### **Probabilistic Methodologies for Autonomous Mobile Robot Localization**

### **Speaker**

**Dr. Akin Tatoglu | tatoglu@hartford.edu Assistant Professor, Mechanical Engineering** 



**Autonomous Mobile Robotics Research Group | D-121**

#### **UNIVERSITY OF HARTFORD**

**COLLEGE OF ENGINEERING.** TECHNOLOGY, AND ARCHITECTURE

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### 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,



**Autonomous Mobile Robotics** 

• 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.







- **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 Mapping D) Future of Engineering Education E) Q&A

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### Our platforms: "We create!"



Advanced Sensor S Differential Drive 3D Mapping **Exploration Missio** 



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 2.7inches 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!



They are designed to work collectively and in tune with each other.

New Design: 3D Printed **Body** 

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**



**Working Principle** 21 18 12 straight ahead 18 12 concerning urn of rear axis

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







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

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## 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



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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 remote controlled. 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 autonomous systems. 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  $\odot$ 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 more than 70% of the world's coastline, 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]



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# Control: Motion Planning

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



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

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### Alternate Mechanisms





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### 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.  $\Omega_H$ : angular velocity of the overall body  $\Omega_H$  is equal to first derivative of vehicle orientation  $ψ$  given by

$$
\dot{\psi}=\varOmega_{H}
$$

### 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

$$
u_1 = F_L + F_R = m\dot{u} - m\nu\Omega_H + d_\nu u
$$

This follows the second equation on the sway direction:

 $m\dot{v} + m u\Omega_H + d_v u = 0$  $d_v$ : 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.

### Controller Improvements with a more advanced system model

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



### 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

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**





#### **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 High Disturbance (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.



# 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

 $W$ att – 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** 

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

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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.

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# Mapping and Localization

*GPS + MAP*

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



Courtesy of Google and Bing

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## Accuracy of GPS

• Junior, DARPA Challenge: 3D Point Clo *GPS, Known map And multi sensors*

### **Velodyne laser Applanix INS SICK LMS laser Riegl laser** WO **BOSCH Radar DMI IBEO** laser **SICK LDLRS laser Junior Sensor Suite Localization Laser typical accuracy: +/- 2cm GPS Error: ~2-20 meters**

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

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### Indoor robot: No GPS

• Indoor Localization



**Indoor Ground Robot**

*No GPS, Known map And multi sensors*

**IMU+Camera Navigation**



**JAMES: SIT Indoor Quad-rotor University of Minnesota**

**GPS-denied navigation**



### No or Obsolete Map

### • Mine and disaster area search missions ....

**Snake like Search Robot Mining Area**

**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 doors A, B and C from left to right. Distances in between them is known(because we have a map).**

Probabilistic Localization



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





# How do robots navigate?

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







# How do robots navigate?

- Robot keeps moving. Couple seconds later….
- It see another door.
- Now, where can it be? **Where am 1?**







# How do robots navigate?

- Robot keeps moving. Couple seconds later….
- It see another door.
- Now, where can it be? **Where am 1?**





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### Are images sufficient?



**3D Printed BiPad Robot**



**Blurred Image**

### Humanoid



## Are images sufficient ?

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

3D Point Cloud Attributes:





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### 3D Point Clouds



Point Clouds and Intensity

[VIDEO-1, Fly Through]

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### **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 Conference, Hofstra University, 2016





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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.

### **Dr. Akin Tatoglu tatoglu@hartford.edu**



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### **"We create!"**

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