FairDen: Fair Density-based Clustering

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International Conference on Learning Representations 2025

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Group-Level Fairness: Non-binary groups

Gender







Age









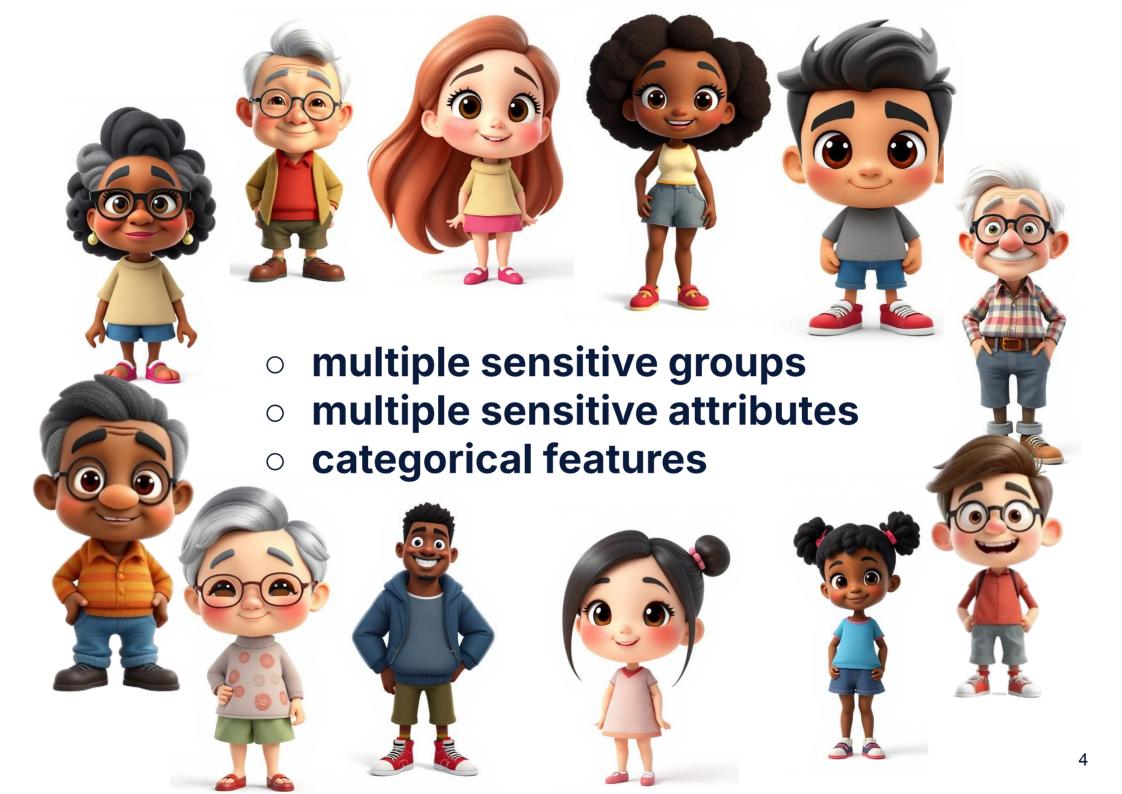
Group-Level Fairness: Multiple sensitive attributes



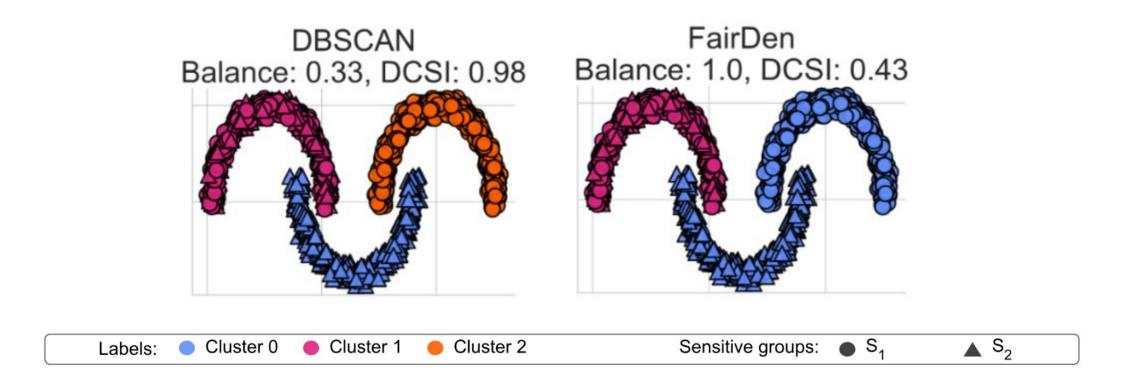








Group-level fairness for density-based clusters



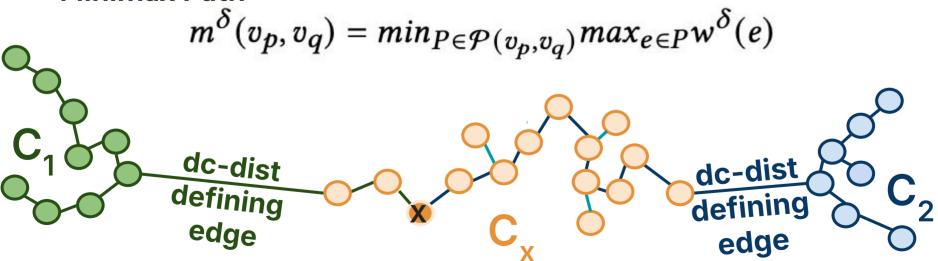
Use the density-connectivity distance [1] to capture structure.

$$d_{dc}^{\mu}(p,q) = m^{d_r^{\mu}}(v_p,v_q) = min_{P \in \mathcal{P}(v_p,v_q)} max_{e \in P} w^{d_r^{\mu}}(e)$$

Mutual Reachability distance

$$d_r^{\mu}(p,q) = max(d_E(p,q), d_{core}^{\mu}(p), d_{core}^{\mu}(q))$$

Minimax Path



Construct fairness matrix [2] to ensure balance.

