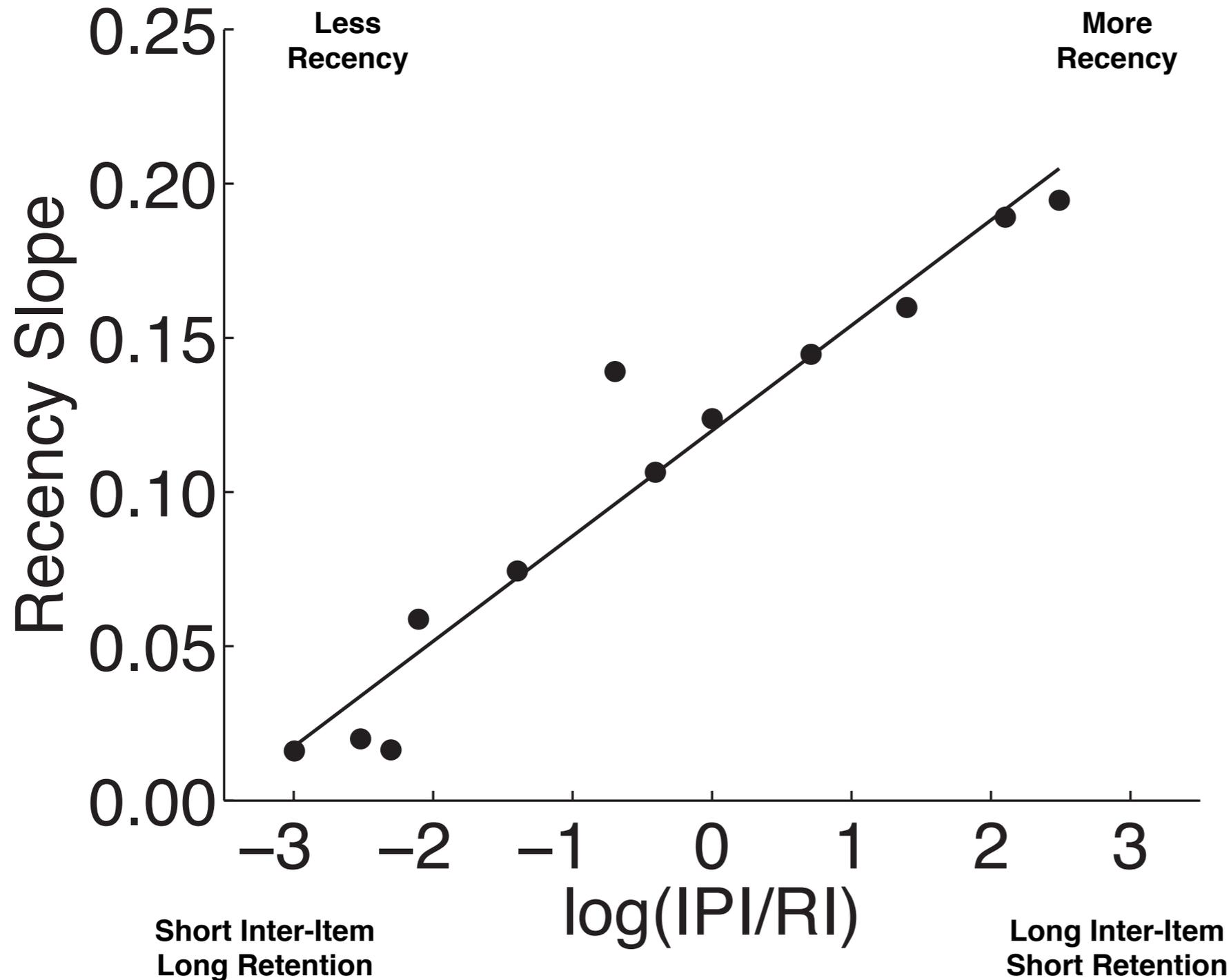


# Recap

- Dual store models (SAM)
  - Short term store
  - Long term store
  - Recency, primacy, contiguity
  - Continual distractor free recall

# The Ratio Rule



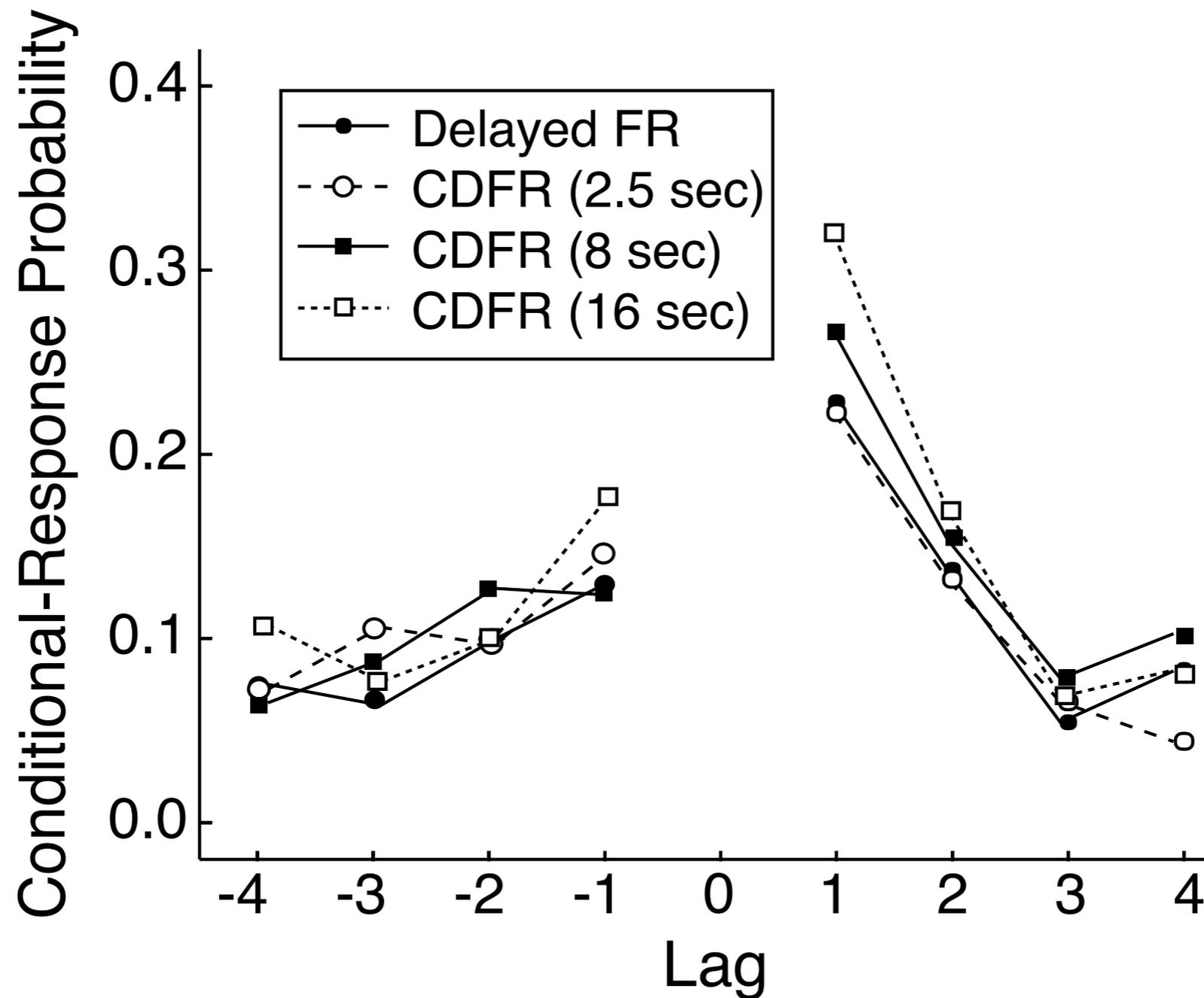
Here, log is "natural log" (base e).

At 1 on the x-axis, the IPI is ~2.7 times longer than the RI.

At -1 on the x-axis, the RI is ~2.7 times longer than the IPI.

At 2 on the x-axis, the IPI is ~7.4 times longer than the RI.

# Long-range contiguity



# Long-range contiguity

- Studied lists (minutes)
- Dinners you've eaten (days)
- People you've had long conversations with (hours to years)
- Teachers you've had in school (years to decades)

# Challenges to SAM

- The long-term recency and long-range contiguity effects present a critical problem for SAM
- If the short term buffer gets filled by the distractor in delayed free recall, why would it not get filled by the same distractor in continual distractor free recall?



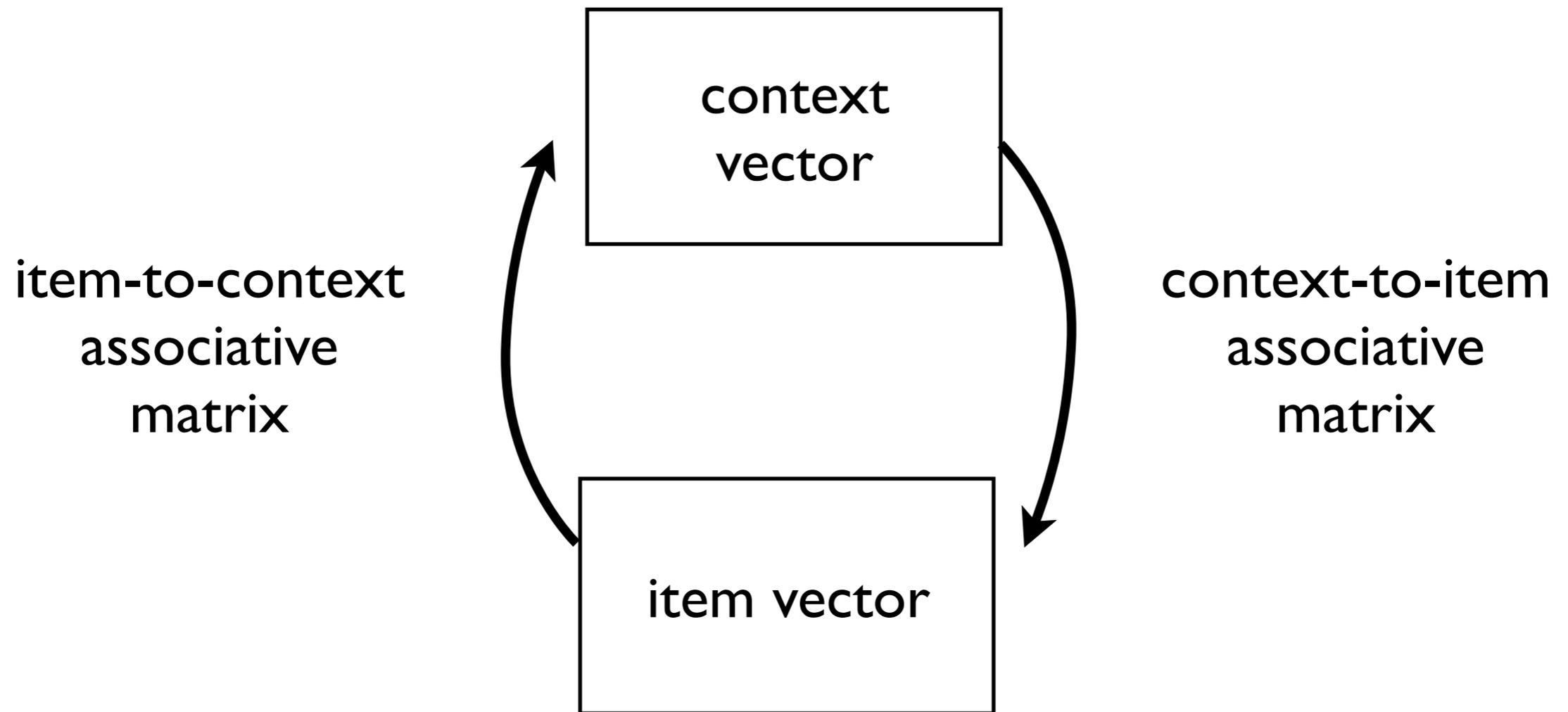
# Challenges to SAM

- Another challenge: the Law of Parsimony tells us to select the simpler model, given two alternatives that explain the data equally well
- Do we “need” to assume there are separate short term and long term memory systems to explain free recall data?

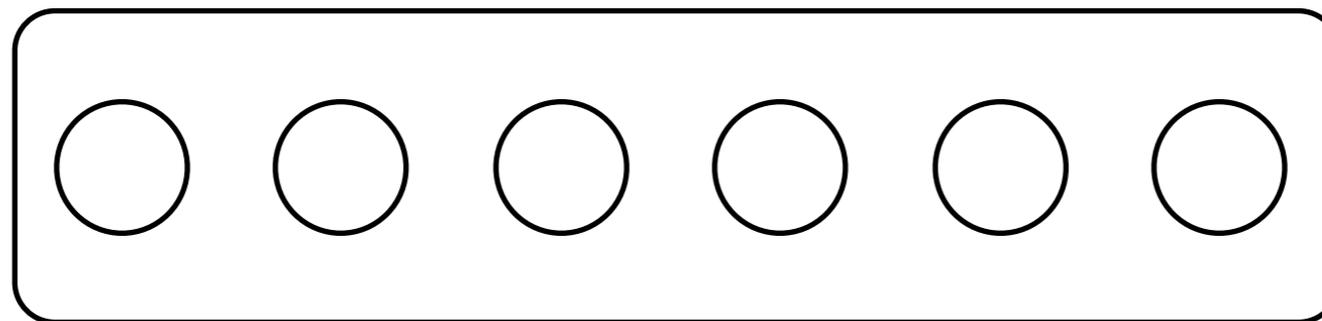
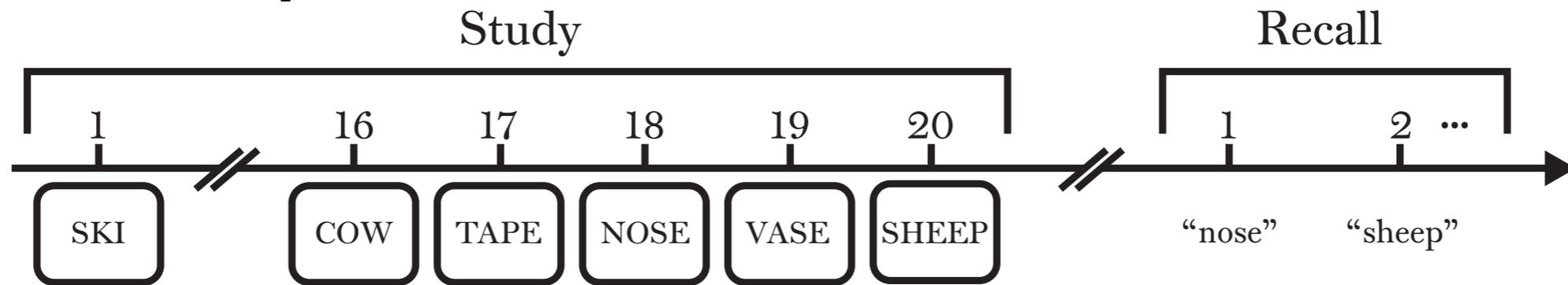
# Single store models

- Single store models (AKA "context-based models") use **context** as a single mechanism that you use to search through your memories

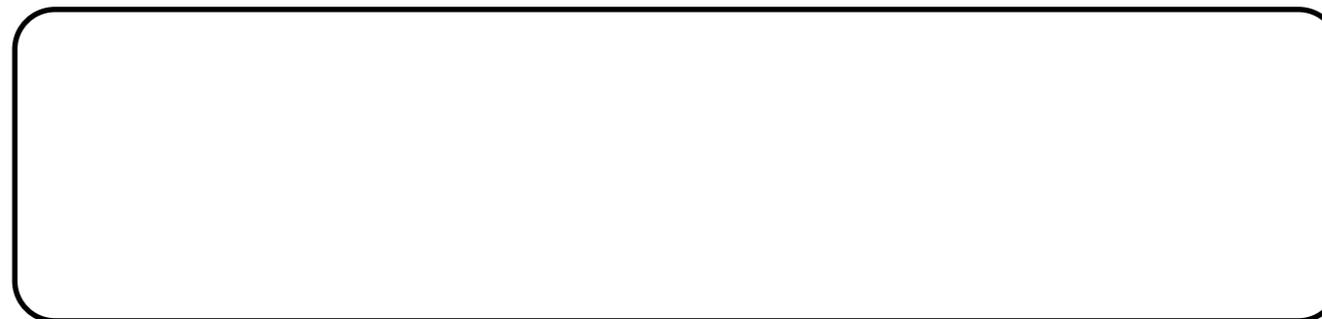
# Single store models



# Temporal context model

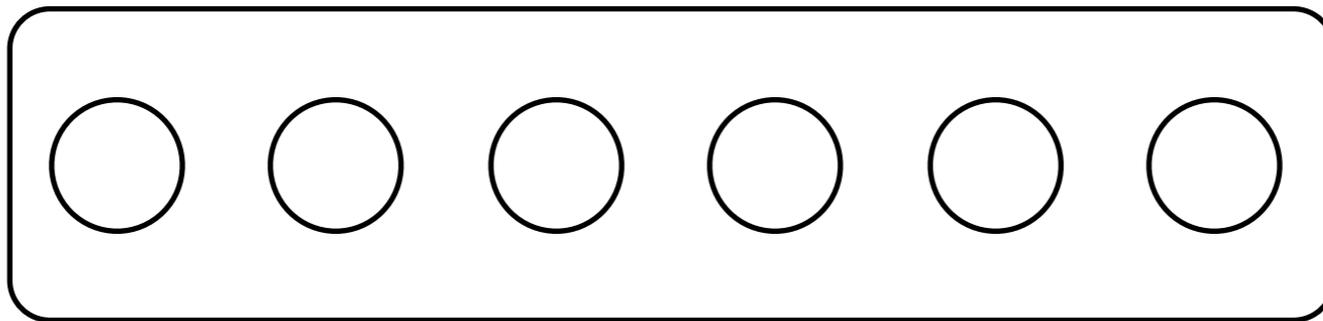
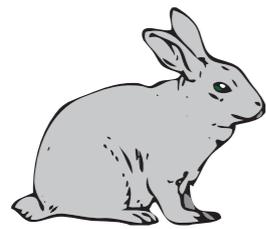
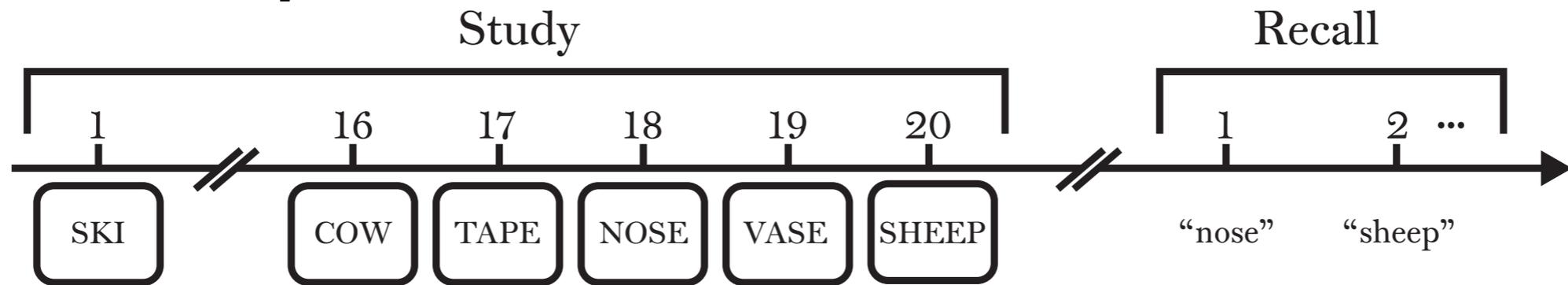


ITEMS

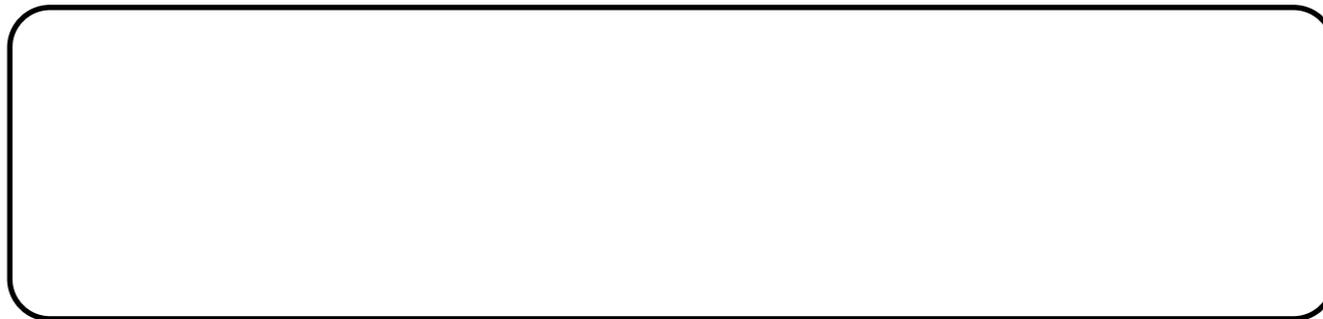


CONTEXT

# Temporal context model

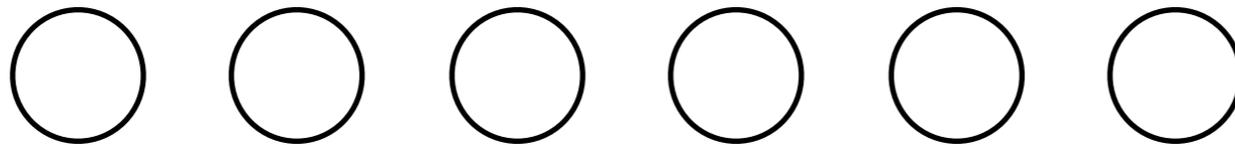
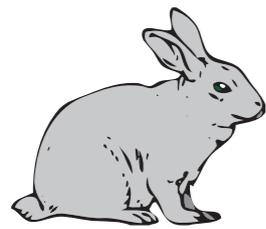
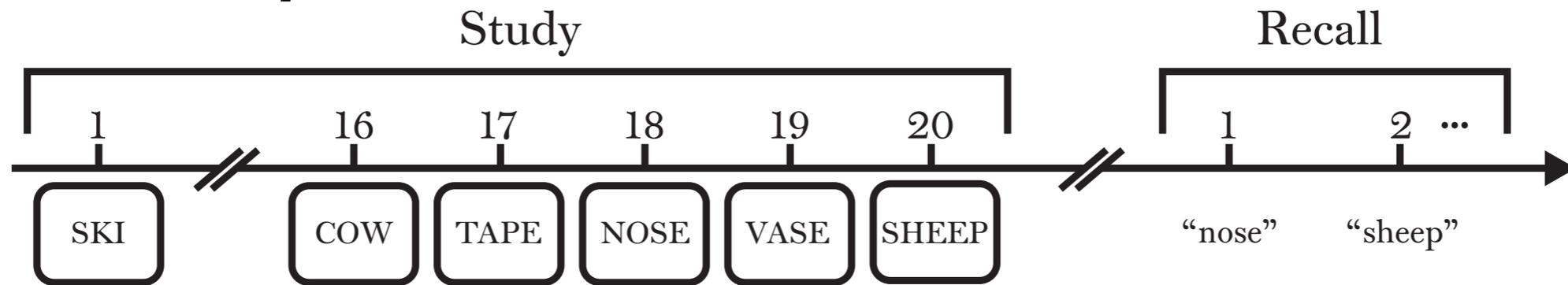


ITEMS

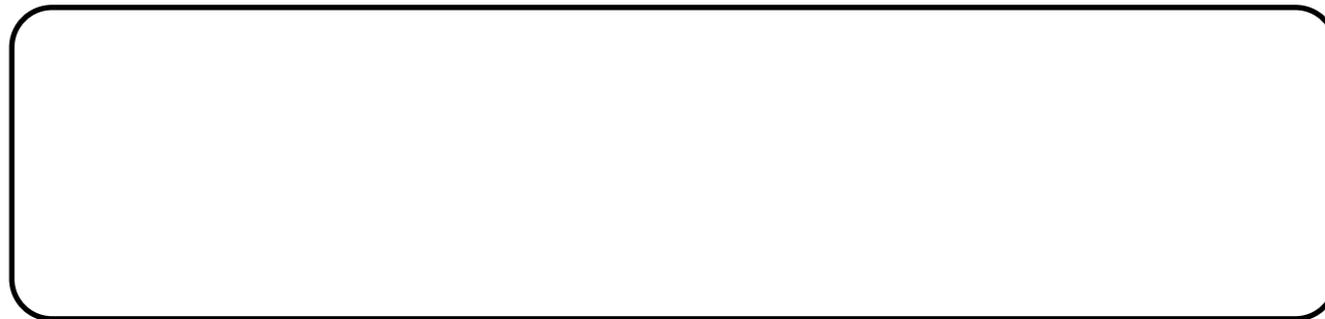
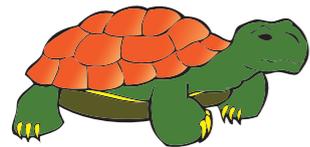


