

# Pragmatic Reasoning Unlocks Quantifier Semantics of Foundation Models

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    - For all/any
  - Generalized Quantifiers (*Mostowski 1957*) **fuzzy scopes**
    - Few/some/most etc.
    - Indicate the proportion that predicates satisfy.

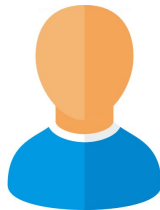
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    - Few/some/most etc.
    - Indicate the proportion that predicates satisfy.
  - Abundant in communications (*Joshi et al. 2020, Cui et al. 2022*).

# Question to Answer

**Some** birds can fly.  $\longrightarrow$   **$X\%$  ( $0 < X < 100$ )** birds can fly.

Understanding/Reasoning



Implicit functionalities

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(Bommasani et al. 2021)

# Contributions

- An annotated dataset QuRe targeting real-world quantifier understanding.
- A pragmatic reasoning based framework PRESQUE for understanding quantifier semantics.

# Task Definition

- Quantifier Understanding
  - Predicting the percentage scope (with an interval width) of a quantified sentence.
    - Splitting  $[0-1]$  into intervals  $W$ , e.g.  $\{0\%, 5\%, 10\%, \dots\}$
    - A quantifier understanding model predicts percentage scope from  $W$  that the predicate in the quantified sentence holds true (e.g. 5% - 30%).



# Dataset

- Limited number of datasets with **human annotated quantifications**.
- HVD (*Herbelot and Vecchi 2015*)
  - quantifier annotation on the <concept, feature> pairs.

CONCEPT	FEATURE	ANNOTATIONS	SENTENCE BASED ON TEMPLATE
rock	has_minerals	all, all, most	All rocks have minerals.
van	has_sliding_doors	most, most, most	Most vans have sliding doors.
sandpaper	has_fine_sand_covering_it	some, some, all	Some sandpapers have fine sand covering it.
banana	is_round	no, no, no	No bananas are round.
tricycle	used_for_transportation	all, few, few	Few tricycles are used for transportation.

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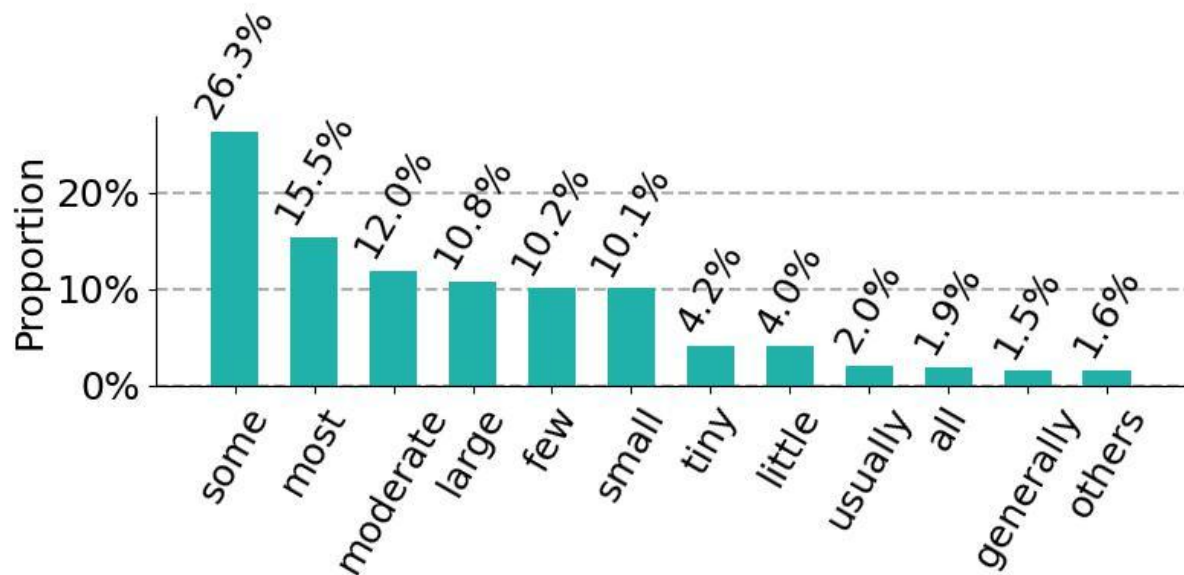
**No golden percentage scopes**

# QuRe Dataset

- QuRe
  - **More** generalized quantifiers.
    - {all, generally, most, usually, some, likely, few, little, occasionally, none, seldom, tiny, small, moderate, large}
  - **Specificity levels** in quantifier understanding
    - How hard it is to reason the percentage scope from the context.
  - **Golden percentage scopes** available.
    - The average age of the 304 drummers at Waterloo was 25, with some being boys under 16.
    - The average age of the 304 drummers at Waterloo was 25, with about 10% being boys under 16.
  - Sentence **topics**.

