MEASURING AND REDUCING DISPARITIES

SURESH VENKATASUBRAMANIAN UNIVERSITY OF UTAH

DISPARITY IN REPRESENTATION

VS

"Harms of allocation"

How the decisions made by a system affect different groups differently

Demographic parity Equal odds Disparate impact



Kate Crawford, NeurIPS 2017

"Harms of representation"

How the way we represent groups might cause harm

Stereotyping Ex-nomination Under-representation Denigration

How can we identify and measure disparities in representation?

Can we identify bias in existing (learned) representations and correct it?

Can we identify bias in existing (learned) representations and correct it?

man

doctor

woman

nurse

Can we identify bias in existing (learned) representations and correct it?



Can we identify bias in existing (learned) representations and correct it?



PROBLEM WITH GEOMETRY AS BIAS

Lipstick on a Pig: Debiasing Methods Cover up Systematic Gender Biases in Word Embeddings But do not Remove Them - Gonen and Goldberg, ACL2019



STEREOTYPING

"associations and beliefs about the characteristics and attributes of a group and its members that shape how people think about and respond to the group"

— SAGE handbook of prejudice, stereotyping and discrimination.

A specific mechanism for stereotyping:

...the tendency to assign characteristics to all members of a group based on stereotypical features shared by a few...

REPRESENTATIVENESS [Bordalo et al 16]

"distorted perception of the relative frequency of a type in the stereotyped group compared to the base group"

$$R(t,G) = \frac{\Pr(t \mid G)}{\Pr(t \mid \overline{G})}$$
 "representativeness"

 $Pr'(t \mid G) \propto Pr(t \mid G)R(t, G)^{\rho}$

 $\rho\,$ is the degree of stereotyping

EXEMPLARS

We can model stereotyping as a process by which groups experience non-uniform variance reduction [AFSV19]



Points regress towards an exemplar

 $\mathbf{p}_{\alpha} = (1 - \alpha)\mathbf{p} + \alpha \mathbf{c}$

EFFECTS OF STEREOTYPING

• Linear regression (via exemplar stereotyping):

$$y = X\beta + \epsilon$$
$$\beta = (X^{\mathsf{T}}X)^{-1}X^{\mathsf{T}}y$$

- If we set $X' \leftarrow X + \Delta_{\alpha}$, then β' depends quadratically on α
- Can show similar behavior for other tasks (clustering, classification)

Distortion in representation has measurable effect on decision.

BROADER GOALS



DISPARITY IN ACCESS

- Social standing within a network confers utility on an individual.
- Social "position" in a network is a class marker defined by the *network*, not the individual. [Coleman]
- Should we be considered about discrimination based on social position? [boyd, Marwick and Levy]

INFORMATION ACCESS

- Social networks grow through recommendations as well as organically
- Network position confers advantage ([Granovetter])
- Access to information that improves network position relies on network position
- "edges in social network" == "biased input data"













INFORMATION CASCADES

An *active* node sends information on each edge with probability α and then becomes *inactive*.



INFORMATION CASCADES

An *active* node sends information on each edge with probability α and then becomes *inactive*.



INFORMATION CASCADES

An *active* node sends information on each edge with probability α and then becomes *inactive*.



WELFARE FUNCTIONS

 $p_v =$ Probability that v gets the information.

Welfare function
$$\mu : [0,1]^n \to \mathbb{R}$$

Welfare of
vertex set $V = \{v_1, v_2, \dots, v_s\}$
in graph *G* with seed set *S*:
 $\mu_G(S,V) = \mu(v_1, v_2, \dots, v_s)$
 $\mu_{r} = 1/n \sum_{v} p_v$
 $\mu_{-\infty} = \min_{v} p_v$

Find S^+ , $|S^+| = k$ so that $\mu_G(S^+ \cup S)$ is maximized.

ACCESS GAPS

Definition: Access gap of a partition V, V' of G with seed set S is

 $\mu(S,V) - \mu(S,V')$

Definition: In a graph G, the *rich get richer* if there is a partition (V, V') such that the optimal intervention S* satisfies

 $\mu(S^*, V') - \mu(S^*, V) > \mu(S, V') - \mu(S, V) > 0.$

The access gap increases after intervention

THE RICH ALWAYS GET RICHER

Proposition

Suppose μ is symmetric, increasing, and for any x_1, \ldots, x_m in [0, 1], there is some $1 \le \phi < \infty$ such that

$$\min_{i} x_i \le \mu(x_1, \dots, x_m) \le \left(\frac{1}{m} \sum_{i=1}^m x_i^{\phi}\right)^{1/\phi}$$

Then if $0 < \alpha < \frac{1}{2\phi}$, $\exists G, S$ where the rich get richer.

