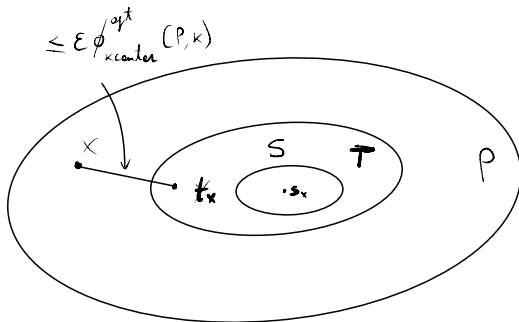


# Coreset Technique

(Part 1 - Exercises)

## Exercise

Let  $P$  be a set of  $N$  points in a metric space  $(M, d)$ , and let  $T \subseteq P$  be a coreset of  $|T| > k$  points such that for each  $x \in P$  we have  $d(x, T) \leq \epsilon \phi_{\text{kcenter}}^{\text{opt}}(P, k)$ , for some  $\epsilon \in (0, 1)$ . Let  $S$  be the set of  $k$  centers obtained by running the Farthest-First Traversal algorithm on  $T$ . Prove an upper bound to  $\phi_{\text{kcenter}}(P, S)$  as a function of  $\epsilon$  and  $\phi_{\text{kcenter}}^{\text{opt}}(P, k)$ .



$$\phi_{\text{kcenter}}(P, S) = f(\epsilon) \cdot \phi_{\text{kcenter}}^{\text{opt}}(P, k)$$

$$\forall x \in P: d(x, S) \leq f(\epsilon) \phi_{k_{\text{center}}}^{\text{opt}}(P, k)$$

let  $t_x$  be the closest point to  $x$  in  $T$

let  $s_x$  be the closest point to  $t_x$  in  $S$

$$\begin{aligned}
 d(x, S) &\stackrel{\text{def } d}{\leq} d(x, s_x) \stackrel{\text{triangular inequality}}{\leq} d(x, t_x) + d(t_x, s_x) \leq \\
 &\leq \epsilon \phi_{k_{\text{center}}}^{\text{opt}}(P, k) + d(t_x, s_x) \\
 &\leq \epsilon \phi_{k_{\text{center}}}^{\text{opt}}(P, k) + 2 \phi_{k_{\text{center}}}^{\text{opt}}(P, k) = (2 + \epsilon) \phi_{k_c}^{\text{opt}}(P, k) \\
 &\quad \downarrow \\
 &\quad 2 + \epsilon \text{ - approximation}
 \end{aligned}$$

We now have to prove that  $\forall t \exists s' \in S$  s.t.

$$d(x, s') \leq 2 \phi_{k_c}^{\text{opt}}(P, k)$$

$S = \{s_1, \dots, s_k\} \rightarrow$  point  $s_i$  found at iteration  $i$  of the FFT

Run the FFT once more:  $q = s_{k+1} \rightarrow \hat{S} = S \cup \{q\} = \{s_1, s_2, \dots, s_k, s_{k+1}\}$

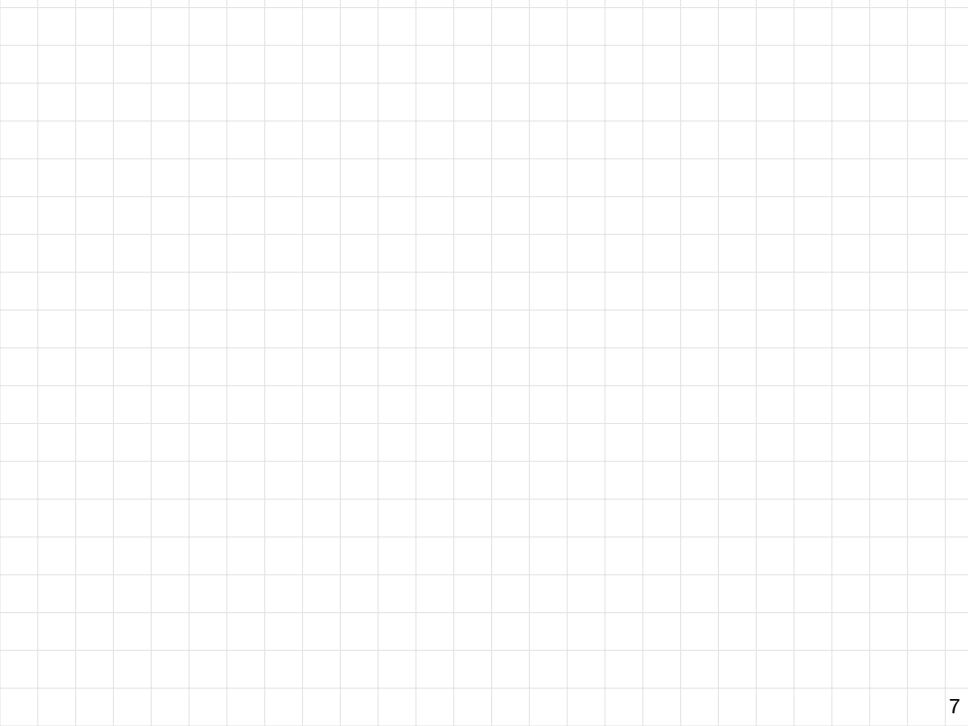
Then  $\exists s_i, s_j$  that are in the same cluster  $C_\ell^*$

