Gender and Representation: Investigations of Bias in Natural Language Processing

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Akademisk avhandling

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Abstract
Natural Language Processing (NLP) technologies are a part of our everyday realities. They come in forms we can easily see as ‘language technologies’ (auto-correct, translation services, search results) as well as those that fly under our radar (social media algorithms, ‘suggested reading’ recommendations on news sites, spam filters). NLP fuels many other tools under the Artificial Intelligence umbrella – such as algorithms approving for loan applications – which can have major material effects on our lives. As large language models like ChatGPT have become popularized, we are also increasingly exposed to machine-generated texts.

Machine Learning (ML) methods, which most modern NLP tools rely on, replicate patterns in their training data. Typically, these language data are generated by humans, and contain both overt and underlying patterns that we consider socially undesirable, comprising stereotypes and other reflections of human prejudice. Such patterns (often termed ‘bias’) are picked up and repeated, or even made more extreme, by ML systems. Thus, NLP technologies become a part of the linguistic landscapes in which we humans transmit stereotypes and act on our prejudices. They may participate in this transmission by, for example, translating nurses as women (and doctors as men) or systematically preferring to suggest promoting men over women. These technologies are tools in the construction of power asymmetries not only through the reinforcement of hegemony, but also through the distribution of material resources when they are included in decision-making processes such as screening job applications.

This thesis explores gendered biases, trans and non-binary inclusion, and queer representation within NLP through a feminist and intersectional lens. Three key areas are investigated: the ways in which “gender” is theorized and operationalized by researchers investigating gender bias in NLP; gendered associations within datasets used for training language technologies; and the representation of queer (particularly trans and non-binary) identities in the output of both low-level NLP models and large language models (LLMs).

The findings indicate that non-binary people/genders are erased by both bias in NLP tools/datasets, and by researchers attempting to address gender biases. Men and women are also held to cisheteronormative standards (and stereotypes), which is particularly problematic when considering the intersection of gender and sexuality. Although it is possible to mitigate some of these issues in particular circumstances, such as addressing erasure by adding more examples of non-binary language to training data, the complex nature of the socio-technical landscape which NLP technologies are a part of means that simple fixes may not always be sufficient. Additionally, it is important that ways of measuring and mitigating ‘bias’ remain flexible, as our understandings of social categories, stereotypes and other undesirable norms, and ‘bias’ itself will shift across contexts such as time and linguistic setting.

Keywords
NLP, natural language processing, gender bias, social impact of AI, gendered pronouns, neopronouns, gender studies, topic modeling