THE RICH ALWAYS GET RICHER

Proposition

Suppose μ is symmetric, increasing, and for any x_1, \ldots, x_m in [0, 1], there is some $1 \le \phi < \infty$ such that

 $\min_i x_i \le \mu(x_1)$

Then if $0 < \alpha < \frac{1}{2\phi}$, $\exists G, S$ where th

UNDER ANY WELFARE FUNCTION, THE RICH GET RICHER

 $1 < (1 m) 1/\phi$

 μ is *k*-imbalanced if $\exists G$, seed set *S*, partition *V*, *V'* and optimal interventions *S*^{*} and *S_V* (for *V* only) with no more than *k* seeds such that

- 1. $\mu(S, V) \leq \mu(S_V, V)$ (there's a way to improve welfare of *V*)
- 2. $\mu(S_V, V) \leq \mu(S, V')$ (...but *V*''s welfare is then better)
- 3. $\mu(S, V') \leq \mu(S^*, V')$ (V''s welfare can be improved by the global optimum)
- 4. $\mu(S^*, V) \leq \mu(S, V)$ (...but *V*'s welfare does not improve)

 μ is *k*-imbalanced if $\exists G$, seed set *S*, partition *V*, *V'* and optimal interventions *S*^{*} and *S_V* (for *V* only) with no more than *k* seeds such that

- 1. $\mu(S, V) \le \mu(S_V, V)$ (there's a way to improve welfare of *V*)
- 2. $\mu(S_V, V) \leq \mu(S, V')$ (...but *V*''s welfare is then better)
- 3. $\mu(S, V') \leq \mu(S^*, V')$ (V''s welfare can be improved by the global optimum)
- 4. $\mu(S^*, V) \leq \mu(S, V)$ (...but *V*'s welfare does not improve)

Minimax welfare $\mu_{-\infty}$ is balanced (improve minimum access)

 μ is *k*-imbalanced if $\exists G$, seed set *S*, partition *V*, *V'* and optimal interventions *S*^{*} and *S_V* (for *V* only) with no more than *k* seeds such that

- 1. $\mu(S, V) \le \mu(S_V, V)$ (there's a way to improve welfare of *V*)
- 2. $\mu(S_V, V) \leq \mu(S, V')$ (...but *V*''s welfare is then better)
- 3. $\mu(S, V') \leq \mu(S^*, V')$ (V''s welfare can be improved by the global optimum)
- 4. $\mu(S^*, V) \leq \mu(S, V)$ (...but *V*'s welfare does not improve)

Minimax welfare $\mu_{-\infty}$ is balanced (improve minimum access)

Let $\phi > -\infty, \alpha < 1$ Then μ_{ϕ} is $\Omega(n)$ -imbalanced

 μ is *k*-imbalanced if $\exists G$, seed set *S*, partition *V*, *V'* and optimal interventions *S*^{*} and *S_V* (for *V* only) with no more than *k* seeds

- 1. $\mu(S, V) \leq \mu(S_V, V)$ (there's a way to '
- 2. $\mu(S_V, V) \le \mu(S, V')$ (...but V''s well
- 3. $\mu(S, V') \leq \mu(S^*, V')$ (V''s welfare commum)

AND NO OTHER WELFARE FUNCTION IS BALANCED

sti-

4. $\mu(S^*, V) \leq \mu(S, V)$ (...but V's welfare

.ot improve)

Minimax welfare $\mu_{-\infty}$ is balanced (improve minimum access)

Let $\phi > -\infty, \alpha < 1$

Then μ_{ϕ} is $\Omega(n)$ -imbalanced

- Questioning the frame
 - CS does not naturally take a critical stance towards its problems
 - But we are seeing more of it now

What are the **traps** we encounter if we ignore the **social** in a sociotechnical system [SbFVV19]

What are effective ways to harness the power of computer science for social change? [ABKLRR20]

- Questioning the frame
 - CS does not naturally take a critical stance towards its problems
 - But we are seeing more of it now

Fairness and Abstraction in Sociotechnical Systems

What are the **traps** we encounter if we ignore the **social** in a sociotechnical system [SbFVV19]

What are effective ways to harness the power of computer science for social change? [ABKLRR20]

- Questioning the frame
 - CS does not naturally take a critical stance towards its problems
 - But we are seeing more of it now

Fairness and Abstraction in Sociotechnical Systems

What are the **traps** we encounter if we ignore the **social** in a sociotechnical system [SbFVV19]

Roles for computing in social change

What are effective ways to harness the power of computer science for social change? [ABKLRR20]

- CS is not merely a "have data, will compute" field.
- The computational lens allows for precision, and therefore allows us to articulate limits.

The (im)possibility of fairness: different value systems require different mechanisms for fair decision making*

[FSV16]

Inherent Trade-Offs in the Fair Determination of Risk Scores

[KMR16]

• The "technology is problematic" frame assumes a fixed formulation of technical questions. We can change that!



