

Seminarium Filozofia  
Kognitywistyki  
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# Representing as compressing.

Bridging the biological and artificial perceptual systems: a case  
for information-based explanations in cognitive science

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# Overview/recap

1. Motivations
2. Compression in cognition?
3. Trouble with information?
4. From information to representation
5. Structural similarity and infomorphism – tbd.

# Main motivations

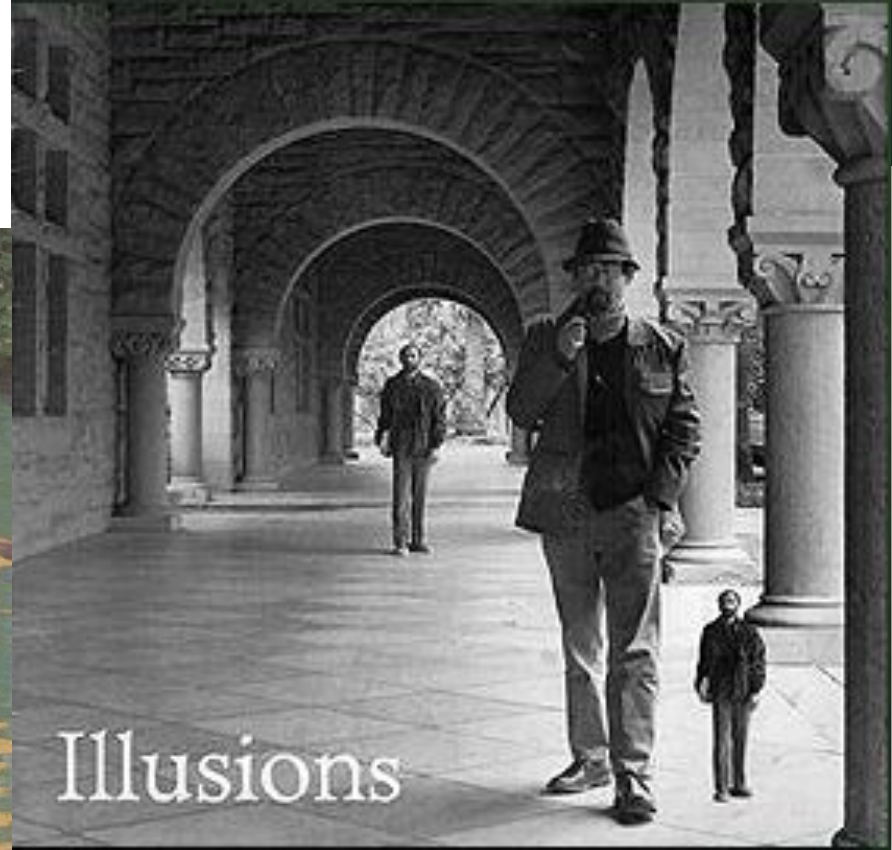
- ❖ Real simple.
- ❖ Analysis of information-talk (and according tools) for explaining perceptual processes
  - It's widely used anyway! Let's delve deeper under the platitude.
- ❖ Assessment of its promise as a currency in which both artificial and biological cognition can be cashed out. Successful comparisons and cognitive cartography?
- ❖ Tackling the question: can information-processing result in representation?

