

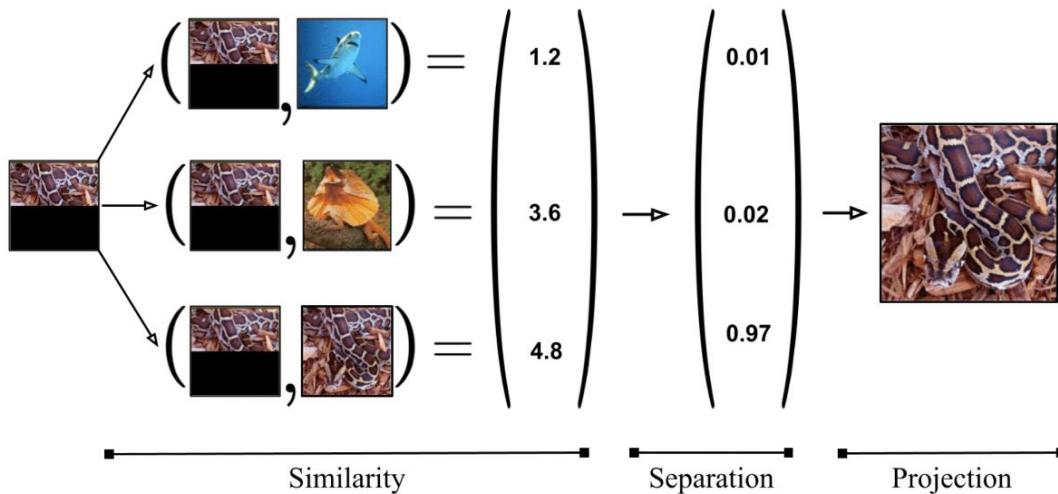
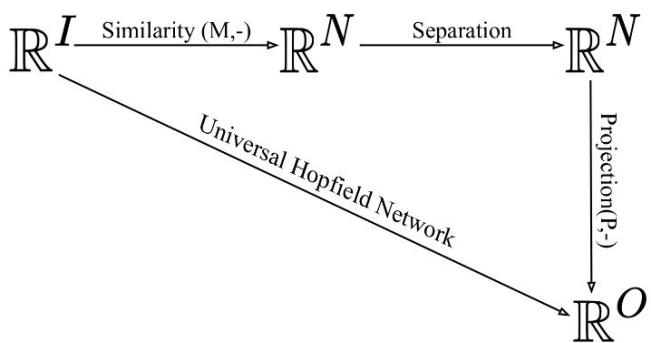


# UNIVERSAL HOPFIELD NETWORKS: A GENERAL FRAMEWORK FOR SINGLE-SHOT ASSOCIATIVE MEMORY MODELS

Beren Millidge, Tommaso Salvatori, Yuhang Song, Thomas Lukasiewicz, Rafal Bogacz

# General Framework

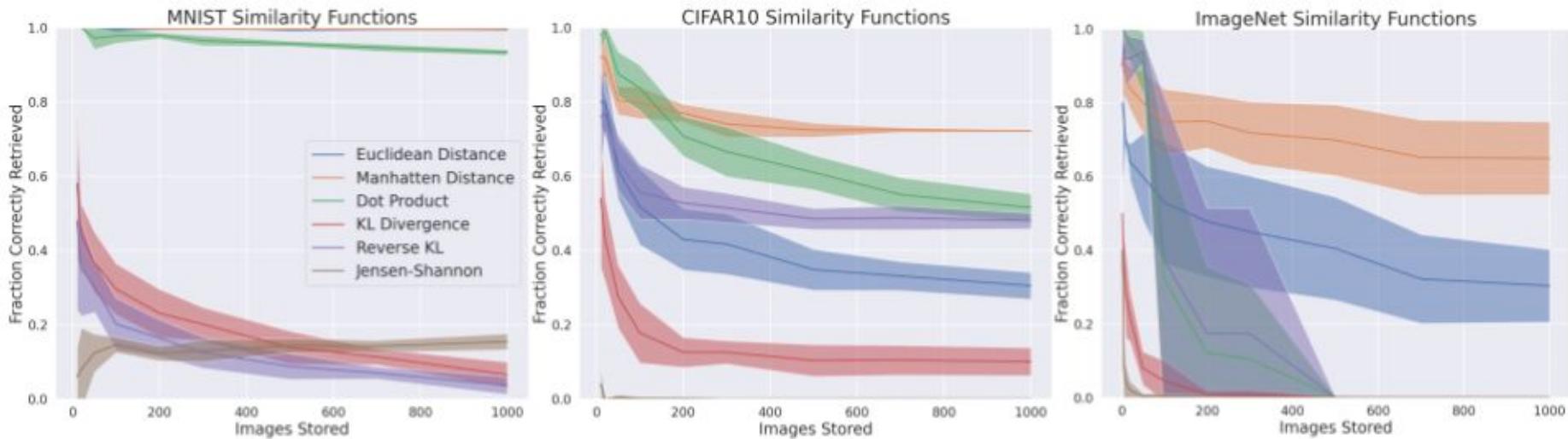
$$z = \underbrace{P}_{\text{Projection}} \cdot \underbrace{\text{sep}}_{\text{Separation}} \underbrace{(\text{sim}(M, q))}_{\text{Similarity}}$$



