

Constraints on Definite Descriptions with Vague Adjectives: Towards a Computational Model

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Definite descriptions with gradable adjectives

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the *big* triangle

the *cheap* laptop

the *short* basketball player

- Referential expressions used to pick up an entity in a relevant domain

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- Vague expressions are efficiently used in communication

Kees van Deemter (2006) Generating Referring Expressions that involve Gradable Properties. *Computational Linguistics* (32) 2: 195–222.

Overview

- Motivation: natural language generation
- Focus: what contexts licence the positive form?
- Proposal based on relative similarity
- Problems
- Conclusions

Generation of Referring Expressions

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- In Computational Linguistics, the production of referring expressions is studied in the subfield of Natural Language Generation (NLG)
- The goal is to generate a **distinguishing description** of a **target** object that singles it out from the set of **distractors**
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- Usually a description is considered optimal if it is *minimal* (Grice's Maxim of Quantity) – eq. to minimal set cover, i.e. NP-hard

Dale & Reiter (1995) Computational interpretation of the Gricean maxims in the generation of referring expressions. *Cognitive Science*, 19(2): 233–263.

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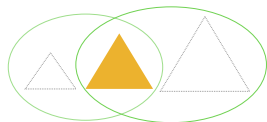
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- **What kind of contexts license the production of a definite description involving a vague adjective in the positive form?**

Context dependence

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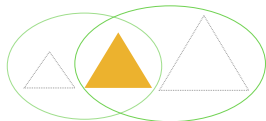
- Well known fact: The positive form can be used to refer to an object that differs from others only in the extent to which it possesses the property expressed by the adjective — does not hold for absolute adjectives, e.g. *“the full glass”*



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the big triangle / the small triangle

- Lots of psycholinguistics research confirming this

Ebeling & Gelman (1994) Children's Use of Context in Interpreting "Big" and "Little." *Child Development*, 65: 1178–1192.

Syrett, Lidz & Kennedy (to appear) Meaning and Context in Children's Understanding of Gradable Adjectives, *Journal of Semantics*.

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- (i) The definite article introduces a presupposition of **unique existence**, where uniqueness is with respect to a contextually determined domain of quantification $\exists x \forall y (P(y) \leftrightarrow x = y)$
- (ii) What counts as **A** depends on a contextually determined **comparison class** C , where C is a partially ordered set of elements: $\langle D, \geq_{Dim(\mathbf{A})} \rangle$

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Constraints on C (Klein 1980):

- (C1) \mathbf{A} partitions C into two sets \mathbf{A}^+ and \mathbf{A}^- corresponding to the positive and the negative extensions of \mathbf{A} , respectively;
- (C2) \mathbf{A}^+ and \mathbf{A}^- must be non-empty and disjoint, hence $|C| \geq 2$
- (C3) $\forall u, u' \in C$ if $u \in \mathbf{A}^+$ and $u' > u$ then $u' \in \mathbf{A}^+$ (and acc. for \mathbf{A}^-)

Klein (1980) A semantics for positive and comparative adjectives. *Linguistics & Philosophy* 4: 1–45

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- The interpretation of a description like *the big triangle* can be seen as a byproduct of (i) and (ii) above: given a comparison class C , the boundary between \mathbf{A}^+ and \mathbf{A}^- can flexibly shift as long as (i) and (ii) are satisfied

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- But this predicts that in a definite description the positive form is equivalent to the superlative. . .

the big triangle = the biggest triangle

Kyburg & Marreau (2000) Fitting Words: Vague Language in Context. *Linguistics & Philosophy*, 23: 577–597.

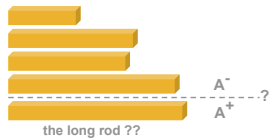
Saliency

Saliency

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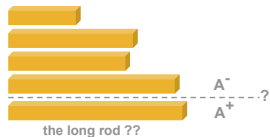
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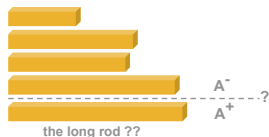
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The boundary between A^+ and A^- can only fall between entities that are *saliently dissimilar*

- Pointed out more or less explicitly by Graff 2000, Kennedy 2007 (“the referent needs to stand out”), Rayo 2008, . . .
- The gap that singles out the intended referent needs to be salient with respect to the other gaps: The positive form presupposes the existence of a salient boundary within C

Graff (2000) Shifting Sands: An interest-relative theory of vagueness, *Philosophical Topics*, 20: 45–81.
Kennedy (2007) Vagueness and Grammar, *Linguistics & Philosophy*, 30(1): 1–45.
Rayo (2008) Vague Representation, *Mind*, 117(466): 329–373.

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Let \sim be a reflexive, symmetric, and non-transitive relation holding between entities $u \in C$ that satisfies the following conditions:

- $\forall u, u', u'' \in C$ such that $u > u' > u''$, if $u \sim u''$ then $u' \sim u$ and $u' \sim u''$
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- This defines a set of clusters of elements that belong to either \mathbf{A}^+ or \mathbf{A}^- and a set of *gaps* that can serve as potential denotational boundaries.

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- Related to the semi-order approach of van Rooij: constraints on what constitutes an pragmatically appropriate context

van Rooij (2008) Revealed Preference and Satisficing Behaviour, in *Proc. of the 8th Conf. on Logic and the Foundations of Game and Decision Theory*.

Computing salience using clustering

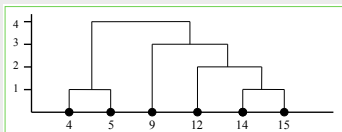
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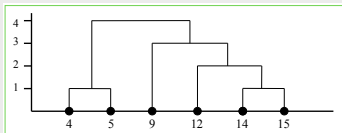
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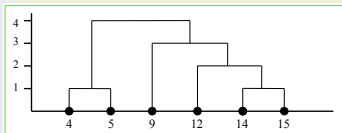
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- We can define the degree of salience δ of a gap between u and u' in C as $\delta(u, u' \in C) = m + 1/n$, where m is the highest level in the dendrogram where u and u' belong to different clusters
- Since δ is defined as a ratio, it accounts for the *relative salience* of the gaps given the distribution of elements in C

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- We can express restrictions on \sim in terms of heuristic constraints that need to be determined empirically

$$u \sim u' \leftrightarrow \delta(u, u' \in C) < k$$

- The value of k may depend on several factors: the nature of the elements under consideration, the task at hand, the goals of the agents, ...

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- We can compute the salience of the relevant boundary to decide if the positive form is licenced – but whenever it is licenced, should it always be preferred over the superlative?

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- Experiments by van Deemter seem to support this:
 - the positive form is dispreferred with very similar elements
 - but for abstract figures some subjects use the positive form across the board
- A technical “solution”: cluster the combined set $\mathcal{C} = C \cup C'$, where the elements in C' are randomly generated from the relevant frequency distribution

Fernández (2009) Saliency and Feature Variability in Definite Descriptions with Positive-form Vague Adjectives, in *Proc. of PRE-CogSci, the Annual Meeting of the Cognitive Science Society*.

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- Importance of contrast and similarity
- Possible influence of patterns of feature variability
- Importance of experimental work and computational evaluation