# A Simple Algorithm for r-gather-clusterings on the Line 

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#### Abstract

In this paper we study a recently proposed two variants of the facility location problem, called the $r$-gatherclustering problem and the $r$-gathering problem. Given a set $C$ of $n$ points on the plane an $r$-gather-clustering is a partition of the points into clusters such that each cluster has at least $r$ points. The $r$-gather-clustering problem finds the $r$-gather-clustering minimizing the maximum radius among the clusters, where the radius of a cluster is the minimum radius of the disk which can cover the points in the cluster. A polynomial time 2-approximation algorithm for the problem is known. When all $C$ are on the line, an $O(n \log n)$ time algorithm, based on the matrix search method, to find an $r$-gatherclustering is known. In this paper we give an $O\left(n \log ^{*} n\right)$ time algorithm to solve the problem. We also give an algorithm to solve a similar problem, called the $r$-gathering problem.


## 1. Introduction

The facility location problem and many of its variants are studied[5], [6].

In this paper we study recently proposed two variants of the problem, called the $r$-gather-clustering problem and the $r$ gathering problem [1], [4].

Given a set $C$ of $n$ points on the plane an $r$-gather-clustering is a partition of the points into clusters such that each cluster has at least $r$ points. The cost of an $r$-gather-clustering is the maximum radius among the clusters, where the radius of a cluster is the minimum radius of the disk which can cover the points in the cluster. The $r$-gather-clustering problem [1] is the problem to find the $r$-gather-clustering minimizing the cost. The problem is NPcomplete in general, however a polynomial time 2 -approximation algorithm for the problem is known[1]. When all $C$ are on the line, an $O(n \log n)$ time algorithm, based on the matrix search method[2], [7], for the problem is known[3].
In this paper we give an $O\left(n \log ^{*} n\right)$ time algorithm to solve the problem, by reducing the problem to the min-max path problem[9] in a weighted directed graph.

Assume that $C$ is a set of residents and we wish to locate emergency shelters for the residents so that each shelter serves $r$ or more residents. Then $r$-gather clustering problem computes optimal locations for shelters which minimizung the evacuation time span, where each shelter for a cluster is located at the center of the cluster.
In this paper we consider one more similar problem. Given sets $C$ and $F$ of points on the plane an $r$-gathering of $C$ to $F$ is an assignment $A$ of $C$ to open facilities $F^{\prime} \subset F$ such that $r$ or more customers are assigned to each open facility.

[^0]The cost of an $r$-gathering is the maximum distance $d(c, f)$ between $c \in C$ and $A(c) \in F^{\prime}$ among the assignment, which is $\max _{c \in C, A(c) \in F^{\prime}}\{d(c, A(c))\}$.

Assume that $F$ is a set of possible locations for emergency shelters, and $d(c, f)$ is the time needed for a person $c \in C$ to reach a shelter $f \in F$. Then an $r$-gathering corresponds to an evacuation assignment such that each opened shelter serves $r$ or more people, and the $r$-gathering problem finds an evacuation plan minimizing the evacuation time span.

Armon[4] gave a simple 3-approximation algorithm for the $r$-gathering problem and proves that with the assumption $P \neq$ $N P$ the problem cannot be approximated within a factor of less than 3 for any $r \geq 3$. When all $C$ and $F$ are on the line an $O((|C|+|F|) \log (|C|+|F|))$ time algorithm[3] and an $O(|C|+$ $\left.|F| \log ^{2} r+|F| \log |F|\right)$ time algorithm[10] to solve the $r$-gathering problem are known.
In this paper we give an $O\left(|C|+r^{2}|F| \log ^{*}|C|\right)$ time algorithm to solve the problem, where $\log ^{*}|C|$ is the number of times the $\log$ must be iteratively applied before results in less than 1 . Since in typical case $r \ll|F| \ll|C|$ holds our new algorithm is faster than the known algorithms.

The remainder of this paper is organized as follows. Section 2 gives an algorithm for the $r$-gather-clustering problem. Section 3 gives an algorithm for the $r$-gathering problem. Finally Section 4 is a conclusion.

## 2. r-gather-clustering on the line

In this section we give an algorithm for the $r$-gather-clustering problem when all points in $C$ are on the line. Let $C=$ $\left\{c_{1}, c_{2}, \cdots, c_{n}\right\}$ be points on the horizontal line and we assume they are sorted from left to right. Our idea is to reduce the $r$-gather-clustering problem to the mix-max path problem in a weighted directed (acyclic) graph[9]. First we have the follow-


Fig. 1 the weighted directed path $D$.

(a)

(b)

Fig. 2 (a)an $r$-gather clustering (b)its corresponding min-max path of $D$.
ing two lemmas.
Lemma 2.1 One can assume the points in each cluster in a solution are consecutive.
Proof. Otherwise repeat swapping some points between the clusters until the condition holds, which never increase the cost.
Q.E.D.

Lemma 2.2 One can assume the number of points in each cluster in a solution is at most $2 r-1$.
Proof. Otherwise devide such clusters into two (or more) clusters, respectively, which never increase the cost.
Q.E.D.

