

Week 1: Course Overview

10.02.2016

Wednesdays 14:00 – 17:00, BM A5

Emre Ugur, ETA-36

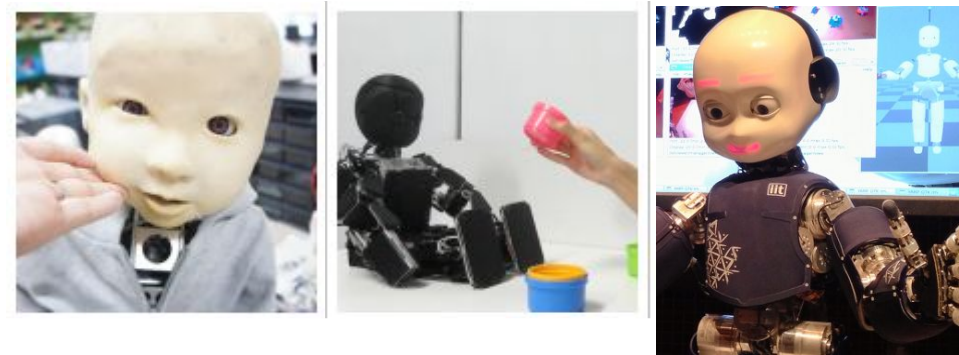
emre.ugur@uibk.ac.at

<http://emreugur.net/courses/robot-learning.html>

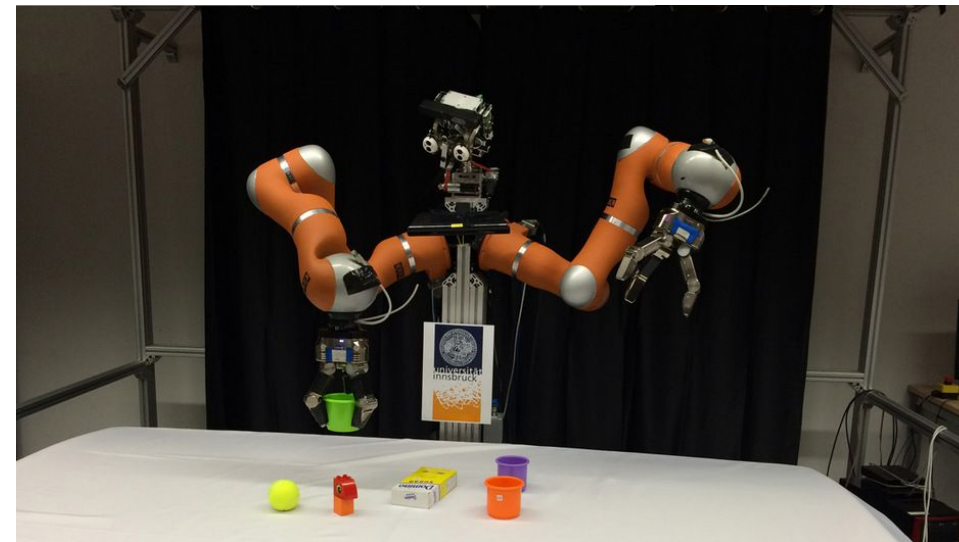
► Bogazici University



► Osaka University



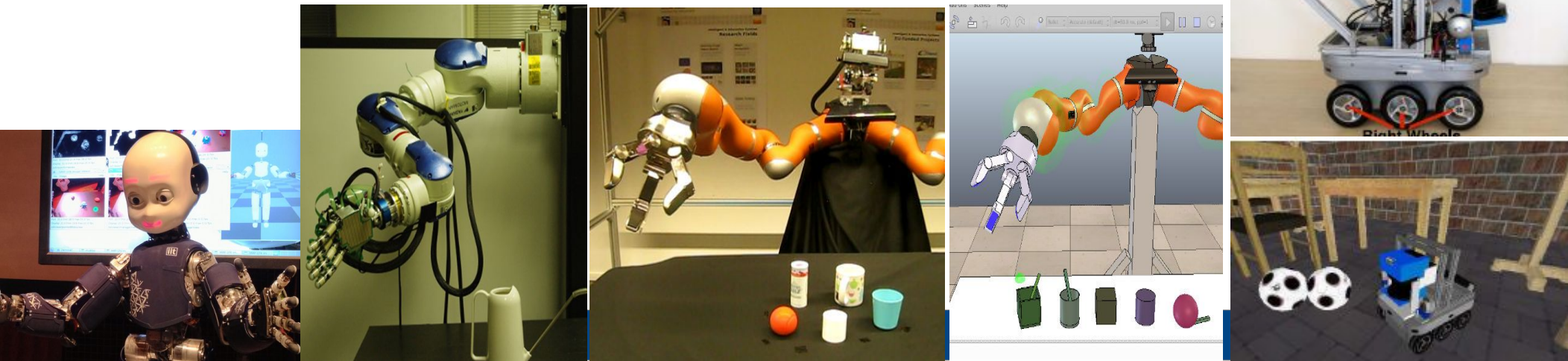
► Innsbruck University



learning environment robot
 through traversability features over
 relations information prediction interactions learned
 performance learn world shape perception hand training
 position distance relevant experts imitation real
 method execution end range studies
 effect image phase al effects perceptual predicted
 parameters Robotics state entity initial
 system next prov example robots robot's space control
 primit feature approach development results cm
 computed new action number vector actions
 complex relation interaction human Psychology
 developmental grasp planning affordances
 behavior affordance category

WordItOut

My topics



Aim or why do we need learning

What type of learning



- ▶ High-level understanding
- ▶ Advanced sensorimotor skills

Aim

- ▶ For high-level understanding:
 - ▶ Learning like babies: Developmental robotics
 - ▶ Exploring smart: Intrinsic motivation
 - ▶ Learning from mistakes: Reinforcement learning
 - ▶ Acquiring higher-level abstractions: Symbol acquisition
- ▶ For advanced sensorimotor skills:
 - ▶ Learning by demonstration
 - ▶ Policy search
 - ▶ Grasping
 - ▶ Probabilistic modeling
 - ▶ Deep learning



Format

- ▶ Mixed lecture-seminar format, instructor leads
- ▶ Basics and concepts are introduced by the instructor.
- ▶ Research papers are presented by students.
 - ▶ One participant presents the paper & leads discussion
 - ▶ All students write a report on each paper
- ▶ Project implementation
 - ▶ Suggest a completely novel approach or choose replicating and extending one high-impact paper
 - ▶ A final project report written in conference/workshop paper format is expected.
