

# Experimental Analysis of Advanced Control and Estimation Systems for Autonomous Ship Landing

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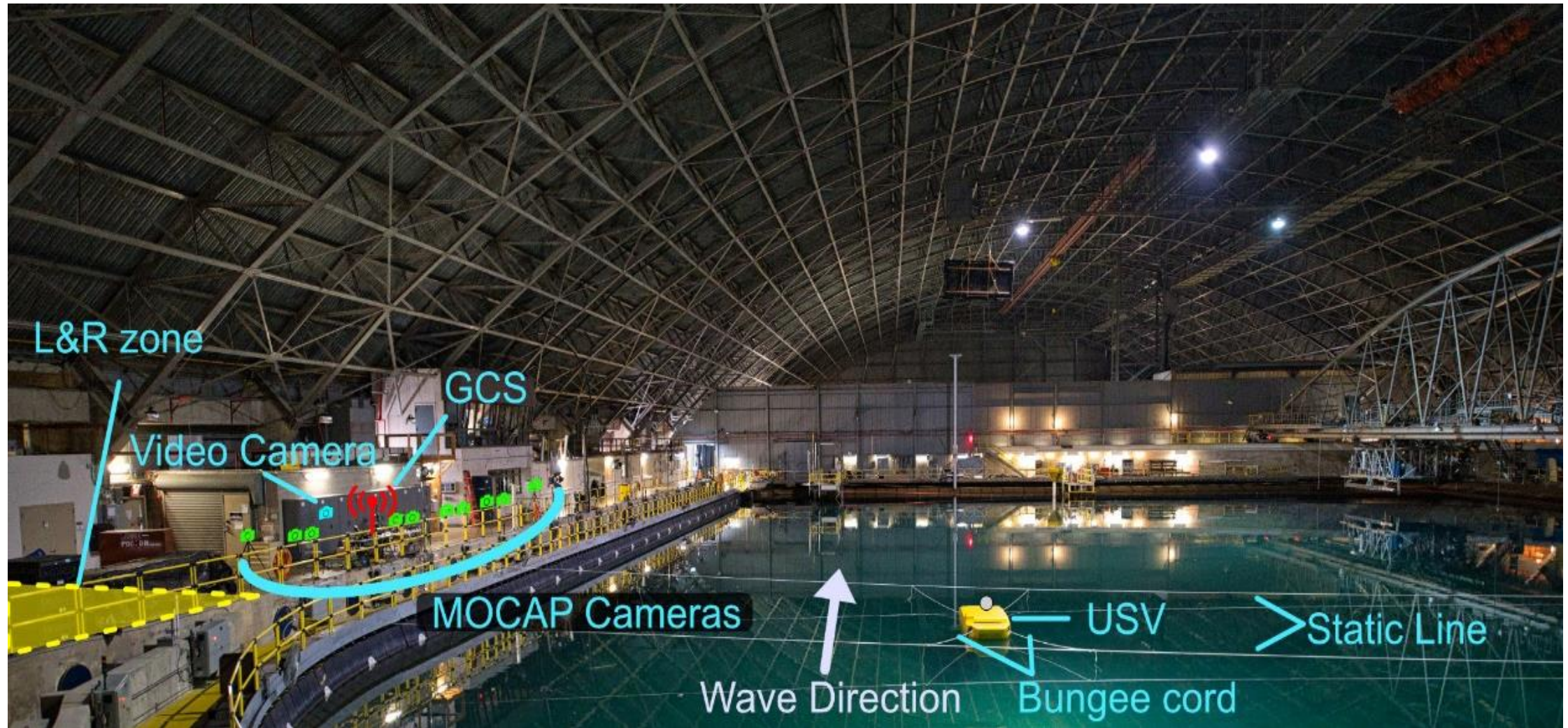
Distribution Statement A: Publication Unlimited

# Introduction

- Landing rotorcraft during high sea states is difficult
- Lack of public domain experimental data on ship landing systems
  - Full scale testing is difficult, risky, and expensive
- Testing at model scale is desirable
  - Low risk and cost
  - Controllable environment
  - High volume testing
- Research goals:
  - Sensitivity of landing algorithms to aircraft response characteristics
  - Comparisons between different path planners
  - Vision based deck state estimation

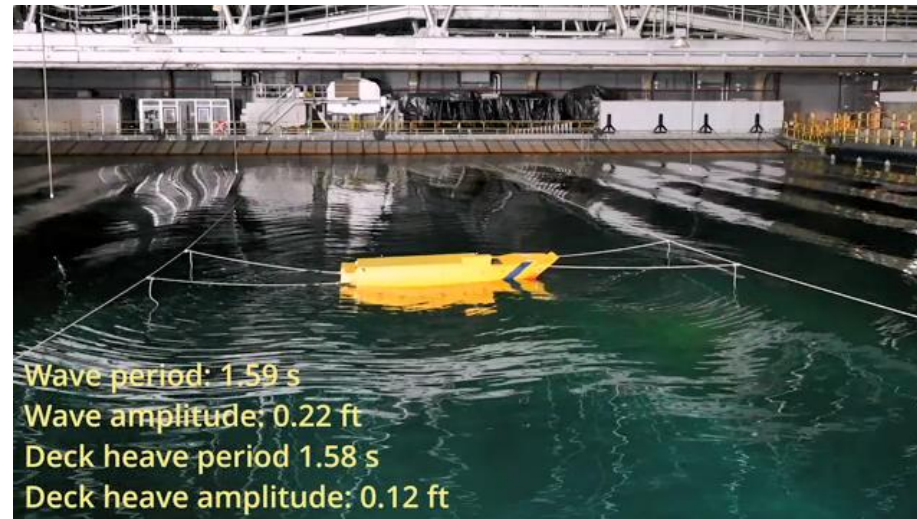
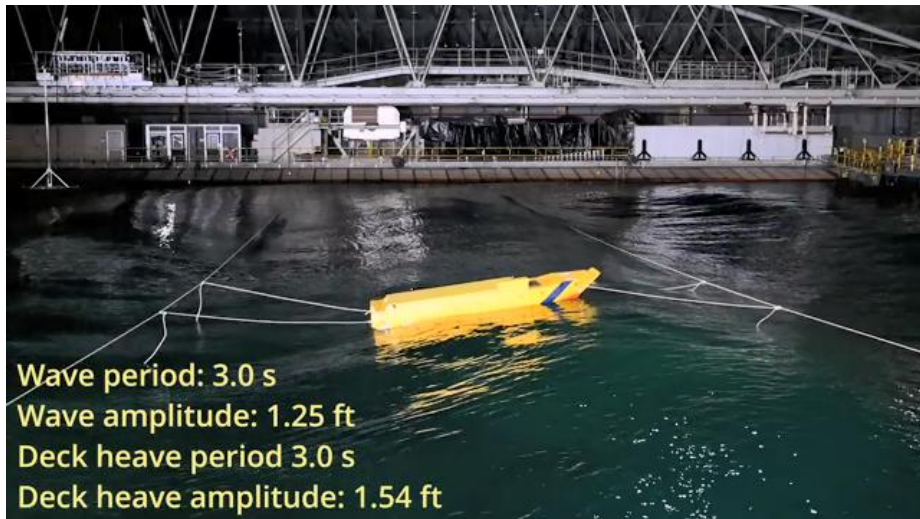
# Experimental Setup

# Maneuvering and Sea Keeping Basin



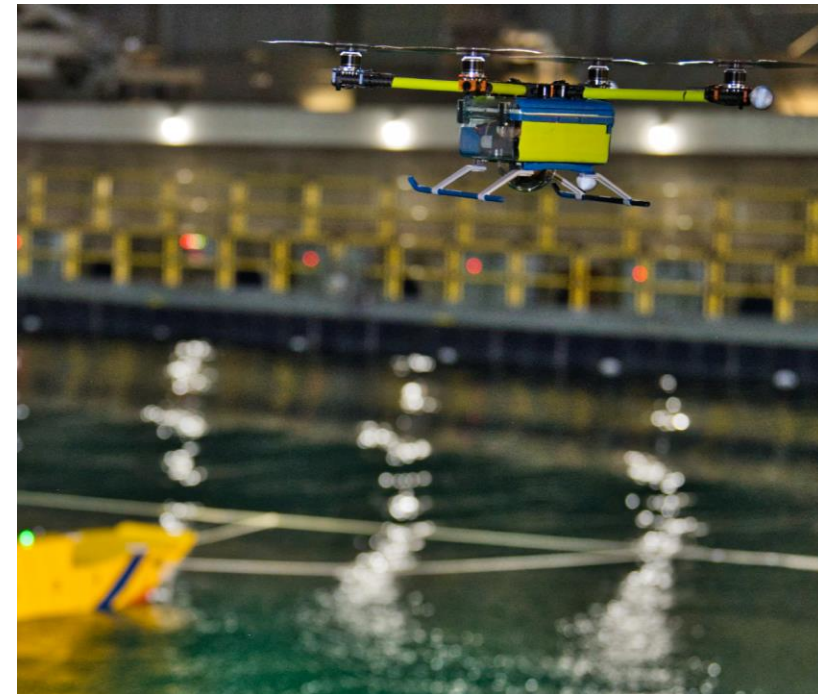
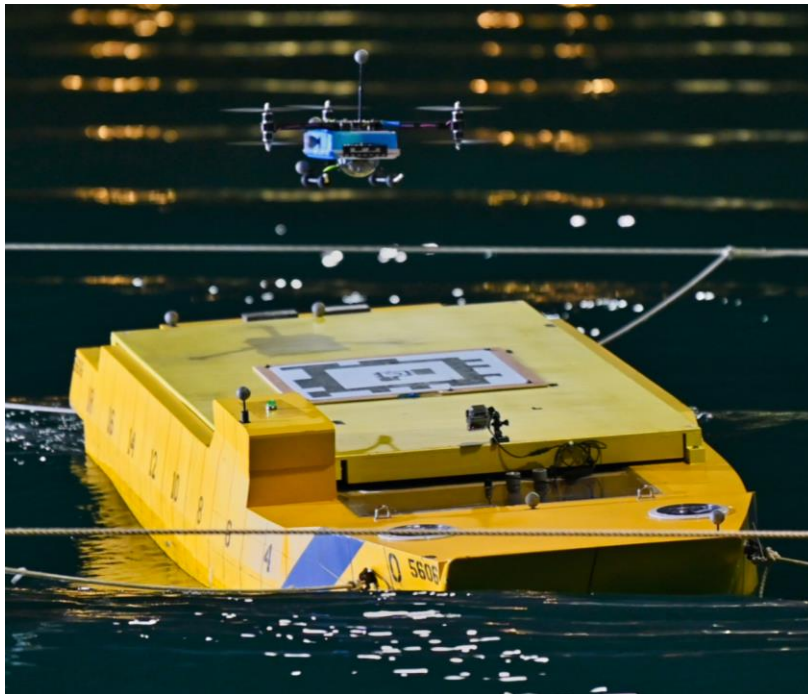


