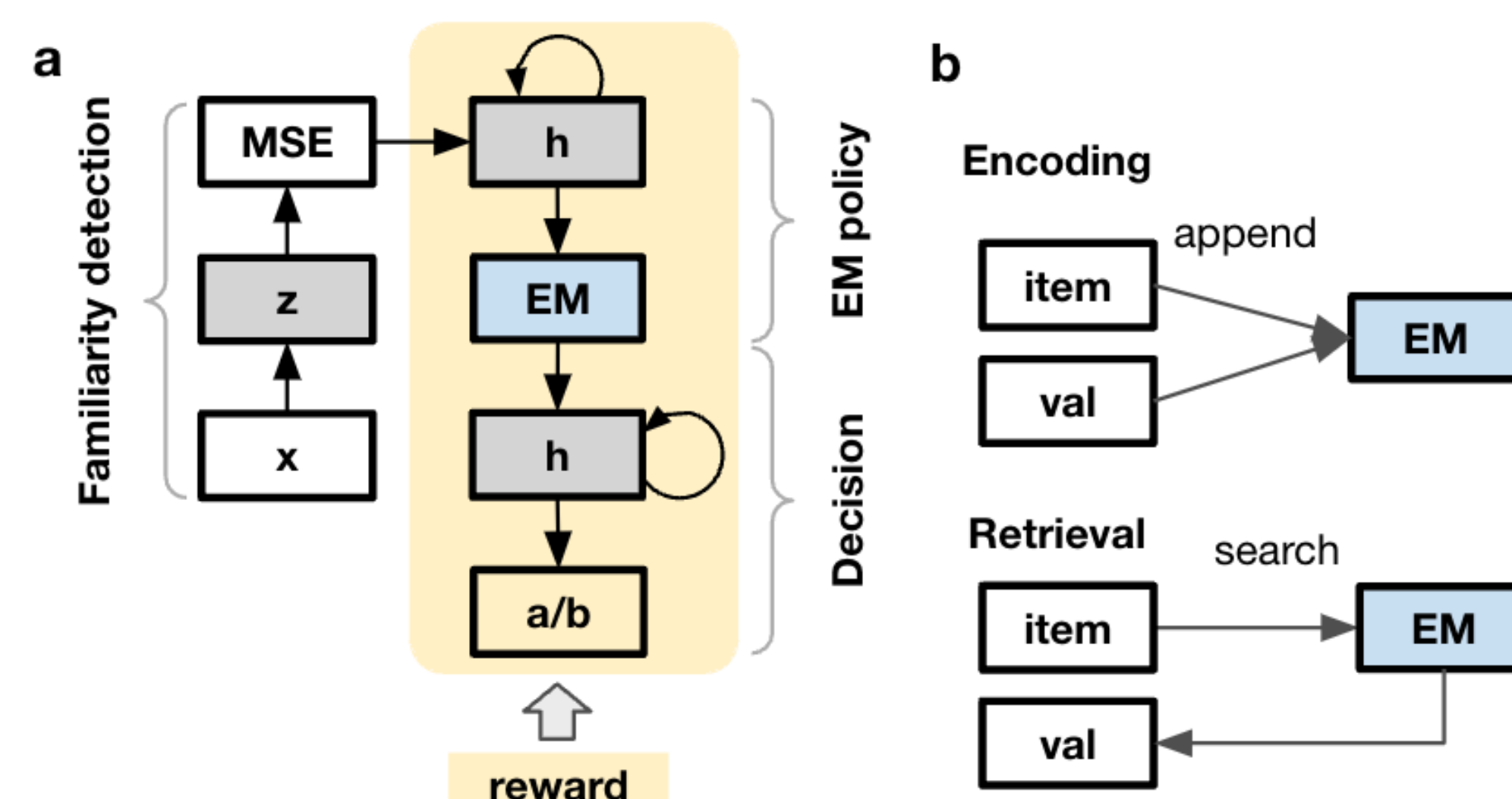


Figure 2: a) A neural net model of EM policy trained with RL. b) Episodic encoding and retrieval.

- Familiarity detector: an autoencoder that minimizes reconstruction mean squared error (MSE) so that lower MSE for experienced stimuli indicates greater familiarity.
- EM policy layer: an RNN that takes familiarity signal as input, outputs binary action – retrieve or encode (Fig. 2b); it learned to recall/encode when the stimulus is familiar/novel (Fig.3a).
- Decision layer: an RNN that takes the retrieved value to produce choices; it learned to choose higher-valued items (Fig. 3b).
- Learning: The EM policy and the decision layer were trained with RL. CA via retrieved memory reinforces encoding actions retrospectively when retrieval leads to rewards (Fig.1a); this is critical for learning an optimal encoding policy, as learning to encode involves long-range CA.

