

# An investigation of deep imitation learning for mobile robot navigation

Shubham Juneja Supervisor: Dr. Virginijus Marcinkevičius Semester 1

## Plan of studies & implementation summary

Study year	Exams		Conference participati		Publications		
	Planned	Completed	Planned	Completed	Planned	Completed	
l (2020/2021)	2	1					
II (2021/2022)	2	0					
III (2022/2023)			1	0	1	0	
IV (2023/2024)			1	0	1	0	

### **Report of activity plan**

Exams		<b>Conference Partic</b>	Publications		
Planned	Status	Planned	Status	Planned	Status
Machine	Passed with	Participation in	Planned to participate in ALLSENSORS	<b>Review</b> of	
Learning	score of 9/10	conferencein	2021 conference	researchon	
		Lithuania	https://www.iaria.org/conferences2021/ALL	topic of the	
			SENSORS21.html	dissertation (In	
				conference	
				proceedings)	
Research	Planned to				
methods and	complete until				
methodology	September				
of informatics	2021				
and computer					
engineering					

#### Workshops participated in

Workshop	ECTS
MOKSLINIŲ REZULTATŲ PUBLIKAVIMAS PAGAL FORMALAUS VERTINIMO REIKALAVIMUS	0.1
MOKSLINĖS INFORMACIJOS IŠTEKLIAI, PAIEŠKA, IR ĮRANKIAI	0.1
MENDELEY PRAKTINIS UŽSIĖMIMAS	0.15
Total:	0.35/3

## Stages of research and dissertation preparation

	Name of task	Duration	Notes
1.	<ol> <li>Review and analysis of scientific research on the topic of the dissertation (in Lithuania and abroad):</li> <li>Defining and describing the objectives of the dissertation research topic.</li> <li>Overview of deep imitation learning and deep reinforcement learning for mobile robot navigation.</li> <li>Summary of methods overview and presentation on the description of the analytical part of the dissertation.</li> <li>Formation of research goal.</li> </ol>	September 2020 – August 2021	Performed literature review on imitation learning and navigation methods, which is currently ongoing as planned.
2.	<ul> <li>Carrying out research:</li> <li><b>2.1. Development of research methodology:</b></li> <li>1. Identification and specification of problems arising in currently available methods.</li> <li>2. Specification of tasks to conduct which address to identified problems.</li> <li>3. Specification of navigation environments which will be analysed further.</li> <li>4. Selection of appropriate research methodology.</li> <li>5. Planning of theoretical and empirical research.</li> </ul>	September 2021 – October 2021	
	<ul> <li>2.2. Theoretical research:</li> <li>1. Analysis of reactive imitation learning methods for sensorimotor control and strategy functions, which utilize deep neural networks, such as behaviour cloning, inverse reinforcement learning, generative adversarial imitation learning, etc.</li> <li>2. Research on new reactive mobile robot navigation trajectory controller, based on learning from experience (e.g. imitation learning, reinforcement learning).</li> <li>3. Research on hierarchical goal-directed visual navigation system for mobile robots, based on aforementioned reactive component.</li> </ul>	October 2021 – April 2022	

#### **Research Object and Aim**

Research object:

- Deep imitation learning methods.
- Application of deep imitation learning methods for mobile robot navigation.

Research aim:

• To develop, implement and research an autonomous navigation system for mobile robots based on imitation learning and deep neural networks

#### **Objectives of Research**

- 1. To **develop and investigate** new sensorimotor reflex algorithms based on deep neural networks and various simulation learning paradigms (e.g. behaviour cloning, generative adversarial imitation learning) (e.g. trajectory following, obstacle avoidance, approach to a recognized object).
- 2. To **compose and implement** a new navigation system for mobile robots from the obtained sensorimotor reflexes.
- 3. To **compare** the obtained navigation system with alternative robot navigation algorithms.
- 4. To **prepare publicly available datasets** for the research of autonomous robot navigation algorithms based on the principles of deep neural networks and imitation training.

