Fair Ranking with Noisy Protected Attributes (NeurIPS 2022)

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1. Model of fair ranking with noisy attributes

2. Theoretical results

- 1. Ranking problem
- 2. Ranking problem with fairness constraints (Fair-ranking problem)
- 3. Ranking problem with fairness constraints, but groups are random variables (Ranking problem with noisy attributes)

Model of fair ranking with noisy attributes

1. Ranking problem

• **Ranking** : given m items, one has to select a subset of n items and output a *permutation* of the selected items. This permutation is said to be a ranking. We denote rankings by assignment matrices:

$$R \in \{0,1\}^{m \times n}$$

where $R_{ij} = 1$ indicates that item *i* appears in position *j*, and $R_{ij} = 0$ indicates otherwise.

- Utility : given an $m \times n$ matrix W, such that placing the *i*-th item at the *j*-th position generates utility W_{ij} .
- The utility of a ranking is the sum of utilities generated by each item in its assigned position. In this notation, the utility of a ranking is

$$\langle R, W \rangle := \sum_{i=1}^{m} \sum_{j=1}^{n} R_{ij} W_{ij}$$

1. Ranking problem

• Ranking problem : to solve

$$max_{R\in R} \langle R, W \rangle$$

where ${\mathcal R}$ is the set of all assignment matrices denoting a ranking:

$$\mathcal{R} := \left\{ X \in \{0,1\}^{m \times n} : \forall i \in [m], \sum_{j=1}^{n} X_{ij} \le 1, \, \forall j \in [n], \sum_{i=1}^{m} X_{ij} = 1 \right\}$$

Here, the constraint $\sum_{i=1}^{m} X_{ij} = 1$ ensures position j has exactly one item and the constraint $\sum_{j=1}^{n} X_{ij} \leq 1$ ensures that item i occupies at most one position.

• The algorithmic task in the ranking problem is to output a ranking with the highest utility.

• **Assumptions** : with $p \ge 2$ socially-salient groups

 $G_1, G_2, \dots, G_p \subseteq [m]$ (e.g., the group of all women or all Black people) which are often protected by law. Each of the *m* items belongs to *one or more* of these socially-salient groups.

• Fair-ranking problem : to output the ranking with maximum utility subject to satisfying certain fairness criteria with respect to these groups.

Definition 3.1 (Fairness constraints). Given a matrix $U \in \mathbb{Z}_{+}^{n \times p}$, a ranking R satisfies the upper bound constraint if $\sum_{i \in G_{\ell}} \sum_{j=1}^{k} R_{ij} \leq U_{k\ell}$, for all $\ell \in [p]$ and $k \in [n]$.

Definition 3.2 (Noise model). Let $P \in [0,1]^{m \times p}$ be a known matrix. The groups $G_1, \dots, G_p \subseteq [m]$ are random variables, such that, for each $i \in [m]$ and $\ell \in [p], Pr[G_\ell \ni i] = P_{i\ell}$. Moreover, for different items $i \neq j$ the events $G_\ell \ni i$ and $G_k \ni j$ are independent for all ℓ , $k \in [p]$.

3. Ranking problem with noisy attributes

Definition 3.4 ((ϵ, δ)-constraint). For any $\epsilon \in \mathbb{R}^n \ge 0$ and $\delta \in (0, 1]$, a ranking R is said to satisfy (ϵ, δ)-constraint if with probability at least $1 - \delta$ over the draw of G_1, \dots, G_p ,

$$\forall k \in [n], \forall \ell \in [p], \sum_{i \in G_{\ell}} \sum_{j=1}^{k} R_{ij} \leq U_{k\ell}(1 + \epsilon_k).$$

Problem 3.5 (Ranking problem with noisy attributes). Given matrices *P*, *U*, and *W*, find the ranking *R* such that, for some small ϵ and δ ,

 $\max_{R \in \mathcal{R}} \langle R, W \rangle$ s.t. *R* satisfies (ϵ, δ) - constraint.

• Note that solving **Problem 3.5** is NP-hard.

Theoretical results

Optimization framework

- Input: Matrices $P \in [0,1]^{m \times p}$, $W \in \mathbb{R}_{>0}^{m \times n}$, $U \in \mathbb{R}^{n \times p}$
- **Parameters**: Constant c > 1, failure probability $\delta \in (0, 1]$, and $k \in [n]$, relaxation parameter

$$\gamma_k := 12 \cdot \log\left(\frac{2np}{\delta}\right) \cdot \max_{\ell \in [p]} \sqrt{\frac{1}{U_{k\ell}}} \tag{1}$$

• Our Fair-Ranking Program (2) :

 $\max_{R \in \mathcal{R}} \langle R, W \rangle, \quad \text{(Noise Resilient)}$ s.t. $\forall \ell \in [p] \quad \forall k \in [n]$ $\sum_{\substack{i \in [m], \\ j \in [k]}} P_{i\ell} R_{ij} \leq U_{k\ell} \left(1 + \left(1 - \frac{1}{2\sqrt{c}} \right) \gamma_k \right).$ **Theorem 4.1** Let $\gamma \in \mathbb{R}^n$ be as defined in Equation (1). There is an optimization program (Program (2)), parameterized by a constant c and failure probability δ , such that for any c > 1 and $\delta \in (0, \frac{1}{2}]$ its optimal solution satisfies $(c\gamma, \delta)$ -constraint and has a utility at least as large as the utility of any ranking satisfying $((c - \sqrt{c})\gamma, \delta)$ -constraint.

• Note that solving Program (2) is a polynomial complexity.

Theorem 4.2 There is a family of matrices $U \in \mathbb{Z}_{+}^{n \times p}$ such that for any U in the family and any parameters $\delta \in [0, 1)$ and $\epsilon_1, \dots, \epsilon_n \geq 0$, if for any position $k \in [n], \epsilon_k \leq 1$ and $\epsilon_k < \max_{\ell \in [p]} \sqrt{\frac{1}{2U_{k\ell}} \log \frac{1}{4\delta}}$ then there exists a matrix $P \in [0, 1]^{m \times p}$, such that it is information theoretically impossible to output a ranking that satisfies (ϵ, δ) -constraint.

 Since γ_k is O (log (np/δ) · max_ℓ √ 1/U_{kℓ}), Theorem 4.2 shows that Theorem 4.1's fairness guarantee is optimal up to log-factors.