

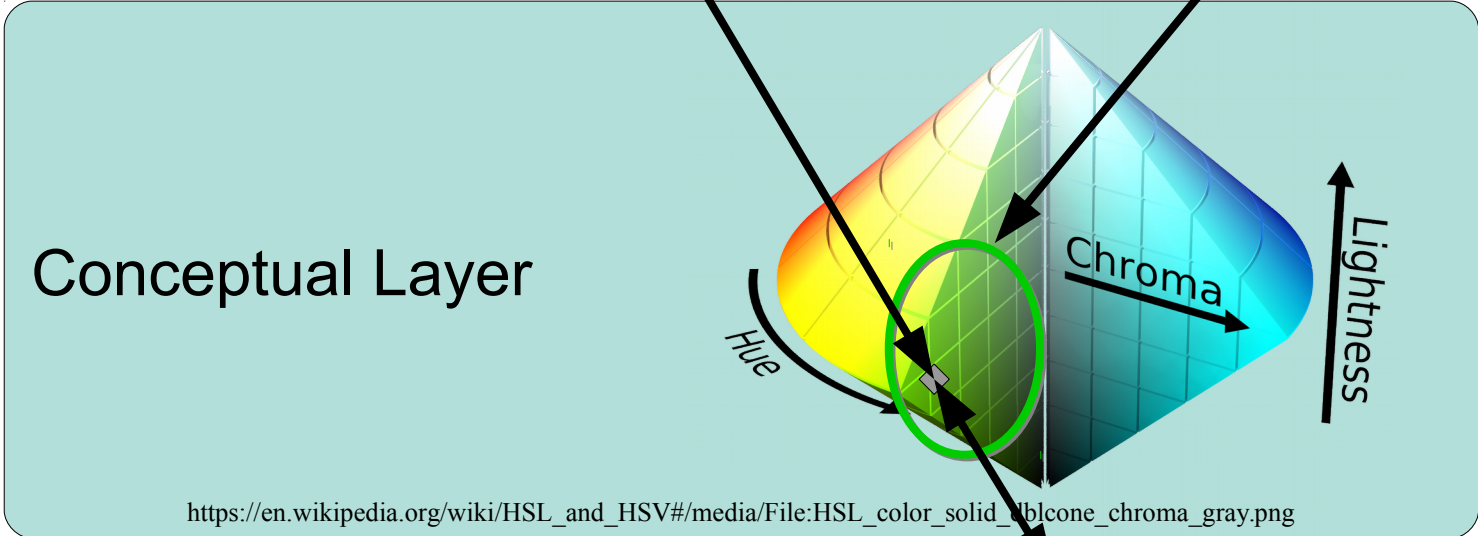
# Grounding Psychological Shape Space in Convolutional Neural Networks

Lucas Bechberger  
Institute of Cognitive Science  
Osnabrück University

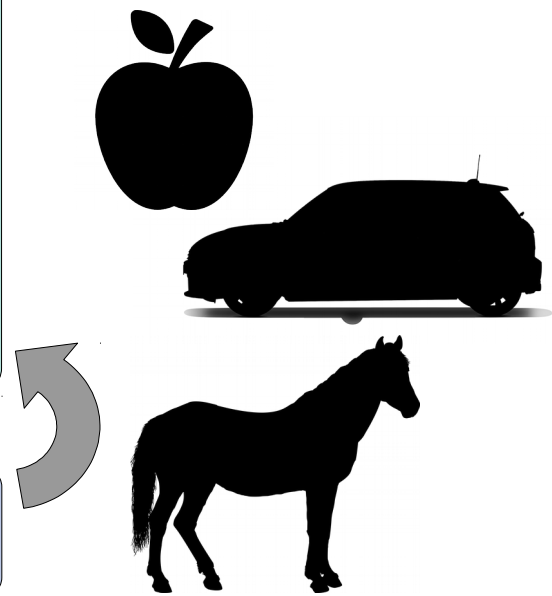
<https://www.lucas-bechberger.de>



Symbolic Layer  $\forall x: \text{apple}(x) \wedge \text{sour}(x) \Rightarrow \text{green}(x)$



Subsymbolic Layer [0.42; -1.337, 1.95583; ...]



