

FAST AND ACCURATE FAIR k -CENTER CLUSTERING IN DOUBLING METRICS

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Will I get
the loan?



Am I getting
hired?



What about
my diagnosis?



Problem definition

Disparate impact

People in different protected classes should not experience disproportionately different outcomes.

Problem definition

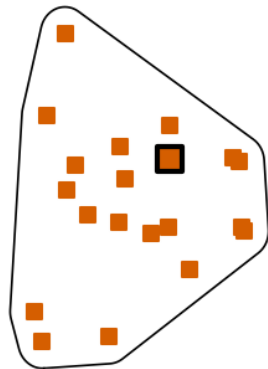
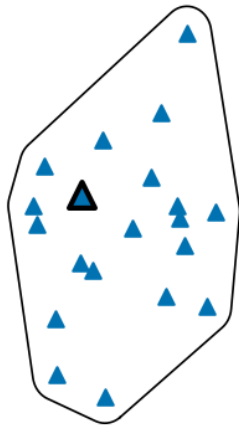
Disparate impact

People in different protected classes should not experience disproportionately different outcomes.

Unawareness does not help

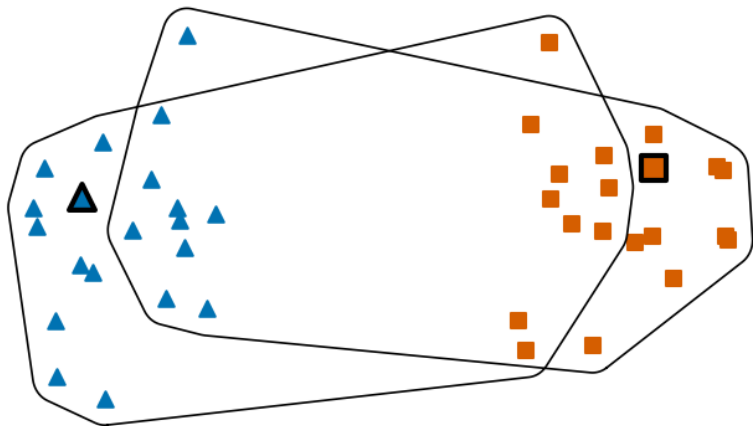
Blindly ignoring protected attributes is no solution: correlated features (e.g. height which correlates with biological sex) can leak information about the protected attributes and thus lead to *unfair* solutions.

Problem definition



Classic k -center assigns each point to the closest center.

Problem definition



If we want to balance the colors in each cluster, we possibly have to assign points to farther away centers.

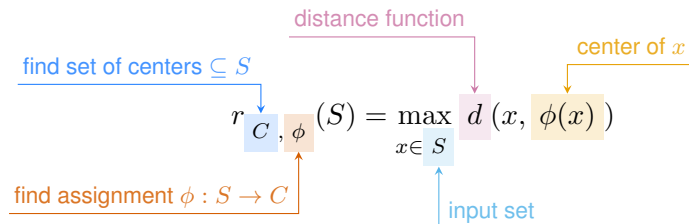
Problem definition

- ▶ Metric space (\mathcal{X}, d)
- ▶ Set of points S
- ▶ Each point has one (or more) colors out of a set Γ
- ▶ Parameter k

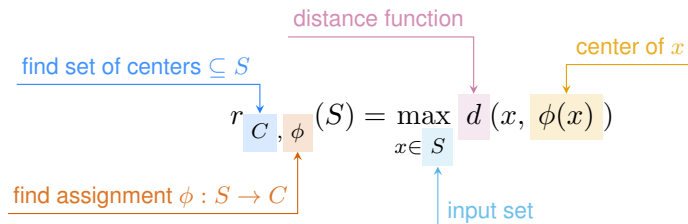
Goal

Build a clustering such that the *proportion* of points from each protected group is the same as the proportion in the entire dataset.

