#### Centralized State Estimation of Distributed Maritime Autonomous Surface Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan {ralch,nickoe,fjuul,afodor12,tmures12}@es.aau.dk

> Department of Electronic Systems, Aalborg University, Denmark

> > December 19, 2012

### Agenda



Introduction Purpose

Path planning
Waypoint generation
Verification

Modeling and Control
Theory
Verification

State estimation
Theory
Verification

Final test

Kalman estimator

Controller

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Introduct

Purpose

ath planning

/aypoint generatior erification

Modeling and Control

erification

State estimat

Verification

Final test

Kalman estim

21

#### Introduction

Purpose

 Little to no research are currently devoted to maritime autonomous crafts.



Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

ntroduction

#### Purpose

Purpose

Path planning
Waypoint generation
Verification

Modeling and Control

Theory Verification

State estimation

Theory Verification

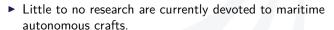
Kalman estimat

Controller

21

#### Introduction

Purpose



▶ During the 2012 Fukushima accident in Japan, no measurements of the spread of radioactivity was available in the coastal zones, thus relying only on estimates.



Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Introduction

Purpose

Path planning

Waypoint generation

Modeling and Control

heory

/erification

State estimation

Verification

Final test

Kalman estimato Controller

#### Introduction

Purpose



- ► Little to no research are currently devoted to maritime autonomous crafts.
- During the 2012 Fukushima accident in Japan, no measurements of the spread of radioactivity was available in the coastal zones, thus relying only on estimates.
- ► The coastal area around Greenland has no up-to-date baymethric maps available, and with the growing interest in Greenland (both industrially and commercially) this poses a threat to the ships going in and out of the fjords.

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Introducti

#### Purpose

Path planning

Waypoint generation Verification

Modeling and Control

Theory

Control of the

Theory

Verification

Kalman estimator

# Path Planning Considerations and Description



#### Considerations

- ▶ The path planner must be memory efficient.
- The path planner should consider the coastal line to be covered.
- ▶ The path planner should create an efficient path.

#### Straight and Turning segments

- ▶ Straight segments are generated by having a fixed minimum distance from the shore line, and then define a distance between the measuring lines. This cannot be optimized, as straight lines are, well straight lines.
- ► Turning segments are however more interesting, as these can be optimized.

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Introduction Purpose

4 Path planning

Waypoint generation

Waypoint generation Verification

Modeling and Control

erification

State estimation

Verification

Kalman estim

ntroller



#### Theory

► The path planner is based around the Train transition problem, originally solved by. INSERT REF!

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Purpose

Path planning

Waypoint generation

Verification

Modeling and Control

/erification

State estimation

Theory Verification

Kalman estin

21

Controller



#### Theory

- ► The path planner is based around the Train transition problem, originally solved by. INSERT REF!
- ► The problem describes that a vehicle in motion, can maintain a linear angular acceleration if the amount of jerk *j* experienced by the ship is kept constant.

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

ntroduction

5 Path planning
Waypoint generation

Verification

Modeling and Control

reory /erification

State estimation

Verification

Final test

21

Kalman estimator



#### Theory

- ► The path planner is based around the Train transition problem, originally solved by. INSERT REF!
- ► The problem describes that a vehicle in motion, can maintain a linear angular acceleration if the amount of jerk *j* experienced by the ship is kept constant.
- ► To keep the jerk *j* constant, the theory defines the path using the two normalized Fresnel integrals, that when plotted produces the Euler spiral. The Fresnel integrals are given as:

$$C_F(x) = \int_0^x \cos(t^2) dt, \quad S_F(x) = \int_0^x \sin(t^2) dt$$
 (1)

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Purpose

Path planning

Waypoint generation

Waypoint generation Verification

Modeling and Control

verification

State estimation

Verification

Kalman estimator

# Path Planning Threshold definition



- ▶ The path planner is also based on the curvature of the turn the ship is to make  $\kappa$ . This defines if the path should consist of a transition onto a circle, or two Euler spirals (dependent on the curvature).
- ▶ The threshold  $\varepsilon_{max}$  is defined to be:

$$\varepsilon_{\mathsf{max}} = \frac{\kappa_{\mathsf{max}}^2}{2 \cdot \eta} \tag{2}$$

lacktriangle Where  $\eta$  is a function described by the highest amount acceleration the ship can experience and the velocity at which it traverses this.

$$\gamma = \frac{\alpha_{\text{max}}}{v_{\text{max}}^2} \tag{3}$$

► This function allows the ship to preserve as much energy as possible, without veering out of course.