# USV Platform



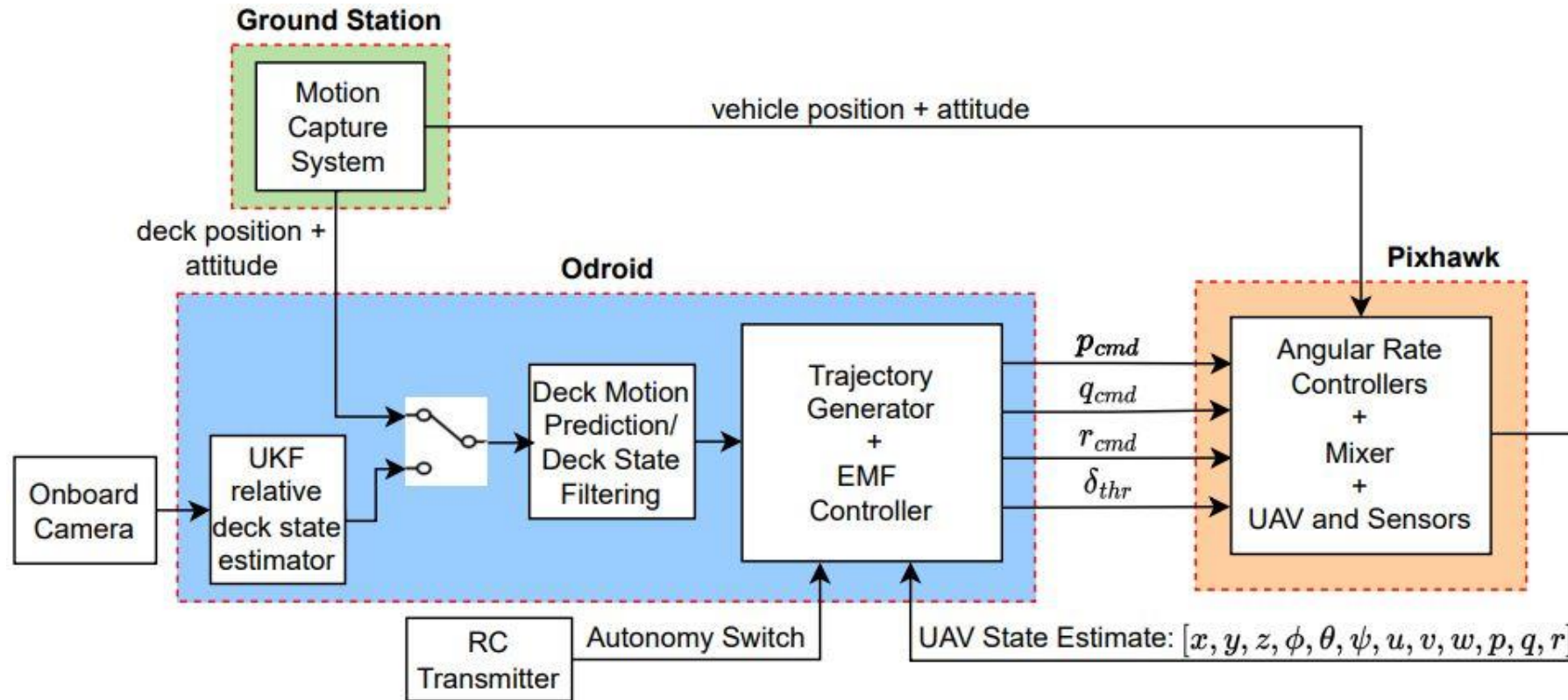
# UAV Platforms

- 2 UAVs: hexacopter and quadcopter
  - Odroid XU4 single board computer
  - Jevios a33 camera
  - Pixhawk Cube Orange autopilot + PX4 Firmware





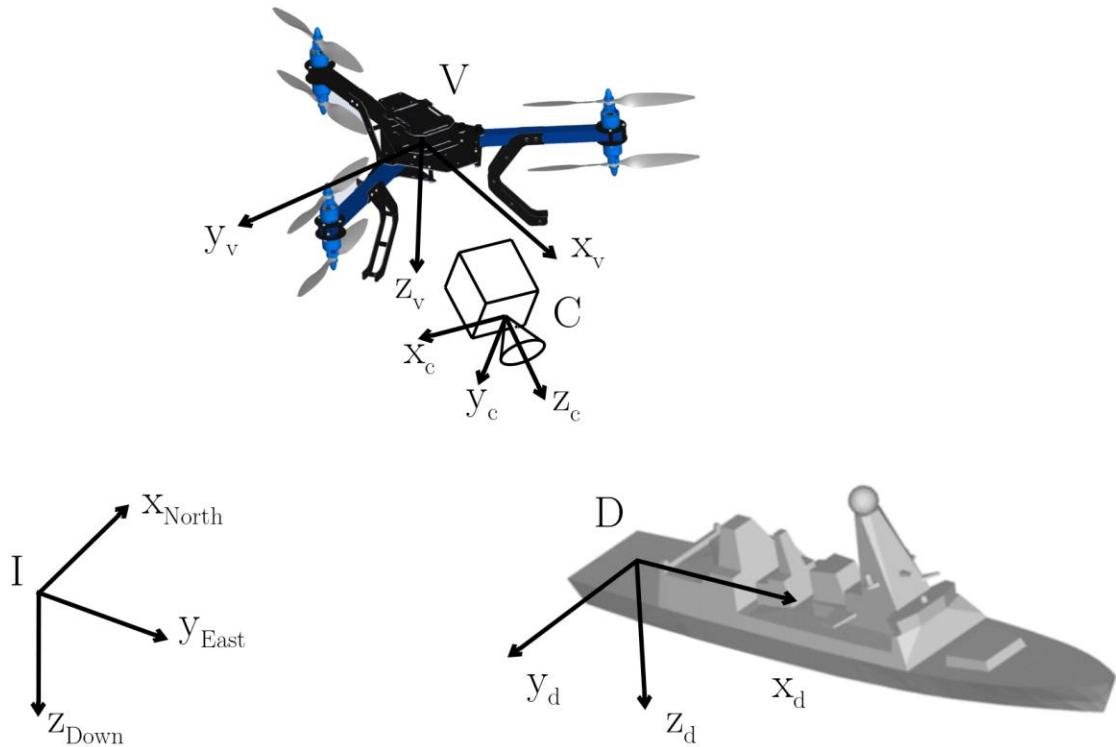
# Hardware and Software Integration



# Relative Deck State Estimator



# Vision Based Unscented Kalman Filter



$$\hat{\mathbf{x}} = \begin{bmatrix} \mathbf{r}_d^v \\ \Phi_{d/v} \\ \mathbf{v}_d^v \\ \boldsymbol{\omega}_{d/i} \end{bmatrix}$$

- Relative deck **position**
- Relative deck **attitude**
- Relative deck **velocity**
- Deck **angular velocity**

## Process Model

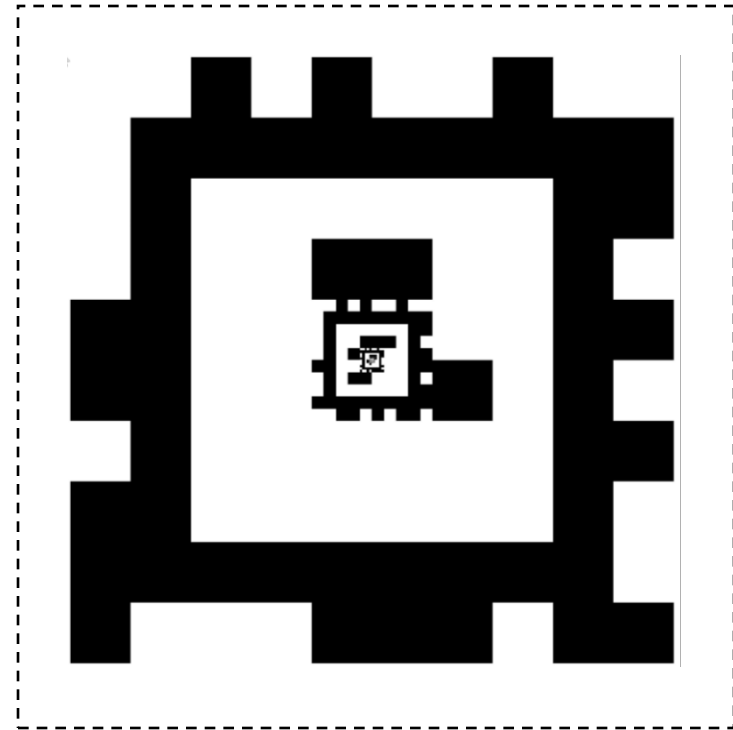
$$\ddot{\mathbf{r}}_{d/h}^h = \ddot{\mathbf{r}}_{d/h}^I - \dot{\boldsymbol{\omega}} \times \mathbf{r}_{d/h} - 2\boldsymbol{\omega} \times \dot{\mathbf{r}}_{d/h} - \boldsymbol{\omega} \times (\boldsymbol{\omega} \times \mathbf{r}_{d/h}^h)$$

# Scalable Fiducial Marker Arrays

- Robustly identify desired landing area
- Measure deck pose accurately from a wide range of distances



$$\mathbf{z}_{cam} = \begin{bmatrix} \mathbf{r}_{d/c}^c \\ \mathbf{q}_{d/c} \end{bmatrix}$$



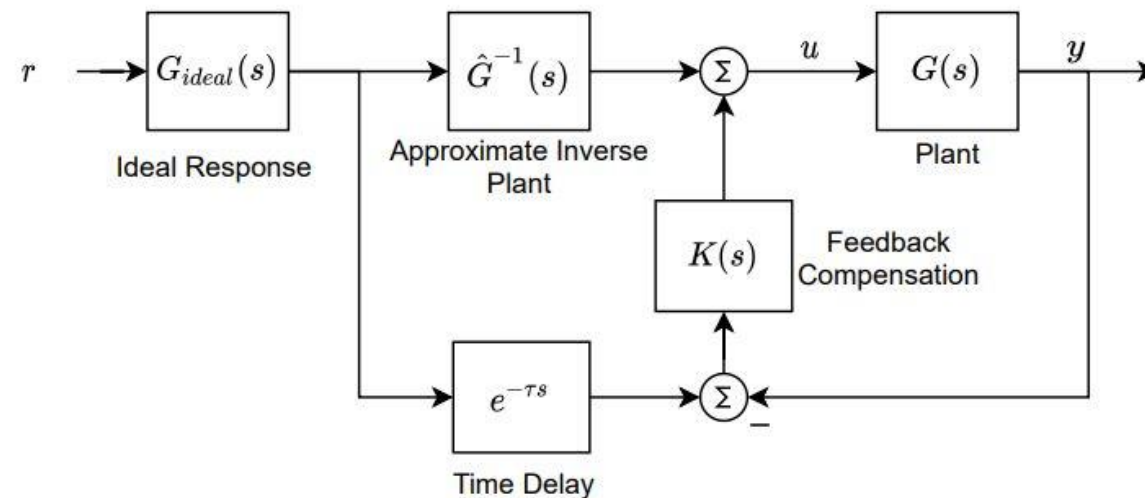
# Flight Control and Autonomy Algorithms



