#### Polarconference 2016 – DTU 1-2 Nov 2016

## DTU

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## **Unmanned Surface Vessels**

- Opportunities and Technology



#### **Mogens Blanke**

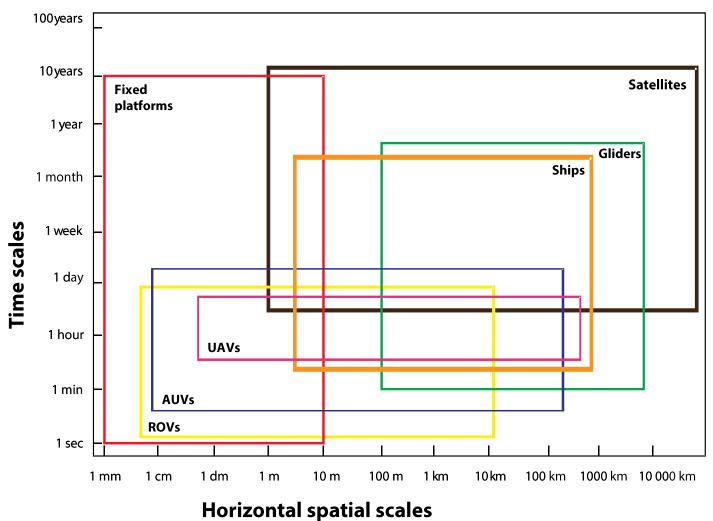
DTU Professor of Automation and Control, DTU-Elektro Adjunct Professor at AMOS Center of Excellence, NTNU, Norway E-mail: <u>mb@elektro.dtu.dk</u>

 $f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}$ 

DTU Elektro Institut for Elektroteknologi



## **Time – space coverage of technologies**



## Mapping and monitoring of marine resources and environment for governance and decision making. Territory surveillance, security.





## **Unmanned surface vehicles USV**

#### USVs:

- Own missions
  - Surveillence
  - Intervention
  - Rescue
- Integrated missions air
- Support for underwater
- Mission specific design: speed, range, instruments

Lower cost than manned

#### 24/7 and long endurance

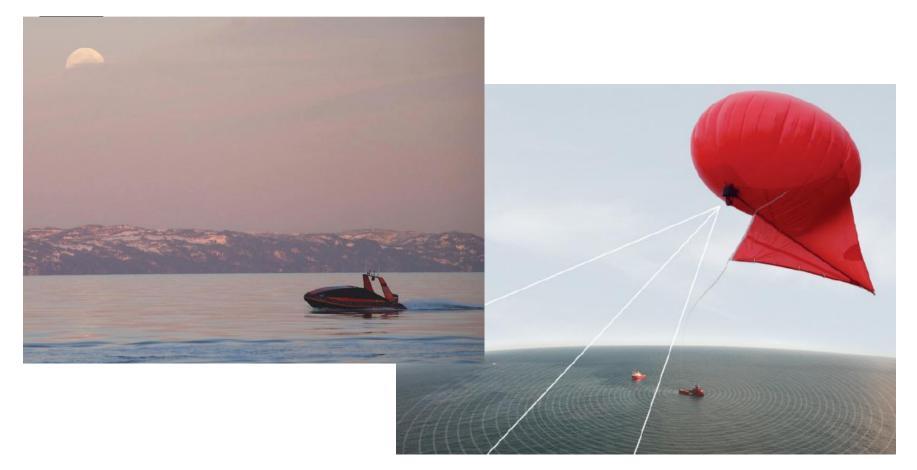
#### Multi-vehicle operation

Excelent for tedious tasks

Smaller environment footprint



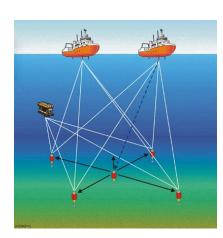
## Autonomy in operation: Maritime Robotics (Trondheim)



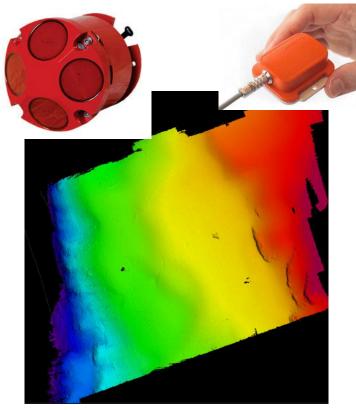
Images used by courtsey of Maritime Robotics (patented technologies)

## **Navigation sensors -**

- Position:
  - GPS at surface for position fix
  - Acoustics
  - Optics (images, video, laser)
- Depth (pressure)
- Altitude and relative velocity to water or seafloor (Doppler Velocity Log)
- Orientations and accelerations, (Inertial Measurement Units)
- Radar (various bands to distinguish different objects)
- Vision systems visible, infrared, multi-spectral, stereo-vision.