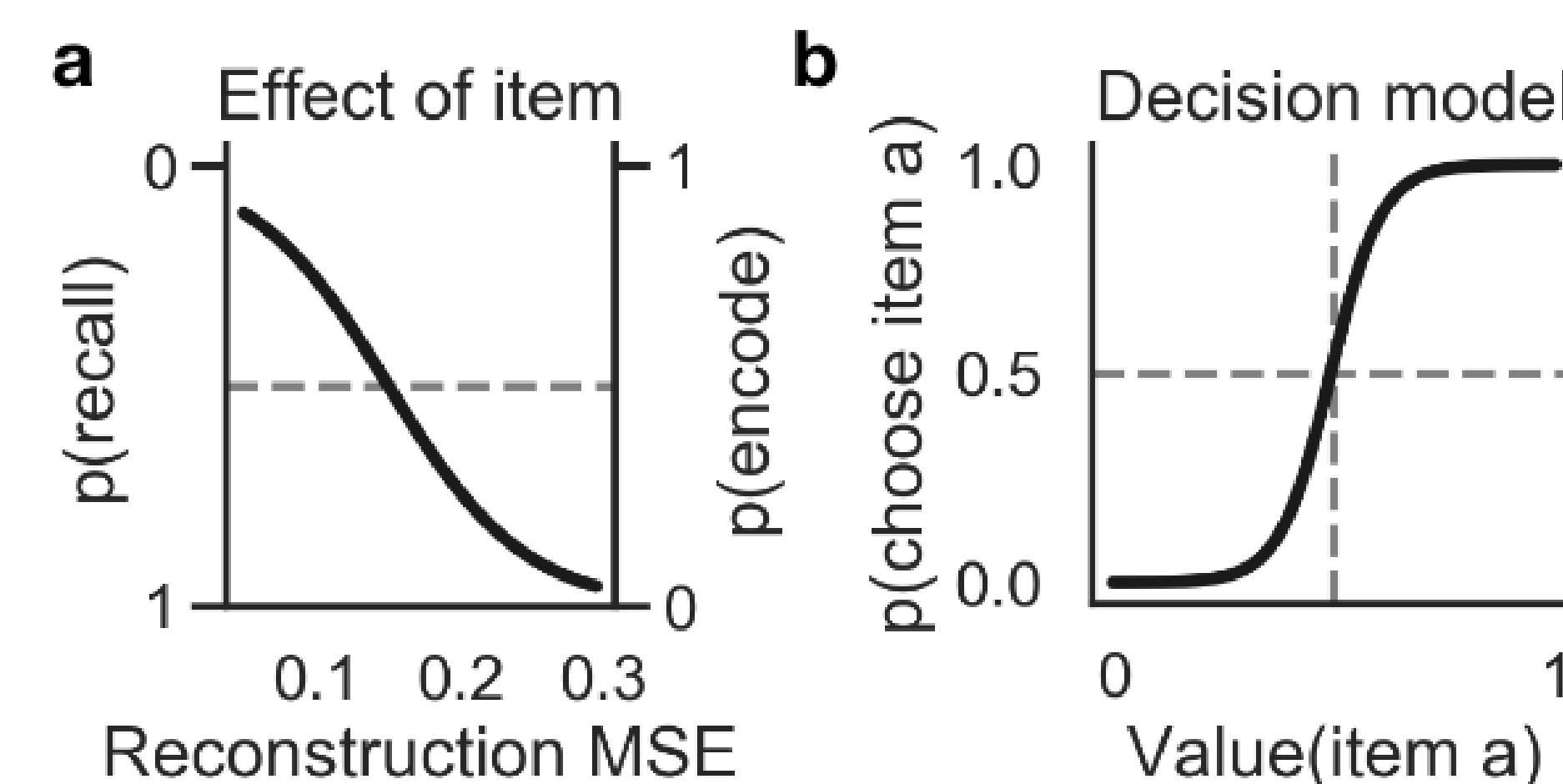


Figure 3: The probability of a) recall/encoding as a function of stimulus familiarity and b) choosing an item as a function of its value of trained models.

## Model: Duncan et al. (2016)

- Hypothesis: Context familiarity/novelty biases are adaptations to environmental autocorrelation.
- Method: To test the hypothesis, models were either trained in i) an autocorrelated environment, where familiar/novel stimuli precede other familiar/novel stimuli or ii) a non-autocorrelated environment.

