A Crow's Beak is not Yellow – Investigations on Cognitively Salient Concept Properties

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Extraction Method

Conclusion

## Describing a Concept...



- ✓ has 2 legs
- $\checkmark$  has feathers
- 🗸 is black

versus

- ✤ has a heart
- ↓ can see

## **Topic Outline**

## Feature norms (e.g. McRae et al.'s)

Concept representations - used in simulations of cognitive tasks

#### Efforts on extracting such descriptions

... using text corpora

(getting norms without experiments;

better models based on more data)

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#### $\sim$ Study aims

- Empirical basis: behavioural experiments (parallel: DE + IT)
- Method for harvesting cognitively salient properties

# Experiments

## Production

"Describe the concept."

- Annotation of answers: property types
- Observation: preferred use of types depending on concept class

## Perception

"Is the word plausible for describing the concept?"

- Recording of RTs and errors
- Findings: (inconsistent) differences between class-type pairs

## **Distributions of Produced Property Types**







#### Cognitively Salient Concept Properties

# Experiments

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## **Extraction Targets**

Composite (adj- modifier + noun) part properties of concepts ,e.g. rabbit : long ears

#### **Corpus excerpt**

```
Die mittelgroßen Affen leben in Gruppen von etwa 15 Tieren auf
Bäumen im Regenwald.
[...]
Die Kipunji sind verwandt mit anderen Mangaben, doch sie weisen
einige Besonderheiten auf. Sie haben braunes oder hellbraunes
Fell und geben Töne von sich [...]
```

# Approach

#### • Aim:

Extract cognitively salient modifiers for given concept-part pairs

#### • Idea:

Create ranked list based on corpus frequencies and select 5 highest ranked modifiers

#### • Resource:

WaCky web corpus

• Evaluation against feature production norms

# **Best Rank List Methods**

- **1.** Modifier–Part pair frequencies ("contextless")[Adj]?[Adj]?[Adj]?[Noun]
- 2. Frequencies of modifier-part pairs in concept context[part]? (20 sent.) [concept] (20 sent.) [part]?
- **3.** Productwise combination of frequencies

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# Example:

## Concept "Bear" With Part "Fur"

| rank | contextless |          | in concept context |          |
|------|-------------|----------|--------------------|----------|
|      | freq        | modifier | freq               | modifier |
| 1    | 507         | thick    | 16                 | thick    |
| 2    | 209         | dense    | 14                 | white    |
| 3    | 204         | soft     | 11                 | small    |
| 4    | 185         | black    | 11                 | soft     |
| 5    | 175         | long     | 9                  | dense    |

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# **Best Rank List Methods**

- 1. Modifier–Part pair frequencies ("contextless")

   [Adj]?
   [Adj]?
   [Adj]?
   [Noun]
- 2. Frequencies of modifier-part pairs in concept context[part]? (20 sent.) [concept] (20 sent.) [part]?
- 3. Productwise combination of frequencies

Robust performance (precision  $\approx$  14 % at recall  $\approx$  43 %)

- across languages,
- for production and perception,
- and for concepts previously unseen by the algorithm

## Conclusion

Types of concept properties produced by native speakers

- ... have similar distributions across languages
- ... are preferably from a type set depending on broad concept class

Automatic corpus-based extraction (of *part* modifiers)

- ... works best when combining in-context and contextless list
- ... performs similarly well across languages
- ... works comparably well based on both production and perception

#### Thank you.



#### German



#### Italian



German



type category quality function location part Pearson residuals: mammal - 23.3 bird fruit class vegetable 4.0 bodypart 2.0 0.0 clothing -2.0 -4.0 implement vehicle furniture -12.6 p-value = < 2.22e-16 building

Italian

#### Algorithm Performances



recall

