

Energy-Aware Path Planning for Autonomous Mobile Robot Navigation

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November 28, 2018

Overview

- 1 Introduction
- 2 Energy-Aware Planning Domain
- 3 Practical Implementation
- 4 Experiments and Results
- 5 Related Work
- 6 Conclusions
- 7 References

Introduction

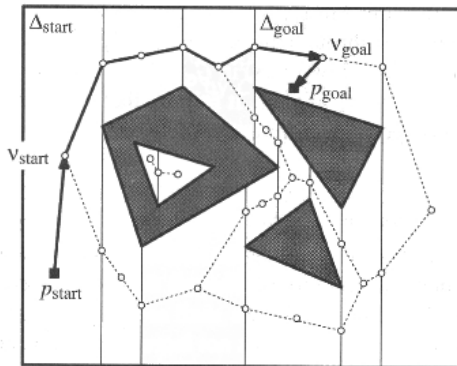
- Embedded processors and Systems-on-Chip have gained attention in mobile robotics recently
 - **Reduced size and weight;**
 - **High performance-per-watt;**
- Many state-of-the-art solutions in mobile robotics require significant computing power
- As SoCs become more powerful to meet this requirement, **the average energy consumption of these solutions increase**
 - Battery life is not a concern for most state-of-the-art algorithms

Introduction

- In this paper, we propose a **path planning** solution for mobile robots, which produces **energy-aware** plans
 - *STRIPS* planning domain with **high-energy zones**;
 - Numerical planning which **minimizes an energy variable at each plan step**;
- We implement a ROS package to integrate the planner with a simulated mobile ground robot, and test it in two experiments:
 - We verify that, due to the energy minimization, our planner **successfully avoids high-energy zones**;
 - We compare the battery discharge curve to a conventional path planning solution, in which our solution **extends the robot's battery life by 1.5 hours, or 42.8%**;

Energy-Aware Planning Domain

- Typically, path planning for autonomous navigation is treated as a graph search problem in a *Configuration Space* (C-Space) [Siegwart et al., 2011]



Energy-Aware Planning Domain

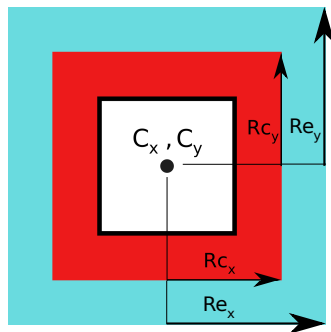
- Being able to describe **different types of actions** is an advantage
 - Both **movement and energy** actions must be considered when planning
- Our objective is to model a planning domain **complex enough to allow planning in an acceptable time** (e.g., ≤ 30 seconds), while **considering movement and energy actions**

World Model

- We use the C-Space representation to describe the world in terms of an **occupancy grid** [Moravec and Elfes, 1985]
 - Spaces can be either occupied by an obstacle or free
- In the STRIP domain, we model the world's **boundaries**, the **resolution** and the **obstacles**
 - Due to dimensionality problems, the obstacles are represented as **geometric spaces**

Obstacle Model

$$Obstacle = \begin{cases} \text{Center : } C_x, C_y \in \mathbb{R}^2 \\ \text{Clearance Radii : } Rc_x, Rc_y \\ \text{Energy Radii : } Re_x, Re_y \end{cases}$$



Robot State Model

- The robot's state is modelled as its **2D position (x,y)** and an **energy variable e**, which indicates the energy requirements at each position, according to the obstacles' energy zones
 - If the robot is in a high energy zone, e is increased, and vice-versa

$$Robot = \begin{cases} \text{Position : } x, y \in \mathbb{R}^2 \\ \text{Energy : } e \end{cases}$$

Possible Actions

- **Energy:** 2 actions
 - **Increase** robot's e variable, no preconditions
 - **Decrease** robot's e variable, no preconditions
- **Movement:** 16 actions, split between two sets of 8
 - In **low energy** zones: Up, Down, Left, Right, and diagonals
 - In **high energy** zones: Up, Down, Left, Right, and diagonals

Possible Actions

- Separating movement into high and low energy sets induces the planner to **perform energy actions** before entering or exiting a high energy zone
- As the energy is minimized at each plan step, the planner will **choose plans with less energy actions**, thus avoiding movement actions in the high energy set

