

# Grounding Psychological Shape Space in Convolutional Neural Networks

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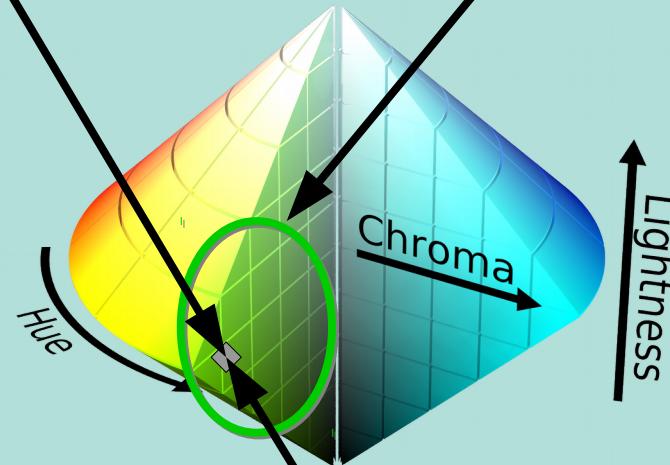
<https://www.lucas-bechberger.de>



Symbolic Layer

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

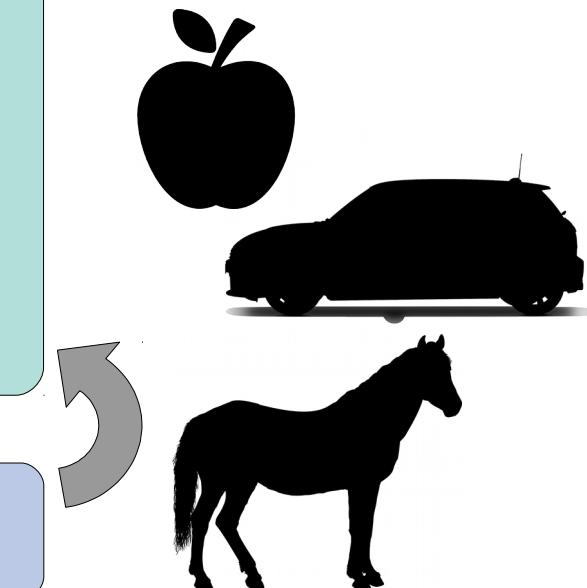
Conceptual Layer



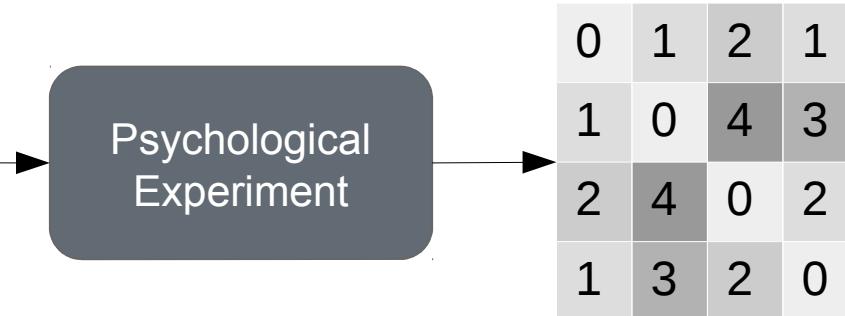
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Subsymbolic Layer