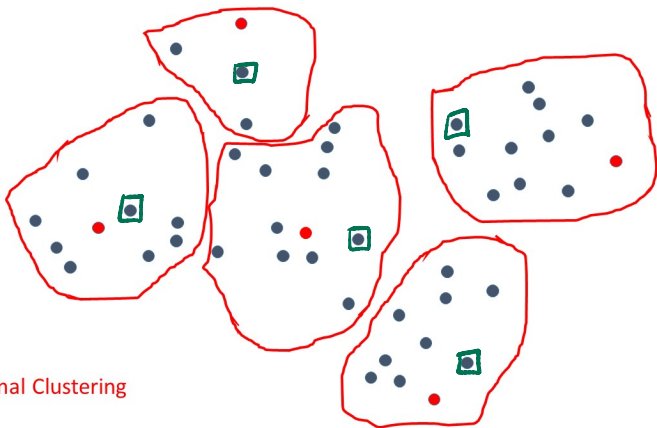
$$\begin{aligned}
 d(s_i, s_j) &\leq d(s_i, c_\ell^*) + d(c_\ell^*, s_j) \\
 &\stackrel{\text{Optimality of the cluster}}{\leq} \phi_{k_c}^{\text{opt}}(P, k) + \phi_{k_c}^{\text{opt}}(P, k) \\
 &\leq 2 \phi_{k_c}^{\text{opt}}(P, k)
 \end{aligned}$$

$$\begin{aligned}
 \forall x \in T, d(x, S) &= d(x, \{s_1, \dots, s_k\}) \leq d(x, \{s_1, \dots, s_{j-1}\}) \leq d(s_j, \{s_1, \dots, s_{j-1}\}) \\
 &\leq d(s_j, \{s_i\}) \leq 2 \phi_{k_c}^{\text{opt}}(P, k)
 \end{aligned}$$

## Exercise

Let  $P$  be a set of points in a metric space  $(M, d)$ , and let  $T \subseteq P$ . For any  $k < |T|, |P|$ , show that  $\Phi_{\text{kcenter}}^{\text{opt}}(T, k) \leq 2\Phi_{\text{kcenter}}^{\text{opt}}(P, k)$ . Is the bound tight?





Optimal Clustering

for each cluster take 1 point  $\rightarrow T'$

$$\phi^{\text{opt}}(\tau, k) \leq \phi(\tau, \hat{\tau})$$

↳ not generated  
by taking  $k$  points



1)  $C^* = \{c_1^* \dots c_k^*\}$  opt sol in  $P$  with centers  $\hat{c}_1^*, \dots, \hat{c}_k^*$