# Explicit Model Following Position Controllers

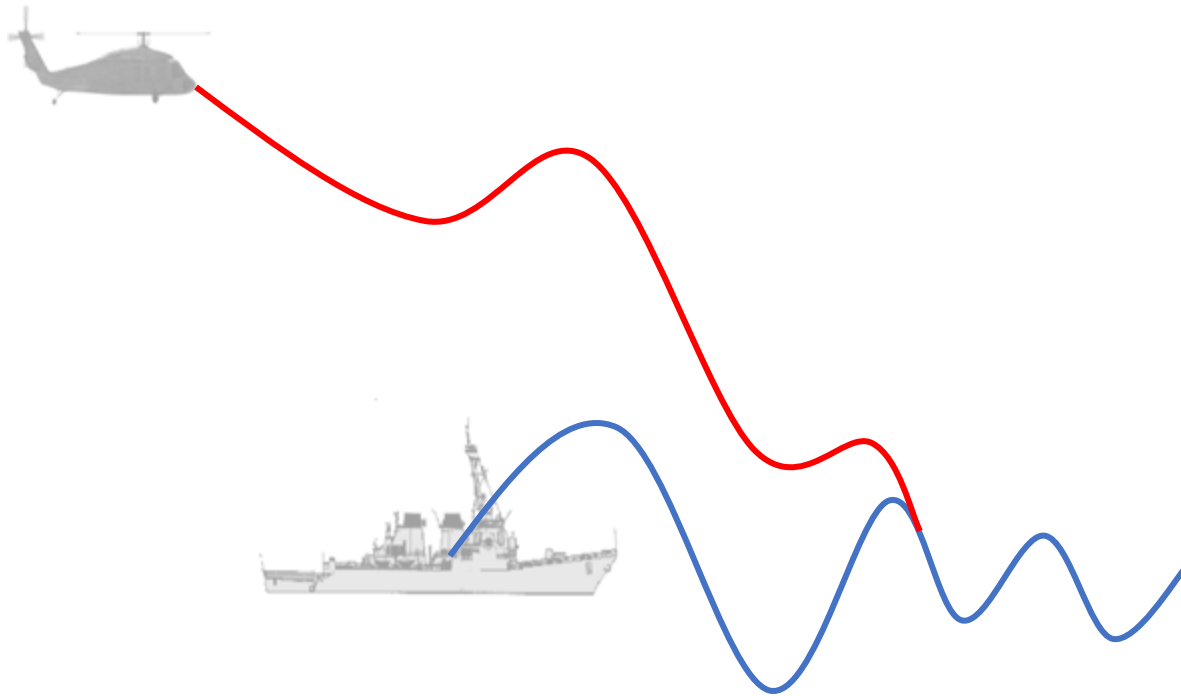
- Motivation – easily varied bandwidths
  - Reference tracking tuned through  $G_{ideal}(s)$
  - DRB tuned through  $K(s)$
- Froude scaled control:

$$N_F = \left( \frac{M_{fs}}{M_{ms}} \right)^{1/3} \rightarrow \omega_{ms} = \omega_{fs} \sqrt{N_F}$$

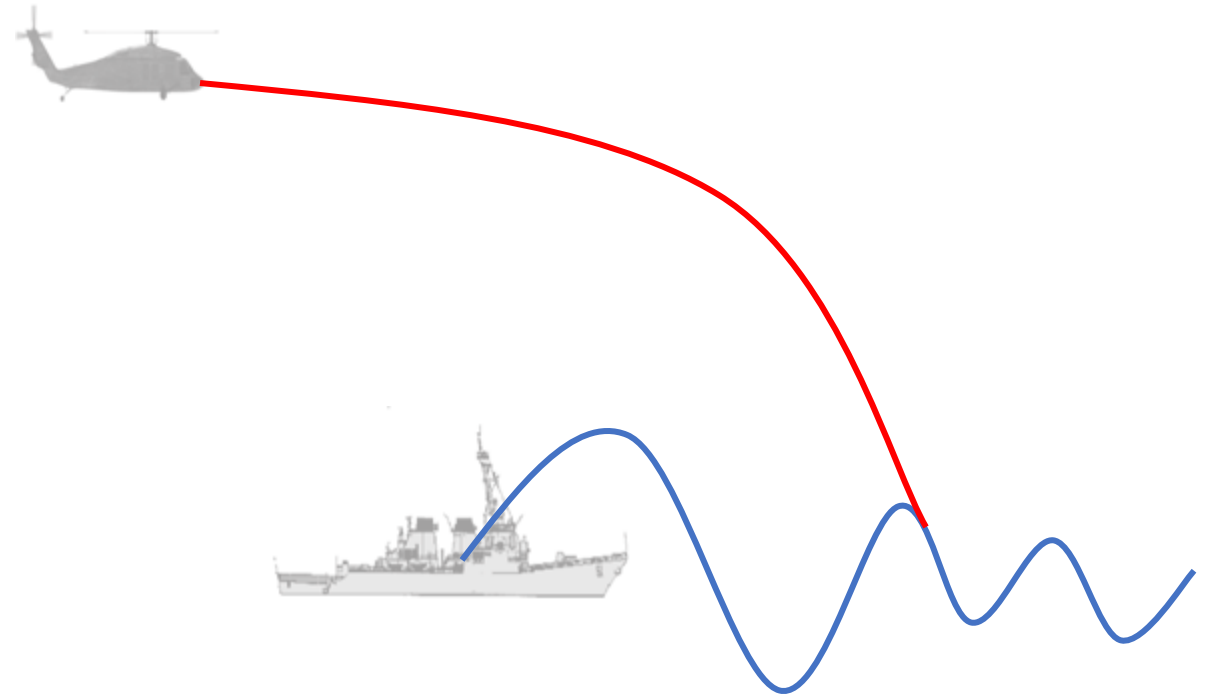


# Path Planning Algorithms

Deck Relative Commands

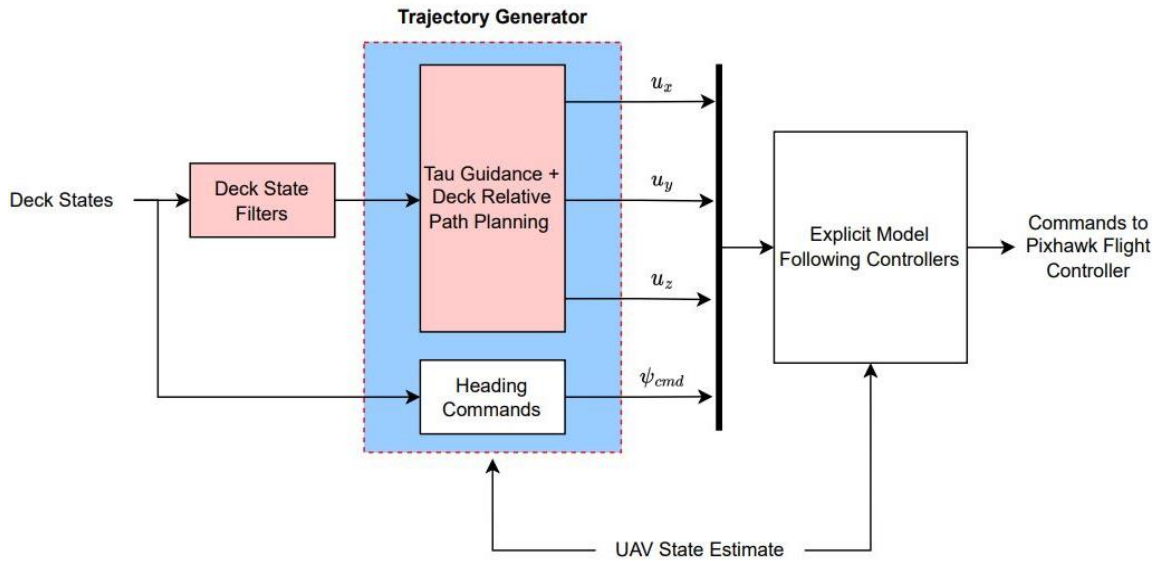


Planning to Predictions

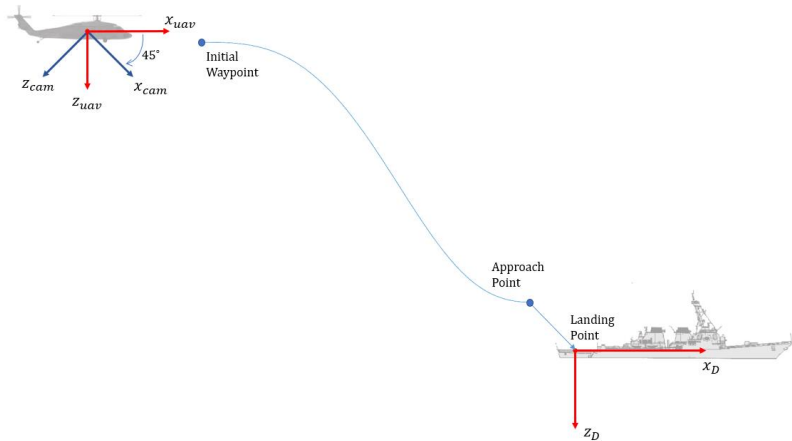
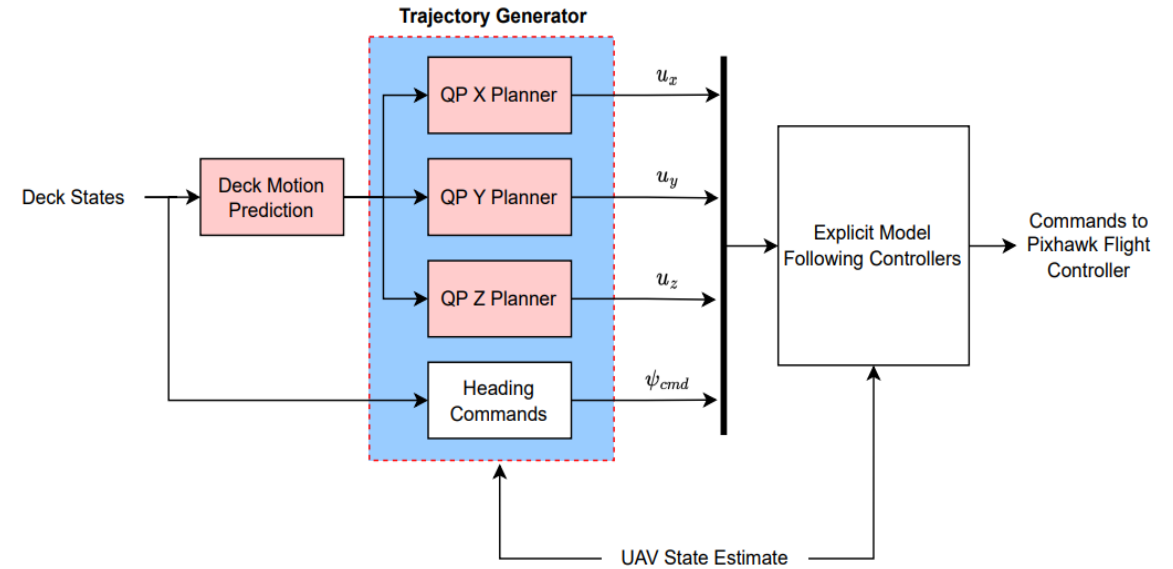


