Proceedings of the 1999 IEEE
International Conference on Robotics & Automation
Detroit, Michigan • May 1999

Path Planning and Control for AERCam, a Free-flying Inspection Robot in Space.

Howie Choset, Ross Knepper, Joleen Flasher, Sean Walker, Andrew Alford, and Dean Jackson
Carnegie Mellon University
Scaife Hall
Pittsburgh, PA 15213

David Kortenkamp, Robert Burridge, and Jaime Fernandez

Texas Robotics and Automation Center
Metrica, Inc.
Houston, TX 77058

ABSTRACT

This paper describes a prototype robot and the necessary path planning and control for space inspection applications. The robot is the first generation of a free-flying robotic camera that will assist astronauts in constructing and maintaining the Space Station. The robot will provide remote views to astronauts inside the Space Shuttle and future Space Station, and to ground controllers. The first part of the paper describes a planar robot prototype autonomously moving about an air bearing table. The second part of this paper introduces a method for determining paths in the three-dimensions for efficient fuel use. Finally, this paper describes the software simulation of the path planner with the future space station.

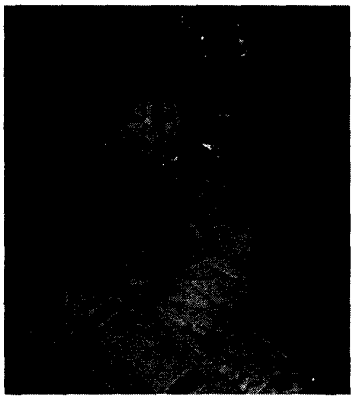

Fig. 1. AERCam space deployment.

1 INTRODUCTION

AERCam (Autonomous Extra-vehicular Robotic Camera) is designed to provide astronauts and ground control camera views of the space shuttle and station. The first generation of AERCam, called AERCam Sprint, flew on a shuttle mission in December 1997. AERCam Sprint was teleoperated by an astronaut inside the space shuttle. AERCam's only autonomy was its ability to automatically stop its rotation when commanded to do so. The next generation of AERCam, called AERCam II, is currently in development at NASA Johnson Space Center. This robot will have additional autonomous functionality and will be controlled by an intelligent layered control architecture called 3T [2].

This paper describes the path planning and control algorithms that direct AERCam's motions. First, we will describe the planar prototype for AERCam and the planar path planning algorithm that directs it. This path planning algorithm is based on the generalized Voronoi diagram (GVD), which has been commonly used in the motion planning field [13]. One of the contributions of this paper is that we use the GVD to locally optimize fuel usage of AERCam on the air bearing table.

The second part of this paper then upgrades the planar path planning approach to three-dimensions using a new motion planning structure called the generalized Voronoi graph (GVG), already described in prior work [4]. Again, we use the GVG initially to first find a path and then apply a similar optimization technique to minimize fuel usage. Software simulations validate this approach for the space station.

2 PLANAR HARDWARE TESTBED

AERCam IGD is a testbed for a free-flying space robot. The AERCam IGD consists of several hardware components and is shown in Figure 2. Although this paper is primarily about the software control algorithms of AERCam, in this section we will briefly describe the major components of the hardware testbed.

- **Flotation sled.** In order to emulate the effects of a friction-less environment on our control algorithms, the AERCam IGD robot floats on a thin layer of air above an air bearing table (ABT). This cushion of air is provided by a flotation sled, which bears the weight of AERCam. An on-board reservoir supplies air, providing tether-free operation.
- **Thrusters.** AERCam IGD moves by firing small nitrogen thrusters. There are eight thrusters: two in each of four directions. By combining different thruster firings rotations and movements can

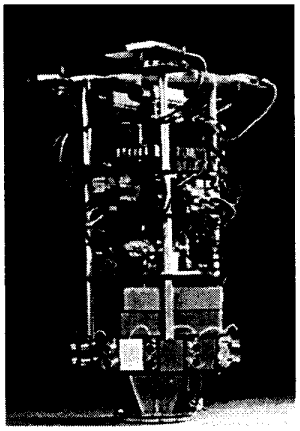

Fig. 2. The AERCam air bearing table robot.

be performed. An on-board tank provides nitrogen for tether-free operation.

