

Reinforcement Learning for Tracking Control in Robotics

Yudha Prawira Pane

Literature Survey

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LITERATURE SURVEY

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The implementation work in this thesis was done at DCSC's robotics lab.



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Abstract

This is an abstract.

Table of Contents

Preface	ix
Acknowledgements	xi
1 Introduction	1
1-1 Problem Definition	2
1-2 Goal of the Thesis	2
1-3 Literature Study Approach	2
1-4 Nomenclature	2
2 Reinforcement Learning Preliminaries	3
2-1 Markov Decision Process	3
2-2 Value and Policy Iteration	3
2-3 Reinforcement Learning for Continuous Space	3
2-3-1 Function Approximation	3
2-4 Actor-Critic Structure	3
3 Reinforcement Learning for Tracking Problem: A Survey	5
3-1 Dynamic Tuning via Reinforcement Learning	5
3-1-1 Case Study: PI Tuning using Reinforcement Learning	5
3-2 Nonlinear Compensation for Tracking via Reinforcement Learning	5
3-2-1 Case Study: 1-DOF Robot Gravity Compensation	5
3-3 Reinforcement Learning for Optimal Tracking Control	5
3-4 Self-Proposed Controller [tentative]	5
4 Simulation & Verification	7
4-1 Simulated Setup	7
4-2 Simulation Result and Analysis	7
4-3 Discussion	7

5 Future Work and Experiments Plan	9
5-1 Experimental Setup: UR5 Robot	9
6 Conclusion	11
A Appendix	13
A-1 Simulation Program	13
A-1-1 A MATLAB listing	13
Glossary	15
List of Acronyms	15
List of Symbols	15
Bibliography	15

List of Figures

List of Tables

Preface

According to WIKIPEDIA, a preface (pronounced “*preffus*”) is an introduction to a book written by the author of the book. In this preface I can discuss the interesting story of how this thesis came into being.

This document is a part of my Master of Science graduation thesis. The idea of doing my thesis on this subject came after a discussion with my good friends Tweedledum and Tweedledee...

Acknowledgements

I would like to thank my supervisor for his assistance during the writing of this thesis...

By the way, it might make sense to combine the Preface and the Acknowledgements. This is just a matter of taste, of course.

Delft, University of Technology
December 15, 2014

Yudha Prawira Pane

“In the future, airplanes will be flown by a dog and a pilot. And the dog’s job will be to make sure that if the pilot tries to touch any of the buttons, the dog bites him.”

— *Scott Adams*

Chapter 1

Introduction

Reference or trajectory tracking is one of the building blocks to perform a complex task in robotics. Given a desired path/trajectory, the robot must be able to follow it as quickly as possible with minimum error. Capability to perform this precise tracking is crucial for robots that are to be deployed at manufacturing industries such as semiconductor, automotive, and recently, the emerging application of 3D printing.

Statistics by International Federation of Robotics (IFR) [1] shows that the global sales of industrial robots continues to increase steadily. In 2014, it is expected that the total number of industrial robots installed reaches 205,000 units, a rise of approximately 15 % from previous year. The survey points out that the mature markets such as automotive, electronics, and metal are responsible for such growth.

Meanwhile, there is also a growing interests in applying robots to relatively new applications such as 3D printing, architecture, and art. For instance, research done by Gramazio et. al [2] aims to push the capability of industrial robots to make the direct fabrication based on CAD model a reality. The advantage of using robots over conventional CNC machines lies on their flexibility, easy-to-adapt feature, and high degree of freedom (DOF) to enable execution of difficult configuration in 3D space. These aforementioned applications demands high precision since a minuscule of error could lead to a defect product or even worse, a disaster. Therefore, a precise, accurate reference tracking capability is inevitable.

In order to achieve this, a tracking controller is needed. However, robots are identical with non-linearities, noises, and external disturbances that are difficult to model, let alone compensate. This unknown properties often hinders the controller to perform optimally. A class of controllers which solely depends on the system's model will surely suffer a poor tracking accuracy. The natural answer to this problem is to introduce a controller capable of adjusting its parameter overtime by comparing the reference to the actual trajectory. By doing so, the controller will have an extra degree of freedom to compensate for the unknown properties hence improves the tracking quality. The controller of such characteristic belongs to the class of adaptive controller.

In this thesis, we will propose a method to improve the performance of nominal controller by using reinforcement learning. RL can

1-1 Problem Definition

The fundamental problem in this literature study concerns with the non-optimal performance of nominal controller with respect to reference tracking task. Therefore the research question can be raised as follows.

"Is it possible to integrate Reinforcement Learning technique to a nominal controller in a certain structure such that reference tracking performance of the controlled system significantly improves?"

While conducting a research, it is considered wise to restrict oneself to a simpler context, but still captures the essential elements of the original problem. Therefore, in answering this question, some general assumptions are made.

1. The system to be controlled is fully actuated
2. The system to be controlled is both controllable and observable
3. Nominal, stabilizing controller is available

1-2 Goal of the Thesis

The goal of this thesis are as follows:

1. To provide a general framework of improving tracking control using reinforcement learning
2. To apply and compare existing method of RL for tracking application to the 3D printing robotic setup
3. To come up with modification of improvement of previous methods

1-3 Literature Study Approach

1-4 Nomenclature

Reinforcement Learning Preliminaries

2-1 Markov Decision Process

This chapter will cover figures and math.

2-2 Value and Policy Iteration

2-3 Reinforcement Learning for Continuous Space

2-3-1 Function Approximation

2-4 Actor-Critic Structure

Reinforcement Learning for Tracking Problem: A Survey

This is real chapter for Delft Center for Systems and Control (DCSC), ok? We will use it as a demo for the different headings you can use to structure your text.

3-1 Dynamic Tuning via Reinforcement Learning

This is the first section .

3-1-1 Case Study: PI Tuning using Reinforcement Learning

This is the subsection of the first section.

3-2 Nonlinear Compensation for Tracking via Reinforcement Learning

This is second section.

3-2-1 Case Study: 1-DOF Robot Gravity Compensation

3-3 Reinforcement Learning for Optimal Tracking Control

This is third section.

3-4 Self-Proposed Controller [tentative]

Simulation & Verification

4-1 Simulated Setup

This chapter will cover figures and math.

4-2 Simulation Result and Analysis

4-3 Discussion

Future Work and Experiments Plan

5-1 Experimental Setup: UR5 Robot

Chapter 6

Conclusion

Appendix A

Appendix

Appendices are found in the back.

A-1 Simulation Program

A-1-1 A MATLAB listing

```
1 %  
2 % Comment  
3 %  
4 n=10;  
5 for i=1:n  
6     disp('Ok');  
7 end
```

Glossary

List of Acronyms

DCSC Delft Center for Systems and Control