CONTEXT

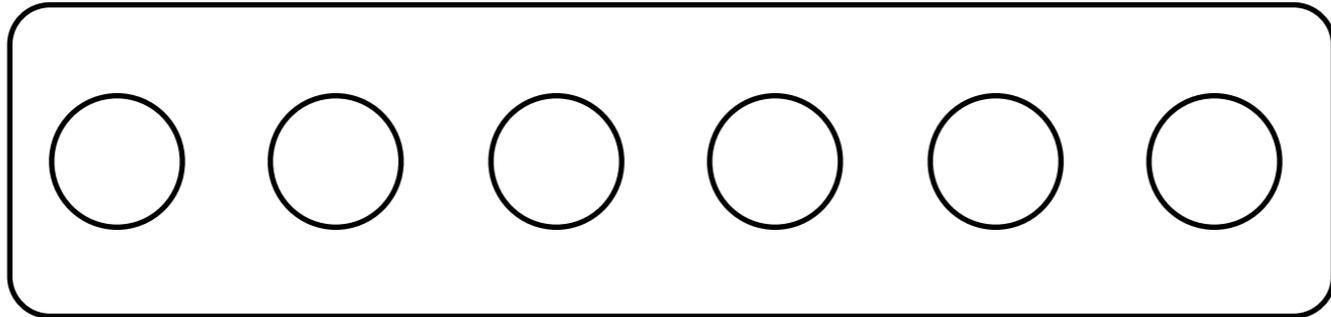
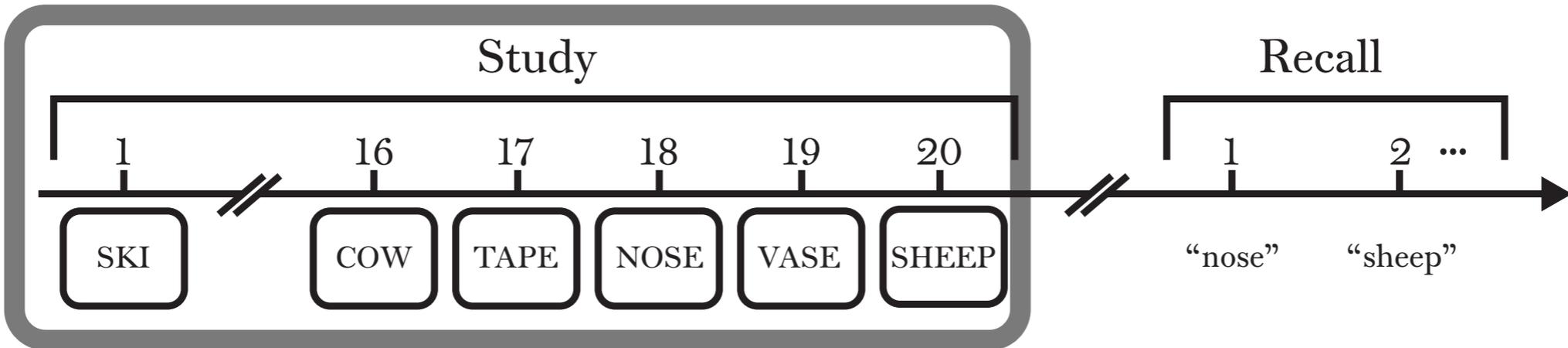
# Temporal context model



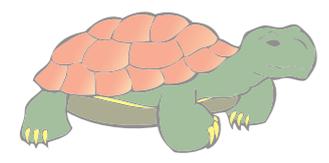
ITEMS



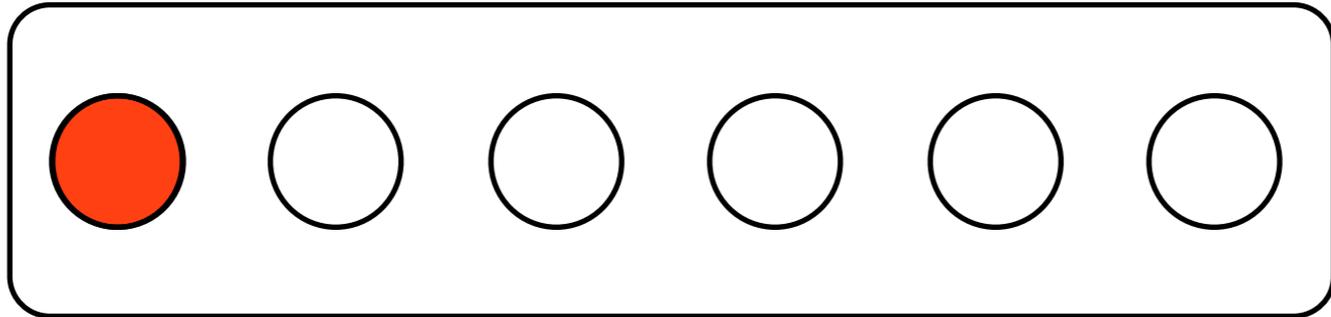
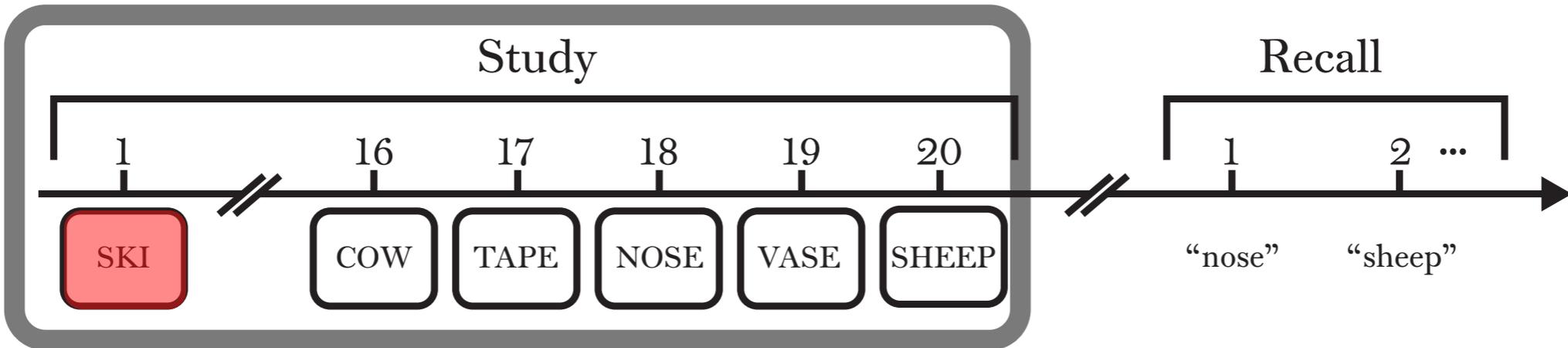
CONTEXT



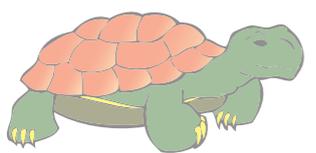
ITEMS



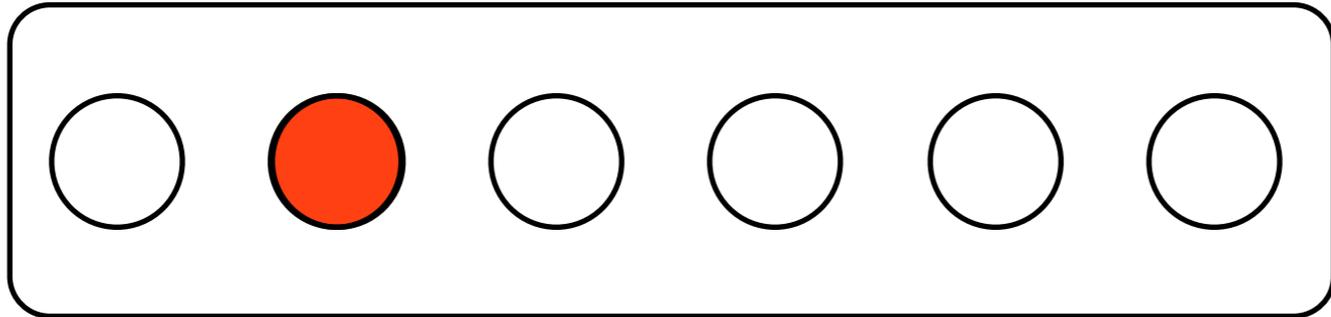
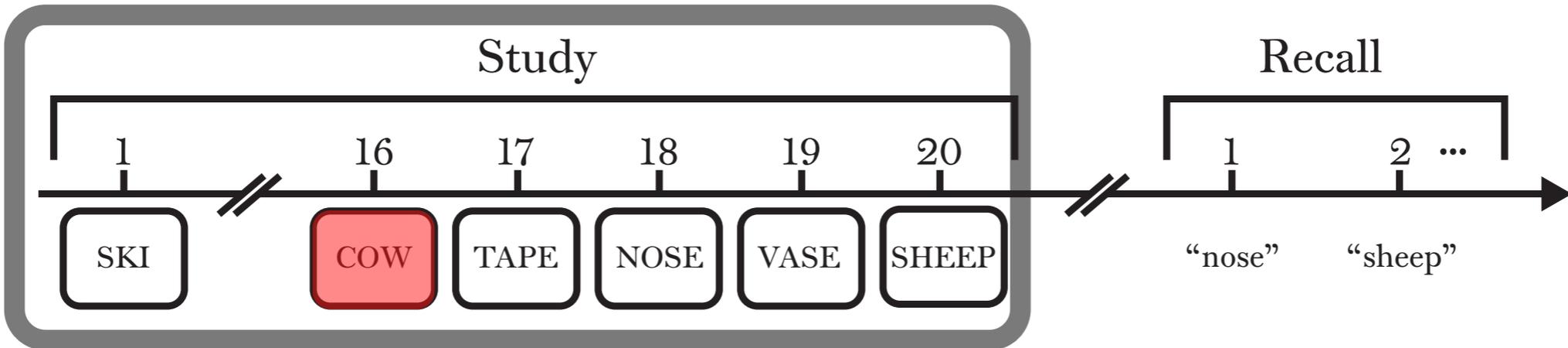
CONTEXT



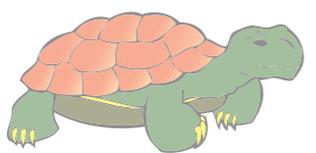
ITEMS



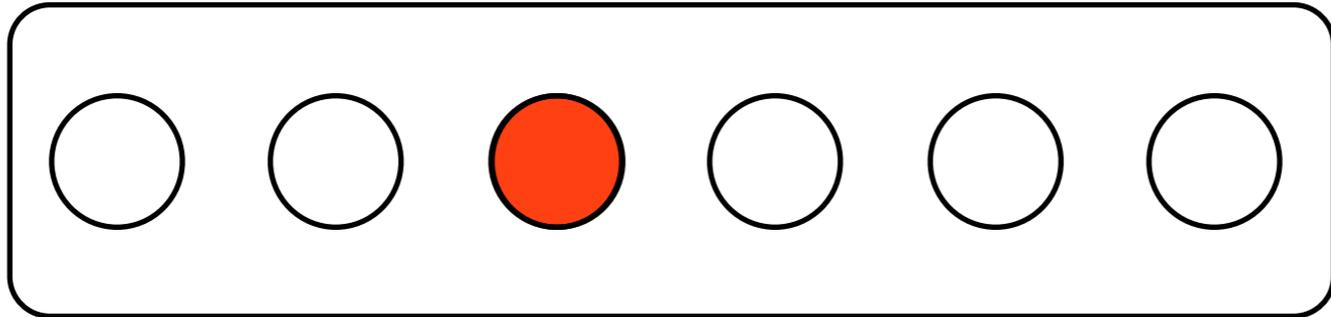
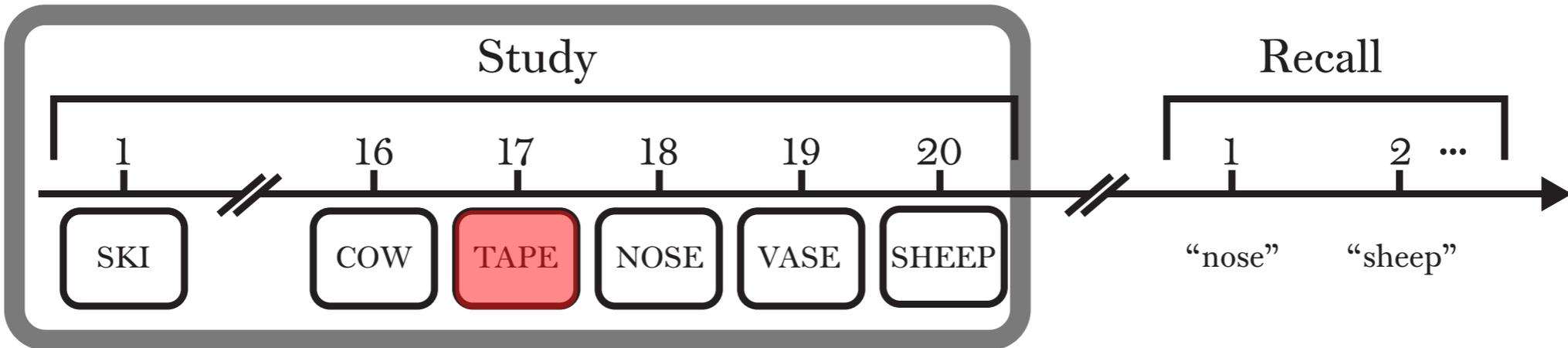
CONTEXT



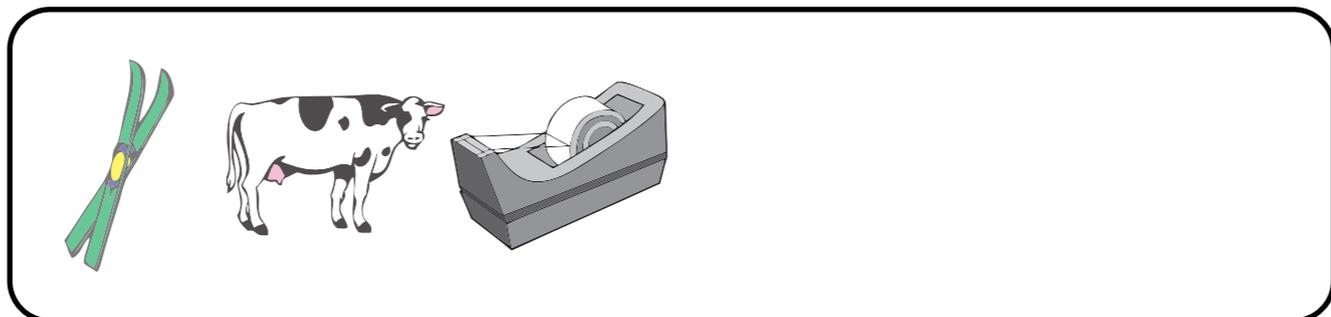
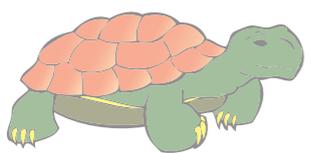
ITEMS



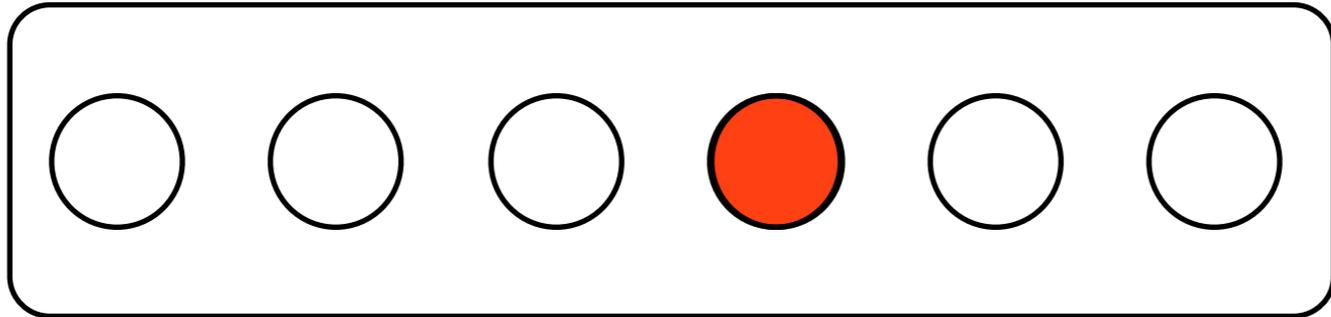
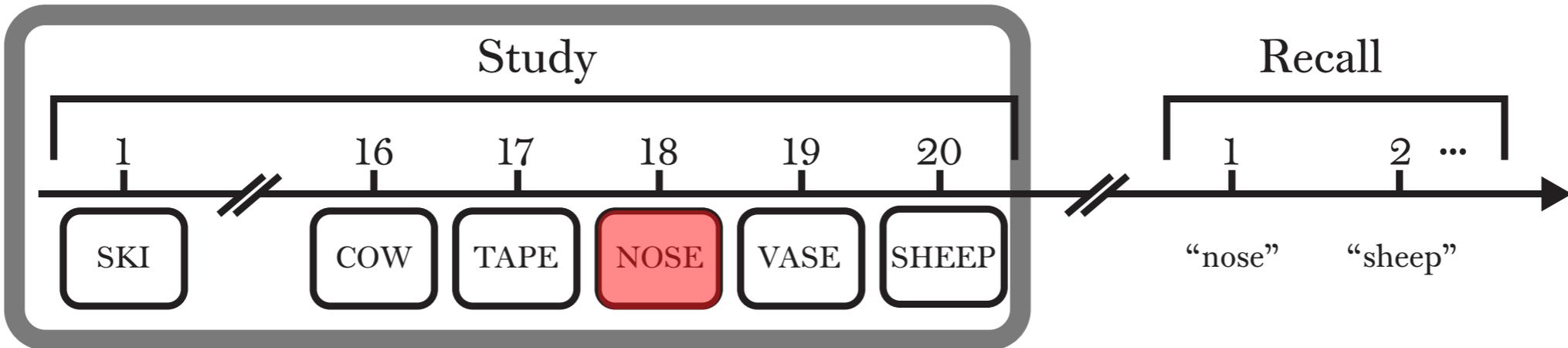
CONTEXT



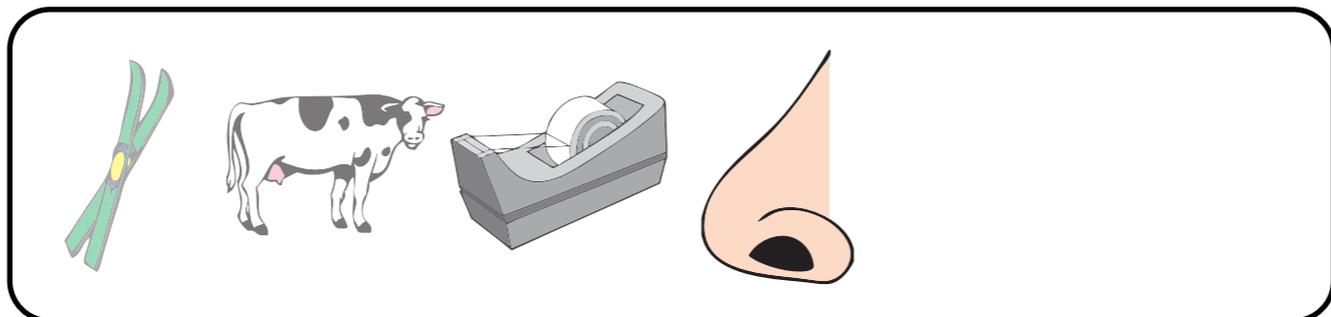
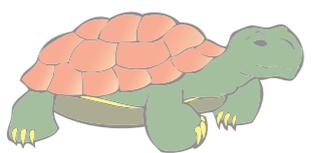
ITEMS



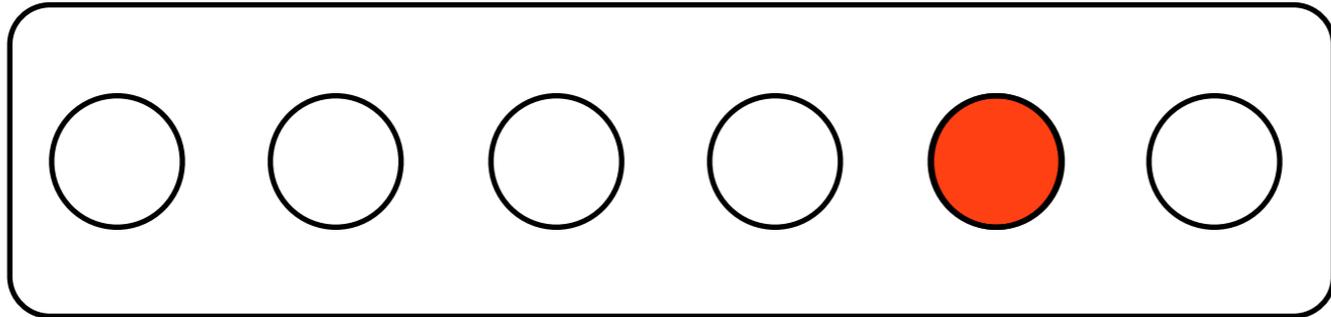
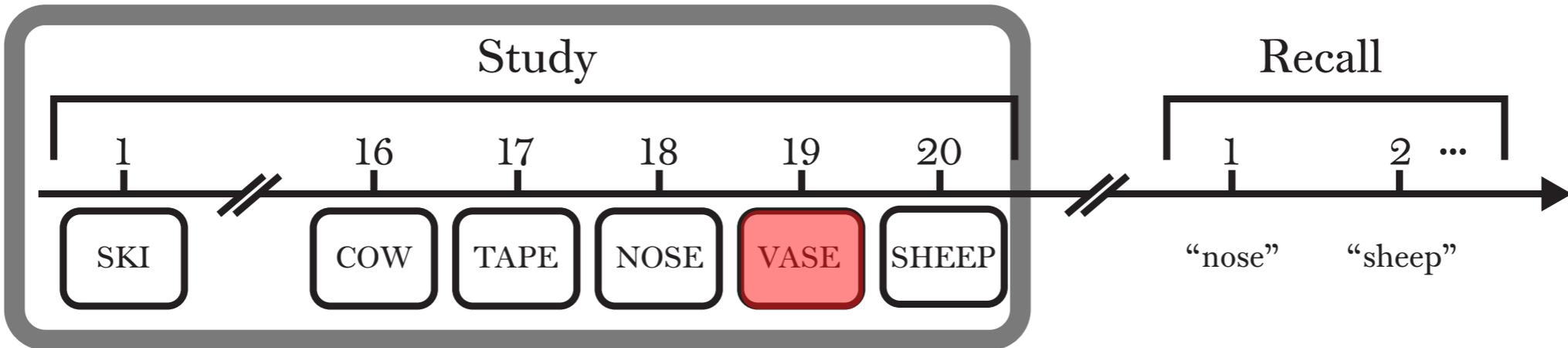
CONTEXT



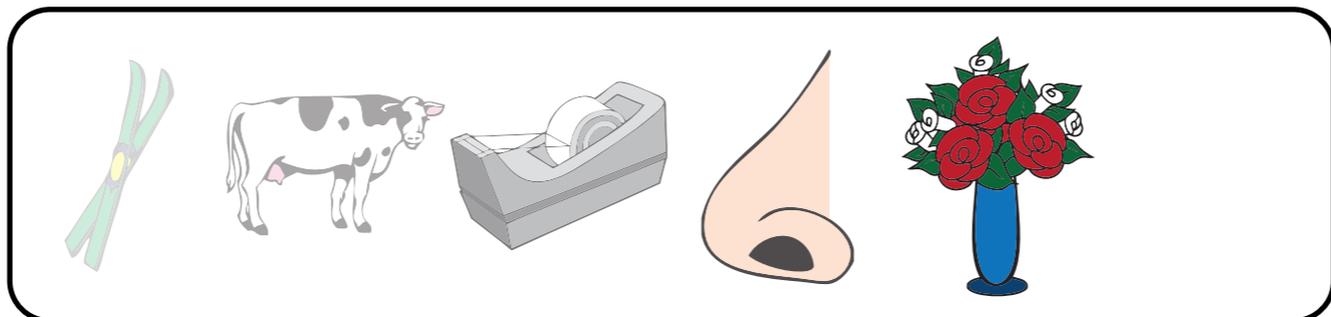
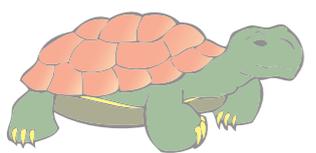
ITEMS



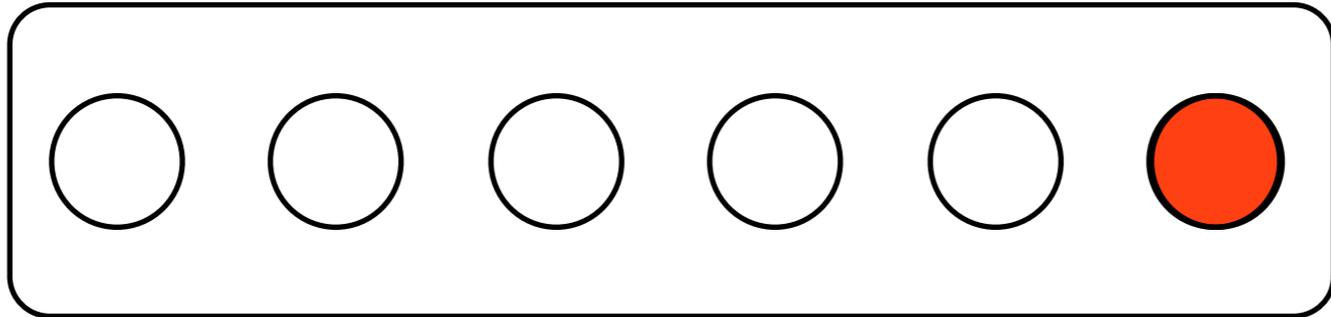
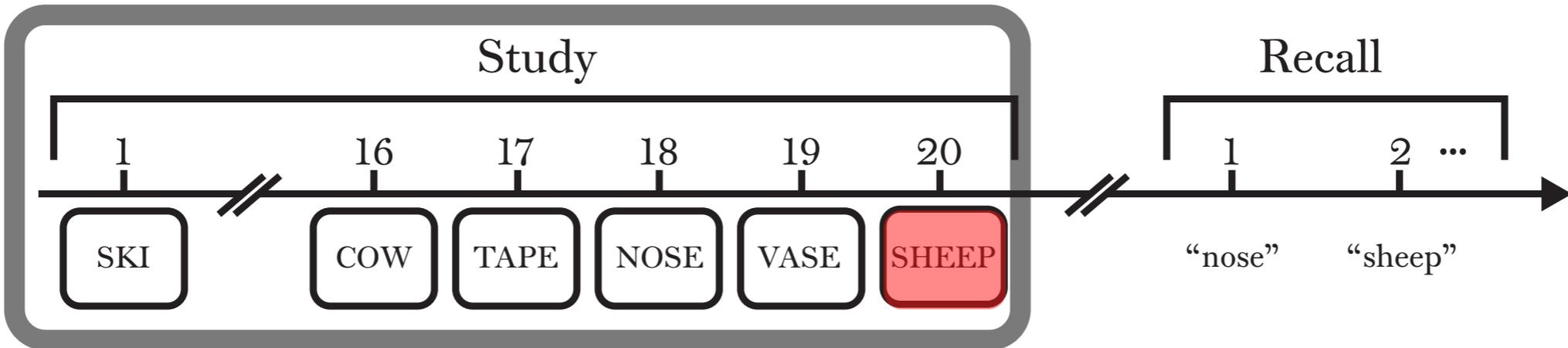
CONTEXT