# Compression in cognition?

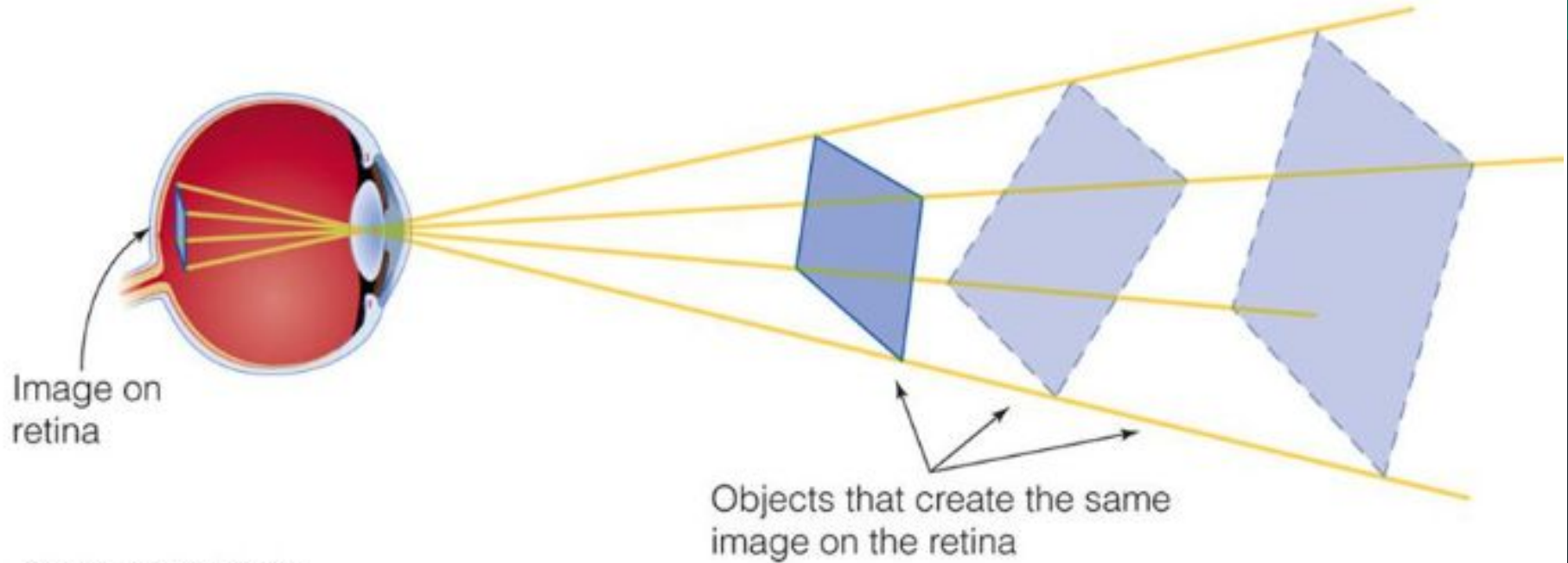
- ❖ Wilkenfeld's (2019) idea of understanding as 'maximizing output-to-scaffolding-ratio'
- ❖ Conceptualizes the idea of generalizing-generative nature of cognition (Buckner 2018)
- ❖ Structure enables compression. Perception thrives on regularities (-> Natural Scene Statistics; Hyvarinen et al. 2009)
  - **Constructivist** (Helmholtzian; Rock 1983) vs. **ecological** (Gibsonian; Orlandi 2014) perception
- ❖ **Regularity=redundancy=predictability**
- ❖ Postulates of compression in semantic pointers theory (Eliasmith 2011, Blouw et al. 2015)

# Where does the question come from?

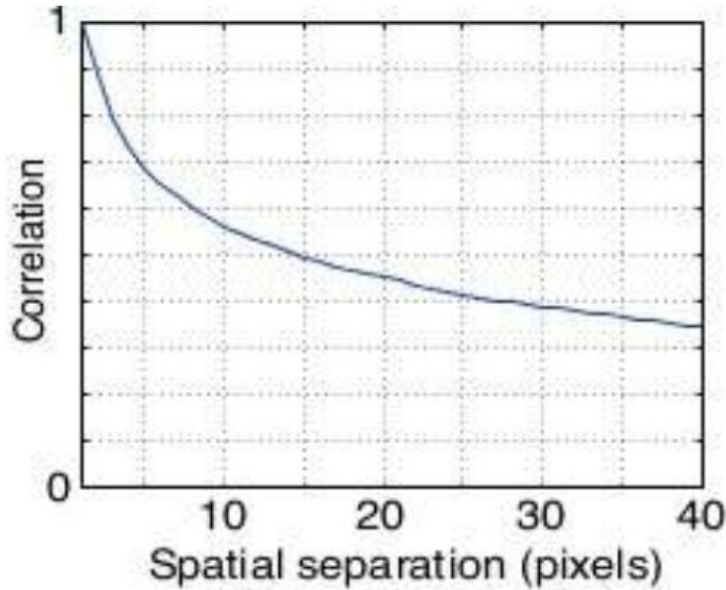
- Ambiguity (underdetermination) of stimuli - e.g. size-constancy
- Assumptions? Inference? Knowledge?  
The temptation of constructivism



