



Word Learning in the Wild: What Natural Data Can Tell Us

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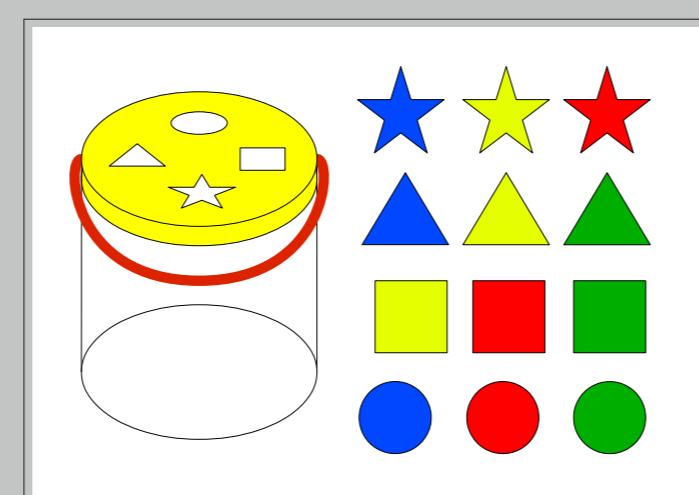
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Word-meaning acquisition

- ▶ **Cross-situational word learning:**
 - ▷ Learner tracks co-occurrence of words and situationally available meanings
 - ▷ Depends on cross-situational availability of meaning
- ▶ **Problem:**
 - ▷ Computational models of word-meaning acquisition assume a high degree of availability of the cross-situational information for all words
 - ▶ Including relational meanings (harder to learn from the situational context than object-labels [2,3])
 - ▷ **How valid is this assumption?**
- ▶ **Our approach:**
 - ▷ Annotate video data of caregiver-child interaction
 - ▷ Investigate situational availability assumption by using computational word learning model

Word learning in the wild: video-taped interaction

- ▶ 32 pairs of Dutch mothers and daughters (16mo)
 - ▷ playing a game of putting blocks in holes
 - ▷ 152 minutes in total
 - ▷ 7,500 words of child-directed speech
- ▶ For every 3-second interval, annotated features:
 - ▷ active game-related objects & participants (child, block)
 - ▷ properties of the objects (red, round)
 - ▷ the participant's actions on the objects (grab, move)
 - ▷ changes in spatial relations among the objects (in, off)
- ▶ High inter- and intracoder reliability ($\kappa > 0.8$)



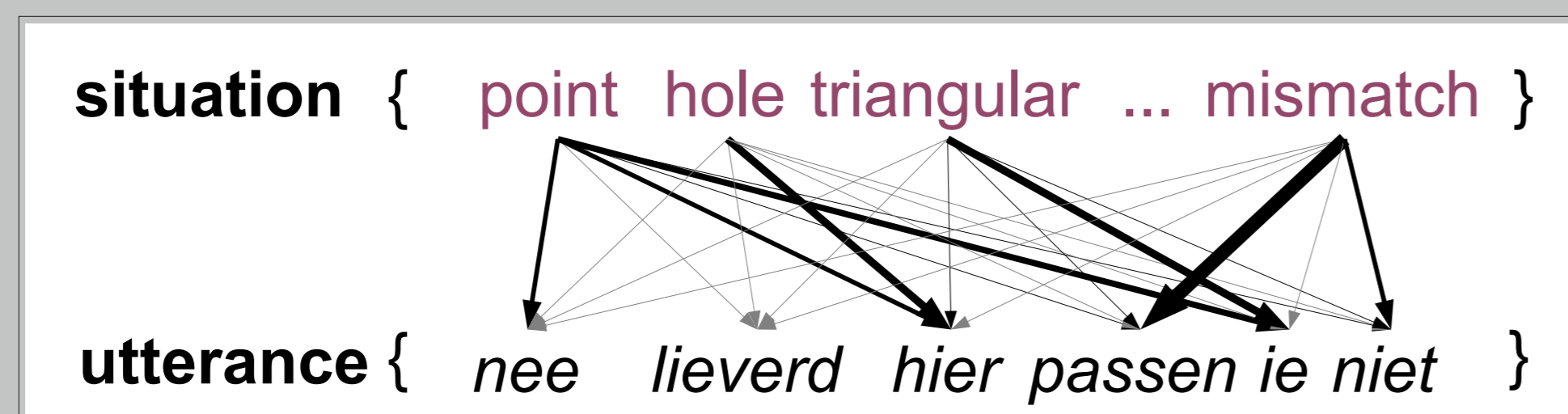
tier	coding/transcription
sit.	<nothing happens>
utt.	een. nou jij een. "One. now you (try) one."
sit.	position(mother, toy, on(toy, floor)), grab(child, b-ye-tr), move(child, b-ye-tr, on(b-ye-tr, floor), near(b-ye-tr, ho-ro)), mismatch(b-ye-tr, ho-ro)
utt.	nee daar. "No, there."
sit.	point(mother, ho-tr, child) position(child, b-ye-tr, near(b-ye-tr, ho-ro)) mismatch(b-ye-tr, ho-ro)
utt.	nee lieverd hier past ie niet. "No sweetie, it won't fit in here."