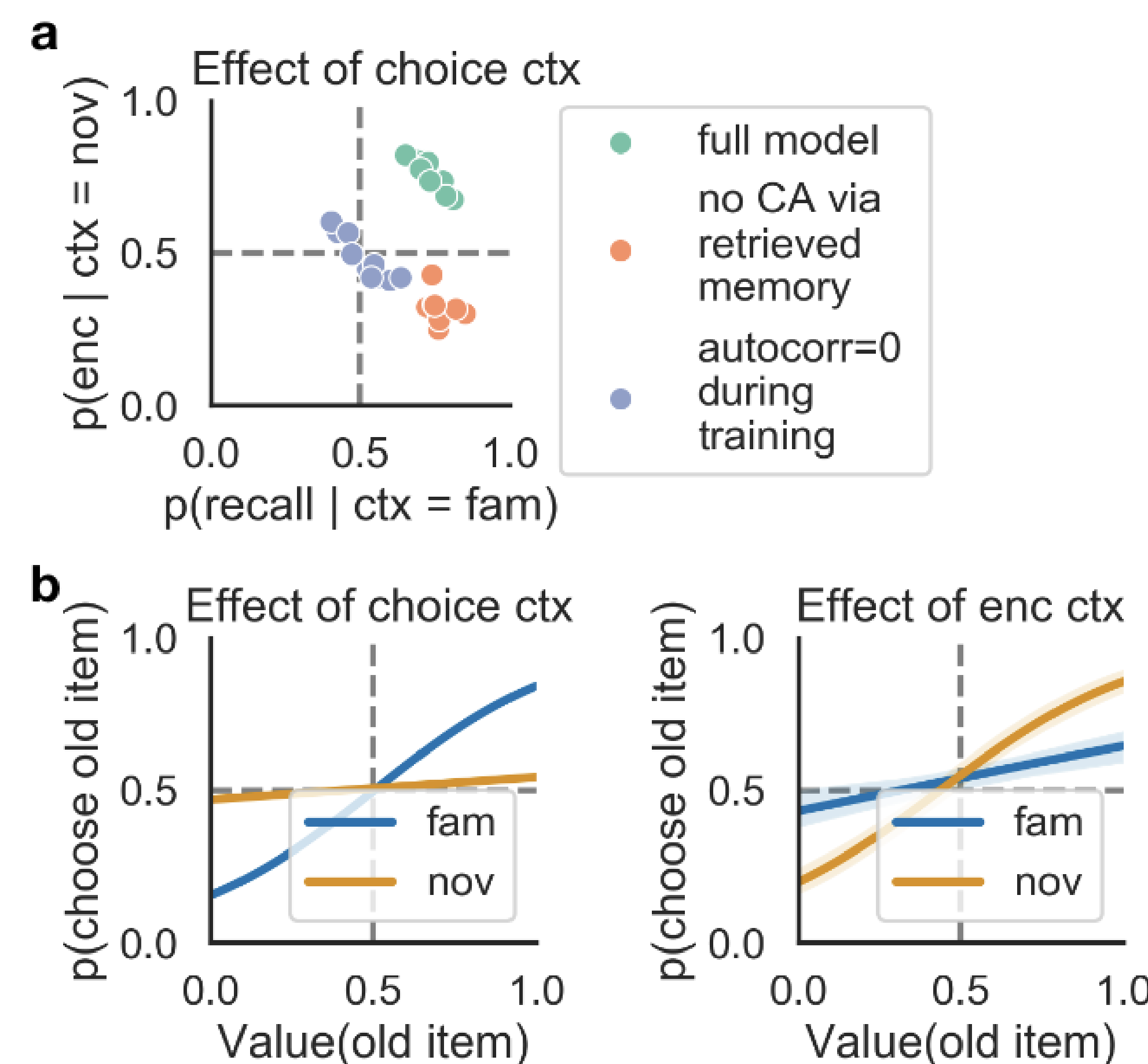


Figure 4: The influence of the familiarity/novelty of the preceding context on a) EM policy and b) decision policy.

Models trained in the autocorrelated environment ...

- ... were more likely to recall/encode when the preceding context was familiar/novel (Fig.4a); this was not the case for models trained in non-autocorrelated environment or models without CA via retrieved memory.
- ... captured the human data (Duncan et al. 2016; Fig.4b) – the level of influence of retrieved value was stronger i) when the preceding context was familiar and ii) when a novel context was presented at encoding, right before the value of the item was initially observed.

Models trained in the non-autocorrelated environment were not sensitive to context familiarity.

## Model: Duncan & Shohamy (2019)

- In Duncan & Shohamy (2019), subjects had to coordinate between episodic and incremental values to make optimal decisions (Fig. 5a).
- Again, models trained in the autocorrelated environment qualitatively captured the human data (Fig. 5b), showing that choices in a familiar/novel context were more/less episodic-dependent.