#### About the problem to solve

- Learning sensorimotor skills to drive and navigate based on visual input.
- It can be done with traditional methods such as SLAM, but it would require expensive sensors and extensive programming.
- The idea of imitation learning promises to solve this problem by learning from human demonstrations.
- Yet, it remains unsolved due the unpredictability of the real world causing the problem of covariate shift.
- To compare the ability between methods NoCrash benchmark has been established.
- NoCrash benchmark uses CARLA simulator to seed vehicles in different parts of a map and tests the ability of reaching from point A to B, under different sets of conditions.



#### What has been carried out so far

- Literature study from papers on imitation learning for mobile robot navigation
- Took courses on Machine learning (at VU) and Reinforcement learning (Online)
- Trying out Simulators (CARLA and OpenAI gym)
- Attempted to run state of the art methods in simulation

#### Takeaways

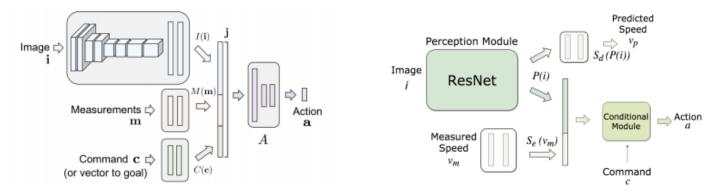
- Mobile robot navigation is being tried to solve in various ways, i.e. modular pipelines, end to end supervised learning, deep reinforcement learning, etc.
- Mainly mobile robots require following of trajectories to go from point A to B and avoid obstacles.
- The problem of urban driving seems to be solving a similar issue, but some research focusing on mobile robot navigation doesn't acknowledge the methods of urban driving, while some of the research acknowledges and utilizes them. And vice versa.
- As per imitation learning, there are methods in both areas of research which utilize imitation learning.
- CARLA seems to be a simulator common between trajectory following focused methods (which intersect with urban driving methods).
- Meanwhile some of the research prefers not to utilize simulators and preferring real world data to report results, leading to forming their own datasets.

#### **Methods for Trajectory Following**

- Conditional Imitation Learning (CIL, 2018) uses imitation learning with high level commands conditioned to the input to learn the skill of trajectory following.
- Conditional Affordance Learning (CAL, 2018) learns affordances in the form of low dimensional intermediate representations from videos, while conditioning with high level commands.
- Conditional Imitation learning with Resnet and speed branch (CILRS, 2019) is an extension of CIL with change in neural network architecture and using a separate branch to predict speed.
- Learning by Cheating (LBC, 2019) proposes training an agent in a twostep process, once with privileged information and once from a teacher network without privileged information.

#### Methods for Trajectory Following

- Implicit Affordances (IA, 2020) uses a encoder to learn to predict affordances and then uses reinforcement learning to learn to navigate based on the affordances.
- Affordances based reinforcement learning (IRL, 2021) experiments with combining implicit and explicit affordances and training with reinforcement learning.



### **Comparison of report results**

Traffic levels	CIL	CAL	CILRS	LBC	ΙΑ	AT (Autopilot)	IRL
Empty	48 ± 3	36 ± 6	51 ± 1	100 ± 0	99	100 ± 0	100 ± 0
Regular	27 ± 1	26 ± 2	44 ± 5	94 ± 3	87	98 ± 1	98 ± 1
Dense	10 ± 2	9 ± 1	38 ± 2	51 ± 3	42	60 ± 3	91 ± 1

Table shows results reported by Agarwal, T., Arora, H., & Schneider, J. (2021). Affordance-based Reinforcement Learning for Urban Driving. *arXiv* preprint arXiv:2101.05970

### Work plan for rest of the year

Review and analysis of scientific research on the topic of the dissertation (in Lithuania and abroad):

- Defining and describing the objectives of the dissertation research topic.
- Overview of deep imitation learning and deep reinforcement learning for mobile robot navigation.
- Summary of methods overview and presentation on the description of the analytical part of the dissertation.
- Formation of research goal.

Passing exam:

- Research methods and methodology of informatics and computer engineering Publication plan:
- Review of research on topic of the dissertation (in conference proceedings)
   Conference Participation:
- Participation in ALLSENSORS conference.



## Thank you