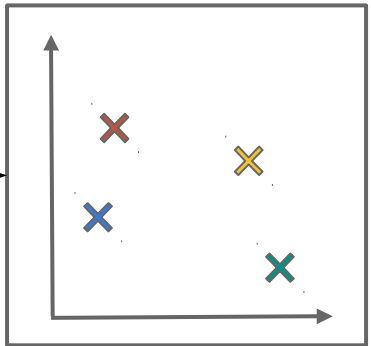
# Multidimensional Scaling on Psychological Data



Psychological Experiment

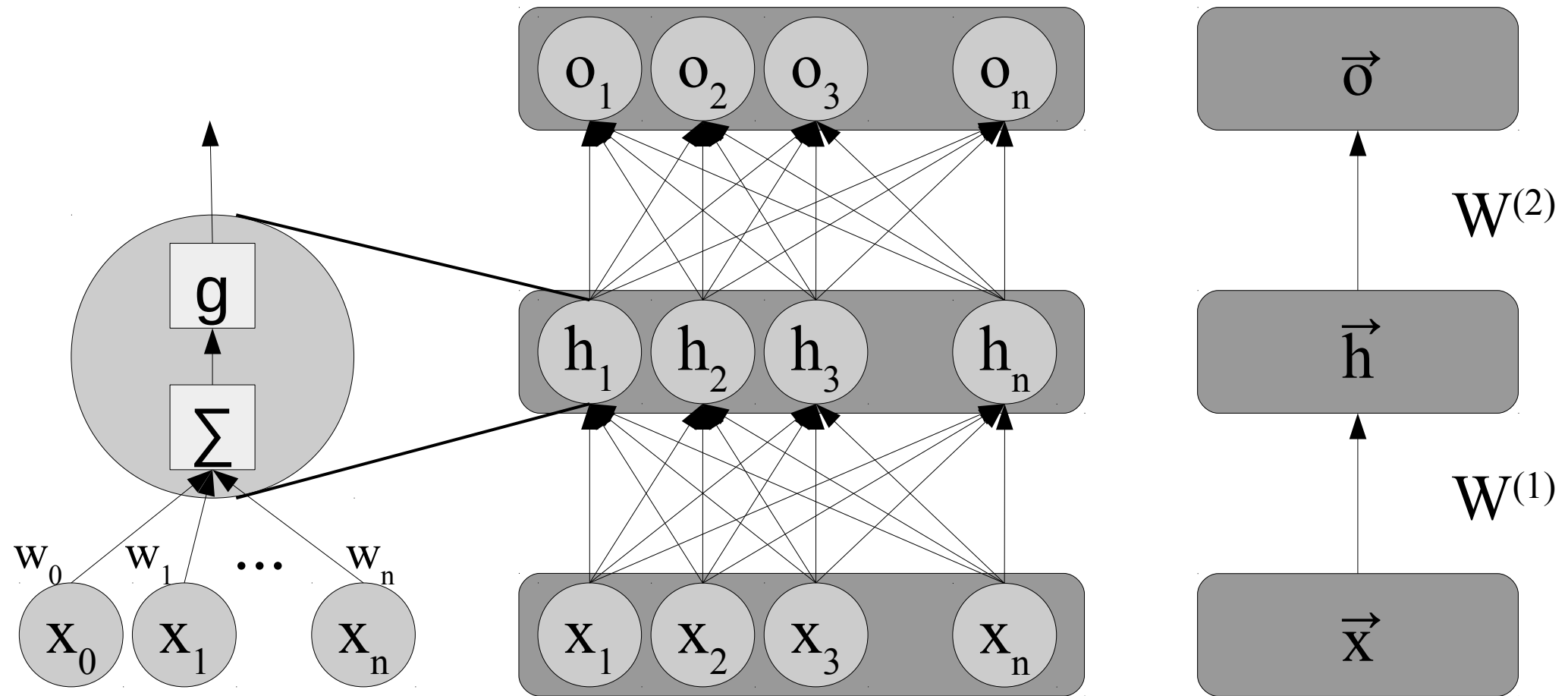
