Real-time Adaptive Robot Motion Planning in Unknown and Unpredictable Environments

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Outline

- Problem and Context
- Basic RAMP Framework
- Extension and Application of RAMP
 - OMultiple Mobile Manipulators
 - ONONholonomic Robot Vehicle
 - OMOBILE Robots in Pursuit and Evasion
 - Continuum Manipulator

Basic Problem

- Plan the motion of a robot from a starting location to reach a goal in an environment with unknown obstacles or unknown obstacle motions.
- The robot can have high degrees of freedom and redundancy (e.g., a mobile manipulator).
- Robot motion *planning has to be on-line* based on sensing.

Motion planning for robot manipulators (DOF \ge 6): existing work

- Off-line planning in known environments using sampling-based approaches, such as variants of PRM and RRT.
- On-line modifying pre-planned paths in largely known environments to avoid new obstacles or obstacle motion.
- OMotion objectives or goals are often fixed.
- Relatively little work on real-time motion planning in unknown and unpredictable environments.

Real-time Adaptive Motion Planning (RAMP) Paradigm [Vannoy&Xiao 04-08]

- Interweave planning, sensing, and execution of motion:
 - plan path and trajectory simultaneously
 - plan and execute motion simultaneously
- Global planning is inspired by the general anytime and parallel idea of evolutionary computation
- Adapt or change motion objectives on the fly based on need

Information from Sensing

- Target objects, obstacles, and their poses at each sensing instant. Obstacles may not have to be segmented or identified.
- Two types of approaches to handle obstacles of unknown motion:
- Predict short-term obstacle trajectory based on time history of sensed obstacle poses – requiring obstacle identification
- Perceive if a robot at a pose at a future time will not collide with any obstacle no matter how obstacles move based on the concept of dynamic envelopes [Vatcha&Xiao08] – not requiring obstacle segmentation or identification

RAMP Algorithm

Initialize a set of trajectories P leading to the goal(s) and evaluate P based on initial sensed information γ← best trajectory in P;

while goal is not reached do

simultaneously **move** and **plan**:

move along γ unless forced to stop to avoid collision;

plan:

modify P

if <u>new sensing cycle</u> then evaluate P;

if <u>new adaptation cycle then</u>
update starting configurations of P;
evaluate P and update γ;

end while

The algorithm allows the robot to <u>smoothly</u> switch to a better trajectory at any time during execution



Trajectory Initialization

Generate *m* trajectories in *P*:

- Oeither randomly or deliberately generate a path in terms of a sequence of knot configurations in the configuration space of the robot.
- O(optional) generate goal configuration(s)
- Ogenerate trajectory under max. velocity and acceleration constraints
- Trajectory diversity in set *P* is important.

Evaluation

- Optimization criteria are different for completely feasible or partially feasible trajectories.
- Feasible (i.e., collision-free and singularity free): e.g.,
 - Minimize distance or time to goal
 - Minimize energy
 - Maximize manipulability
 - Maximize goodness of goal if there are alternative goals
- Partially feasible: e.g.,
 - Time or distance to first collision or singularity
 - Cost as if feasible + penalty
- Optimization criteria can be changed on the fly based on circumstances to steer planning in more effective directions.

Modification

Modify the shape of a path or create a new path by <u>generically random</u>
 insert, *delete*, *change*, *swap*, *crossover* and some <u>robot-specific random</u> operators

 Modified trajectory is evaluated and replaces randomly a non-best trajectory in P to promote the best while preserving diversity.



RAMP on Mobile Manipulators

Trajectory: cubic-splined for arm and linear-with-parabolic-blends for base

Special modification operator: random stop modifies the velocity profile to allow loose-coupling of the arm and the base for adaptive redundancy resolution

Loose Collaboration

- Multiple Mobile manipulators pick up a large set of objects of unknown quantity and arrangement and move them to some destinations or re-arrange them.
- Every robot has its own RAMP planner and view other robots as obstacles with unknown motions. Real-time, distributed motion planning is achieved.
- Spontaneous division of work each robot decides a target object for pick-up and a destination to put it down on the fly based on availability.

Examples



Tight Coordination

- Planning motion trajectories for two mobile manipulators moving and manipulating an object together through an environment with obstacles of unknown motion.
- Each mobile manipulator in the two-robot team is equipped with its own instance of the same realtime *leader-oriented planner*.



The Leader-oriented Planner

- Each mobile manipulator plans its motion as the <u>leader</u> with RAMP and then the other robot's motion as the <u>helper</u>'s motion constrained by the leader's.
- The two robots run their planners in parallel at the same time as they move, but the actual motions that the robots execute are determined by a simple *coordinator* algorithm running on top of the two parallel planners.
- The coordinator <u>constantly</u> decides which robot's leaderhelper motion plan is better and let the two-robot team to execute the better motion plan.
- At any time, the roles can be switched seamlessly as the robots always follow the better trajectories.

RAMP on Nonholonomic Robot Vehicle

- A robotic vehicle moves autonomously in an unknown environment guided by a GPS navigator and local sensing.
- The GPS system does not indicate the actual geometry of the road as well as small obstacles.
- The robot has to conduct on-line planning based on local sensing to produce collision-free nonholonomic trajectories for it to follow.

Approach

 Introduce a set of parameterized, basic maneuver patterns that our RAMP planner can use to build nonholonomic trajectories with arbitrary turns quickly in a piece-wise fashion.

 The maneuver patterns use Bezier curves, which allow both forward and backward driving of a vehicle with great flexibility,

- can be modified analytically via control points an advantage over clothoids, and
- Illow higher vehicle speeds than Reeds & Shepp curves.

Special modification operators

Assumptions

- A sequence of knot points are given (from GPS) that capture the topology of the path.
- Environment is unknown, but obstacles are visible if within sensing range.

Implementation Results



Note the quasi-static moving obstacle above

RAMP on Pursuit and Evasion

- Two mobile robots: one pursuer and one evader.
 Each moves in an initially unknown environment and relies on line-of-sight sensing to discover the environment and the other agent.
- The pursuer aims to catch the evader without knowing its motion and destination.
- The evader aims to reach its destination as quickly as possible while avoiding being caught by the purser.

Approach

- Each agent has a RAMP-based motion planner. Each agent is unaware of the other's motion strategies.
- Optimization objectives are dynamically changed based on an agent's need:
 - escaping, goal-seeking, or hiding for the evader, and
 - chasing based on recent sighting or exploring possible evader locations for the pursuer.



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RAMP on Continuum Manipulators

A continuum manipulator is inherently smooth and compliant and can deform almost everywhere.



A three-section OctArm designed by Walker *et al.* at Clemson University

Configuration variables for each section: κ , *s*, ϕ

Special Features

- Goals must be generated and adapted along with the motion to reach a goal by the planner.
- Path/trajectory evaluation includes a unique distance metric and a goodness measure for goals.
- Special modification operators: repair and curl-up.



Some real experiments

Real Experiment #1

Vmax: 1 cm/s



7-DOF manipulator in unknown environment with stereo vision sensing. Obstacles are not segmented/identified.

Conclusions

- Introduced a general RAMP paradigm for on-line motion planning of high-DOF robots or teams of robots in unknown and unpredictable environments.
- Further work is on effective sensing and real-world testing of RAMP.
- Extend RAMP to incorporate compliant motion, especially for continuum robot manipulation.

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