# QuRe Dataset

- Quantifier distribution



# QuRe Dataset

## - Metadata examples

[WIKI ENTITY] ORIGINAL SENTENCE	[SPECIFICITY, EXPRESSION] QuRe SENTENCE	TOPICS
<b>[Human]</b> Most humans (61%) live in Asia; the remainder live in the Americas (14%), Africa (14%), Europe ( <u>11%</u> ), and Oceania (0.5%). Within the last century, humans have explored challenging environments such as Antarctica, the deep sea, and outer space.	<b>[Fully, 0.11]</b> Most humans (61%) live in Asia; the remainder live in the Americas (14%), Africa (14%), <u>some</u> Europe, and Oceania (0.5%). Within the last century, humans have explored challenging environments such as Antarctica, the deep sea, and outer space.	population continents exploration
<b>[The Jungle Book (2016 film)]</b> The Jungle Book was shown across 4,028 theaters of which 3,100 theaters ( <u>75%</u> ) were in 3D, including 376 IMAX screens, 463 premium large format screens, and 145 D-Box locations.	<b>[Fully, 0.75]</b> The Jungle Book was shown across 4,028 theaters of which <u>most</u> (3,100) theaters were in 3D, including 376 IMAX screens, 463 premium large format screens, and 145 D-Box locations.	theaters movie release 3D technology
<b>[Electric car use by country]</b> The EV market share of total new and used cars first registered during 2018 was <u>2.8%</u> based on 5,557 out of a total of 198,600 first registered cars. 7,542 vehicles were registered in this country over 2019.	<b>[Fully, 0.028]</b> The EV market share of total new and used cars first registered during 2018 was <u>small</u> based on 5,557 out of a total of 198,600 first registered cars. 7,542 vehicles were registered in this country in 2019.	electric vehicles market share registration numbers

# Pragmatic Reasoning in Quantifier Understanding

- **Pragmatic Reasoning for Semantics of Quantifiers: PRESQUE**
  - NLI backbone for text understanding.
  - Adaptation of Rational Speech Act (RSA).
  - **No training data needed**

# Natural Language Inference (NLI)

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- In PRESQUE, *quantified premises* to *percentaged hypotheses*.
  - e.g. *few staircases have a spiral structure*, *20% staircases have a spiral structure*.



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  - e.g. *few staircases have a spiral structure*, *20% staircases have a spiral structure*.

weak description (less exact)

strong description (more exact)

# Limitations of NLI in Quantifier Understanding

- Implicit percentage value in quantifiers (*Horowitz et al. 2018*)
- Sentence-level relation nature, impacts of linguistic and social clues (*Bergen et al. 2016*).
- Deficiencies in ambiguous premises (*Thukral et al. 2021*) and quantative reasoning (*Naik et al. 2018; Ravichander et al. 2019*)

# Beyond Direct Interpretation

- Pragmatic Theory (*Grice 1975*)
  - Locates the semantic meaning in interpretation considering the communication goal.
  - Reduced the complexity of semantic theories required for interpretation (*Bergen et al. 2016*)

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- Quantifier understanding through Rational Speech Act (RSA, *Frank and Goodman 2012*)

# Rational Speech Act (RSA)

- World states  $W$  and utterances  $U$ .
- Lisenter  $L$  and speaker  $S$ .
- Bayesian approach of the pragmatic theory (iteratively modeling the mental state of  $L$  and  $S$ ).

# Quantifier Understanding through RSA

- $W = \{0\%, 10\%, 20\%, \dots, 100\%\}$  percentage value basis
- $U = \{\text{no, few, some, most, all}\}$  quantifier basis

# Quantifier Understanding through RSA

- $W = \{0\%, 10\%, 20\%, \dots, 100\%\}$
- $U = \{\text{no, few, some, most, all}\}$
- premise  $\bar{p}$  : "All airplanes have engines."
- hypothesis  $\bar{h}$  : "90% airplanes have engines."