0	1	2	1
1	0	4	3
2	4	0	2
1	3	2	0

Multidimensional Scaling

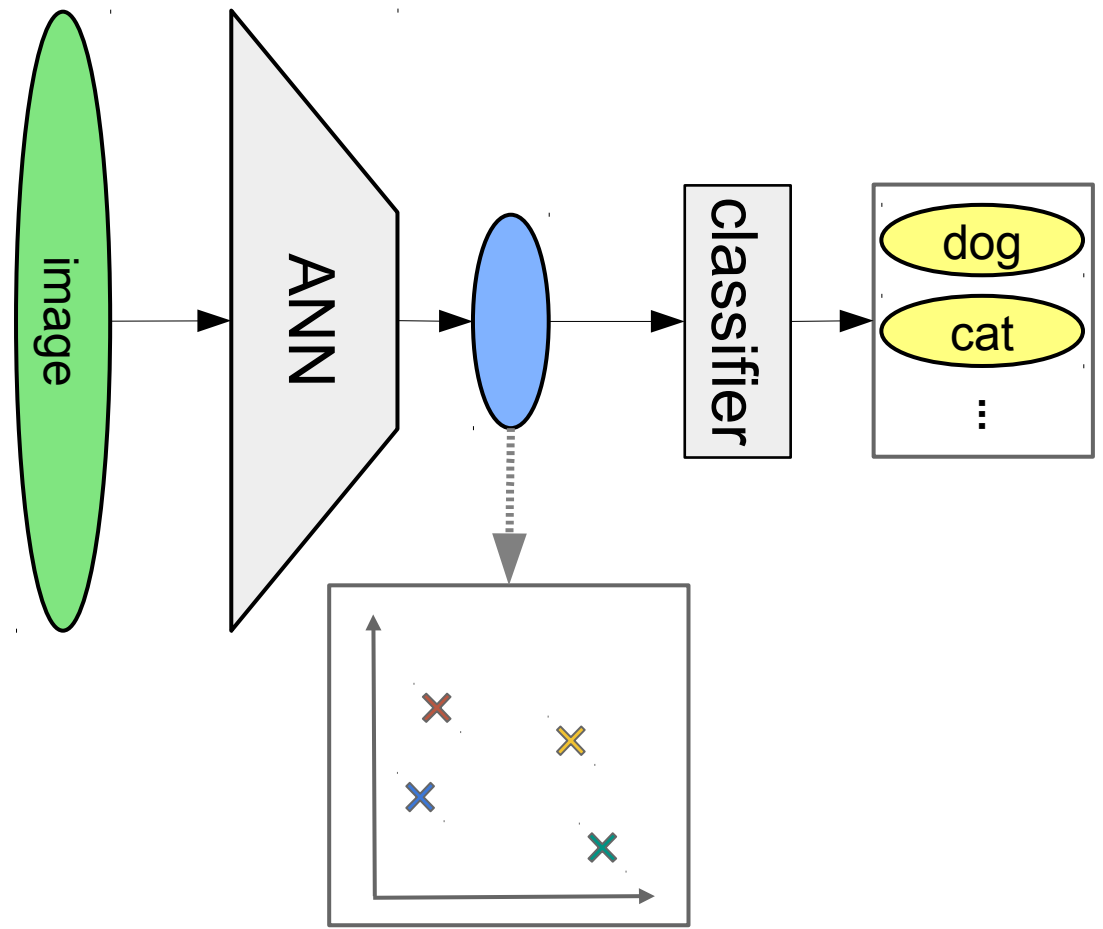
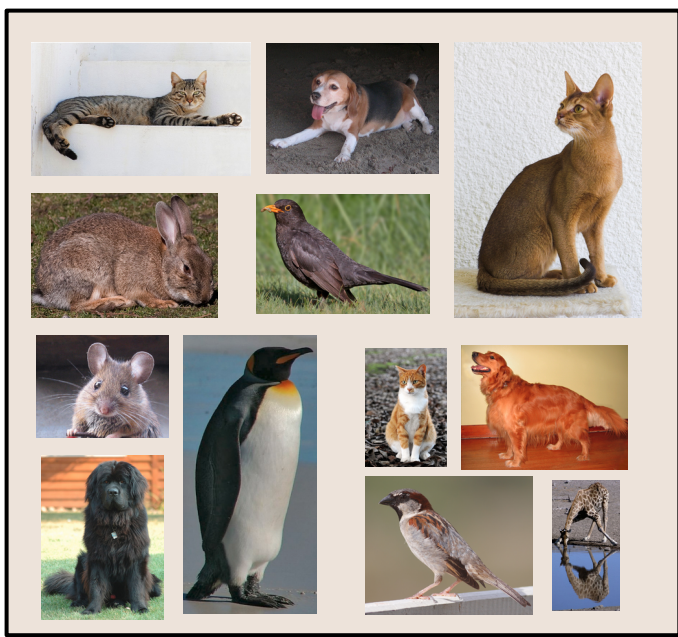