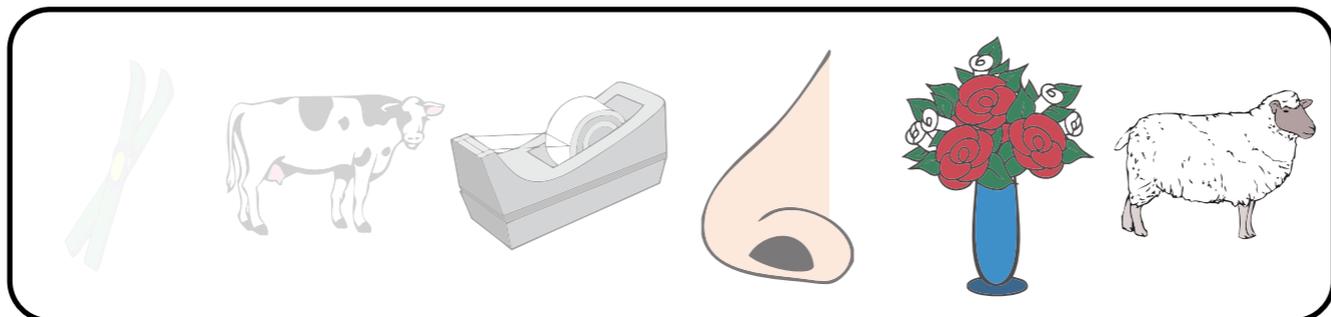
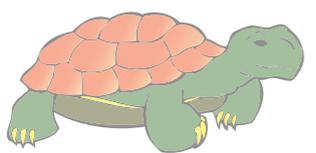
ITEMS



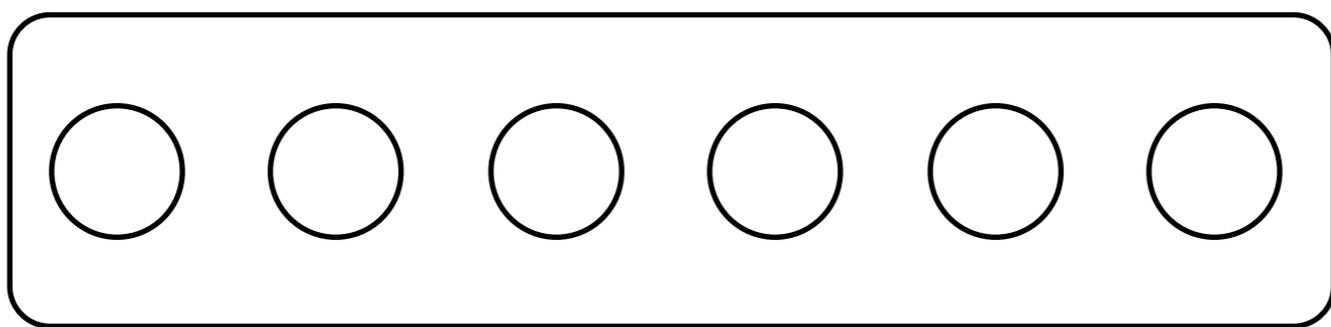
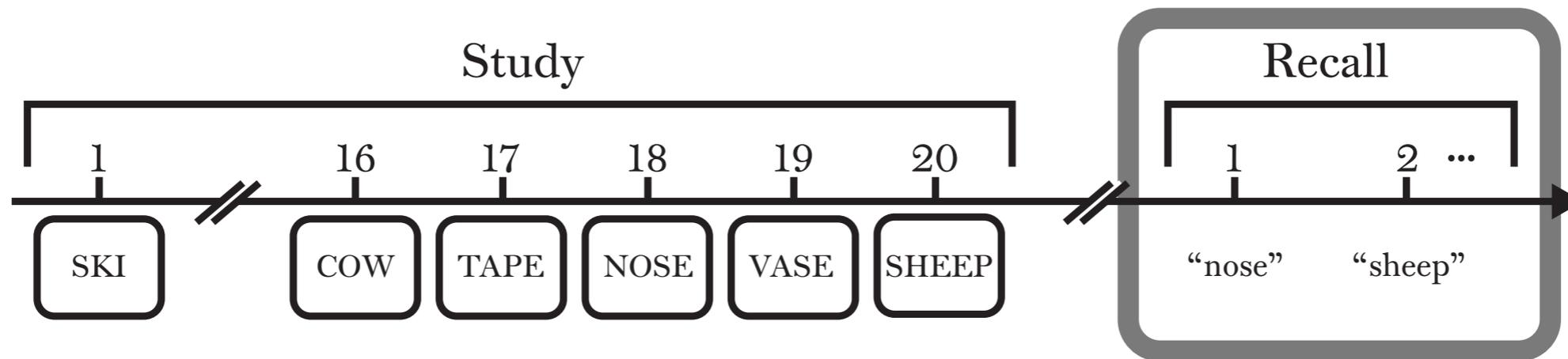
CONTEXT



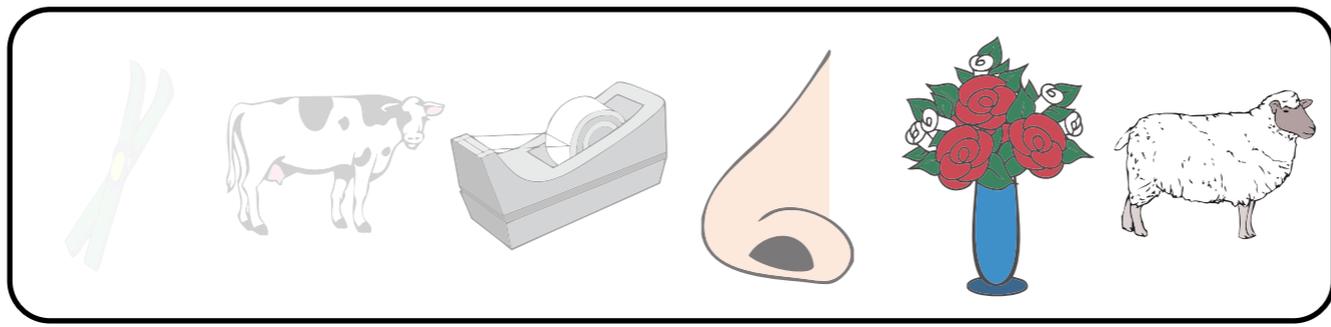
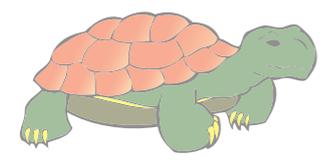
ITEMS



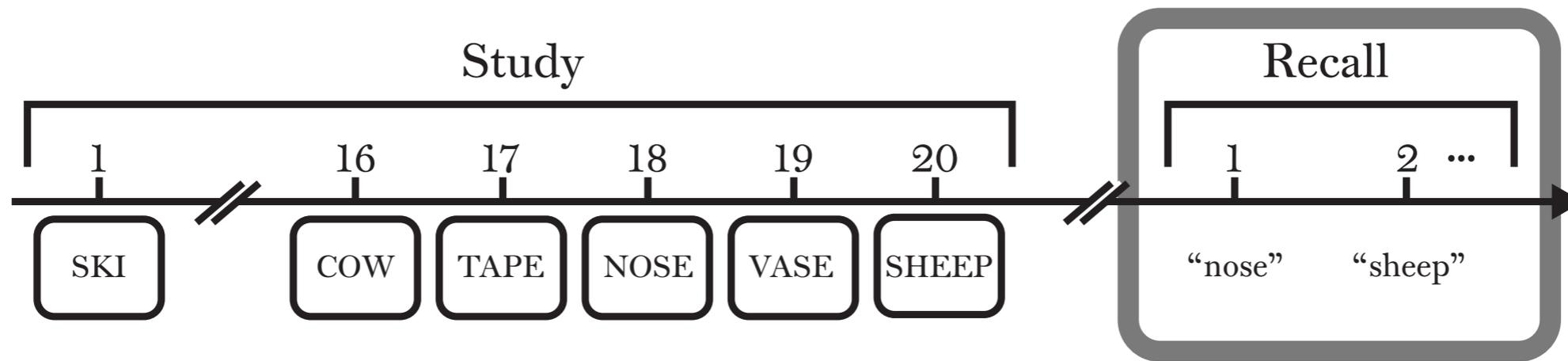
CONTEXT



ITEMS



CONTEXT



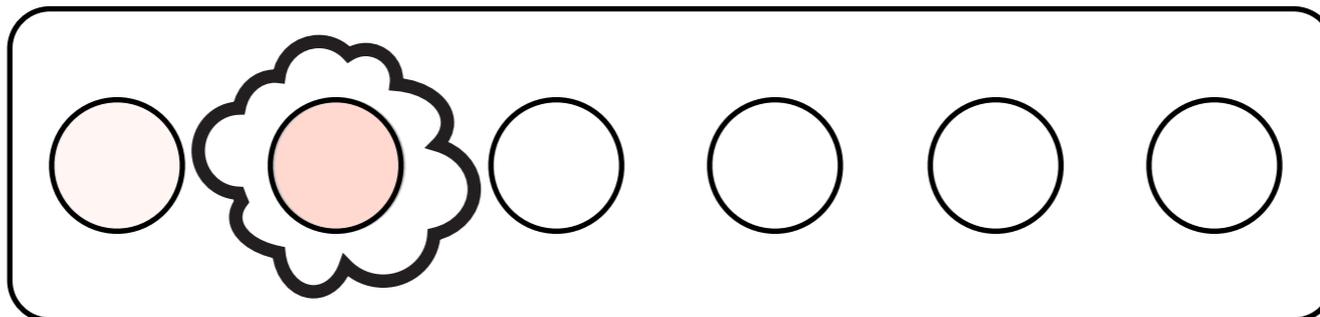
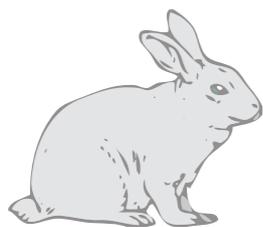
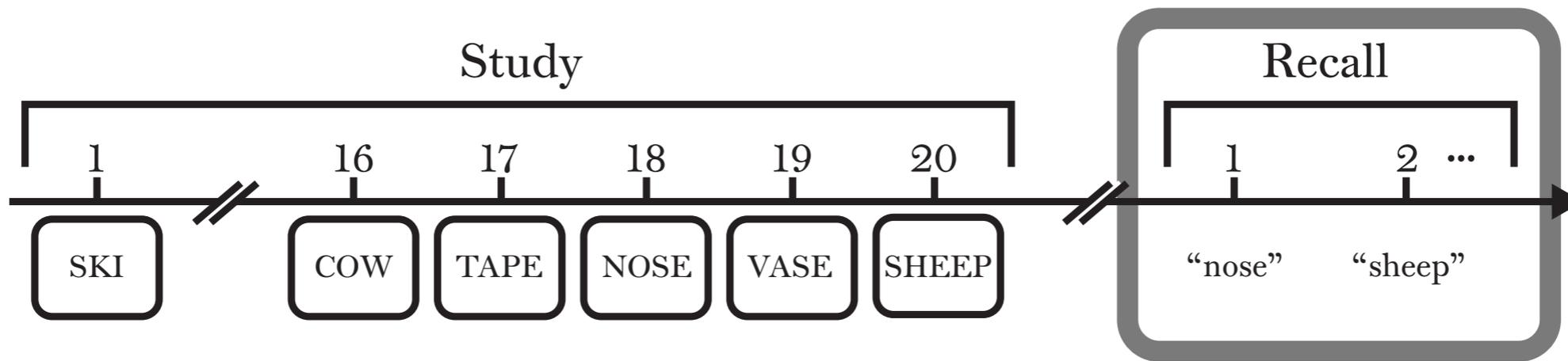
ITEMS

A row of six items for recall: a rabbit, a flower, and four empty circles.

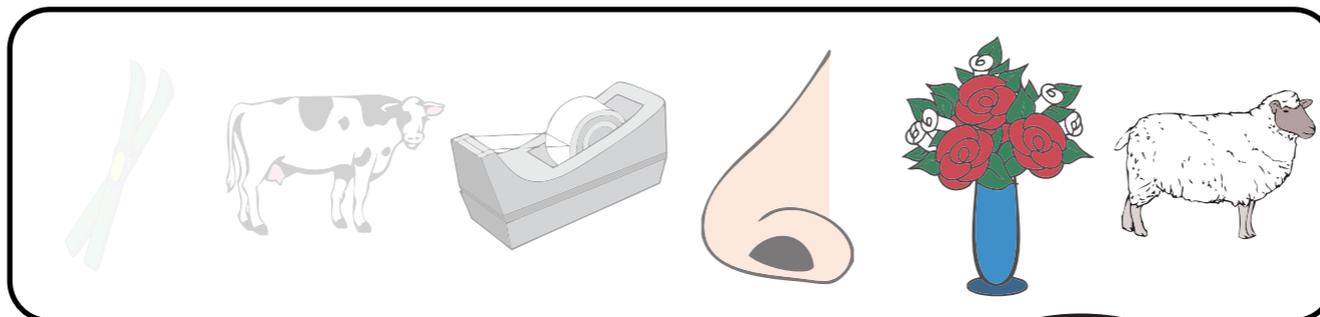
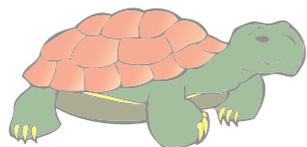
CONTEXT

A row of six context images: a turtle, a pair of glasses, a cow, a tape dispenser, a nose, a vase of flowers, and a sheep.

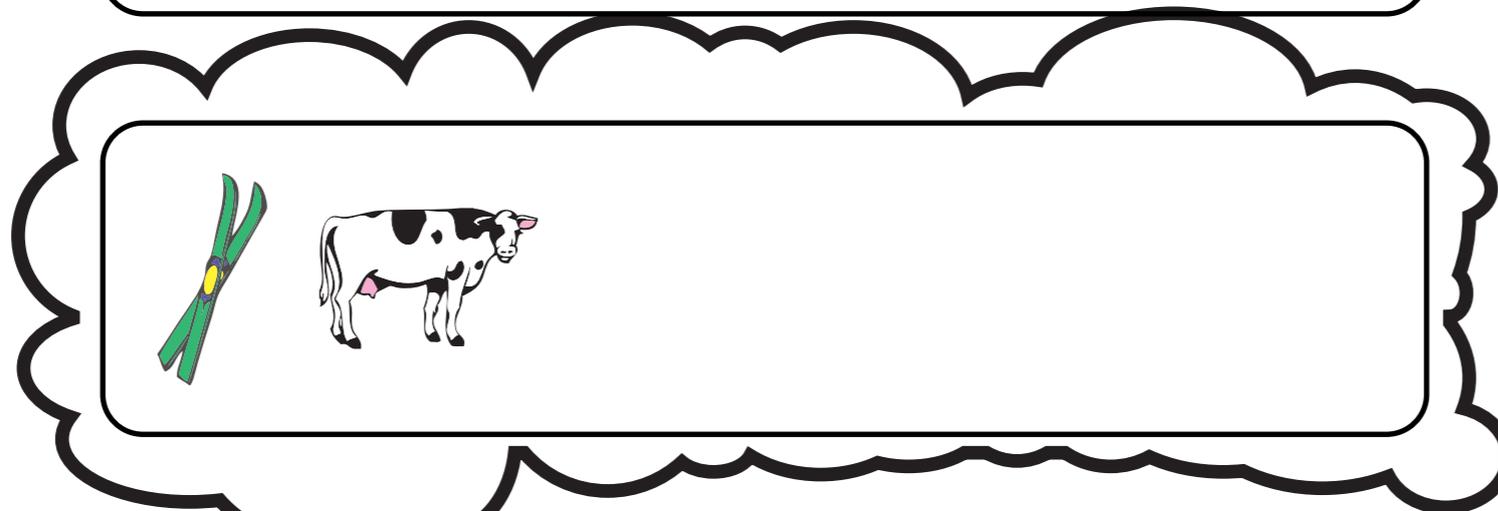
A thought bubble containing a pair of green ski poles.

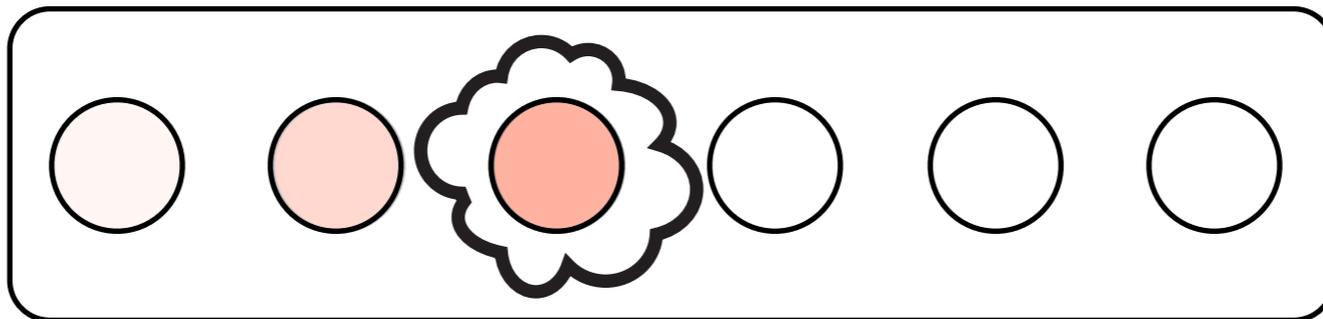
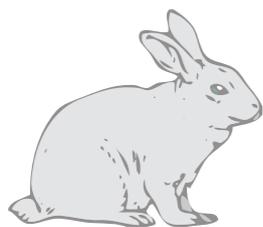
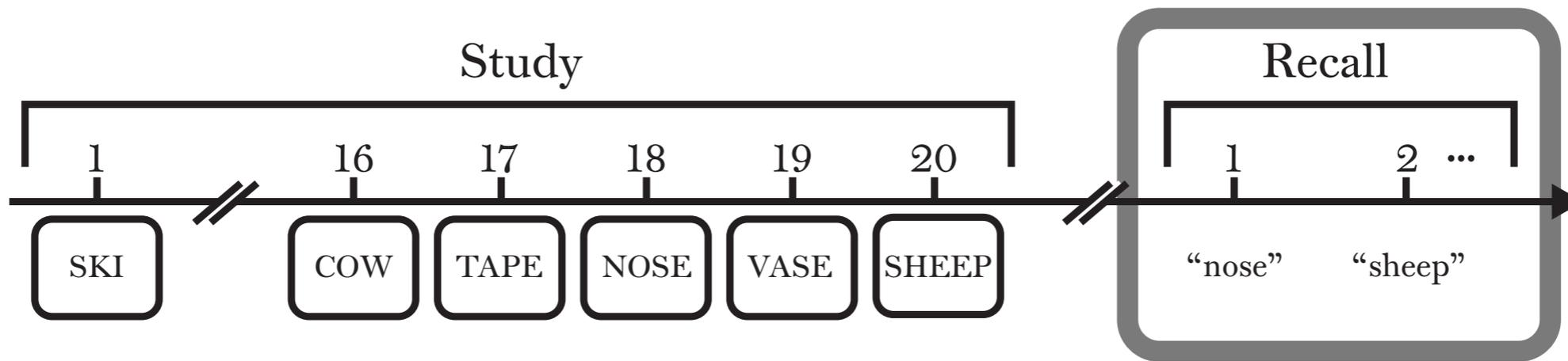


ITEMS

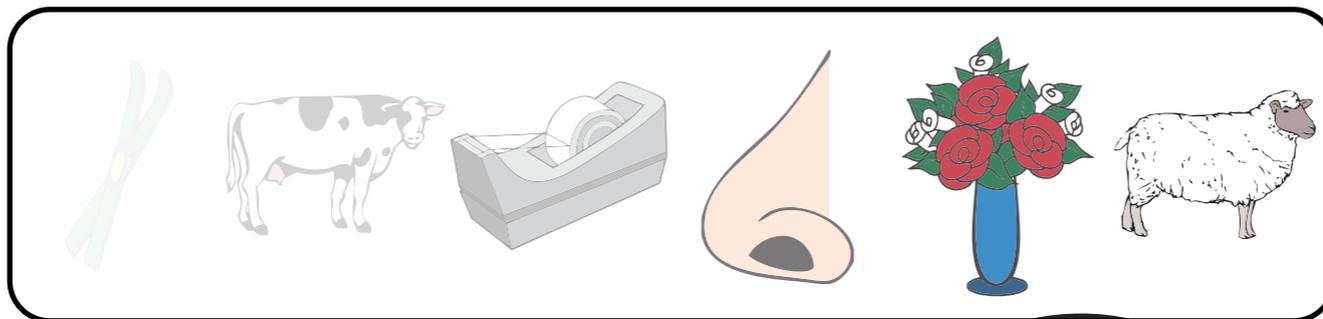
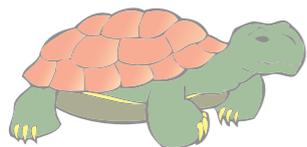


CONTEXT

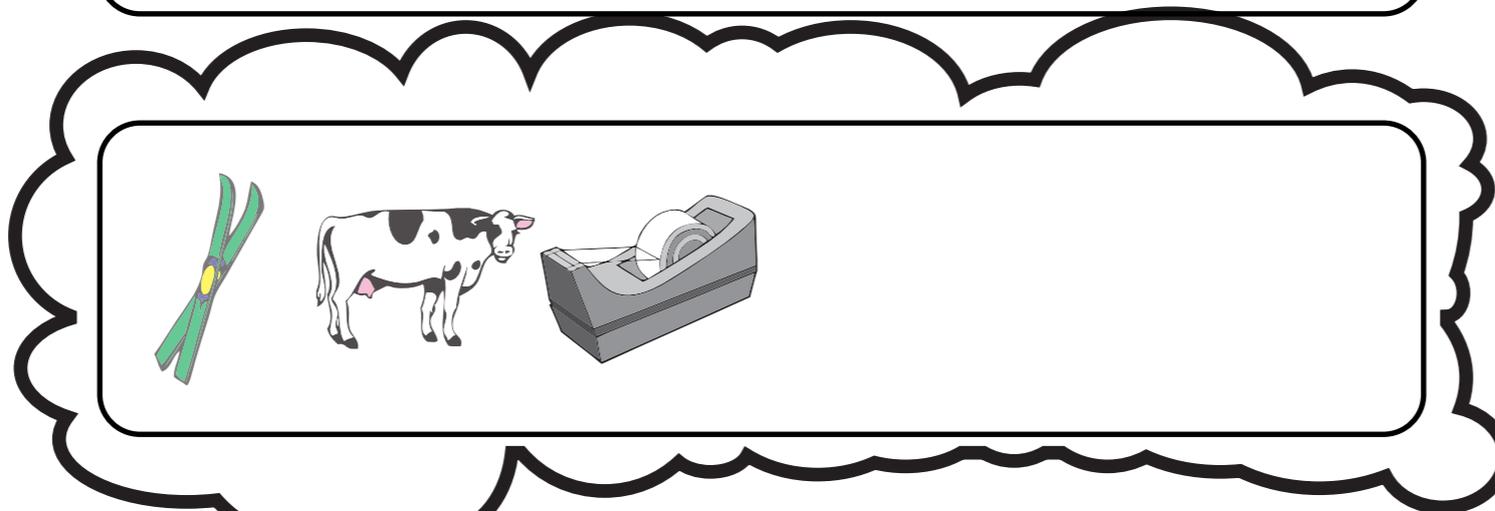


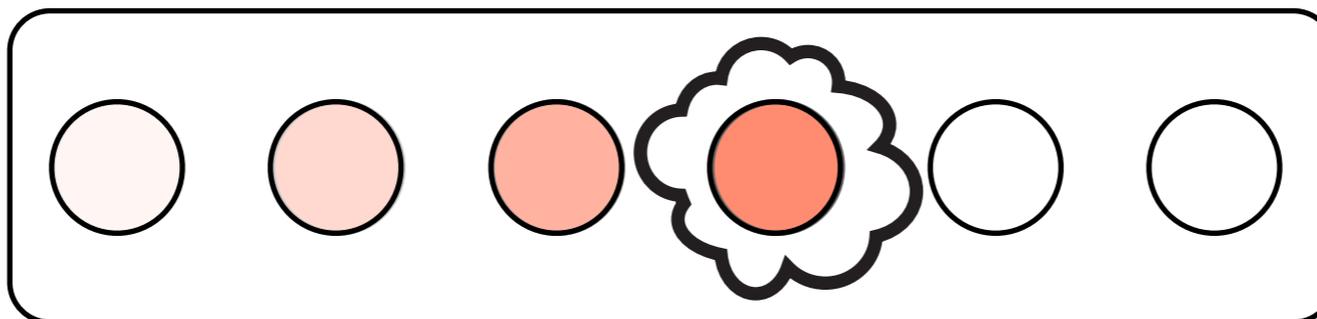
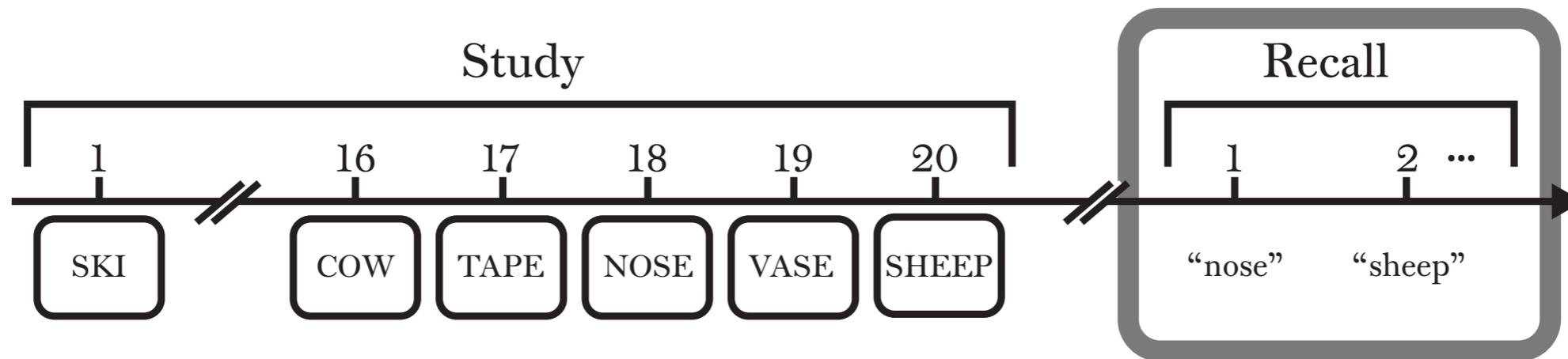


