



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

S. Pezzelle, R. Shekhar, R. Bernardi

{firstname.lastname@unitn.it} CIMeC, DISI (University of Trento)

### **Motivation**

#### Language

Noun-noun compounds (*clipboard, bagpipe,* etc.) are (to some extent) **compositional**: using even simple additive cDSMs we obtain composed vectors that approximate the ones extracted from text corpora

#### Vision

Can we do the same in vision? That is, does the visual representation of a nn-compound rely on the **combination of its parts**? If so, to what extent (for which cases) does it work?

#### **Visual Composition**





clipboard

## Hypothesis

We should be able to obtain the visual representation of a *clipboard* by combining *clip* and *board* with a **compositional function**, similarly to language

Our hypothesis is that a **simple additive model** should work for some cases, i.e. when the parts are still visible in the representation of the whole:

- superimposed cases (object in a background, e.g. *airplane*)
- concatenation of parts (whole resulting from the parts, e.g. *clipboard*)

#### Experiment

- 1. We manually built a dataset of images of visually-compositional nn-compounds (*compositional group*)
- 2. We randomly chose a *control group*
- 3. We extracted visual features for NN, N1, N2 with ConvNets (VGG-19 model)
- 4. We applied additive model to N1,N2 to obtain composed representations N+N
- 5. We evaluated over:

a) Cosine similarity between observed and composed NN vectors (Sim)

b) Retrieval (**Rec@k**) with k=1 and 5

c) **Complnfo**: Sim - similarity observed NN/closest N (working when >0)

## Vision vs Language

For each N, NN in the dataset, we built CBOW linguistic vectors with **word2vec** 

We experimented with the same additive model and the same evaluation measures

#### Results

Vision:

Composition works in *compositional group* (76.6%) but not in *control group* 

#### Language:

Same performance in *compositional group* but higher in *control group* (58.3%)

Dataset	Avg.Similarity		$\left  \%(CompInfo > 0) \right $		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

#### Conclusions

Simple additive model in Vision **works for manually selected cases** (superimposed or concatenated items) but not for combinations where subtler, more abstract interactions are involved (eg. *corkscrew*)

Need for large, annotated datasets

Need for new, more complex compositional models (perhaps combining the 2 modalities) designed for solving the task

#### Thank you for your attention!