 - ▶ The topics are not limited to the ones discussed in the lectures as long as they are related to robot learning.

Assessment

Project (35%)

Reading assignments 15%
Quizzes 10%

Midterm (15%)

Presentations (15%)

In-class activity (10%)

Project

- ▶ Propose a completely novel approach or choose from existing papers.
 - ▶ Implementation in a simulator.
 - ▶ e.g. learning from demonstration without a real robot.
 - ▶ e.g. learning probabilistic relationships between concepts obtained from robots
 - ▶ Depending on difficulty, replicate or extend.
 - ▶ Final report in conference paper format (TORK?, RSS 2017 workshops?)
- ▶ Better to start soon. Discuss all steps together.
- ▶ Deadlines for proposals and progress reports: TBA

Reading Assignments

Please read the papers before the class and submit an **outline** of paper (via email) prior to the class in PDF format. 1 page

- ▶ The addressed problem.
- ▶ Main contributions.
- ▶ Brief technical details.
- ▶ The relation to prior-work and state-of-the-art.
- ▶ The most important experimental results
- ▶ Do conclusions follow the experimental results?

Quizzes

- ▶ In the beginning of each lecture, a 5-10 min quiz
- ▶ Simple questions about the paper
 - ▶ Trivial to reply if you have read the paper
 - ▶ Please bring your **outline**

Paper presentations

- ▶ Papers will be selected **at least one week** prior to the presentation.
- ▶ Aim for 30 minutes presentation, 15 minutes discussion time
 - ▶ Who are the authors?
 - ▶ What are the problems?
 - ▶ Why are the problems important, interesting, and/or hard?
 - ▶ Who did what before?
 - ▶ What are the key contributions?
 - ▶ How did they do it? - algorithmic and technical details
 - ▶ How does it perform?
 - ▶ What are/might be the limitations?
 - ▶ What is next?
- ▶ Do not limit yourself with the selected paper, do/present some background, works of other groups, etc.
- ▶ Any ideas to have stimulating discussions?

