





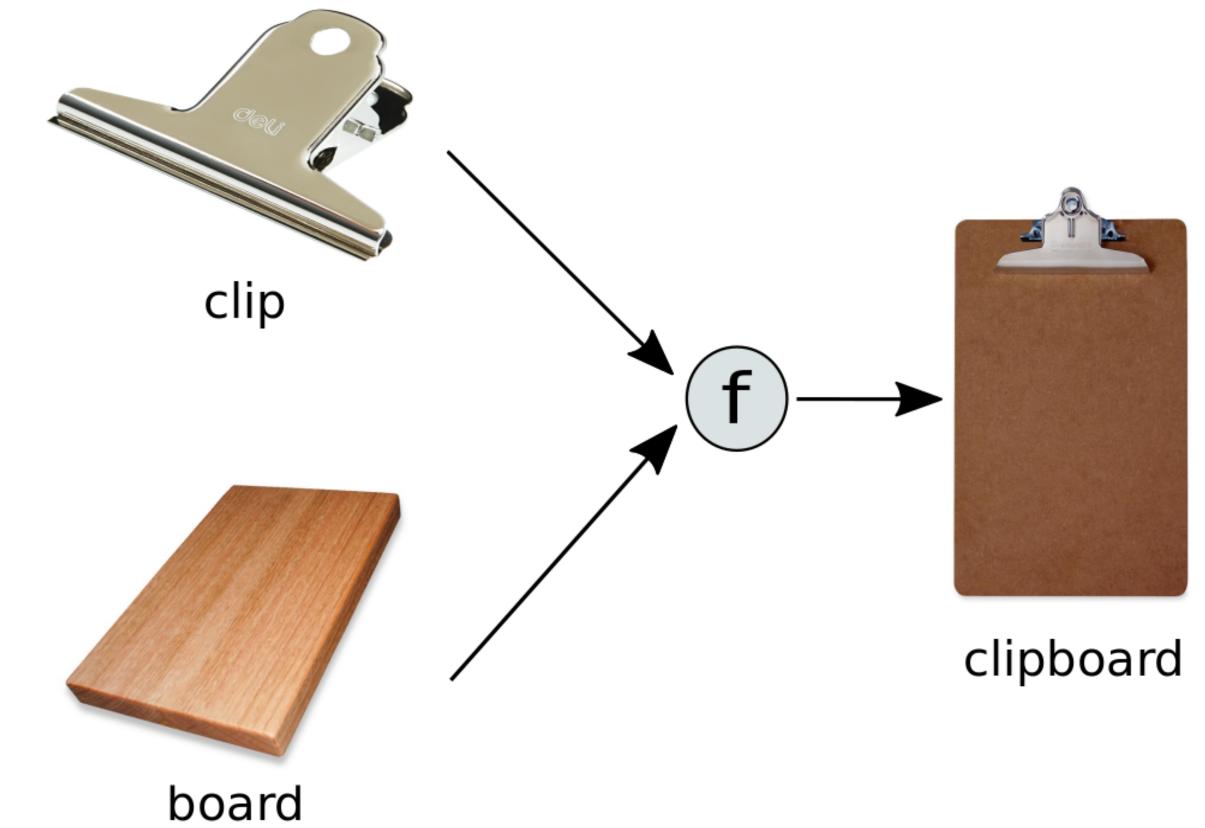
Building a bagpipe with a bag and a pipe: **Exploring Conceptual Combination in Vision**