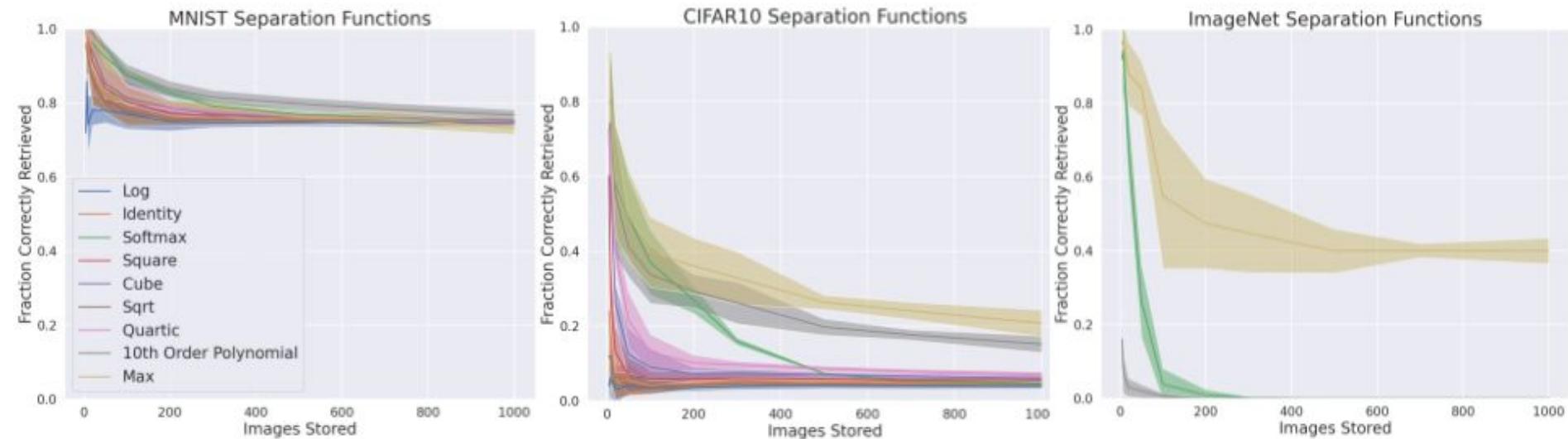
# Charting the Literature

<b>Memory Network</b>	<b>Similarity Function</b>	<b>Separation Function</b>
(Classical) Hopfield Network (HN)	Dot Product	Identity
Sparse Distributed Memory (SDM)	Hamming Distance	Threshold
Dense Associative Memory (DAM)	Dot Product	Polynomial
Modern Continuous Hopfield Network (MCHN)	Dot Product	Softmax

# Effect of Similarity Function



# Effect of Separation Function



# Summary and Future Work

- We have devised a general framework that lets us easily understand associative memory models proposed in the literature
- Closely related to transformer attention, and explains difference between heteroassociative and autoassociative memories
- Comes equipped with a neurobiological process theory
- Future work could apply insights into importance of similarity and separation function to transformer networks
- Investigate scaling up associative memories by encoding inputs into a latent space
- Extend the framework to understanding iterative associative memories