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



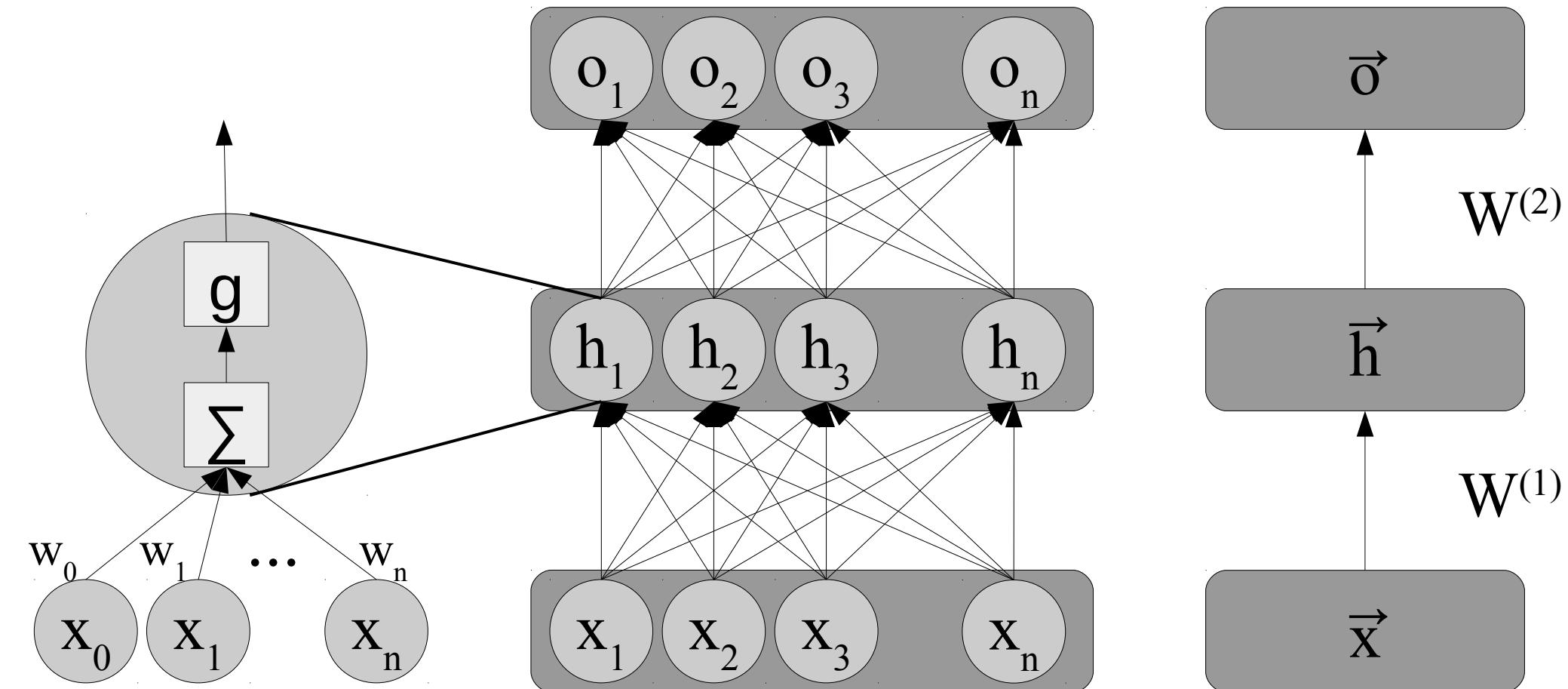
# Multidimensional Scaling on Psychological Data