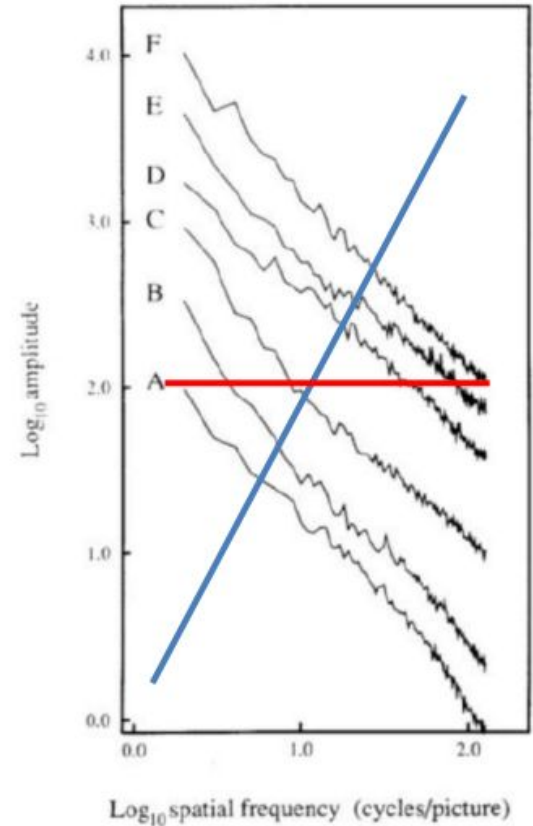
# The inverse problem of perception



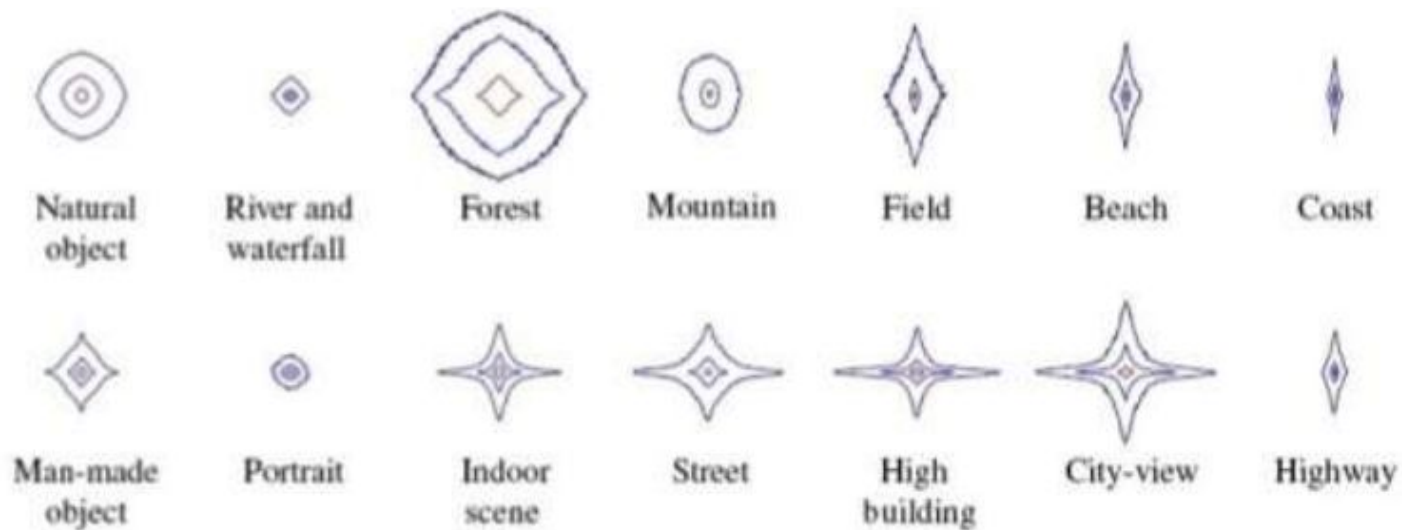
# Auto-correlation in natural images



Power spectrum of  
natural images  
(Field, 1987)



