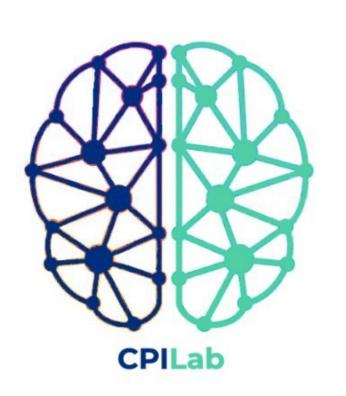




Learning Structure from the Ground-Up— Hierarchical Representation Learning by Chunking

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D Evaluation representation of the sequence of

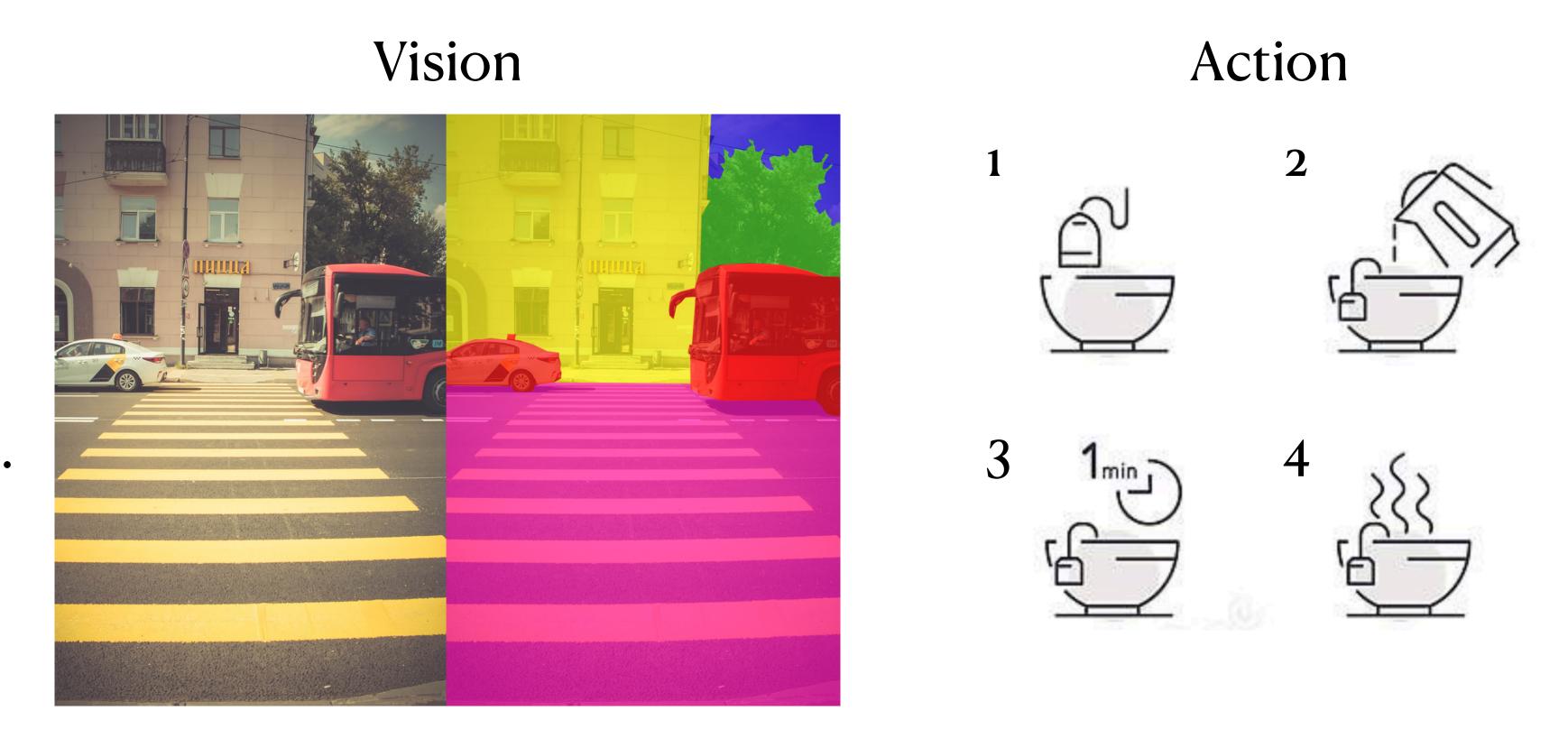
Chunking is a ubiquitous learning phenomenon

Language

... As you might know ...