# Baseline and QP Landing Algorithms

## Baseline Path Planner



## Quadratic Programming Path Planner



**Discrete Model**

$$\begin{aligned} \vec{x}_{k+1} &= A\vec{x}_k + Bu_{k-\tau_d} \\ \vec{y}_{k+1} &= A\vec{x}_k + Bu_{k-\tau_d} \end{aligned}$$

**Cost Function**

reference trajectory and control input

$$J = \sum_{k=0}^{N-1} \left[ (\vec{y}_{ref,k} - \vec{y}_k)^T Q (\vec{y}_{ref,k} - \vec{y}_k) + u_k^T R u_k \right]$$

terminal cost

$$+ u_N^T R u_N + N \left[ (\vec{y}_{ref,N} - \vec{y}_N)^T S (\vec{y}_{ref,N} - \vec{y}_N) + j_N^T S_{\Delta} j_N \right]$$

**Standard QP Form**

$$J = \frac{1}{2} \bar{U}^T H \bar{U} + F^T \bar{U}$$

s.t.  $A_c \bar{U} \leq b_0$

**Constraints**

$$\vec{y}_{low,k} \leq \vec{y}_k \leq \vec{y}_{up,k}$$



# Autoregressive Models For Deck Forecasting

## Define Output Vectors

$$\vec{y}_{long} = [X_d^{dhf} \quad \dot{X}_d^{dhf} \quad \theta_d \quad Z_d^l \quad \dot{Z}_d^l]^T$$

$$\vec{y}_{lat} = [Y_d^{dhf} \quad \dot{Y}_d^{dhf} \quad \phi_d \quad \psi_d]^T$$

↓

## AR Model

$$\vec{y}_k = \alpha_1 \vec{y}_{k-1} + \alpha_2 \vec{y}_{k-2} + \dots + \alpha_{N_{lag}} \vec{y}_{k-N_{lag}} + \vec{v}_k$$

↓

## Estimate

$$\alpha_1 \quad \dots \quad \alpha_{N_{lag}}$$

↓

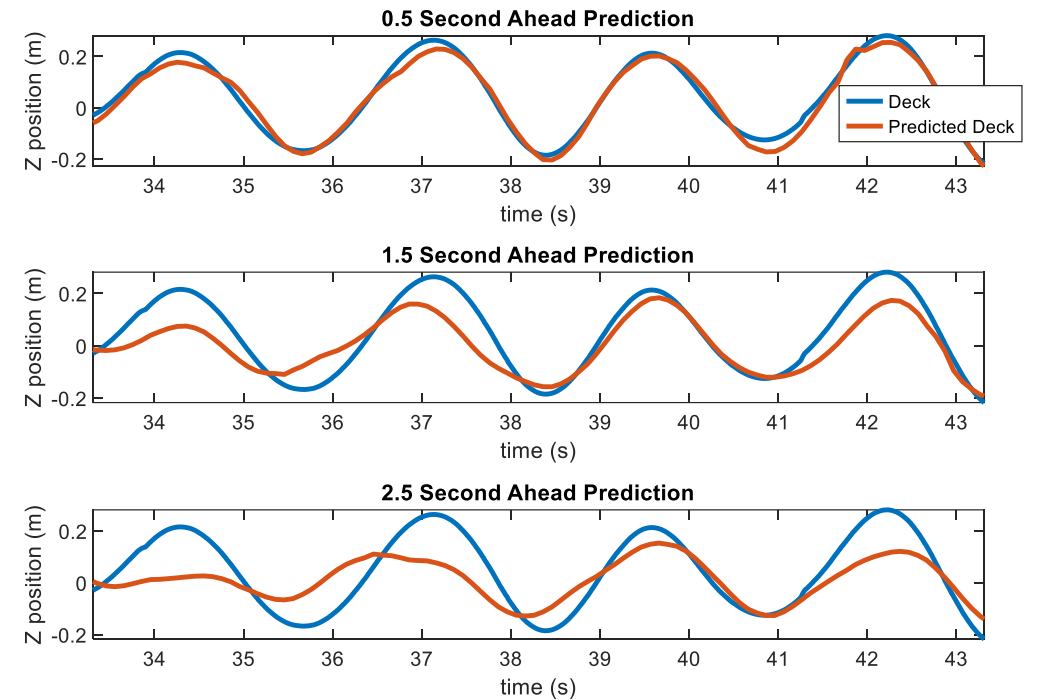
## Propagate

$$\vec{y}_{k+1} = \alpha_1 \vec{y}_k + \alpha_2 \vec{y}_{k-1} + \dots + \alpha_{N_{lag}} \vec{y}_{k-N_{lag}+1}$$

⋮

$$\vec{y}_{k+N} = \alpha_1 \vec{y}_{k-1+N} + \alpha_2 \vec{y}_{k-2+N} + \dots + \alpha_{N_{lag}} \vec{y}_{k-N_{lag}+N}$$

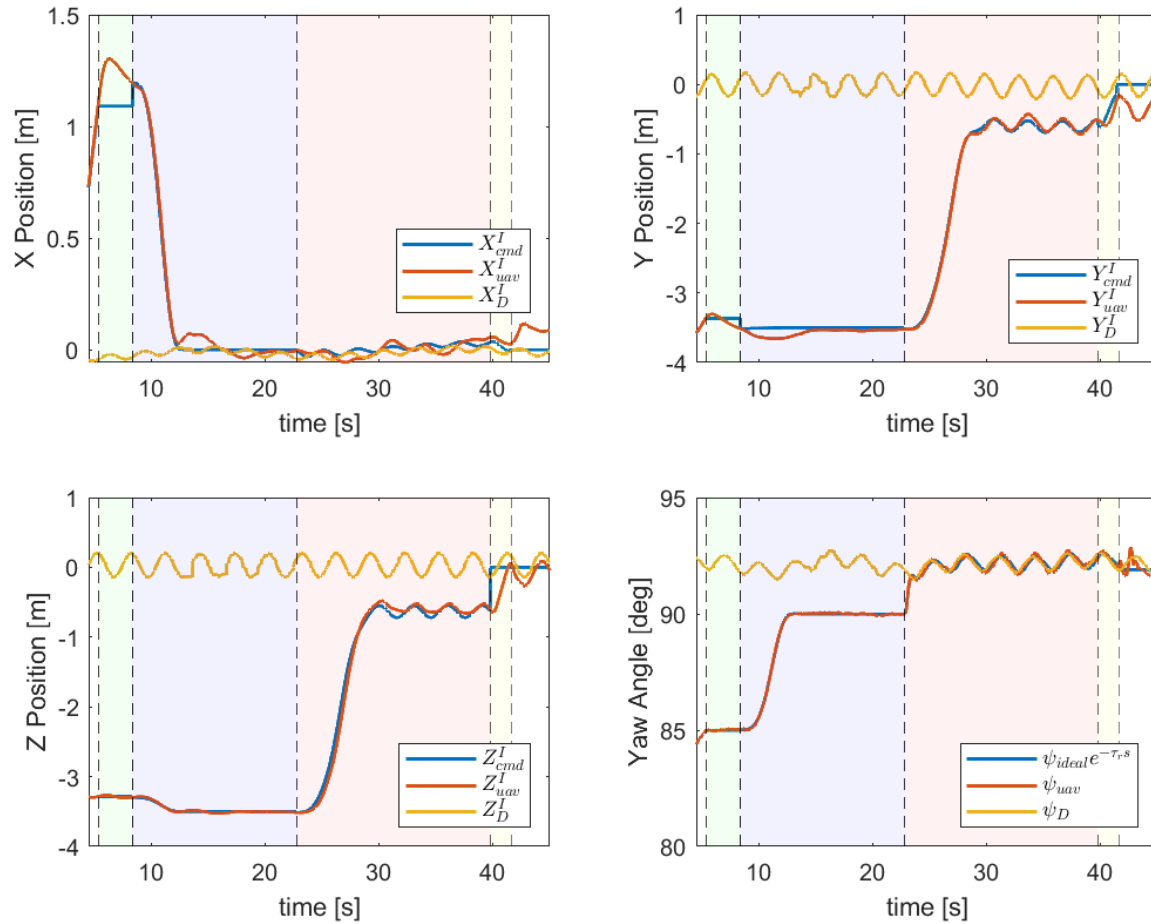
Deck Heave Predictions, Wave Condition 2, Flight 38