- **Infrared detectors.** AERCam IGD has a ring of twelve infrared detectors. These detectors provide a range to objects ranging from six inches to 36 inches away. The range is given as one of sixteen steps between these two distances. Each sensor provides data at approximately ten hertz.
- **Processor.** AERCam has an on-board Pentium 100 MHz processor on a PC-104 bus. The processor runs VxWorks 5.3. The PC-104 bus provides slots for attaching additional cards, including cards for data acquisition.
- **Inertial Measurement Unit.** AERCam IGD has a laser gyroscope that can measure accelerations in both rotation and translation. Control software uses these measurements (plus additional data) to determine AERCam's position and velocity.
- **Vision.** AERCam IGD has two color cameras mounted on its top surface. These camera have a fixed position and a fixed verge (angle between the cameras). They are attached to a wireless video transmitter, described in the next section.
- **Communication.** While a great deal of AERCam's real-time control is done on-board, some high-level control, some perception and the user interface are off-board the robot. AERCam has a wireless Ethernet communication system that connects its on-board Pentium processor with off-board workstations. AERCam also has a wireless stereo video transmitter to provide a video feed to the user interface and off-board processors.

3 SOFTWARE COMPONENTS

The AERCam IGD contained a large number of software components written many different people that has been combined to allow for both teloperation and autonomous operation. Some of the software ran on-board the IGD robot, but most ran on off-board computers.

3.1 Motion Control System

The Motion Control System (MCS) of AERCam provides two functions: 1) determining AERCam's position; and 2) performing real-time control of AERCam. Basic commands that the MCS can implement include: 1) moving in any axis and rotating; 2) stopping; 3) moving to an x,y position; and 4) turning to an angle.

The MCS is a separate process running on-board the AERCam IGD under VxWorks. It communicates via shared memory with the intelligent control system (see Section 3.5). The control laws and position determination algorithms were written by experts on control of free-flying spacecraft and tested extensively in simulation and on the air bearing table. This paper does not focus on the MCS, although its functioning is crucial to a successful robot system, but instead focuses on the higher level planning and control that results in commands to the MCS.

A key component of the MCS is a Global Positioning System (GPS). This GPS consists of six pseudolites (beacons that mimic GPS satellites indoors) and two receiving antenna's, one on AERCam and one off-board as a reference receiver. The GPS system uses differential measurements between the on-board and reference receiver to determine AERCam's position. This position is combined with information from the Inertial Measurement Unit and used to control AERCam. The GPS system was developed at Stanford University and modified for use at NASA JSC.

3.2 Path planning

This path planning algorithm used for the AERCam IGD is based on a geometric structure called a roadmap [3]. Sometimes, roadmaps are called skeletons because they capture the salient geometric structure of an environment much like an animal's skeleton reflects the geometry of its body. Roadmaps have the following properties: *accessibility*, *connectivity*, and *departability*. A robot uses a roadmap to plan a path by finding a path from the start onto the roadmap (accessibility), then along the roadmap (connectivity) and then from the roadmap to the goal (departability). This is similar to how people use roadways, hence the term roadmap structure. Once the robot determines the path from the roadmap, the path is then optimized for fuel usage. The first roadmap used in this paper is the generalized Voronoi diagram (GVD), which has been commonly used in the motion planning field [14]. One of the contributions of this paper is that we use the GVD to locally optimize fuel usage of AERCam on the planar air table.

Later, we will extend the GVD to three dimensions by introducing the generalized Voronoi graph.

3.2.1 The generalized Voronoi diagram

The planar roadmap used in this work is the *generalized Voronoi diagram* (GVD). Ó'Dúnlaing and Yap [14] first applied the GVD, which is the locus of points equidistant to two or more obstacles, to motion planning for a disk in the plane.

To define the GVD, assume the robot is a point operating in a work space, $\mathcal{W}$, which is populated by convex obstacles $C_1, \ldots, C_n$. Non-convex obstacles are modeled as the union of convex shapes. The distance between a point and an obstacle is the shortest distance between the point and all points in the obstacle. The distance function, and its "gradient," respectively are

$$d_i(x) = \min_{c_0 \in C_i} \|x - c_0\| \quad \text{and} \quad \nabla d_i(x) = \frac{x - c_0}{\|x - c_0\|},$$

where (1) d_i is the distance to obstacle C_i from a point x, and (2) the vector $\nabla d_i(x)$ is a unit vector in the direction from x to c_0, where c_0 is the nearest point to x in C_i.

