Robotics

Erwin M. Bakker| LIACS Media Lab

15-2 2019



Universiteit Leiden

Bij ons leer je de wereld kennen

Organization and Overview

 Period:
 February 15th - May 10th 2019

 Time:
 Friday 09.00 - 10.45

 Place:
 LIACS, Room 401 (Workshops Room 303)

 Lecture:
 Dr Erwin M. Bakker (erwin@liacs.nl)

 Assistant:
 Andrius Bernatavicius

NB E-mail your name and student number to erwin@liacs.nl

Schedule:

15-2	Introduction and Overview
22-2	Locomotion and Inverse Kinematics
1-3	Sensors and Algorithms
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Website: http://liacs.leidenuniv.nl/~bakkerem2/robotics/



Grading (6 ECTS): Presentations and Robotics Project (60% of grade). Class discussions, attendance, workshops and assignments (40% of grade). It is necessary to be at every class and to complete every workshop.



Robotics in the News



University of California at Berkeley, Feb. 2011



https://foldimate.com/ , CES 2019

Philo of Byzantium (~280 – 220 BC) Al-Jazari (1136 – 1206)

- Mechanisms and methods for automation
- Water-raising machines
- Clocks
- Automata
 - Drink-serving waitress
 - Hand-washing automaton with flush mechanism
 - Peacock fountain with automated servants
 - Musical robot band

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Leonardo da Vinci (1452 – 1519)

- Robotic Carts
- Studies on locomotion
- Robotic Soldier
- Robotic Lion





Pictures from: http://www.leonardo3.net http://brunelleschi.imss.fi.it

The Turk Constructed and unveiled in 1770 by Wolfgang von Kempelen (1734–1804)



Der Schachforder im Spiele begriffen Le konen Hickers til gein le ont pradant le per

Pictures from: http://en.wikipedia.org/wiki/The_Turk

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EARLY ROBOTS

LOCOMOTION & INVERSE KINEMATICS



South Pointing Chariot by Ma Jun (c. 200–265)

Autonomous Robots for Artificial Life (MIT, T. Braunl, Stuttgart University) 'Rug Warrior'



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Autonomous Robots for Artificial Life

- Sensors
- Bumper
- Photoresistors (2)
- Infrared Obstacle Detectors w. 2 infrared LED's
- Microphone
- Two Shaft-Encoders



Tekening van: http://ag-vp-www.informatik.uni-kl.de

Autonomous Robots for Artificial Life

Software (PC, Macintosh, UNIX) Interactive C Compiler and Libraries

- motor(0,speed), motor(1,speed)
- music: tone(), analog(micro)
- get_left_clicks(), get_right_clicks()
- analog(photo_left), analog(photo_right)
- left_ir, right_ir
- left_, right_, back_bumper



- Note: Microsoft Robotics Studio 4: development environment for different robotic platforms (Lego Mindstorm, Fischertechnik, Lynxmotion, Parallax Boe-Bot, Pioneer P3 DX, iRobot Roomba), Kinect (2014⁺);
- ROS (Robot Operating System) 50+ robots, etc.

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Autonomous Robots for Artificial Life



Straight ahead





Finding the Light



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Finding the Light

```
if ( analog(photo_right) < analog(photo_left) )</pre>
{ motor(0, speed);
                      /* rechtsdrehen */
  motor(1, -speed);
} else
{ motor(0, -speed);
                      /* linksdrehen */
 motor(1, speed);
}
clicks = 0;
while( ( (clicks += (get_left_clicks() + get_right_clicks()) / 2)) < 37
        && !all_bumper ) /* eine Umdregung machen solange kein Bumper
betaetigt */
{ printf("FIND MAX %d %d\n", clicks, light);
 light = get_light();
                               /* Lichtwert holen */
                                /* maximum merken */
 if ( light > max_light )
 { max_light = light; }
  sleep(0.2);
 }
```

Finding the Light 2



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Finding the Light 2

- Drive along the wall until the light source is found.
- Drive with a left curve until the IR-sensors detect an obstacle, then make a correction to the right until no sensor input is read.
- If an obstacle is found that cannot be resolved this way, then drive 1.5 seconds backwards and start over again.



















How to move to a goal?

Problem: How to move to a goal?

• Grasp, Walk, Stand, Dance, Follow, etc.

Solution:

- Program step by step.
- Inverse kinematics: take end-points and move them to designated points.
- Trace movements by specialist, human, etc.
- Learn the right movements: Reinforcement Learning, give a reward when the movement resembles the designated movement.



https://pybullet.org/wordpress/





ROBOTICS SENSORS

- Bumper switches
- Acceleration, Orientation, Magnetic
- IR/Visible Light
- Pressure, Force
- Ultrasonic, Lidar, Radar
- Camera's, stereo camera's
- Structured Light Camera's



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Location & Navigation

Problem:

How to locate yourself? How to navigate?

- In unknown or known environment. With sensors:
- internal, passive, active, gps, beacons, etc.
 With or without reference points.

Solution:

- Collect data to determine starting position, or determine your location.
- · Move around while collecting data from your environment.

- OpenCV.org
- Sensor data is noisy => location and map building is a stochastic process.
- SLAM



Landmark Navigation

Only proprio-perception

- dead reckoning
- no-dead reckoning (no known starting position)
- drift errors

With extero-ception

- global odometry
- corrections by sighting of known environmental features, etc.