Then we difine the directed (acyclic) graph $D(V, E)$ and the weight of each edge, as follows.

$$
\begin{gathered}
V=\left\{p_{0}, p_{1}, p_{2}, \cdots, p_{n}\right\} \\
E=\left\{\left(p_{i}, p_{j}\right) \mid i+r \leq j \leq i+2 r-1\right\}
\end{gathered}
$$

See Fig. 1. Note that the number of edges is at most $r n$. The weight $w$ of an edge $w\left(p_{i}, p_{j}\right)$ is the half of the distance between $c_{i+1}$ and $c_{j}$, and denoted by $w\left(p_{i}, p_{j}\right)$.

The cost of a directed path from $p_{0}$ to $p_{n}$ is defined by the weight of the edge having the maximum weight in the directed path. The min-max path from $p_{0}$ to $p_{n}$ is the directed path from $p_{0}$ to $p_{n}$ with the minimum cost.

Now $C$ has an $r$-gather-clustering with cost $k$ iff $D(V, E)$ has a directed path from $p_{0}$ to $p_{n}$ with cost $k$. See Fig. 2.

Thus if we can compute the min-max path in $D$ then it corresponds to the solution of the $r$-gather-clustering problem. Intuitively, each (directed) edge in the min-max path corresponds to a cluster of an $r$-gather-clustering.

We can construct the $D(V, E)$ in $O(r n)$ time. Then compute the min-max path from $p_{0}$ to $p_{n}$ in $O\left(r n \log ^{*} n\right)$ time, since an $O\left(|E| \log ^{*}|V|\right)$ time algorithm for the min-max path problem for a
directed graph $D=(V, E)$ is known [9].
Thus we have the following theorem.
Theorem 2.3 One can solve the $r$-gather-clustering problem in $O\left(r n \log ^{*} n\right)$ time, when all points in $C$ are on the line.

## 3. $r$-gathering

In this section we give an algorithm for the $r$-gathering problem when all points in $C$ and $F$ are on the line, by reducing the problem to the min-max path problem for a weighted directed graph.

Let $C=\left\{c_{1}, c_{2}, \cdots, c_{n}\right\}$ and $F=\left\{f_{1}, f_{2}, \cdots, f_{m}\right\}$ be points on the horizontal line and we assume they are sorted from left to right, respectively. Similar to Lemma 2.1 we can assume the points assigned to a facility are consecutive in a solution.

For consecutive three facilities $f_{j-1}, f_{j}$ and $f_{j+1}$ in $F$ let $m_{L}$ be the midpoints of $f_{j-1}$ and $f_{j}$, and $m_{R}$ the midpoints of $f_{j}$ and $f_{j+1}$. We have the following two lemma.

Lemma 3.1 If $C$ has $2 r$ or more points on the left of $m_{L}$, then $c_{i^{\prime}}$ with $i^{\prime}<i$ is never assigned to $f_{j}$ in a solution of the $r$-gathering problem, where $c_{i}$ is the $2 r$-th point in $C$ on or left of $m_{L}$.
Proof. Assume for a contradiction such $c_{i^{\prime}}$ is assigned to $f_{j}$. We have two cases.

If the rightmost point assigned to $f_{j}$ is on the left of $m_{L}$ then reassigning the points assigned to $f_{j}$ to $f_{j-1}$ results in a new $r$ gathering and since it does not increase the cost the resulting $r$ gathering is also a solution of the given $r$-gatheing problem.

Otherwise, the rightmost point assigned to $f_{j}$ is on or right of $m_{L}$. Then at least $2 r$ points on or left of $m_{L}$ are assigned to $f_{j}$ (possibly with other points on the right of $m_{L}$ ) Let $C^{\prime}$ be the subset of $C$ consisting of the points (1) assigned to $f_{j},(2)$ on or left of $m_{L}$, and (3) but not the rightmost $r$ points on or left of $m_{L}$. Note that $\left|C^{\prime}\right| \geq r$ holds and $C^{\prime}$ contains $c_{i^{\prime}}$. Reassigning the points in $C^{\prime}$ to $f_{j-1}$ results in a new $r$-gathering and the resulting $r$-gathering is also a solution since it does not increase the cost.
Q.E.D.

Intuitively if $c_{i^{\prime}}$ is too far form $f_{j}$ then $c_{i^{\prime}}$ is never assigned to $f_{j}$. Symmetrically we have the following lemma.

Lemma 3.2 If $C$ has $2 r$ or more points on the right of $m_{R}$, then $c_{i^{\prime}}$ with $i^{\prime}>i$ is never assigned to $f_{j}$, where $c_{i}$ is the $2 r$-th point in $C$ on or right of $m_{R}$.

We have more lemma. Let $C^{\prime}$ be the set of points between $m_{L}$ and $m_{R}$ except the leftmost $2 r$ points and the rightmost $2 r$ points.