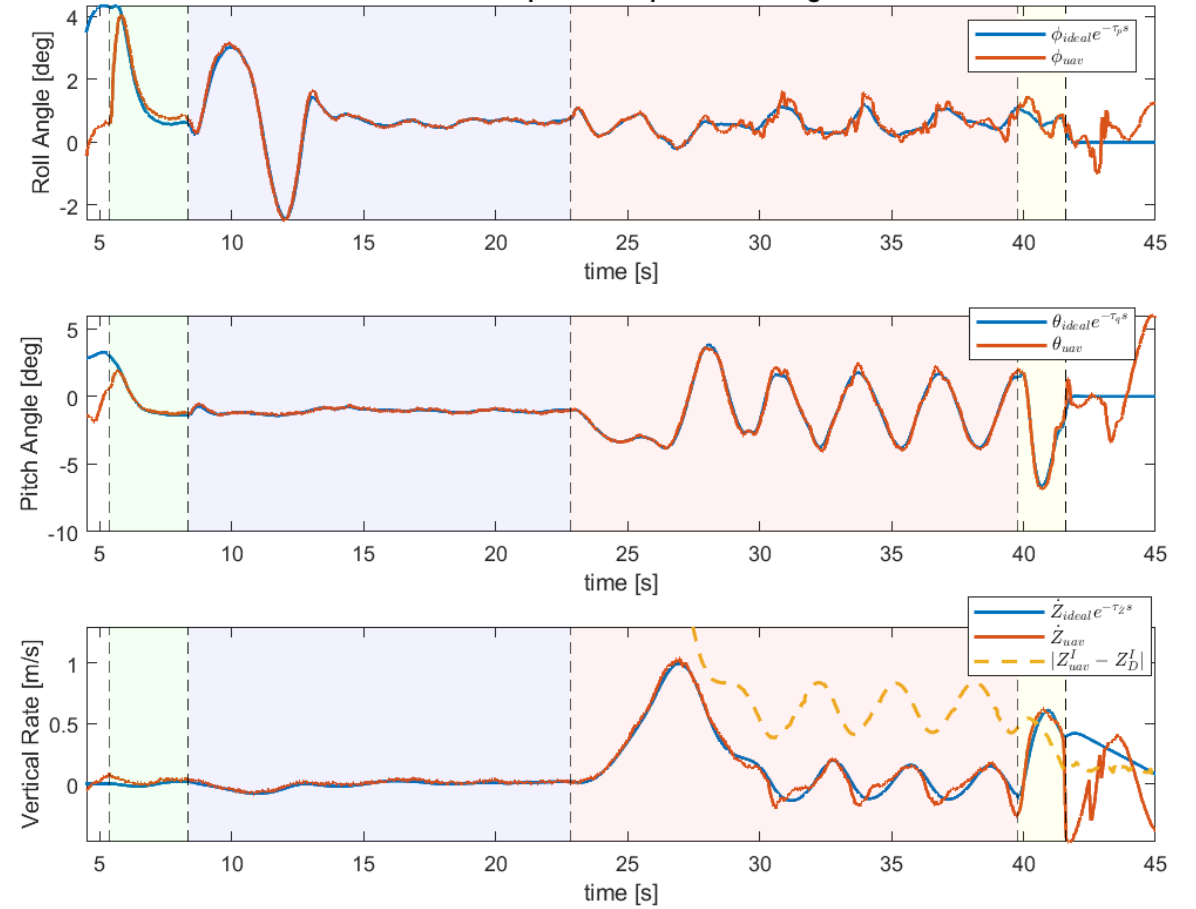
# First Tests at the MASK

# Control Verification

### Landing Sequence, High Amplitude Waves



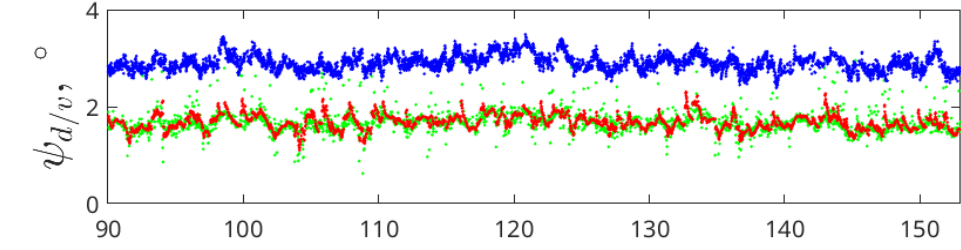
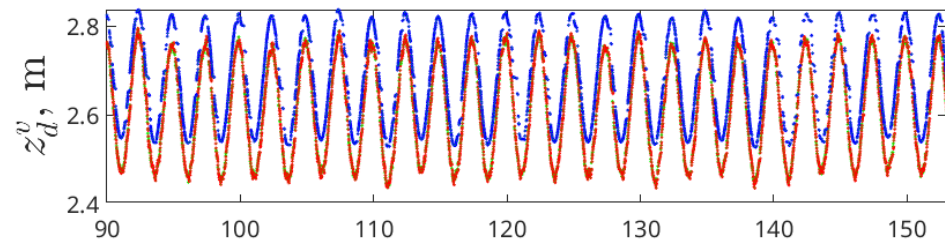
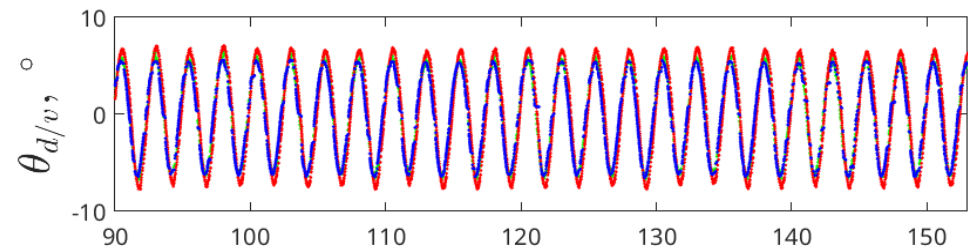
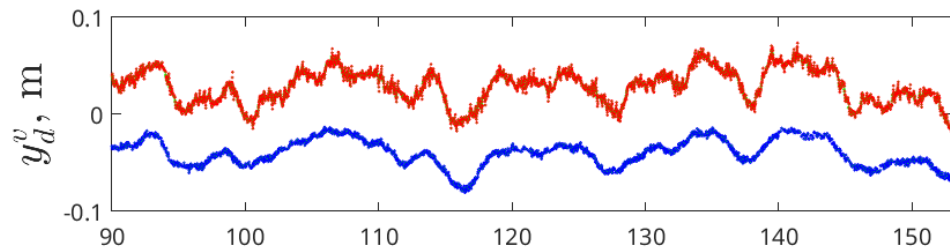
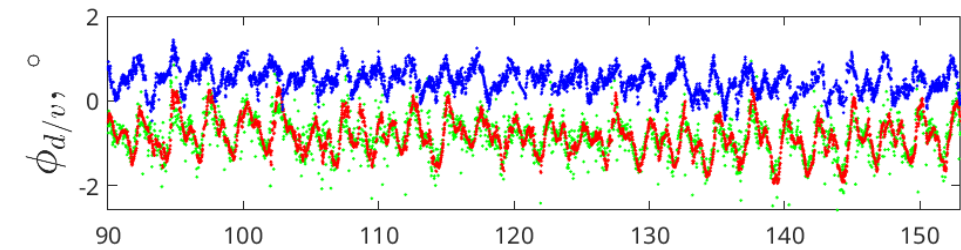
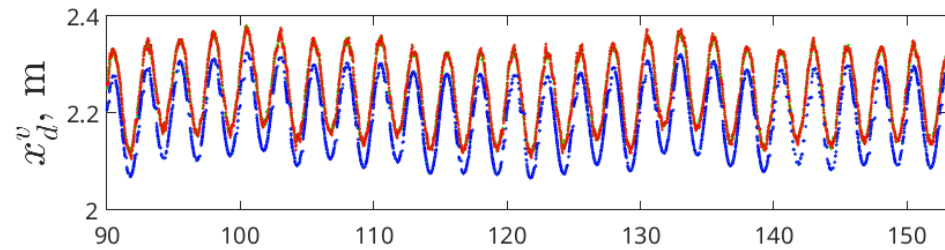
### Inner Loop Ideal Response Tracking