- + Psychological grounding
- Dealing with unseen inputs

Borg, I. & Groenen, J. F. Modern Multidimensional Scaling: Theory and Applications Springer-Verlag New York, 2005



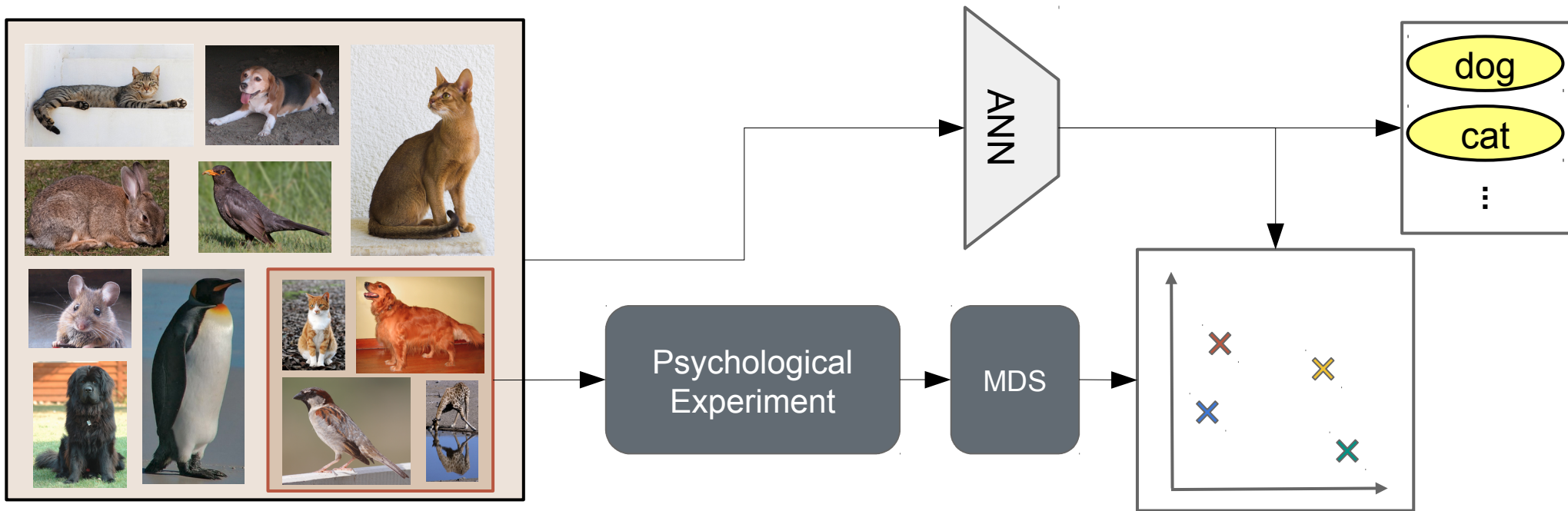
# Representation Learning with Neural Networks



- Psychological grounding
- + Dealing with unseen inputs

Bengio, Y.; Courville, A. & Vincent, P. Representation Learning: A Review and New Perspectives IEEE Transactions on Pattern Analysis and Machine Intelligence, Institute of Electrical and Electronics Engineers (IEEE), 2013, 35, 1798-1828

# A Hybrid Approach



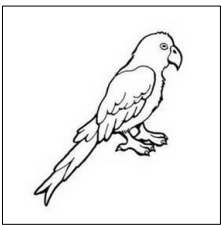
- + Psychological grounding
- + Dealing with unseen inputs