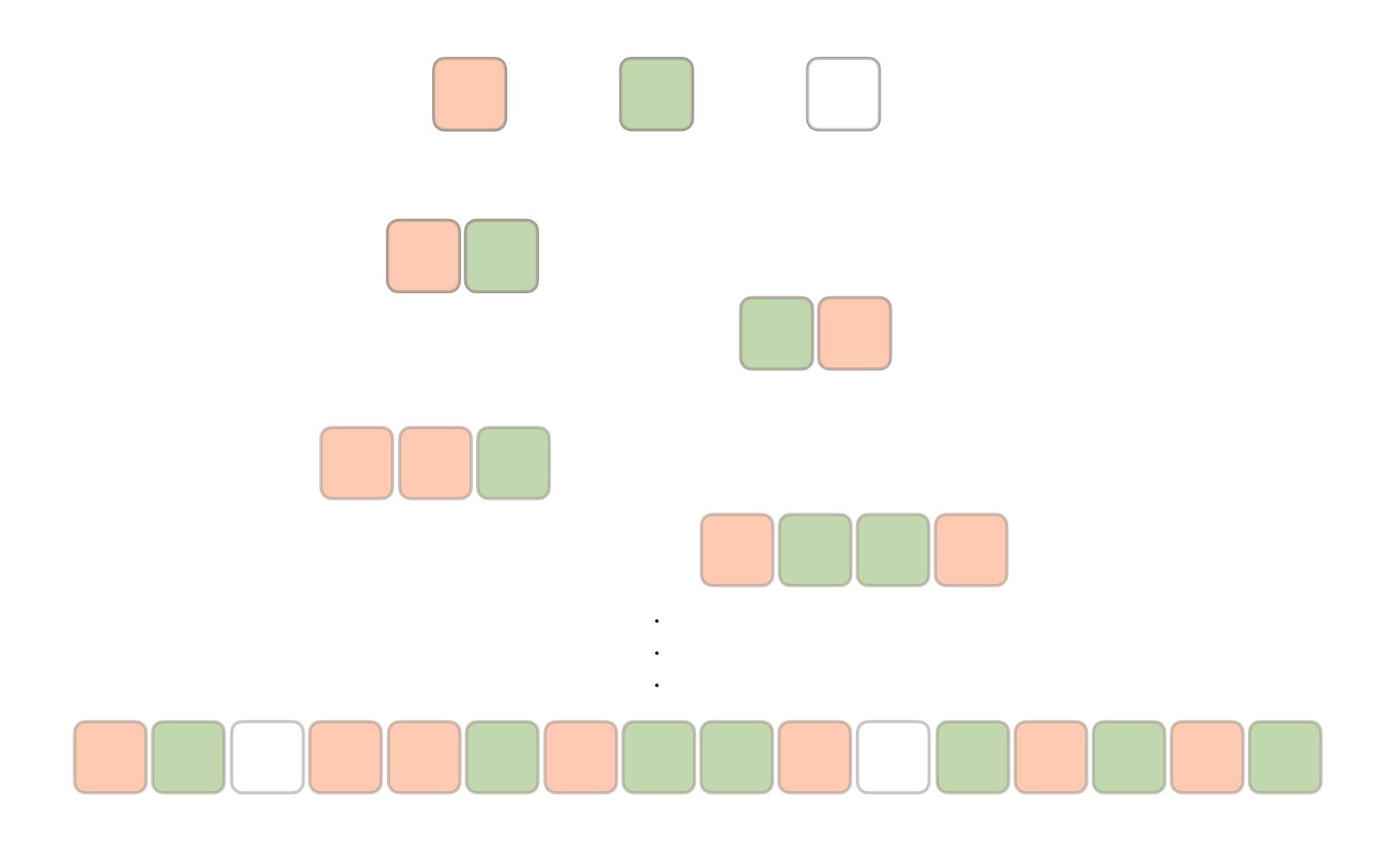
... the thing is ...

It is a bit like ...



