







# Some of Them Can be Guessed! Exploring the Effect of Linguistic Context in Predicting Quantifiers

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Motivation	Task & Datasets			
Quantifiers ('few', 'some', 'all') are interesting because:	Cloze test	Quantifiers		

- They are typically considered as function words (as opposed to nouns, verbs, etc.), but they have a rich semantics
- They are of central importance in linguistic semantics and its interface with cognitive science [1,2,3]
- Their choice depends on both local and global context [4]
- Larger contexts are claimed to be detrimental for the prediction of function words in cloze test [5]

### Hypotheses

- Human performance **boosted** by more context (proportional Qs)
- Models (very) effective with local context, hurt by broader context

< <b>qnt&gt;</b> the island's breeding birds are endemic.	
	all of
The island is one of the world's most biologically	almost all of
	few of
diverse areas, with many endemic species.	many of
<pre><qnt> the island's breeding birds are endemic.</qnt></pre>	more than half of
Other endemic species include the red-bellied	most of
	none of
lemur, the indri, and the aye-aye.	some of

# **Datasets**

**1–Sent** 10350 target sentences (quantifer+ of at beginning):  $\langle s_t \rangle$ 

**3-Sent** 10350 preceding +  $s_t$  + following:  $< s_p, s_t, s_f >$ 

### Human Evaluation

Humans vs Models

### **Models & Results**

## Crowdsourcing

### **Models**

- Two experiments, one per condition (1-Sent, 3-Sent)
- 506 examples from validation set (same in two conditions)
- 3 judgments/datapoint; correctly-guessed w/ agreement > 0.66
- Higher accuracy in 3-Sent (0.258) compared to 1-Sent (0.221)



8 models tested: 3 BoW baselines, 1 CNN, 4 LSTMs 2 conditions: 1-Sent, 3-Sent Data: 80% train, 10% val, 10% test

	1-Sent		3-Sent	
	val	test	val	test
chance	0.111	0.111	0.111	0.111
BoW-conc	0.270	0.238	0.224	0.207
BoW-sum	0.308	0.290	0.267	0.245
fastText	0.305	0.271	0.297	0.245
CNN	0.310	0.304	0.298	0.257
LSTM	0.315	0.310	0.277	0.253
bi-LSTM	0.341	0.337	0.279	0.265
Att-LSTM	0.319	0.324	0.287	0.291
AttCon-LSTM	0.343	0.319	0.274	0.288
Humans	0.221*		0.258*	

### **Discussion & References**



### **Discussion**



Humans **do better** w/ broader contexts especially on proportional Qs; models **suffer** due to their inability to handle longer sequences

Models capitalize more on **lexical** cues compared to humans: 41% cases in 3-Sent (hum. 24%) and 50% cases in 1-Sent (hum. 44%)

### References

[1] Jon Barwise and Robin Cooper. 1981. Generalized Quantifiers and Natural Language. Linguistics and Philosophy 4(2):159-219.

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[4] Kevin B. Paterson, Ruth Filik, and Linda M. Moxey. 2009. Quantifiers and Discourse Processing. Language and Linguistics Compass.

[5] Frank Smith. 1971. Understanding reading: A psycholinguistic analysis of reading and learning to read. Holt, Rinehart & Winston.