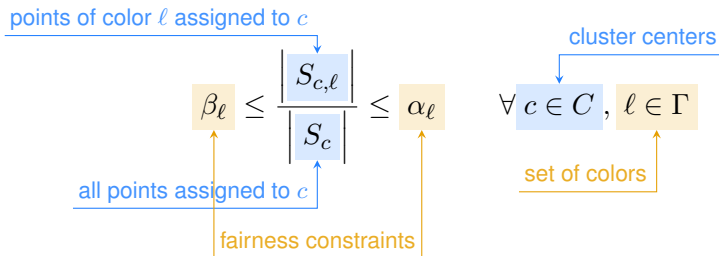
Problem definition



Problem definition



Minimize the above, where the assignment is subject to



State of the art

- ▶ Find a set of k centers, ignoring fairness
- ▶ Build the assignment function by means of linear programming, imposing fairness

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- ▶ 3 approximation [Ber+19; HL20]

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- ▶ 9 approximation in MapReduce, $7 + \varepsilon$ in Streaming [Ber+22]

State of the art

- ▶ Find a set of k centers, ignoring fairness
- ▶ Build the assignment function by means of linear programming, imposing fairness

- ▶ 3 approximation [Ber+19; HL20]
- ▶ 9 approximation in MapReduce,
 $7 + \varepsilon$ in Streaming [Ber+22]
- ▶ large linear program of size $O(k \cdot n)$

Our contribution

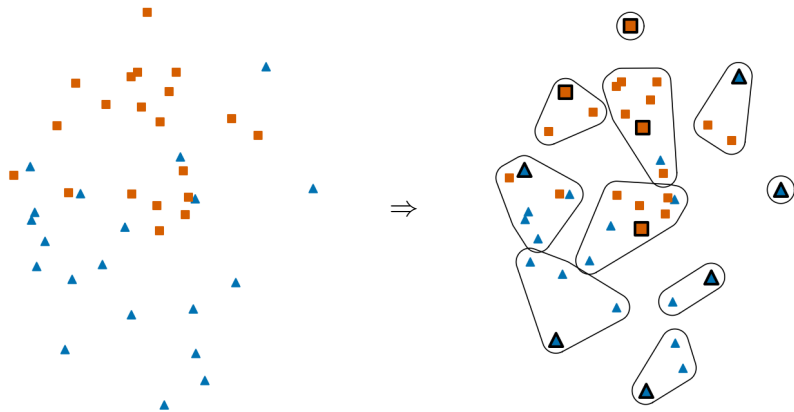
$3 + \varepsilon$ approximation algorithms

- ▶ **Sequential:** Linear time in the input size
- ▶ **Streaming:** 2 passes and memory $O\left(\log \frac{d_{max}}{d_{min}}\right)$
- ▶ **MapReduce:** 5 rounds and memory $O(\max\{|S|/p, p\})$, where p is the number of processors

Coreset

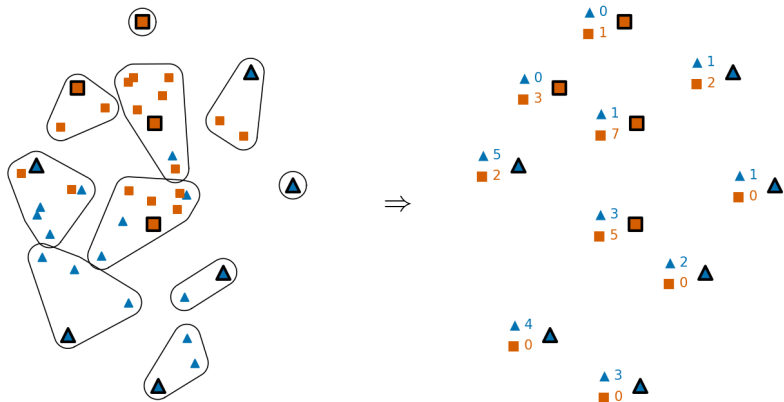
- ▶ Set $T \subseteq S$
- ▶ $|T| \ll |S|$
- ▶ Proxy function $\pi : S \rightarrow T$
- ▶ Weight function $w : T \rightarrow \mathbb{N}$