The basic building block of the GVD is the set of points equidistant to two sets C_i and C_j, such that each point in this set is closer to the objects C_i and C_j than any other object. We term this structure the *two-equidistant face*,

$$\mathcal{F}_{ij} = \{x \in \mathbb{R}^m : 0 \leq d_i(x) = d_j(x) \leq d_h(x) \quad \forall h \neq i, j \text{ and } \nabla d_i(x) \neq \nabla d_j(x)\}.$$

A two-equidistant face has co-dimension one in the ambient space, and thus in the plane, a two-equidistant face is one dimensional [4]. The union of the two-equidistant faces forms the generalized Voronoi diagram, i.e.,

$$\text{GVD} = \bigcup_{i=1}^{n-1} \bigcup_{j=i+1}^{n} \mathcal{F}_{ij}.$$

See Figures 3 for examples of the GVD. Note GVD edges terminate on the boundary at nodes termed *boundary points* and at nodes termed *meet points* where other GVD edges terminate.

Now, we show how the robot can use the GVD for motion planning. Given an arbitrary start location for AERCam, there always exists a path from the start to at least one point on the GVD. This path is described as

$$\dot{c}(t) = \nabla d_i(c(t))$$

where C_i is the closest object to the robot. Essentially, this path directs AERCam away from the closest obstacle until it encounters another obstacle which is equidistant to the first closest obstacle. This accessibility property of the GVD was show in [14] but is restated here in terms of the distance function. Loosely put, the GVD has the accessibility property because when the robot moves away from its closest object it must encounter another object to which its is equidistant due to boundedness.

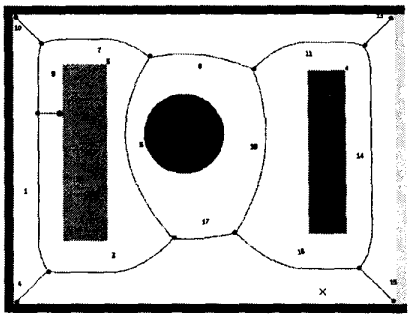

Fig. 3. The solid curve segments are the edges of GVD.

It is shown in [5], that all points in the free space are within at least one point on the GVD. This means, when the robot negotiates the entire GVD, it is guaranteed to "see" the entire work space. This feature furnishes the GVD with the property of departability.

So, we have shown that the robot can access the GVD and depart from it, and now we describe the intermediate step — traversing the GVD from the access to the depart locations on the GVD. Since the GVD is connected [14], we are guaranteed that there will be a path on the GVD between the access and depart locations. This path is determined via a graph search of the GVD. The robot traverses an edge until it encounters a meet point. Here, the robot search branches and another edge is traversed. If the robot encounters a boundary point, it simply returns to a meet point with an un-searched edges associated with it. If the robot encounters an already visited meet point, then there is a cycle in the graph and again, the robot returns to a meet point with an unexplored edge associated with it. If a path exists, this procedure will terminate when the robot becomes within line of sight of the goal, at which point the robot departs the GVD for the goal. See Figure 4.

If no path exists between the start and goal, then this procedure terminates when all meet points have no un-searched edges associated with them. The power of this procedure is that it determines a path when one exists or reports failure when no path exists. Such a path planner is called *complete*.

3.2.2 Minimizing Fuel Usage

The planar AERCam uses its thrusters to move on the air table, thereby emulating space-like deployment. Naturally, AERCam must fire its thrusters several times to follow an arbitrary path. In Figure 4, AERCam would have to fire its thrusts several times to negoti-

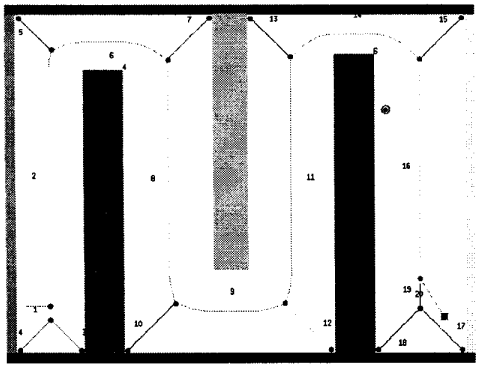

Fig. 4. AERCam determined a path from a start to a goal location by accessing the GVD, traversing the GVD, and then departing the GVD.