ITEMS

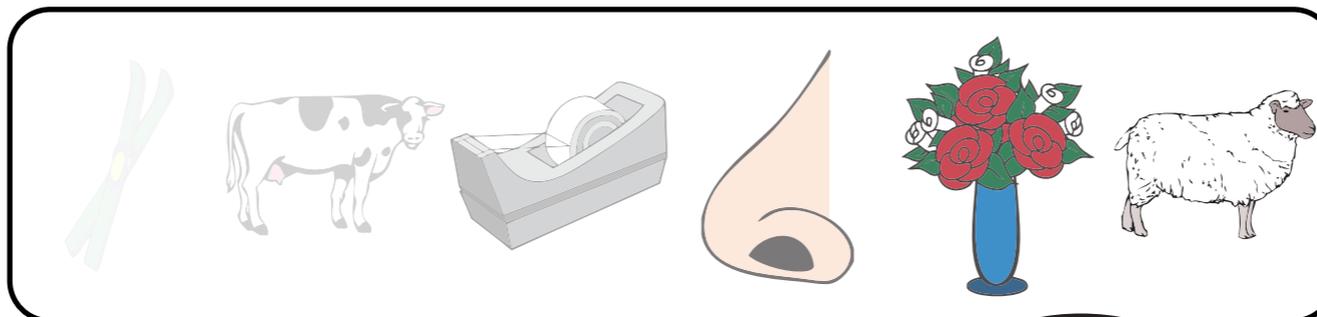
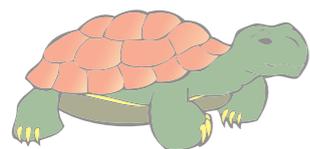


CONTEXT

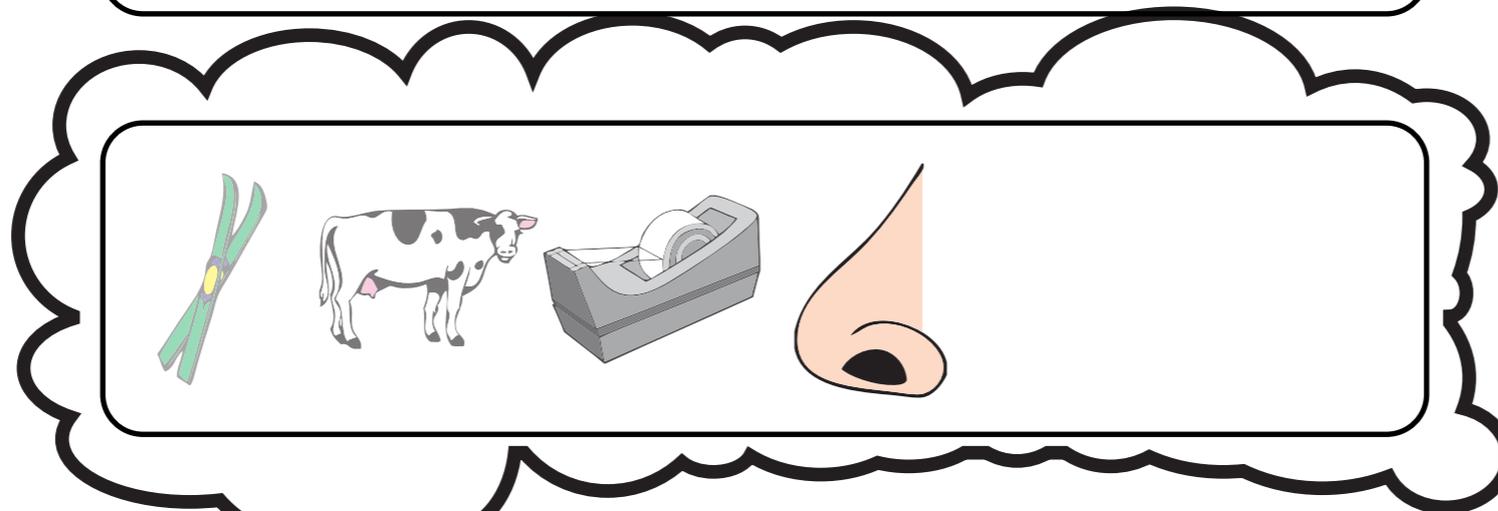


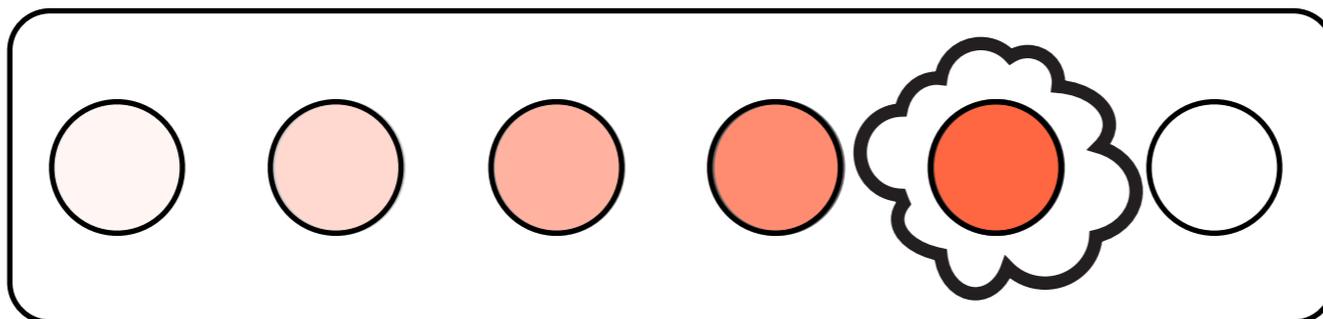
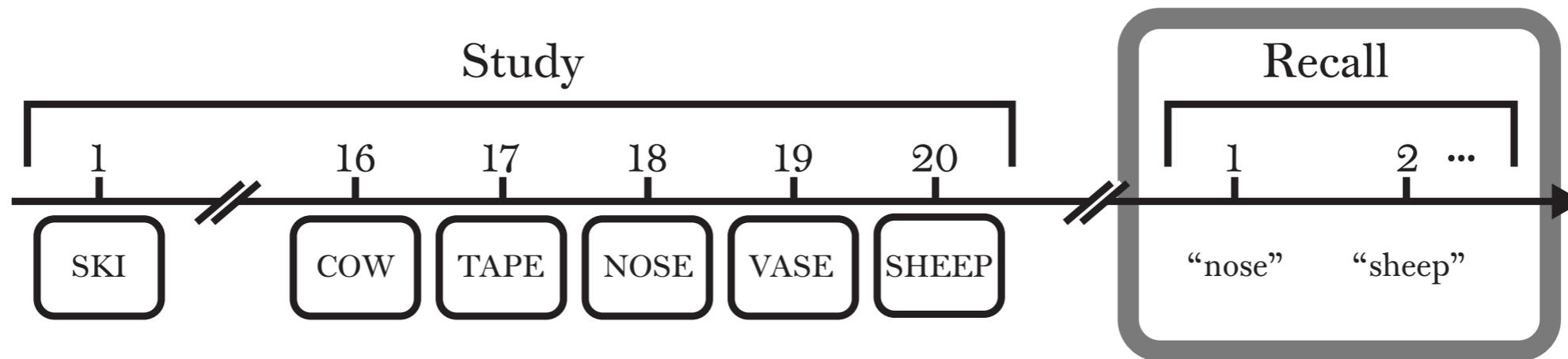


ITEMS

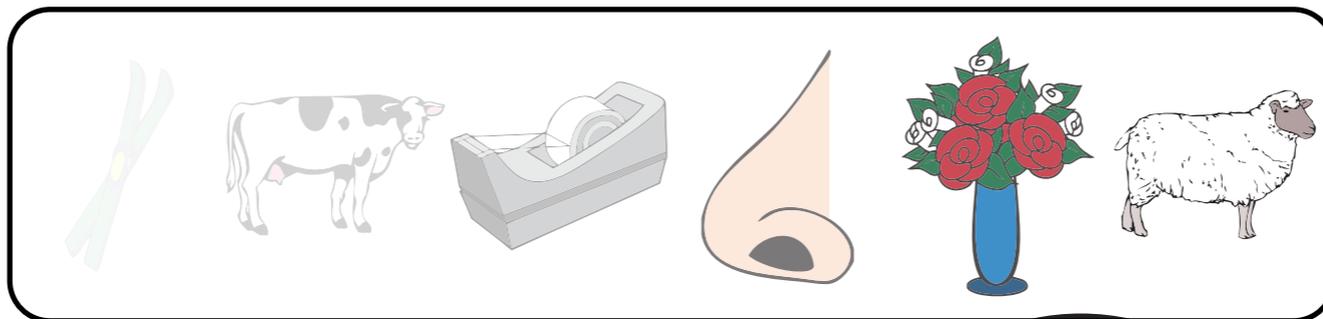
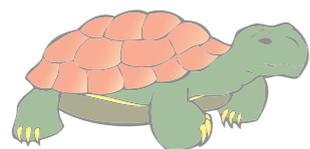


CONTEXT

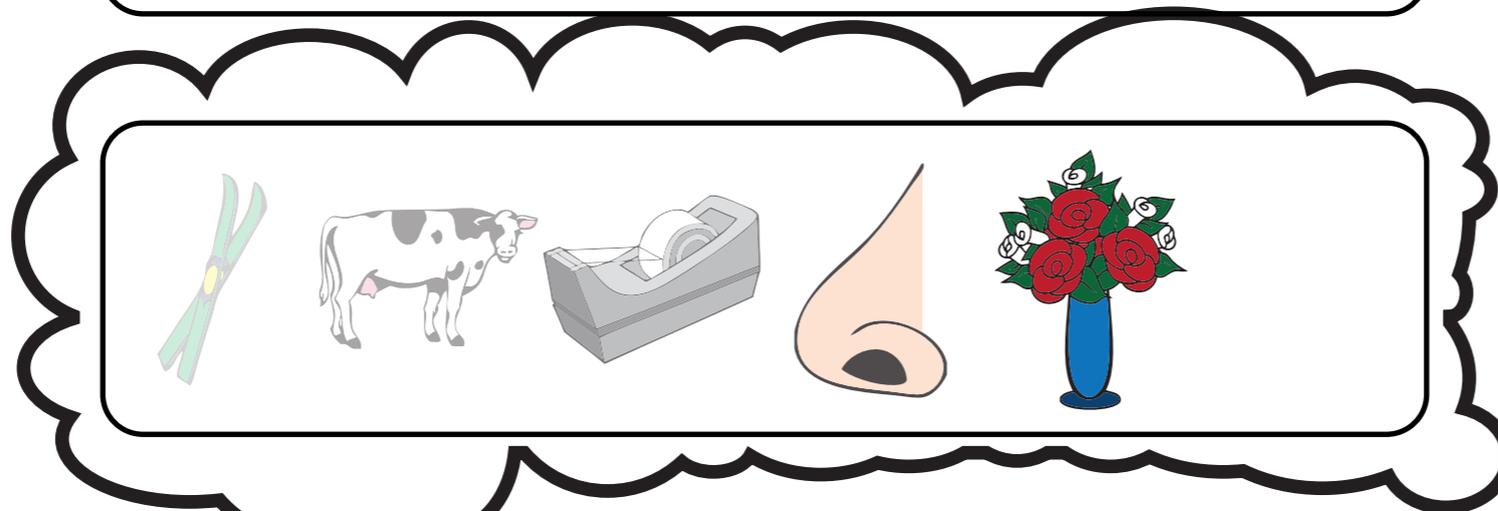


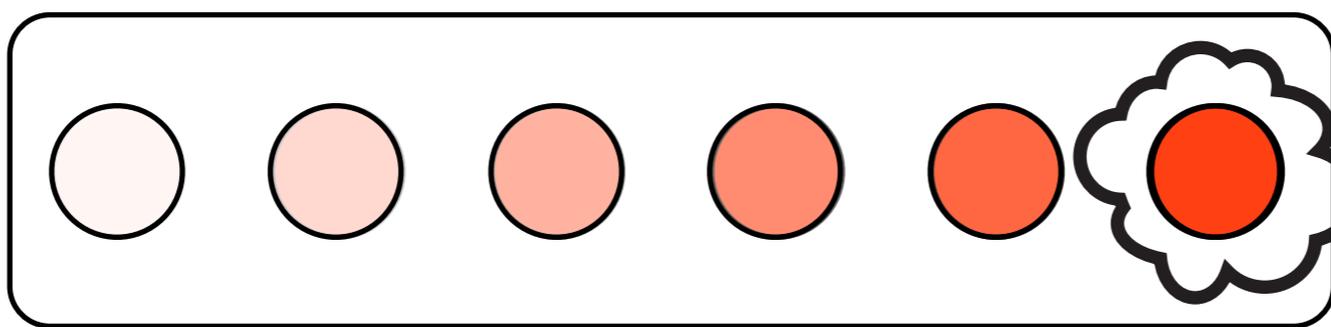
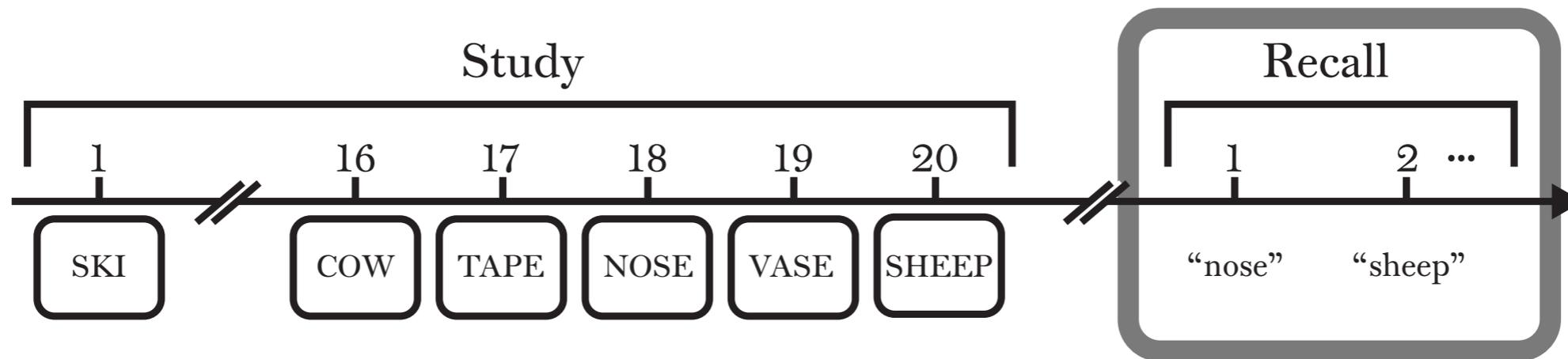


ITEMS

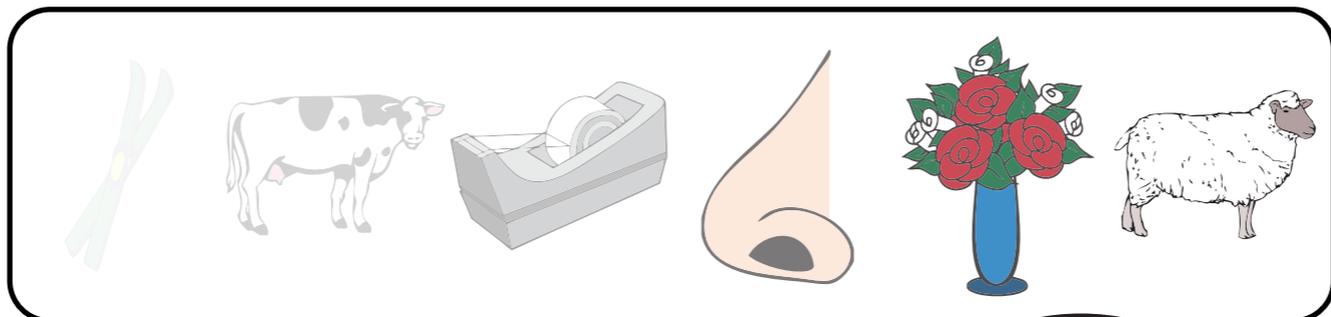
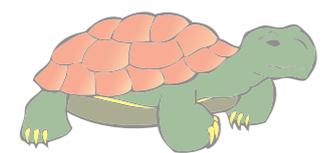


CONTEXT

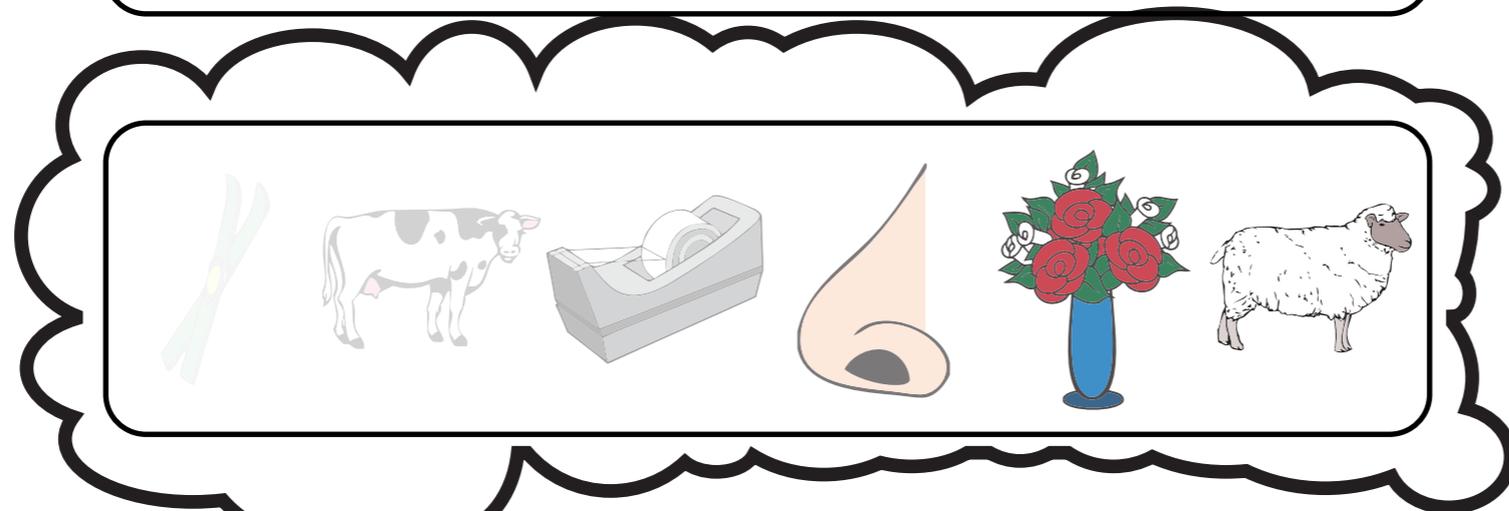


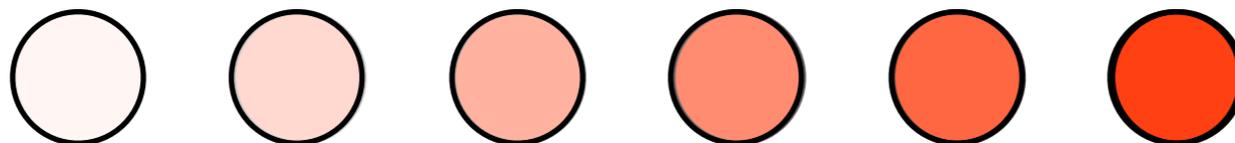
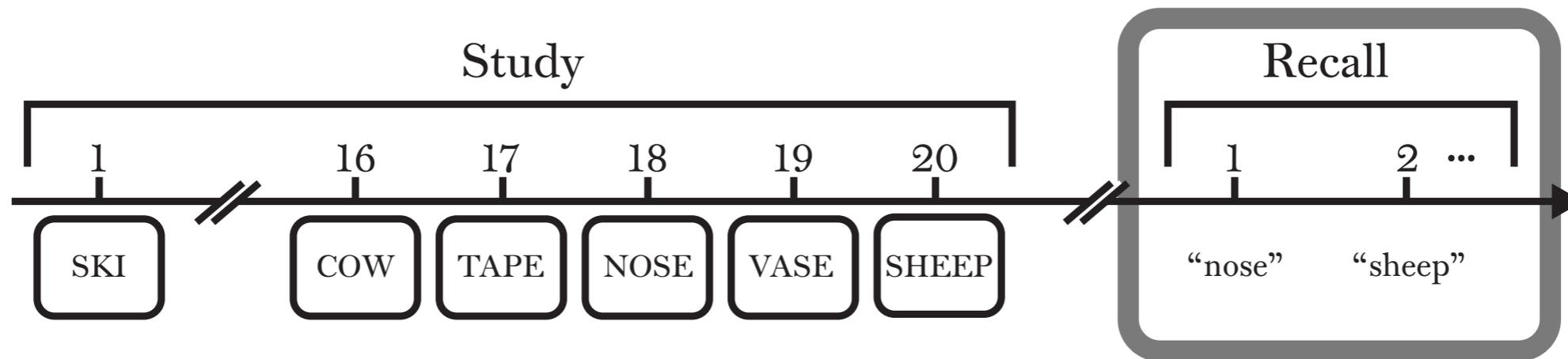


ITEMS

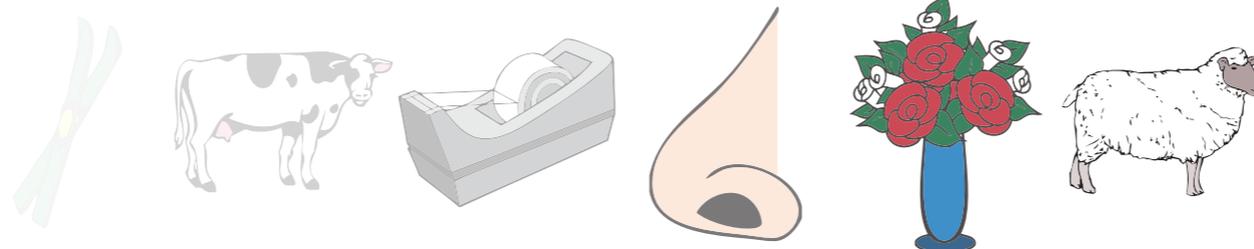
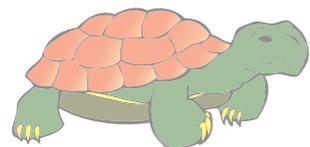


CONTEXT

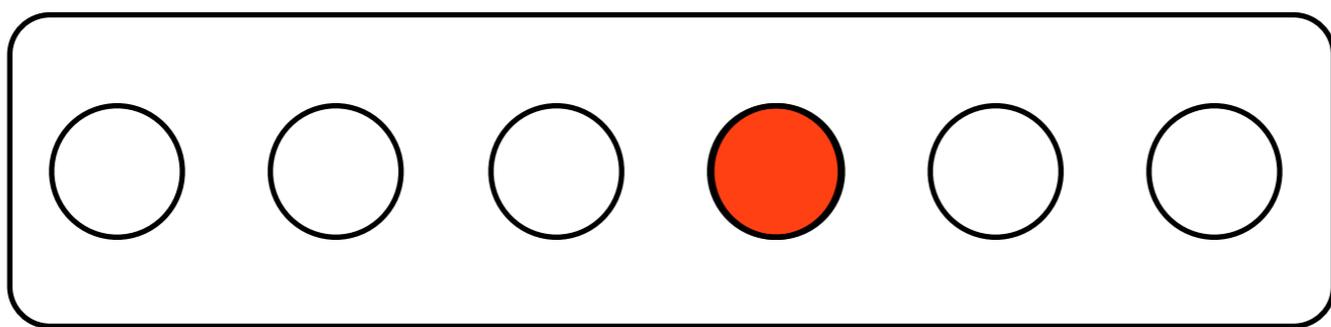
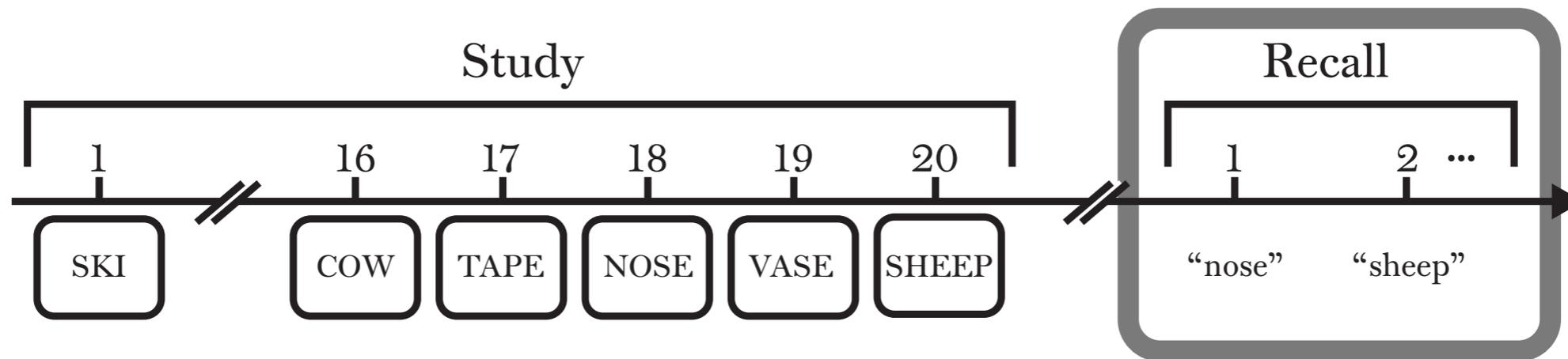




