

Navigation and the Kalman Filter

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AREUS Lab Seminar Series Week 2

February 18, 2019



Mechanical and Aerospace Engineering

Introduction

Guidance

Introduction

Guidance

Navigation

Introduction

Guidance

Navigation

Control

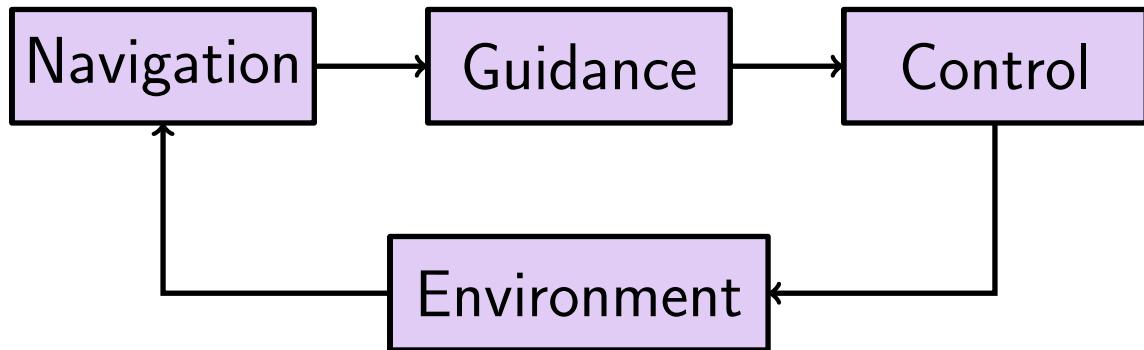
Introduction

Navigation

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Control