Table 1: A sample of the dataset. The dash-separated abbreviations denote blocks (b) and holes (ho) and their properties, color (only blocks) ({red,green,blue,yellow}) and shape ({round,star,square,triangular})

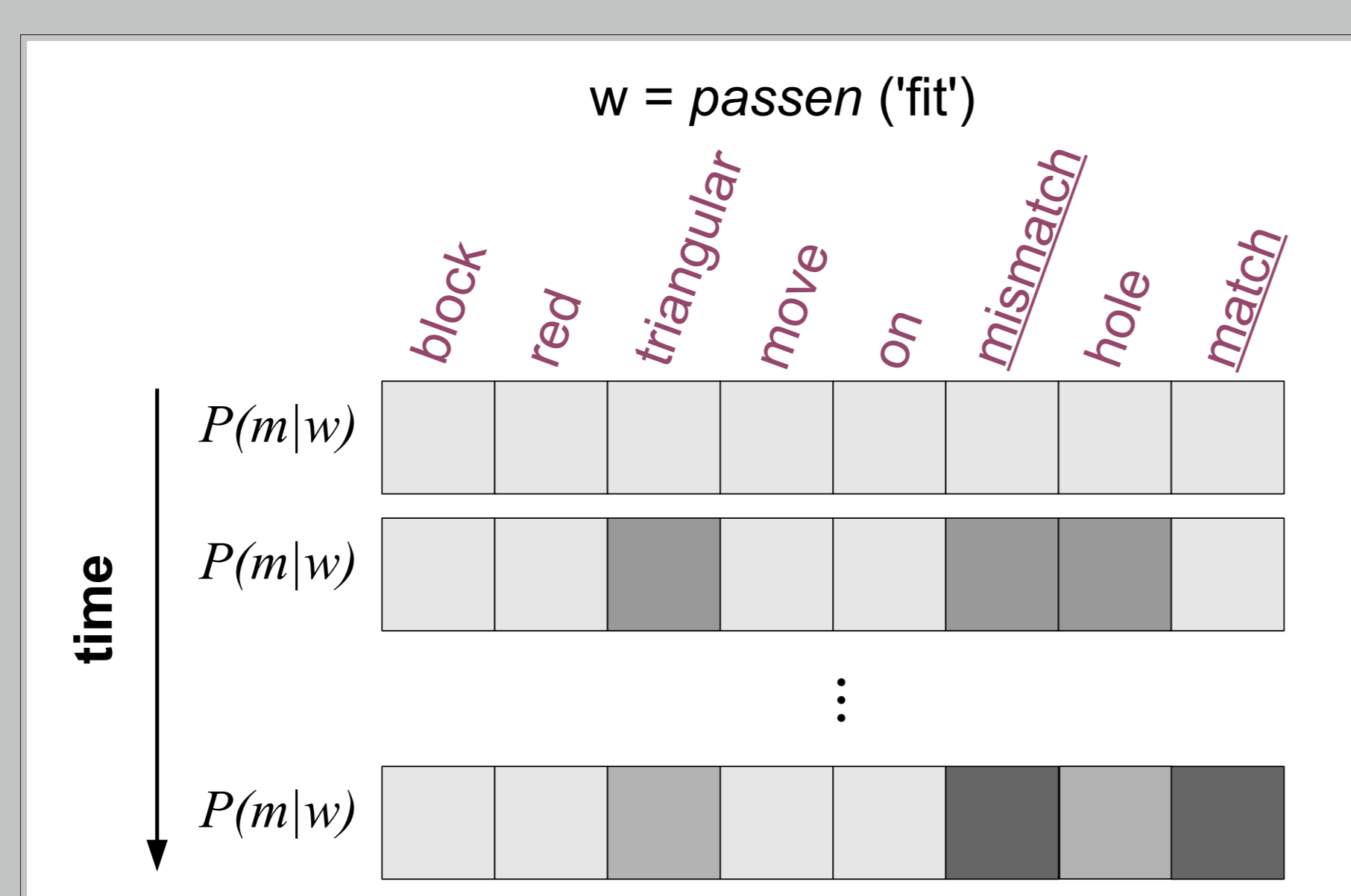
Exploring the availability assumption