Locate a set of *proxy points*



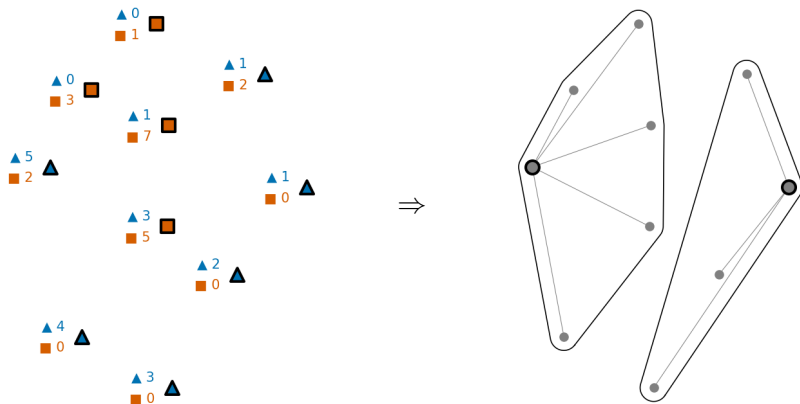
Goal: find a good compact representation of the input

Assign weights to coreset points



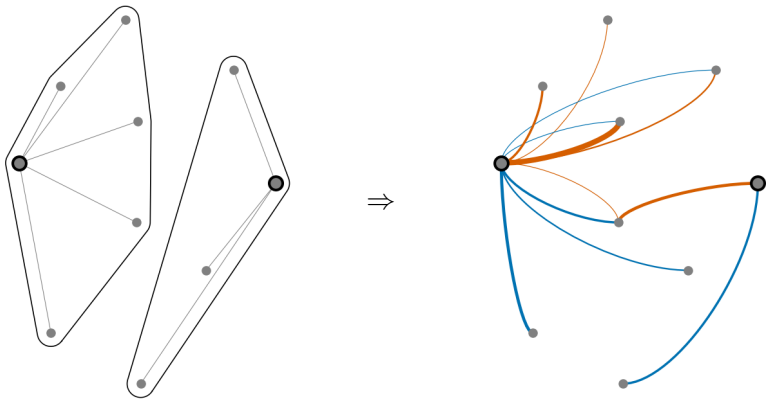
Goal: enable addressing fairness later

Find an unfair k -center clustering on the coresets



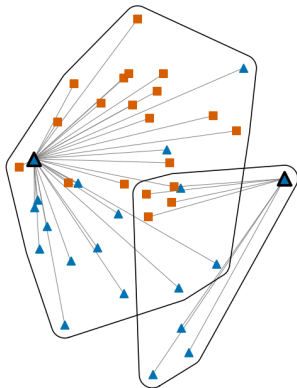
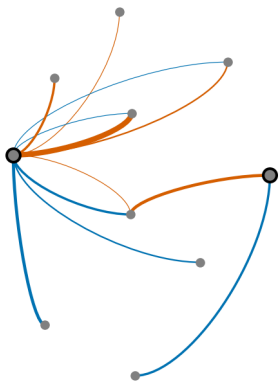
Goal: optimize the placement of centers

