UTILISING SYNTHETIC DATA FROM LLM FOR DATA BIAS DETECTION AND MITIGATION

IN RECRUITMENT SYSTEMS

ABSTRACT

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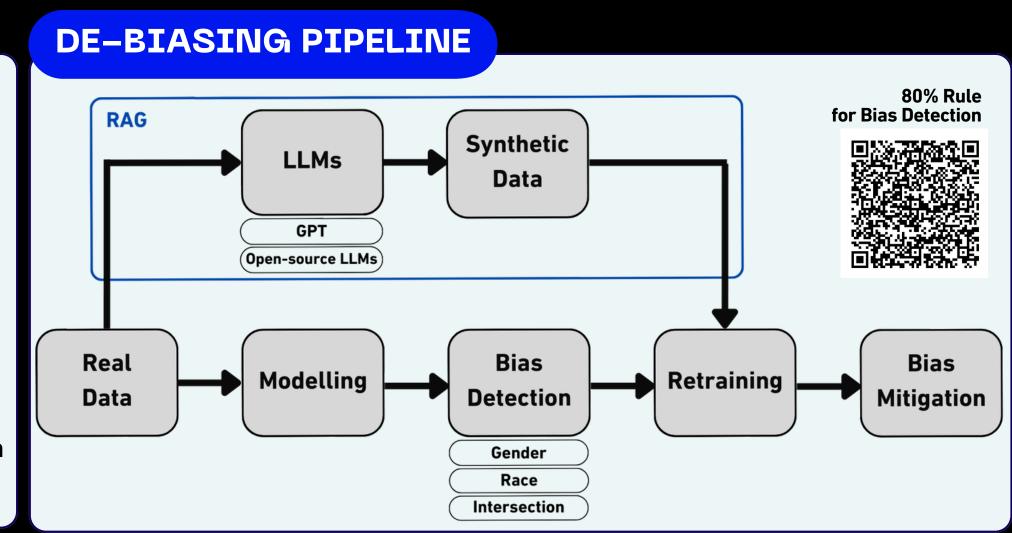




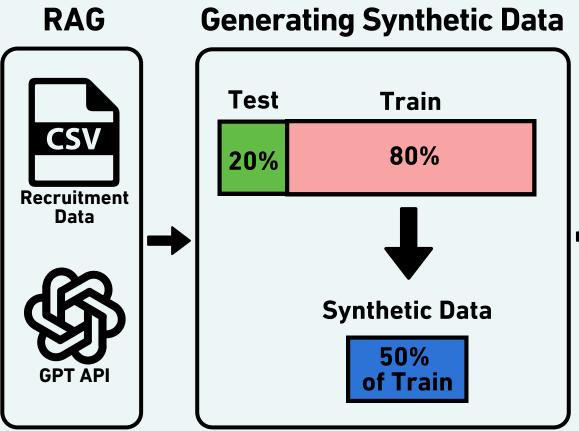
In the current landscape, diversity and inclusion are highly emphasised. This research develops a methodology to identify and mitigate biases in AI recruitment systems. EU AI Act classifies AI recruitment systems as high-risk areas, imposing numerous limitations on the use of real customer data to effectively address biases within these systems (GDPR). The process includes identifying biases, generating synthetic data with GPT, mitigating these biases, and comparing results before and after mitigation. Using GPT, the research creates high-quality, diverse synthetic data to retrain AI systems, addressing ethical and privacy concerns of using real data. The goal is to establish a comprehensive framework for ensuring fairness and ethical practices in AI-driven recruitment, promoting a diverse and inclusive employment environment.

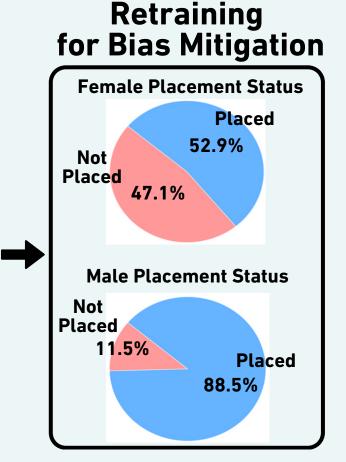
INTRODUCTION

As the use of synthetic data, free from data privacy issues, emerges as a hot topic, this research focuses on developing a de-biasing pipeline for recruitment, where diversity and inclusion are particularly emphasised. When biases are detected in existing models, this pipeline can quickly and cost-effectively reduce those biases. It enables data privacy protection, reduces data collection costs, and allows for rapid bias resolution. This approach utilises LLMs for bias mitigation.



METHODOLOGY





RESULT

Before retraining

	Testset Count	Placed Count	l	80% Rule	Bias Detection
Female	17	9	52.94%	70.77%	✓
Male	26	23	88.46%		
Total	43	32			

After retraining

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	Testset Count	Placed Count	Placed Rate		Bias Detection			
Female		13		70.77%				
Male	26	23	88.46%					
Total	43	36						

CONCLUSION

The development of a de-biasing pipeline using synthetic data and LLMs represents a significant step forward in creating fairer Al systems. By addressing biases in recruitment, we can promote a more diverse and inclusive employment environment. This research not only contributes to the academic understanding of bias mitigation but also provides practical solutions for organizations striving for ethical Al.

IMPROVEMENTS

- Generate synthetic data considering correlations in real data.
 (High Quality Synthetic Data)
- Utilise advanced prompt engineering techniques.
 (Accurate Synthetic Data)
- Compare the results with open-source LLMs.
 (LLM Performance Comparison)
- Identify better real data that includes race column.

Grant
As Part

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