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## **Increasing the Level of Autonomy**



Level		Descriptor	Guidance	Navigation	Control	EEM
	10	Fully autonomous	Human level decision making	Human like navigation capabilities	Same or better performance as for a piloted vehicle in the same conditions	Very High
	4	Real-Time Obstacle/Event Detection and Path Planning	Hazard avoidance, Real- time path planning and re-planning	Perception capabilities for obstacles, targets and environment, low fidelity situation awareness	Robust trajectory tracking capabilities	Mid-low
	3	Fault/Event Adaptive USV	Low-level decisions and execution of pre- programmed tasks	Detection of hardware and software faults	Robust adaptive controller	Low
	2	ESI Navigation (e.g. non-GNSS)	Waypoint guidance of pre-planned paths	Sensing and state estimation by the USV, all perception and situational awareness by the operator	Control commands are computed by the autopilot	Low
	1	Automatic Control	Waypoint guidance of pre-planned paths	Sensing and state estimation by the USV, all perception and situational awareness by the operator	Control commands are computed by the autopilot	Low
	0	Remote Control	Performed by external system (mainly human operator)	Sensing done on-board the vehicle, data are processed externally (human operator)	Control commands are given by a remote external system	Very Low

#### Kendoul, F., **Survey of Advances in Guidance, Navigation, and Control of Unmanned Rotorcraft Systems**, J. Field Robotics, 29, 2012

7 DTU Electrical Engineering Technical University of Denmark Smart Sensor Based Obstacle Detection for High-Speed Unmanned Surface Vehicle

IFAC MCMC2015, Copenhagen August 24-26

### **Obstacle Detection for High-Speed Unmanned Surface Vehicle**

- High-speed unmanned surface vehicle
- Desired <u>Autonomy Level 4</u>
  - Robust adaptive controller
  - Perception capabilities for obstacles/environment
  - Hazard avoidance/path re-planning



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#### **DETECTION REQUIREMENTS**

#### **Class of obstacles**

 Boats, yachts, and buoys with radar reflectors

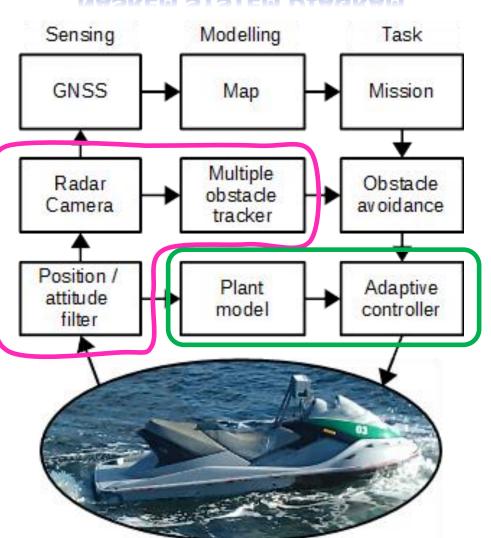
#### Range of detection (30m/s)

- Safety: 60m
- Evasive: 30m

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### **Dan Herman: Overall System Architecture**





NASREM SYSTEM DIAGRAM

#### Previous work

- Plant model
- Way-point controller
- Station keeping controller

#### Contributions

- Vision assisted position and attitude filter
- Multiple obstacle tracker

Herman, Galeazzi, Andersen and Blanke: Smart Sensor Based Obstacle Detection for High-Speed Unmanned Surface Vehicle. IFAC-Papers Online vol 48 (16), pp190-197, DOI: 10.1016/j.ifacol.2015.10.279

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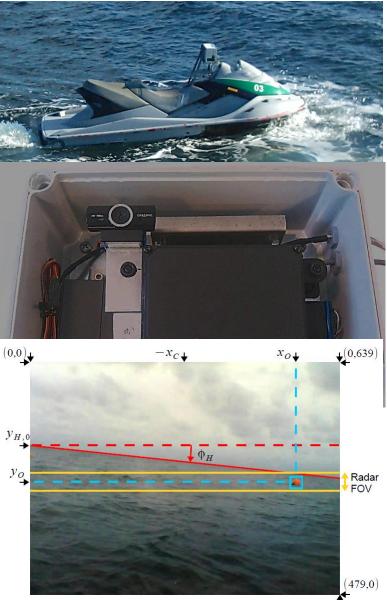
## Vehicle & Sensors Suite

#### . Unmanned marine craft

- Modified for autopilot and remote control
- . Sensors for obstacle detection
  - Automotive scanning radar
    - Mid range: 50m +/- 45°
    - Long range: 175m +/- 15°
    - Vertical FOV: 5°
    - Fs = 20Hz
  - Onboard low-cost camera
    - Resolution 640 x 480 pixels
    - FOV: 52° x 39°
    - Fs = 10 fps

