CONSTRUCTION OF THE VORONOI DIAGRAM BY A TEAM OF COOPERATIVE ROBOTS

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Abstract: This paper presents a method for cooperation in the construction of Voronoi diagrams which is suitable for use in dangerous tasks performed by a team of robots. The algorithm has been implemented on a network of eight workstations using the MPI library. Two implementation approaches have been used. In the first one, no communication among the robots is required but some degree of redundancy in the work performed by the robots may result. In the second approach, a more cooperative scheme is adopted and, as a consequence, communication among the robots increases but the work performed by each one is reduced. In both approaches, the calculation time decreases almost linearly when adding robots to the team. Nevertheless, the second approach, more cooperative, has consistently produced better results. With the achieved speed-up, it is possible to use this algorithm in applications where the obstacle configuration within the robot team working area changes with time.

1 INTRODUCTION

This paper presents a message-passing cooperative algorithm for the construction of Voronoi diagrams which is suitable for use within a team of cooperative robots. The algorithm implementation allows the required storage to be evenly distributed among the robots and demands little or no communication at all among the robots. However, in this latter case, the robots may end up doing some redundant work.

Different approaches, such as visibility graphs (Pere, 1979), potential fields (Kim, 1991) (Aude, 1999) and Voronoi diagrams (Lato, 1996), can be used in the solution of the path-planning problem. This paper is focused on a cooperative implementation of the Voronoi diagram by a robot team.

More recently, parallel algorithms for the construction of Voronoi diagrams have also been presented (Tzio, 1997) (Sudh, 1999). Both algorithms have been conceived for implementation on dedicated VLSI cellular architectures which are based on fine grain parallelism and tight coupling between neighbor cells consisting of very simple hardware. Therefore, these algorithms are not suitable for use in distributed environments such as a team of cooperative robots.

The algorithm works for arbitrarily shaped robots, as long as the distance from the robot center to its most external point is known, and assumes that the arrangement of obstacles within the working area is known a priori. However, since the algorithm scales very well with the number of robots, it can be used to compute new Voronoi diagrams very fast whenever the configuration of obstacles change. Therefore, the proposed algorithm can be very useful in applications where teams of cooperative robots work in time-varying environments requiring path re-planning due to changes in the obstacle configuration.
Section 2 of this paper describes the basic sequential algorithm used to construct the Voronoi diagrams and its required data structures. Section 3 discusses two different approaches for the cooperative implementation of this algorithm. In Section 4, performance results are presented considering implementations of the algorithm on an Ethernet cluster of workstations and the use of MPI (Message Passing Interface). Finally, in Section 5, the main conclusions of the paper are summarized and the proposals for future work are presented.

2 BASIC ALGORITHM

The proposed algorithm works on a grid of square cells that represents the plane working area of the robots. The obstacles which are present in this area are known a priori and a separate data structure holds the coordinates of the obstacle corners and the equations of the obstacle edges. In the two-dimensional plane, the obstacles must be represented as a single convex figure or as a set of convex figures, including circles. Any non-convex figure must initially be broken into two or more convex figures before the algorithm starts its operation. Circular objects are described by the center coordinates and the value of the radius.

The grid data structure holds status information for each cell and the closest obstacle identification for every corner of a grid cell. In a typical implementation, both the cell status and the obstacle identification can be stored in a single byte. Since for a grid consisting of \( N \times N \) cells there are \( (N+1)^2 \) corners, the number of bytes needed to store the information associated with grid cells status and corner information is given by: \( N^2 + (N+1)^2 \), which is close to \( 2N^2 \) for large values of \( N \), thus giving \( O(N^2) \). Therefore, for a 1000 x 1000 grid the required amount of storage is around 2 Mbytes.

The proposed algorithm is based on three very simple ideas. The first one is that all the vertices of the working area bounding polygon are Voronoi vertices. The second one derives from the observation that the Voronoi diagram is always a connected diagram, in which the Voronoi arcs always have at least one intersection point. And the third is that if at least 2 cell corners are closer to different obstacles then the cell belongs to the Voronoi Diagram. Therefore, given a starting point for the diagram, it is possible to draw it on a grid by checking at each step which is the next neighboring cells to be visited. Such cells will be the ones whose common edge with the current cell connects two corners labeled with different obstacles and is not the cell entrance edge. This procedure is shown when a Voronoi vertex is reached on Figure 1.