# Quantifier Understanding through RSA

- premise  $\bar{p}$  : "All airplanes have engines."
- hypothesis  $\bar{h}$  : "90% airplanes have engines."
- Literal listener **baseline**

$$L_0(p|q) \propto \text{Entailment}(\tilde{p}, \tilde{h})$$

- Pragmatic speaker

$$S_0(q|p) \propto \text{Entailment}(\tilde{h}, \tilde{p})$$

- Pragmatic listener

$$L_1(p|q) \propto S_0(q|p)P(p)$$

$$P(p) = \sum_{q \in \mathcal{U}} P(p|q)P(q)$$




# Model Choices of PRESQUE

- Foundation models
  - ALBERT-XXLarge (*Lan et al. 2020*)
  - XLNet-Large (*Yang et al. 2019*)
  - BART-Large (*Lewis et al. 2020*)
  - RoBERTa-Large (*Liu et al. 2019*)
- NLI finetuning datasets
  - SNLI (*Bowman et al. 2015*)
  - MNLI (*Williams et al. 2018*)
  - NLI-style FEVER (*Nie et al. 2019*)
  - Adversarial NLI (*Nie et al. 2020*)

# Baselines

- Randomly ranking percentage values (Rnd)
- Literal listener ( $L_0$ ): direct interpretation of NLI.

# Evaluation Metrics

- HVD
  - Cross Entropy: The similarity between human and model perception of quantifier semantics.
- QuRe (starting from classification)
  - HIT@1: Topmost percentage value lies in the golden percentage scope.
  - Mean Reciporal Rank (MRR): The general ranking of the golden scope.
  - Cross Entropy: likelihood of the scope predictions.
  - Minimal Scope Distance (MSD@K): The distance of scope prediction of top K values and the golden scope.

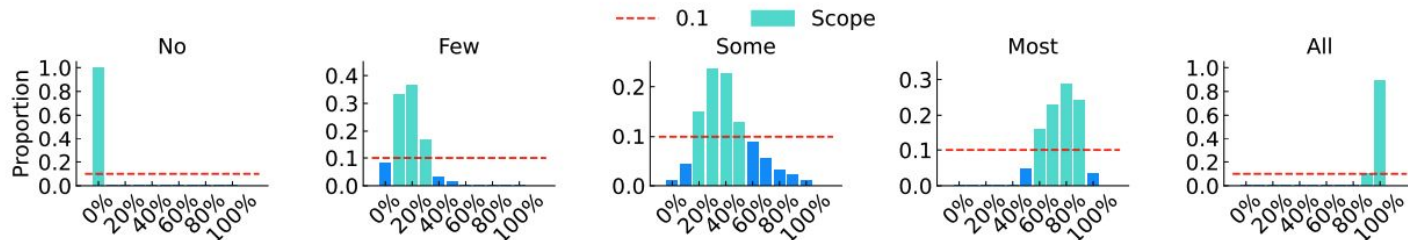
# Human Perception

- Instruct the annotator to define the percentage scope of the given quantifier (e.g. "Some stands for?").

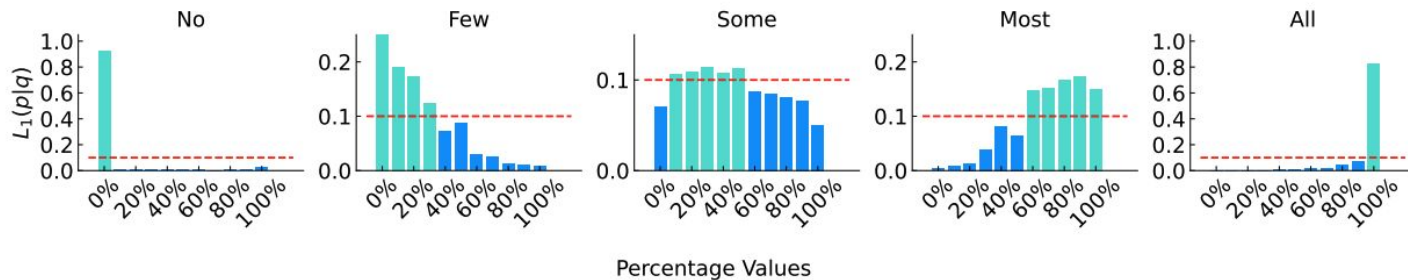
# Model Perception - HVD

- Human perception (**H**) is similar to PRESQUE (**P**)

