# technische universität dortmund

**Predicting Human Responses to Syllogism Tasks Following the Principle of Maximum Entropy** Marco Wilhelm, Gabriele Kern-Isberner

## Maximum Entropy Reasoning

Beliefs are represented by probabilistic conditionals

(B|A)[p]

"If proposition A holds, then B follows with probability p."

- ► Probabilities reflect reasoner's degrees of belief in possible worlds  $\omega \in \Omega$ ( $\widehat{=}$  interpretations)
- Belief states (\$\hightarrow\$ probability distributions) are inferred from belief bases \$\mathcal{R}\$ (\$\hightarrow\$ sets of conditionals) following the principle of maximum entropy:

#### Syllogisms

- Syllogisms are quantified inference schemas:
  - If  $\{all, no, some, some not\}$  *M*'s are *P*'s and  $\{all, no, some, some not\}$  *S*'s are *M*'s,

then {all, no, some, some not} S's are P's.

Varying the quantifiers and the order of M, P, S leads to 256 syllogisms, of which 24 are logically valid

$$\mathsf{ME}(\mathcal{R}) = rg\max_{\mathcal{P} \models \mathcal{R}} - \sum_{\omega \in \Omega} \mathcal{P}(\omega) \cdot \log(\mathcal{P}(\omega))$$

Acceptance of syllogisms is investigated to better understand human understanding of quantification (and of uncertainty in general)

#### PreCORE Challenge

Initial situation: The premises of syllogisms are shown to study participants and the participants are asked to draw the conclusion among

{all, no, some, some not} { S's are P's, P's are S's}

they do accept most ("no valid conclusion" (nvc) is also allowed)

- Challenge: Submit a model which predicts the choices of the participants (training data and answers of individuals to previous syllogism tasks are provided)
- Models compete on unknown data against cognitive models (cf. diagram) (similar data available on web to test models beforehand)

## Results:<sup>1</sup> (training data: Veser2018.csv, test data: Ragni2016.csv)

Accuracy of Model Predictions



#### Prediction accuracy of MaxEnt: 35.45%

<sup>1</sup>wrt. data available on web (cf. refs) showing the same tendency as in the challenge

## MaxEnt Model – Basic Version

- 1. Translate premises of syllogism into probabilistic conditionals  $(\rightarrow \mathcal{R})$ : {all, no, some, some not} M's are P's  $\rightarrow (M|P)[p]$
- **2.** Calc. MaxEnt probabilities  $q = ME(\mathcal{R})(P|S)$  for possible conclusions

" $\alpha$  S's are P's" with  $\alpha \in \{$ all, no, some, some not $\}$ 

**3.** Return conclusion for which q fits best to  $\alpha$  (if none fits, return nvc)



#### Parameter selection:

- Translate  $\alpha \in (\text{all, no, some, some not})$  to  $p \in (1, 0, 0.65, 0.15)$
- Tolerance of  $\pm 0.1$  when saying that q fits to  $\alpha \in \{$ some, some not $\}$
- If several conclusions fit, return the first one

## Variant submitted to challenge:

**Portfolio Model**  $\rightarrow$  Test if participant is a MaxEnt reasoner beforehand

## References

#### Maximum Entropy Reasoning: J.B. Paris (2006): The Uncertain Reasoner's Companion: A Mathematical Perspective. Cambridge University Press

#### Syllogisms:

S. Khemlani, P.N. Johnson-Laird, P.N. (2012): **Theories of the Syllogism: A Meta-Analysis.** Psychological Bulletin, 138, 427-457

#### PreCORE Challenge:

M. Ragni, N. Riesterer, S. Khemlani (2019): **Predicting Individual Human Reasoning: The PRECORE-Challenge.** Proceedings of CogSci2019 (Workshop)

https://www.cognitive-computation.uni-freiburg.de/modelingchallenge
(PreCORE Challenge)

https://orca.informatik.uni-freiburg.de/ccobra (CCobra Framework)

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- https://www.hybrid-reasoning.org/

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