



# On Formalizing Fairness in Prediction with Machine Learning

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# Fairness: From Social Science to Machine Learning

## ❑ Machine learning

→ Critical decision making affecting **human** lives

- ❑ ML algorithms should be prevented from systematic discrimination

**“ML with Fairness”**



# Fairness: From Social Science to Machine Learning

- ❑ However, people might have **diverse understandings** of fairness...

The New York Times

## *The Harvard Bias Suit by Asian-Americans: 5 Key Issues*



The basic claim by the plaintiffs, a group representing Asian-American students rejected by Harvard, is that the university has systematically discriminated against Asian-Americans by holding them to a higher standard than other applicants. Harvard argues that in trying to compose a diverse class, it considers each applicant as an individual and does not discriminate.

<https://www.nytimes.com/2018/12/20/us/harvard-asian-american-students-discrimination.html>

# Fairness: From Social Science to Machine Learning

This paper...

- ❑ Introduce **different notions of fairness** and how they are **formalized** in machine learning literature
- ❑ Provide theoretical and empirical **critiques** of each notion from **social sciences**
- ❑ Determine the **suitability** of each formalization of fairness in the context of **machine learning**

# Formalizations of Fairness: Taxonomy

- ❑ **Parity or preference?** : whether fairness means achieving parity or satisfying the preferences.
- ❑ **Treatment or impact?** : whether fairness is to be maintained in treatment or impact (results).

7 existing notions of  
fairness in ML literature

	Parity	Preference
Treatment	Unawareness Counterfactual measures	Preferred treatment
Impact	Group fairness Individual fairness Equality of opportunity	Preferred impact

# What is Fair: Fairness through Unawareness

## □ Definition 1

*A predictor is said to achieve **fairness through unawareness** if protected attributes are not explicitly used in the prediction process.*

**Social science (SS) notion:** being “**blind**” to counter discrimination

### *Protected Attributes*

certain demographic attributes protected by law against discrimination (e.g. sex, gender, race, etc.)

# What is Fair: Fairness through Unawareness

## ❏ Critiques

- ❑ Protected attributes may be no longer blind when **additional information** is available
- ❑ Discriminatory practices have been observed following race-blind approach in SS studies

## ❏ Suitability

- ❑ problematic for domains in which protected attributes can be **deducted from easily available non-protected attributes**

### *Protected Attributes*

certain demographic attributes protected by law against discrimination (e.g. sex, gender, race, etc.)

# What is Fair: Counterfactual Measures

## □ Definition 2

A predictor  $\mathcal{H}$  is **counterfactually fair**, given protected attributes  $A = a$  and non-protected attributes  $Z = z$ , iff for all outcome  $y$  and  $a \neq a'$ ,

$$\mathbb{P}\{\mathcal{H}(A, Z) = y | A = a, Z = z\} = \mathbb{P}\{\mathcal{H}(A, Z) = y | A = a', Z = z\}$$

**SS notion: “counterfactual reasoning”**

- The outcome still remains the same even if the protected attributes were flipped



# What is Fair: Counterfactual Measures

## ❏ Critiques

- ❏ Hindsight bias & Outcome bias
- ❏ Negatively influence the process of causality

## ❏ Suitability

- ❏ problematic for domains where **the above mentioned biases are frequently observed**, e.g., health-care and judicial systems

### *Hindsight Bias*

the tendency for people to perceive events that have already occurred as having been more predictable than they actually were

### *Outcome Bias*

evaluating the quality of a decision when the outcome of that decision is already known

# What is Fair: Group Fairness

## □ Definition 3

A predictor  $\mathcal{H}: X \rightarrow Y$  achieves **group fairness** with bias  $\epsilon$  with respect to groups  $S, T \subset X$  and  $O \subseteq Y$  being any subset of outcomes iff

$$|\mathbb{P}\{\mathcal{H}(x_i) \in O | x_i \in S\} - \mathbb{P}\{\mathcal{H}(x_j) \in O | x_j \in T\}| \leq \epsilon$$

**SS notion: “collectivist egalitarianism”**

→ Affirmative Action Policies (US, India, etc.)

## Affirmative Action

the policy of promoting the education and employment of members of groups that are known to have previously suffered from discrimination

# What is Fair: Group Fairness

## ❏ Critiques

- ❏ It is not meritocratic
  - ❏ Group fairness is blind to “ground truth” → discrimination against “**qualified**” candidates
  - ❏ The predictor can select anyone **within a group** as long as it maintains statistical parity
- ❏ It reduces **efficiencies**

## ❏ Suitability

- ❏ The controversies above limits its applicability

## *Meritocracy*

certain things, like economic goods or power, should be vested in individuals on the basis of talent, effort, and achievement

# What is Fair: Individual Fairness

## □ Definition 4

A predictor achieves **individual fairness** iff

$$D(\mathcal{H}(x_i)_Y, \mathcal{H}(x_j)_Y) \approx 0 \mid d(x_i, x_j) \approx 0$$

where  $d : X \times X \rightarrow R$  is a distance metric for individuals and  $D$  is a distance measure for distributions.