- ▶ Using a cross-situational word learning model [1]

Aligning features in situation with words



Using alignments to learn probability distributions of features given words



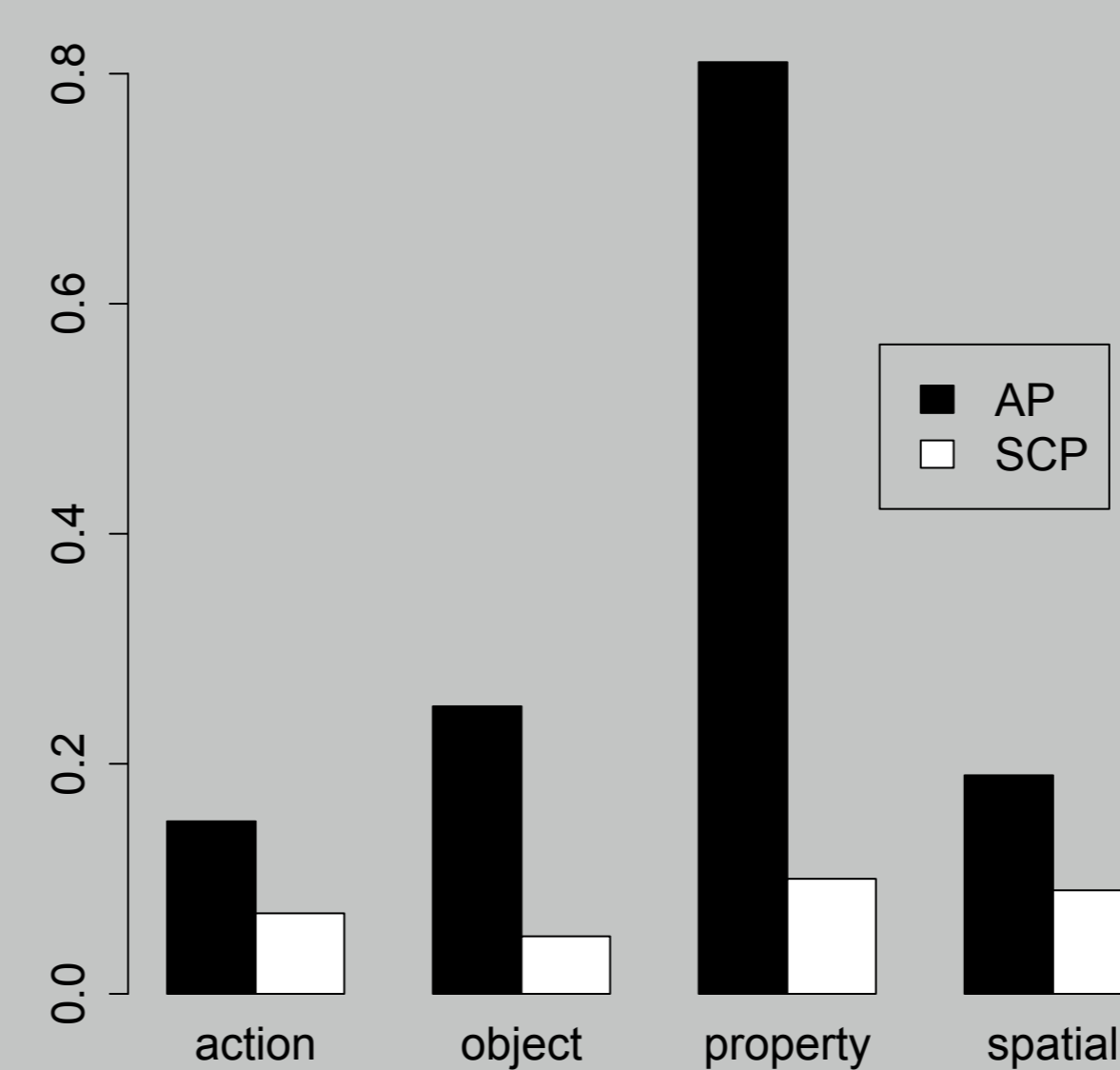
- ▶ Trained on 2500 Utterance-Situation pairs

