Probabilistic Methodologies for Autonomous Mobile Robot Localization

Speaker

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Autonomous Mobile Robotics Research Group | D-121

UNIVERSITY OF HARTFORD

COLLEGE OF ENGINEERING, TECHNOLOGY, AND ARCHITECTURE

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Introduction Mapping Model Generation MonoSLAM-R Future Agenda Teaching Exp.

Autonomous Mobile Robotics Research Group

- What do we do? We are a research group focusing on design and development of robotics, industrial automation systems and advanced mechanisms.

- Who are we? We have 32 active members from all majors.

- What do we offer?

Our group offers free courses about:

- Robotics Design,
- Embedded Control,



• Software Development (Arduino, Raspberry Pi, Matlab)

which will be useful for your education and future scientific research.

Our lab is at D-121.

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Autonomous Mobile Robotics Research Group

We have 32 active members from all majors.





		Solid Works	Electronics	Arduino IDE	Linux C++	Matlab	Java	TEAM-I	TEAM-II
1	Ackeifi, Ross		Х	Х	Х		Х	A	D
2	Bandarupalli, Gouthamsai	х		Х		Х		D) Robotic Arm	В
3	Bituin, AmielAndrew	х		Х	Х			B) Self Drving Golf Ca	D
4	Dai, Shuang	х	Х					В	С
5	Darlington, HestonDavid	х			Х			В	С
6	DeRosa, Stephen	х	Х	Х	Х			B) Self Drving Golf Ca	E
7	DeVeau, Adam		х	х	х			A) OmniDirectional R	C or E
8	Ferrera, Amber	х	Х		Х	х		A) OmniDirectional R	в
9	Garcia, Josephine				Х	Х	Х	A) OmniDirectional R	В
10	Jacobson, Eric	x	х	х	х	х	х		
11	Karuturi, JitendranathCh	х						В	С
12	Klesczewski, Peter		Х	х	х		х	B) Self Drving Golf Ca	A
13	Kobos, Alexander							C) Hovering Robot	В
14	Kodali, Madhukanth	х	Х					C) Hovering Robot	B)
15	Kral, Jacob								
16	Likhitha Mullapudi	х						B)	C)
17	Malempati, PoornaPruthvi	х						В	C)
18	Maynor, Kaelaan		х	х	х			B) Self Drving Golf Ca	D
19	Melecio, Javier	х						C) Hovering Robot	В
20	Merrikin, Ryan		Х	Х				A	E
21	Pagadala, SaiAditya	x						В	С
22	Severino, Jeffrey	х				Х		В	С
23	Simko, Justin	х		Х	Х		Х	A) OmniDirectional R	D
24	Tamboli, Parth								
25	Woodard, Matthew		Х		,Х	Х	Х	B) Self Drving Golf Ca	С
26	Dion, Scott	х	Х	Х	Х	Х	Х		
27	Day Moo	х		х	х	х		В	
28	Nigel Otis							B) Self Drving Golf Carl	
29	Christopher Jaramillo							C) Hovering Robot	A)
30	Toby Poole	х	х			х		D) Robotic Arm	С

- We welcome all experience levels.
- If you would like to join our meetings, please drop your name and email address.

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Today, we will talk about

A) Our research projects,

- B) Fundamental Robotics Concepts:
 - Feedback
 - Sensor Fusion
 - Perception
 - Platforms
 - Advanced Locomotion
 - Basic Localization
- C) Probabilistic Localization and MappingD) Future of Engineering EducationE) Q&A

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Our platforms: "We create!"



Underwater

Domain

Educational

Platform

Swarm Robotics

Study

Automated 3D

Scanner

3D Mapping

Exploration Missions

How can I start ? Let's start with couple concepts...