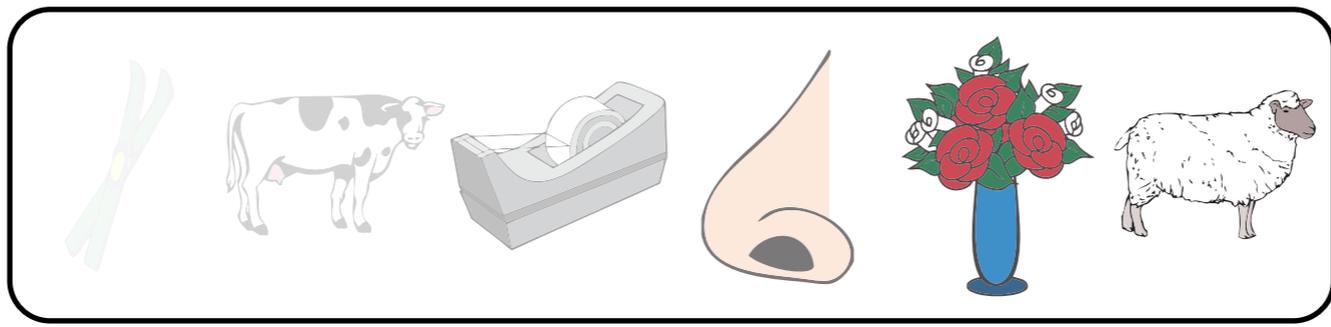
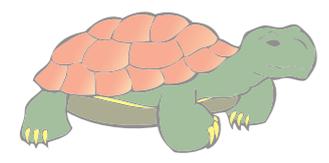
ITEMS



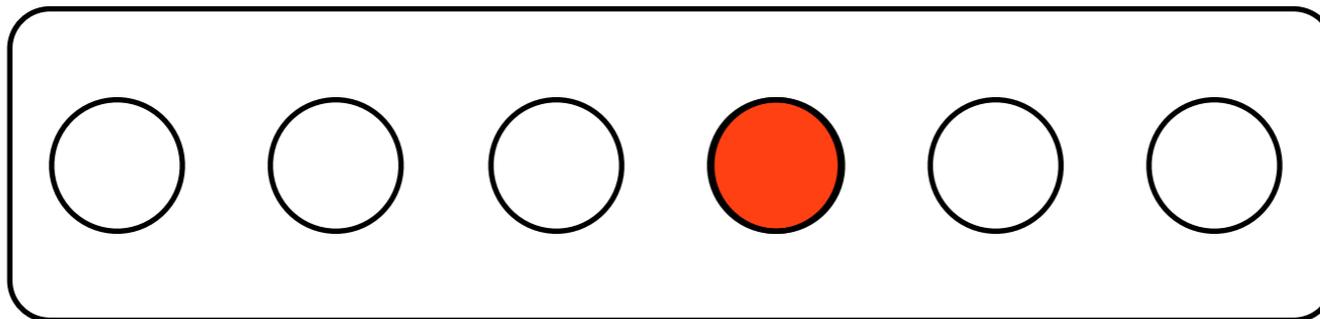
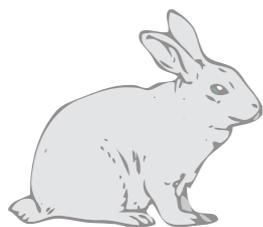
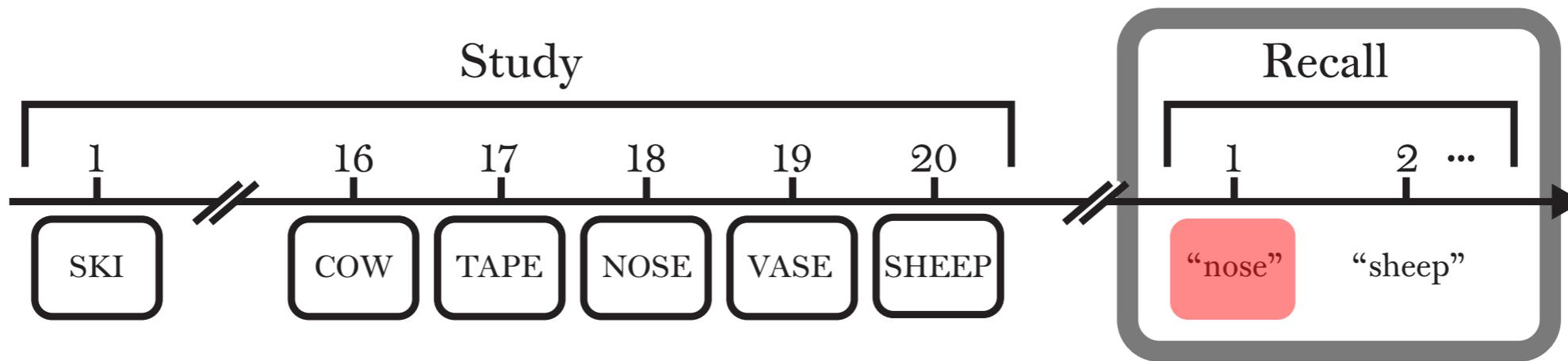
CONTEXT



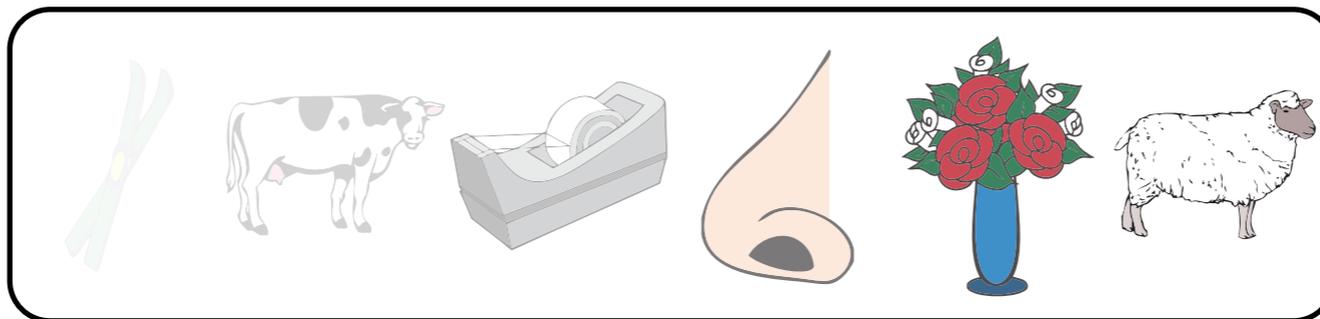
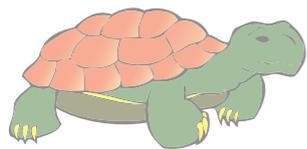
ITEMS



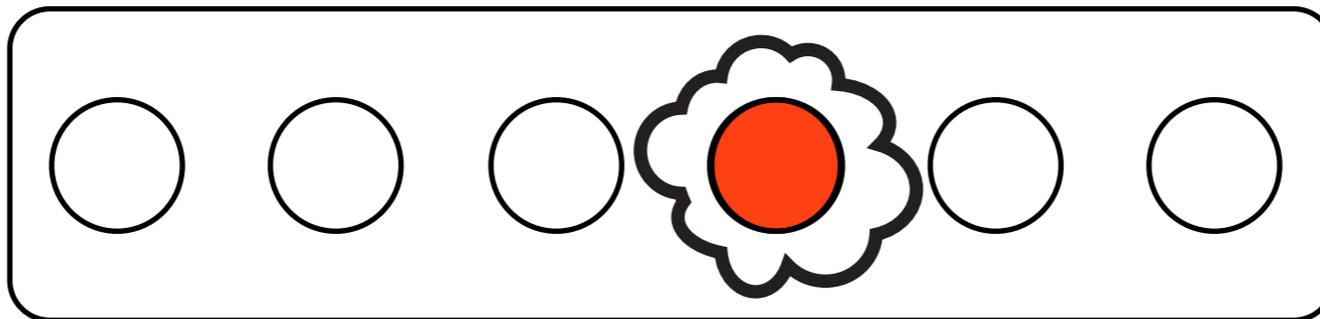
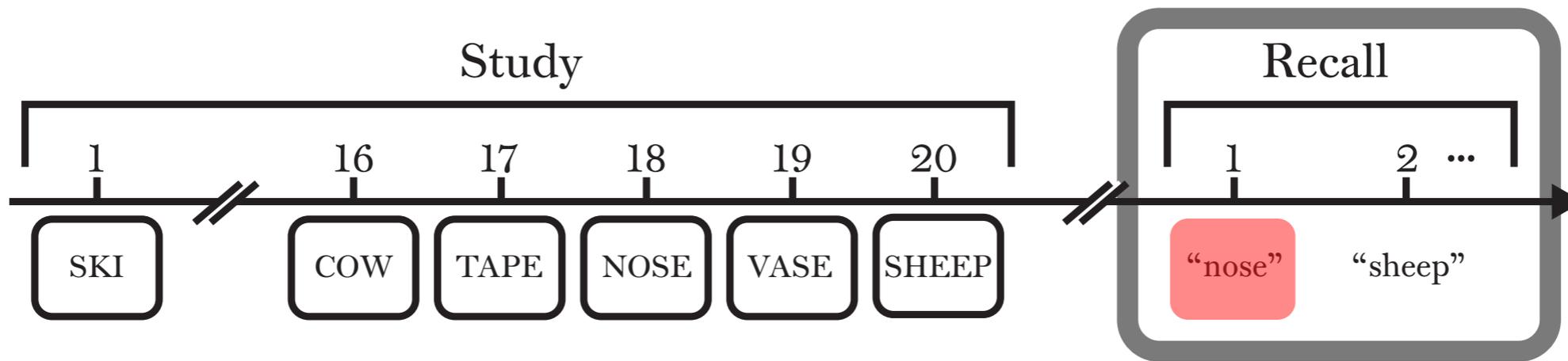
CONTEXT



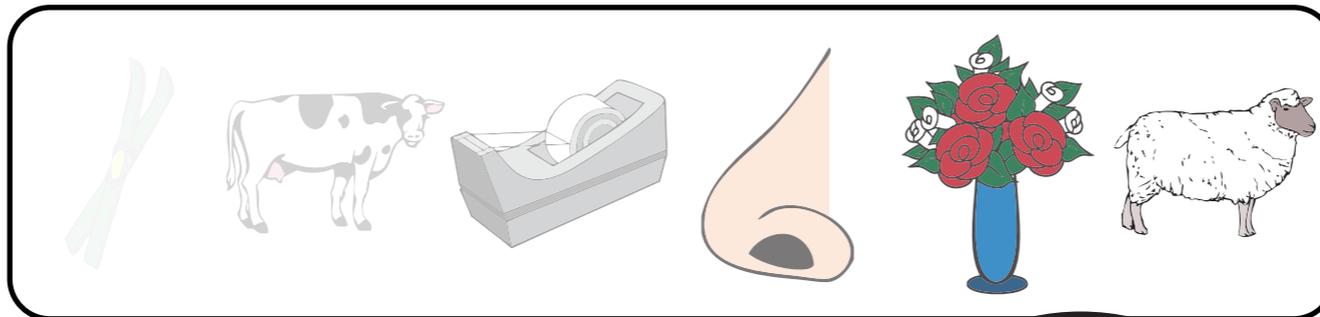
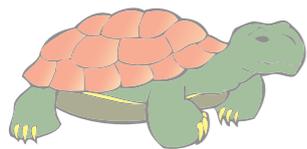
ITEMS



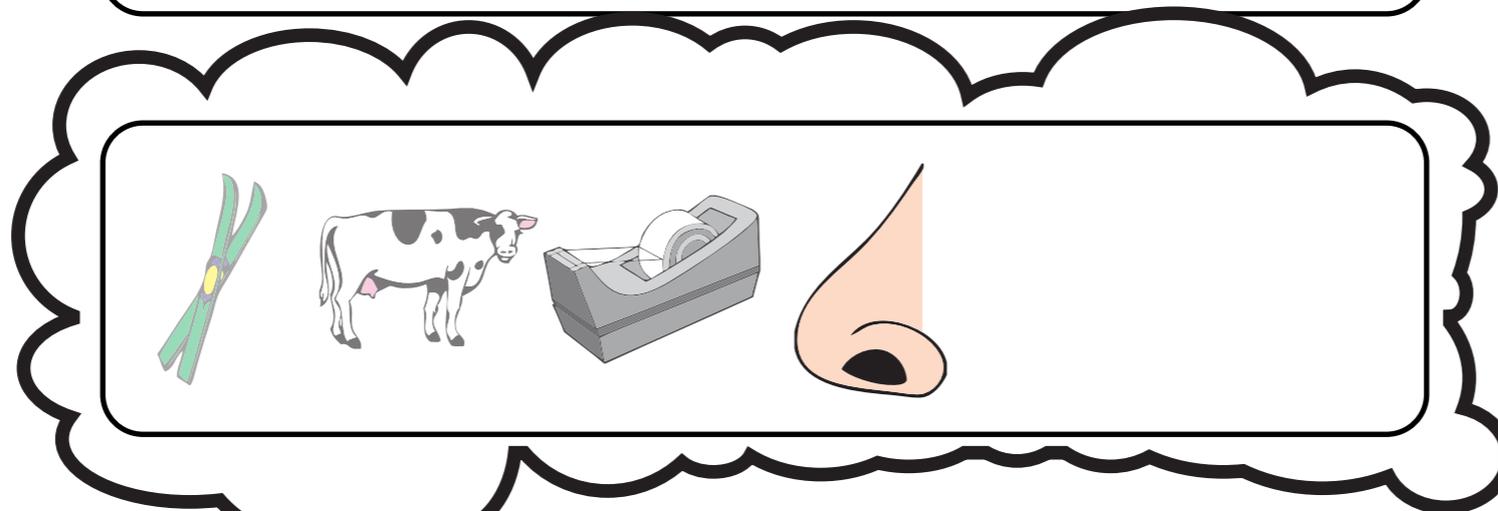
CONTEXT

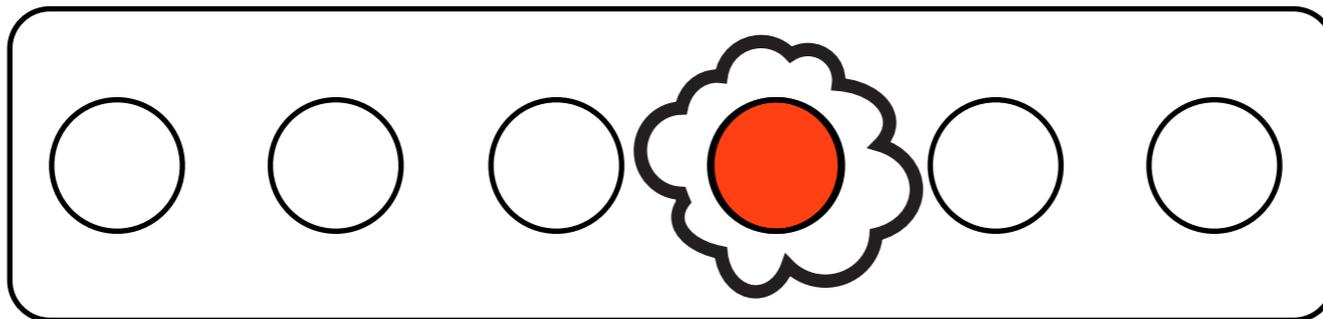
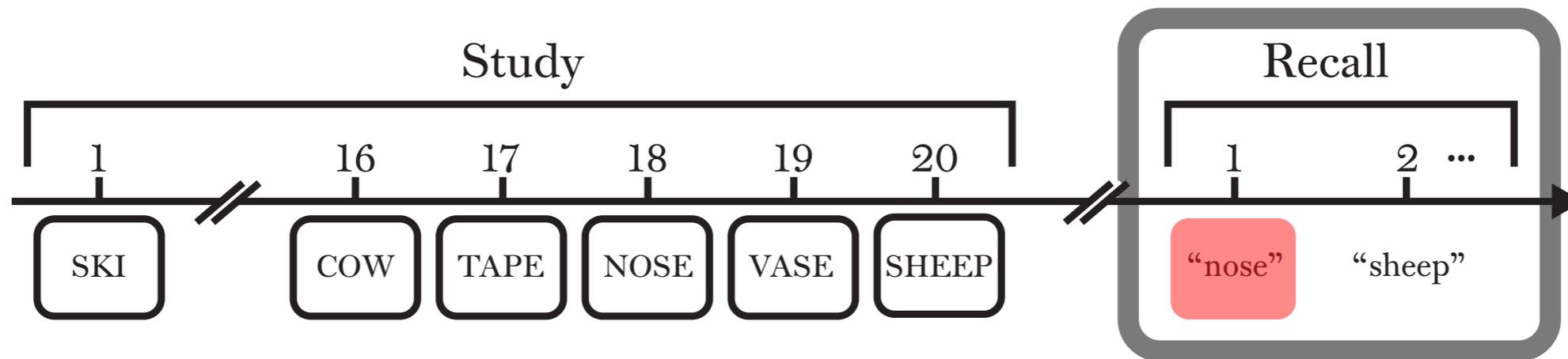


ITEMS

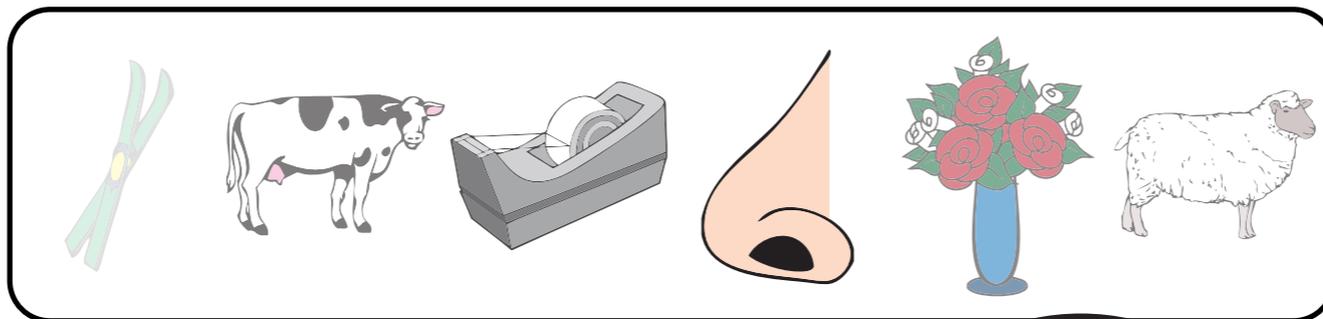
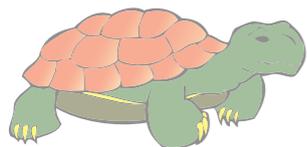


CONTEXT

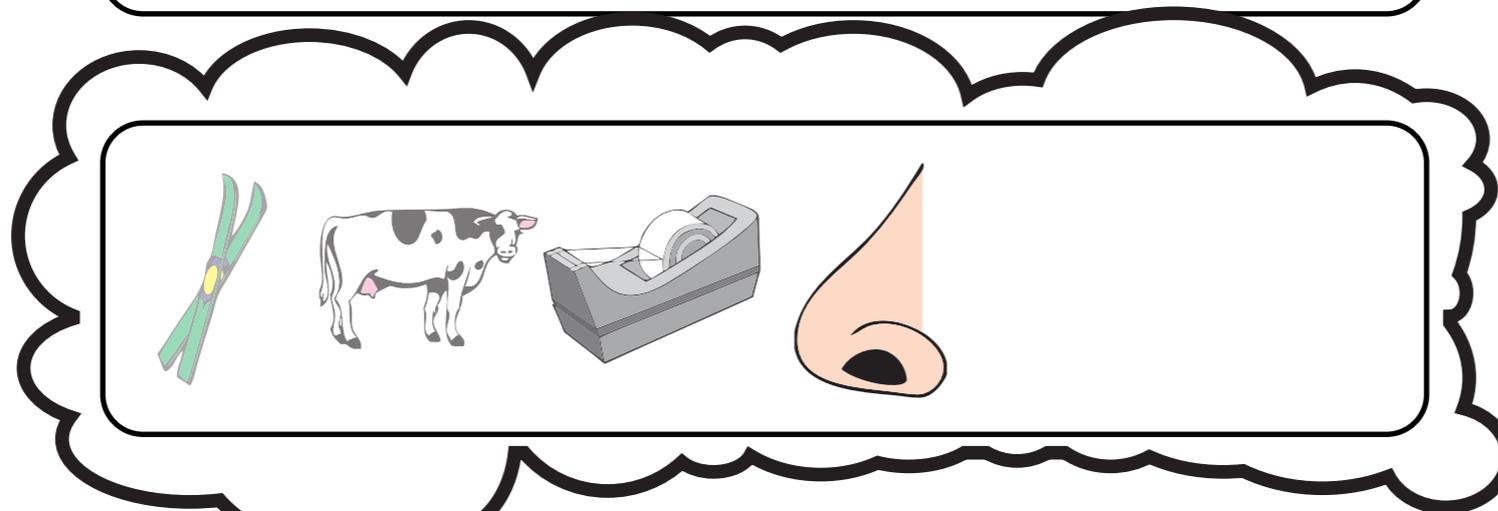




ITEMS



CONTEXT

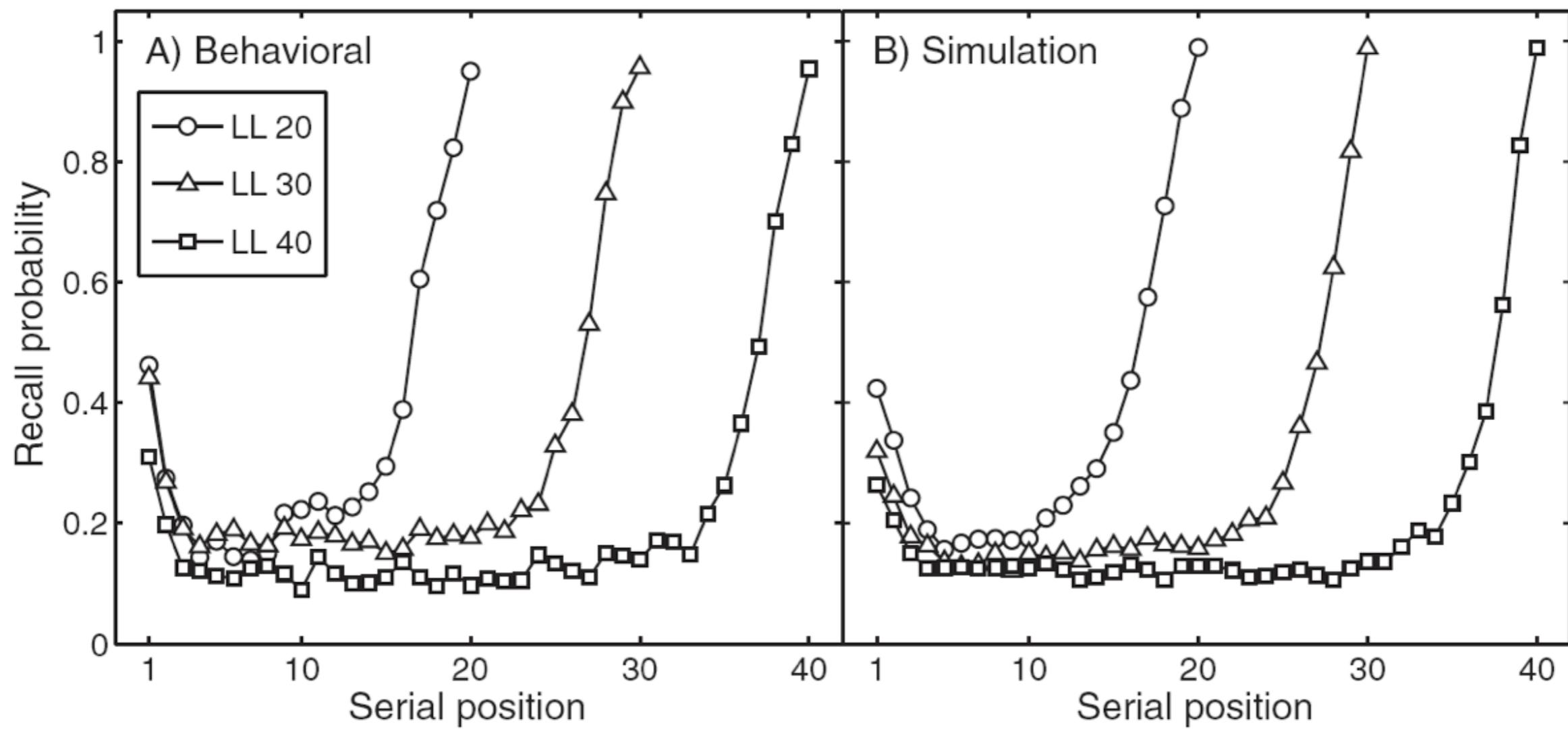


# TCM mechanisms

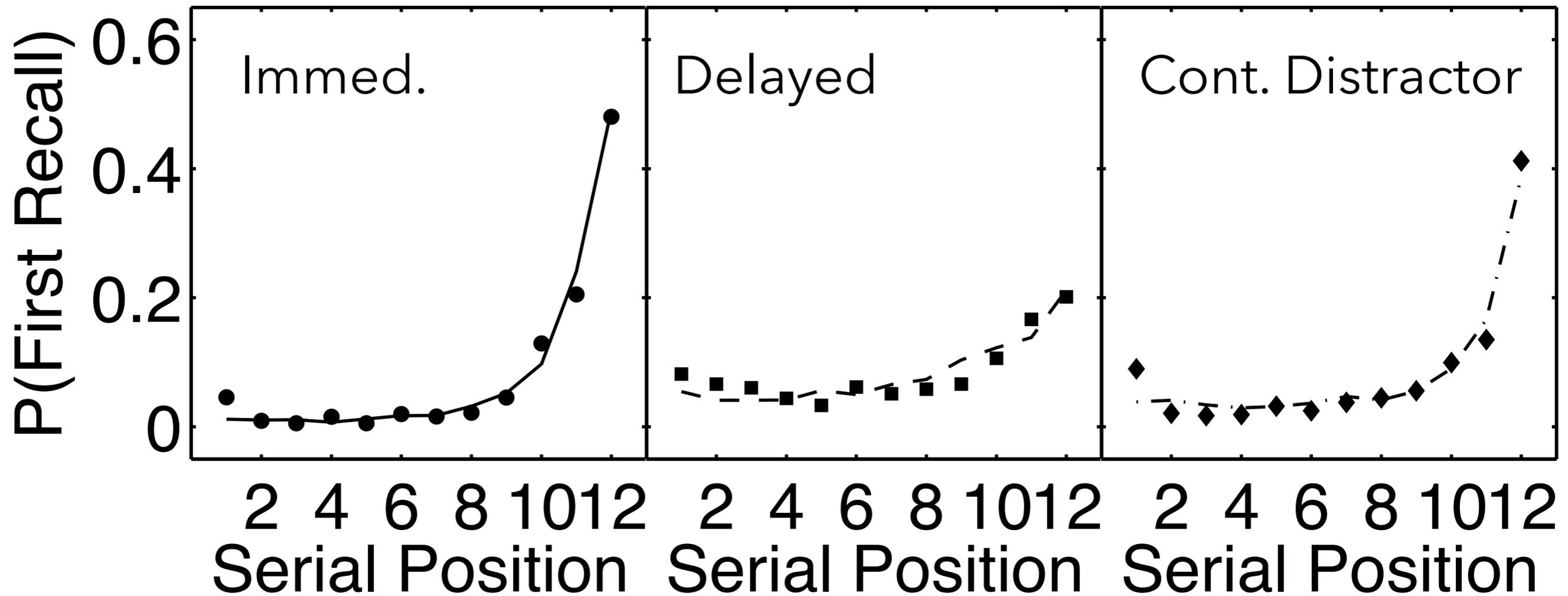
- TCM assumes that each occurrence of an item is (bidirectionally) associated with the context *at the time the item is presented*
- Information association with an item is retrieved and added to context. Thus, context represents a **recency-weighted average of prior thoughts.**

$$C_t = \rho C_{t-1} + (1 - \rho) f_t$$

Evaluating TCM...

