

Conceptual Spaces

A Bridge Between Neural and Symbolic Representations?

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Peter Gärdenfors, "Conceptual Spaces: The Geometry of Thought", MIT press, 2000

The Different Layers of Representation



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OSNABRÜCK Dimensions & Domains

- Quality dimensions
 - Different ways stimuli are judged to be similar or different
 - Interpretable by a human
 - E.g., temperature, weight, brightness, pitch
- Domain
 - Set of dimensions that inherently belong together
 - Color: hue, saturation, and brightness
- Distance in this space is inversely related to similarity
 - Within a domain: Euclidean distance
 - Between domains: Manhattan distance



• A natural property is a convex region within a domain.

OSNABRÜCK Example: The Color Domain



https://en.wikipedia.org/wiki/HSL_and_HSV#/media/File:HSL_color_solid_dblcone_chroma_gray.png

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- Example: "apple"
 - Color: red

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- Shape: spherical
- Texture: smooth
- Taste: sweet



- Defined across multiple domains: combination of properties
 - Different "importance" to the concept (influenced by context)
 - Potentially correlated

Connection to Psychology

- Prototype theory of concepts
 - Each concept is mentally represented by a prototype
 - e.g., the most typical instance
- Conceptual spaces
 - Central point of convex region can be interpreted as prototype
 - Voronoi-tesselation based on prototypes results in convex sets





https://commons.wikimedia.org/wiki/File:CentroidalVoronoiTessellation2.png

OSNABRÜCK Reasoning in Conceptual Spaces

- Geometric relationships between regions
 - Sky blue is a subset of blue
 - Projecting *apple* onto the color domain results in *red* region
- Concept combination: "green banana"
 - Narrow down the color region of *banana* to green region
 - Correlations between domains yield further updates:
 - Consistency is *solid*, taste is *bitter*
- Use betweenness and similarity for plausible reasoning
 - Both bachelor and PhD students have to pay a certain fee
 - What about Master students?

J. Derrac and S. Schockaert, "Inducing Semantic Relations from Conceptual Spaces: A Data-Driven Approach to Plausible Reasoning", Artificial Intelligence, 2015

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OSNABRÜCK Learning Dimensions with ANNs

 Information Maximizing Generative Adversarial Networks (InfoGAN)



X. Chen et al., "InfoGAN: Interpretable Representation Learning by Information Maximizing Generative Adversarial Nets", Advances in Neural Information Processing Systems, 2016

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Learning Concepts Through Clustering

- We look for meaningful regions in the conceptual space
 - Concepts = clusters of data points
- Observed objects usually come without class information
 - unsupervised

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- Observing one object at a time, limited memory
 - Stream of data points, incremental processing



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Thank you for your attention!

Questions? Comments? Discussions?



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