Distribute weight to centers



Goal: address fairness

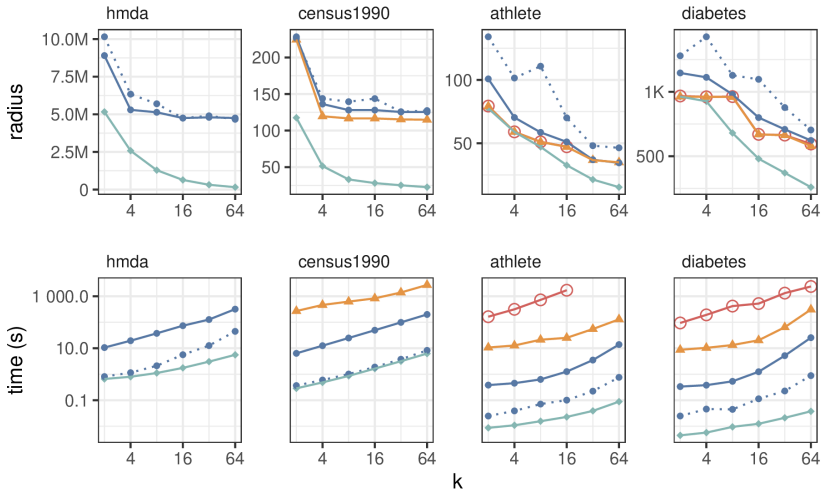
Assign original points



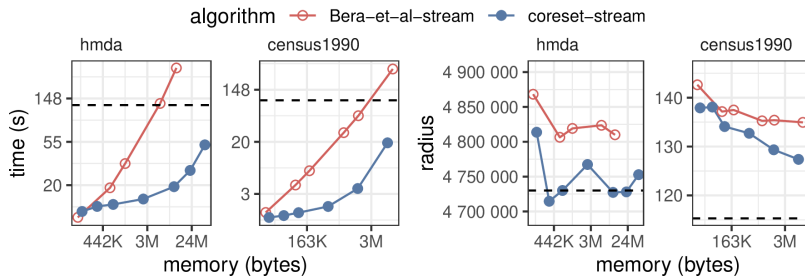
Goal: compute the solution

Experiments (sequential)

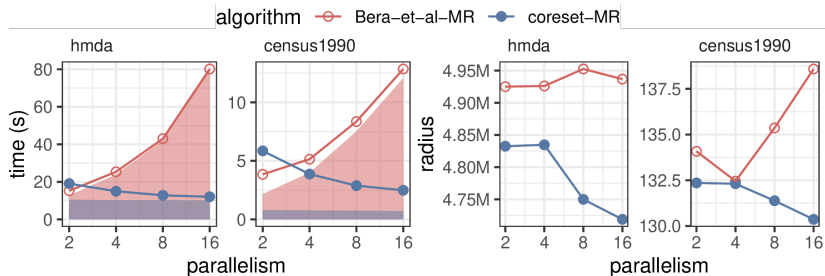
algorithm ○ Bera-et-al ⋯ coresets (1k) — coresets (32k) ▲ KFC — unfair



Experiments (Streaming)



Experiments (MapReduce)



Thank you!

Appendix

Problem definition

Additive violation

The additive violation of an assignment, w.r.t. the fairness constraints α_ℓ, β_ℓ , is the minimum \mathcal{E} s.t. $\forall c \in C, \ell \in \Gamma$

constraints on the number of points with color ℓ in C

The diagram shows the inequality $\beta_\ell |S_c| - \mathcal{E} \leq |S_{c,\ell}| \leq \alpha_\ell |S_c| + \mathcal{E}$. The terms $\beta_\ell |S_c|$ and $\alpha_\ell |S_c|$ are enclosed in light orange boxes. The terms $-\mathcal{E}$ and $+\mathcal{E}$ are enclosed in light blue boxes. An orange bracket above the equation connects the two orange boxes, with the text "constraints on the number of points with color ℓ in C " above it. A blue bracket below the equation connects the two blue boxes, with the word "violation" below it.

$$\beta_\ell |S_c| - \mathcal{E} \leq |S_{c,\ell}| \leq \alpha_\ell |S_c| + \mathcal{E}$$

Our algorithm provides an additive violation of $4\Delta + 3$, where Δ is the maximum number of colors per point.

Preliminaries: GMM

Classic algorithm for *unfair* k -center.

Input: Set S , parameter k

$C \leftarrow \{\text{arbitrary point from } S\};$

while $|C| < k$ **do**

$c \leftarrow \arg \max_{x \in S} d(x, C);$
 $C \leftarrow C \cup \{c\};$

return $C;$

Provides a 2-approximation in time $O(k \cdot n)$

Starting point [Ber+19; HL20]

Algorithm

1. Find a set C of centers with the GMM algorithm
2. Do a binary search on guesses over the possible clustering radii
3. Instantiate a linear program and find a fractional solution (if any)
4. If the linear program has a feasible solution, then iteratively round it (will not see today).

