

RE to Avoid Gender Bias

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Outline

- Background
- Research Questions
- Problem
- Detection
- RE as a Solution

Biased Computer System

- A computer systems that systematically and unfairly discriminate against certain individuals or groups of individuals in favor of others
- Example: An automated credit advisor

Unbiased System	Biased System
<ul style="list-style-type: none">• Denies credit to individuals with consistently poor payment records• Appropriate for a credit company to want to avoid extending credit privileges to people who consistently do not pay their bills	<ul style="list-style-type: none">• System that systematically assigns poor credit ratings to individuals with ethnic surnames• Discriminates on grounds that are not relevant to credit assessments and, hence, discriminates unfairly

Types of Bias

- Pre-existing Bias
 - has its roots in social institutions, practices, and attitudes
 - can enter a system
 - explicitly and conscious efforts of individuals or institutions, or
 - implicitly and unconsciously, even despite the best of intentions
 - Individual
 - Societal

Types of Bias

- Technical Bias
 - Arises from technical constraints or technical considerations
 - Caused due to
 - Computer Tools - limitation of the computer hardware, software or peripherals
 - Decontextualized Algorithms – an algorithm that fails to treat all groups fairly
 - Random Number Generation - imperfections in or misuse pseudorandom number generation
 - Formalization of Human Constructs - Bias that originates from attempts to make human constructs

Types of Bias

- Emergent Bias
 - This bias typically emerges some time after a design is completed, as a result of changing societal knowledge, population, or cultural values
 - UIs are mostly prone to this
 - Caused due to
 - New Societal Knowledge
 - Mismatch between Users and System Design

Gender Bias

Unfair differences in the way a person is treated because of their gender

Research Questions

- RQ₁
 - A systematic study of gender bias in software engineering
 - What leads to introduction of gender bias in a software?
 - How to detect gender bias in pre-existing system?

- RQ₂
 - RE to avoid gender bias
 - How can RE prevent/avoid gender bias in software?

Problem of Gender Bias

Study 1(a): Graphical Software

- *City County Opposites* by Random House
- Animals were used to make it gender neutral
- Male Animals – doctors, Policemen
- Female Animals – Cooking, cleaning

Study 1(b): Linguist Software

- *Streets* by Microcomputer Workshops
- Story telling for young kids
- Female character – Crazy old witch, less frequent
- Male character – Working and earning, more frequent

Problem of Gender Bias

- Study 2: Amazon's AI recruiting tool
- Global workforce is 60% male
- Men hold 74% of the company's managerial positions
- Dataset used was based on resumes submitted over 10-year period – mostly white male
- Software penalized any resume that contained the word "women's" in the text and downgraded the resumes of those who attended women's colleges

Problem of Gender Bias

- Another example, AI based targeted ads show better paying jobs to men
- Gartner predicts that by 2022, 85% of AI projects will deliver erroneous outcomes
- This will be caused due to bias in
 - data
 - algorithms
 - the teams responsible for managing them

What leads to introduction of gender bias in a software?

- No software is intentionally designed to be sexist
- Key component - Pre-existing bias in society
 - Lack of awareness
- Bias can creep in during all phases of a project
 - Problem Description
 - Considering an insufficiently rich set of factors
 - Lack of diverse data
- Lack of strict code towards gender neutrality starting from requirement gathering phase
- No focus on countering the issue which leads to further growth

Detecting Gender Bias

- Study 1(a) and 1(b) (Cont.)
- A study with 2 sections
 - Section A : 11 Students (1 male, 10 Female)
 - Section B: 7 Students (All female)
- Evaluation method
 - Expanded Sexism Checklist (ESC)– Supported characters, title, subject matter etc
 - Computers, Reading and Language Arts (CRLA) – yes/no
- Both sections chose 2 software and evaluated using both the methods

Detecting Gender Bias

- Most accurate method would be - Manual Method
 - High accuracy
 - Time consuming
 - Cannot be done for every piece of software
 - Prone to pre-existing bias
- Automated detection
 - Can be less accurate than manual
 - Fast
 - Prone to technical bias
 - Mostly successful for language issues
- Compare outputs for different groups

RE as a Solution

- Inclusive Design
 - Expanded sexism checklist should be included in each user manual
 - Requirement engineer should be made aware of gender bias issue so that it is continuously countered
 - UI design should be carefully defined as it directly impacts user psychology, including attitudes and intentions
 - Test cases specified during requirement gathering phase should include all minority cases

RE as a Solution

- Inclusive Design Method: The GenderMag Project
- Evaluates a system's gender inclusiveness
- Specialized personas
- Method
 - Group of software professionals walk through a scenario in their system, step by step, through the eyes of one of the GenderMag personas
 - At each step, they decide whether their persona (e.g., "Abby") will
 - know what to do and,
 - if Abby performs the action, whether she will know that she is progressing toward her goal

RE as a Solution

- Problem Description
 - Clearly worded
 - Preference should be given to gender neutral terms
 - Should be defined in a way that acknowledges different classes and how they will be impacted
- Dataset Description
 - Flawed data is a big problem!
 - Recognize and address statistical bias
 - Data should represent all groups
 - Sufficient data should be gathered (What do you call “sufficient”?)

RE as a Solution

CS 106A

Welcome to Programming Methods! This course is an introduction to computer programming and software development.

ASSIGNMENTS

PROJECTS

Name	Due
Send Out Email (REDESIGN)	4/14
Character Filter (P3)	4/24
Hangman (P4)	5/1
Breakout (P5)	5/8
Nobody Player + Image Algorithms (P6)	5/27
Critique (P8)	5/28
Handcuffs (P9)	6/3

SECTIONS

Name	Week
Section 1: Send Out Email, Basic Java	4/12 - 4/14
Section 2: Java Control Statements, Parameters	4/19 - 4/20
Section 3: Strings, Character File Processing	4/26 - 4/28
Section 4: Graphics, Animations, and Audio	5/5 - 5/7
Section 5: Arrays, Images, Trade	5/12 - 5/12
Section 6: Classes and Enums, Mathematical Objects	5/27 - 5/28
Section 7: Arrays/lan and Handicap	5/28 - 5/28
Section 8: Graphical User Interfaces	5/28 - 6/2

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Source: Metaxa-Kakavouli, D., Wang, K., Landay, J. A., & Hancock, J. (2018). *Gender-Inclusive Design*. Proceedings of the 2018 CHI Conference on Human Factors in Computing Systems - CHI '18. doi:10.1145/3173574.3174188

Future Work

- Extend the work on Cause, Detection and Solution of gender bias from
 - Standard SE Perspective
 - Inclusive Design focusing on RE
 - AI & ML perspective
 - NLP
- Work can be extended to other issues such as racism

References

- Nina Ribak Rosenthal & Diana Mayer Demetrulias (1988) Assessing Gender Bias in Computer Software, *Computers in the Schools*, 5:1-2, 153-164, DOI: 10.1300/J025v05n01_13
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- https://www.ey.com/en_gl/wef/why-we-need-to-solve-the-issue-of-gender-bias-before-ai-makes-it
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Thank You!