#### Navigation sensors

- GPS (Fs = 3-4 Hz)
- 6DOF IMU (Fs = 100Hz)
- 3DOF Magnetometer (Fs = 100Hz) set



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## **Image Object Detection: Challenges**

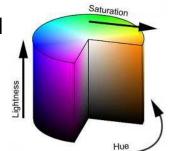
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#### Illumination

- Overcast
  - Constant illumination
  - Detectable: RGB and saturation
- Sun reflections
  - Numerous false detections
  - Detectable: RGB
- Good illumination
  - Detectable: Saturation and hue

#### **Post-processing detection**

- Adaptive threshold





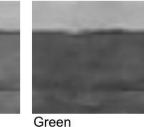
Input

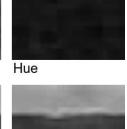
Red





Saturation





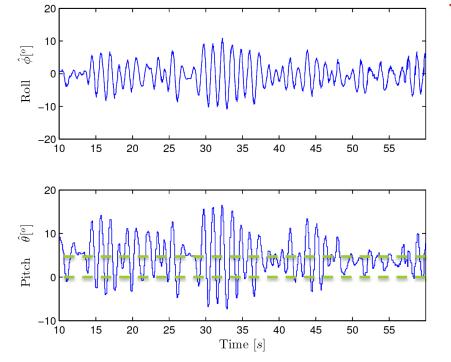


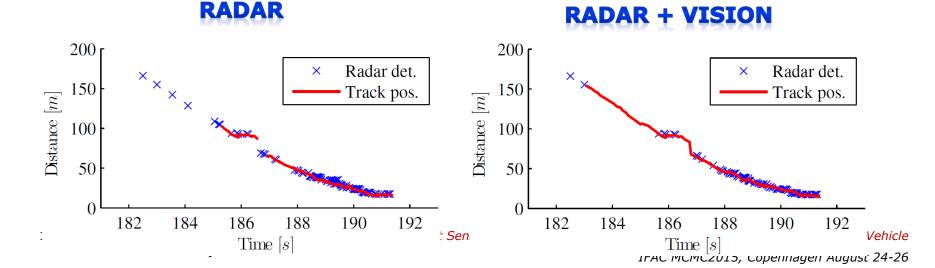
11 **DTU Electrical Engineering** Technical University of Denmark Smart Sensor Based Obstacle Detection for High-Speed Unmanned Surface Vehicle

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## Sea Trial – Track Persistence Assessment

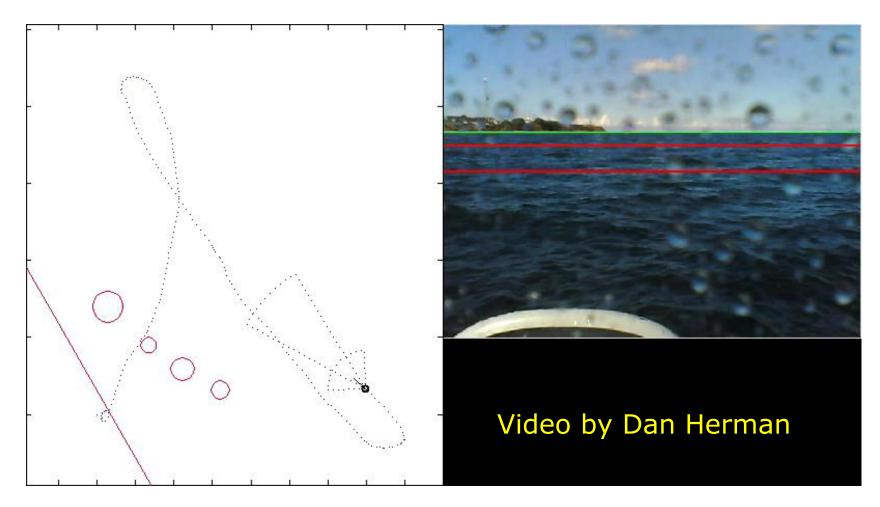
- Pitch and roll motion
  - Exceed radar vertical FOV
- Radar maintained
- . Radar assisted by vision
  - Range of detection increased (doubled)
  - Increased track persistence



