PAP: A Dataset for Physical and Abstract Plausibility and Sources of Human Disagreement

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INTRODUCTION

- Discerning plausible from implausible events: crucial building block for NLP
- Previous work mostly focused on semantic knowledge to distinguish
 - *physically* plausible vs. implausible events
 - events with mostly conceptually *concrete* participants

physically plausible cat—eat—sardine law—prohibit—discrimination rain—break—belly humour—requires—merger concreteness abstractness

RESEARCH GOALS & CONTRIBUTIONS

- Create novel dataset for physical and abstract plausibility of events in English,
 capturing abstractness to the same extent as concreteness for the first time
- Systematically examine plausibility across levels of abstractness
- Explore and represent disagreement in plausibility annotation

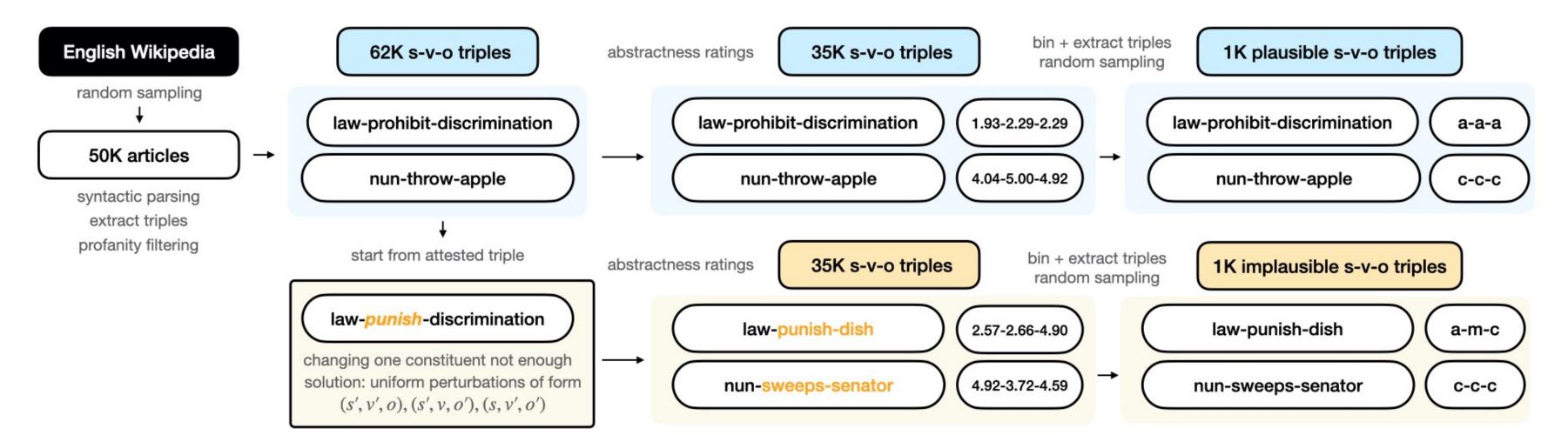
CAPTURING (SEMANTIC) PLAUSIBILITY

PLAUSIBILITY

- Captures non-surprisal in a given context child-sleep vs. tree-sleep
- Includes both what is preferred (and probably most plausible) and what is unusual (but still very much plausible), child-eat-banana vs. child-eat-pebble
 - → in contrast to selectional preference / thematic fit
- Can be estimated as a matter of degree with events assessed corresponding to perceived plausibility child-eat-banana vs. child-eat-pebble vs. child-eat-skyscraper
- Denotes what is likely in a given world but not necessarily attested in a given corpus

human-dies vs. human-breathes

CONSTRUCTING EVENT TARGETS



Simplified Illustration of Dataset Construction

PLAUSIBLE EVENTS (marked in blue)