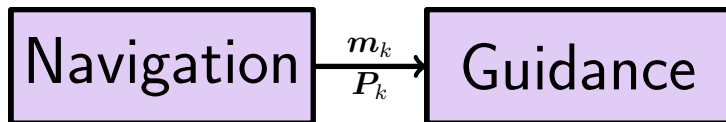
Introduction



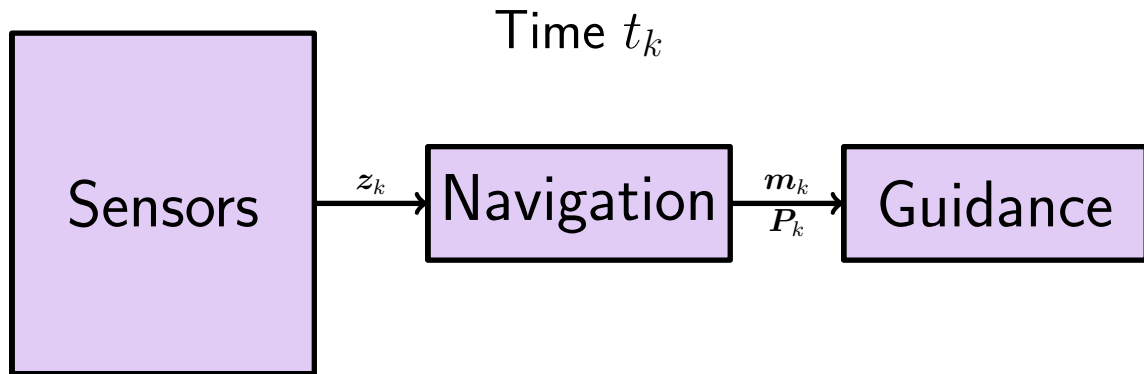
Navigation I/O

Time t_k Navigation

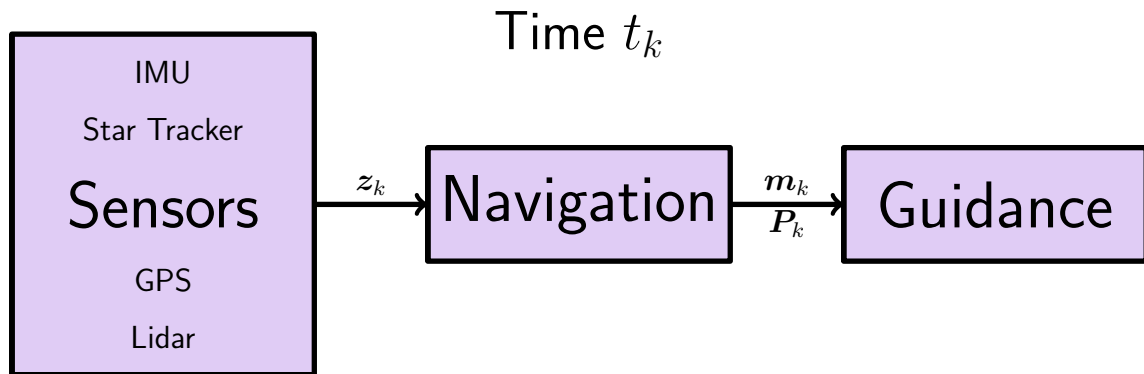
Navigation I/O

Time t_k 

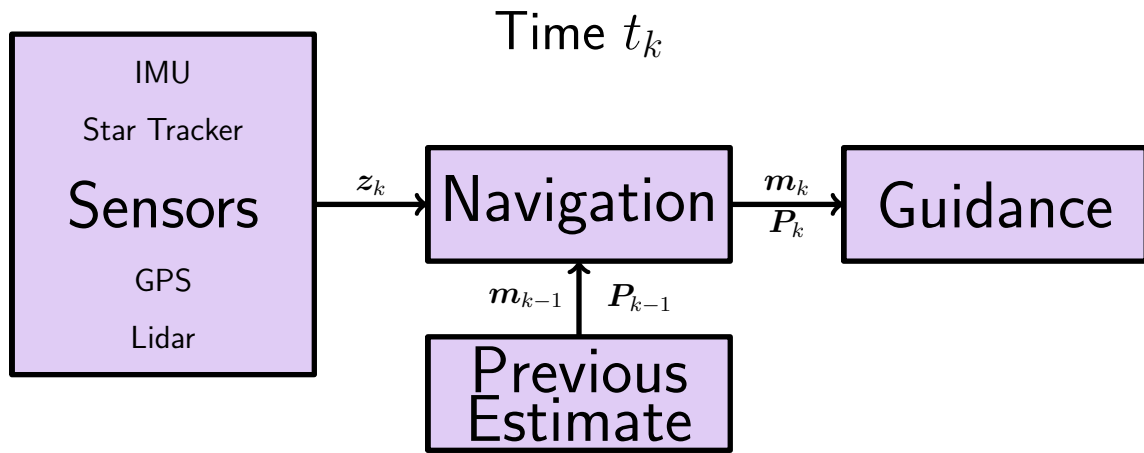
Navigation I/O



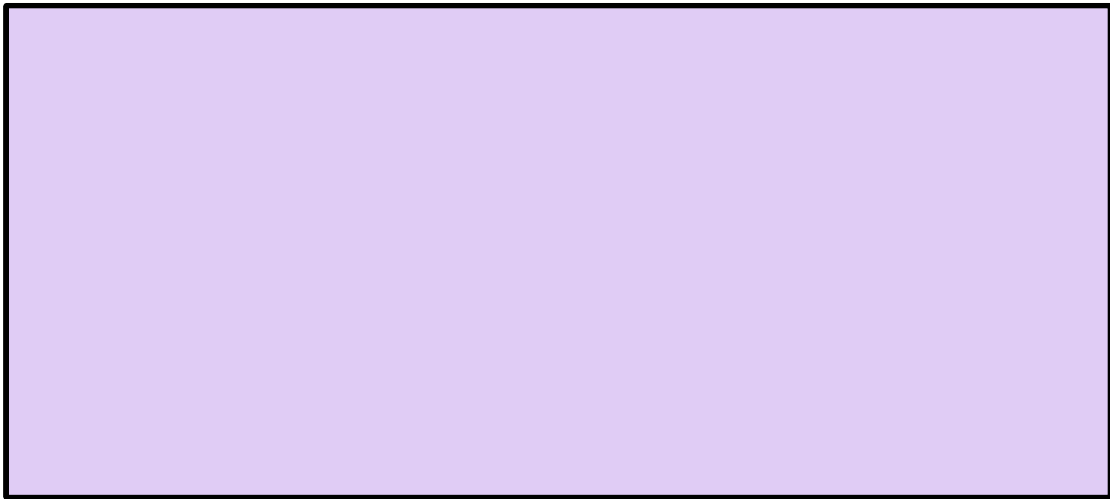
Navigation I/O



Navigation I/O



Kalman Filter



Kalman Filter

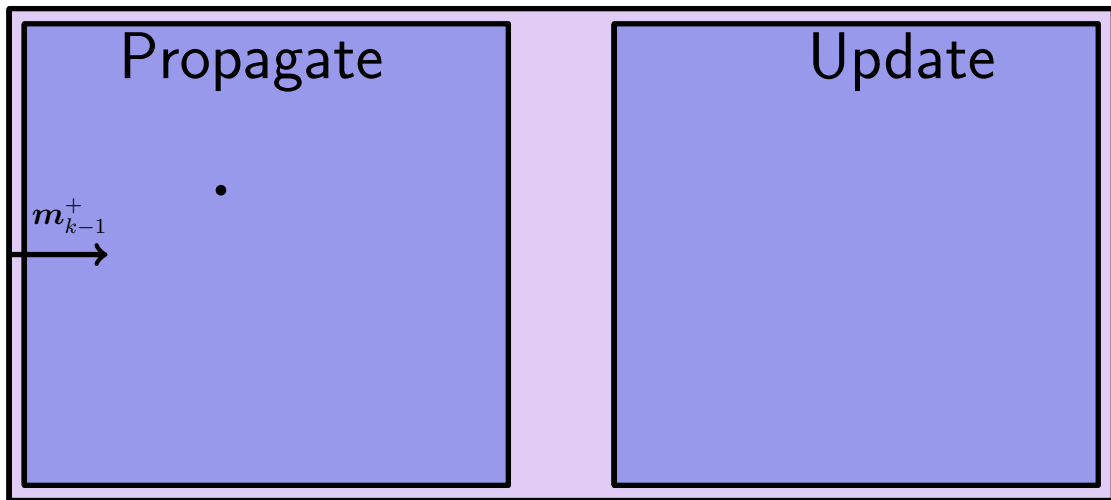