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1. Find a set C of centers with the GMM algorithm
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Guarantees

- ▶ The algorithm provides a 3 approximation to the radius
- ▶ The fairness constraints have an additive violation up to 7

Outline of our approach

1. Build a *weighted* coreset T out of the input set S
 2. Compute a fair k -center clustering on T , whose centers are C
 3. Build a fair assignment of S to C by using information from the solution on the coreset
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1. Build a *weighted* coreset T out of the input set S
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3. Build a fair assignment of S to C by using information from the solution on the coreset

In MapReduce, steps 1. and 3.
are carried out in a single
parallel round each

In Streaming, steps 1. and 3.
require each a pass on the data

Sequential coresets construction

Input: Set S , parameter k , parameter ε

$T \leftarrow \{\text{arbitrary point from } S\};$

while $|T| < k$ **do** $T \leftarrow T \cup \{\arg \max_{x \in S} d(x, T)\} ;$

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$r_k \leftarrow \max_{x \in S} d(x, T);$

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while $\max_{x \in S} d(x, T) > \frac{\varepsilon}{6} \cdot r_k$ **do**
 $T \leftarrow T \cup \{\arg \max_{x \in S} d(x, T)\}$

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for $t \in T$ **do**

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for $x \in S$ **do**

$t' \leftarrow \arg \min_{t \in T: \text{col}(t) = \text{col}(x)} d(x, t);$
 $w(t') \leftarrow w(t') + 1;$
 $\pi(x) \leftarrow t';$

return $T, w, \pi;$

Properties of the coresets

Proxy radius

Let T be a coreset on S constructed as above, and let π be its proxy function. Then

$$d(x, \pi(x)) \leq \frac{\varepsilon}{3} OPT_{unf} \leq \frac{\varepsilon}{3} OPT_{fair}$$

Size

If S belongs to a metric space with doubling dimension D , then

$$|T| \leq |\Gamma| \cdot k \cdot \left(\frac{12}{\varepsilon}\right)^D$$

Properties of the cores set

Proxy radius

Let T be a cores set on S constructed as above, and let π be its proxy function. Then

$$d(x, \pi(x)) \leq \frac{\varepsilon}{3} OPT_{unf} \leq \frac{\varepsilon}{3} OPT_{fair}$$

Size

If S belongs to a metric space with doubling dimension D , then

$$|T| \leq \underbrace{|\Gamma|}_{\text{One copy per color}} \cdot \underbrace{k}_{\text{Clusters}} \cdot \underbrace{\left(\frac{12}{\varepsilon}\right)^D}_{\text{Balls covering each } k\text{-cluster}}$$

A revised linear program, on the coresot

Let $C \subseteq T$ be a set of centers found by GMM *on the coresot* and a radius guess R

How much of the weight of t is assigned to c

$$\begin{aligned} z_{t,c} &\geq 0 && \text{Assign all the weight} && \begin{matrix} t \in T \\ c \in C \end{matrix} \text{ if } d(t, c) \leq R \\ \sum_{c \in C} z_{t,c} &= w(t) && && \forall t \in T \\ \beta_\ell \sum_{t \in T} z_{t,c} &\leq \sum_{t' \in T_\ell} z_{t',c} \leq \alpha_\ell \sum_{t \in T} z_{t,c} && && \forall c \in C, \ell \in \Gamma \end{aligned}$$

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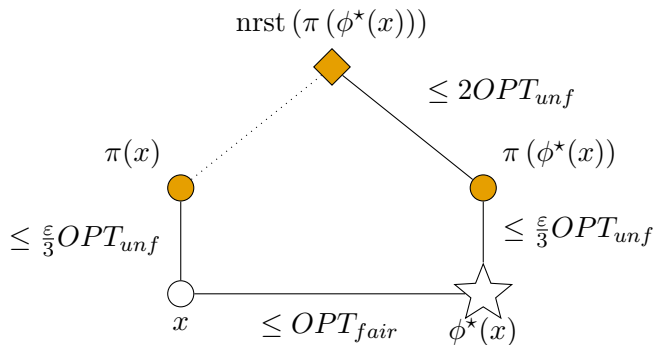
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Which radius guess R allows for a feasible solution?