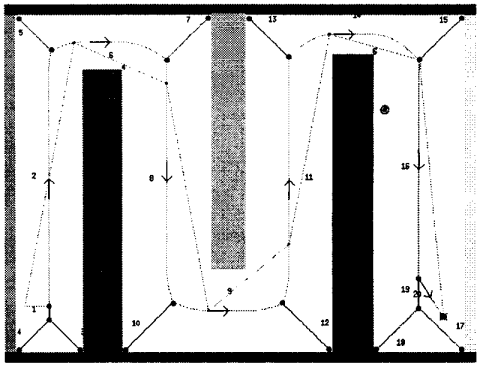

Fig. 5. Using the GVD, AERCam determined a path as a sequence of straight line segments from a start to a goal location.

ate the curved portions of the GVD. Firing thrusters requires fuel; fewer bursts requires less fuel thereby making the mission more cost-effective. Our goal is to minimize the number of firings AERCam requires to traverse from a start to a goal location.

Initially, we considered orbital mechanics calculations to optimize fuel usage. This proved to be unnecessary because recently NASA researchers at the Johnson Space Center have shown that taking orbital dynamics into account when doing AERCam-type navigation does not significantly reduce fuel usage [8]. The scientists concluded that a sequence of straight line paths from start to goal suitably approximates an optimal solution for fuel usage.

Once the GVD-path is determined from the GVD, the straight-line-path-sequence optimization procedure is quite simple: move along the GVD-path until the start location can no longer be seen. The last point where the start is within line of sight is a "way-point" for AERCam. This procedure is then repeated with the way-point serving as the start location until the goal is reached. The result is a sequence of way points that describe a piece-wise linear path from start to goal.

This optimization procedure will produce paths that bring AERCam unacceptably close to obstacles, so the operator specifies a safety parameter describing the minimum distance to an obstacle AERCam is allowed to achieve. The robot moves along the GVD-path from the start. Whenever the distance of the line segment defined by the current robot location and start location falls below the safety threshold, the robot defines a way point. This procedure is then repeated with the way point serving as the start location until the goal is reached. See Figure 5.

3.3 Stereo vision

One goal of the AERCam IGD was to be able to maintain a fixed distance and heading to a person. To do this an active stereo vision system was used. This vision software has been developed over a number of years at NASA Johnson Space Center and Metrica/TRACLabs [9], [10], [12]. When tracking an object it can provide the location of the object with respect to the camera. It runs under Windows NT on a 300 Mhz Pentium II using four C80 DSP chips and operates at approximately ten frames per second.

In order to efficiently process the enormous amount of information available from stereo cameras, the system uses techniques that have recently been developed by the active vision research community [1], [6]. In particular, the system addresses the issue of *gaze control*, i.e., where to focus attention and visual resources. In our vision system a Laplacian of a Gaussian convolution is performed on the image data streams as they are simultaneously acquired. Only the sign of the LOG output is stored in memory. Then, a search is performed in which a patch from the left LOG image is compared with a portion of the right LOG image, producing correlation measurements. This search produces a series of correlations from which the strongest (the "peak") is chosen as the best. At the same time, the right LOG image from the frame before is compared with the current right LOG image to measure motion. This correlation data is used to assess information within a bounded volume of space called a proximity space. Each proximity space is controlled by a set of behaviors. These behaviors keep the proximity space on an object that is being tracked and the object's position is reported to the intelligent control system so that the robot's motion can be determined. There is not enough room in this paper to give a detailed account of the stereo vision system – see [11] for more information.

3.4 User interface

All commands to AERCam flow through a user interface. The user interface is implemented in C++ and

runs on a Pentium laptop running Windows 95. The user interface displays telemetry from AERCam and allows the user to teleoperate AERCam using a joystick, to command AERCam to a position and to begin stereo tracking.

3.5 Intelligent control

In the AERCam IGD, high-level control of the robot was done using an intelligent control architecture called 3T [2]. The 3T architecture separates the general intelligent control problem into three interacting layers or tiers (and is thus known as 3T).

- A set of hardware-specific situated skills that represent the architecture's connection with the world. The term "situated skills" [15] is intended to denote a capability that will achieve or maintain a particular state in the world.
- A sequencing capability that can activate the situated skills in order to direct changes in the state of the world and accomplish specific tasks. This tier of the architecture is implemented using Reactive Action Packages (RAPs) [7].
- A deliberative planning capability which reasons in depth about goals, resources and timing constraints. The planning layer was not used in the AERCam IGD.