# Scene categorization according to Natural Scene Statistics

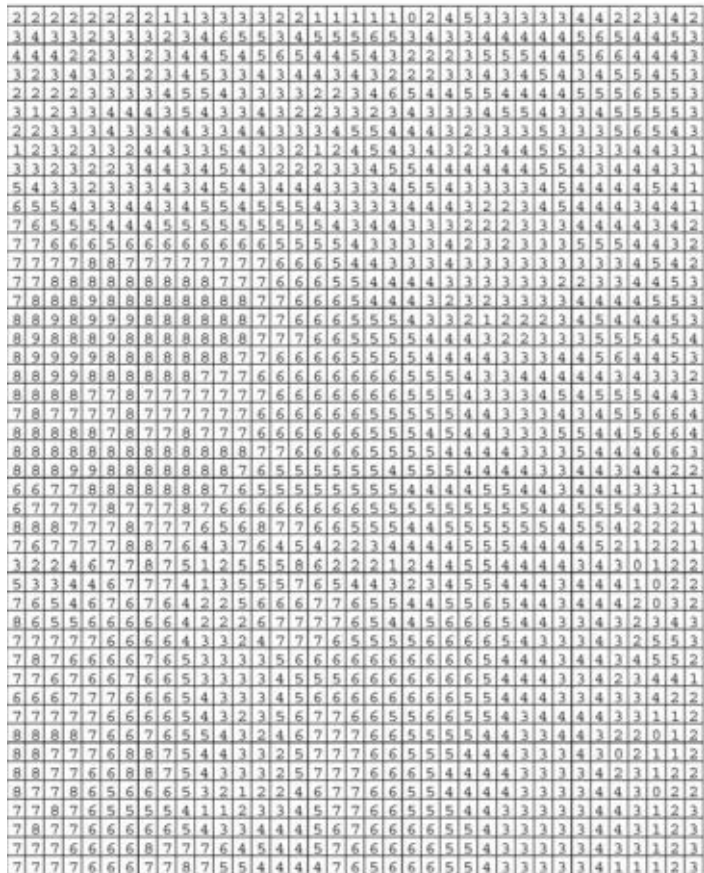


[Torralba and Oliva, 2003]

# Putative problems with 'information'

- ❖ The notion is a *free-for-all*
- ❖ Alternative frameworks: classical (Shannonian; 1948), algorithmic (Chaitin 1977), semantic, teleosemantic (Milikan 2005), transmission (Bergstrom & Rosvall 2008), thermodynamic (Colgate & Ziock 2010; -> **equivocation?**)
- ❖ What about neural information?
  - *Hydraulic* metaphor (information-flow)
  - Receiver-relativity (Rathkopf 2017, 2018)
  - What's a vehicle? (Timme & Lapish 2018)
- ❖ Format matters (Hyvarinen et al. 2009)
- ❖ Hard Problem of Content (Myin & Hutto 2017 -> Isaac 2019, Martinez, in press)
- ❖ Information-in-the-system vs information-about-the-system (Roskies, in press)
- ❖ Setting up the coding game (Skyrms 2010:35)

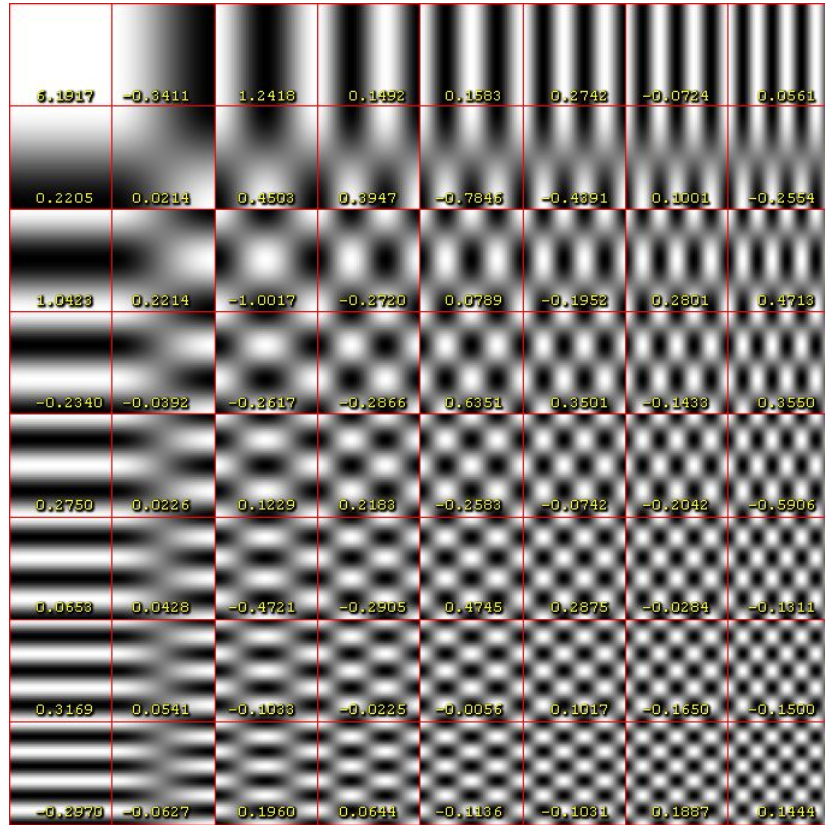
# The invisible president - vehicle's important