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Sensor Suite

Bionic Egg: Ruggedized Remote Sensor Suite for Impact and Ambient Conditions

Design Challenges:

- a) 2 2.7 inches in length
- b) 1.5 2 inches' wide
- c) 5 inch average circumference
- d) 114 grams of approximate weight





Students: Electrical Engineering: Simon Darius | Computer Engineering: Eric Jacobson, Mechanical Engineering: Theresa DeFreitas, Maegan Hall, Jerrod Sutcliffe Paper: "Bionic Egg: Sealed mobile sensor packaging design with adaptive power consumption, E. Jacobson, S. Darius, A. Tatoglu and P. Mellodge, IEEE Long Island Systems, Applications and Technology Conference (LISAT), Farmingdale, NY, 2017, pp. 1-6. 2017"

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Swarm Robotics

Vibron: A new approach to the coordination of multirobot systems which consist of many small physical robots. No moving parts!



New Design: 3D Printed Body



- They are designed to work collectively and in tune with each other.
- Primary focus is pointed at controlling the motion of the robots and possibly make them communicate with each other



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Environmental Decisions: Harsh Environments

Unmanned Underwater Vehicle: Create a simple autonomous robot that travels underwater following predetermined cube like path.



Bill-of-Materials

- Pelican 1020 Waterproof Micro Case, Arduino UNO, Motor Drive Shield, 12V Submersible Water Pumps
- Plastic Submersible Cord Grip, Adafruit Water Flow Sensors, Zip ties, Styrofoam

Students: Jason Carter, Jamie Dolan, Tiffany Pauley, Troy Solt, & Jeremy Stager

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Perception

3D Robotic Arm Scanner: A device that uses a robotic arm along with a hand held scanner to make digital 3D model of objects.



Students: Mason Paul ME, Gabriel Valero ME, Hector Ortiz CET, Justin Simko ME

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Locomotion: Alternate Mechanisms

OmniBot: SPARK: A ground vehicle with use of mecanum wheels that can move in all directions.

Design Iterations





Students: Nikhil Rametra, Yeshwanth Kumar Abburi

Zero Radius Rotation

Can nature help me ? Of course! Robotics and Biomimetic.

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Locomotion: Mimicking Nature

SpiderBot: Locomotion of the robot imitating spider

walk.



Students: Gabriel Valero • Fasi Mohammed • Saranjog S. Sukhija

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Robotics and Biomimetic

MicroSwimmer: It has long been known that swimming at the microscale requires techniques that are very different from those used by macroscale swimmers, such as fish and humans [1].

Can we use these techniques to develop a robot ?
 Locomotion of the robot imitating spider walk.



Propulsion Control Structure Design for Micro Underwater Robot, IEEE International Energy and Sustainability Conference'2015

Mapping Model Generation CMPISLAM Concluding Remarks

Robotics and Biomimetic

Microorganisms are able to swim at low Re using a variety of techniques[1], none of which look like those used by macroscale swimmers.

All of the swimming methods utilized by microorganisms are fairly inefficient, which is not a problem because microorganisms' source of energy (food) is so plentiful.

20 µm and have a diameter around 0.25 µm





Propulsion Control Structure Design for Micro Underwater Robot, IEEE International Energy and Sustainability Conference'2015

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Robotics and Biomimetic



Tatoglu A., Propulsion System for Micro Underwater Robot, IEEE International Energy & Sustainability Conference, 2015

Can we build a self-driving car? We are working on it!

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Full Scale Self-Driving Car Project

Stephen	Matthew	Evan Gerard,	David Dai, ME,	Peter
DeRosa, ME, Jr	Woodard, ECE,	MET, Sr	Sr	Klesczewski,
	Jr			ECE, Fr
Digno Iglesias	Jeff Severino,	Nigel Otis, ME,	Eric Jacobson,	Day Moo
ME, Jr	ME, Jr	So	ECE, Gr	ME, Sr

Fr. Freshman, So. Sophomore, Jr. Junior, Sr. Senior, Gr. Graduate

Current Team Members

Autonomous Mobile Robotics Research Group

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Full Scale Self-Driving Car Project

Phase 1: [Completed]

Brainstorming and designing the system that will be implemented with the cart. It also must take all safety measures into account. [Completed.]

Phase 2: [Mid-Spring Semester]

Is when the designed system will actually be implemented with the golf cart. At this point the golf cart will be made <u>remote controlled</u>. This this will allow for testing of the systems implemented in a safe and controlled manner.