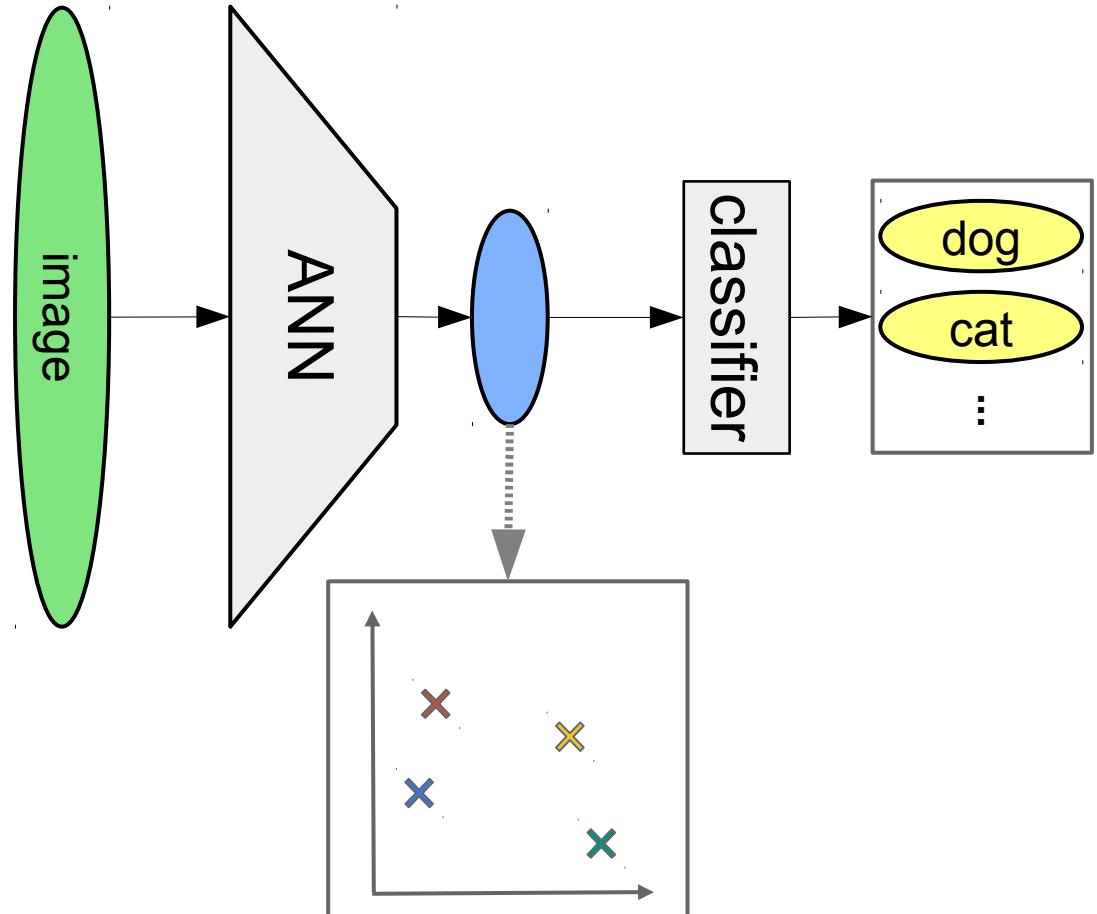
- + Psychological grounding
- Dealing with unseen inputs