Bechberger, L. & Kypridemou, E. Mapping Images to Psychological Similarity Spaces Using Neural Networks. AIC 2018

Sanders, C. A. & Nosofsy, R.M. Using Deep-Learning Representations of Complex Natural Stimuli as Input to Psychological Models of Classification. CogSci 2018

Bechberger, L. & Kühnberger, K.-U. Generalizing Psychological Similarity Spaces to Unseen Stimuli. In: Concepts in Action: Representation, Learning, and Application, Springer International Publishing, 2021, 11-36

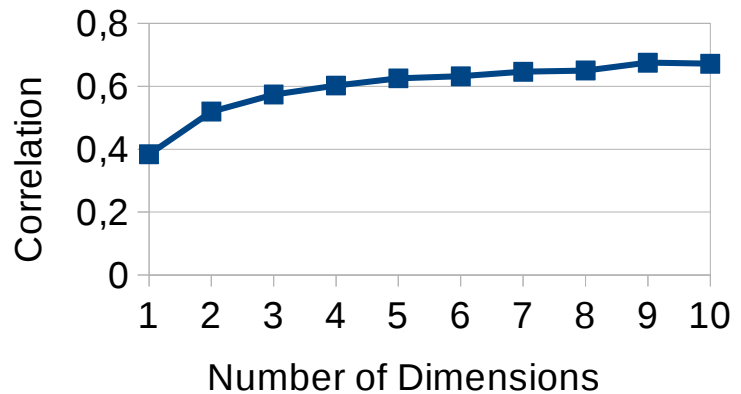
# The Similarity Spaces



Bechberger, L. & Scheibel, M. Alam, M.; Braun, T. & Yun, B. (Ed.) Analyzing Psychological Similarity Spaces for Shapes Ontologies and Concepts in Mind and Machine, Springer International Publishing, 2020, 204-207  
 Bechberger, L. & Scheibel, M. Modeling the Holistic Perception of Everyday Object Shapes with Conceptual Spaces. In preparation

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- Dissimilarity = Distance ✓
- Small Convex Regions ✓
- Interpretable Directions ✓

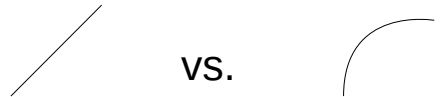


**FORM**



vs.

**LINES**



vs.