The skills and skill manager reside on-board the free-flyer running under VxWorks. The sequencing layer runs off-board the free-flyer on a Windows 95 machine. The two communicate via TCP/IP using the wireless ethernet.

3.6 Communication

All interprocess communication in this project was done using the IPC software package from Carnegie Mellon University (approval to use IPC on this project was given by CMU). IPC supports the VxWorks, SunOS, IRIX, Linus, System 7, MacOS 8.0, and Windows 95/NT operating systems. It supports the C, C++, Allegro Common Lisp and Macintosh Common Lisp programming languages. It allows for dynamic reconfiguration of processes and both primitive and complex data structures.

IPC works on a publish/subscribe paradigm. That means that processes that have data publish that data as predefined messages with an identifier. Processes that need data subscribe to these predefined messages using the identifier. Whenever a message is published, the subscribing processes are notified immediately. A central router (called the central server) keeps track of which processes are publishing and subscribing to which messages and routes the messages among them. IPC handles differences in byte ordering between different computer architectures.

4 EXPERIMENTS

Over a period of several months the AERCam IGD was tested on the air bearing table performing a variety of tasks. These tasks included:

- Teleoperating the AERCam IGD using a joystick. At any time, AERCam can be commanded to halt.
- Teleoperating the AERCam IGD using voice commands. At any time, AERCam can be commanded to halt.
- Autonomous navigation of the AERCam to a specified position using path planning.
- Tracking a moving human.
- Obstacle avoidance while navigating to a specified position.

All of these tasks relied on a key set of skills. The sequencer (see Section 3.5) enabled the appropriate set of skills depending on the task and on the situation. First, we will list the core set of skills. Then we will look at each task and examine how the different software components come together to perform each task.

4.1 Skills

Skills are 3T's interface to hardware and low-level robot processes. AERCam has a number of skills that can be activated by the sequencer. The skills all run in VxWorks on-board the robot. A skill manager schedules the skills and manages their data. The following are the core implemented skills:

- **mcs_interface**: communicates with the motion control system (see Section 3.1) on-board the robot.
- **thruster_control**: manages the eight AERCam thrusters.
- **teleop**: interprets joystick commands and produces appropriate AERCam commands.
- **voice_cont**: interprets the voice system commands and produces appropriate AERCam commands.
- **attitude_hold**: maintains the current AERCam attitude.
- **translate_hold**: maintains the current AERCam position.
- **turn_to**: generates the appropriate MCS command to turn AERCam to a given angle.
- **move_to**: generates the appropriate MCS commands to move AERCam to a given position.
- **find_object**: initiates a visual search using the stereo tracking system.
- **track_object**: takes data on the object's location from the stereo tracking system and determines the appropriate movement for AERCam in order to track the object.
- **obstacle_avoid**: examines the infrared sensor data and generates appropriate AERCam movement to avoid any obstacles.

- **battery_state:** monitors the battery and alerts the sequencer when the battery is running low.
- **propellant_level:** monitors the on-board thruster propellant and alerts the sequencer when the propellant is low.

There are a number of other skills that perform more mundane operations such as interfacing to various hardware and software components. However, the skills given above allow the sequencer to task AERCam through a wide variety of activities as will be described in the following subsections.

4.2 Teleoperation

The user interface contained a six degree-of-freedom joystick that could be used to command AERCam. The user first needs to click on the user interface to request control of the robot. This sends an IPC message to the sequencer, which stops any current robot activity and enables the `teleop` skill. The `teleop` skill then waits for IPC messages generated by the user interface corresponding to joystick directives. This continues until the teloperation mode is disabled at the user interface. Once teleoperation is disabled, the robot will not respond to any joystick commands.

4.3 Voice

The voice system is commercial-off-the-shelf voice recognition software. The user requests voice command from the user interface. The sequencer shuts down other robot activity and turn on the `voice_control` skill. While the voice recognition system is always active, any commands it receives are be ignored by the robot unless it was in the proper mode. Once in the proper mode, commands like 'turn right', 'move forward', 'stop', etc. can be given. The voice system generates an IPC message for each utterance and this is interpreted by the `voice_control` skill and an appropriate robot action is performed.

