

Improved Algorithms for Placing Undesirable Facilities *

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Abstract

We improve a number of existing algorithms for determining the location of one or more *undesirable* facilities amidst a set P of n demand points, under various constraints and distance functions. We assume that the demand points reside within some given bounded region R . Applying concepts and techniques from Computational Geometry, we provide efficient algorithms for the following problems:

1. **Brimberg Mehrez '94:** Locate k undesirable facilities within R under the constraints that the smallest distance between each demand point and the facilities is greater than a given r , and the distance between any two facilities is greater than a given d . Under the L_∞ (L_1) norm we present efficient algorithms for any k and under the L_2 norm we can locate efficiently two such facilities. In all cases R is assumed to be an axis-parallel rectangle.

2. **Wesolowsky '94:** Given a set of weighted demand points contained in an axis-parallel rectangular region (resp. circular region) R , and given a smaller axis-parallel rectangle (resp. circle) r , locate r within R such that the sum of the distances from the demand points in r is minimized.

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These problems deal with *undesirable* or *obnoxious* facilities [BKS1, BKS2, BM]. A facility is called undesirable or obnoxious if it may pose a danger to the surrounding area, may have an adverse effect on property values, or may cause lower quality of life through pollution. Examples of obnoxious facilities are nuclear power plants, garbage dump sites, mega-airports, and chemical plants.

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In this paper we consider a number of problems, all of which are concerned with locating an undesirable facility (or a number of facilities) amidst demand points, under various distance functions and varying constraints. We survey previously known algorithms for these problems, and propose more efficient algorithms which are based on concepts and techniques from Computational Geometry.

Problem 1: Maxmin multiple facility location. In their paper "On the Maxmin multiple facility location problem using a maxmin criterion and rectangular distances", Brimberg and Mladenović [1] consider the Maxmin multiple facilities location problem under the L_∞ distance as follows. Given a set $P = \{p_1, \dots, p_n\}$ of n points in the plane, a distance r and another distance d , locate k undesirable facilities such that the smallest distance between each demand point and the nearest facility is at least r , and the distance between any pair of facilities is at least d . Brimberg and Mladenović give a bound algorithm and their algorithm runs in time $O(n^2)$. In this paper we give an algorithm which runs in time $O(n \log n)$. Our algorithm can be extended to handle different separation constraints. In [2] we show that for the Euclidean distance

Problem 2:

there exist a set of k site locations in V with pairwise distance at least d ” If there is one we report the locations. The associated optimization problem will output the largest square size r^* for which the answer for the decision problem is “yes”. We will describe it to the end of this section.

We solve several variants of the decision problem Under the L_∞ norm we find $O(n \log n)$ -time algorithms for $k=2,3$, and then present a general scheme for k that allows us to obtain efficient algorithms that run in time $O(n^k)$ significantly improving the $O(n^{2k})$ (for any $k \geq 1$) solution proposed by Mehzrez [BM].

We also present an algorithm that locates two obnoxious facilities, in $O(n \log n)$ time. The space requirements of our algorithms are considered in [BM]. Specifically, the algorithms in [BM] require $O(n^2)$ space, while the space requirements of our algorithms vary between $O(n)$ and $O(n \log n)$.

$k=2$. It is well known that the combinatorial complexity of the boundary of n squares is linear in n [PS]. In other words, the boundary has $O(n)$ vertices and edges and can be computed in time $O(n \log n)$ (see [M]). Thus the boundary of U (and therefore the boundary of V) can be computed in time $O(n \log n)$ and space $O(n)$. The problem of locating two obnoxious facilities at least d under the L_∞ norm boils down to finding the largest and smallest x -coordinate, and the largest and smallest y -coordinate. Since we can choose these coordinates in time $O(n)$, by going over

Theorem 2.1 *The two-facility problem can be solved in $O(n \log n)$ time and $O(n)$ space.*

$k=3$. As in the previous case, we can compute the boundary of U in time $O(n \log n)$ and space $O(n)$.

Claim 2.2 *If there are no three obnoxious facilities in U that satisfy the constraints above, then the boundary of V is a simple polygon.*

Proof. Assume that there are three obnoxious facilities in U that satisfy the constraints above, and let c be the center of the largest square in U . Then c is inside the boundary of V and is not a vertex of the boundary of V . Thus the boundary of V is a simple polygon.