**ORIENTATION**



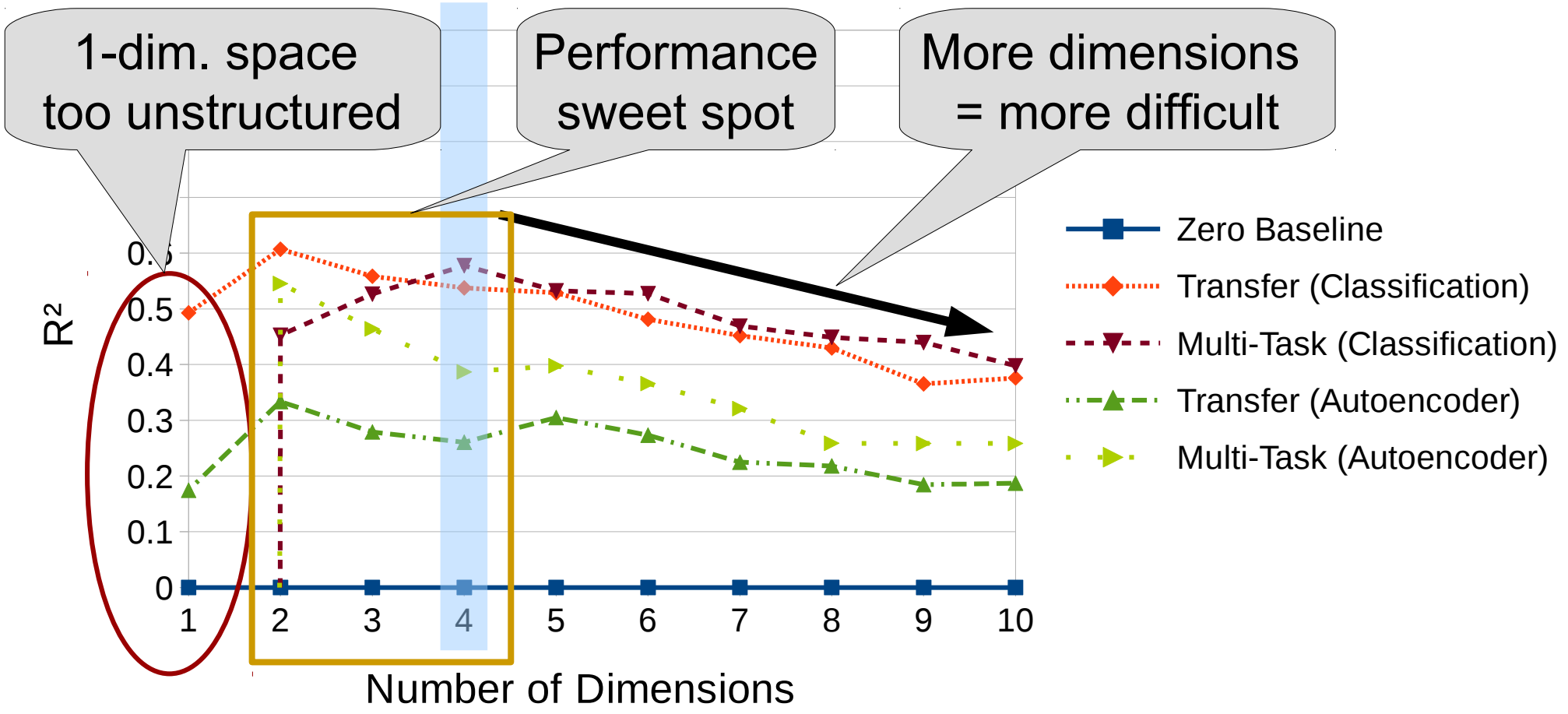
vs.

vs.

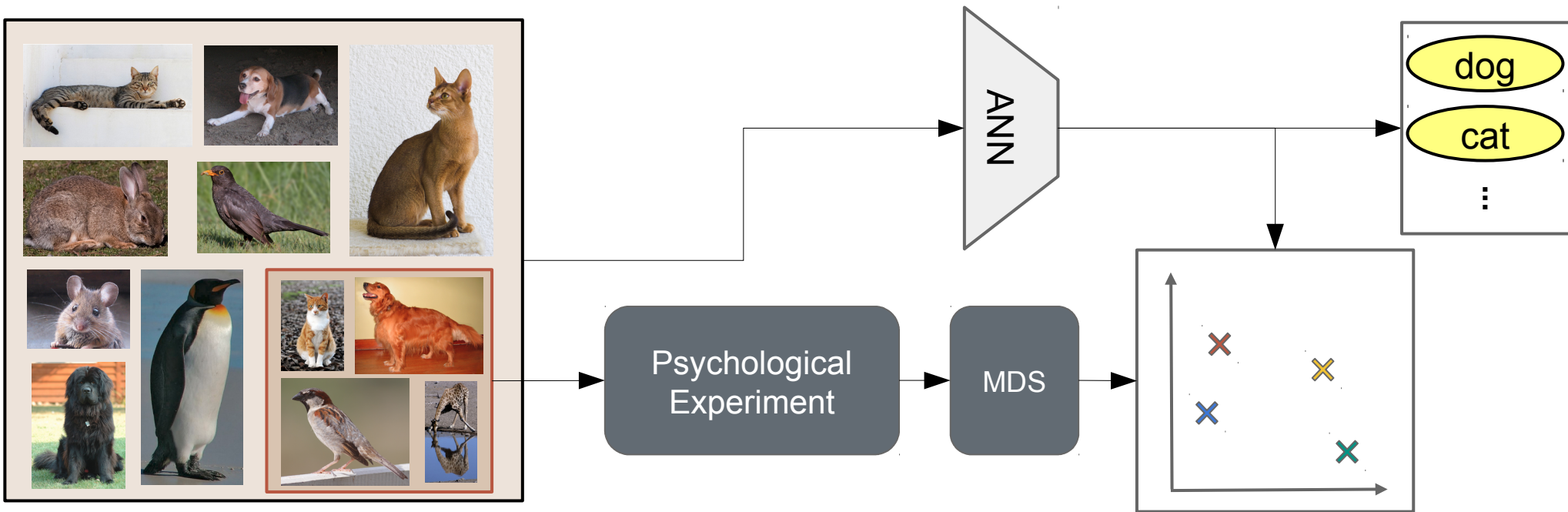
	Classification	Autoencoder
Transfer Learning	<p><math>R^2 \approx 0.52</math></p>	<p><math>R^2 \approx 0.26</math></p>
Multi-Task Learning	<p><math>R^2 \approx 0.58</math></p>	<p><math>R^2 \approx 0.42</math></p>