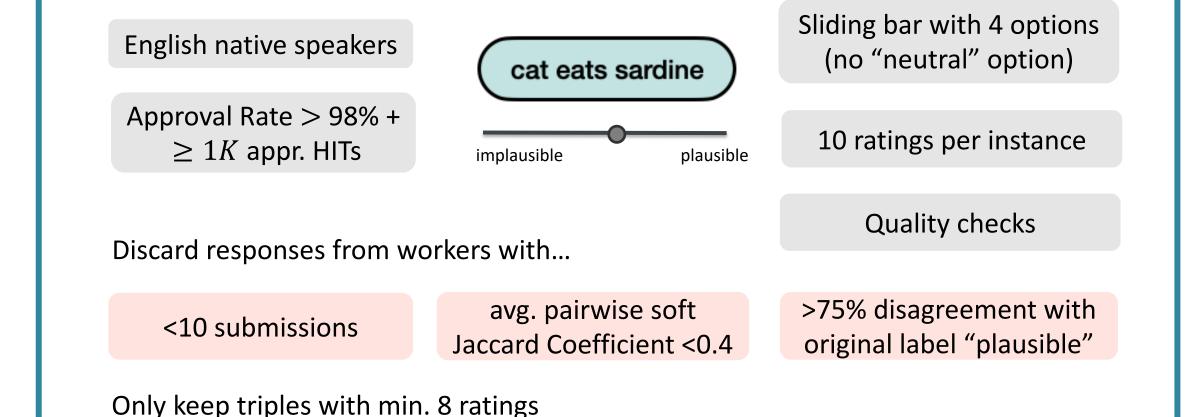
From English Wikipedia sample: Extract attested events, filter for profanity, assign abstractness ratings, bin according to abstractness, and sample 1,080 plausible events

(PSEUDO-) IMPLAUSIBLE EVENTS (marked in yellow)

- Based on extracted attested triples:
 - (i) Automatically generate pseudo-implausible events by perturbating event constituents
 - (ii) Construct 1,080 pseudo-implausible event similarly to plausible event construction

COLLECTING HUMAN ANNOTATIONS

TASK: Collect plausibility judgements on AMT for 2,160 plausible and implausible triples

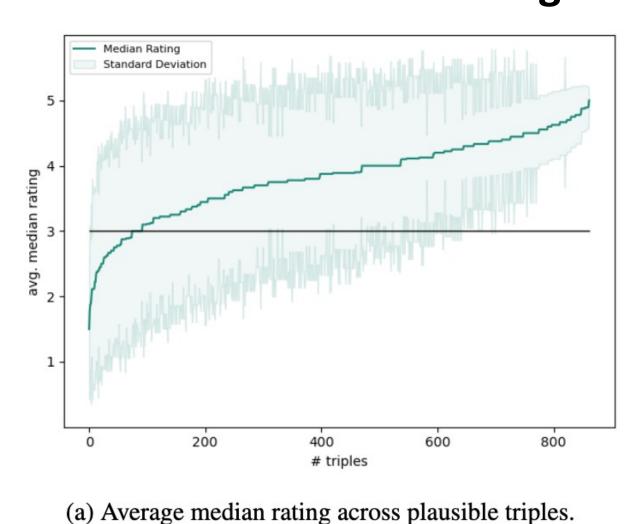


DATASET STATISTICS

- 15,571 plausibility ratings for 1,733 triples
- Ø IAA: Soft Jaccard Coefficient of 0.64
 - → reasonable agreement among annotators with indication of disagreement to be examined

ANALYSIS OF HUMAN JUDGEMENTS AND DISAGREEMENT

What can we learn from rating distributions?