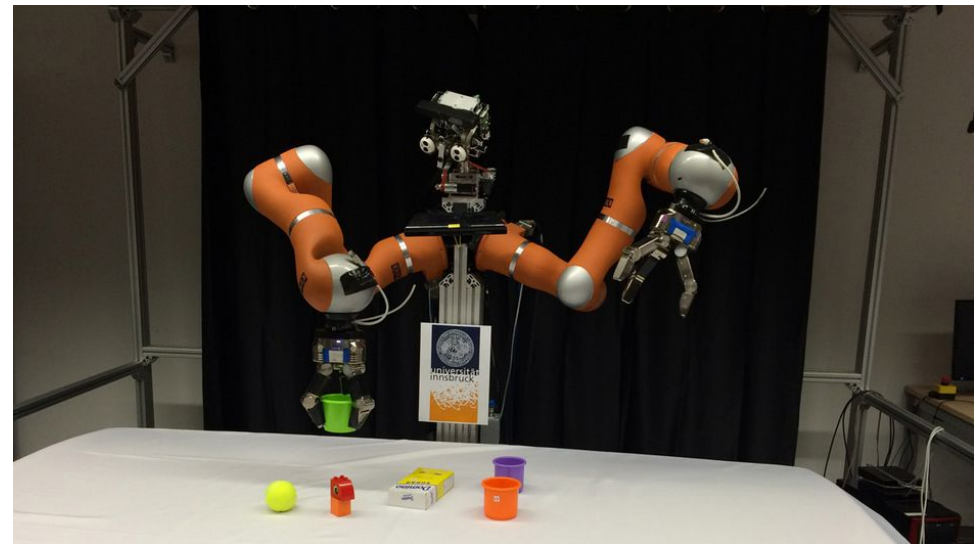
Midterm

- ▶ Responsible from the material covered in the class

In-class activity

- ▶ Active participation to the discussions in the class
- ▶ Evaluation of the presentations

Bonus

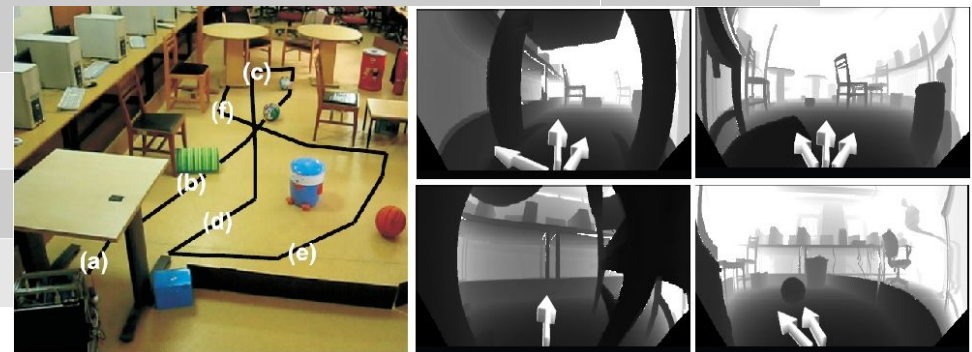
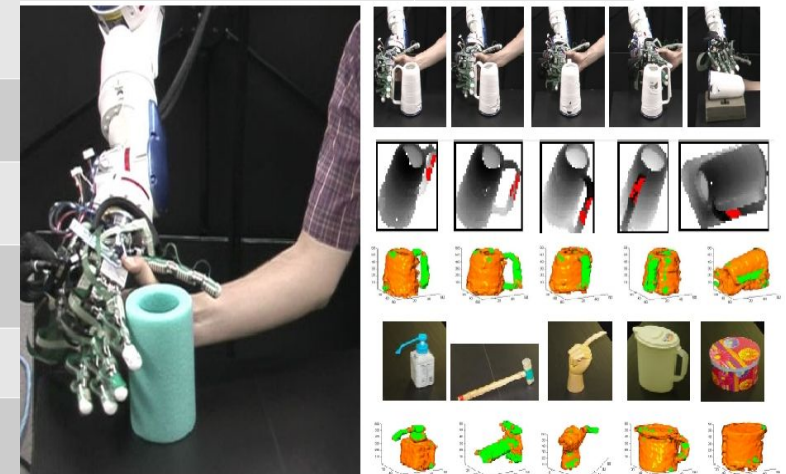
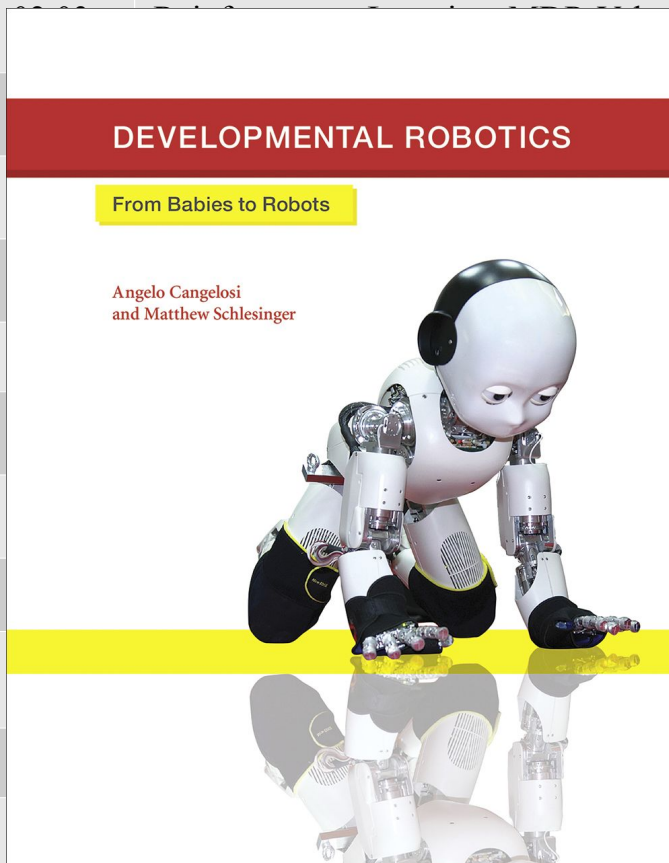