Chunking facilitates memory compression (Gobet et al., 2001; Miller, 1956), compositional generalization (Schulz et al., 2017), predictive processing (Koch & Hoffmann, 2000; Müssgens & Ullén, 2015), and communication (Schulz et al., 2020)

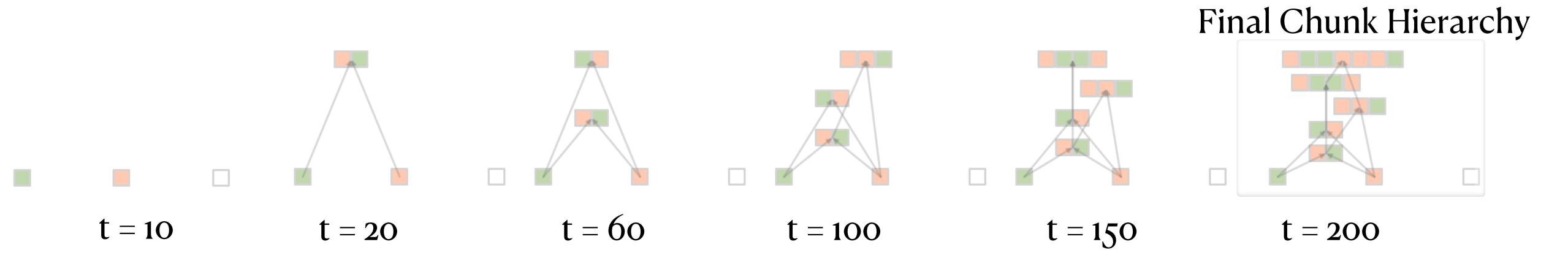
What if perceptual sequences come from a hierarchical structure?



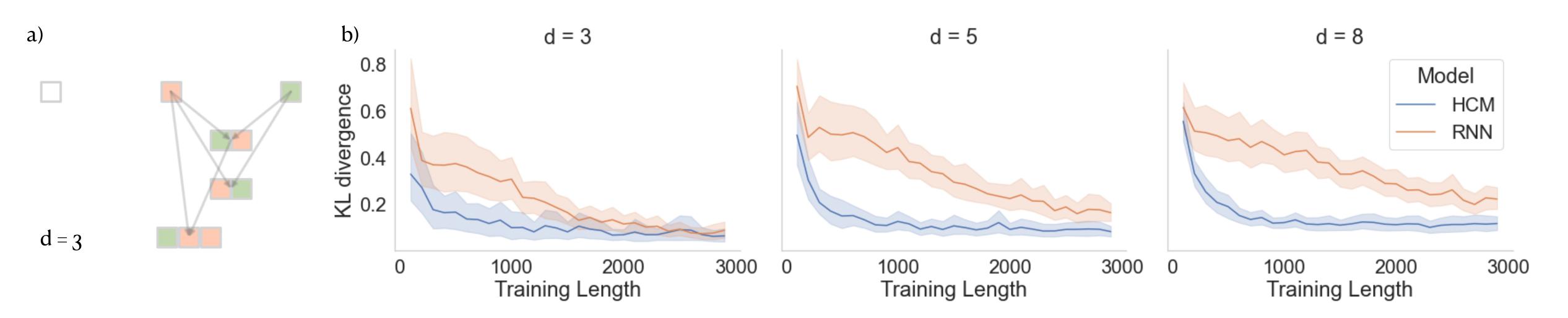
The Hierarchical Chunking Model (HCM)

Sequence: Inspiration: the Gestalt principle of grouping by proximity $c_L \oplus c_R$