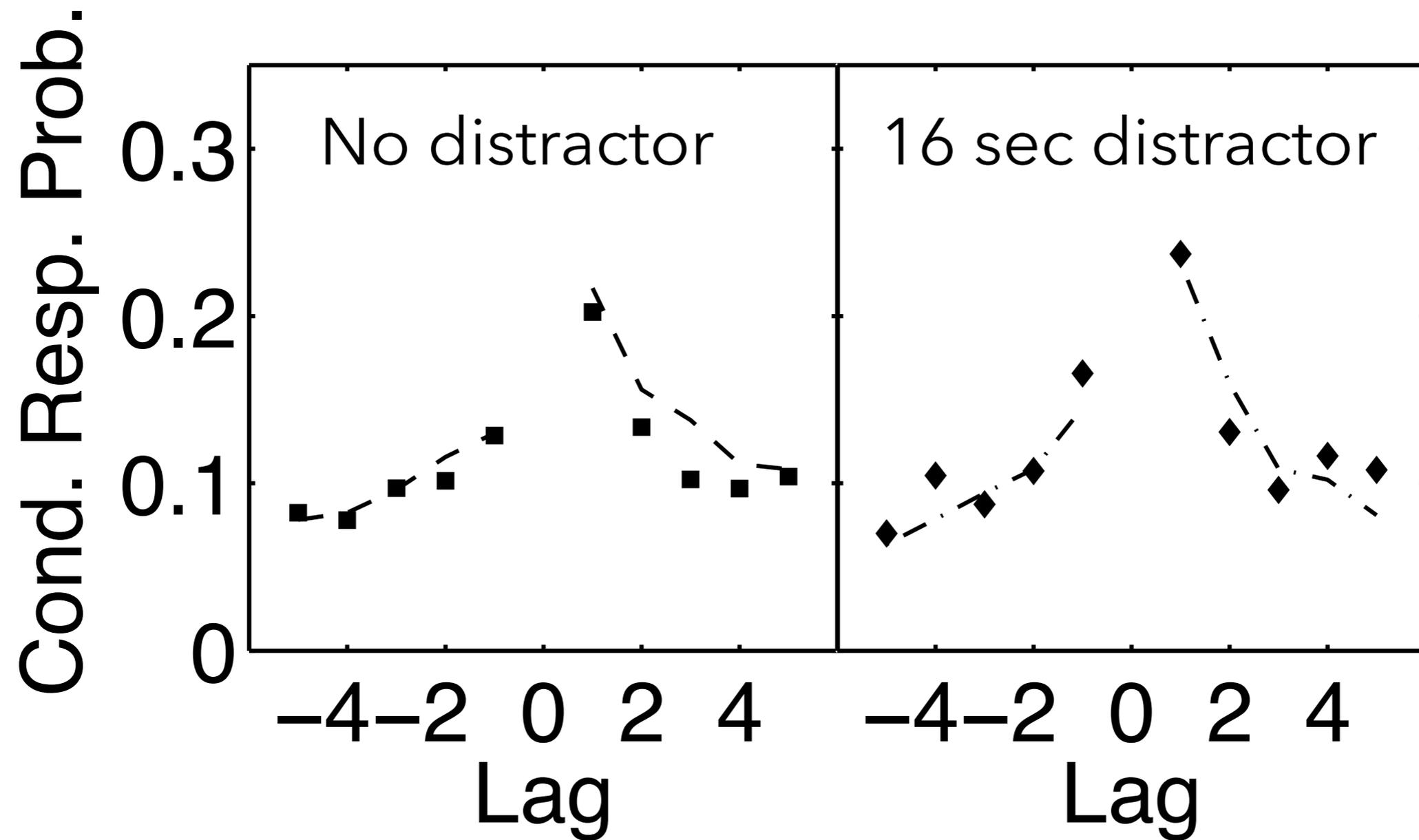


# Long-term Recency



Sederberg et al. (2007)

# Long-range associations

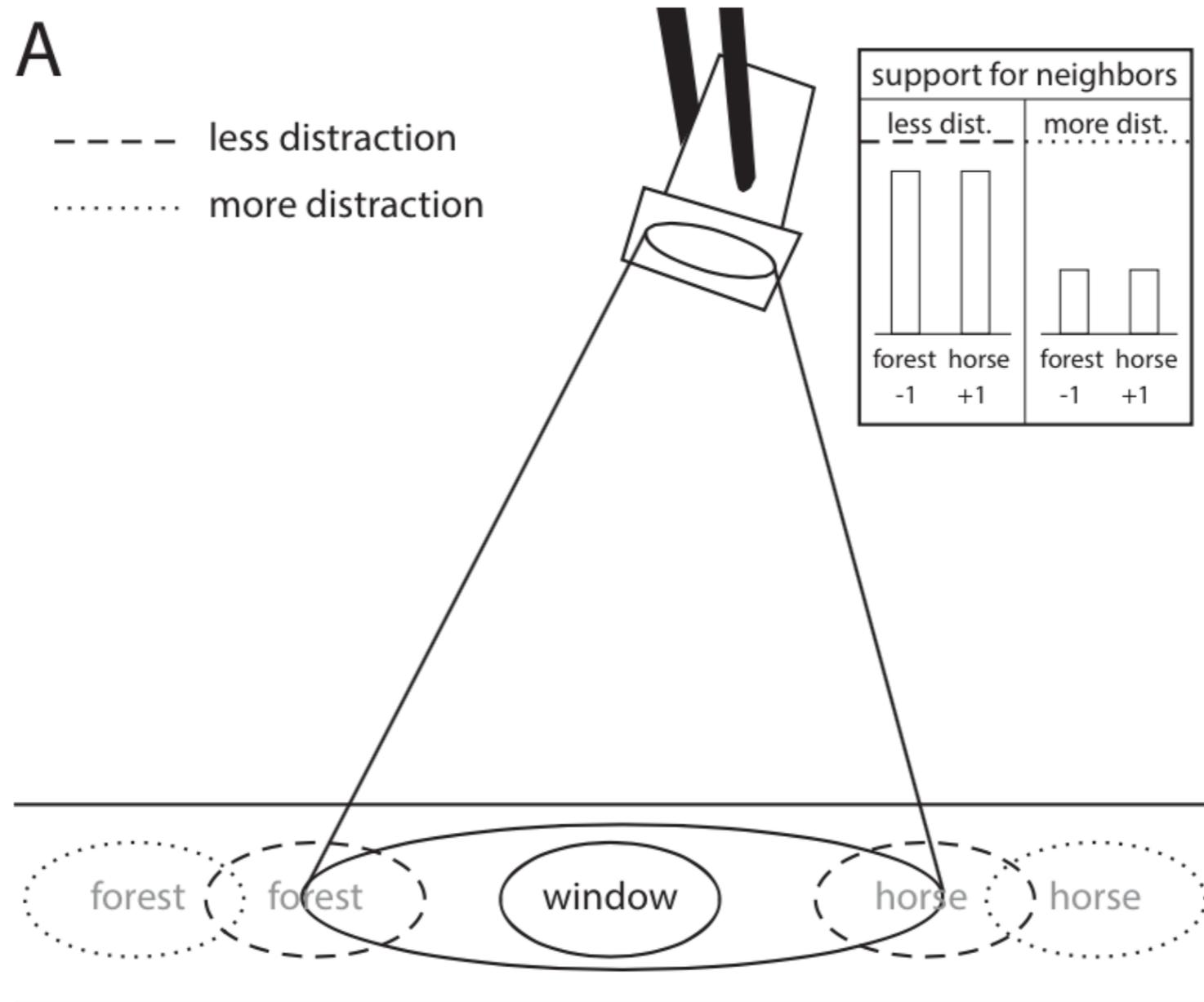


Sederberg et al. (2007)

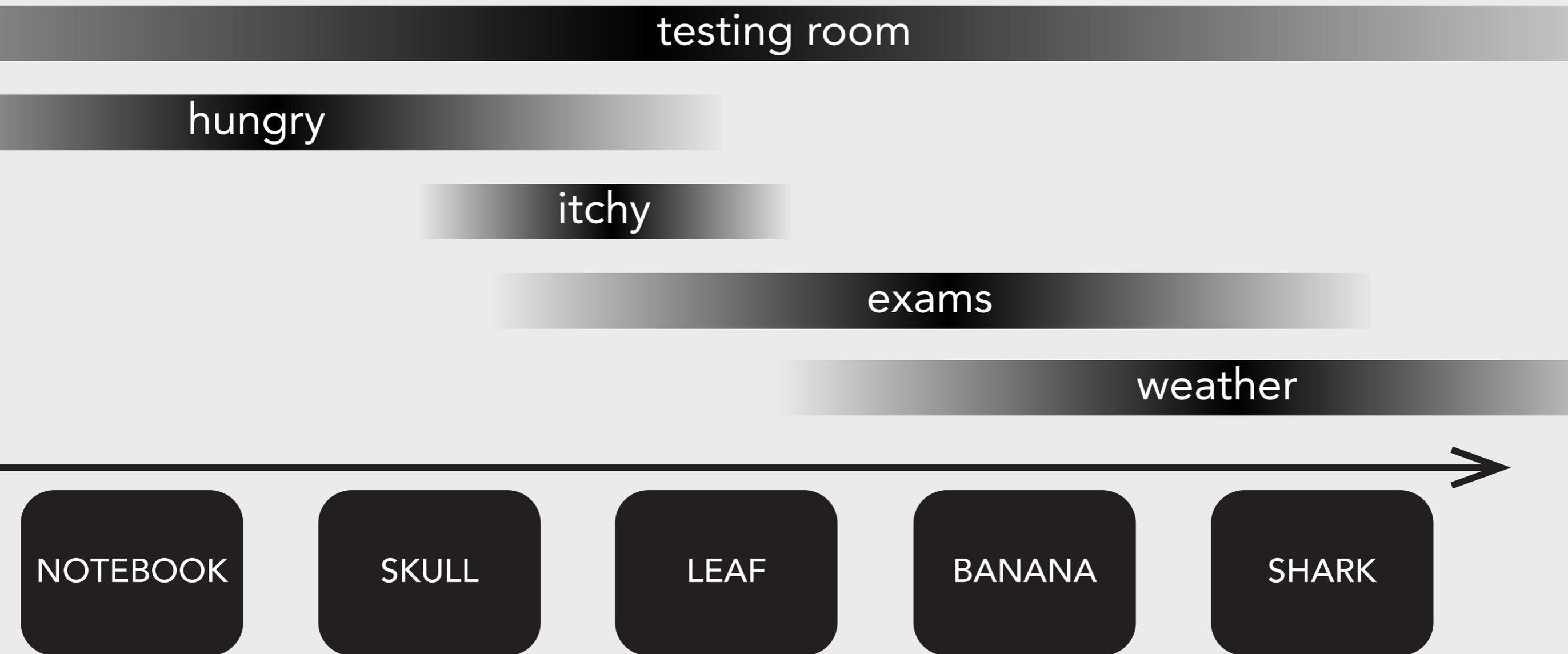
# TCM and long-range associations

A

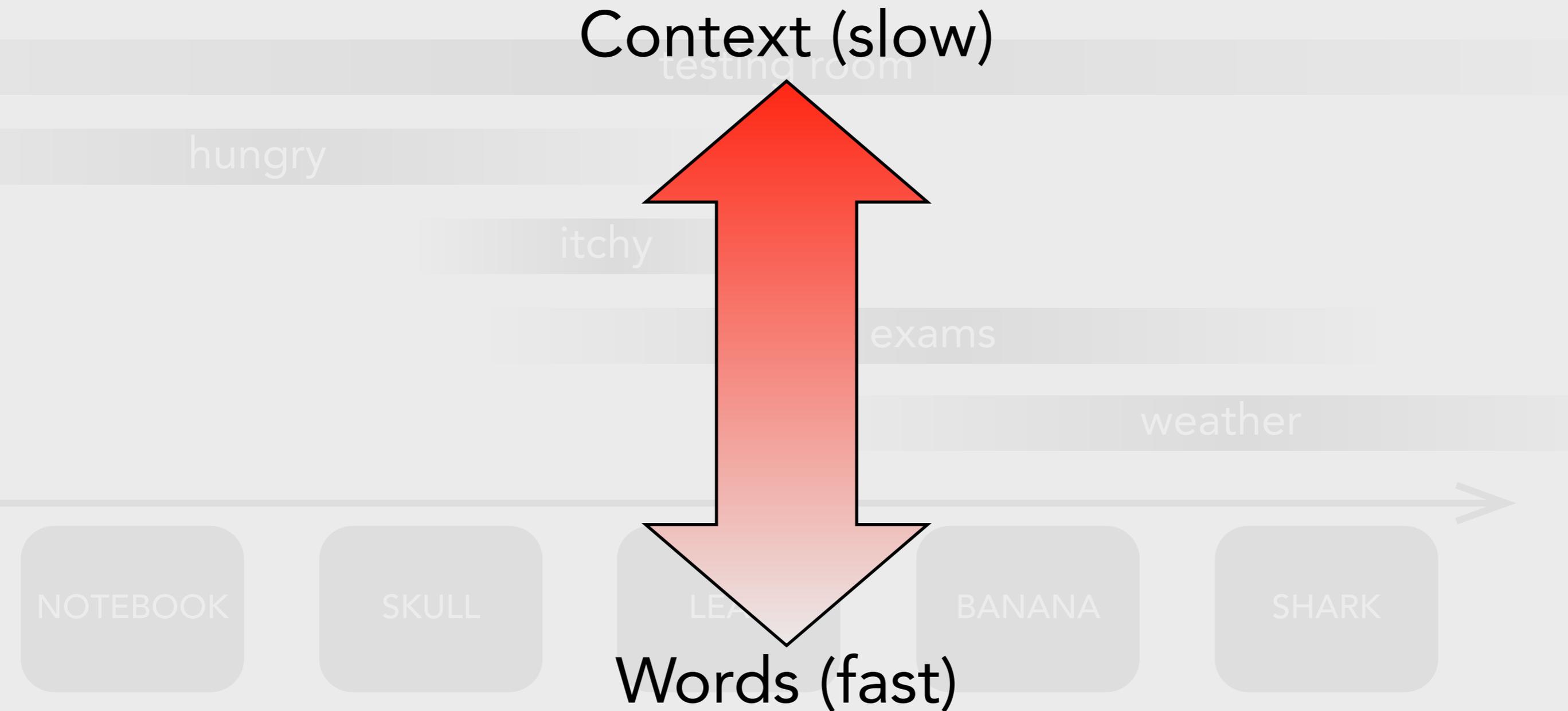
----- less distraction  
..... more distraction



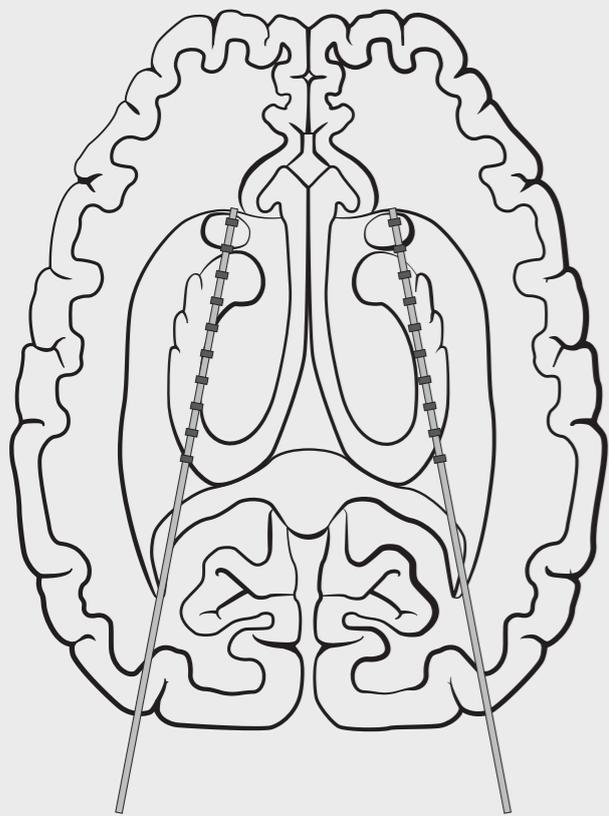
# Teasing out the neural representation of context



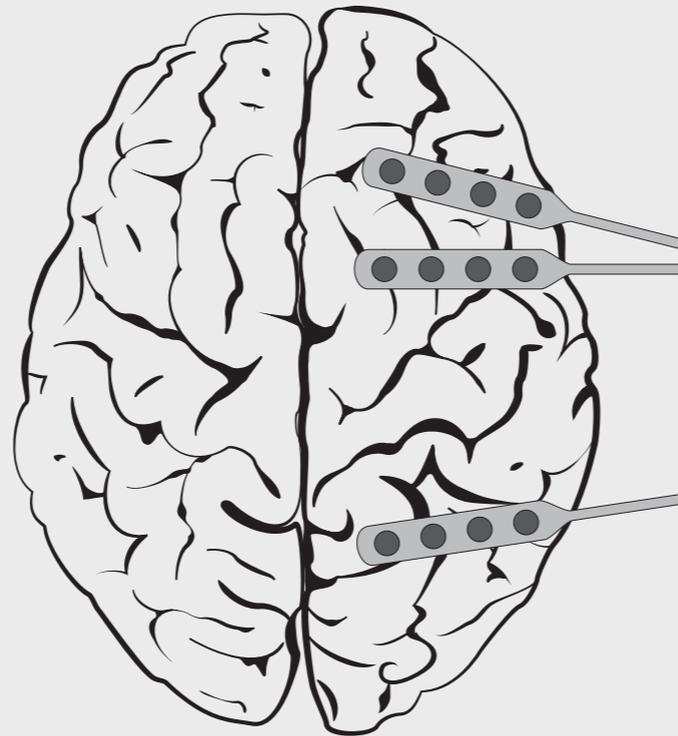
# Teasing out the neural representation of context