Based on these three ideas the algorithm finds the Voronoi diagram (VD) using the following procedure:

```
begin
    create a queue to temporarily hold the Voronoi cells;
    mark all the cells inside the working area as free and all the cells external and around the working area as blocked;
    choose one of the working area bounding polygon vertices as the VD starting point;
    mark the corresponding cell as belonging to the VD;
    insert the starting point in the queue;
    while the queue is not empty do
        remove a cell from the queue;
        mark it as belonging to the VD;
        for each edge whose corner labels are different
            if it’s not the entrance edge and not belonging to the VD;
                mark it as belonging to the VD;
                for each edge whose corner labels are different
                    insert this neighbor cell in the queue and fill in its corner’s label;
                end if;
            end for;
        end while;
    end;
end;
```

The algorithm starts by choosing one of the working area bounding polygon vertices as the Voronoi diagram starting point and inserts it in the temporary queue. Then, while this queue is not empty, the algorithm removes cells from the queue and marks them as belonging to the diagram. For each cell marked, the next cell choice procedure is used to determine which cells will be put in the queue. All cells selected in the previous procedure have their four corners labeled with the Euclidean distance to the closest obstacle. Finally, the selected cells are entered into the queue. The procedure is repeated until no more cells can be found in the queue. At this point the Voronoi diagram is completely determined.

The amount of computation required for finding the closest obstacle to a particular cell corner is reduced if the algorithm computes only the distances to obstacle edges which are “visible” from that...
corner. A limited but effective visibility test of an obstacle edge can be performed by finding the inner product between a perpendicular outward vector to the obstacle edge and a vector which starts at the cell corner under consideration and points to any point inside the obstacle polygon or on its edges. If the inner product is negative, the edge is visible, otherwise it is invisible and no distance calculations need to be performed.

The procedure implemented by the described algorithm would be sufficient to find the Voronoi diagram for solving the path planning problem for robots which are reduced to a point. In order to make the algorithm work for arbitrarily shaped robots, a slight modification has to be introduced. Let us consider that the distance from the robot center to its most external point is \( r \). Then, when the algorithm finds a cell which should belong to the Voronoi diagram, it must perform a proximity test to verify if the distances from all the cell corners to the closest obstacles are greater than \( r \). If this is true then the cell is marked as belonging to the Voronoi diagram and inserted in the queue. Otherwise the cell is marked as blocked but it is still inserted in the queue, because its neighbors may still belong to the Voronoi diagram which will establish possible paths.

The algorithm error in defining the Voronoi diagram is less than the size of a cell diagonal. If, for instance, the robot working area is a 10m x 10m square and the grid structure is defined as a 1000 x 1000 cell array, the maximum algorithm error is less than 1.41 cm since the cell side is 1 cm long.

3 COOPERATIVE STRATEGY

The proposed cooperative implementation of the algorithm described in Section 2 divides the working area of the robots into \( \text{Nr} \) slices, where \( \text{Nr} \) is the number of available robots. Each robot has a working area slice associated with it and is responsible for finding the Voronoi diagram within that slice. Therefore, only a fraction \( (1/\text{Nr}) \) of the grid data structure needs to be stored by each robot. The data structure containing information on the obstacle corner coordinates and on the equations of the obstacle edges is broadcast to all the robots. Two approaches have been adopted to solve this problem. The first one does not require any communication, but leads to some redundant work performed among the robots. In the second approach, some communication between the robots working on neighbor slices is required but, as a consequence, a more cooperative work pattern is adopted.

If we assume the working area is rectangular and the cut lines are vertical lines, it is possible to say that the Voronoi diagram will certainly cross the vertical borders of a slice. So, in the first approach, every robot searches both vertical border lines of its assigned slice for grid cells representing Voronoi points, as shown in Figure 2, where the use of four processors (P0 to P3) is considered.

Once a Voronoi point is found, the Voronoi diagram segment starting at the corresponding cell and within robot slice is determined using the algorithm described in Section 2. This procedure is repeated until all the cells on both vertical border lines have been visited. The two robots with the leftmost and the rightmost slices assigned to them can do less work because they know a priori the working area corners are Voronoi vertices and, therefore, they can draw the Voronoi diagram segments starting from these two known points. So, each vertical border line is analyzed twice by two different robots as can be seen in Figure 2.

