# What Machine Learning Tells Us About the Mathematical Structures of Concepts



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## <u>Question: what kind of mathematical structure is needed to model concepts?</u>

"What are concepts?" is one of the fundamental questions in philosophy, where Aristotle, Kant, Wittgenstein, etc. have proposed different models of concepts. Meanwhile, machine learning literature has developed powerful methods to learn *representation* from data, as well as mathematical models to analyze such representations. This presentation aims to bridge between these two traditions by identifying mathematical models & machine-learning counterparts of the (1) Aristotelian, (2) Wittgensteinian, (3) Functional, and (4) Symmetry-based theories of concept.

## <u>Aristotelian Abstractionism</u>

Idea: concepts are formed by extracting common properties of individual things. Cf. Aristotle, Locke, etc. (see Heis 2008).

## <u>Cluster Concept Theory</u>

Idea: concepts are groups of similar things or impressions.

## Functional Model

Idea: concepts are a set of properties having a certain functional relationship. Cf. Lotze & Cassirer's Funktionsbegriff; Gärdenfors's topological concept.

# <u>Symmetry</u>

Idea: concepts are a set of properties

Modeled by...

Boolean Algebra

with abstraction ∨ and instantiation ∧ operators



Cf. Wittgenstein's family resemblance; Rosch's prototype theory.



Vector Spaces

Concepts are vectors in a high-dim vector space, with their similarity being measured by some metric, e.g.  $\cos(\theta)$ .





# Manifolds

Possible data points "live" on a submanifold in a non-euclidean space, where the submanifold (=concept) structure is determined by a set of functions.



invariant under a certain group of transformations.

Cf. Berkeley, van Fraassen, Jantzen



Groups

Representations transform in accordance with changes in objects, i.e. the representation map f commutes with object transformations/groups g, s.t.

 $g \cdot f(x) = f(g \cdot x)$ 



#### Pros & Cons

✓ Intuitive

- ✓ Well-studied syntax & semantic relationship (Stone duality)
- X Allows arbitrary abstraction (e.g. beef V cherry = red juicy food?)

X Difficult to learn from data (GOFAI)

## Pros & Cons

- ✓ Easy to learn from data (e.g. wordembeddings such as word2vec)
- ✓ Allows some algebraic operations (e.g. king - man + women = queen)
- X Difficult to deal with other logical connectives, such as negation or disjunction.
- X Difficult to learn complex expressions (e.g. sentences, etc.)

## Pros & Cons

- ✓ Learnable from data (e.g. generative models such as VAE)
- ✓ One can create new images by sliding them upon a submanifold.



each dimension of the submanifold represents distinct characteristics?

X These characteristics are not always distinct but are sometimes "entangled" Moreover, representations must be *disentangled*, so that one can tweak one characteristic without affecting others (each dimension encodes a distinct characteristic).



- There is a decomposition  $Z = Z_1 \times ... \times Z_n$  of the latent space corresponding to the decomposition  $G = G_1 \times ... \times G_n$  of transformations
- s.t.  $G_i$  acts on  $Z_i$  but not other  $Z_j$ ,  $(i \neq j)$

(Higgins+ 2018)

#### <u>Geometrical vs. Algebraic sides of concepts</u>

#### Geometrical

- "Similarity" measured by a metric on the conceptual space (Wittgenstein)
- A "concept" must form a connected region (Gärdenfors)
- The space of possible concepts must be constrained (Lotze, Cassirer)

#### Algebraic

- Concepts are related to each other by algebraic operations (abstraction, addition, negation, transformation, etc.)
- Each concept embodies a set of rules that prescribe how its properties are transformed.

An adequate mathematical framework for modeling concepts must combine their geometrical & algebraic aspects.

#### <u>References</u>

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