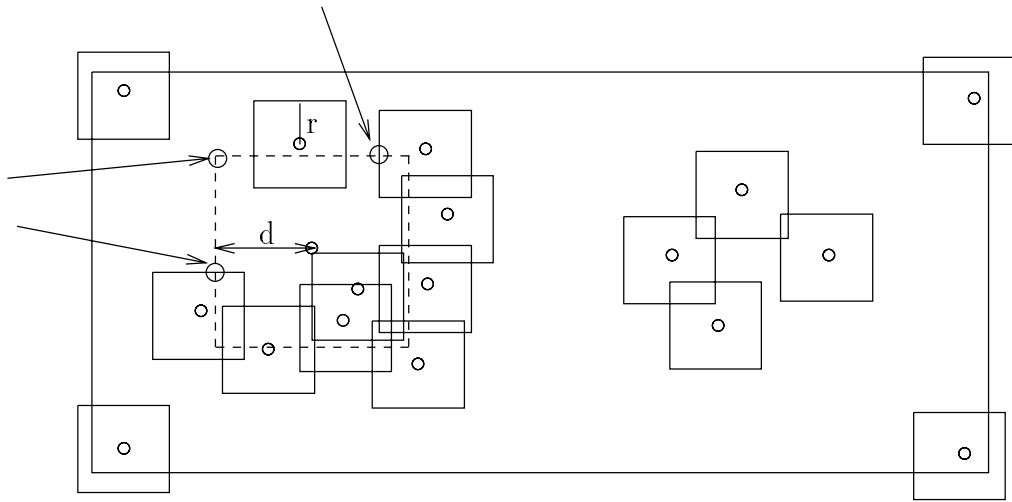


Figure 1: The extremal points of B

This approach might naively lead to a roughly quadratic runtime, since Q might intersect the boundary of V in $O(n)$ points, and we need to compute for each vertex $v \in V$ the region $V - Q$ and identify two locations in it. However, we observe that if one of the other two facilities u_1 and u_2 is located on a vertex of $B = \partial(V - Q) \cap \partial Q$, then it can be dragged to an extremal point of B namely, the highest or lowest point in B (alternatively, leftmost or rightmost point in B), while maintaining the distances between the three facilities greater than d (see Figure 2). This is because the first chosen facility v is certainly within d from the dragged u_1 and also the distance between u_1 and u_2 remains greater than d since we are working with L_∞ (and, thus, there must be some direction (alternatively, left or right) for which the distance between u_1 and u_2 will increase).

The above process can be performed more efficiently by preprocessing the boundary of V for orthogonal range searching with the framework of [BKOS, CH1, CH2]. For each boundary vertex v , perform a range query of size $2d$ centered at v , and obtain the vertices of the boundary of B as a collection of $O(\log n)$ canonical sets.

In addition we need to determine for each of the vertices in B For this, we preprocess the horizontal segments of U together with the two horizontal (resp. vertical) segments of U of logarithmic time vertical (resp. horizontal) range searching and does not lie outside of R . For each vertex v we perform an orthogonal ray query of length ϵ containing ϵ to detect the extremal point of B . We proceed by solving a linear programming problem in logarithmic time. As described in [CH2], we use canonical sets which are of size $O(\log n)$ (see [CH2, y]). The fact that

We also consider the at most 8 extreme points that were computed on the boundary of Q .

Theorem 2.3 *The three-facility location problem can be solved in $O(n \log n)$ time and $O(n \log n)$ space.*

$k \geq 4$. In this case we claim that

Lemma 2.4 *At least one of the sites is a vertex of $R-U$*

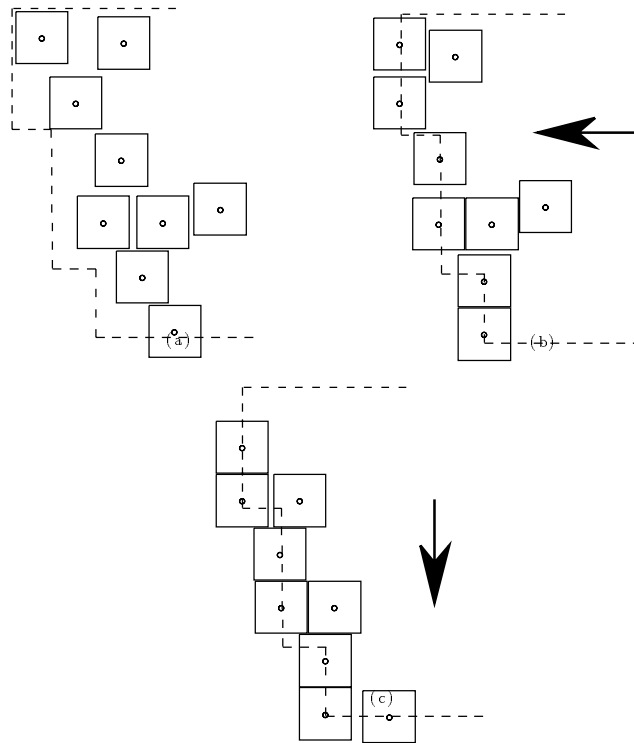


Figure 2: The movement of the squares

Proof. Let us consider the rectilinear free space $V=R-U$. Assume that there is an initial positioning for k locations such that none of them is on a vertex of V . Our approach is to move the facilities, maintaining the $\geq d$ distance requirement, such that at least one of them will be on a vertex. Denote the facilities by $F=\{f_1, \dots, f_k\}$. About each facility f_i , consider an axis-parallel square c_i with side size d (f_i being the square's center). In the initial positioning the squares do not intersect. We push all the squares as far to the right as possible, so that they still do not intersect, and their right side is on the right boundary. If at some point during this stage, one of the squares' center is at a vertex, then we are done. At the end of this stage, the squares are positioned along the right boundary of the boundary of V (if there are several vertices on the right boundary). Next we push them as much down as possible.