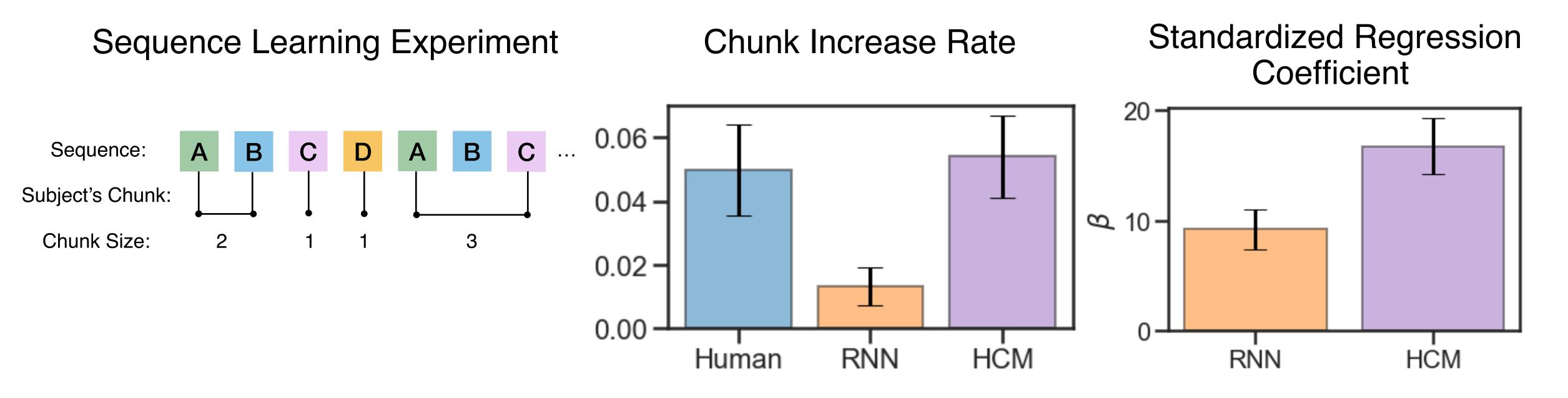




Comparative Advantage to RNN when the sequence contains underlying hierarchy

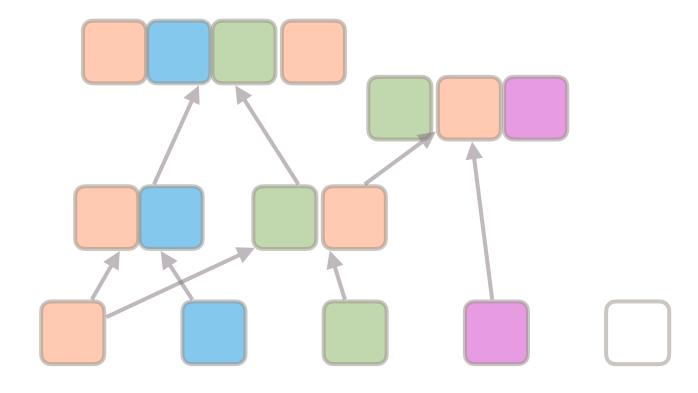


HCM resembles more to human chunk learning

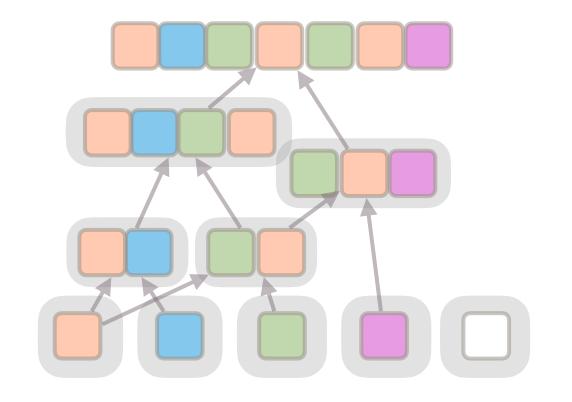


HCM permits transfer between environments

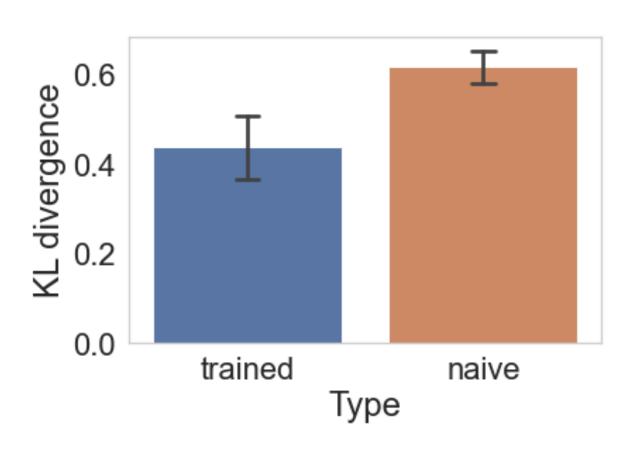
Learned Representation



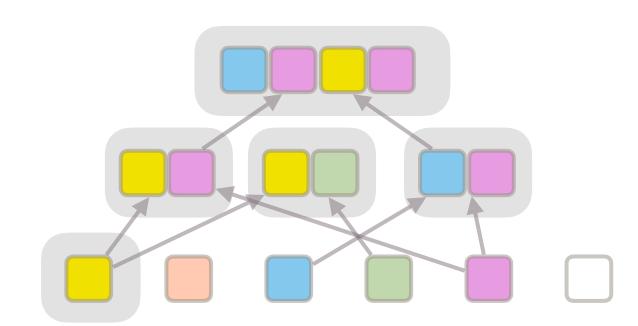
Facilitative Environment