In the second approach, this redundant work is avoided by assigning, for instance, the job of searching a vertical border line to the robot which is responsible for defining the Voronoi diagram within the border right slice. The cell coordinates corresponding to Voronoi points which have been found on the border line are sent to the robot processing the neighbor slice on the left. When the coordinates of a cell are received, the robot puts the cell on its queue. This cell will be used as the starting point for finding a Voronoi diagram segment. With this second approach, each robot searches for Voronoi points on a single vertical border line and receives the information on Voronoi points on the other border line from its right side neighbor. Therefore, each robot does less work and uses communication to help its left side neighbor to do its work.

4 EXPERIMENTAL RESULTS

The parallel algorithm described in Section 2 has been implemented in C with the use of MPI (Message Passing Interface) for implementing the communication functions on an Ethernet cluster consisting of 6 workstations, with 2.4 GHz Intel Pentium 4 processors and 128 Mbytes of memory. This environment has been chosen for the experiments because it is similar to the actual...
environment in which the cooperative robots use Ethernet-like wireless communication.

Two obstacle arrangements have been considered in the experimental work. In the first one, a sea of triangles is placed within the working area. Figure 3 shows the resulting Voronoi diagram produced for this arrangement by the proposed algorithm. The lighter lines represent Voronoi diagram segments which have failed the proximity test.

In the second example, less obstacles are present but they have different types of shapes, including circles, dots and non-convex polygons. Figure 4 shows the resulting Voronoi diagram for this obstacle arrangement (mixed shapes).

For both obstacle arrangements shown in Figures 3 and 4, the application of the visibility test has been able to reduce by nearly 40% the amount of computational work performed by the algorithm.

For each example situation, three grid sizes have been considered in the experiments: a small (1024 x 1024), a medium (2048 x 2048) and a large (4096 x 4096) one. For all combinations of obstacle arrangements and grid sizes, both approaches for the algorithm parallel implementation described in Section 3 have been evaluated considering the use of 1, 2, 4 or 6 workstations in the network.

As a higher grid resolution is used, more work has to be done by the processors in both approaches (particularly in the no communication approach) since the number of cells per slice border increases and the number of Voronoi points in the diagram also increases. However, the amount of work grows linearly with the grid dimension. Therefore, the algorithm performance can scale very well.

It should also be noticed that the computational work associated with the determination of Euclidean distances between each candidate Voronoi point and the visible edges is greater when the sea of triangles arrangement is considered since the number of obstacles is higher. This issue is important because, with the no communication approach, the number of Euclidean distances that are calculated is approximately twice as big as the number of distances evaluated by the cooperative approach. So, when the amount of calculation is reduced, the no communication approach was able to produce smaller running times than the cooperative approach in two situations in which only two processors were in use, the grid size was not the largest one and the mixed shape obstacle arrangement was considered.

Figures 5 and 6 show the speed up achieved by the parallel algorithm based on both approaches.

The figures show that, in most cases, the cooperative approach achieved a slightly higher speed-up. Nevertheless, for both approaches the speed up increases almost linearly with the number of processors when the sea of triangles obstacle arrangement is used. With this arrangement more computational work is available to be done and, as a consequence, better load balancing among a larger number of processors can be achieved.
5 CONCLUSIONS AND FUTURE WORK

Two approaches for the cooperative implementation of an algorithm for the construction of Voronoi diagrams which is suitable for use within a team of cooperative robots have been discussed in this paper. The first one does not require any communication among the robots. However, with this approach some redundant work is performed by the robots. The second approach requires some communication among the robots but implements a more cooperative strategy for the construction of the Voronoi diagram.

Experiments performed on an Ethernet cluster of workstations have demonstrated that both approaches for the algorithm cooperative implementation scale well with the number of processors particularly when the number of obstacles is big. Nevertheless, the approach based on cooperation has produced slightly better results for two quite different obstacle arrangements and for different grid resolutions. With the achieved speed-ups and running times, the parallel algorithm can be used in applications where the obstacle configuration changes with time within the robot team working area and, consequently, real time path re-planning is often needed.

Future work will include the actual implementation and evaluation of the proposed parallel algorithm within a team of cooperative mobile robots under development in our laboratory (Lopes, 2001) (Aude, 2003).

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REFERENCES