- Binary encoding for each *sensitive group:*

$$\triangle \bigcirc \bigcirc \triangle \\
 \triangle (1 0 0 1) \\
 \bigcirc (0 1 1 0)$$

- Columns of fairness matrix:

$$\triangle$$
 (1 0 0 1) - $\frac{|\triangle|}{n}$ · 1_n

Combining Density-Connectivity and Fairness leads to a problem we can solve with spectral clustering.

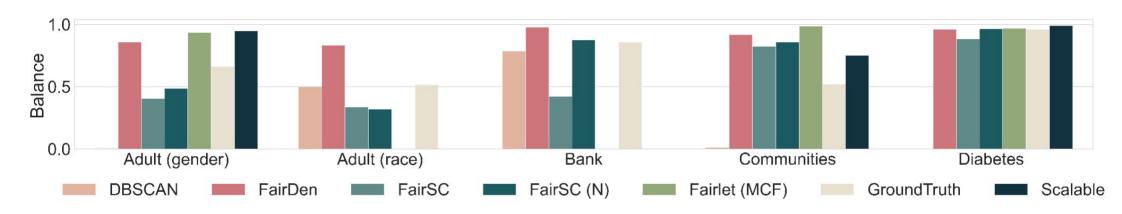
Density-connectivity objective

$$\min_{\mathcal{H} \in \mathbb{R}^{n imes k}} Tr(\mathcal{H}^{ op} \mathcal{L} \mathcal{H})$$
 $ext{subject to } \mathcal{H}^{ op} \mathcal{D} \mathcal{H} = \mathcal{I}_k$

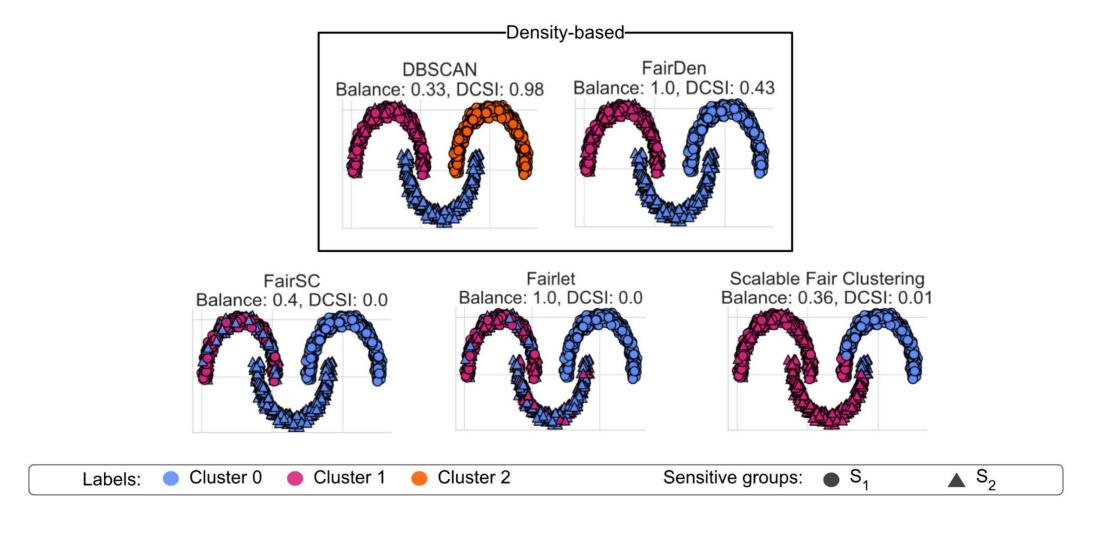
Fairness Constraint

and
$$\mathcal{F}^{\top}\mathcal{H}=0$$

FairDen achieves balanced clusterings w.r.t. sensitive attributes.



FairDen detects density-based clusters better than other fair methods.



FairDen

Fair Density-based Clustering

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FairDen is the first density-based group-level fair clustering algorithm.

It can even cluster data with:

- multiple sensitive groups
- multiple sensitive attributes
- categorical features

Code



Paper



References

- [1] Anna Beer, Andrew Draganov, Ellen Hohma, Philipp Jahn, Christian MM Frey, and Ira Assent. "Connecting the Dots--Density-Connectivity Distance unifies DBSCAN, k-Center and Spectral Clustering." In *Proceedings of the 29th ACM SIGKDD Conference on Knowledge Discovery and Data Mining*, pp. 80-92. 2023.
- [2] Matthäus Kleindessner, Samira Samadi, Pranjal Awasthi, and Jamie Morgenstern. "Guarantees for spectral clustering with fairness constraints." In *International conference on machine learning*, pp. 3458-3467. PMLR, 2019.

Pictures of people generated with https://deepai.org/machine-learning-model/3d-cartoon-generator