Phase 3: [Summer and Fall Semesters] sees the remote controls being handed off to the <u>autonomous systems</u>. Trials will be run under different circumstances the golf cart will encounter, to ensure proper and safe operation.



Cars are good, they can't even swim ^(S) Well, we have a solution for this!

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Multi Terrain Vehicles

Hovercraft can travel over almost any non-porous surface:

- even or uneven terrain sandy and icy grounds
- Ideal for disaster relief situations





Landing Craft Air Cushion (LCAC) is delivering supplies to the citizens of Meulaboh Indonesia after the 2004 Indian Ocean tsunami.

A hovercraft docking to a ship.

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Alternate Locomotion: Hovering

- The hovercraft's ability to distribute its laden weight evenly across the surface below it makes it well suited to the role of amphibious landing craft.
- Hovercrafts can transport materials from ship to shore and can access <u>more than 70% of the world's coastline</u>, as opposed to conventional amphibious landing craft, which are only capable of landing along 17% of that coastline.

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Control: Advanced Dynamics

- An Hovercraft is controlled by commands below.



Angular Displacement

MOTION COMMANDS:

- STOP[1s]
- MOVE FORWARD [2 s]
- TURN LEFT [0.5 s]
- MOVE FORWARD [2 s]
- STOP[1s]



Utilizing Reaction Wheels to Incl

Control: Motion Planning

- Initial Simulations: Motion Planning and Execution
- Different capture radius values are tested.
- Rotation takes time and not accurate



Tatoglu A, Greenhalge S,

Windheuser K., ASME International Mechanical Engineering Congress and Exposition, Volume 4B: Dynamics, Vibration, and Control, 2016

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Alternate Mechanisms





Introduction Mapping

Model Generation CMPISLAM Concluding Remarks

Alternate Mechanisms





RESULTS:

- Momentum wheel substantially increased **rapid angular displacement** ability of hovering body.
- System is **less sensitive** to the terrain/ground shape.

Controller Improvements with a more advanced system model

- Blows the air underneath the craft
- Rubber cushion—skirt—traps the air and inflates



The velocities on x and y axes are given by

$$\dot{x} = u\cos\psi - v\,\sin\psi \tag{1}$$

$$\dot{y} = u\sin\psi + v\cos\psi \tag{2}$$

 ψ : is projection angle between frames. u (surge speed) and v (sway speed) represent the velocities on x and y directions. Ω_H : angular velocity of the overall body Ω_H is equal to first derivative of vehicle orientation ψ given by

$$\dot{\psi}=arOmega_H$$

Parameter Identification and Closed Loop Control of a Flywheel Mounted Hovering Robot", Tatoglu A., ASME International Mechanical Engineering Congress and Exposition, 2017

Controller Improvements with a more advanced system model



The controller input u_1 is the sum of forward thruster fan forces which is given by

$$u_1 = F_L + F_R = m\dot{u} - mv\Omega_H + d_v u$$

This follows the second equation on the sway direction:

 $m\dot{v} + mu\Omega_H + d_v u = 0$ d_v : the coefficient of viscous friction. Second controller input u_2 is given by

$$u_2 = \frac{r}{2}(F_L - F_R) + M_w r = J \dot{\Omega_H} + d_r \Omega_H$$

J: is the overall vehicle inertia, M_w : rotational torque released by the flywheel d_r : the coefficient of rotational friction.

Parameter Identification and Closed Loop Control of a Flywheel Mounted Hovering Robot", Tatoglu A., ASME International Mechanical Engineering Congress and Exposition, 2017

Controller Improvements with a more advanced system model

- Feedback Control system of the differential drive forward thrusters
- Flywheel break engages at the waypoint.



Parameter Identification and Closed Loop Control of a Flywheel Mounted Hovering Robot", Tatoglu A., ASME International Mechanical Engineering Congress and Exposition, 2017

Fig.6 Global frames, individual axis linear and angular positions with feedback control system.