2) Partition  $P$  using  $C^*$

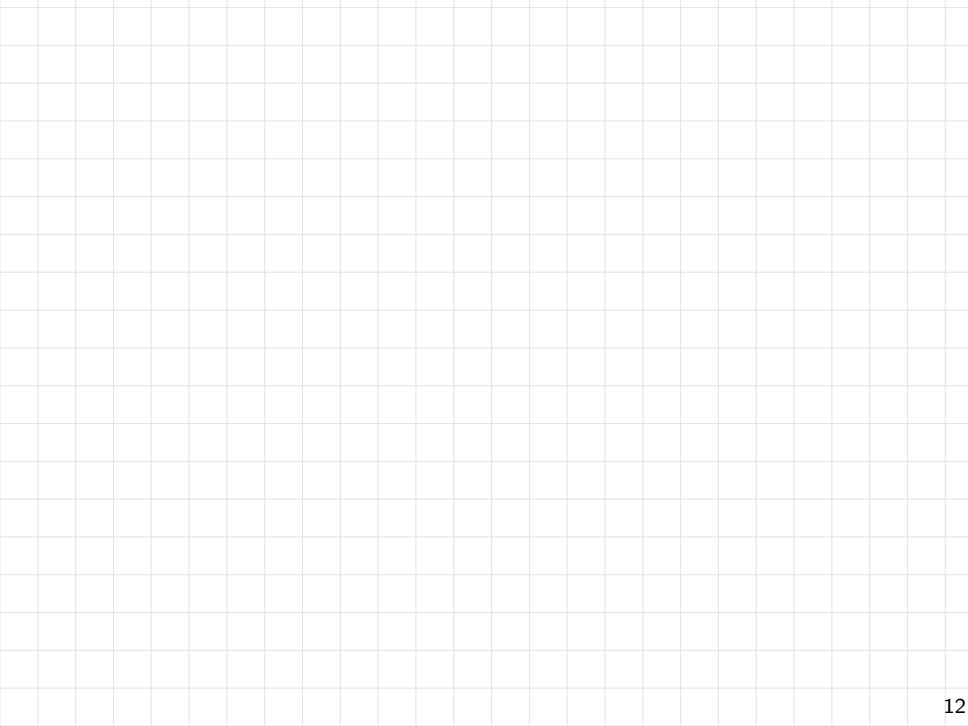
3) Remove all points not in  $T$

4)  $\forall c_i^*$  take one point in  $T \rightarrow \hat{t}_i$

5)  $\hat{T} = \{\hat{t}_1, \dots, \hat{t}_k\}$

}

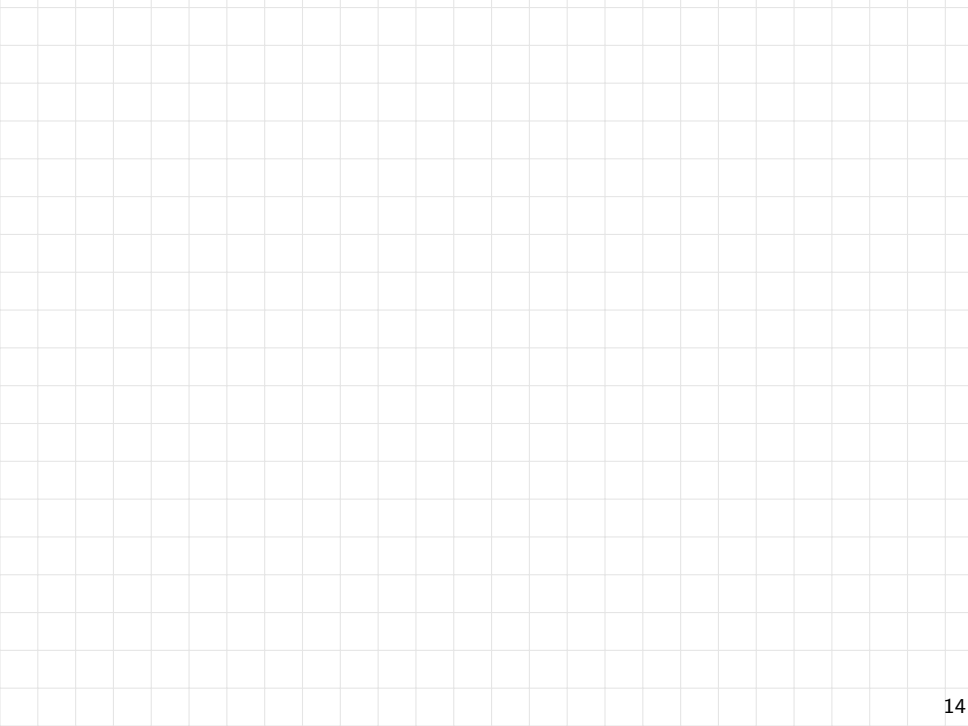


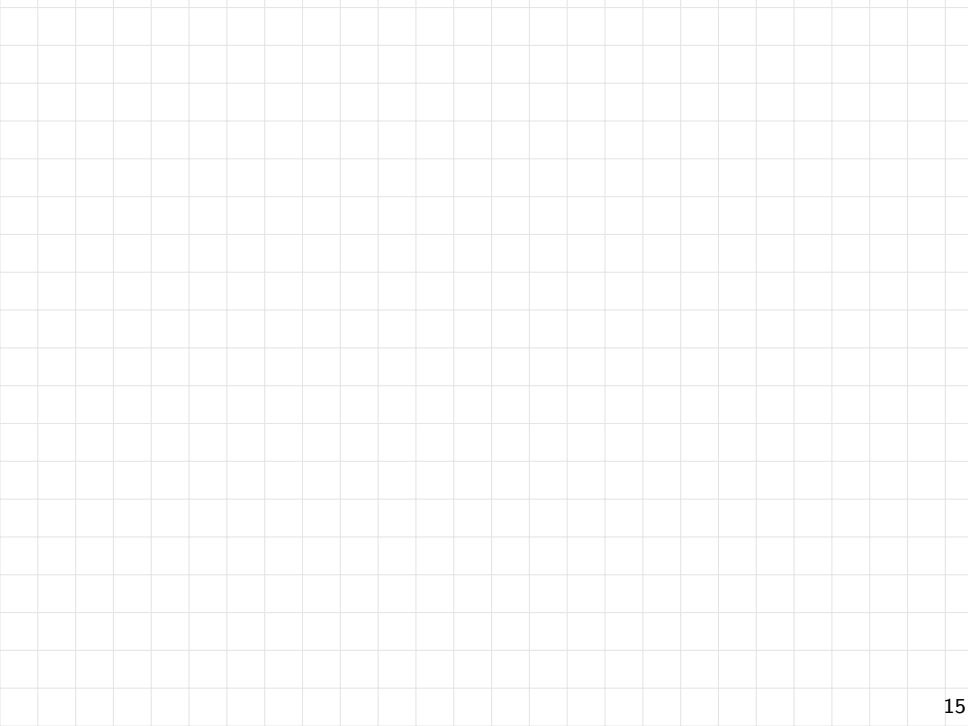


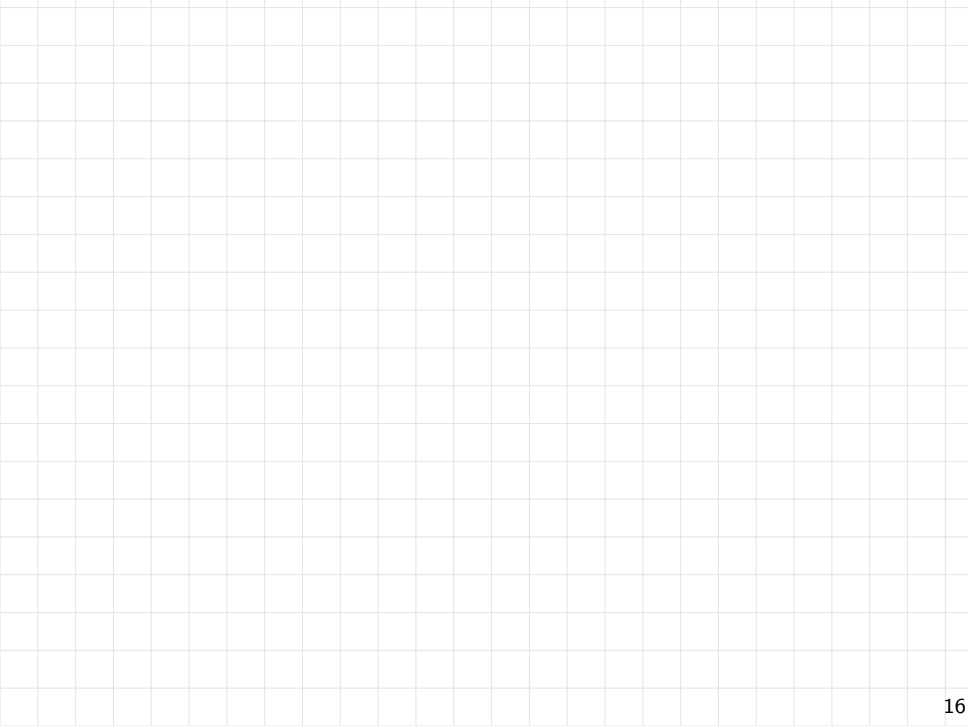
## Exercise

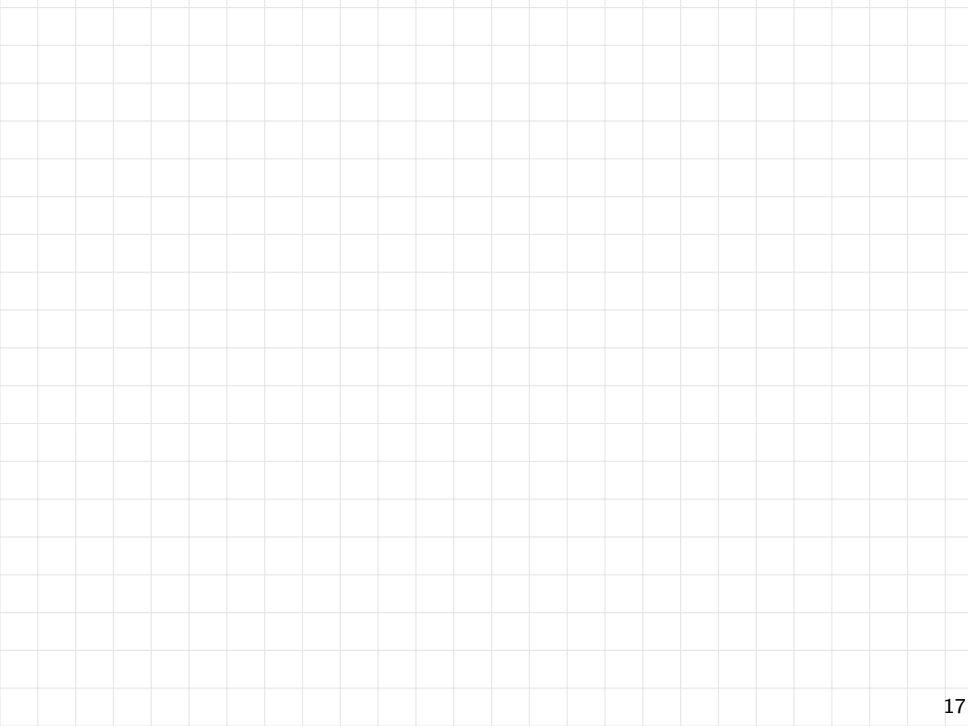
Let  $P$  be a set of  $N$  points in a metric space  $(M, d)$ , and let  $\mathcal{C} = (C_1, C_2, \dots, C_k; c_1, c_2, \dots, c_k)$  be a  $k$ -clustering of  $P$ . Initially, each point  $q \in P$  is represented by a pair  $(\text{ID}(q), (q, c(q)))$ , where  $\text{ID}(q)$  