**H**



**P**



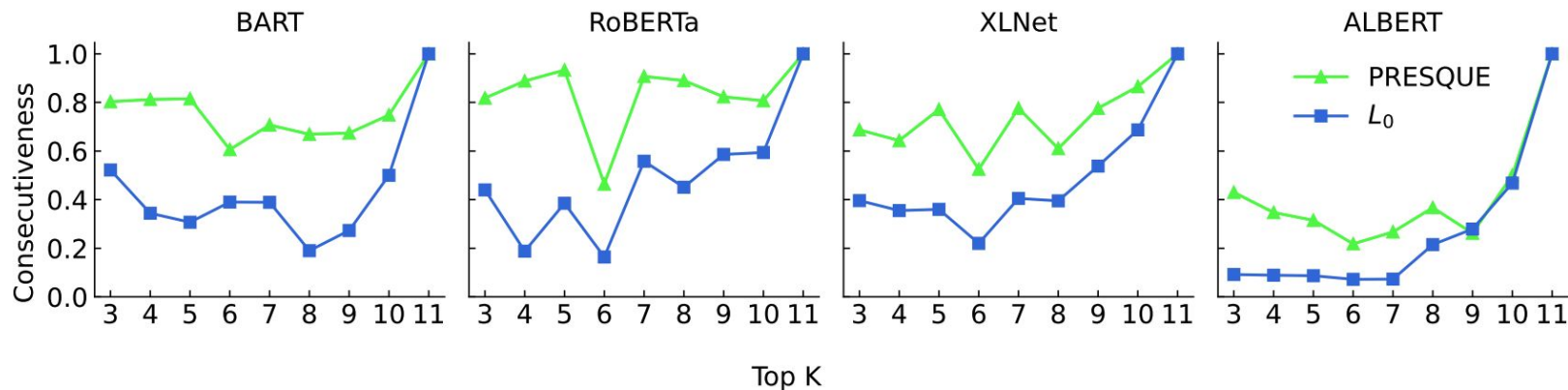
## Result - HVD

- PRESQUE is better than the literal listener ( $L_0$ ).
- RoBERTa generally performs best among model choices.

BASE MODEL(#PARAM.)	CROSSENTROPY↓	
	$L_0$	PRESQUE
ALBERT (Lan et al., 2020) (222M)	1.76	1.48
XLNet (Yang et al., 2019) (361M)	<b>1.64</b>	1.35
BART (Lewis et al., 2020) (407M)	1.89	1.32
RoBERTa (Liu et al., 2019) (355M)	1.69	<b>1.29</b>

# Consistency

- Consecutiveness of the Top K percentage inferences.
  - {10%, 20%, 30%}: consecutive (10%-30%)
  - {10%, 30%, 50%}: not consecutive
- PRESQUE has higher consecutiveness than  $L_0$ .



# Result - QuRe

- PRESQUE generally performs better than  $L_0$  among all specificity levels.