 ϕ , rad

20

40

3

2

0

0

40

Controller Improvements with a more advanced system model



Fig.8 Object Tracking to Generate the Path followed



Fig.12 Rotation with flywheel and fans, feedback controller on after rotation

Parameter Identification and Closed Loop Control of a Flywheel Mounted Hovering Robot", Tatoglu A., ASME International Mechanical Engineering Congress and Exposition, 2017

It looks complicated. Is there an easier way to learn the control logic? Yes, of course! Pyro

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Educational Platform: Pyro

Electromechanical Design

	Design, Prototyping & Testing	
	Design & Testing Components	
	Microphone: detecting the start up frequency at	
	3.8KHz	
	Arduino Uno: Single-board microcontroller with	
	Analogue and Digital Inputs/Outputs	y.
	Heat/IR Sensor: detects the fire at a certain	
	distance and using an algorithm to be closer and	
	extinguish it	1
	Servo Motors: coding a rectangle shape to	1
- Resident	calculate the percentage error	1
	Ultrasonic Sensors: Having 180 degree rotation to	
	read obstacles and their distances	



Robot Controller



Trinity College, Firefighting Robot Competition 2017.

Students: Electrical Engineering: Heather Volkens, Mechanical Engineering : Yousef Bahman Ali Alsulaiman Bryant Miranda

Problem Statement

The tournament expects Pyro to avoid obstacle, solve the maze and extinguish a fire with fastest amount of time possible.

Solution

Using highly sensitive sensors like Ultrasonic sensors to avoid the obstacles and any walls present in Pyro's way, Left/Right hand rule so that Pyro follows the walls until it solves the maze and Heat/IR sensor to detect the fire and use a blowing fan to extinguish it.

How about UAVs? Yup!! It is time!

UAV Path Planning

Localization algorithms are also used to follow a predetermined path.



High Altitude: Mostly Linear Path Plan

(This problem is kind of solved.)

Continuously Varying Path Plan: Fixed distance from ground



Obstacle Avoidance

During the mission, path plan needs to be updated locally once an obstacle is met.



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Mission Types

Constant vs Variable input signal

- Linear Path Plan vs Continuously Varying Path Plan
- Obstacle avoidance, Rapid Moving Object tracking
- Little Disturbance vs <u>High Disturbance</u> (i.e. wind)



[1] Aggressive Maneuvers for UAV Flight, GRASP Lab, UPenn, Mellinger, IJRR 2012

[2] High-speed Flight in an Ergodic Forest, MIT, Karaman , ICRA 2012

Stereo Imaging: Mimicking Human Vision System

How can a UAV/robot perceive the environment ? Visual Navigation: It can perceive the environment including depth with a stereo camera system, same as human beings. $\times \times \times$



Visual Navigation

For a UAV and its visual navigation system: We want to develop a 2 DOF gimbal controller for continuously variable controller input for mission types discussed.

HOW CAN WE DECIDE GIMBAL CONTROLLER PARAMETERS IF WE ACCOUNT FOR:

Landmark Tracking Quality

Steady State Error

LQM(Landmark Quality Metric)

 e_{ss} (error, steady state)

Energy Consumption

Watt-second

Landmark Detection Algorithms

Sobel, Roberts, Canny, LoG, Prewitt, FAST(Features from Accelerated

Segment Test)



Wait!!! I am lost! What is a landmark ? OK, let's start again, from the beginning.

What is the common task for all the robots discussed ? If we ask them to go to the nearest Starbucks and get a coffee...

Why is Localization important?

• The first question required to be answered for all these robotic systems is "Where am I? "





Where am I

Current State of Research Efforts

- SLAM(Simultaneous Localization and Mapping) is a stochastic(probabilistic localization algorithm.
- It is defined as a chicken and egg problem:
 - Robot moves, generates a map.
 - Then try to localize itself within this map.
 - By using this new location, it decides where to go.
 - And again generates a map to localize itself in it.
- It corrects itself and reduces uncertainty.
- Currently, SLAM is the most advanced localization algorithm.

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Current State of Research Efforts

 SLAM(Simultaneous Localization and Mapping) methodology offers a probabilistic solution as an answer to the localization problem [Durrant-Whyte, 2006].