4.4 Autonomous navigation

To do autonomous navigation, the user selects a goal position by clicking their mouse on a drawing of the air bearing table on a laptop computer. The goal position is sent to the sequencing layer of the architecture. The sequencer requests a path from the path planner (described in Section 3.2). The path planner returns a list of via points (or way points) that will move the AERCam robot safely through the environment to the goal position (or it returns an error message stating that no path was found). The sequencer then steps through the via points, activating the appropriate skills in the skill tier. Before sending a new via point, the sequencer checks to make sure that the robot had successfully reached the current via point. If the robot was not successful, the sequencer re-sends the original via point or takes other action (such as asking for a new path). The user can stop the robot's navigation at any time by pressing a button on the laptop.

A sequencer script obtains a path, navigates the robot to each via point in the path and then turns the robot to a final orientation. The actual RAP scripts used in AERCam are more complicated with more checking of robot state and conditional execution paths. The actual RAP scripts also conform to the RAP language, while the example in the figure is in a pseudo-code scripting language. Each step in the RAP script may be another RAP that has its own script or it may be a primitive skill that is then activated.

4.5 Tracking

To track an object or person, the AERCam operator opens the tracking window on the graphic user interface (GUI), which shows the video image from the robot's left camera. The operator positions and sizes a selection box around the object to be tracked and clicks the tracking button. This sends an IPC message to the RAPs sequencer with the coordinates for the box and the command to try to track the object there. The sequencer enables the `find_object` skill with these coordinates (top, bottom, left, right), narrowing the search space for the skill. Then it enables the `track_object` skill, which holds a constant distance from the tracked object.

The stereo vision system tracks the moving person and sends their location in camera coordinates via IPC to the skill manager. A new update is available approximately ten times a second. The `track_object` skill converts the location into world coordinates and determines a location for the robot to move to in order to stay a fixed distance away from the person and in order to stay facing the person. This location is passed the the MCS for execution.

In our experiments, we could track a person who moved at a very slow walk for periods of up to fifteen minutes (limited by the amount of on-board propellant). If the vision system lost the person it automatically performed a search and reacquired the person. Unfortunately, space does not permit a complete detailing of how the vision system acquires and reacquires a moving person. See [11] for more details.

4.6 Obstacle avoidance

AERCam has a ring of twelve infrared sensors that give a distance to obstacles around AERCam. Obstacle avoidance mode can be initiated from the user interface. This causes the sequencer to start the `obstacle_avoidance` skill, which reads all twelve infrared sensors and determines if there is an obstacle in the path of AERCam. An obstacle is signalled if any in-

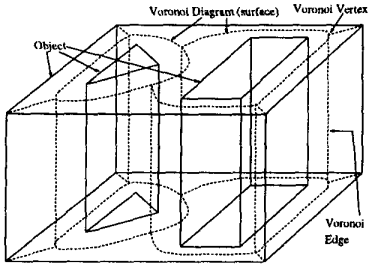

Fig. 6. Two-dimensional GVD in a three-dimensional world.

frared sensor reads an obstacle within thirty inches for three straight readings. When an obstacle is detected, the MCS is signalled to immediately halt the robot. The user then needs to reselect the goal location after verifying that the robot's path is free of obstacles.

5 THREE-DIMENSIONAL PATH PLANNING

The two-dimensional testbed described in this paper is only a first step in producing a free-flying robot for space applications. Many important research issues need to be addressed in the move to three-dimensional environments. Path planning in three dimensions is both computationally complex and non-intuitive. A free-flying robot, such as AERCam, or a person teleoperating this robot, can eliminate this complexity and acquire an intuitive solution using a geometric structure termed a roadmap. With the roadmap in-hand, path planning becomes computationally efficient because the search takes place on the one-dimensional roadmap as opposed to the ambient three-dimensional space.

5.1 Roadmap Definition: The Generalized Voronoi Graph

The roadmap defined in the previous section is limited to the planar case. In three dimensions, the set of points equidistant to two obstacles is a two-dimensional set. Figure 6 has a rectangular and triangular based prism enclosed by a rectangular prism. The dotted lines outline the locus of points equidistant to two obstacles. These two-dimensional sheets comprise the GVD.