Topics (tentative) – new course new topics

1	10.02	Developmental Robotics, Intrinsic Motivation, Perceptual Learning	I.
2	17.02	Motor Learning, Affordances in Psychology, Neuroscience and Robotics	I.
	24.02	No class	
3	02.03	Reinforcement Learning, MDP, Value Iteration, TD Learning	I.
4	09.03	Reinforcement Learning, Policy Search	I. + S. (2)
5	16.03	Learning from Demonstration – Trajectory & Probabilistic	S.
6	23.03	Learning from Demonstration – Trajectory & Probabilistic	G.+ S.(1) + I.
7	30.03	Grasping and grasp learning	S. (3)
8	06.04	Introduction to Probabilistic Approaches	I. + S. (2)
9	13.04	Probabilistic Approaches (HMMs, MRFs, DBNs)	S. (3)
	20.04	Spring break	
10	27.04	Human-in-the-loop learning / Deep learning	G. + S. (2)
11	04.05	Skill/symbol acquisition	S. (3)
12	11.05	Final project presentations	S.

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4		h	
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7		ies	
8		RFs, DBNs)	
9			S.
10			
11			
12			



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5

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7

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11

12

Reinforcement Learning
An Introduction

Richard S. Sutton and Andrew G. Barto

– Trajectory & Prob

– Trajectory & Prob

Approaches

IMs, MRFs, DBNs)



G.D. Konidakis, S.R. Kuindersma, R.A. Grupen and A.G. Barto.
Autonomous Skill Acquisition on a Mobile Manipulator. In Proceedings of
the Twenty-Fifth Conference on Artificial Intelligence, pages 1468-1473,
August 2011.

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Motor Skill Learning for Robotics

Jan Peters, Jens Kober, Katharina Mülling
Department of Empirical Inference and Machine Learning
Max Planck Institute for Biological Cybernetics