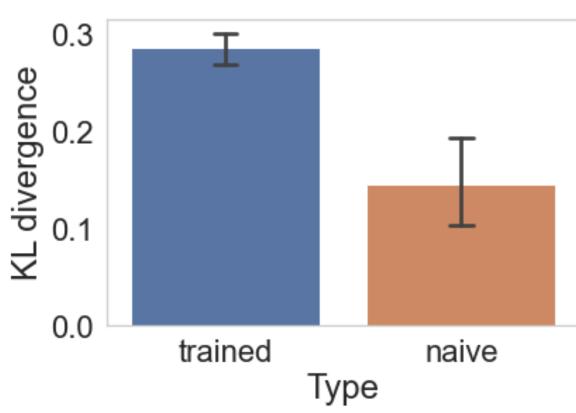
Facilitation



Interfering Environment

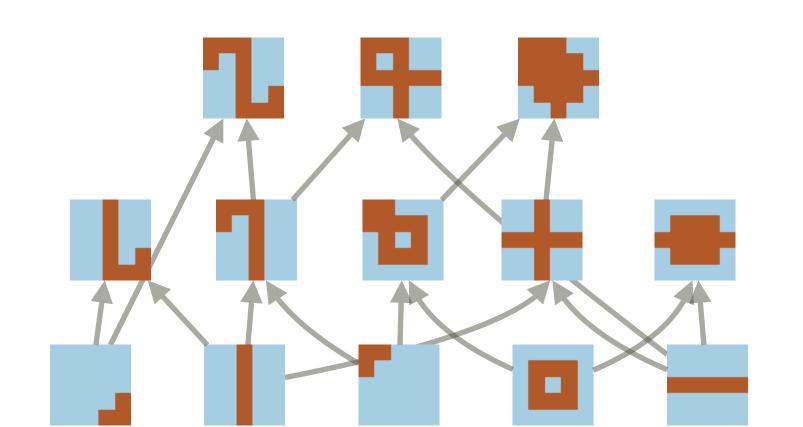


Interference

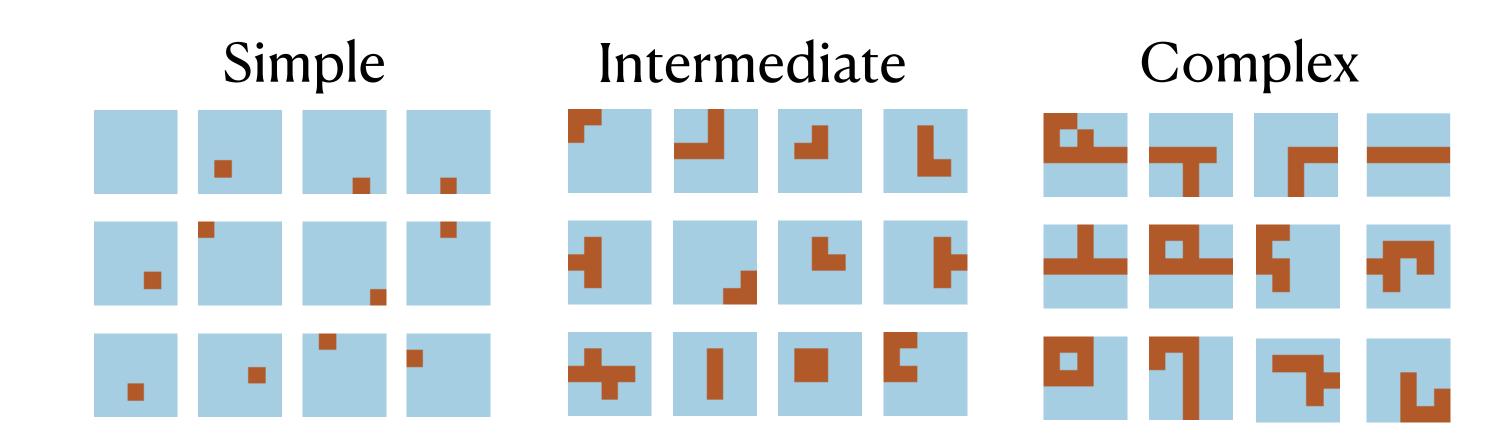


Generalizing Chunk Learning Principles to Visual-Temporal Domain

Visual Hierarchical Model



HCM Learns Compositional Structure in the Visual Domain



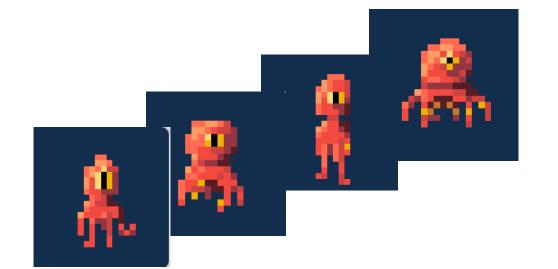
HCM Learns Part-whole Structure that Resembles Object Entities