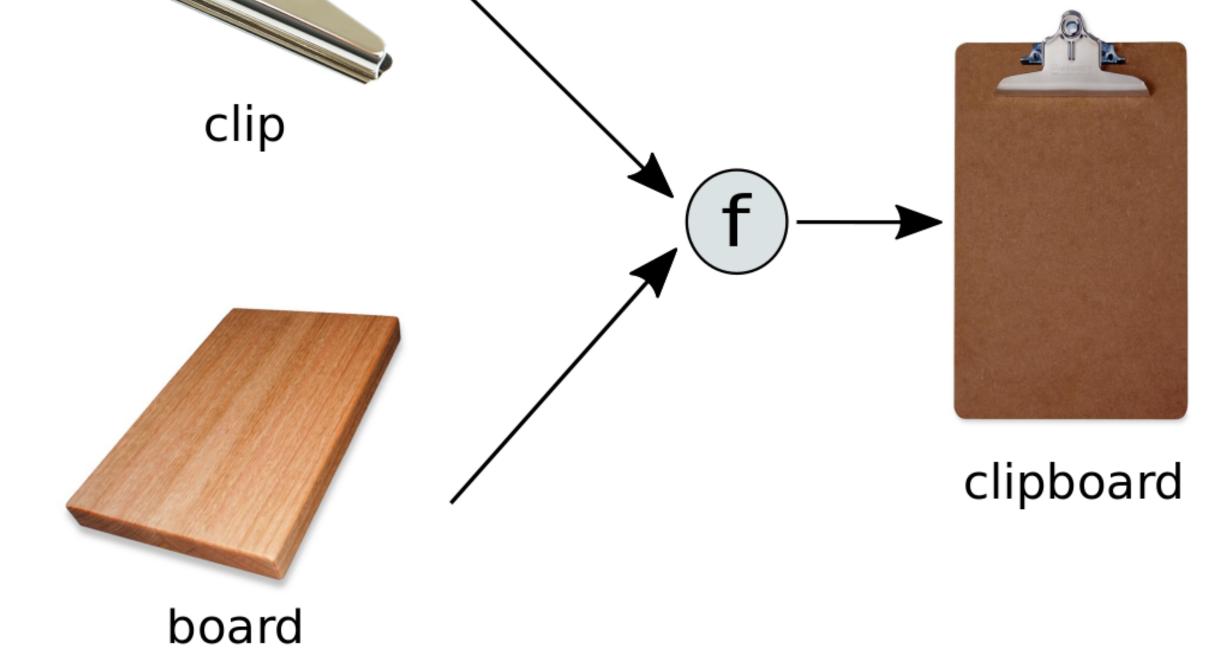
Sandro Pezzelle, Ravi Shekhar, Raffaella Bernardi {firstname.lastname}@unitn.it

Motivation

Conceptual combination is the cognitive process by which two existing concepts are combined to form new complex concepts [1]

In language, this mechanism can be observed in the formation and





lexicalization of compound words like *boathouse*, *swordfish*, etc.

Composition of concepts/words is something more than a simple addition, but **additive models** are effective in language (DSMs) [2]

Research question

Can the visual representation of a complex concept (*clipboard*) be obtained **by summing up** its parts (*clip*, *board*) as in language?

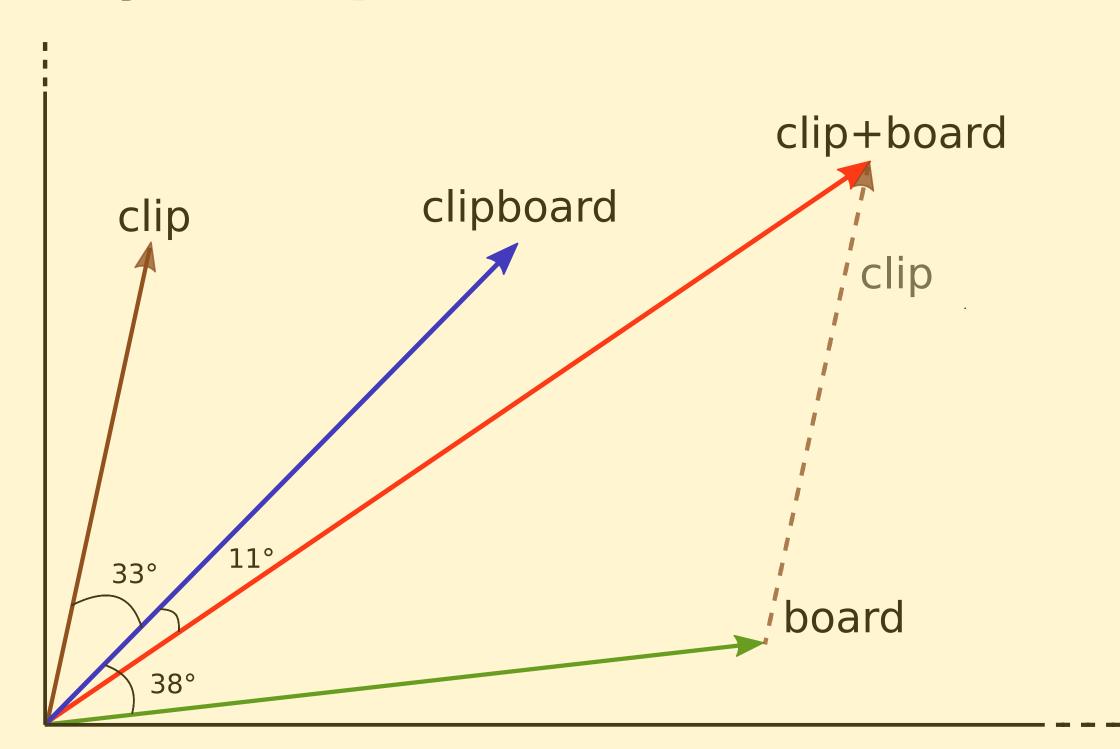
We expect this procedure to work in some cases (parts still visible), but fails where more abstract operations are needed