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

# Representation Learning with Neural Networks



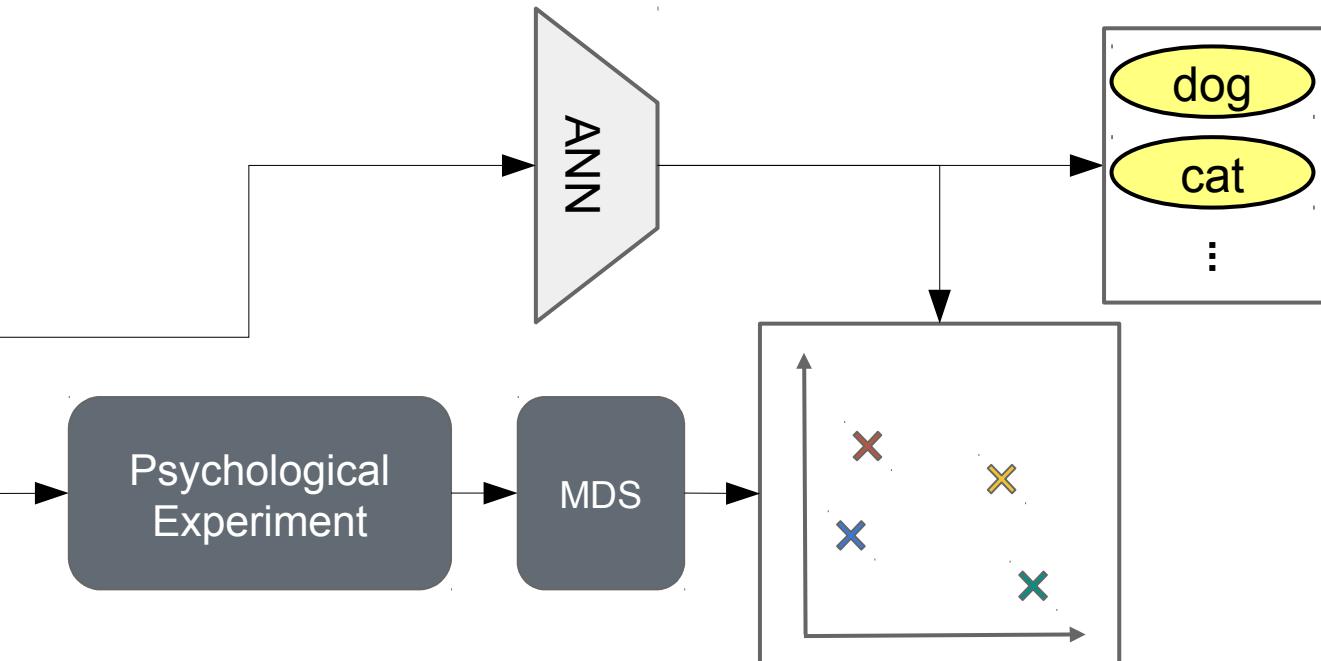
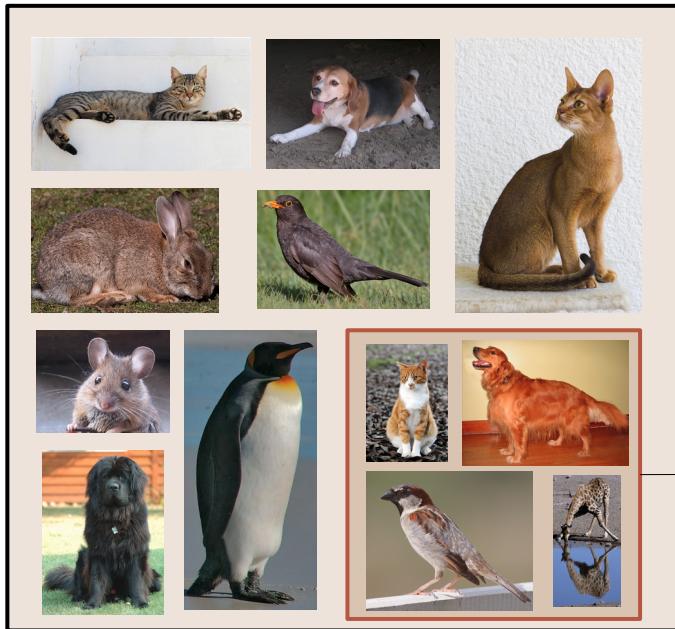
# Representation Learning with Neural Networks



- 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



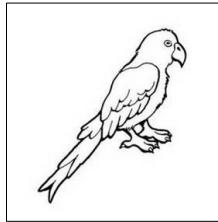
- + Psychological grounding
- + Dealing with unseen inputs

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

Sanders, C. A. & Nosofsky, 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



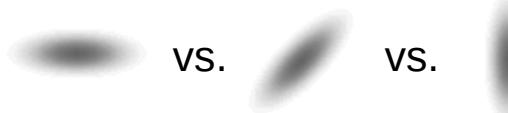
VS.

LINES

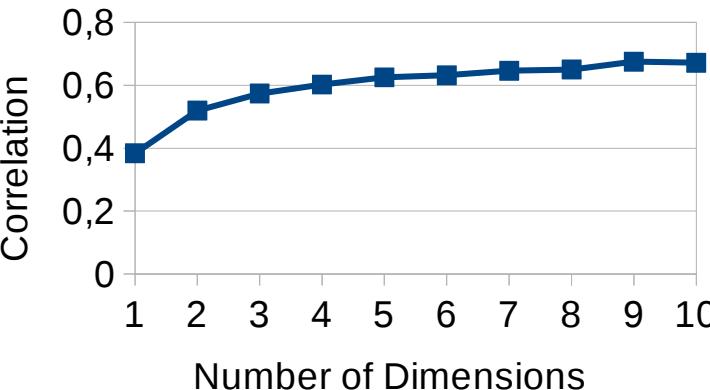


VS.