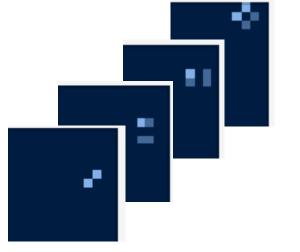






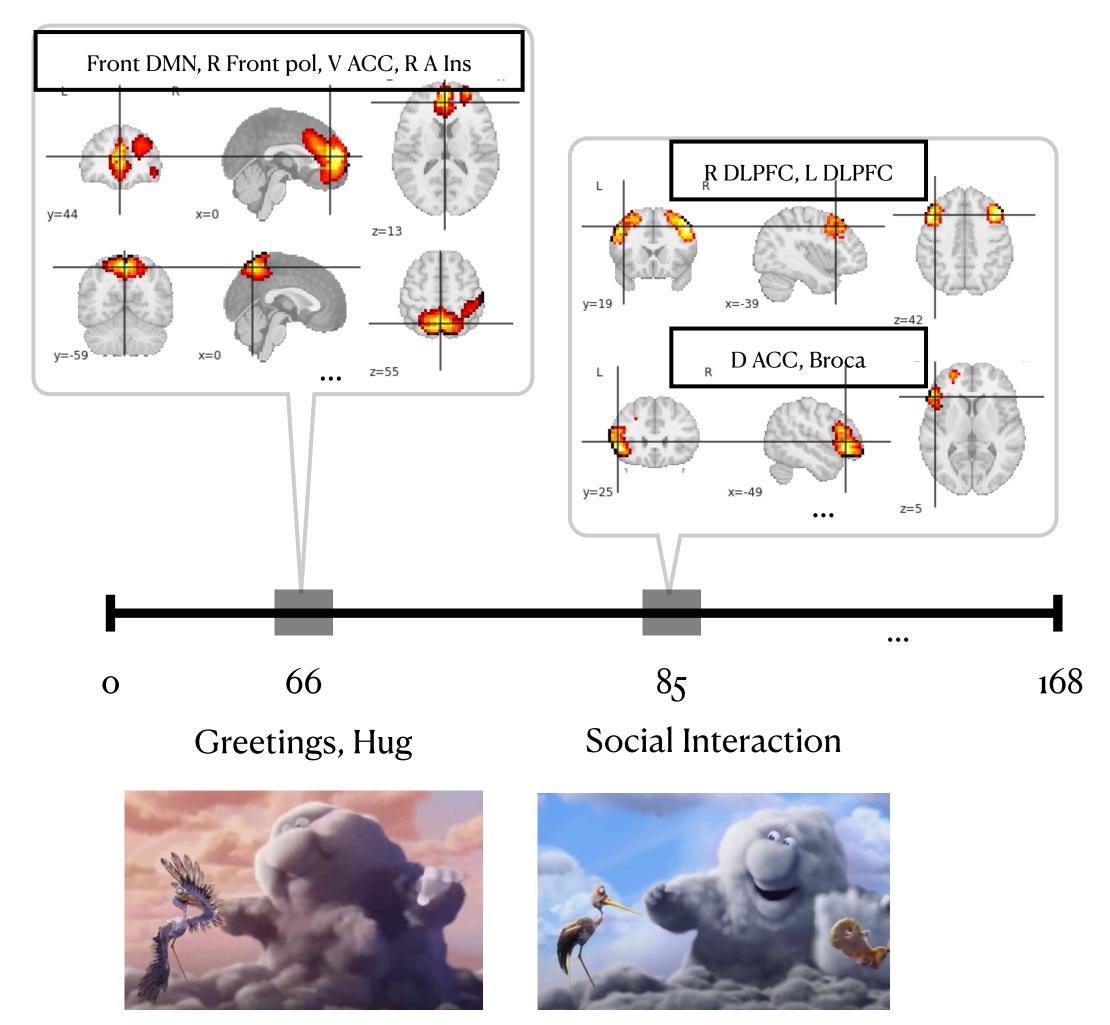




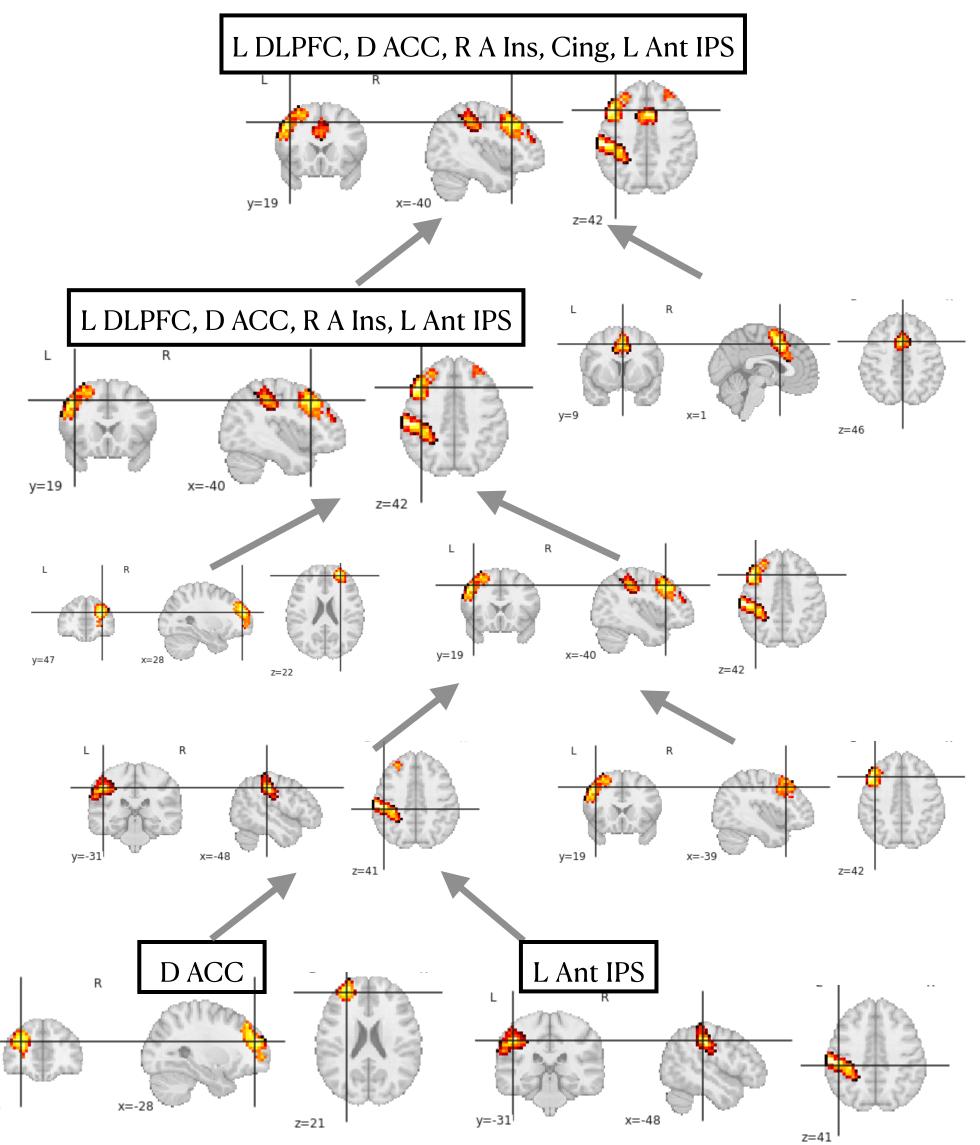


Learning Hierarchies of Brain Activation from Resting-state fMRI data

Discovering patterns of functional activity HCM's Chunks can be Matched with Stimulus Onsets



Learning Hierarchical Activation Patterns





Acknowledgement

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