# Estimator Verification in Hover

- Measurement
- Estimated
- Ground Truth

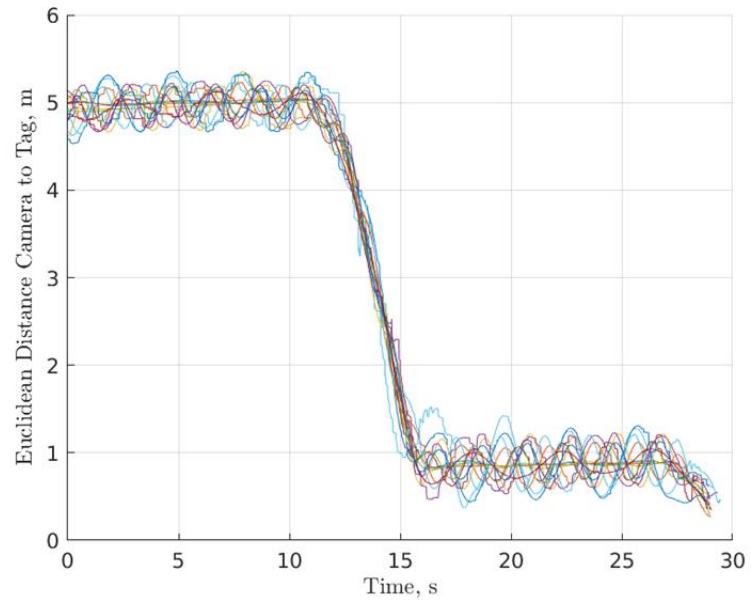


Time, s

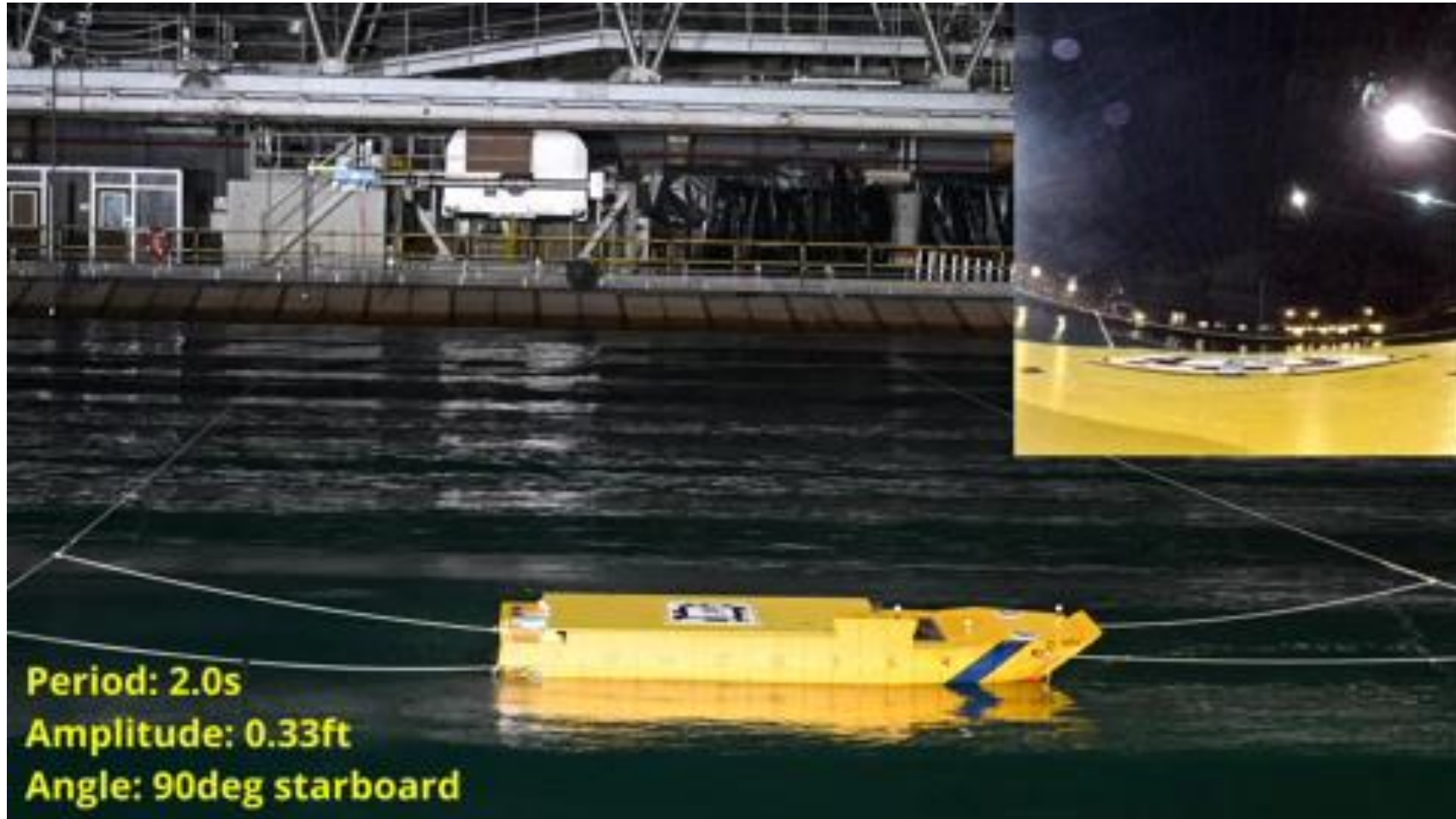
Time, s

# Vision Based Landings

15 Vision Based Landings



# Vision Based Landings

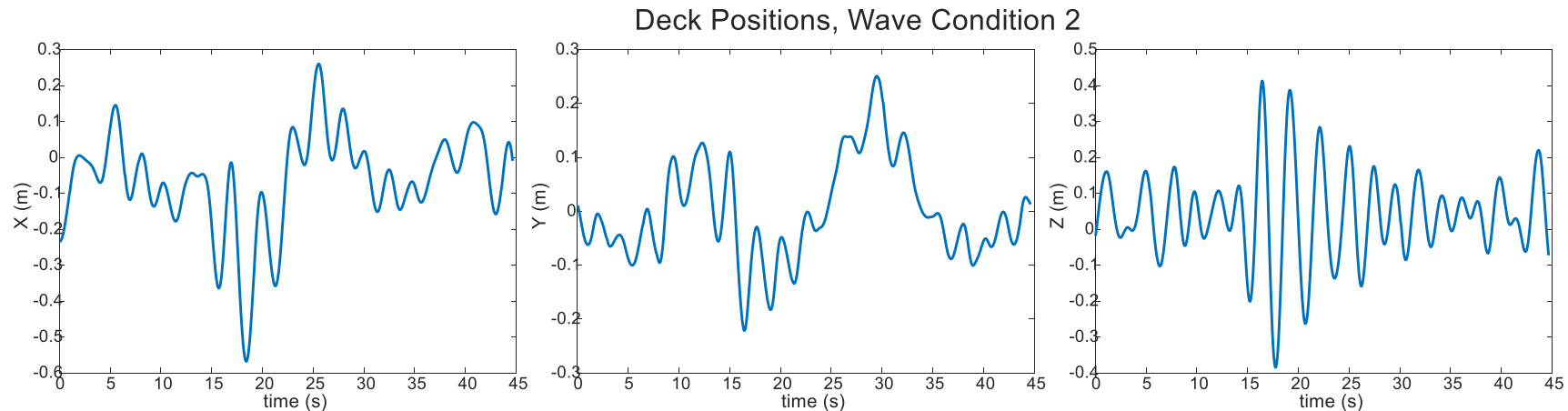
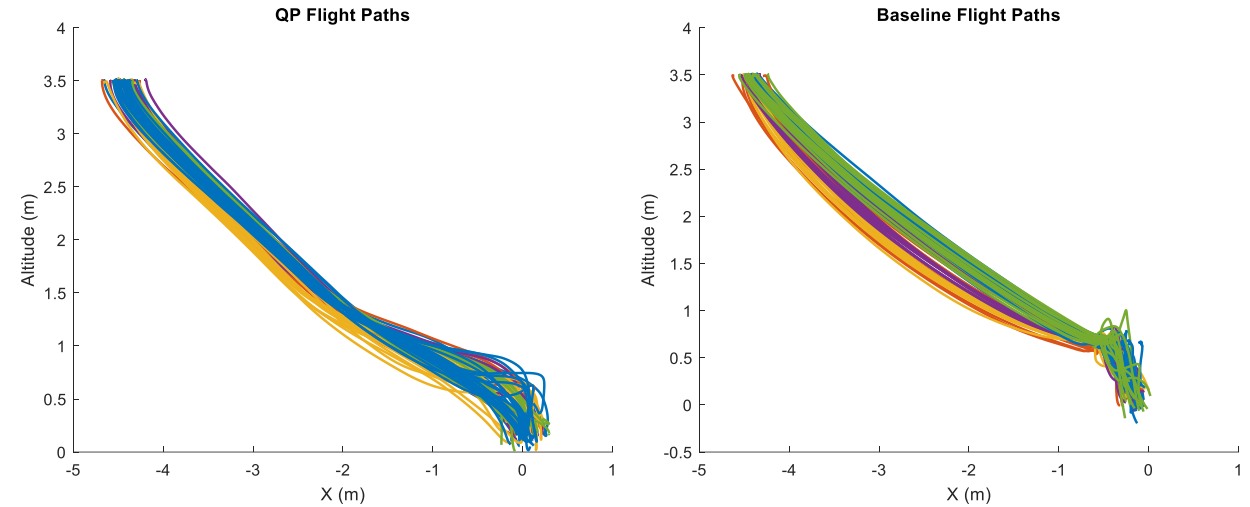


# Recent Tests at the MASK



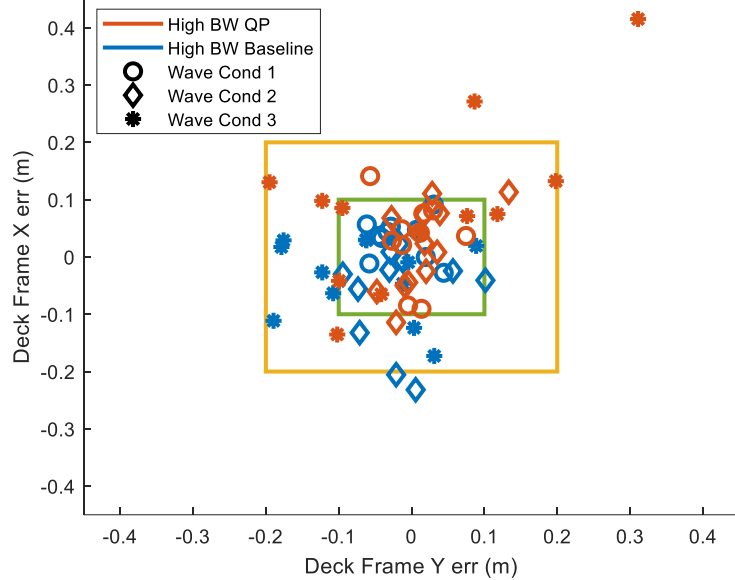
# More Recent Tests

- 162 recorded landings
- Focus on path planning and control
  - QP vs baseline
  - Varied tracking bandwidths
  - 3 different stochastic wave conditions

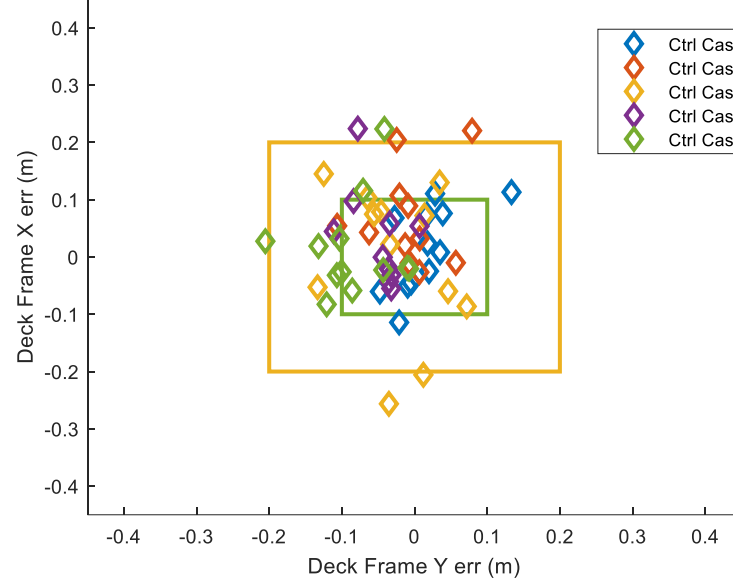


# X and Y Landing Errors

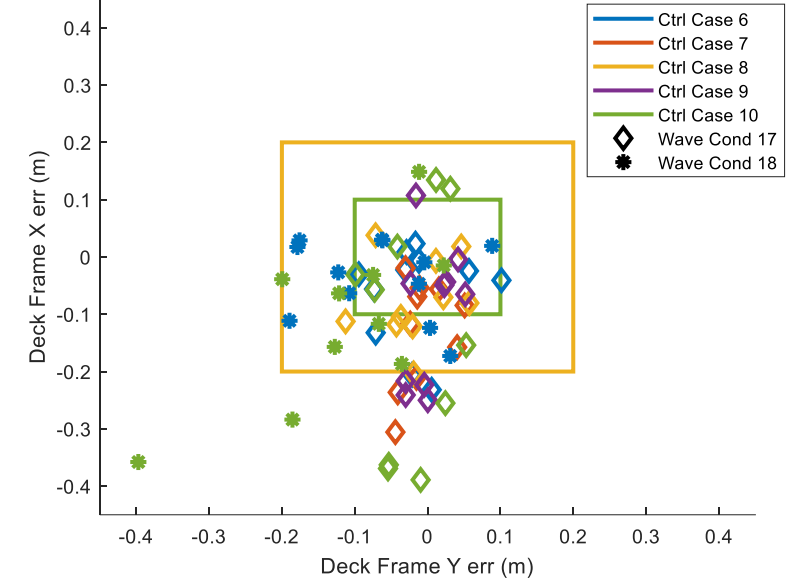
X and Y Position Landing Error, High BW Baseline vs High BW QP