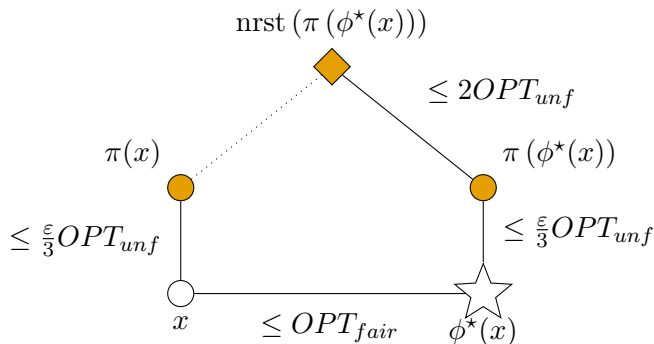
Finding the radius guess



Finding the radius guess



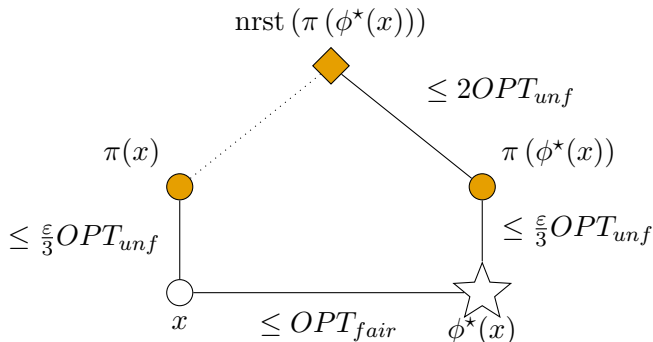
Finding the radius guess



By the triangle inequality, we have that

$$d(\pi(x), \text{nrst}(\pi(\phi^*(x)))) \leq \frac{2\varepsilon}{3}OPT_{fair}$$

Finding the radius guess



Is the assignment fair?

By a charging argument, we can build an assignment of coreset points to centers that respects the fairness constraints.

Summary

- ▶ Set $T \subseteq S$
- ▶ Proxy function $\pi : S \rightarrow T$
- ▶ Weight function $w : T \rightarrow \mathbb{N}$
- ▶ Weight assignment $\hat{\phi}(t, c)$, for $t \in T$ and $c \in C \subseteq T$ such that

$$\hat{\phi}(t, c) > 0 \quad \Rightarrow \quad d(t, c) \leq \left(3 + \frac{2}{3}\varepsilon\right) OPT_{fair}$$

Building the final assignment

Input: The weight distribution $\hat{\phi}(t, c)$, the proxy function $\pi(\cdot)$, the set S , the coreset T , the coreset centers C

for $x \in S$ **do**

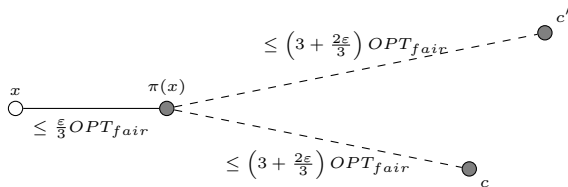
$t \leftarrow \pi(x)$;

$c \leftarrow \text{arbitrary } c \in C : \hat{\phi}(t, c) > 0$;

$\phi(x) \leftarrow c$;

$\hat{\phi}(t, c) \leftarrow \hat{\phi}(t, c) - 1$;

return C, ϕ



Summary

Approximation

$$3 + \varepsilon$$

Linear program size

$$\min\{2^{k-1}k|\Gamma|, k \cdot |\Gamma| \cdot k \cdot \left(\frac{12}{\varepsilon}\right)^D\}$$

State of the art had n here