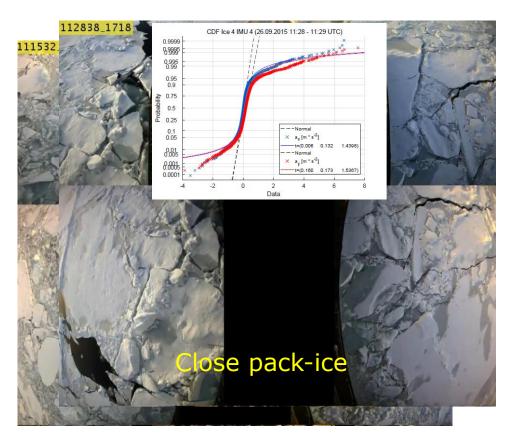


## **Results: Obstacle detection for safe navigation**





## **Vehicles in ice ??**





Hans-Martin Heyn (AMOS) at the North Pole

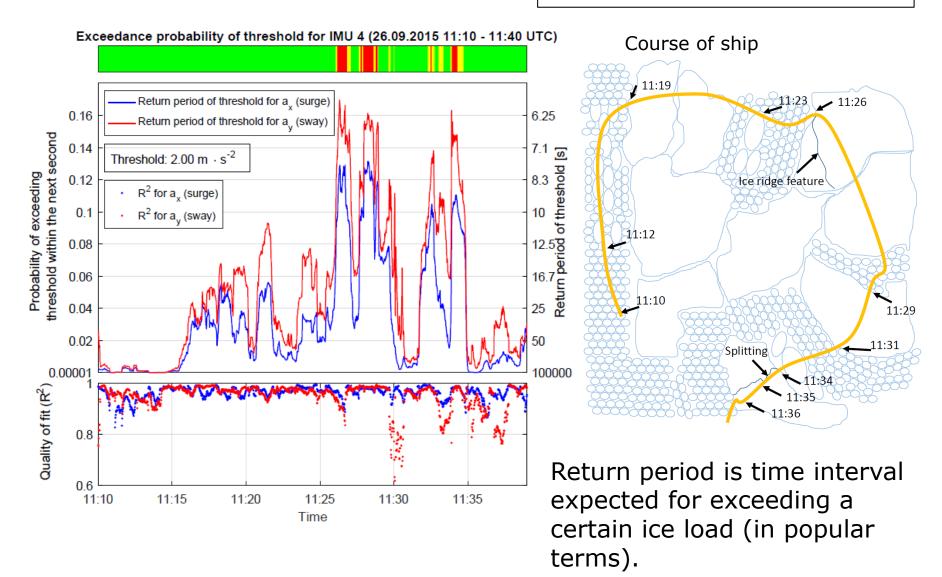


Accelerometers measure ship accelerations. 4 cameras monitor ice Distribution of accelerations disclose type/severity of ice load. Cameras are used for validation

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# Short term Ice-load prediction

Heyn, Blanke, Skjetne: Estimation of extreme ice accelerations based on signal detection. (NTNU – AMOS results)



# Unmanned and Autonomous Vessels:



"I was navigating by sight because I knew the depths well and I had done this maneuvre three or four times" . Captain Schettino Master, Costa Concordia. Source: BBC.com

## **How Can Autonomy Enhance Safety**

Simplify information

Perception -> warn if danger

Suggest solutions

Make autopilot "intelligent" – aware of context

Intervene in critical situations

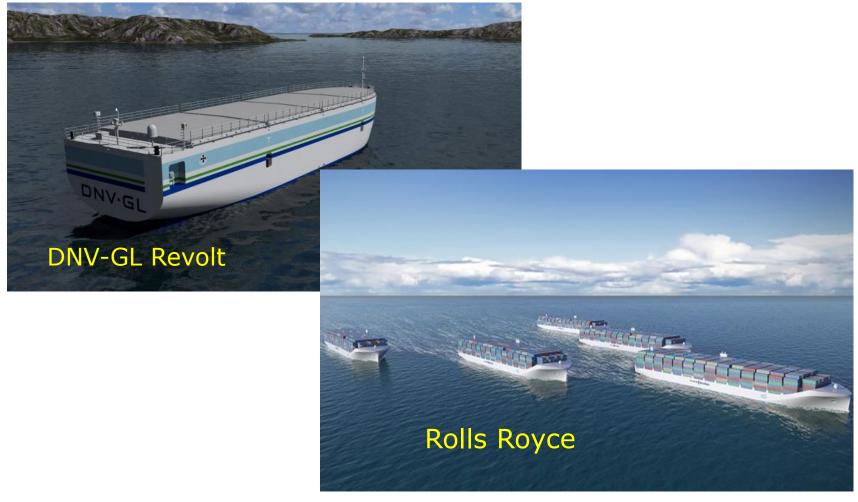
Øystein Engelhardsen: Autonomy at Sea. Plenary at IFAC MCMC'2015 Conference at DTU.

Navigate autonomously

Supervise remotely



## **Development and research in the comercial area**





## Technology is available in the very near future.

## How do we wish to take advantage of its benefits in the arctic areas