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Purpose

6 Path planning

Waypoint generation

Verification

Modeling and Control

Verification

State estimation

Verification

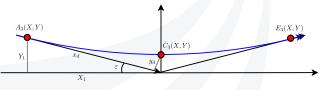
Kalman estim

Controller

3-point scenario



▶ The below figure shows the waypoints generated by the ship when  $\varepsilon < \varepsilon_{\text{max}}$ . This consists of an inward and an outward Euler spiral.



► The waypoints are computed as follows:

$$A_3 = (-X_1, Y_1), C_3 = (0, \frac{y_d}{\cos(\varepsilon)}), E_3 = (X_1, Y_1)$$
 (4)

▶ Where  $X_1$  and  $Y_1$  can be computed as functions of the normalized Euler spirals, with  $x_d$  and  $y_d$  representing the length:

$$X_1 = x_d \cdot \cos(\varepsilon) + y_d \cdot \sin(\varepsilon), \ Y_1 = X_1 \cdot \tan(\varepsilon)$$
 (5)



Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Purpose

Path planning Waypoint generation

Verification

Modeling and Control Theory

State estimation

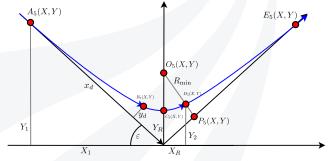
Verification

Kalman estimator

5-point scenario

UNITED STATE OF STATE

▶ The below figure shows the waypoints generated by the ship when  $\varepsilon > \varepsilon_{\text{max}}$ . This consists of a transition from an inward Euler spiral, onto a circle and then a transition from the circle onto an outward Euler spiral.



▶ Adding the two points  $B_5$  and  $D_5$  given as:

$$B_5 = (-R_{\min} \cdot \sin(\varepsilon - \varepsilon_{\max}), Y_1 - x_d \cdot \sin(\varepsilon))$$
 (6)

$$D_5 = (R_{\min} \cdot \sin(\varepsilon - \varepsilon_{\max}), Y_1 - x_d \cdot \sin(\varepsilon))$$
 (7)

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Purpose

Waypoint generation

Verification

Modeling and Control
Theory

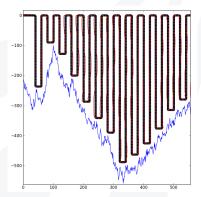
State estimatio Theory

Verification

Calman estimator Controller

UNIVERSITAS ALBURED

► To verify that the planner works properly, a random shoreline have been generated using a random walk, and a bounding box was drawn, the algorithm was then programmed to generate a shoreline, producing the following results:



▶ The dots on the figure represent the waypoints.

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Introduct

Purpose

Path planning

Verification

Andaling and Con

Modeling and Control

Verification

Theory Verification

Final test

Kalman estin

Consideration and Description



#### Considerations

The model is built using a top-down approach, running the main controls on a high-level interface (HLI). A low-level interface (LLI) handles the actuators and reads all the sensors. The link between these are a simplex 19.2 kbps radio link.

#### Top layer - HLI

Receives the sensor readings from the LLI and computes actuator set-points which are transmitted to the LLI.

#### Bottom layer - LLI

Receives the actuator set-points and sets these, then reads the sensors and transmit these readings back.

Centralized State Estimation of Distributed Maritime Autonomous Surface Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Verification

Modeling and Control

Verification

State space representation



#### Assumptions

The model is simplified to a 3-DOF model, with only motion in the x and y direction, as well as rotation about the z-axis defined as  $\theta$ . The model does not take into account the surge generated by the ship as it moves in the water, and the effects of planing is not taken into account, as the ship is not moving faster than  $\approx 1$  m/s. With these assumptions the following continuous time state space model have been derived:

$$\begin{bmatrix} \dot{\mathbf{v}} \\ \theta \\ \omega \end{bmatrix} = \begin{bmatrix} -\beta_{\mathbf{v}} & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & -\beta_{\omega} \end{bmatrix} \begin{bmatrix} \mathbf{v} \\ \theta \\ \omega \end{bmatrix} + \begin{bmatrix} m^{-1} & 0 \\ 0 & 0 \\ 0 & I^{-1} \end{bmatrix} \begin{bmatrix} F \\ \tau \end{bmatrix}$$
(8)

4 non-linear terms appear in the equation. Namely the  $\beta$  terms and the computations of the forward force F and the torque  $\tau$ .