Localisation problem may be formulated as computing the probability distribution

 $P(x_k|z_{0\ldots k},\mathbf{u}_{0\ldots k},\mathbf{m})$

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Current State of Research Efforts

- There are various solution approaches
 - Stereo SLAM
 - RGBD SLAM [Kinect Like Sensors]
 - tinySLAM
 - SLAM with RBPF [Non-linear Solutions]
 - Visual Odometry [Camera + Odometry]
 - MonoSLAM [Single Camera Solutions]

What are the applications ? OK, let's see the applications and finalize with a simple example.

Mapping and Localization

GPS + MAP

When there is a GPS and a map, localizing a robot is easy.



Courtesy of Google and Bing

Introduction Mapping

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Accuracy of GPS

Junior, DARPA Challenge: 3D Point Clo



GPS, Known

And multi

sensors

map

Localization

Junior: The Stanford Entry in the Urban Challenge [Montemerlo, 04]

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Indoor robot: No GPS

Indoor Localization



Indoor Ground Robot JAMES: SIT No GPS, Known map And multí sensors

IMU+Camera Navigation



Indoor Quad-rotor University of Minnesota

GPS-denied navigation

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No or Obsolete Map on

• Mine and disaster area search missions

Mining Area Snake like Search Robot

Rescue Robot: Gemini-Scout

Tohoku University

No-GPS,

NO MAP,

Known

input

controller

Localization at unstructured environments

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Unexplored areas

• Rover's planned path and navigation camera image





Localization at extraterrestrial planets (Courtesy of NASA)

Sooo, is it possible to localize a robot without GPS? Let's discuss a case study.

|--|

 If we have a map and if we know velocity of a robot, can we find where it is ?



Illustration of a map with three <u>doors A, B and C</u> from left to right. Distances in between them is known(because we have a map).

Probabilistic Localization

Introduction Mapping Model Generation CMPISLAM Concluding Remarks	How	do	rot	ots	nav	/igate	?	
_			_	• • •		•	• • •	

Robot's current position is unknown. It is lost!*



Mapping Model Generatio CMPISLAM Concluding Remarks	Introduction
Model Generatio CMPISLAM Concluding Remarks	Mapping
CMPISLAM Concluding Remarks	Model Generation
Concluding Remarks	CMPISLAM
Remarks	Concluding
	Remarks

How do robots navigate?

- Robot's camera sees a door.
- Where can it be ?





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How do robots navigate?

- Robot keeps moving. Couple seconds later....
- It see another door.
- Now, where can it be?





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How do robots navigate?

- Robot keeps moving. Couple seconds later....
- It see another door.
- Now, where can it be?





Are images sufficient?



3D Printed BiPad Robot



Blurred Image

Humanoid

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Are images sufficient ?

• Images might not be sufficient for an accurate localization. Especially for self driving cars.





3D Point Clouds



Point Clouds and Intensity

[VIDEO-1, Fly Through]

Other Applications



Embedded Virtual CAD Models

[VIDEO-2, Elm Street]

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Robotics Research and Education

- 1) One of the most funded topic:
- National Robotics Initiative (NRI)
 The realization of co-robots acting in direct support of individuals and groups
- 2) Future of Engineering and Science
- A Roadmap for US Robotics- From Internet to Robotics
- New multidisciplinary departments
- 3) Self-Directed Learning will be the key of future education since most of the text books will be obsolete in couple years.
- 4) Gap between science and branches of engineering is closing.

Implementing Self Learning Skills with Multidisciplinary Robotics Courses, Tatoglu A., Russell I., ASEE Mid-Atlantic Section

INTRODUCTION INITIAL ANALYSIS SYSTEM DESIGN EXPERIMENTAL RESULTS CONCLUSION



Acknowledgements:

I would like to thank you ASME team, especially Mr. Ziair Deleon, for inviting me.

I also would like to thank ME Department and all students who were part of the projects.

INTRODUCTION INITIAL ANALYSIS SYSTEM DESIGN EXPERIMENTAL RESULTS CONCLUSION

Thanks!

If you would like to learn more about autonomous mobile robots, please join our email list.

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"We create!"

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