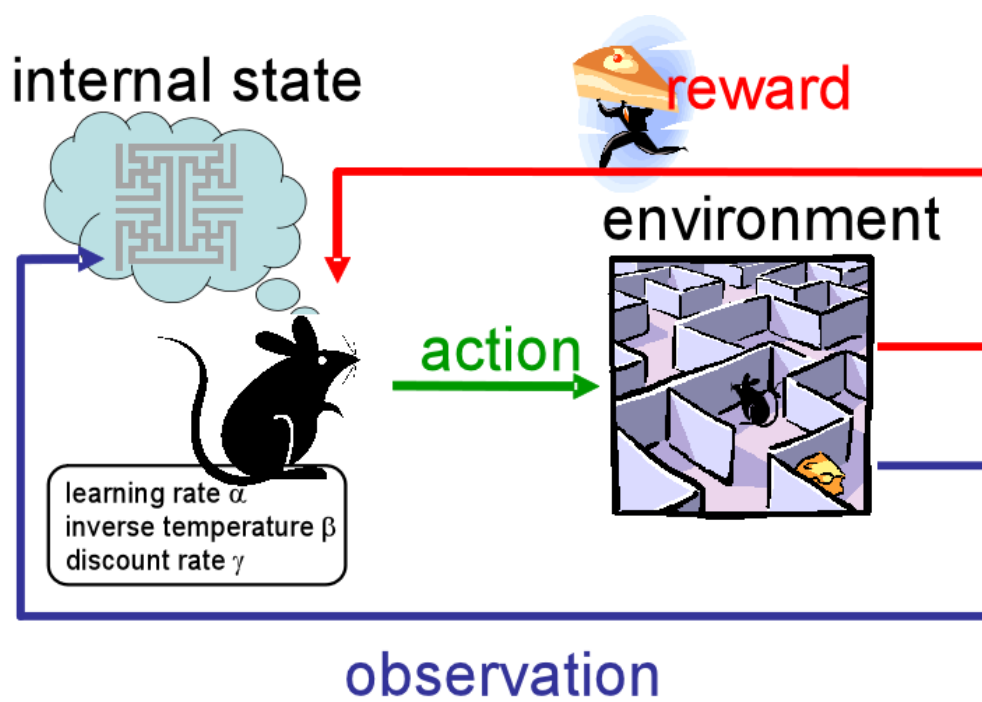




Aarhus University Department of Engineering 2017

Thesis

Title of project



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Project title:

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Contents

1	Introduction	1
2	Theoretical Background and State of the Art	2
2.1	Theoretical Background in Reinforcement Learning	2
2.2	Previous Research	2
2.2.1	Atari 2013	2
2.2.2	Go 2016	2
2.2.3	A3C 2016	2
2.2.4	Framework 2017	2
3	Project Framework	3
3.1	Convolution Neural Networks	3
3.2	Recurrent Neural Networks	4
3.2.1	Long Short Term Memory	4
	Bibliography	5

Chapter 1

Introduction

Getting in touch with the most promising and hottest topic in artificial intelligence represents a challenge

To be continued...

Chapter 2

Theoretical Background and State of the Art

2.1 Theoretical Background in Reinforcement Learning

Reinforcement learning is an approach in artificial intelligence for goal-directed learning from interaction, which makes it different from the other approaches in machine learning.

2.2 Previous Research

...

2.2.1 Atari 2013

...

2.2.2 Go 2016

...

2.2.3 A3C 2016

...

2.2.4 Framework 2017

...

Chapter 3

Project Framework

This project is about learning a car or robot to control and navigate it self. This should be done so the robot don't hit walls or obstacles. To do this a system is created. This system is created as inspiration from [1] Can be seen on Figure 3.1.

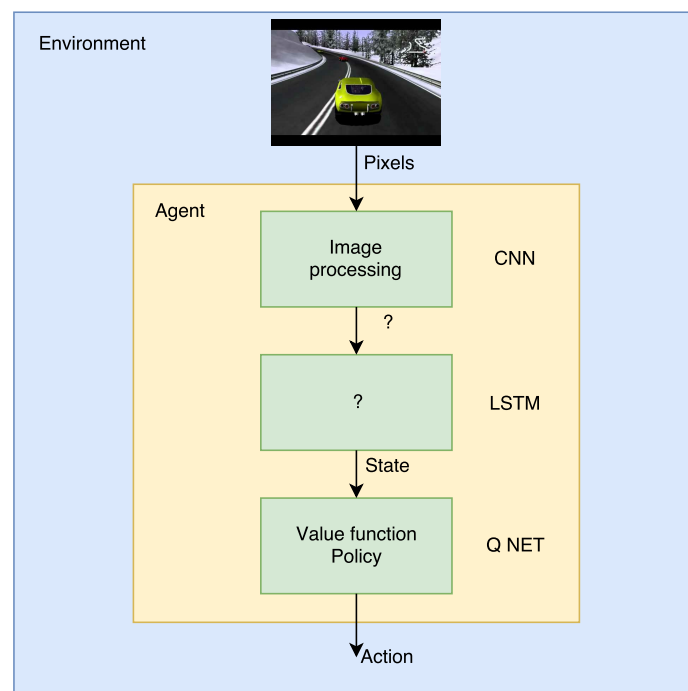


Figure 3.1: *The block diagram of the system*

3.1 Convolution Neural Networks

CNN is here

3.2 Recurrent Neural Networks

...

3.2.1 Long Short Term Memory

...

Bibliography

- [1] V. Mnih, A. P. Badia, M. Mirza, A. Graves, T. P. Lillicrap, T. Harley, D. Silver, and K. Kavukcuoglu, “Asynchronous methods for deep reinforcement learning,” *CoRR*, vol. abs/1602.01783, 2016. [Online]. Available: <http://arxiv.org/abs/1602.01783>