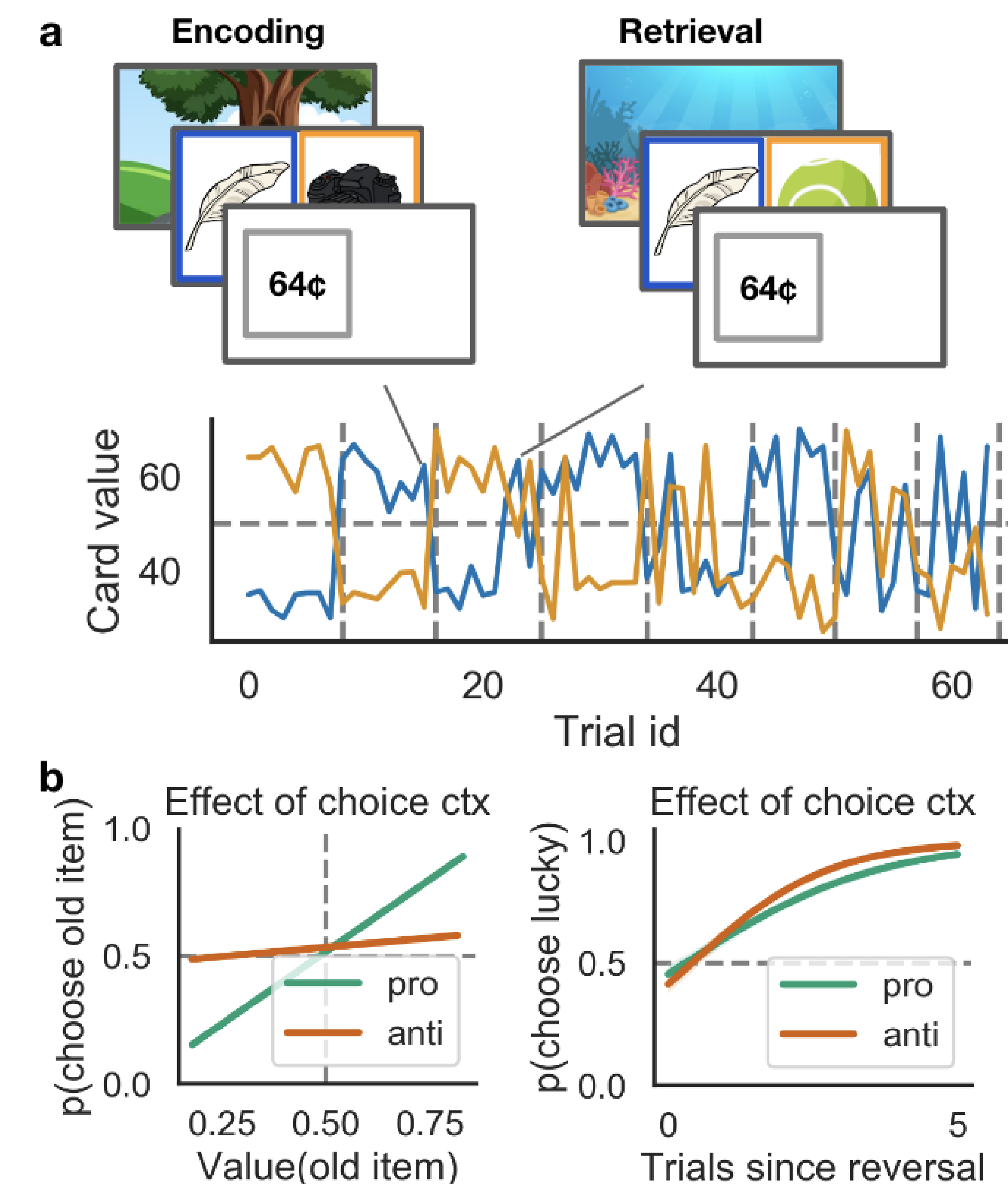


Figure 5: a) In the Duncan & Shohamy (2019) task, each card has a color (blue or orange). At any moment, cards from one of the colors have a higher mean value (the "lucky deck"). b) The modeling results. \*pro: familiar context -> old-new trial, novel context -> new-new trial; anti: novel context -> old-new trial, familiar context => new-new trial.

## References & Acknowledgement

**References:** [1] Duncan et al. (2012). Science; [2] Duncan et al. (2016). J Cog Neuro; [3] Duncan & Shohamy (2019). JEP.G; [4] Anderson & Schooler (1991). Psych Science; [5] Honey et al. (2023). Current Direction in Psych Science; [6] Hasselmo et al. (1996). Hippocampus  
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