ORIENTATION



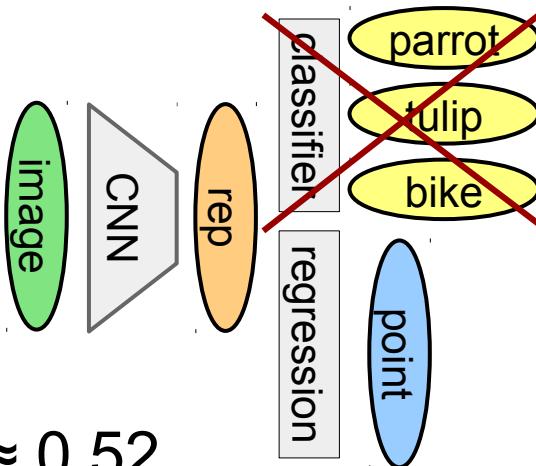
VS.



# Learning a Mapping From Images Into Shape Space

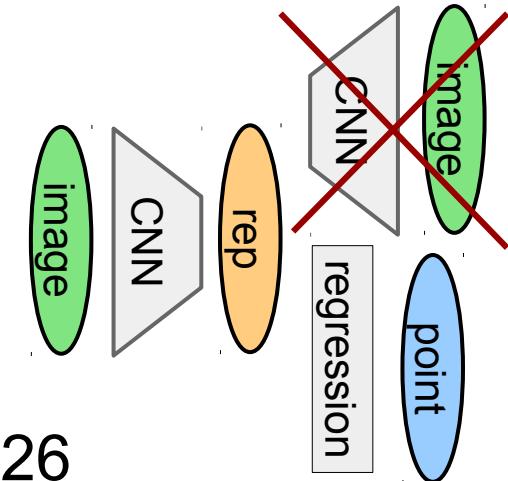
## Transfer Learning

### Classification



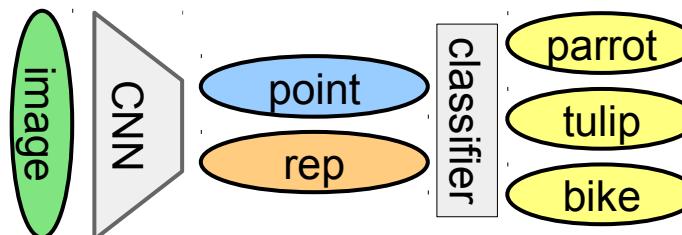
$$R^2 \approx 0.52$$

### Autoencoder

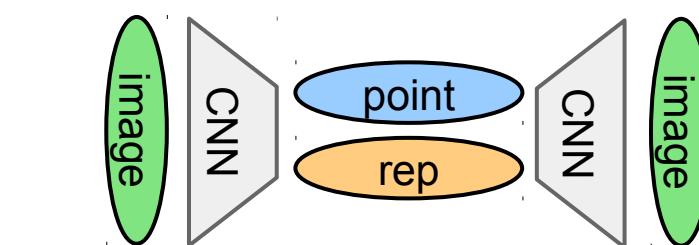