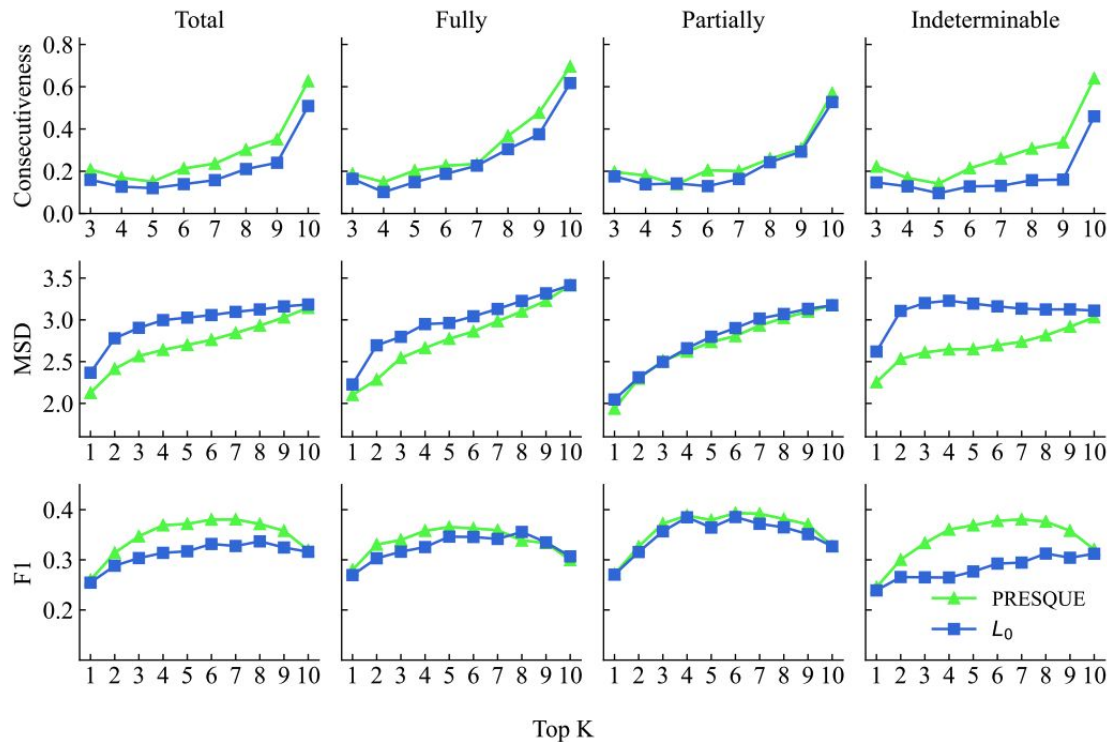
SPECIFICITY	HIT@1↑			MRR↑			CROSSENTROPY↓			F1@{1, 5}↑		
	Rnd.	$L_0$	$L_1$	Rnd.	$L_0$	$L_1$	Rnd.	$L_0$	$L_1$	Rnd.	$L_0$	$L_1$
Fully	4.1	27.3	<b>29.7</b>	12.3	22.1	<b>24.3</b>	6.44	<b>5.64</b>	5.74	2.8/8.6	19.5/24.3	<b>21.5/26.5</b>
Partial	8.2	26.4	<b>28.5</b>	11.6	21.2	<b>21.7</b>	7.78	<b>6.99</b>	7.06	4.3/8.3	16.9/25.9	<b>18.3/27.3</b>
Indeterminable	9.7	21.4	21.4	12.5	18.1	<b>22.7</b>	7.76	7.20	<b>6.69</b>	5.3/10.1	<b>14.9/18.2</b>	14.8/ <b>25.6</b>
Total	7.9	24.0	<b>25.1</b>	11.8	19.8	<b>22.7</b>	7.47	6.86	<b>6.78</b>	4.4/9.3	16.3/21.7	<b>17.1/26.3</b>



# Result - QuRe

- Consistency + distance based scope evaluation

PRESQUE predictions has higher consecutiveness and are more similar to the golden percentage scopes than  $L_0$ .



# Result - QuRe

## - Examples

[GS.] SENTENCE <sub>Q</sub> / [SPC.] SENTENCE <sub>P</sub>	PRIMARY SCOPE	MRR	F1@5	CE
[F] In 57 separate fights, one loss was observed to Neope goschkevitschii, giving V. mandarinia a <u>large</u> winning rate.	L <sub>0</sub> : 5%-20%	0.11	0.00	7.67
[95%-100%] In 57 separate fights, one loss was observed to Neope goschkevitschii, giving V. mandarinia a win rate of <u>98.3%</u> .	L <sub>1</sub> : 85%-100%	<b>0.67</b>	<b>0.67</b>	<b>3.52</b>
[F] In the 2017 Dutch study, only (2 out of the total 27) <u>few</u> school children recognized that the website was a hoax.	L <sub>0</sub> : 0%	0.08	0.00	7.79
[5%-10%] In the 2017 Dutch study only 2 out of the total 27 school children ( <u>7%</u> ) recognized that the website was a hoax.	L <sub>1</sub> : 0%-5%	<b>0.11</b>	<b>0.50</b>	<b>6.36</b>
[P] From 4 locations in different parts of Europe, a <u>large</u> number had clutch size of 2, 41% had size of 3, clutches of 1 and 4 each constituted about 8%.	L <sub>0</sub> : 30%-40%	0.22	0.40	6.29
[40%-45%] From 4 locations in different parts of Europe, <u>43%</u> had clutch size of 2, 41% had size of 3, clutches of 1 and 4 each constituted about 8%.	L <sub>1</sub> : 30%-45%	<b>0.33</b>	<b>0.67</b>	<b>4.92</b>

Paper: <https://arxiv.org/pdf/2311.04659.pdf>

Code: <https://github.com/Nativeatom/PRESQUE>

Thank you