X and Y Position Landing Error, QP Landings

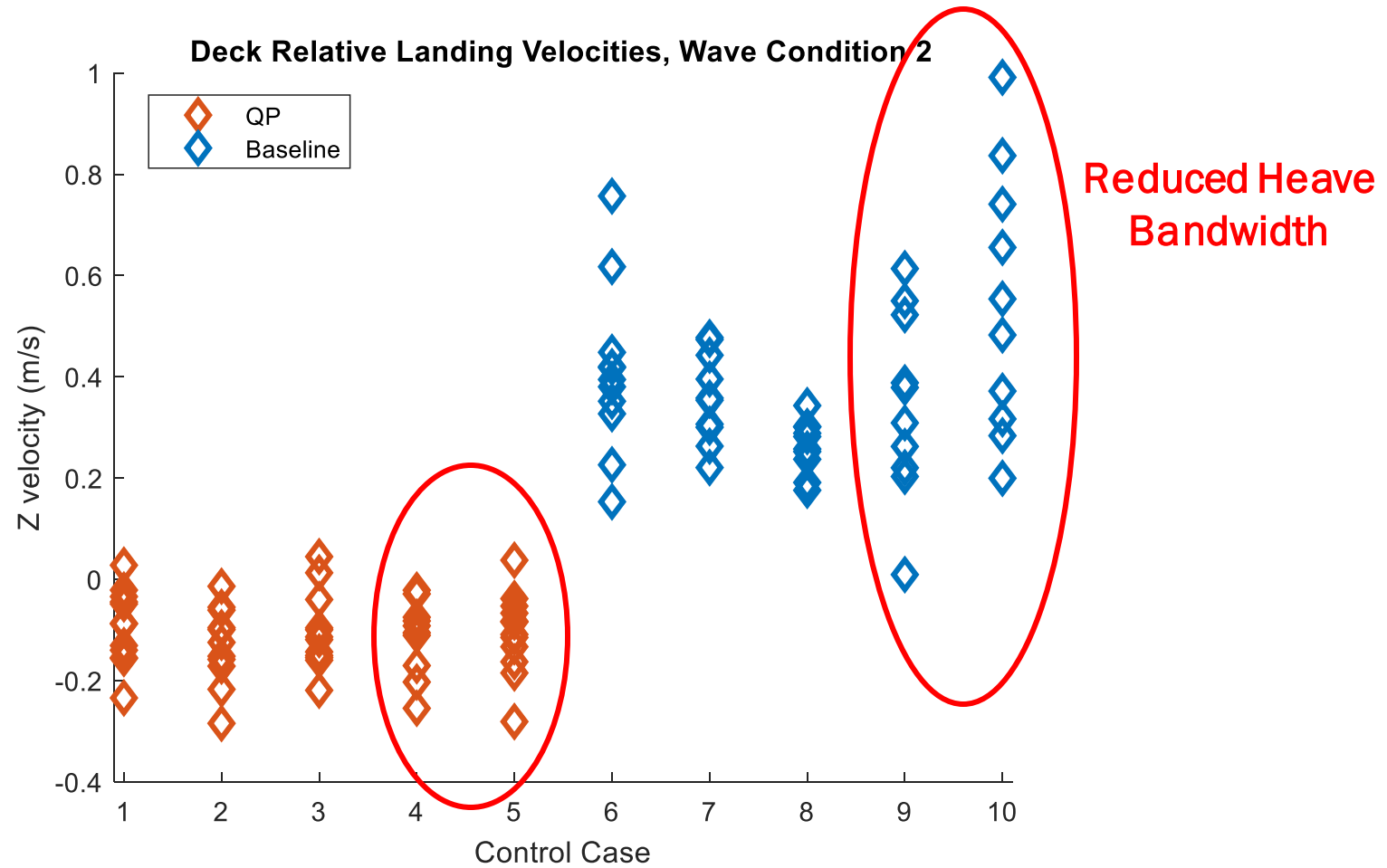


X and Y Position Landing Error, Baseline Landings



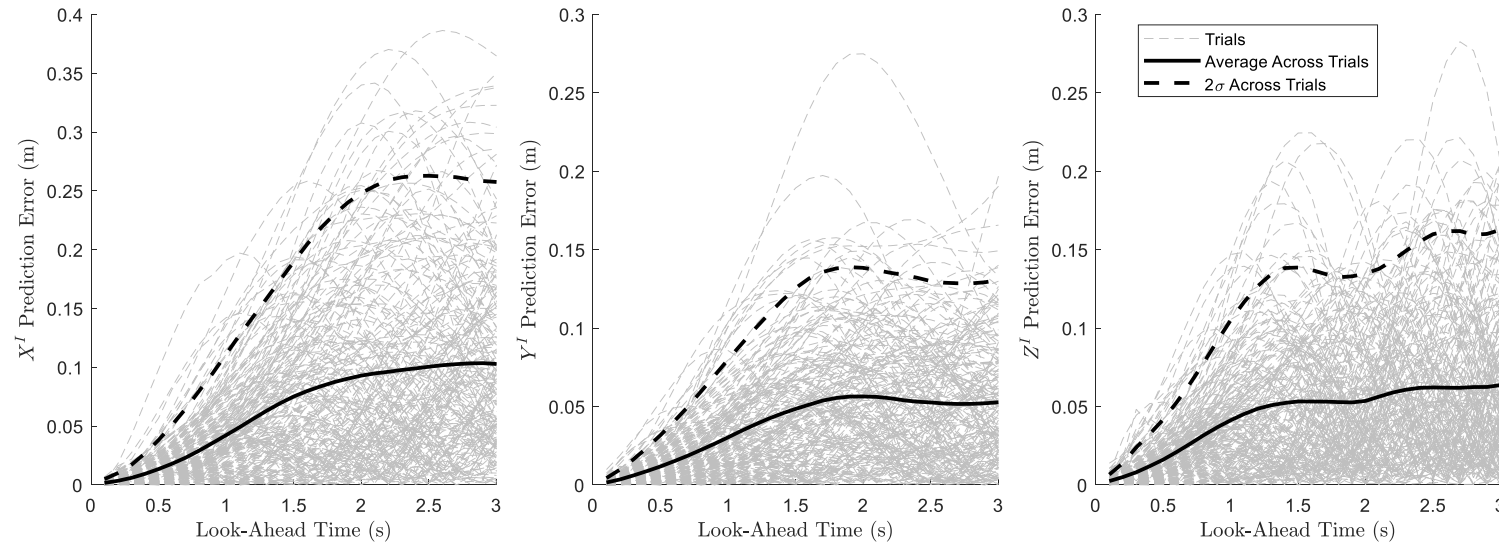


# X and Y Landing Errors

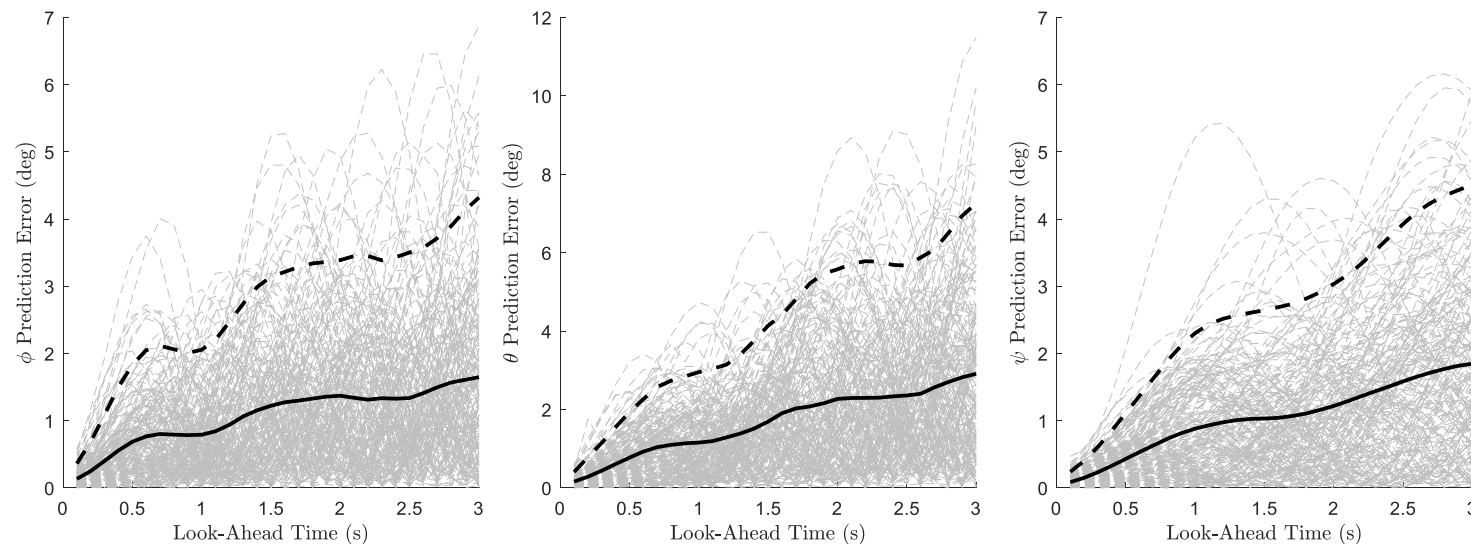


# Deck Prediction Accuracy

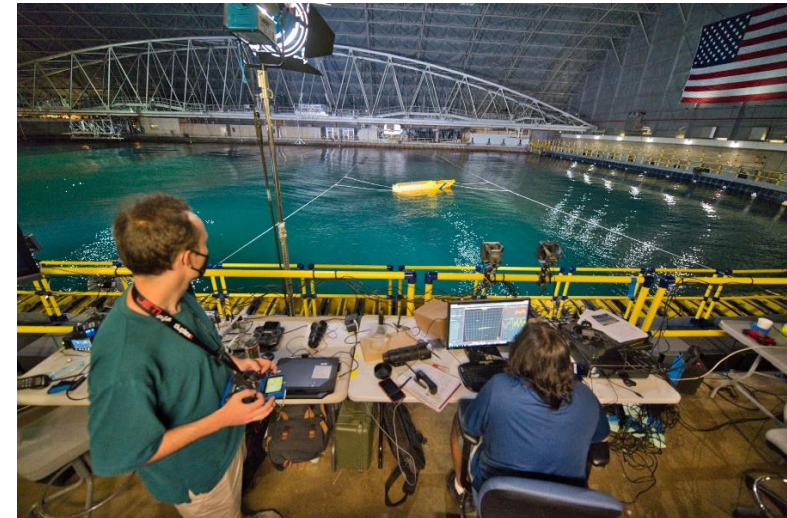
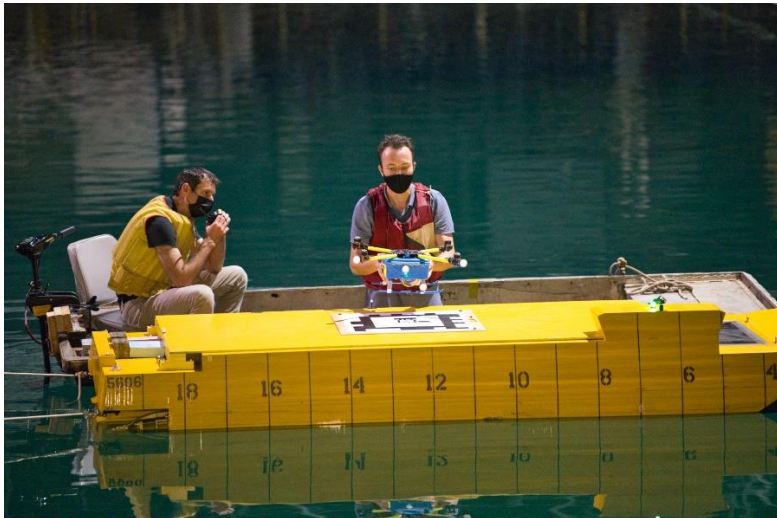
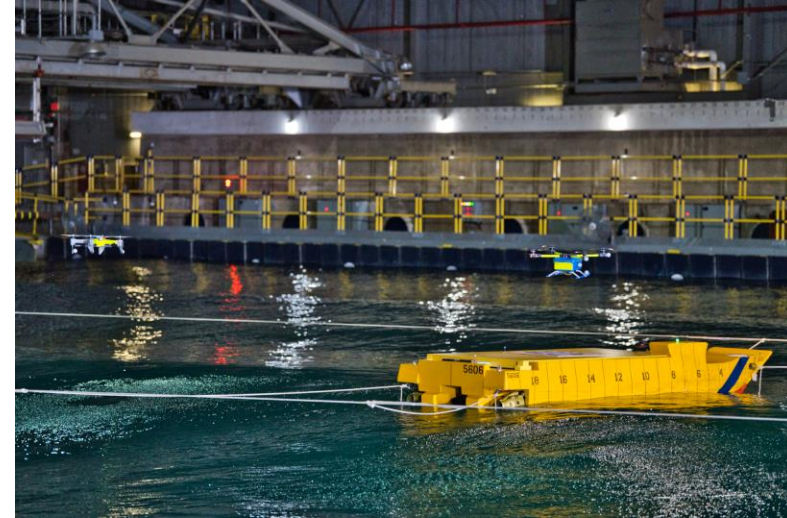
Forecasted Deck Position Errors, Wave Condition 2