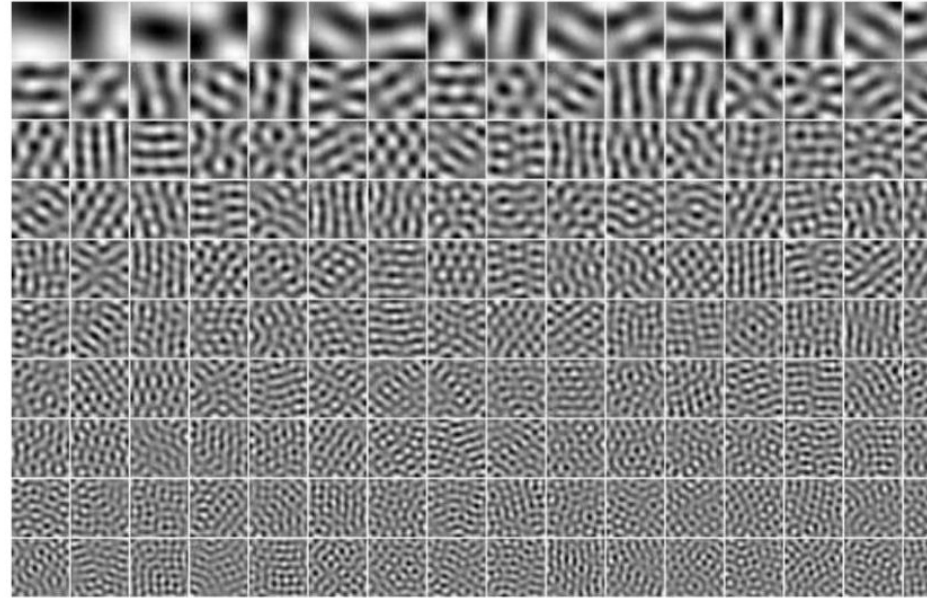
(Hyvaerinen et al.  
2009:4-5)

# Neuropsychological evidence

- ❖ Barlow's maxim of efficient coding (minimize the number of spikes to transmit signal)
  - Task-dependent
  - Level-dependent
  - Constraints
- ❖ Whitening and sparse coding
  - Artificial and biological (Field 1987, Ng 2015)
- ❖ Inter-frame compression? (Nortmann et al. 2013)



Discrete Cosine Transform (JPEG)



PCA Analysis Effect (Field, Olshausen 1995)

Inter-frame  
compression  
scheme?  
(Nortmann et  
al. 2013)