and not letting the facilities to penetrate into U nor leave R and stopping if at some point a facility passes through a vertex of V . At the end of this stage, the bottommost facility f_j must lie on a horizontal edge of the boundary of V (and if there are several bottommost facilities f_j is the leftmost among them). See Figure 2.

Assuming we have not stopped with a facility on a vertex, then we know that since otherwise the corresponding facility lies on a vertex and we would have stopped. Now check whether we can slide the square c_j to the left, under the same condition that its center f_j coincides with a vertex of V . If we can, then we are done. If not, we proceed as follows. Consider the south west quarter plane defined by the bottom edge of c_j and the line through the left edge of c_j .

There is at least one square that is fully contained in this quadrant and whose bottom edge is a square x blocking c_j from below, and there exists a square y such that if $x=y$ then this square is such a square. Otherwise, if $x \neq y$, then x is such a square. And if x does not exist, then y is necessarily below the bottom edge of c_j , and y is fully contained in the above quadrant. If y is not fully contained in this quadrant, then y is not a square, and repeat the whole process. We are not afraid that there are only leftmost squares.

2.1 The optimization scheme

In order to find the largest value r^* for which there still exists a solution to the k -facility location problem (keeping d fixed), we employ the technique of Frederickson and Johnson [FJ], as has been done before (see, e.g., [GS] and many others). Each pair p, q of demand points determines eight critical values, four for each dimension. We list the critical values of the x -difference d_x between p and q : (i) $d_x/2$, (ii) d_x , (iii) $(d_x - d)/2$, and (iv) $(d - d_x)$. In addition, each demand point p determines four critical values; the two horizontal distances between p and the boundary of R and the two vertical distances between p and the boundary of R .

We can represent all these distances as a constant collection of sorted numbers. We perform a binary search on these values using the decision algorithm. As was shown in [FJ], the above scheme adds a multiplicative $O(\log n)$ factor to the running time of the decision algorithm.

3 Minimum coverage

Let P be a set of n points within an axis-parallel rectangle R . Let R' be a smaller rectangle which is smaller than R (both in width and in height) and is contained within R such that the total number of points lying in R' is at least k . Consider the optimal location of r . If the bottom edge of R' does not lie on the bottom edge of R then these conditions does hold, without changing the location of r (we move r downwards and closed from above).

Below we show how to find the optimal location of r on the bottom edge of R' or on an horizontal line segment parallel to the bottom edge of R' . To the above observation, this location is determined by d and its height. The solution is on the bottom edge of R' and by d its height. The solution is on an horizontal line segment parallel to the bottom edge of R' and by d its height.

The segment that consisting of all points r on the bottom edge of R' of length c , centered at the point r .

Initially the tree is empty. We start with the lower edge of R' and the upper side of R' .

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We now start moving the slab upwards until the first of the following events occurs: either the (i) upper or (ii) lower side of the slab hits a point of P , or the upper side of the slab hits the top of B . At event (i) we insert into T the segment corresponding to the point of just encountered, updating the weights in the nodes on the paths to the root. At event (ii) we delete the point and update the weights accordingly. Event (iii) terminates. We choose the event where the minimal weight was achieved at the root to be the event that determines the location of r . Clearly handling an event of type (i) or (ii) takes $O(\log n)$ time. Thus we obtain the following theorem

3.1 A lower bound

We obtain an $\Omega(n \log n)$ lower bound for the above minimum coverage problem in the 1-dimensional case. Besparyatnikh et al. [BKS2] obtained an $\Omega(n \log n)$ lower bound for the following problem. Given n positive real numbers and a number γ , there exist two consecutive numbers in the sequence a_1, \dots, a_n obtained by sorting the numbers, such that their difference is greater than γ . Our reduction is based on the segment $[a_i, a_{i+1}]$, and let r be a segment of length γ . Every point in r is a 1-dimensional point with weight 1. If we can place r within the interval $[a_i, a_{i+1}]$ such that the points lying in r is 0, then two such numbers exist. Thus we conclude that

Theorem 3.1 *Given a set P of n points within an axis-parallel rectangle r which is smaller than $\Theta(n \log n)$ time, such that the total number of points in r is 0, then two such numbers exist.*

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