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IMAGE PROCESSING & UNDERSTANDING

Problem: How to understand your environment?

- · What are obstacles?
- What are the objects?
- Can you locate the objects you have to manipulate?
- Humans, voice, speech, and emotion recognition?

Solution:

- Basic image processing.
- Content based image understanding, classification.
- Speech recognition, speech synthesis.
- Multimodal data processing.
- Natural Language Processing, etc.





Robotics at LIACS Media Lab

Content Based Retrieval

Robotics Research

- · Vision, audio, and passive sensor based
- · Environment and object recognition and understanding
- Paradigms: DNN, RL and ARL

• ...

Education

- Multimedia Systems
- Multimedia Information Retrieval
- Robotics
- · Robotic Projects and many more

Robotics Challenges

DARPA Challenges

- · Autonomous vehicles
- Search and Rescue
- SHRIMP (https://www.darpa.mil/news-events/2018-07-17), Etc.

Robocup

- Goal: A team of robots defeating the human world champion in 2050.
- Yearly event
- Different Leagues:
- Simulation league
- Small-size, Mid-size
- Aibo-, Nao-, Humanoid League



18 WEELOG

scale robots

DARPA has competition plans for insect-

Robocup 2018 Montreal



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DARPA Challenge 2005

\$2 million Prize awarded to

the Stanford Racing Team

Five teams completed the Grand Challenge; four of them under the 10 hour limit. The Stanford Racing Team took the prize with a winning time of 6 hours, 53 minutes.











Outer human loop (counter-clockwise)

Inner human loop (clockwise)

Robot loop

URBAN Challenge 2007

A new DARPA challenge, which took place at the George Air Force Base on a 96 km urban area course, to be completed in less than 6 hours, obeying all traffic regulations while negotiating with other traffic

Results

- 1. Tartan Racing (\$2 million), a collaborative effort by Carnegie Mellon University and General Motors Corporation, with vehicle "Boss", a Chevy Tahoe (average 14 miles/hour)
- 2. Stanford Racing Team (\$1 million) with vehicle "Junior", a 2006 Volkswagen Passat (average 13 miles/hour).
- 3. Victor Tango from Virginia Tech (\$500,000) with their 2005 Ford Escape hybrid, "Odin" (average 13 miles/hour)

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Complex Tasks

Problem: How to train on complex tasks?

- Grasping different objects, clothes.
- Interact with people in natural environments, drive a car, etc.

Solution:

• Training in the real world.

Draw backs: Too costly, too little iterations, etc.

Deep neural networks success in the field of Reinforcement Learning:

- Fast computations
- Fast Simulations
- Improved networks

But, most RL-based approaches fail to generalize, because:

- 1. gap between simulation and real world
- 2. policy learning in real world is hampered by data scarcity









Robust Adversarial Reinforcement Learning (RARL) to close the Reality Gap

Training of an agent in the presence of a **destabilizing adversary**

- Adversary can employ disturbances to the system
- Adversary is trained at the same time as the agent
- Adversary is reinforced: it learns an optimal destabilization policy.

Here policy learning can be formulated as a zero-sum, minimax objective function.





Figure 8. Visualization of forces applied by the adversary on InvertedPendulum. In (a) and (b) the cart is stationary, while in (c) and (d) the cart is moving with a vertical pendulum.

Actions of Adversary



Figure 9. Visualization of forces applied by the adversary on Hop per. On the left, the Hopper's foot is in the air while on the righ the foot is interacting with the ground.

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Robotics Homework I

Assignment:

Give a link to the coolest, strangest, most impressive, most novel, or technologically inspirational robot you could find.

Due: Thursday 21-2 at 14.00 PM.

Email your link to erwin@liacs.nl with subject 'Robotics'.

References

- 1. L. Pinto, J. Davidson, R. Sukthankar, A. Gupta, Robust Adversarial Reinforcement Learning, arXiv:1703.02702, March 2017.
- 2. S. Gu, E. Holly, T. Lillicrap, S. Levine, Deep Reinforcement Learning for Robotic Manipulation with Asynchronous Off-Policy Updates, arXiv:1610.00633v2 [cs.RO], October 2016.
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- 4. L. Pinto, J. Davidson, A. Gupta, Supervision via Competition: Robot Adversaries for Learning Tasks, arXiv:1610.01685, ICRA 2017, October 2016.
- 5. K. Bousmalis, N. Silberman, D. Dohan, D. Erhan, D. Krishnan, Unsupervised Pixel–Level Domain Adaptation with Generative Adversarial Networks, arXiv:1612.05424, CVPR 2017, December 2016.
- 6. A. Banino et al., Vector-based navigation using grid-like
- 7. representations in artificial agents, <u>https://doi.org/10.1038/s41586-018-0102-6</u>, Research Letter, Nature, 2018.
- 8. R. Borst, Robust self-balancing robot mimicking, Bachelor Thesis, August 2017
- 9. Jie Tan, Tingnan Zhang, Erwin Coumans, Atil Iscen, Yunfei Bai, Danijar Hafner, Steven Bohez, and Vincent Vanhoucke, Sim-to-Real: Learning Agile Locomotion For Quadruped Robots, https://arxiv.org/pdf/1804.10332.pdf, RSS 2018.

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