$$R^2 \approx 0.26$$

## Multi-Task Learning



$$R^2 \approx 0.58$$

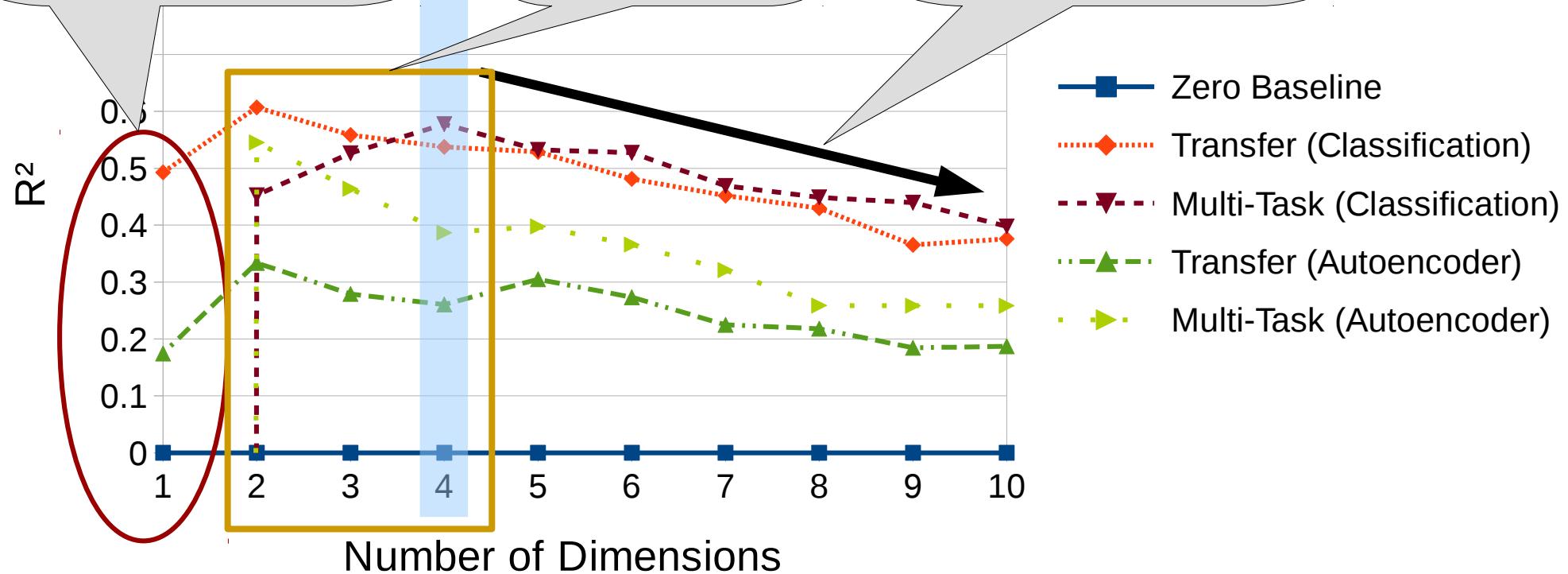


$$R^2 \approx 0.42$$

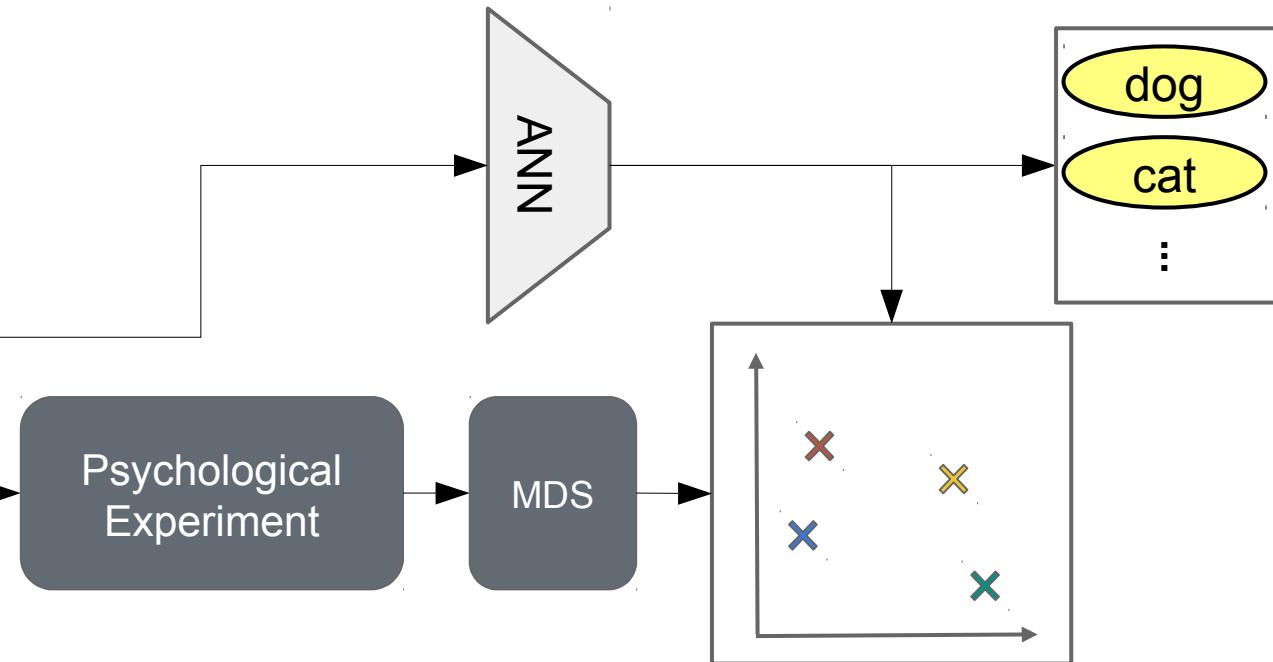
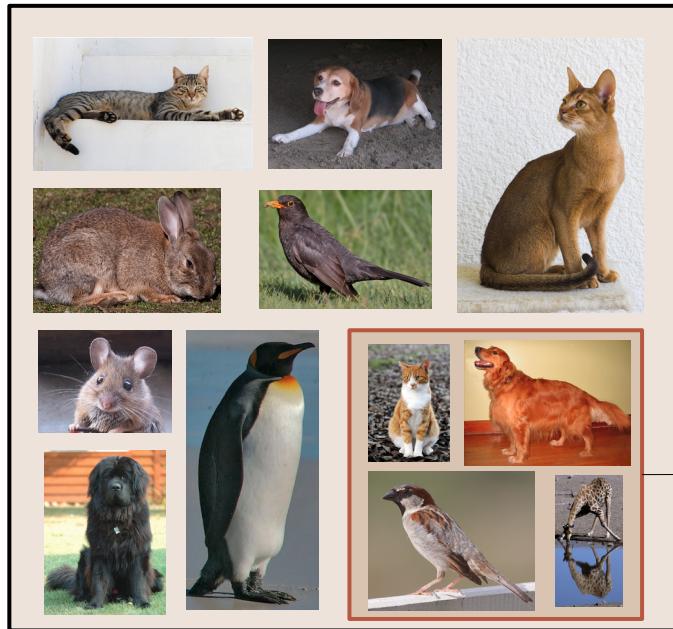
1-dim. space  
too unstructured

Performance  
sweet spot

More dimensions  
= more difficult



# A Hybrid Approach



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

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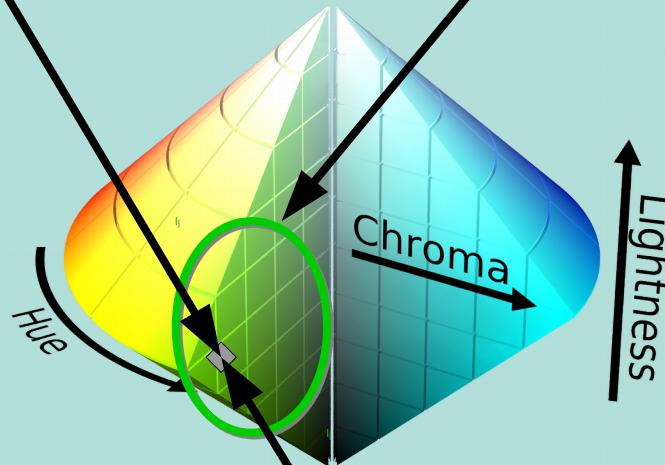
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Reasoning on top of this

Apply to other domains