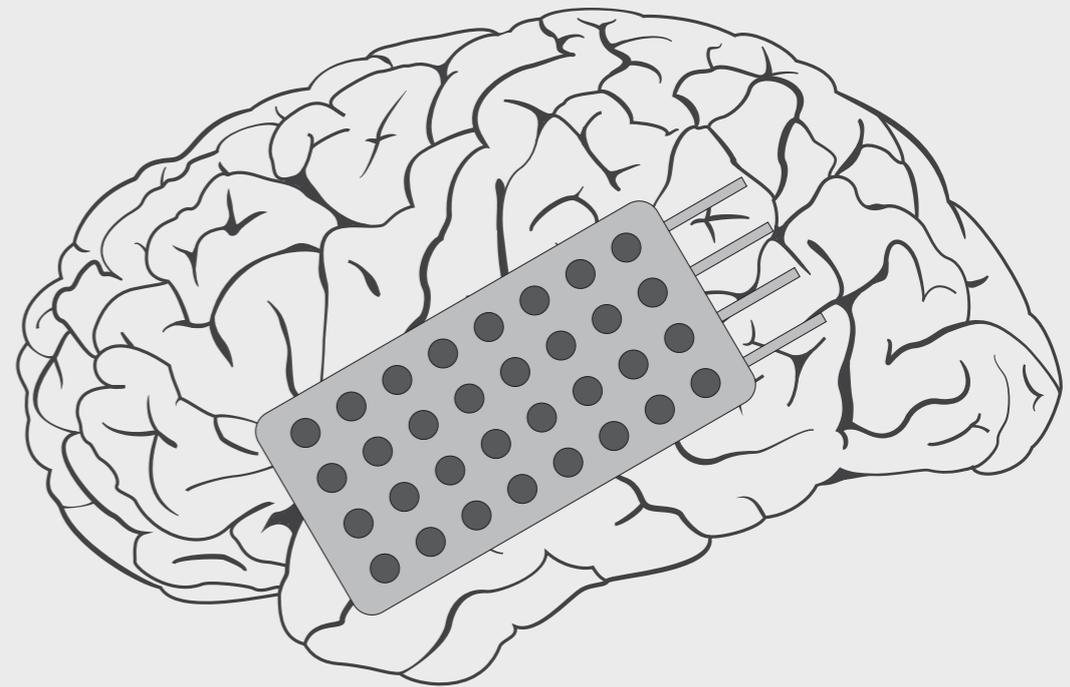
# Human intracranial recordings



Depth electrodes

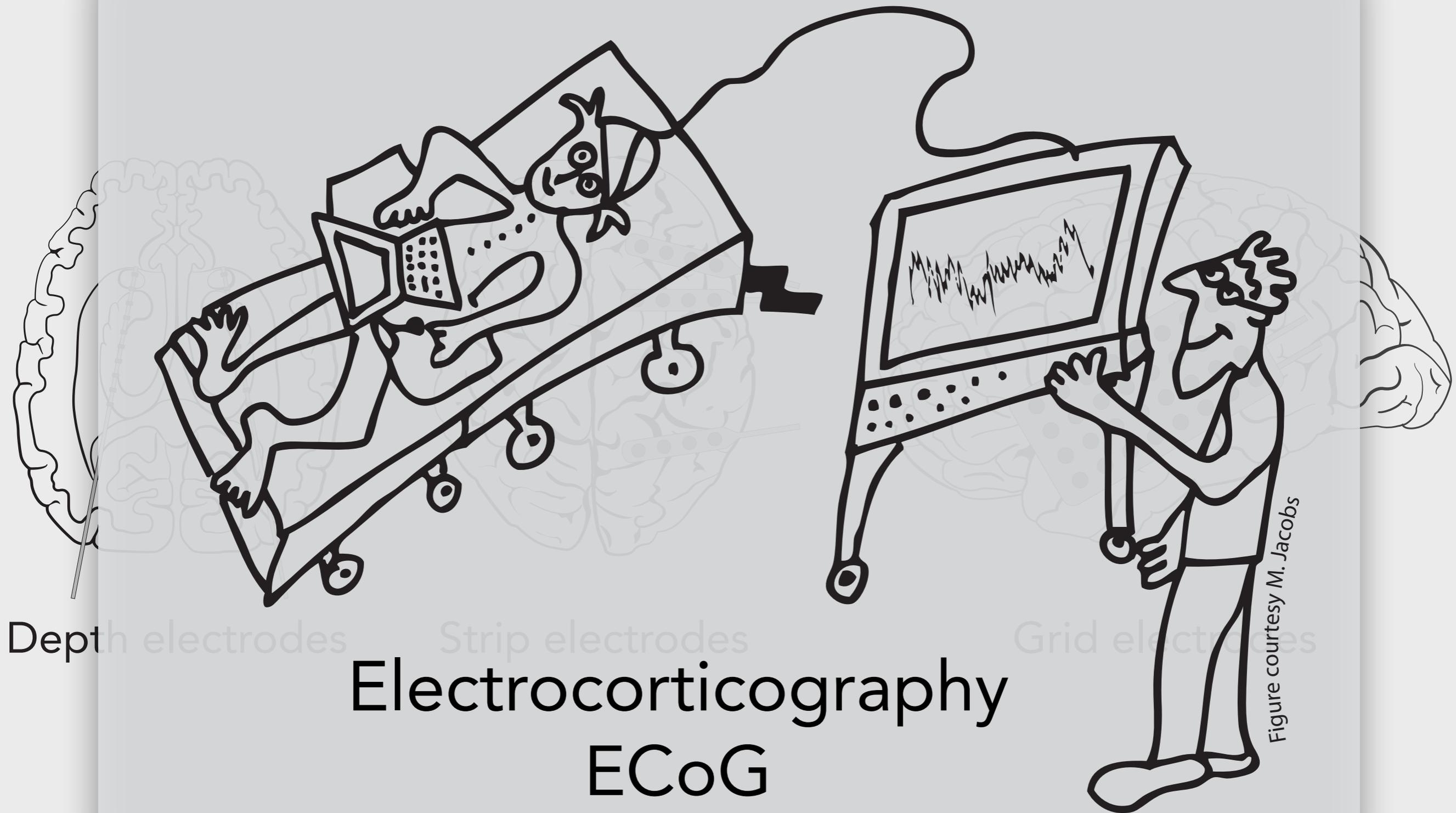


Strip electrodes



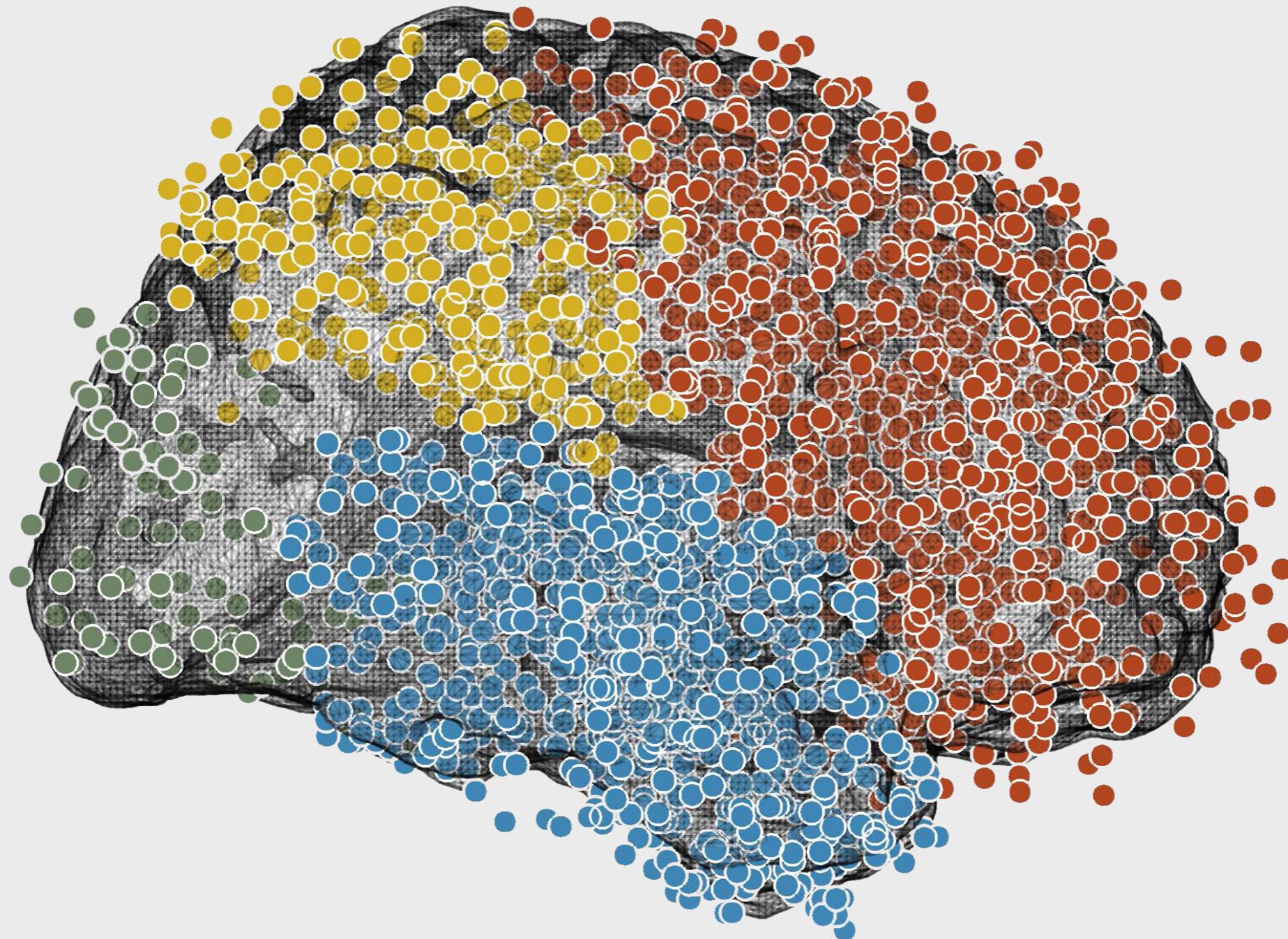
Grid electrodes

# Human intracranial recordings

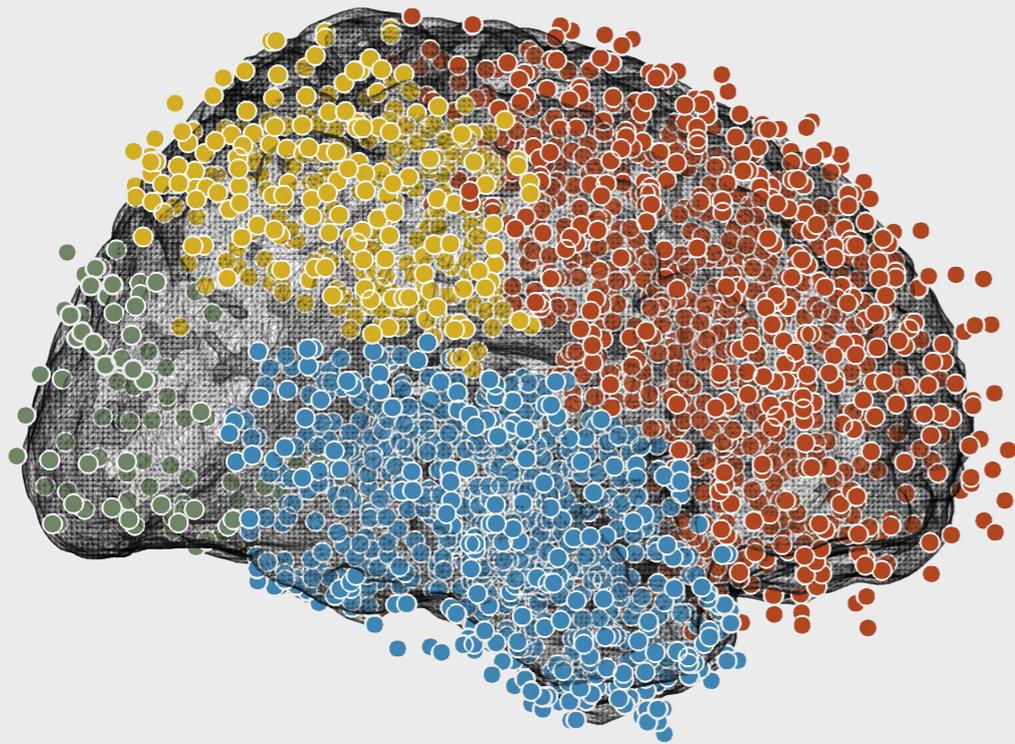


Electrocorticography  
ECoG

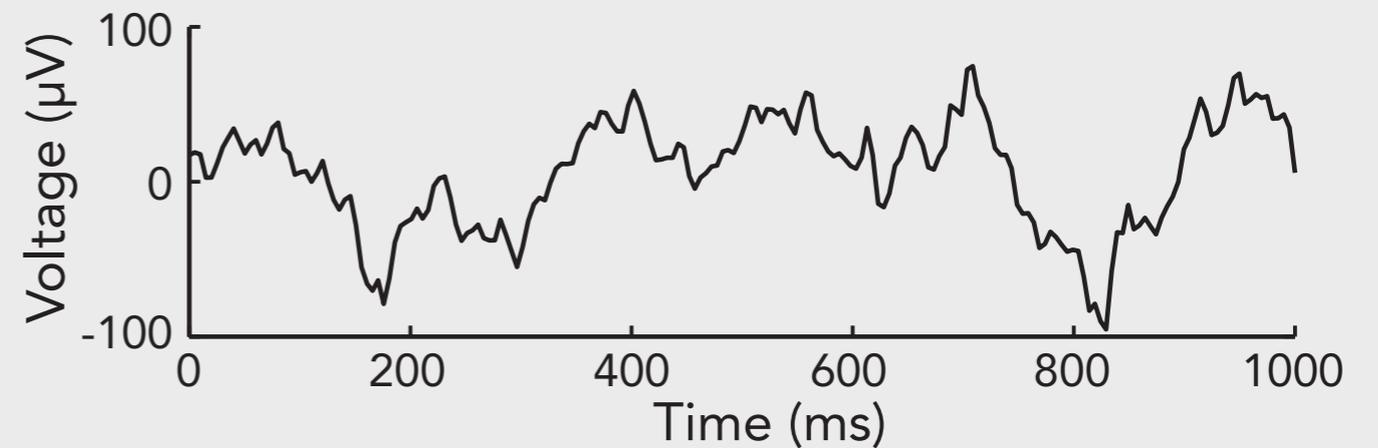
# Human intracranial recordings



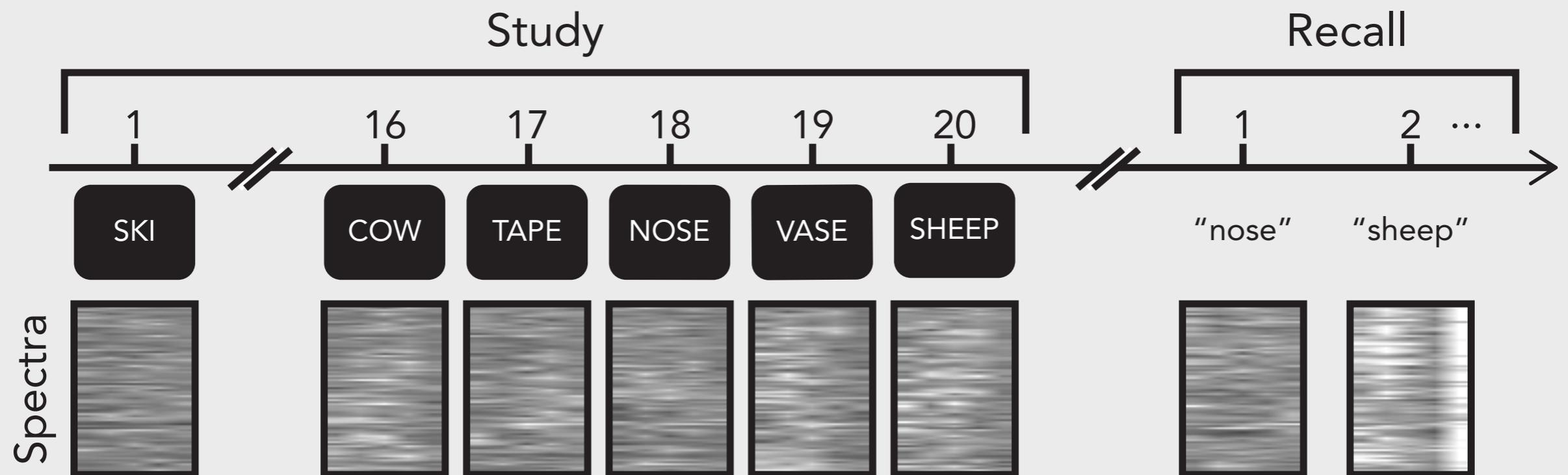
# Human intracranial recordings



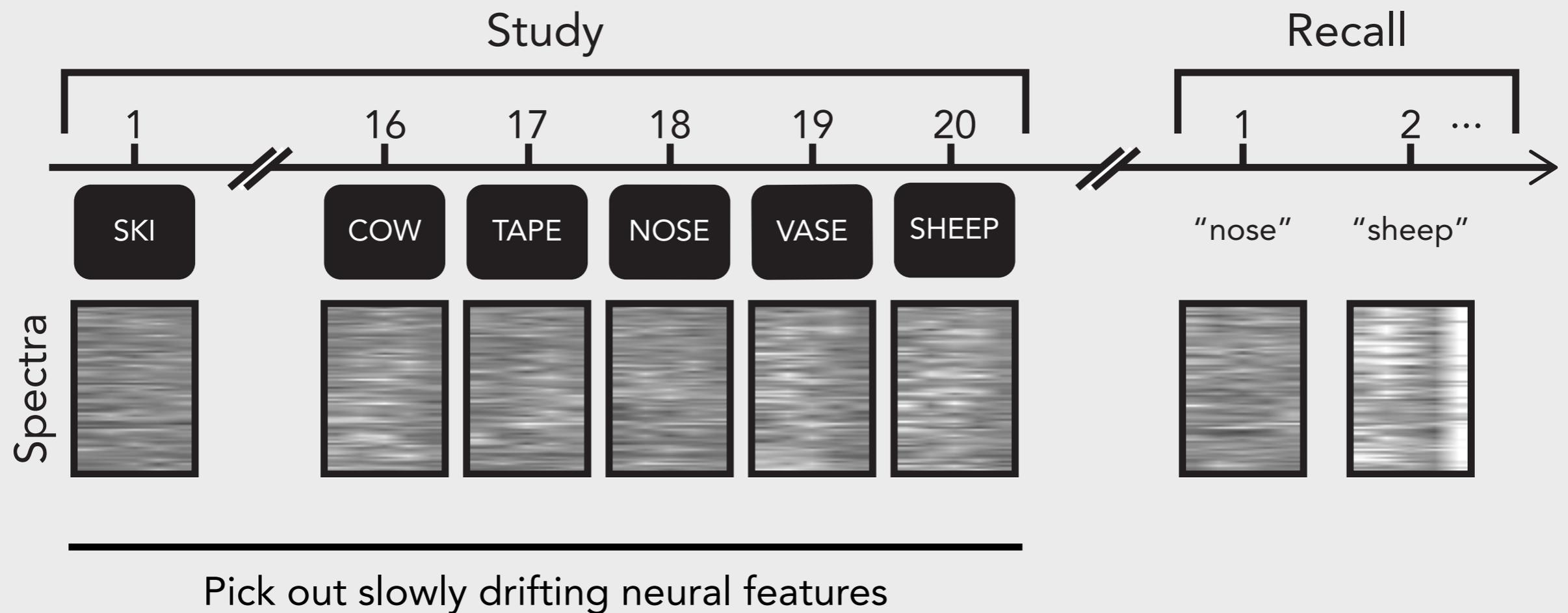
## Local field potential



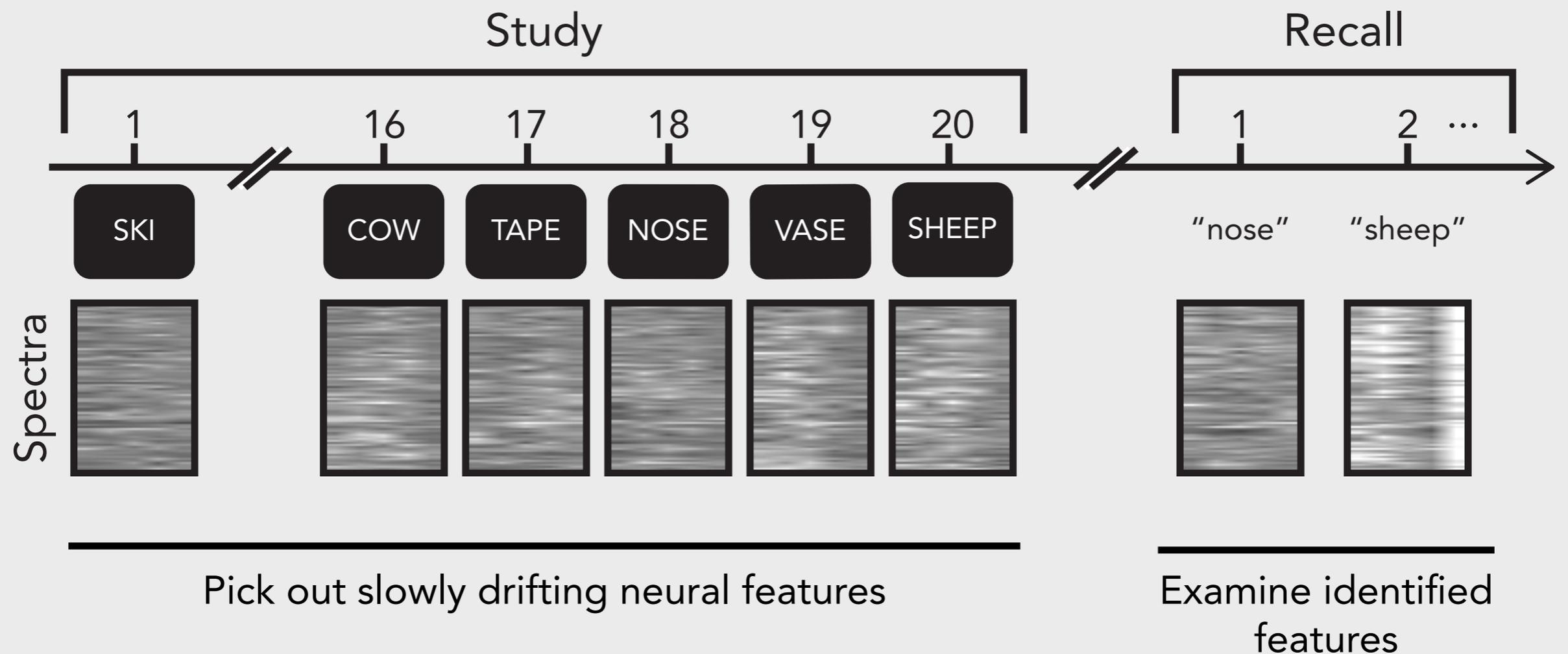
# Human intracranial recordings



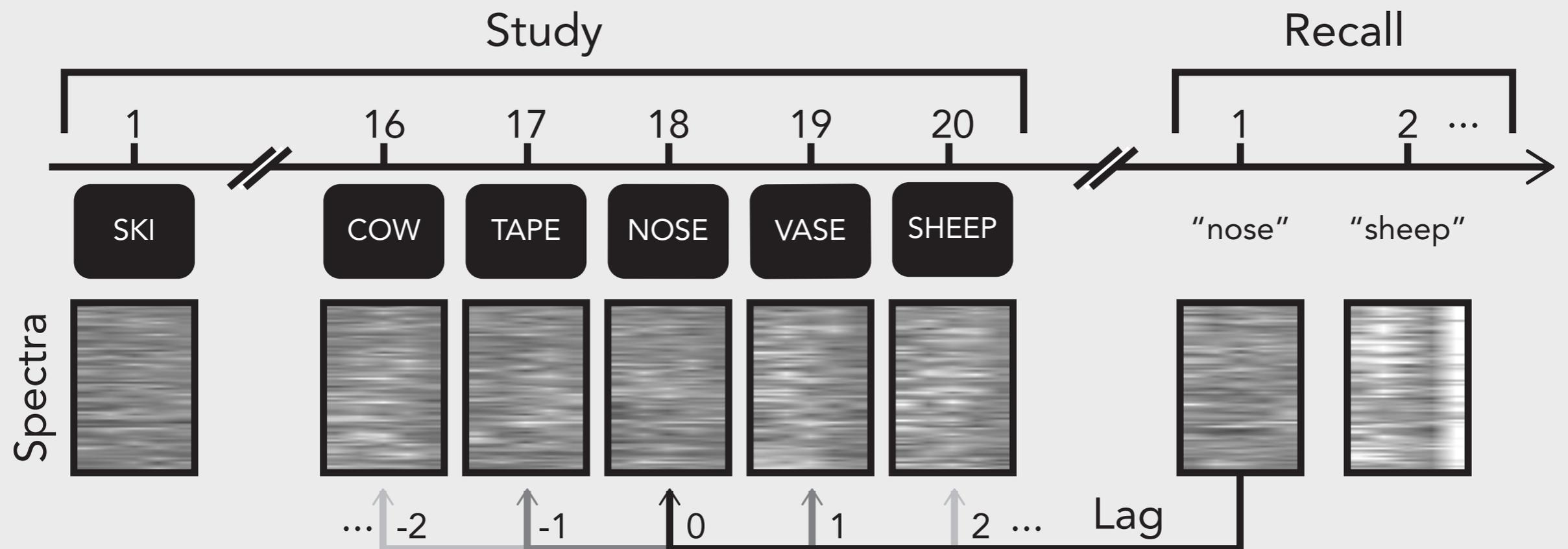
# Isolating neural representations of contextual cues



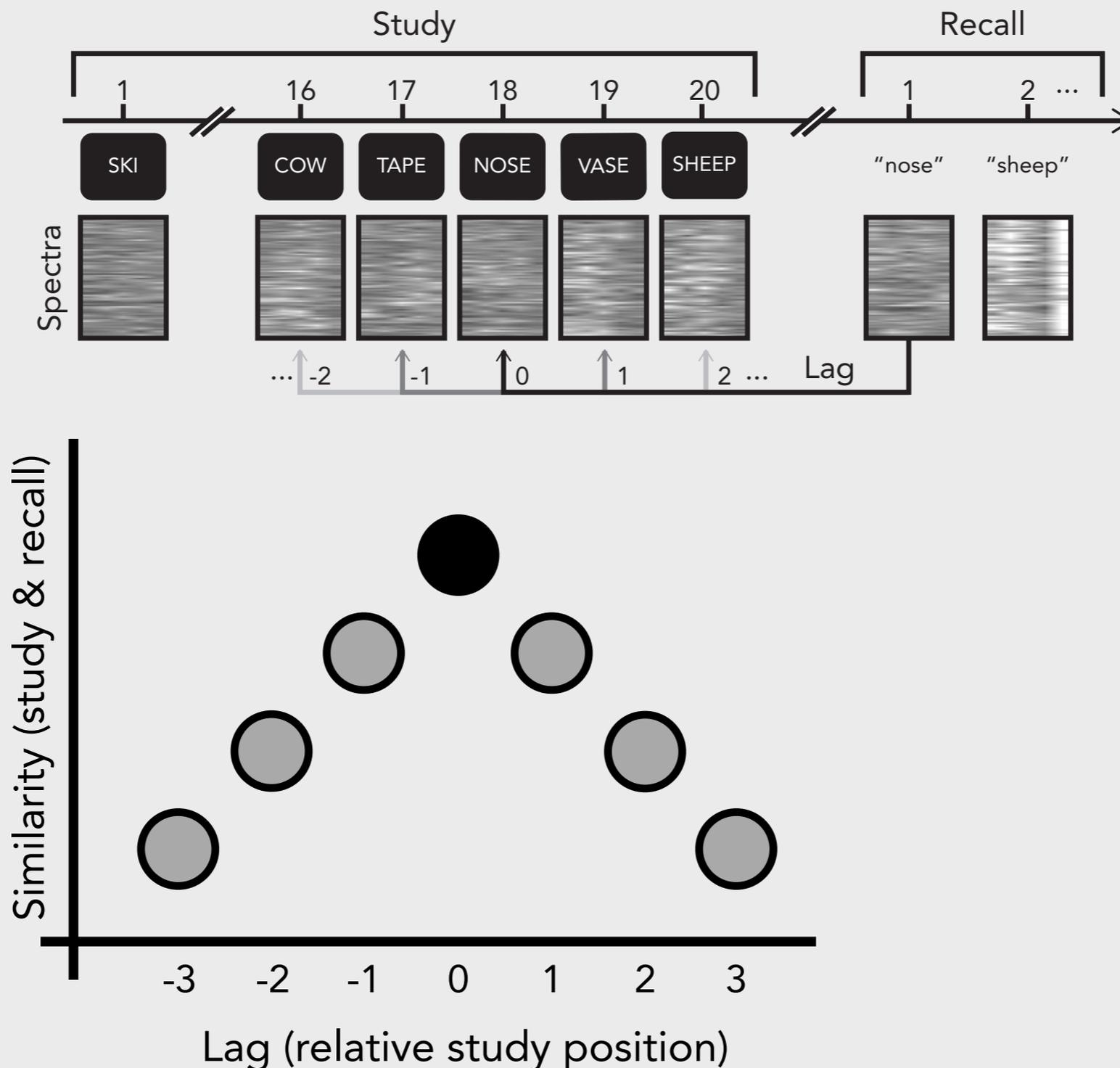
# Isolating neural representations of contextual cues



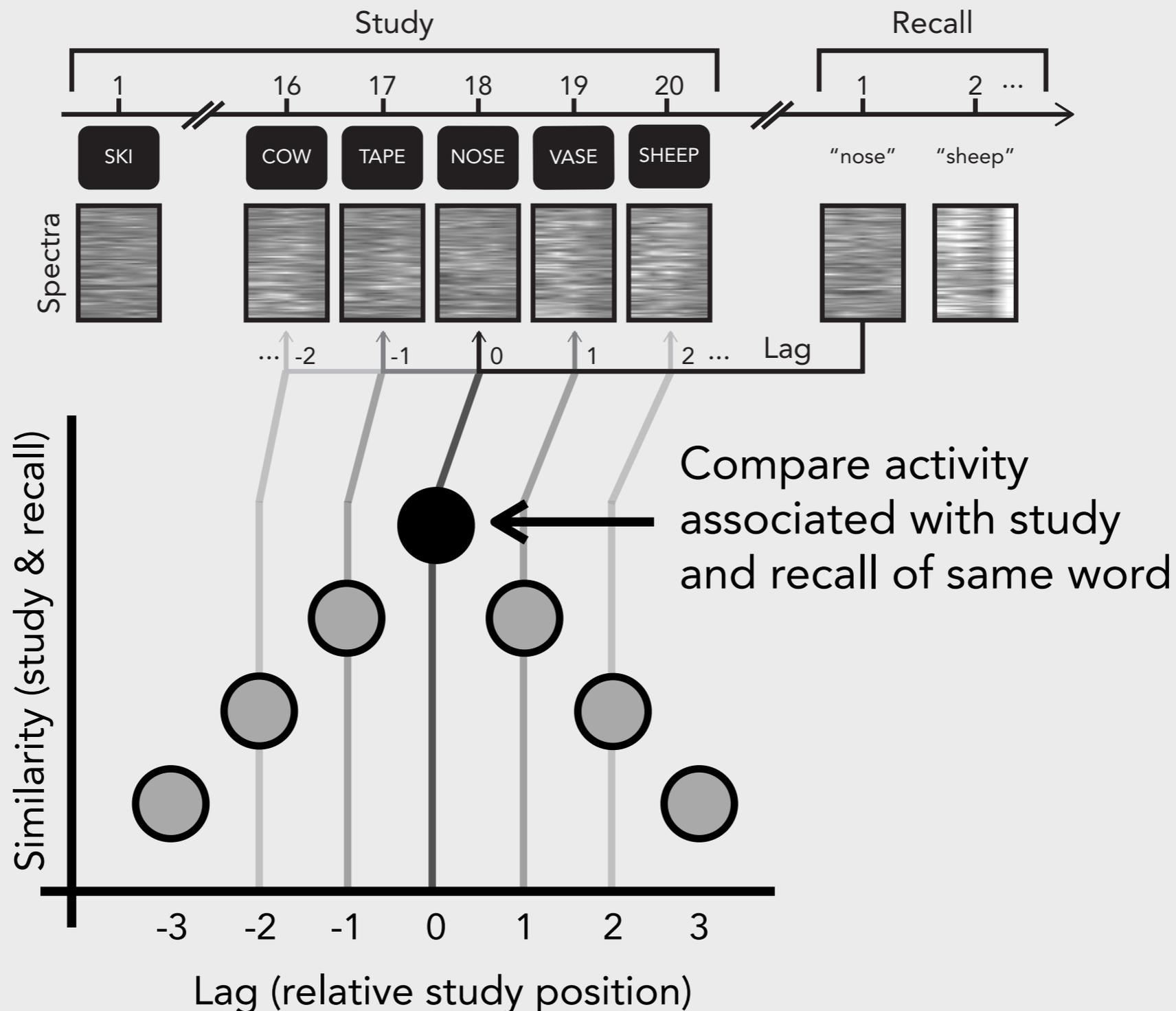
# Isolating neural representations of contextual cues



