

Thesis
Subtitle

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Abstract

Real-time path planning is a necessity for any autonomous mobile robot operating in unknown environments. This report describes an implementation of the Game-Theoretic Optimal Deformable Zone with Inertia and Local Approach (GODZILA) path planning algorithm on the Nao humanoid robot by Aldebaran Robotics. The algorithm is a lightweight local path planner that does not require building an environment map. Forward facing ultrasonic distance sensors on the Nao were used for occlusion detection. Theory, simulation, and implementation results will be discussed.

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Chapter 1

Introduction

The path planning problem consists of moving a mobile robot platform or end-effector from a start location to a goal location. Examples vary from ground robots moving through an office building, UAVs navigating to GPS waypoints, or end-effectors avoiding objects on a cluttered table to grasp an object. Methods of solving such a problem depend largely on the construction and knowledge of the environment and the knowledge the robot has about its position within the environment. For example, a mobile robot operating in the plane within a static environment, *a priori* map, and direct position sensing has different challenges than a mobile robot operating in three-dimensions, within an unknown dynamic environment, and only platform velocity estimation. Popular approaches include graph search algorithms such as A* or D*, Bug algorithms, and potential fields approaches. Potential fields algorithms treat the mobile robot as a point mass that is repelled by obstacles and attracted by the goal. The algorithm detailed here

The problem considered in here is that of a mobile robot operating in the plane in an unknown static office-like environment with range and bearing estimate to a goal location. The robot also estimates the range and bearing to occlusions in the environment.

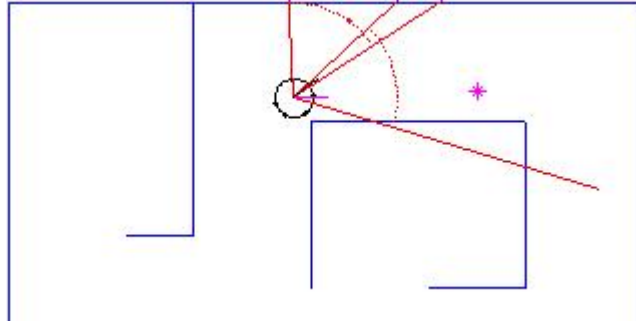


Figure 1.1: Planar robot simulation example. The robot is indicated by a black circle with the red cones simulating ultrasonic sensor beams and their detected range indicated by dotted red arcs in the red cone. Blue lines represent obstacles. The magenta line points towards the goal location, which is indicated by a magenta star.

Chapter 2

Algorithm

The path planning algorithm implemented in this project was the Game-Theoretic Optimal Deformable Zone with Inertia and Local Approach (GODZILA) algorithm. This potential fields algorithm uses the closed form formulation of an optimized cost function that rewards motion towards a goal location and penalizes motion towards obstacles and in directions other than the current heading. In addition to the optimized cost function, GODZILA incorporates a straight line path planner when the goal is in sight of the robot and a randomization function to allow the robot to escape local minima.

Using the closed form formulation of the cost function in quadratic terms, the algorithm effectively generates forces that push the robot from obstacles and pull it towards the goal. The component penalizing motion in directions other than the current heading of the robot acts like an inertia term which helps prevent oscillations and forces any limit cycles to be larger, which helps eliminate certain local minima.

Occlusion forces are generated through use of sensor data from the robot's distance sensors. Each distance reading is used to generate a force magnitude which is applied as a vector to the robot along the orientation of the sensor relative to the robot. The goal force is generated in the same manner, with estimated distance to the goal generating a magnitude applied to the vector along the estimated orientation of the goal relative to the robot. The inertia term provides a constant bias towards the current heading of the robot.

The angle from the sum of these forces is then taken as the desired heading of the robot and converted into an angular rate with is passed to the vehicle.

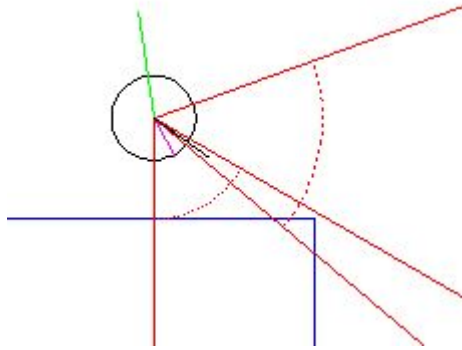


Figure 2.1: Simulation showing component forces. The green line shows the summed occlusion vectors, magenta is the goal vector, black is the inertia vector.

Chapter 3

Platform



Figure 3.1: Nao robot at Control/Robotics Research Lab

The platform utilized on this project was the Nao humanoid experimental robotic platform by Aldebaran Robotics. It is a 25 degree-of-freedom humanoid mobile manipulator with two HD cameras, two ultrasonic distance sensors, four microphones, one three-axis accelerometer, two one-axis gyroscopes, and four force sensors in each foot. The platform is programmed

using a framework called NAOqi allowing development using various languages including C++ and Python.

Two ultrasonic sensors in the chest allowed for distance measurements to occlusions. The transmitters are mounted at an angle of 20 degrees from the forward direction of the Nao and the receivers are mounted at 25 degrees. They have a 60 degree viewing angle with a detection range between 0.25 and 2.55 meters with a 1 cm resolution.

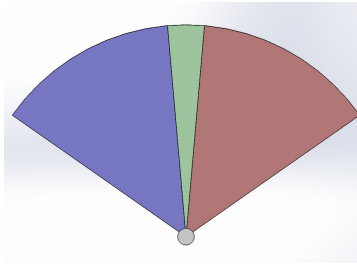


Figure 3.2: Nao sonar cones. The left cone is shown in blue and the right cone is shown in red. The green cone shows the overlap between the two.

As shown in Figure 3.2 the regions covered by the two ultrasonic sensors overlap. As the sensors do not report the angular position of the occlusion within the cone, uncertainty about the location of detected objects is high as they could be anywhere within the sonar cone.

The motion API for the Nao takes three different velocity commands, forward, lateral, and angular, in terms of fraction of maximum step length and step frequency. The maximum step length is 8 cm forward, 16 cm laterally, and 0.523 radians angularly. The maximum step frequency is 2.381 Hz. Due to a stability controller in the built-in gaiting algorithm for the Nao, it takes 0.8 seconds for the robot to react to new commands. At maximum step frequency, this equates to about 2 steps. [1]

Talk about built in red ball tracking for head that returns range and bearing estimates of a red ball 6 cm in diameter.

Chapter 4

Simulation

Algorithm development and robot simulation was done in MATLAB. The simulation constructed a virtual environment using line segments and approximated the robot as a point on the plane. The kinematic model of the robot was taken as a differential drive robot, as modeling the full kinematics of the Nao robot was not necessary to the efficacy of the algorithm. A basic approximation for inertia was made by low pass filtering all velocity commands. Motion noise was injected by adding uniform random noise between zero and one, which was then scaled by a constant, to the linear and angular velocities. The resultant velocities were integrated to gain robot position using RK4 integration. The sonar beams were approximated as cones in the plane with the sensor model returning the distance to the closest object within the region of the cone. These ranges were also perturbed by uniform random noise before being returned to the navigation algorithm.

Bibliography

- [1] Locomotion control. NAO Software 1.14.3 documentation. Retrieved from <http://www.aldebaran-robotics.com/documentation/naoqi/motion/control-walk.html#control-walk>