Evaluation

- ▶ Evaluated learned probability distributions against hand-annotated relevant features:
 - ▷ e.g. *stoppen* - move,in; *blok* - block; *rood* - red; *op* - on
 - ▷ four groups of features: action, object, property, spatial
 - ▷ evaluation metrics, for each word:
 - SCP Summed Conditional Probability (probability mass of relevant features)
 - AP Average Precision (quality of the ranking of relevant features)

Experiment 1

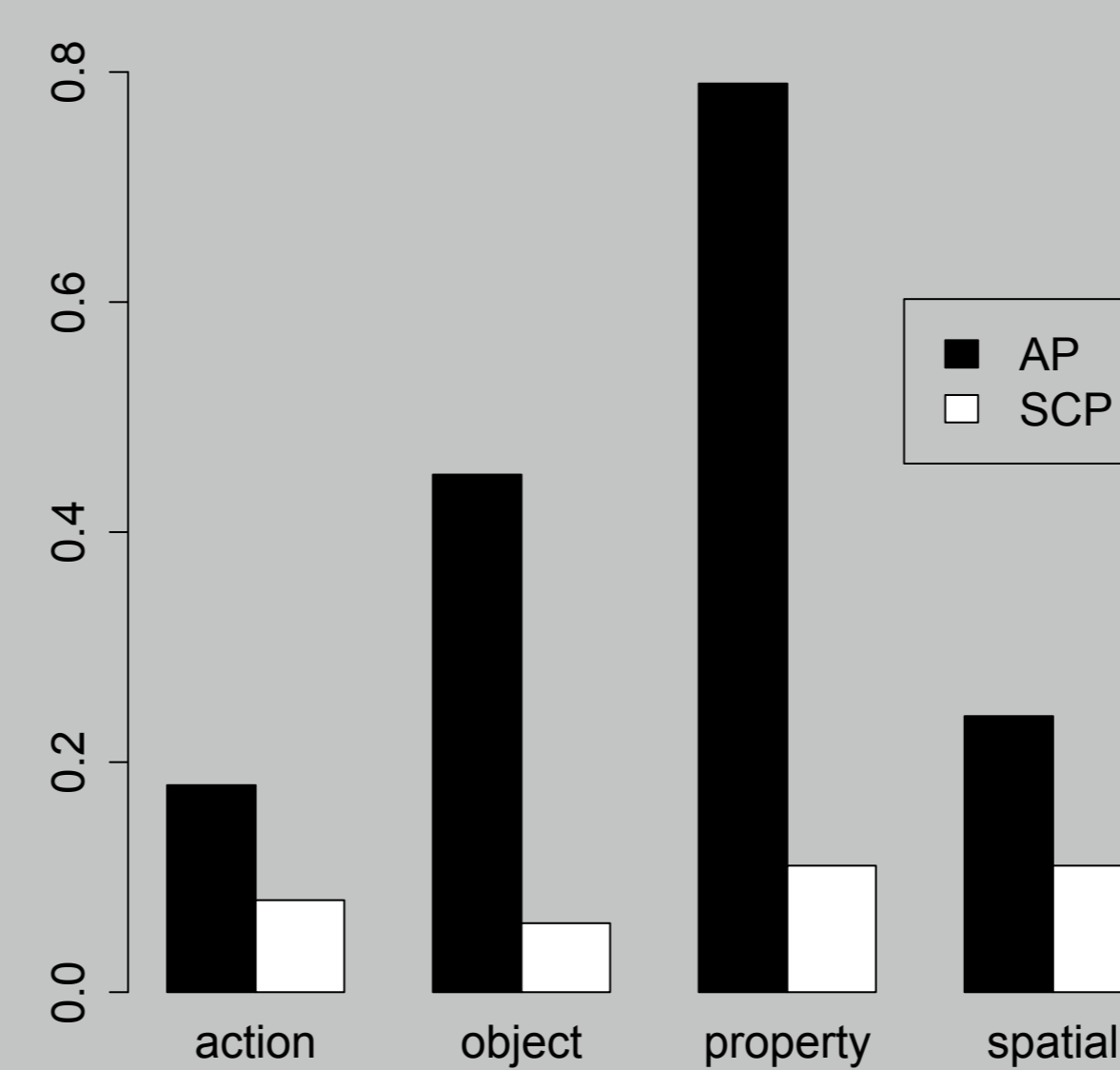
- ▶ Situation is the 3 second interval of the utterance



- SCP low in general
- AP poor, except properties
- Cross-situational availability is problematic, because of:
 - ▶ absence of relevant features,
 - ▶ overwhelming presence of irrelevant features,
 - ▶ low variability across situations

Experiment 2

- ▶ Situation consists of all intervals between the current utterance and the next



- SCP remains low
- AP general increase
- A pragmatically defined attention span
 - ▶ increases the availability of relevant features for words
 - ▶ while not increasing the irrelevant features

Key insights

- ▶ Developing annotations of naturalistic data is possible
- ▶ Cross-situational availability may be low in naturalistic data and the assumption of availability is problematic
- ▶ But: results depend on other assumptions (w.r.t. attention and intentions)
- ▶ Modeling has to move beyond using mere associations between situations and utterances and look into other mechanisms of word learning in order to understand the mechanisms involved
 - ▷ e.g. syntactic bootstrapping, intentions, attention, biases
 - ▷ experiment 2: wider attention span increases performance

References

- A. Fazly, A. Alishahi, and S. Stevenson (2010). A probabilistic computational model of cross-situational word learning. *Cognitive Science*, 34(6):1017–1063.
 - D. Gentner (1978). On Relational Meaning: The Acquisition of Verb Meaning. *Child Development*, 49:988–998.
 - L. Gleitman (1990). Sources of Verb Meanings. *Language Acquisition*, 1(1):3–55.
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