Note that the two-dimensional sheets in Figure 6 intersect on a one-dimensional manifold. These one-dimensional edges form our roadmap in three-dimensions. Let $\mathcal{F}_{ijk}$ be a three equidistant face that is defined by the intersection of two-equidistant faces $\mathcal{F}_{ij}, \mathcal{F}_{ik}$, and $\mathcal{F}_{jk}$, i.e., $\mathcal{F}_{ijk} = \mathcal{F}_{ij} \bigcap \mathcal{F}_{ik} \bigcap \mathcal{F}_{jk}$. The pre-image theorem asserts that the three-equidistant faces are indeed one-dimensional. The union of the three-equidistant face is termed the *generalized Voronoi graph*,

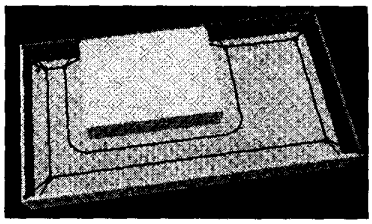

Fig. 7. Generalized Voronoi Graph (GVG) of a simple room with the ceiling removed. The GVG is the set of points equidistant to three obstacles.

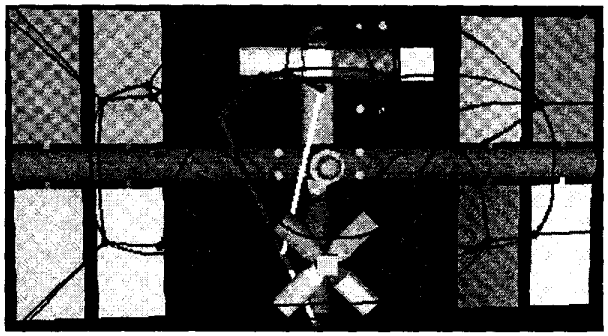

Fig. 8. Space station with GVG.

i.e.,

$$\text{GVG} = \bigcup_{i=1}^{i=n-2} \bigcup_{j=i+1}^{j=n-1} \bigcup_{k=j+1}^{k=n} \mathcal{F}_{ijk}.$$

Figure 7 has a GVG, for a simple three-dimensional rectangular enclosure with a box on the side. The ceiling is removed so the GVG can be viewed.

5.2 Minimize Fuel Usage with the GVG

Variables such as fuel, safety, and time must be optimally budgeted for effective long-term use of AERCam. The AERCam in space will also have a suite of thrusters to effect motion. AERCam must fire its thrusters several times to following an arbitrary path. Hence, the GVG path, by itself, is not optimal for fuel usage (just as the GVD in Section 3.2.2 was not optimal in the planar case).

Again, we do not have to consider orbital mechanics to optimize a path determined by the GVG because of the result found in [8]. Instead, we use the same optimization routine that is described in Section 3.2.2 to determine a sequence of safe piece-wise linear path from start to goal in three-dimensions.

Figure 8 contains the space station, the GVG, and a path that brings AERCam from a start location to a goal location along the GVG. This path is then optimized to minimize the number of thruster firings. The space station, with the GVG path and its optimized path, is shown in Figure 9. Note how the optimized path severely cuts the corners.

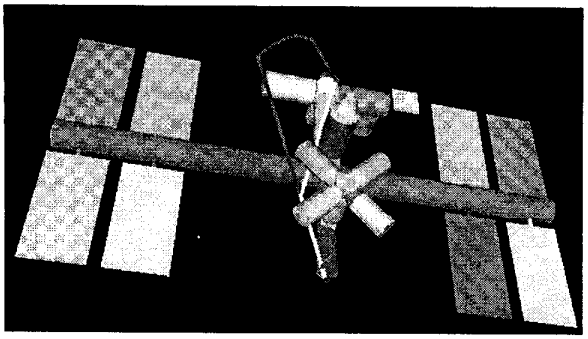

Fig. 9. Space station, path found with GVG, and optimized path.

6 CONCLUSION

The ultimate goal of this work is to enable AERCam, a free-flying robot, to autonomously fly around the space station to look for flaws in its hull. This paper describes two necessary component to achieve this goal: path planning and control. The three-dimensional path planner uses the generalized Voronoi graph (GVG) method and the control was based on an intelligent layered control architecture called 3T.