Dataset

List of noun-noun compounds annotated for **imageability** from [3]

- 1. Filtering based on imageability > 5 (visually well-defined items)
- 2. Only genuine noun-noun combinations were retained
- 3. Selection driven by average quality of top-25 Google images for both the compound (*bagpipe*) and its constituents (*bag*, *pipe*) resulting list including 115 items

Toy example



Dataset construction

- compositional group: **38** manually-selected items involving either superimposition (*air+plane*) or concatenation (*bag+pipe*) • control group: **12** randomly-selected from the 115-item list • full group: compositional + control group (50 items)
- In total: 50 nn-compound images + 79 noun images (129)

Model

We test a **simple additive model** in both Vision and Language:

bagpipe = bag + pipe

Visual features

For each image: 4096-dimension vector extracted using ConvNets (VGG-19 pretrained on ImageNet, fc6 layer) [4]

Linguistic features

• Cosine similarity observed-composed = $cos(11^\circ) = 0.98$ • Complute $\cos(11^\circ) - \cos(33^\circ) = 0.98 - 0.84 = 0.14 > 0$

Results

Vision

1. Composition works in **76.31%** cases in the compositional group vs **16.66%** cases in the control group (*Complnfo* > 0) 2. Both *similarity* and *Rec@k* are higher in compositional group

Language

1. Composition works in **76.31%** and **58.33%** cases, respectively 2. Both *similarity* and *Rec@k* are higher in compositional group

For each word: word2vec 400-dimension vector trained on 3-billion tokens corpus (ukWac + Wikipedia + BNC)

Evaluation

To evaluate the compositional model, we use three measures:

- 1. **Cosine similarity** between observed (*clipboard*) and composed vector (*clip+board*)
- 2. **Complute**: difference between composed-observed similarity and observed-closest noun similarity (e.g., *clipboard* and *clip*) Thus, composition works with *Complate* > 0
- 3. **Rec@k**: retrieving the observed vector in the semantic space using the composed one (with k=1 and k=5)

Dataset	Avg.Similarity		$\ \%(CompInfo > 0) \ $		Rec@1		Rec@5	
	Vision	Lang	Vision	Lang	Vision	Lang	Vision	Lang
Full	0.6283	0.407	62%	72%	0.34	0.52	0.76	0.88
Compositional	0.6476	0.429	76.31%	76.31%	0.3947	0.57889	0.8158	0.9211
Control	0.5671	0.3377	16.66%	58.33%	0.1667	0.3333	0.5833	0.75

References

[1] E. J. Wisniewski. 1996. Construal and similarity in conceptual combination. Journal of Memory and Language, 35(3):434-453 [2] D. Paperno and M. Baroni. To appear. When the whole is less than the sum of the parts...

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[3] B. J. Juhasz, Y-H Lai, and M. L. Woodcock. 2014. A database of 629 English compound words... Behavior research methods, pages 1-16

[4] A. Vedaldi and K. Lenc. 2015. MatConvNet - Convolutional Neural Networks for Matlab. Proceeding of the ACM Int. Conf. on Multimedia