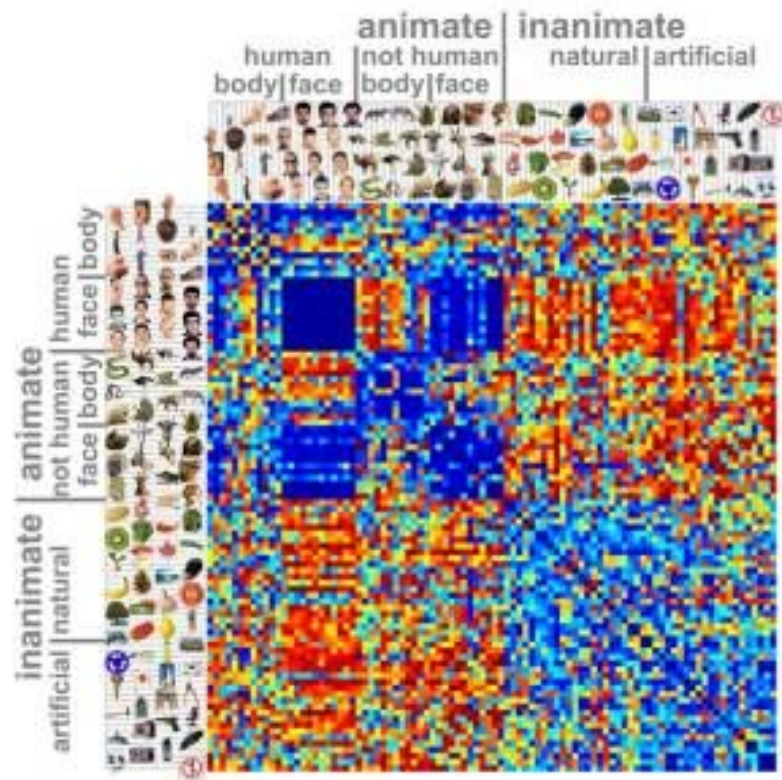


# Representations

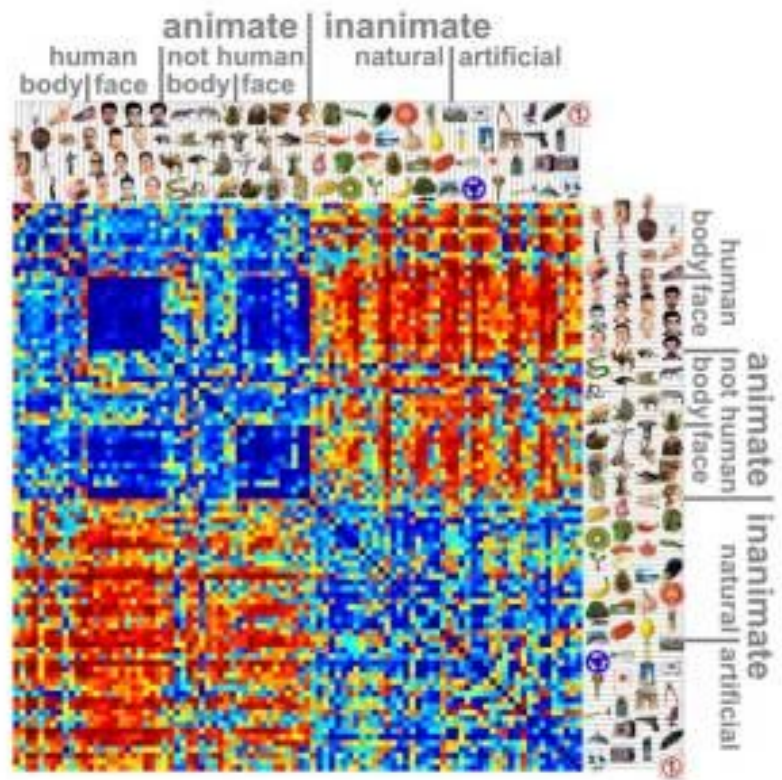
- ❖ Stand-in relations
- ❖ Semantic content (indicative, imperative)
- ❖ Functional role, realized in the virtue of that (Ramsey 2007)
- ❖ Similarity to what's represented (Cummins 2015)
- ❖ No representation without misrepresentation (Dennett 1988)
- ❖ Information is cheap, representation is to be cherished (Orlandi 2014)
- ❖ Absence of stimuli, decouplability, portability? (Clark 2013)
- ❖ Taxonomy? (s-representations, p-representations, r-...and more?)
- ❖ Causally relevant

# New tools

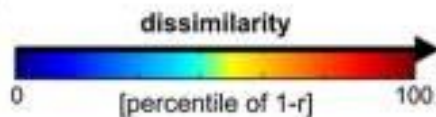
- ❖ Representation similarity analysis (Kriegeskorte et al. 2008, 2015; & Kievit 2013)
- ❖ Observable and manipulable neural representations (Piccinini & Thomson 2018)
- ❖ Deep learning -> deep image synthesis (Bashivan et al. 2019) , building blocks of interpretability (Olah et al. 2018); 'compressed representations'
- ❖ Reciprocal predictive power (Haxby et al 2001; Tong et al. 2005)
- ❖ Incepting percepts (Schalk et al. 2017)