**SS notion: “individualist egalitarianism”**

→ Similar outputs for similar individuals

# What is Fair: Individual Fairness

## ❏ Critiques

- ❏ How to define the similarity of individuals?
- ❏ If the distance metric **uses the protected attributes directly or indirectly**, a predictor satisfying **Definition 4** could still be discriminatory

## ❏ Suitability

- ❏ not suitable for domains where **reliable and non-discriminating distance metric** is not available

# What is Fair: Equality of Opportunity

## □ Definition 5

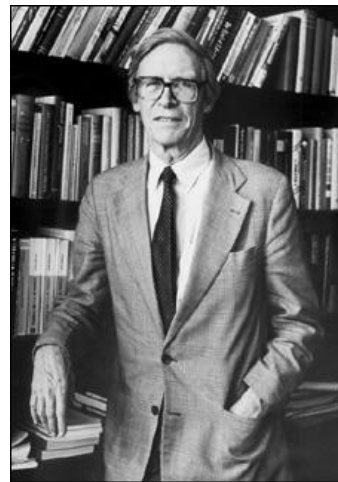
A predictor  $\mathcal{H}$  is said to satisfy **equal opportunity** with respect to group  $S \subset X$  iff (here  $y$  denotes the true label)

$$\mathbb{P}\{\mathcal{H}(x_i) = 1 | y_i = 1, x_i \in S\} = \mathbb{P}\{\mathcal{H}(x_j) = 1 | y_j = 1, x_j \in X \setminus S\}$$

**“equivalence of true positive rate across groups”**

**SS notion:** John Rawls' *A theory of Justice* (1971)

- People with “the same native talent and the same ambition” have the same prospects of success



John B. Rawls (1921 - 2002)

# What is Fair: Equality of Opportunity

## ❏ Critiques

- ❏ “**Stunted ambition**” & “**Selection by bigotry**”
- ❏ Not considering the effect of discrimination due to protected attributes which essentially affect one’s **access to opportunities** (“structural barriers”)

## ❏ Suitability

- ❏ problematic for domains in which there exists vast evidence that **protected attributes do indeed affect one’s prospects**

# What is Fair: Preference-based Fairness

## □ Definition 6

(Preferred treatment) A **group-conditional** predictor is said to satisfy **preferred treatment** if each group receives more group benefit from their respective predictor than they would have received from any other predictor i.e.

$$\mathbb{B}_S(\mathcal{H}_S) \geq \mathbb{B}_S(\mathcal{H}_T) \quad \text{for all } S, T \subset X$$

### Group Benefit

The expected proportion of individuals in the group for whom the predictor predicts the beneficial outcome.

(Alternate def: The expected proportion of individuals from the group who receive the beneficial output for whom the true label is the same.)



# What is Fair: Preference-based Fairness

## ❏ Definition 7

(Preferred impact) A predictor  $\mathcal{H}$  is said to have **preferred impact** as compared to another predictor  $\mathcal{H}'$  if  $\mathcal{H}$  offers **at-least as much benefit as**  $\mathcal{H}'$  for all the groups.

$$\mathbb{B}_S(\mathcal{H}) \geq \mathbb{B}_S(\mathcal{H}') \quad \text{for all } S \subset X$$

- ❏ Individuals in one group may prefer another outcome than the one preferred by the majority of the group.

## SS notion: “Envy-Freeness”

- It can be defined in terms of ordinal preference relations of the utility values of the predictors.

### Envy-Freeness

In an envy-free division, every agent feels that their share is at least as good as the share of any other agent, and thus no agent feels envy.

# What is Fair: Preference-based Fairness

## ❏ Critiques

- ❑ Freedom from envy is neither necessary nor sufficient for fairness.
- ❑ **“Pareto-efficiency”**
- ❑ Deciding whether there is a Pareto-efficient envy-free allocation is computationally very hard even with simple additive preferences.

## ❏ Suitability

- ❑ Limited to the domains where such an effective and envy-free allocation can be computed easily.

### *Pareto-Efficiency*

An allocation is ‘Pareto efficient’ if there is no other allocation in which some other individual is better off and no individual is worse off.

# Prospective notions of fairness: Equality of Resources *NEW!*

## ❏ Definition 1

*Unequal distribution of social benefits is only considered fair when it results from the **intentional decisions and actions** of the concerned individuals.*

- ❏ *Ambition-sensitive*: Each individual's ambitions and choices that follow them ascertains their benefits.
- ❏ *Endowment-insensitive*: Each individual's unchosen circumstances including the **natural endowments** should be offset.

# Prospective notions of fairness: Equality of Capability of Functioning <sup>NEW!</sup>

## ❑ Definition 2

*People should not be held responsible for attributes they had no say in to include personal attributes which cause difficulty in developing **functionings**.*

- ❑ In order to equalize capabilities, people should be compensated for their unequal powers to convert opportunities into functionings.
- ❑ Flexible and widely used in many ways
- ❑ Difference between resource equality and capability equality
  - ❑ **Social endowment & Natural endowment**
  - ❑ **what we can get vs what we can do**

### *Functionings*

“being and doing”:  
various states of  
existence and activities  
that an individual can  
undertake.

# Prospective notions of fairness

## ❏ Critiques

- ❏ To **Def 2**: The failure to identify of valuable capabilities
- ❏ To **Def 1 and 2**: The informational requirement of this approach can be very high
  - ❏ Difficult to make exact mathematical formalizations

## ❏ Suitability

- ❏ Makes the open problem of formalizing them worthwhile.

# Discussion and Summary

- ❑ Fair prediction cannot be addressed without considering **social issues** such as unequal access to resources and social conditioning.
- ❑ It is important to acknowledge their impact and attempt to incorporate them in **fairness formalizations**.
- ❑ **Seven existing notions** in ML society: Fairness through Unawareness, Counterfactual Measures, Group Fairness, Individual Fairness, Individual Fairness, Equality of Opportunity, and Preference-based Fairness (Preferred treatment and Preferred impact)
- ❑ **Two new notions** in ML society: Equality of resources and Equality of capability of functioning.
- ❑ Short but dense, read references to better understand concept.