Before considering the full three-dimensional problem, we used a planar analog to AERCam. This planar robot floated on a thin cushion of air and used small nitrogen thrusters to move. Naturally, if the robot fired a thruster, it would move until either another thruster firing stops it or until the robot collides with an object. This planar system is complicated enough to capture many of the features involved with a free-flying space robot, while simple enough to demonstrate the concepts in a short amount of time.

The first concept demonstrated what the path planning which used a roadmap called the generalized Voronoi diagram (GVD). A robot uses the GVD to plan paths between any two points in an environment by first finding a path onto the GVD, then along the GVD, and then from the GVD to the goal. We then optimize this path with respect to fuel usage and safety.

The second concept demonstrated was the intelligent control architecture. The architecture separates the real-time control of the robot from the more deliberative aspects of intelligent behavior. It is the control architecture that interacts with the robot user and interprets her commands. The control architecture then calls the three-dimensional path planner when necessary. When a path is returned, the control architecture navigates the robot along the path while responding to anomalous conditions.

Upon the successful completion of the planar experiments, we upgraded the technology to handle three-dimensional environments. Therefore, this paper also described a three-dimensional path planning technique and its application to free-flying space robot. The three-dimensional path planner uses the generalized Voronoi graph (GVG). An intelligent control architecture is used to traverse the three-dimensional path and serves to coordinate path planning and execution.

Future work on this project includes using the GVG to localize the AERCam, which would be useful in the case of GPS loses tracking. The localization procedure draws from the previous work of the authors' in mobile robot localization. Essentially, we will exploit geometries encoded in the GVG to determine landmarks that the robot can use for localization. Another avenue of future research is using the HGVG to plan optimal inspection paths for AERCam.

References

[1] Dana H. Ballard. Animate vision. *Artificial Intelligence*, 49(1), 1991.

[2] R. Peter Bonasso, R. J. Firby, E. Gat, David Kortenkamp, D. Miller, and M. Slack. Experiences with an architecture for intelligent, reactive agents. *Journal of Experimental and Theoretical Artificial Intelligence*, 9(2), 1997.

[3] J.F. Canny. *The Complexity of Robot Motion Planning*. MIT Press, Cambridge, MA, 1988.

[4] H. Choset and J.W. Burdick. Sensor Based Planning, Part I: The Generalized Voronoi Graph. In *Proc. IEEE Int. Conf. on Robotics and Automation*, Nagoya, Japan, 1995.

[5] H. Choset and J.W. Burdick. Sensor Based Planning, Part II: Incremental Construction of the Generalized Voronoi Graph. In *Proc. IEEE Int. Conf. on Robotics and Automation*, Nagoya, Japan, 1995.

[6] David J. Coombs and C. M. Brown. Cooperative gaze holding in binocular vision. In *Proceedings of the Fifth IEEE International Symposium on Intelligent Control*, 1991.

[7] R. James Firby. An investigation into reactive planning in complex domains. In *Proceedings of the National Conference on Artificial Intelligence (AAAI)*, 1987.

[8] Rodolpho Gonzalez. Translation maneuver dv costs to the solar array, cw and line trajectories (nasa memo), 1998.

[9] Eric Huber. Object tracking with stereo vision. In *Proceedings of the AIAA/NASA Conference on Intelligent Robots in Field, Factory, Service, and Space (CIRFFSS '94)*, 1994.

[10] Eric Huber and David Kortenkamp. Using stereo vision to pursue moving agents with a mobile robot. In *Proceedings of the IEEE International Conference on Robotics and Automation*, 1995.

[11] Eric Huber and David Kortenkamp. A behavior-based approach to active stereo vision for mobile robots. *Engineering Applications of Artificial Intelligence*, 11(1), 1997.

[12] David Kortenkamp, Eric Huber, and R. Peter Bonasso. Recognizing and interpreting gestures on a mobile robot. In *Proceedings of the National Conference on Artificial Intelligence (AAAI-96)*, 1996.

[13] J.C. Latombe. *Robot Motion Planning*. Kluwer Academic Publishers, Boston, MA, 1991.

[14] C. Ó'Dúnlaing and C.K. Yap. A "Retraction" Method for Planning the Motion of a Disc. *Algorithmica*, 6:104–111, 1985.

[15] Marc G. Slack. Sequencing formally defined reactions for robotic activity: Integrating raps and gapps. In *Proceedings of SPIE's Conference on Sensor Fusion*, 1992.