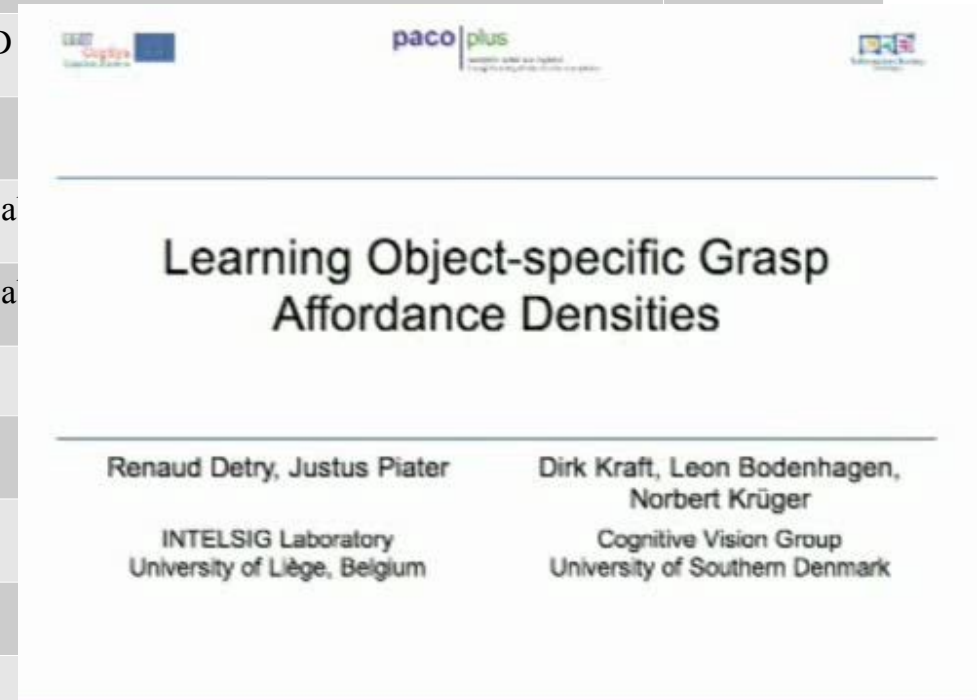
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Anticipating Human Activities for Reactive Robotic Response

Hema Koppula and Ashutosh Saxena
Cornell University

Koppula, Hema S., and Ashutosh Saxena. "Anticipating human activities using object affordances for reactive robotic response." Pattern Analysis and Machine Intelligence, IEEE Transactions on 38.1 (2016): 14-29.

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Supersizing Self-supervision

Learning to Grasp from 50K Tries and 700 Robot Hours

Lerrel Pinto and Abhinav Gupta
The Robotics Institute, Carnegie Mellon University

Pinto, Lerrel, and Abhinav Gupta. "Supersizing Self-supervision: Learning to Grasp from 50K Tries and 700 Robot Hours." arXiv preprint arXiv:1509.06825 (2015).

G. + S.

S.

S.



Topics (tentative)

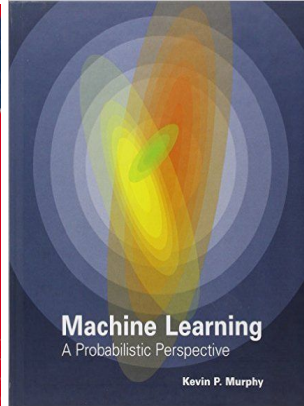
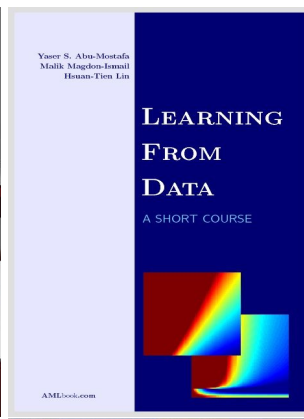
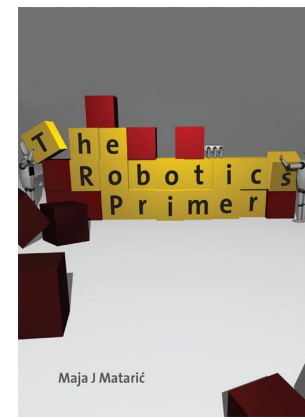
1	10.02	Developmental Robotics, Intrinsic Motivation,	Bottom-Up Learning of Object Categories, Action Effects and Logical Rules: From Continuous Manipulative Exploration to Symbolic Planning Emre Ugur and Justus Piater University of Innsbruck Ugur, Emre, and Justus Piater. "Bottom-up learning of object categories, action effects and logical rules: From continuous manipulative exploration to symbolic planning." Robotics and Automation (ICRA), 2015 IEEE International Conference on. IEEE, 2015.
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Basics

- ▶ Degrees of freedom
- ▶ Depth camera, point cloud
- ▶ Forward Kinematics, Inverse Kinematics
- ▶ Robot, world state
- ▶ Clustering / classification
- ▶ Supervised learning, unsupervised learning
- ▶ Reinforcement learning, Markov Decision Process
- ▶ Support Vector Machine (SVM), Maximum Margin Classifier
- ▶ Bayesian Inference, Bayesian Networks
- ▶ Feed-forward networks, back-prop algorithm



Opinions?

- ▶ Expertise
- ▶ I will need feedback