Lemma 3.3 If $C$ has $5 r$ or more points between $m_{L}$ and $m_{R}$, then the customers in $C^{\prime}$ are assigned to $f_{j}$ in a solution of the $r$-gathering problem.
Proof. Immediate from the two lemmas above.
Q.E.D.

Thus if we can compute the solution for $C-C^{\prime}$ then appending the assignment from points in $C^{\prime}$ to $f_{j}$ results in the solution for $C$. From now on we assume we have removed every such $C^{\prime}$ from C.

We have more lemmas for the boundary case. Let $m$ be the midpoints of $f_{1}$ and $f_{2}$ in $F$.

Lemma 3.4 If $C$ has $2 r$ or more points on the left of $m$, then each $c_{i^{\prime}}$ with $i^{\prime}<i$ is assigned to $f_{1}$ in a solution of the $r$-gathering
problem, where $c_{i}$ is the $2 r$-th customer in $C$ on the left of $m$.
Proof. Immediate from Lemma 3.1.
Q.E.D.

Let $m$ be the midpoints of $f_{m-1}$ and $f_{m}$ in $F$.
Lemma 3.5 If $C$ has $2 r$ or more points on the right of $m$, then each $c_{i^{\prime}}$ with $i^{\prime}>i$ is assigned to $f_{m}$ in a solution of the $r$-gathering problem, where $c_{i}$ is the $2 r$-th customer in $C$ on the right of $m$.

Thus we have the following lemma.
Lemma 3.6 The number of points in $C$ possibly assignning to each facility $f \in F$ is at most $9 r$.
Proof. For each $f_{j}$ with $1<j<m$ define $m_{L}$ and $m_{R}$ as above. The number of points possibly assigning to $f_{j}$ is (1) at most $2 r$ on the left of $m_{L}$, (2) at most $2 r$ on the right of $m_{R}$, and (3) at most $5 r$ between $m_{L}$ and $m_{R}$, by the lemmas above. Similar for $f_{1}$ and $f_{m}$.
Q.E.D.

Now we are going to define a weighted directed graph $D(V, E)$ for $F$ and $C$, and the weight of each edge.

The set of vertices is defined as follows.

$$
V=\left\{p_{0}, p_{1}, p_{2}, \cdots, p_{n}\right\}
$$

For each facility $f_{h}$ with $h=2,3, \cdots, m-1$ and its possible cluster consisting of points $\left\{c_{i+1}, c_{i+2}, \cdots c_{j}\right\}$ we define an edge $\left(p_{i}, p_{j}\right)$. So $\left(p_{i}, p_{j}\right)$ is an edge iff
(1) $i+r \leq j \leq i+2 r-1$
(2) $i \geq i^{\prime}$ where $i^{\prime}$ is the $2 r$-th customer on the left of $m_{L}$, and
(3) $j \leq j^{\prime}$ where $j^{\prime}$ is the $2 r$-th customer on the right of $m_{R}$, where $m_{L}$ and $m_{R}$ are defined for $f_{h}$ as in Section 2. Let $E_{j}$ be the set of edges consisting of edges defined above. Simillary we define $E_{1}$ and $E_{m}$.

Finally,

$$
E=E_{1} \cup E_{2} \cup \cdots E_{m}
$$

Note that $G$ may contain many multi-edges.
The weight $w$ of an edge $\left(p_{i}, p_{j}\right)$ for $f_{h}$ is the maximum of (1) the distance between $p_{i}$ and $f_{h}$, and (2) the distance between $p_{j}$ and $f_{h}$.

The cost of a directed path from $p_{0}$ to $p_{n}$ is defined by the weight of the edge having the maximum weight in the directed path. The min-max path from $p_{0}$ to $p_{n}$ is the directed path from $p_{0}$ to $p_{n}$ with the minimum cost.

We need to compute for each $f_{h}$ the $2 r$-th customer on the left of $m_{L}$ and the $2 r$-th customer on the right of $m_{R}$. By scanning the line we can compute them for all $f_{h}$ in $O(|F|+|C|)$ time in toal. Note that each edge in $E$ corresponds to a pair of customers possibly assigning to a common facility. Thus the number of the edges in $E$ is at most $81 r^{2}|F|$ by Lemma 3.6. Thus we can construct $D(V, E)$ in $O\left(|F|+|C|+81 r^{2}|F|\right)$ time in toal.

Similar to Section 2 we have reduced the $r$-gathering problem to the min-max path problem, and have the following theorem.

Theorem 3.7 When all $C$ and $F$ are on the line one can solve the $r$-gathering problem in $O\left(n+r^{2} m \log ^{*} n\right)$ time, where $n=|C|$ and $m=|F|$.

## 4. Conclusion

In this paper we have presented an algorithm to solve the $r$ gather clustering problem when all $C$ are on the line. The running
time of the algorithm is $O\left(r n \log ^{*} n\right)$, where $n=|C|$. We also presented an algorithm to solve the $r$-gathering problem, which runs in time $O\left(n+r^{2} m \log ^{*} n\right)$, where $n=|C|$ and $m=|F|$.

Can we design a linear time algorithm for the $r$-gathering problem when all $C$ and $F$ are on the line?

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