TBD

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Abstract

To be completed...

2 1 Introduction

2 Problem Statement and Notations

- 4 to formalize well inspired by how this motion planning with obstacles problem is presented in other
- 5 papers
- 6 Motion planning Problem description: Let there be an environment of \mathbb{R}^d where d=2 or 3; obsta-
- 7 cles $\mathcal{O}_i \in \mathbb{R}^d$, $i = \{1, 2, \dots, n\}$, a robot of geometry \mathcal{B} , denote the configuration of the robot as
- 8 $q \in \mathbb{R}^n$, where n is the number of Degree of Freedom (DoF) of the robot. Let the robot has controls
- 9 $u \in \mathcal{U}$, and let there be start $s \in \mathbb{R}^n$ and goal $g \in \mathbb{R}^n$; find a sequence of robot configurations or a
- sequence of controls so that the robot can go from s to g. We can assume there is an oracle function
- 11 $\mathcal{F}: (q \times \cup \mathcal{O}) \to \{0,1\}$ that can return collision detection result in a given environment for any
- 12 given configuration of the robot.
- 13 One common approach is called sampling based motion planning, which is achieved through placing
- samples in \mathbb{R}^n , the configuration space, and retain valid non-collision samples in the configuration
- space by inquiring \mathcal{F} . The invalid samples are discarded (or retained in some cases), and the valid
- samples are connected if the path connecting the configurations is valid (pass the validity check after
- inquiring \mathcal{F}). A good set of samples will lead to solutions much faster compared to random samples,
- though may not always be optimal.
- 19 If we want to use a learning method to create the samples, with bias, we may be able to find paths
- 20 faster.

1 3 Using an Energy Learning Approach

- 22 Formalize and state the learning problem.
- 4 Conclusion

24 A Appendix