Practical Implementation

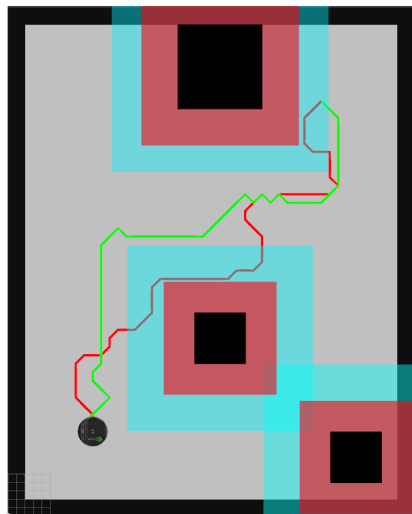
- To test the solution in an autonomous navigation application, a **ROS package** (*ros_enhsp*) **was implemented**
- It wraps a numerical planner called **Expressive Numeric Heuristic Search Planner** (ENHSP)
 - Supports linear and non-linear numerical expressions (e.g., euclidean distance)
 - Supports planning constraints and metric minimization (in this case, energy)

Experiments and Results

- To evaluate the planning domain and implemented package, we perform **two experiments**
 - ① To see **if energy minimization significantly changes the plans**
 - ② To measure **how longer the robot's battery life is extended** with our solution, compared to a typical ROS path planning solution
- The experiments are run in the Stage simulator:
 - Simulated world
 - Simulated Turtlebot 2 robot
 - Running on an **NVIDIA Jetson TX2**, powered with a **11.1V 1300 mAh LiPo battery**

Experiment 1

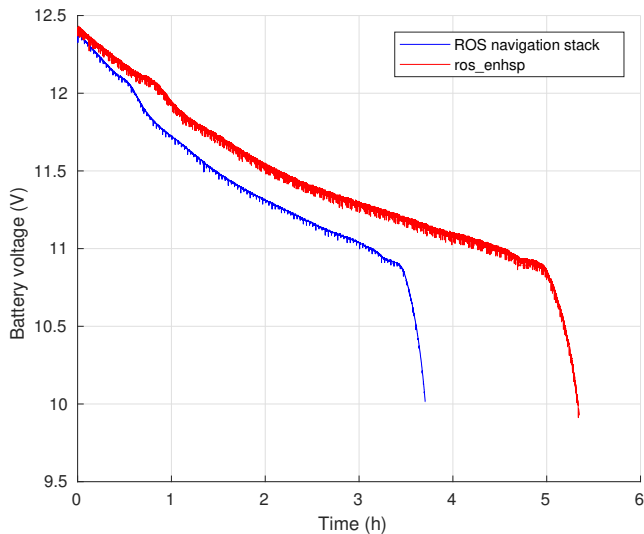
- Red path: **Plan ignoring the energy metric**
- Green path: **Plan considering the energy metric**



Experiment 2

- **Task:** Autonomous navigation
 - The robot must visit 4 user-defined waypoints while autonomously moving and avoiding obstacles
- With a fully charged battery, **we run the task in loop**, once with the conventional path planning and once with our implemented solution, **until the battery runs out**
 - The battery voltage is measured throughout the runs with the Jetson TX2's internal power sensors
 - In `ros_enhsp`, the energy actions correspond to switching the Jetson TX2's **operation mode**

Experiment 2



Related Work

- Energy efficiency is often a bonus of optimal path planning, and **few works consider energy consumption** as a key aspect in their planning domains [Stentz, 1994, Kruger et al., 2007, Mei et al., 2006, Cabreira et al., 2018]
- For example, [Ooi and Schindelhauer, 2009] proposes a path planner to **minimize energy consumption for both mobility and communication** in a robot, by considering the **distance to the goal** as well as the **transmission power** required for communication

Related Work

- [Plonski et al., 2013] use dynamic programming to find **energy-minimal paths** for a solar-powered ground robot, based on a **power draw model** and a "**solar map**" of an environment
- [Franco and Buttazzo, 2015] explore **energy-aware coverage path planning** using an energy model derived from real measurements.

Conclusions

- We proposed a path planning solution which **accounts for high energy zones** and **integrates energy-changing actions** in the plans
 - We model a STRIPS domain to represent obstacles as geometric spaces to be avoided
 - High energy zones are avoided by minimizing the robot's energy at each plan step
- We **integrate the domain in a ROS package** and test it in two experiments
 - We verify the **significance of energy minimization** in the plans
 - We compare the battery discharge curve of our solution to a conventional path planner, **extending battery life by 1.5 hours (42.8%)**

Limitations

- Our solution has **three main limitations**:
 - Compared to the conventional path planner, our solution's **planning time is longer**
 - We assume that **the robot's response time is directly related to the operating mode**, which may not be true
 - The `ros_enhsp` package is **specific to the Jetson TX2** embedded computer

Future Work

- Test the solution in a **real environment and robot**;
- Assert the assumption that **response time and operating mode are directly related**;
- Build a tool to **create geometric space representation** for obstacles directly from ROS' occupancy grid maps;
- **Define other energy actions** (e.g., switching off an unnecessary sensor), to maximize energy efficiency in a generic robot;
- Exploit ENHSP's *processes* abstraction to **model actions as linear/angular velocity** instructions;
 - Possibility to integrate automatic control in the planning domain

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