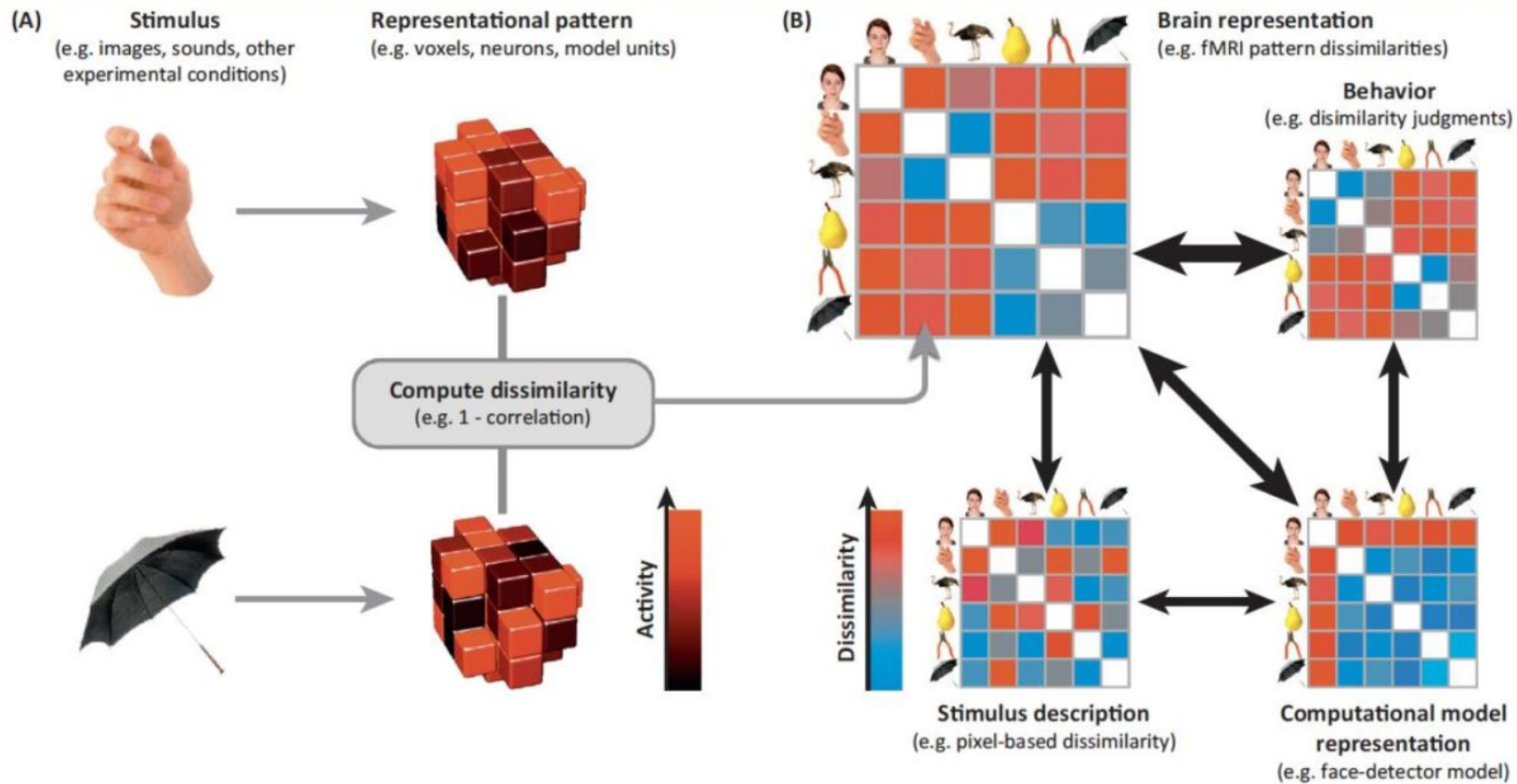
monkey IT



human IT



Kriegeskorte  
et al. 2008



*TRENDS in Cognitive Sciences*

(Kriegeskorte & Kievit 2013)

# Incepting percepts

<https://www.youtube.com/watch?v=x7pWlnZJHVE>

Schalk et al. 2017: Facephenes and rainbows: Causal evidence for functional and anatomical specificity of face and color processing in the human brain, PNAS November 14, 2017 114 (46) 12285-12290;  
<https://doi.org/10.1073/pnas.1713447114>

# Success in similarity(?)

- ❖ Taking advantage of the structure of the informational make-up of the environment (encoding biases)
- ❖ Escaping the curse of dimensionality through 'computational swindle' (Poldrack, in press)
- ❖ The contravariance principle - only very few information-processors can succeed (Cao & Yamins, in press)
- ❖ Learning as arriving at more adequate, compressed representations (-> Schmidhuber's lab)
- ❖ Cue structural similarity and infomorphisms? (Barwise & Seligman 1997; Gładziejewski & Miłkowski 2017)