Forecasted Deck Attitude Errors, Wave Condition 2



# Questions

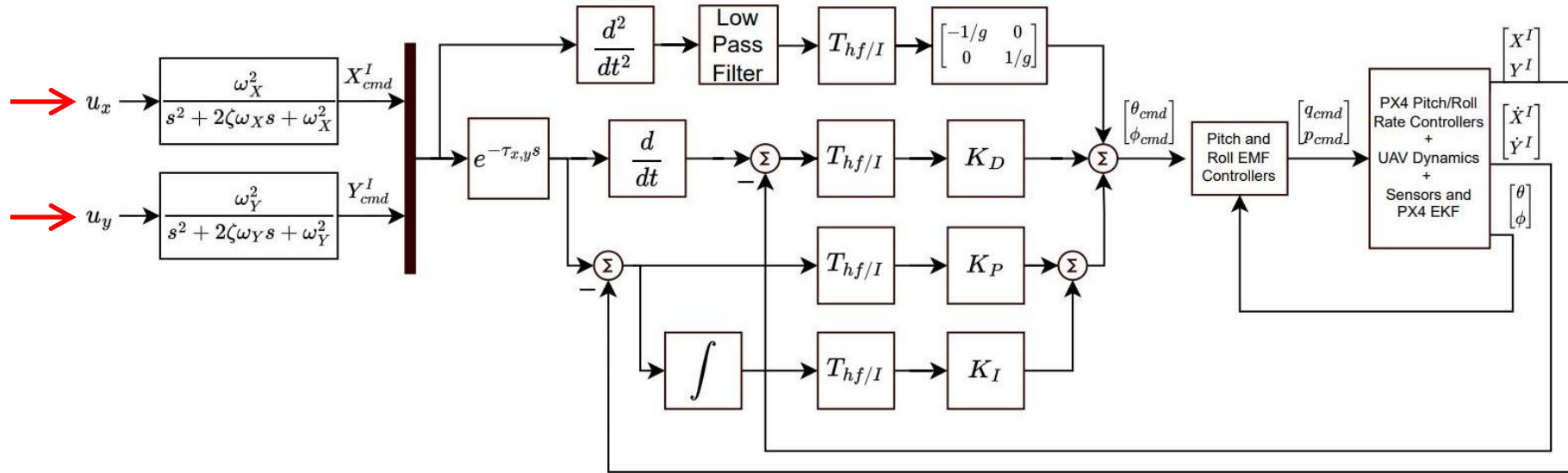


# Appendix



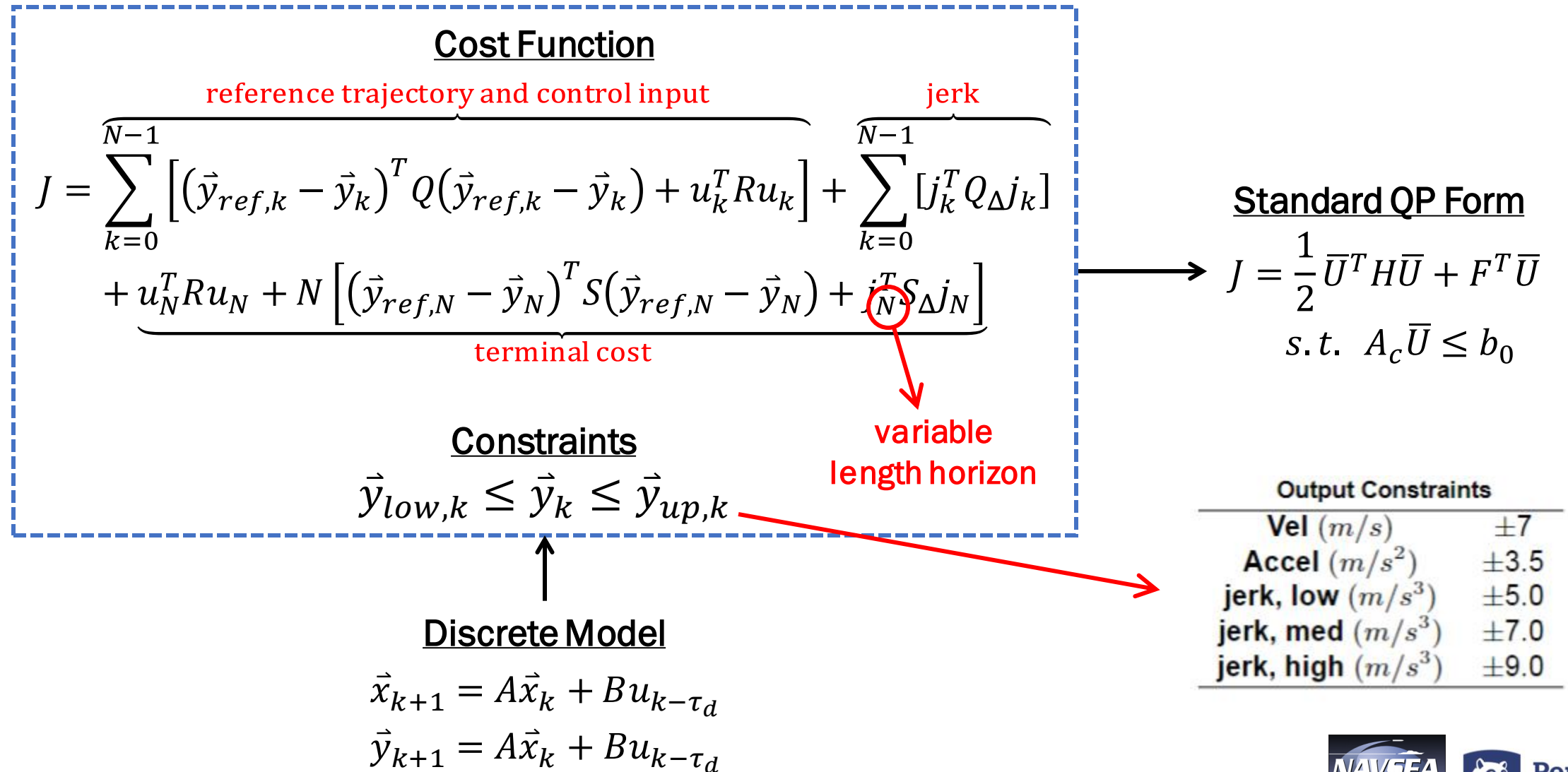
# Outer Loops and Froude Scaled Control

From Path Planner



- Tracking bandwidth:  $\omega_{\theta,fs} \rightarrow \omega_{\theta,cf} = \omega_{\theta,fs} \sqrt{N_F} \rightarrow \omega_{X,cf} = \frac{\omega_{\theta,cf}}{5}$
- Outer loop delay:  $\frac{\theta(s)}{\theta_{cmd}(s)} \approx \frac{\omega_{\theta,cf}^2}{s^2 + 2\zeta\omega_{\theta,cf}s + \omega_{\theta,cf}^2} e^{-\tau_{\theta}s} \rightarrow \tau_x = \frac{1.65}{\omega_{\theta,cf}} + \tau_{\theta}$
- PID gains set to meet DRB for scaled level 1 HQ

# Quadratic Program Transcription





# Discrete Model for QP Trajectory Generation

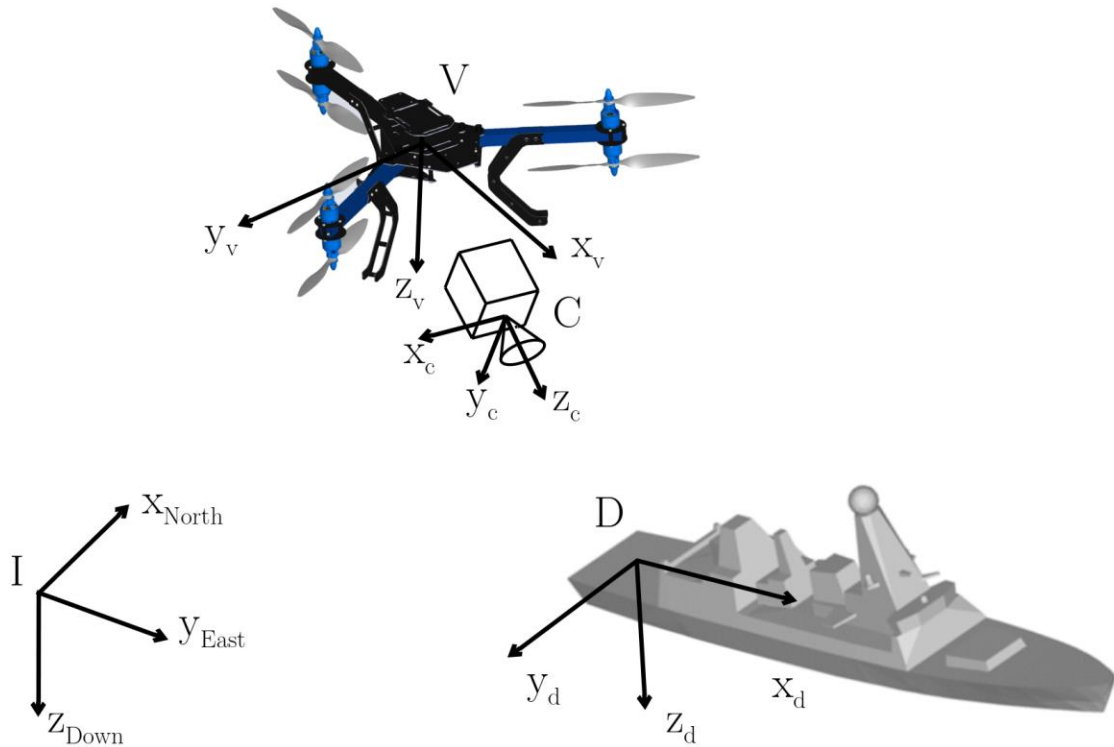
- Separate QP solvers for inertial  $X$ ,  $Y$ , and  $Z$  commands
- Dynamics modeled as theoretical ideal result for EMF position controllers:

$$G_X(s) = \frac{\omega_{X,cf}^2}{s^2 + 2\zeta\omega_{X,cf}s + \omega_{X,cf}^2} e^{-\tau_X s} \rightarrow \begin{aligned} \vec{x}_{X,k+1} &= A_X \vec{x}_{X,k} + B_X u_{X,k-\tau_d} \\ \vec{y}_{X,k} &= C_X \vec{x}_{X,k} + D_X u_{X,k-\tau_d} \end{aligned}$$

- Outputs are pos, vel, and accel
- Approximate jerk with a back difference:

$$j_k = \frac{a_k - a_{k-1}}{\Delta t}$$

# Vision Based Unscented Kalman Filter



$$\hat{\mathbf{x}} = \begin{bmatrix} \mathbf{r}_d^v \\ \Phi_{d/v} \\ \mathbf{v}_d^v \\ \boldsymbol{\omega}_{d/i} \end{bmatrix}$$

- Relative deck position
- Relative deck attitude
- Relative deck velocity
- Deck angular velocity

## Process Model

$$\dot{\mathbf{r}}_{d/h}^h = \underbrace{\dot{\mathbf{r}}_{d/h}^I}_{\mathbf{0}} - \dot{\boldsymbol{\omega}} \times \mathbf{r}_{d/h} - 2\boldsymbol{\omega} \times \dot{\mathbf{r}}_{d/h} - \boldsymbol{\omega} \times (\boldsymbol{\omega} \times \mathbf{r}_{d/h}^h)$$

Assume  $\dot{\mathbf{r}}_d^I = \mathbf{0}$  and capture with process noise

# Pose Estimation During Landing