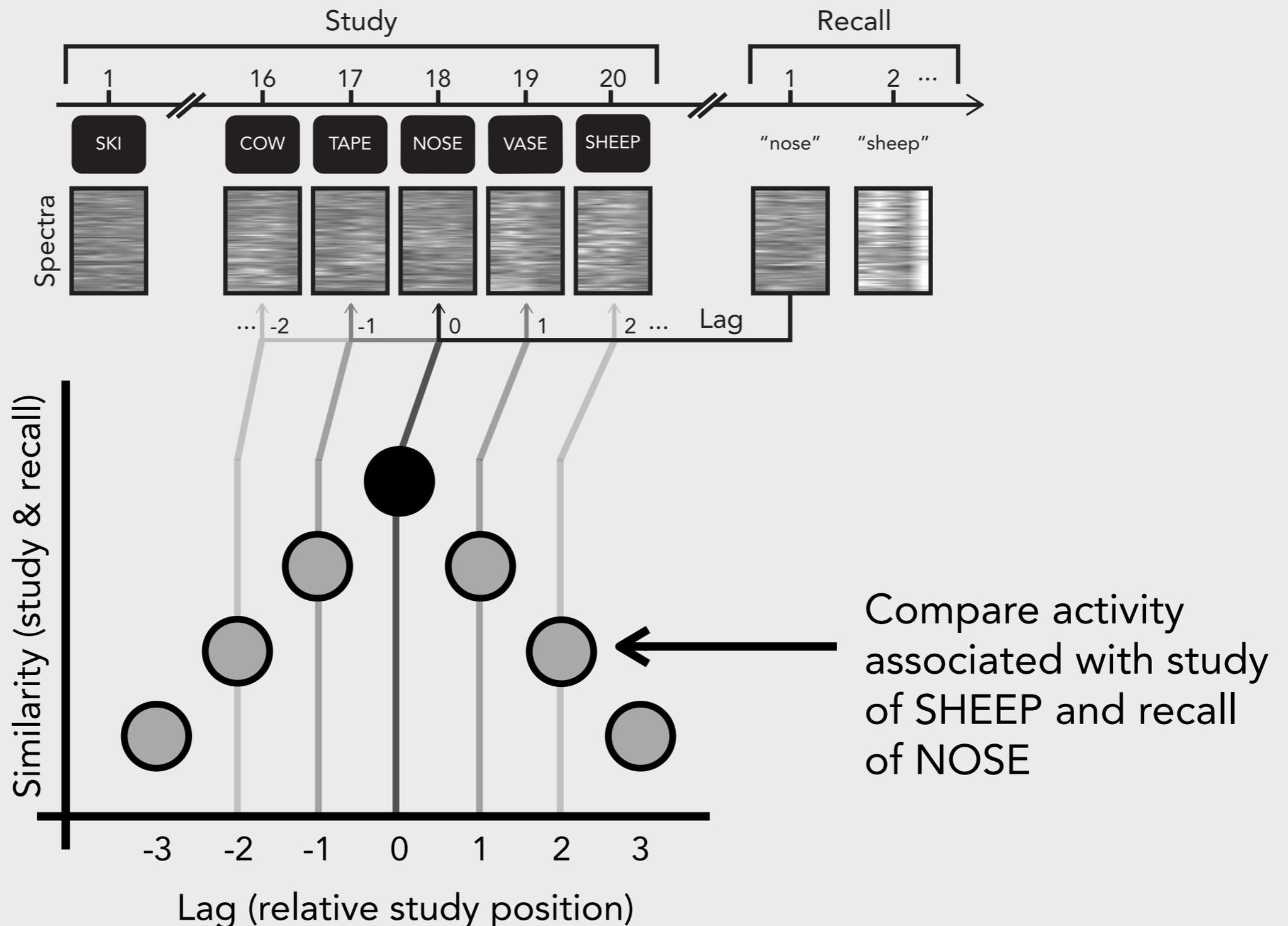
# Isolating neural representations of contextual cues



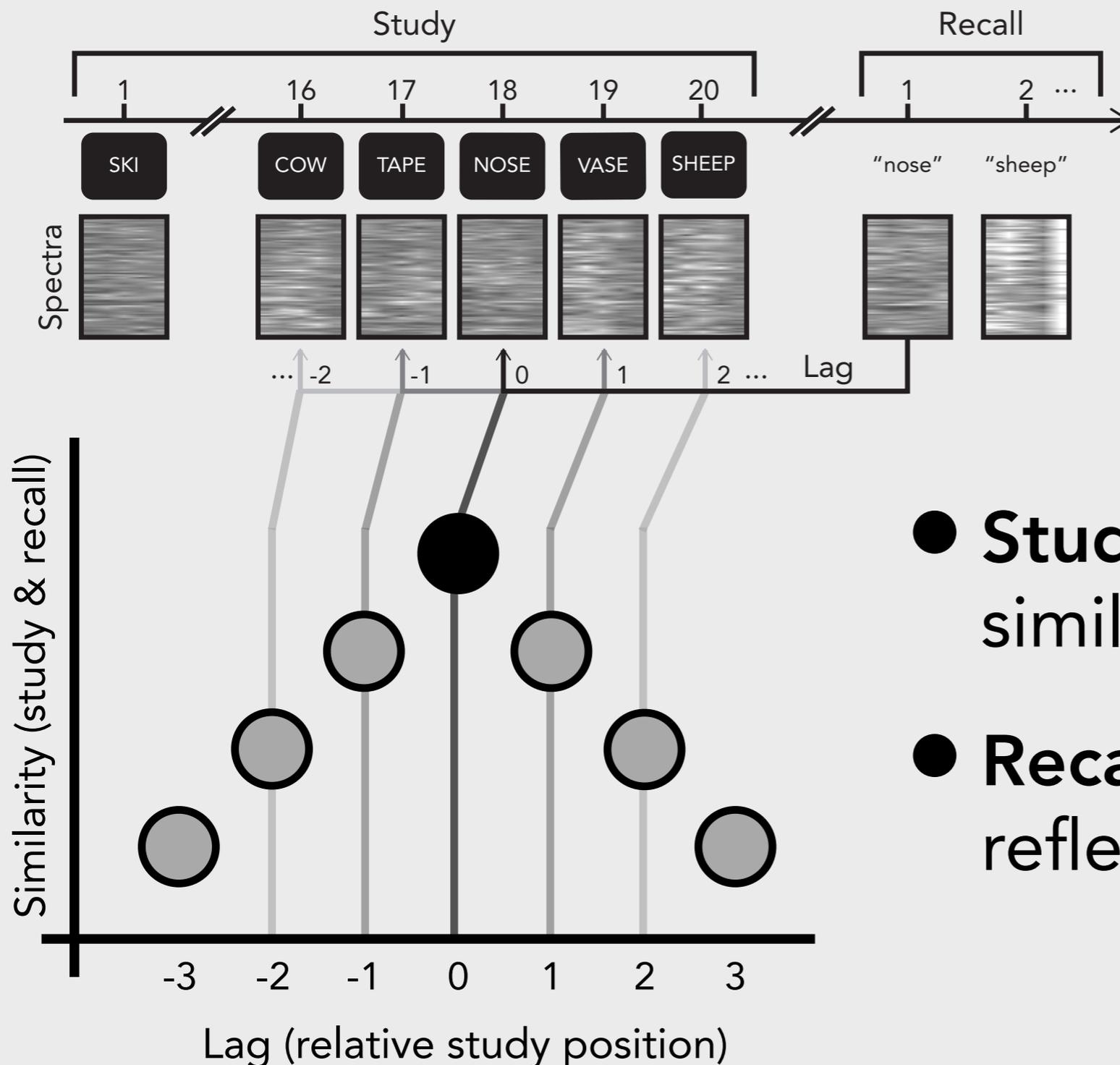
# Isolating neural representations of contextual cues



# Isolating neural representations of contextual cues

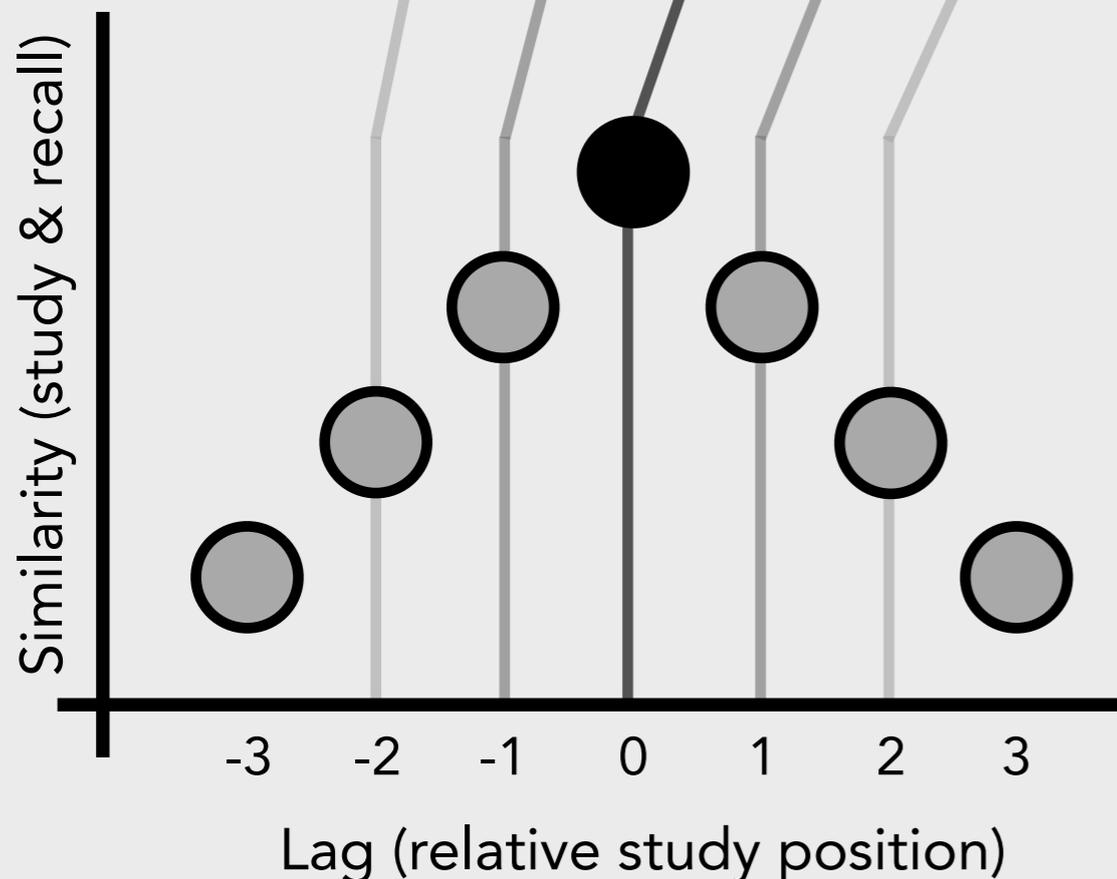
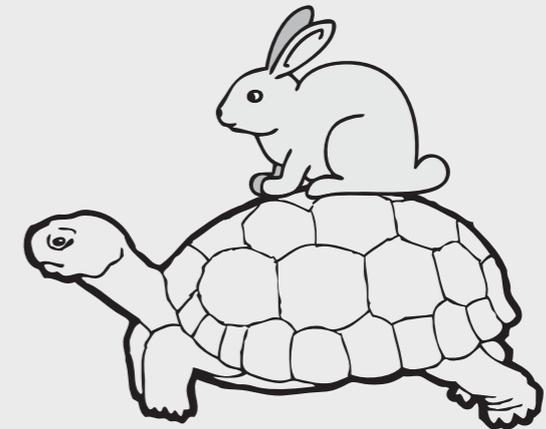
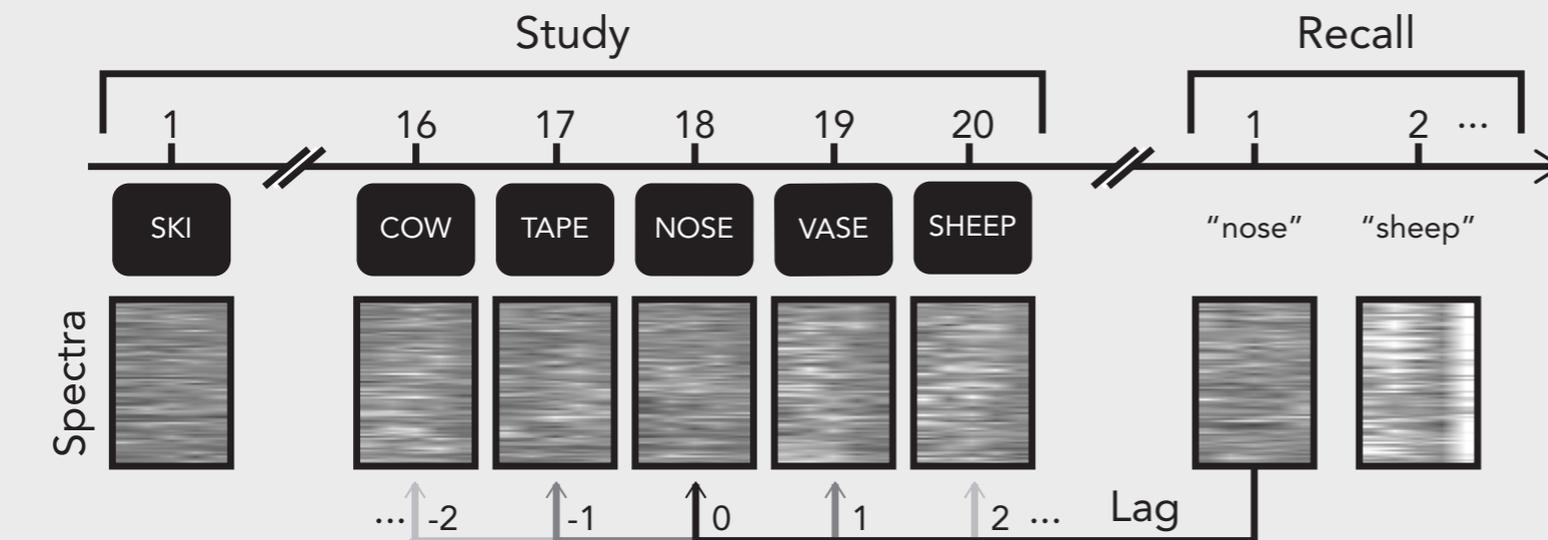


# The upside-down 'V' indicates reinstatement of slow drift (context)



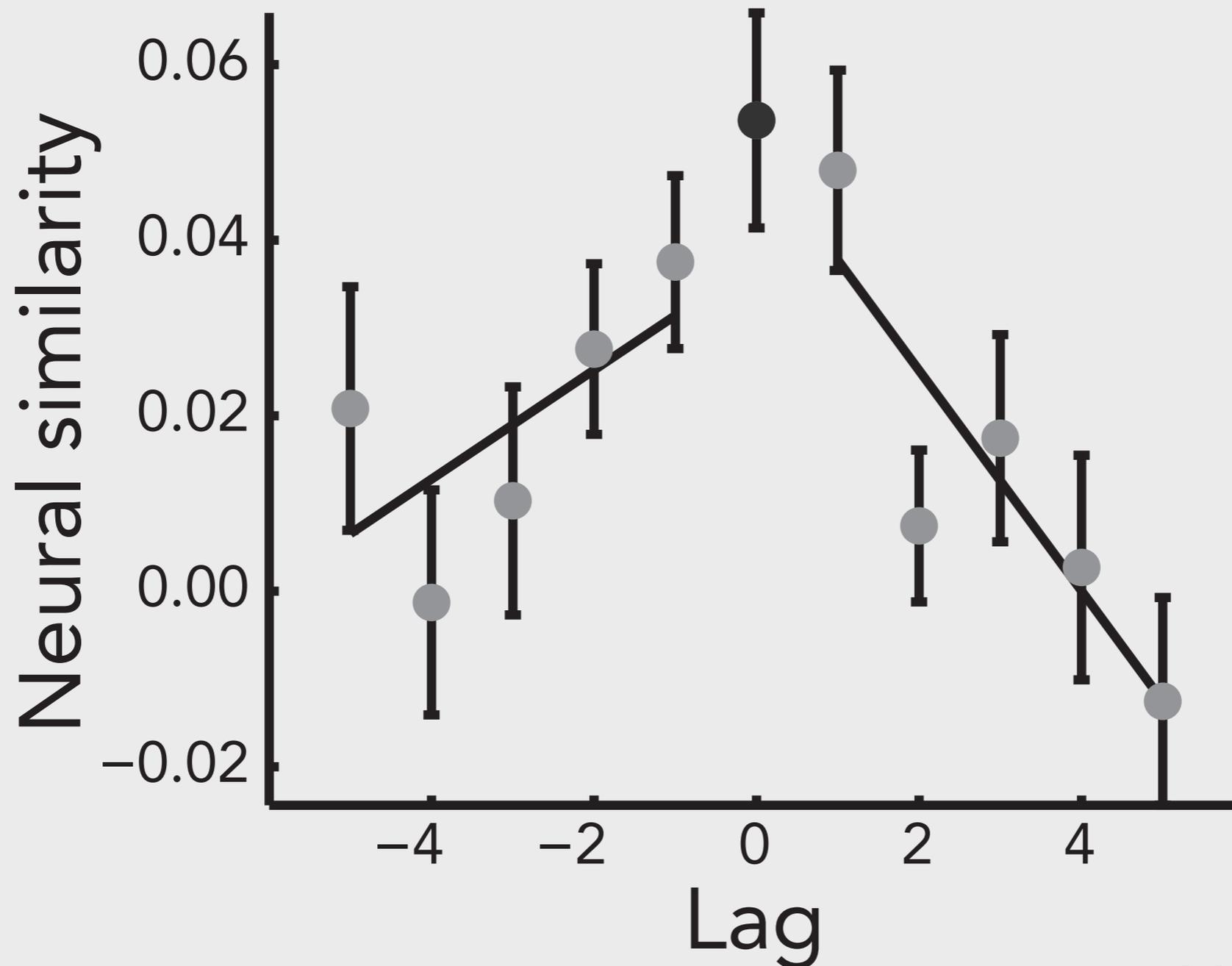
- **Study:** nearby items have similar representations
- **Recall:** similarity to study reflects reinstatement

# The upside-down 'V' indicates reinstatement of slow drift (context)

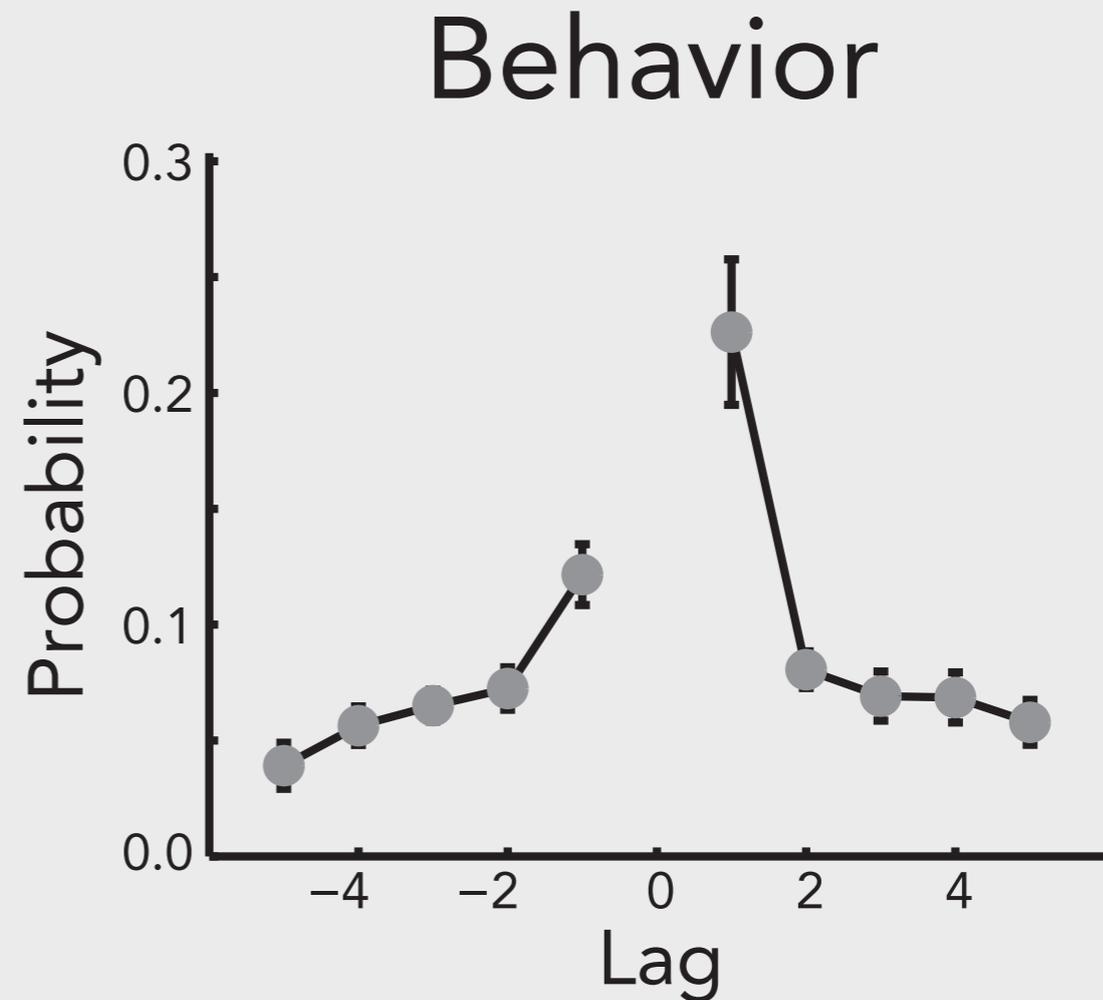
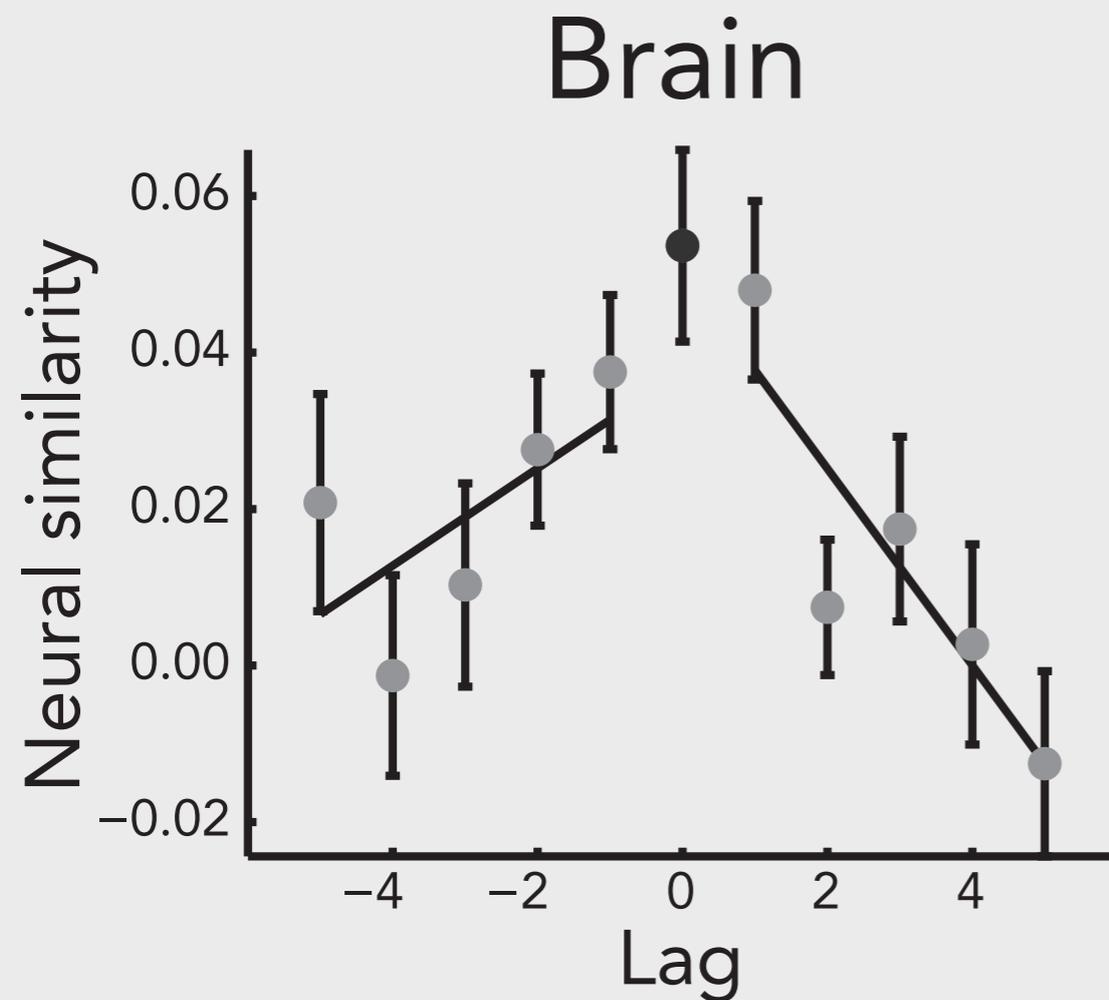


- **Study:** nearby items have similar representations
- **Recall:** similarity to study reflects reinstatement

# The neural underpinnings of temporal clustering



# The neural underpinnings of temporal clustering



# The neural underpinnings of temporal clustering