# Learning a Mapping From Images Into Shape Space



# A Hybrid Approach



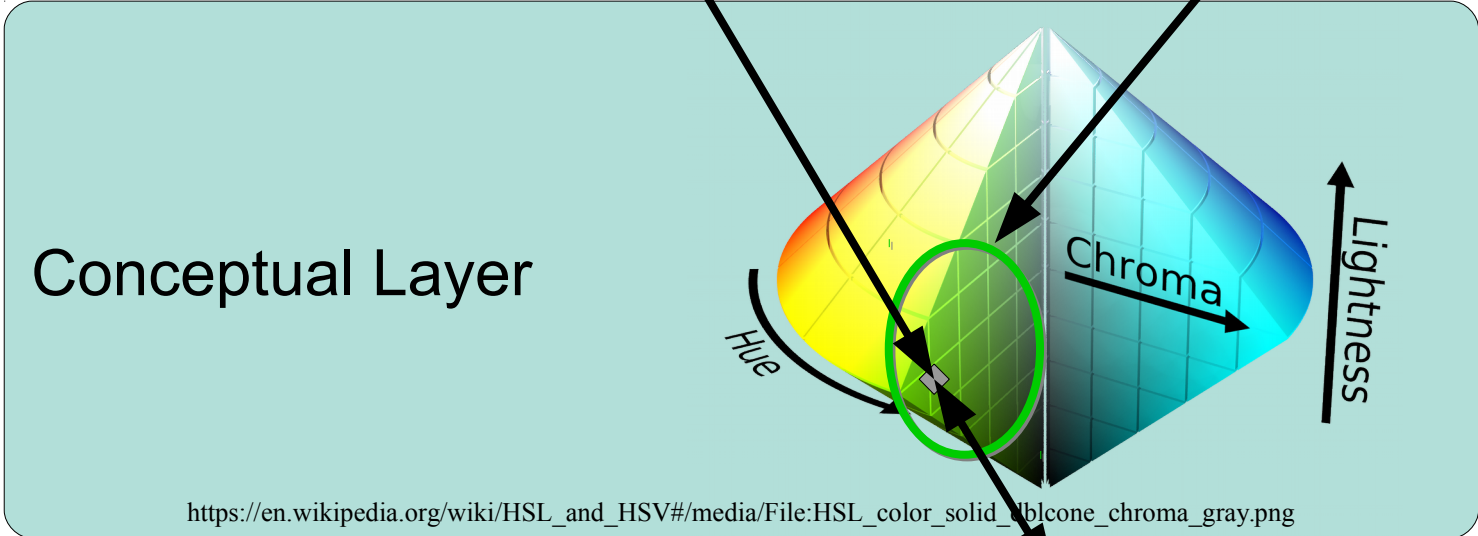
- ✓ It works in principle
- ✓ Experimental setup matters
- ? Performance still limited

Bechberger, L. & Kypridemou, E. Mapping Images to Psychological Similarity Spaces Using Neural Networks. AIC 2018

Sanders, C. A. & Nosofsky, R.M. Using Deep-Learning Representations of Complex Natural Stimuli as Input to Psychological Models of Classification. CogSci 2018

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Symbolic Layer  $\forall x: \text{apple}(x) \wedge \text{sour}(x) \Rightarrow \text{green}(x)$



Subsymbolic Layer  $[0.42; -1.337, 1.95583; \dots]$

- ? Reasoning on top of this
- ? Apply to other domains

