

Reading Assignment 4

Hamza Rashid, 260971031
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The paper, “*Gender Bias in Coreference Resolution*” (Rudinger et al., 2018) proposes dataset schemas for measuring gender bias in Coreference Resolution Systems (CRS). To this end, the authors focus on gender bias with respect to occupations, evaluating the accuracy of Rule-based, Statistical, and Neural Coreference Systems in resolving a pronoun (male, female, or neutral) to a coreferent antecedent that is either an occupation or a participant. They constructed a challenge dataset, Winogender schemas, in the style of Winograd schemas, wherein a pronoun must be resolved to one of two previously mentioned entities in a sentence. The authors followed good practice by validating their hand-crafted dataset on Amazon’s Mechanical Turk (MTurk) with 10-way redundancy, with 94.9% of responses agreeing with their intended answers. This shows that the authors designed test sentences where correct pronoun resolution is not a function of gender. However, they do not report on their MTurk workers’ approval ratings or method for selecting them (e.g., qualification tests). With the Winogender schemas in hand, an unbiased model is expected to exhibit no sensitivity to pronoun gender in its accuracy, and to resolve a male or female pronoun to an occupation or participant with equal likelihood.

To construct the dataset, the authors used a list of 60 one-word occupations obtained from Caliskan et al. (2017), with corresponding gender percentages available from the U.S. Bureau of Labor Statistics. For each occupation, there are two sentence similar templates: one in which the pronoun is coreferent with the occupation, and one in which it is coreferent with the participant. For each sentence template, there are two instantiations for the participant referent (a specific participant, e.g., “the passenger,” and a generic participant, “someone.”). Thus, the resulting evaluation set contains 720 sentences: 60 occupations \times 2 sentence templates per occupation \times 2 participants \times 3 pronoun genders.

A key observation is that when each CRS’s predictions diverge based on pronoun gender, they do so in ways that reinforce and magnify real-world occupational gender disparities. As shown in figure 4 of the paper, the systems’ gender preferences for occupations correlate with BLS and the gender statistics from text (Bergsma and Lin, 2006), which these systems access directly. A notable case is “gotcha” sentences in which pronoun gender does not match the occupation’s majority gender (BLS) if the occupation is the correct answer; all models performed worse in this scenario. The paper discussed a potential case of bias amplification involving the occupation manager: 38.5% female in the U.S. according to BLS, and mentions of “manager” in the B&L resource are only 5.18%, yet no managers were predicted to be female by any of the coreference systems. The mechanism in which dataset bias could be amplified at the system level depends on the system; a Rule-based system is susceptible to the biases of the human designer and the resulting hand-crafted rules, a Statistical system is vulnerable to the biases of a feature function associating an occupation with a gender, and a Neural system’s pretrained embeddings is prone to capturing latent correlations between occupations and genders due to the biases in its pre-training data. In modern, Neural-based systems, system bias can lead to further amplification in society as the market for consumer chat-based LLM’s continues to grow. For example, the integration of Gemini in Google search is prone to gender bias in queries such as “most impactful computer scientists”, where the contributions of Ada Lovelace are likely to be overlooked in comparison with Alan Turing.

References

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