Propagate

Kalman Filter

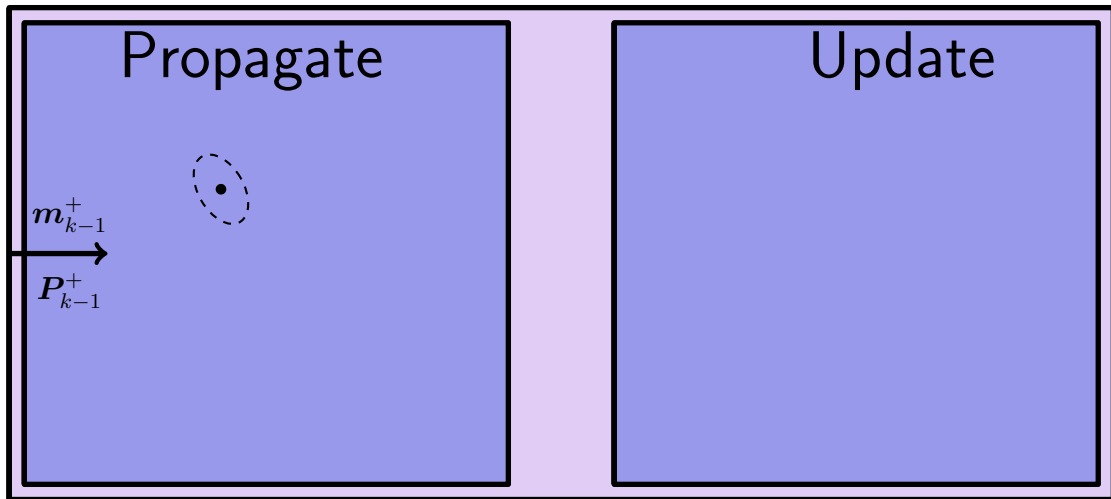
Propagate

Update

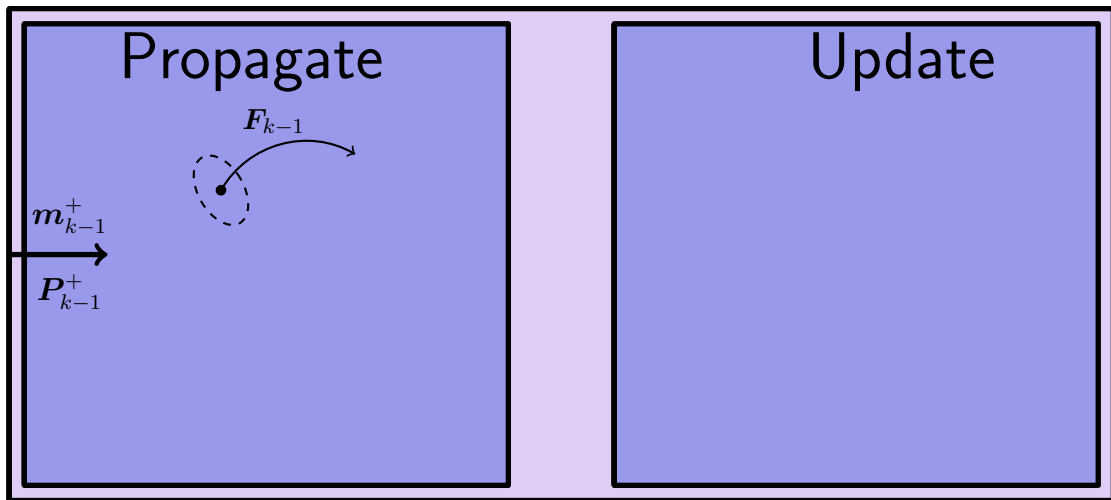
Kalman Filter



Kalman Filter

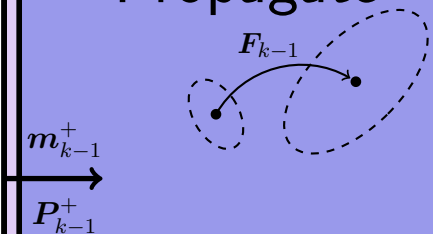


Kalman Filter



Kalman Filter

Propagate

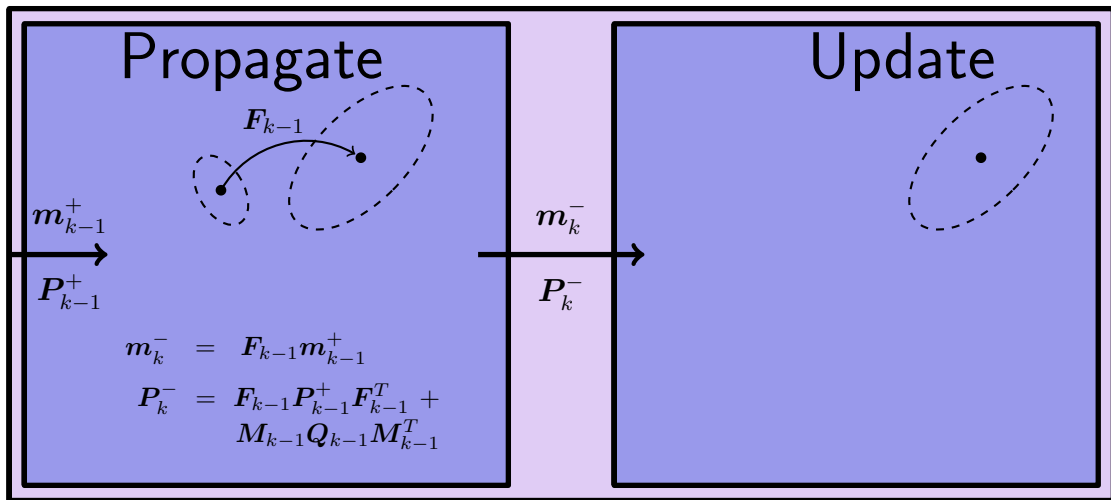


$$m_k^- = F_{k-1} m_{k-1}^+$$

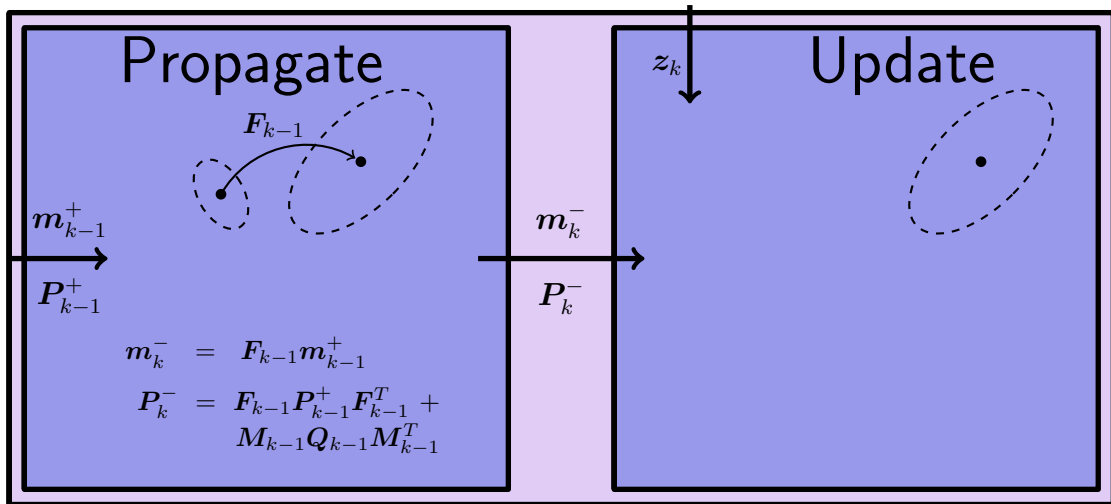
$$P_k^- = F_{k-1} P_{k-1}^+ F_{k-1}^T + M_{k-1} Q_{k-1} M_{k-1}^T$$