is a distinct key in  $[0, N - 1]$  and  $c(q) \in \{c_1, \dots, c_k\}$  is the center of the cluster of  $q$ .

- 1 Design a 2-round MapReduce algorithm that for each cluster center  $c_i$  determines the most distant point among those belonging to the cluster  $C_i$  (ties can be broken arbitrarily).
- 2 Analyze the local and aggregate space required by your algorithm. Your algorithm must require  $o(N)$  local space and  $O(N)$  aggregate space.









## Exercise

Let  $P$  be a set of  $N$  *bicolored points* from a metric space, partitioned into  $k$  clusters  $C_1, C_2, \dots, C_k$ . Each point  $x \in P$  is initially represented by the key-value pair  $(ID_x, (x, i_x, \gamma_x))$ , where  $ID_x$  is a distinct key in  $[0, N-1]$ ,  $i_x$  is the index of the cluster which  $x$  belongs to, and  $\gamma_x \in \{0, 1\}$  is the color of  $x$ .

- 1 Design a 2-round MapReduce algorithm that for each cluster  $C_i$  checks whether all points of  $C_i$  have the same color. The output of the algorithm must be the  $k$  pairs  $(i, b_i)$ , with  $1 \leq i \leq k$ , where  $b_i = -1$  if  $C_i$  contains points of different colors, otherwise  $b_i$  is the color common to all points of  $C_i$ .
- 2 Analyze the local and aggregate space required by your algorithm. Your algorithm must require  $o(N)$  local space and  $O(N)$  aggregate space.