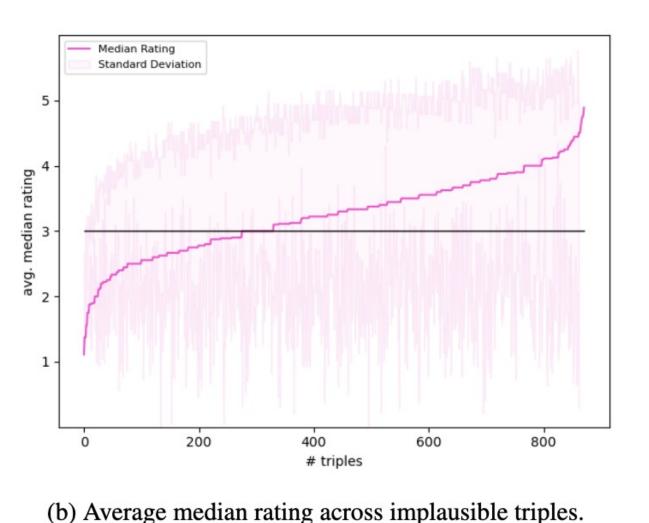


Figure 4: Average median ratings across originally plausible (a) and implausible (b) triples with standard deviation visualized as cloud around average rating lines. Triples are represented numerically on the x-axis. The black horizontal line denotes a median rating of 3. Average median ratings for *plausible* triples *below* the line disagree with the original label, while the opposite is true for average median ratings for *implausible* triples. Here, ratings *above* the line disagree with the original label.

- (i) Humans tend to favor plausibility over implausibility, while avoiding the extreme on the plausibility end of the scale.
- (ii) Implausibility yields higher disagreement as annotators disagree more when rating triples originally labelled implausible.

How does abstractness impact plausibility ratings?

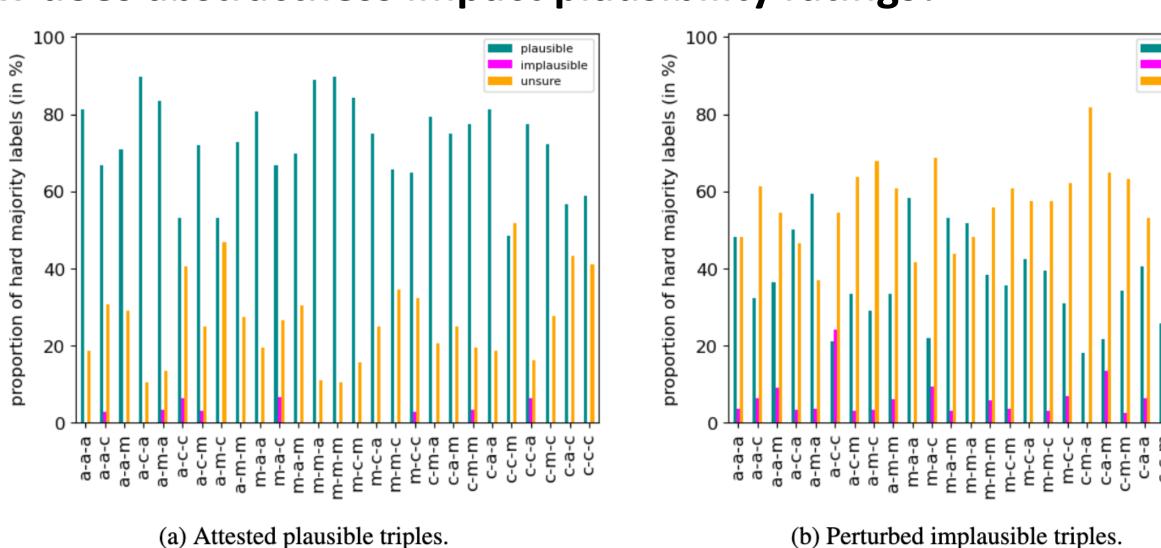


Figure 5: Proportion of strict majority ratings ($\geq 70\%$) across abstractness combinations for attested plausible triples (a) and perturbed implausible triples (b). Green bars denote a majority of plausible ratings $\in \{4, 5\}$, pink bars refer to a majority of implausible ratings $\in \{1, 2\}$, and orange bars capture cases of no clear majority.

- (i) Plausibility tends to be more likely to be assigned in case of more abstract event participants.
- (ii) Implausibility seems to be easier to capture with conceptually concrete words no matter the underlying original label.

ACKNOWLEDGEMENTS



CONCLUSIONS

- Presented a novel human-annotated dataset for physical and abstract plausibility for events in English
- Explored relationship between abstractness and plausibility and analyzed annotator disagreement
- Released both raw and a range of aggregated annotations to foster research on (semantic) plausibility and related notions, disagreement, and relevant downstream tasks such as commonsense reasoning