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

introduct

rurpose

Path planning Waypoint generatio

Verification

Modeling and Control

1) Theory

State estimation

Verification

Final test

Kalman estimator Controller



#### F and $\tau$ terms

The force F and torque  $\tau$  are dependent on the same variables, namely the number of revolutions of the propellers  $n_1$  and  $n_2$ . These can be computed by:

$$\begin{bmatrix} n_1^2 \\ n_2^2 \end{bmatrix} = \begin{bmatrix} C_1 & C_1 \\ C_1 \cdot I \cdot \sin(\theta_{\text{stbd.}}) & C_1 \cdot I \cdot \sin(\theta_{\text{port}}) \end{bmatrix}^{-1} \begin{bmatrix} F_{\text{desired}} \\ \tau_{\text{desired}} \end{bmatrix}$$
(9)

Solving for  $n_1$  and  $n_2$  produces:

$$n_1 = \frac{n_1^2}{\mathsf{abs}\{n_1^2\}} \cdot \sqrt{n_1^2}, \ n_2 = \frac{n_2^2}{\mathsf{abs}\{n_2^2\}} \cdot \sqrt{n_2^2}$$
 (10)

### $\beta$ term

The  $\beta$  terms are linearized using a Taylor approximation. The constant term is removed in the reference gain of the controller.

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Introduction

Path planning
Waypoint generation
Verification

Modeling and Control

Theory

Verification

Verification

Final test

Kalman estimator Controller



Optimal feedback gain and reference tracking The optimal state feedback gain is computed by minimizing the cost function  $\mathcal{J}$ , given as:

$$\mathcal{J} = \int_0^\infty (x^T(t) \cdot Q \cdot x(t) + u^T(t) \cdot R \cdot u(t)) dt \qquad (11)$$

Producing the following feedback gain:

$$F_{\text{opt}} = \begin{bmatrix} 15.1668 & 0 & 0\\ 0 & 2.5165 & 0.7134 \end{bmatrix} \tag{12}$$

The reference gain is computed by augmenting the system as in INSERT REF!, thus producing the reference gain:

$$N_{\text{reference}} = \begin{bmatrix} 24.0668 & 0\\ 0 & 2.5165 \end{bmatrix} \tag{13}$$

Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Centralized State

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Durnoro

Purpose

Waypoint generation Verification

Modeling and Control

Theory

Verification

State estimation

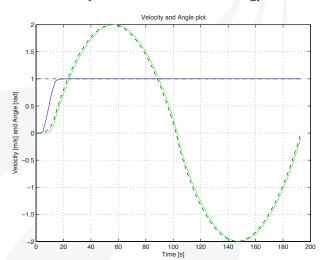
Verification Final test

> Kalman estimator Controller

#### Modeling and Control Verification



Through simulations, the system have produced the following figures used to verify wether the controller strategy.



Centralized State Estimation of Distributed Maritime Autonomous Surface Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Path planning

Verification

Modeling and Control

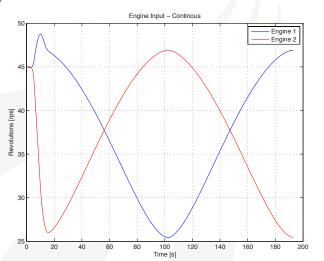
Verification

21

Verification

NI WAS ALBURA

And testing if the conversion from desired force to revolutions work.



Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Introduct

Purpose

Path planning

Waypoint generation Verification

Modeling and Control

5 Verification

#### Verification

State estimation

Verification

Final test Kalman estin

21

ntroller



Estimated states and input state model

The estimated states are defined as:

$${}^{b}\hat{x_{k}} = \begin{bmatrix} x & \dot{x} & y & \dot{y} & \theta & \omega \end{bmatrix}^{T}$$
 (14)

As the ship is fitted with an IMU and a GPS, the measured states becomes:

$$v_k = \begin{bmatrix} x & \dot{x} & \ddot{x} & y & \dot{y} & \ddot{y} & \theta & \omega & \alpha \end{bmatrix}^T$$
 (15)

Giving the state model of the system as (ts = 20 Hz):

$$\Phi = diag\{\Phi_{\mathsf{x}}, \Phi_{\mathsf{y}}, \Phi_{\omega}\} \tag{16}$$

Where the individual  $\Phi$ s are given as:

$$\Phi_{x,y,\omega}(k) = \begin{bmatrix} 1 & t_s & 0\\ 0 & 1 & t_s\\ 0 & -\beta_{x,y,\omega} & 0 \end{bmatrix}$$

$$\tag{17}$$

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

ntroduction

Path planning

Waypoint generation Verification

Modeling and Control Theory

State estimation

Theory Verification

inal test

Kalman estimator

Theory - zero gaining



As the LLI and HLI are run on two seperate computers, the Kalman filter needs to filter out the measurements which are invalid. This is done by multiplying the Kalman gain  $\bar{K}$  by a validity mask-matrix  $\Lambda$ , given as:

$$\Lambda = diag\{\lambda_{x}, \lambda_{\dot{x}}, \lambda_{\ddot{x}}, \lambda_{y}, \lambda_{\dot{y}}, \lambda_{\ddot{y}}, \lambda_{\theta}, \lambda_{\omega}, \lambda_{\alpha}\}$$
(18)

Where the  $\lambda$ s are defined as:

$$\lambda = \begin{cases} 1 & \text{if checksum is valid} \\ 0 & \text{otherwise} \end{cases}$$
 (19)

Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Introduction Purpose

Path planning
Waypoint generation
Verification

Modeling and Control

Theory Verification

State estimation

Theory Verification

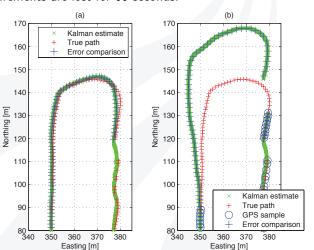
Final test

Kalman estimator

Verification of State estimator



A test of the Kalman filter, to see if it estimates lost packages was carried out producing. Figure (b) represents the case where GPS measurements are lost for 60 seconds:



Centralized State Estimation of Distributed Maritime Autonomous Surface Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Verification

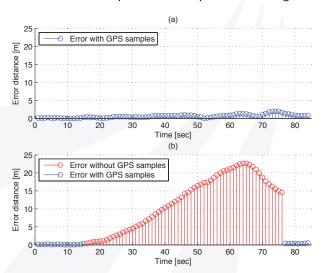
Modeling and Control

Verification

Verification of State estimator



The absolute error of the position are depicted on the figure below:



Centralized State Estimation of Distributed Maritime Autonomous Surface Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Path planning

Verification

Modeling and Control

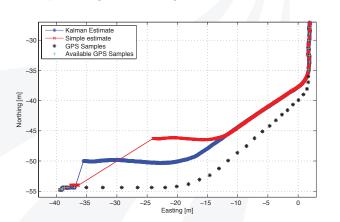
Verification

21

#### Final test

Kalman estimator verification

As the waters around Aalborg have frozen solid, final tests in water have not been carried out, however, tests on land have been carried out, producing the following:



Centralized State Estimation of Distributed Maritime Autonomous Surface Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Path planning

Verification

Modeling and Control

Verification



21

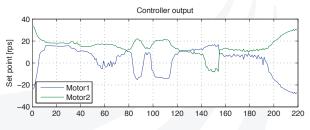
Kalman estimator

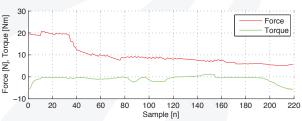
#### Final test

Engine input verification



#### Producing the engine inputs from the controller.





Centralized State
Estimation
of Distributed
Maritime Autonomous
Surface
Oceanographers

Rasmus L. Christensen, Frederik Juul, Nick Østergaard, Attila Fodor, Tudor Muresan

Introduct

Purpose

Path planning

Waypoint generation Verification

Modeling and Control

erification

State estimation

Verification

Final test

Kalman estimator

21 Controller

21