Update

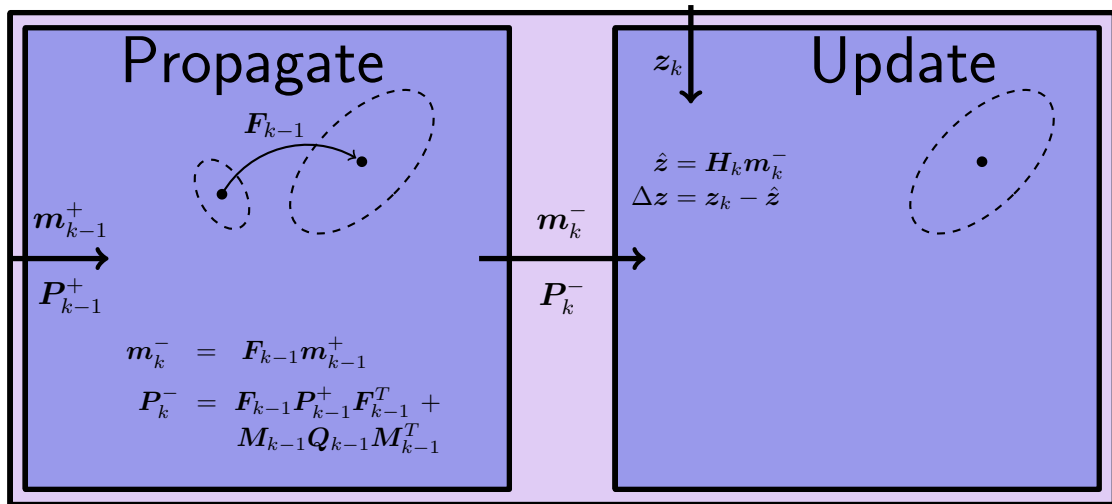
Kalman Filter



Kalman Filter



Kalman Filter



Kalman Filter

Propagate

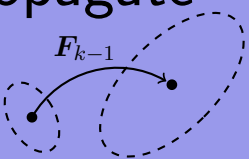


Diagram illustrating the Propagate step. An arrow labeled m_{k-1}^+ and P_{k-1}^+ enters the box. Inside, a curved arrow labeled F_{k-1} points from a small dashed ellipse to a larger dashed ellipse, representing the state transition. Below the diagram, the equations for the predicted state and covariance are given:

$$m_k^- = F_{k-1} m_{k-1}^+$$

$$P_k^- = F_{k-1} P_{k-1}^+ F_{k-1}^T + M_{k-1} Q_{k-1} M_{k-1}^T$$

 m_k^- P_k^-

Update




Diagram illustrating the Update step. A vertical arrow labeled z_k points down into the box. Inside, a dashed ellipse represents the updated state. The equations for the update are given:

$$\hat{z} = H_k m_k^-$$

$$\Delta z = z_k - \hat{z}$$

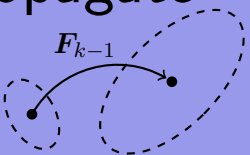
$$W_k = H_k P_k^- H_k^T + L_k R_k L_k^T$$

$$C_k = P_k^- H_k^T$$

$$K_k = P_k^- H_k^T W_k^{-1}$$

Kalman Filter

Propagate




$$m_k^- = F_{k-1} m_{k-1}^+$$

$$P_k^- = F_{k-1} P_{k-1}^+ F_{k-1}^T + M_{k-1} Q_{k-1} M_{k-1}^T$$

 m_k^- P_k^-

Update



$$\hat{z} = H_k m_k^-$$

$$\Delta z = z_k - \hat{z}$$

$$W_k = H_k P_k^- H_k^T + L_k R_k L_k^T$$

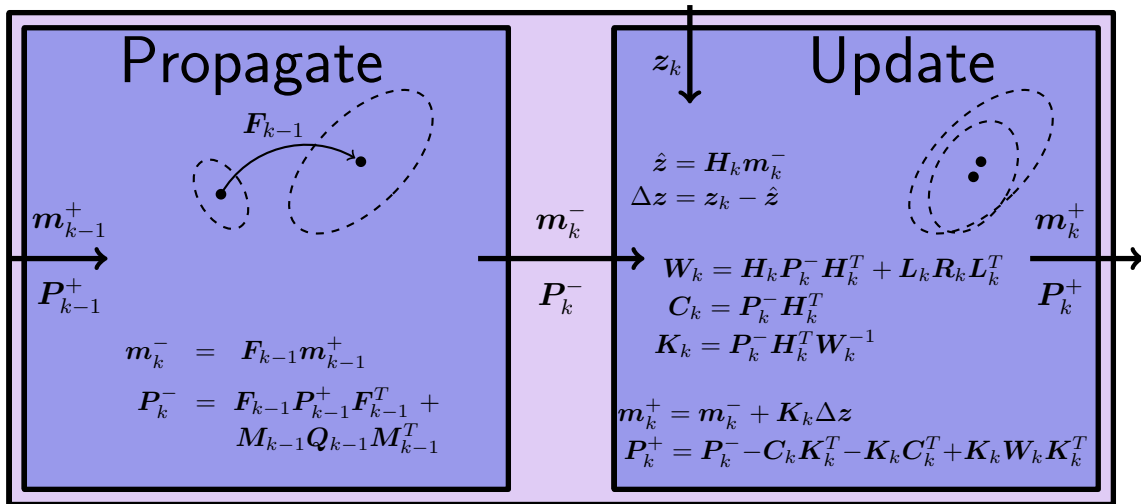
$$C_k = P_k^- H_k^T$$

$$K_k = P_k^- H_k^T W_k^{-1}$$

$$m_k^+ = m_k^- + K_k \Delta z$$

$$P_k^+ = P_k^- - C_k K_k^T - K_k C_k^T + K_k W_k K_k^T$$

Kalman Filter



The Kalman Gain

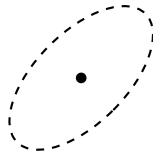
$$K_k = P_k^- H_k^T W_k^{-1}$$

Large Kalman Gain

Small Kalman Gain

Large Residual

Small Residual



$$\delta z = z_k - H_k m_k^-$$

The Kalman Gain

$$K_k = P_k^- H_k^T W_k^{-1}$$

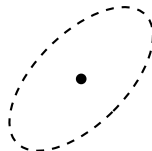
Large Kalman Gain

Small Kalman Gain

Large Residual

Small Residual

$$\delta z = z_k - H_k m_k^-$$



The Kalman Gain

$$K_k = P_k^- H_k^T W_k^{-1}$$

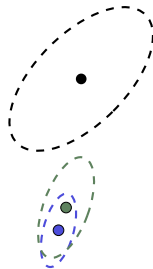
Large Kalman Gain

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Small Residual

$$\delta z = z_k - H_k m_k^-$$



The Kalman Gain

$$K_k = P_k^- H_k^T W_k^{-1}$$

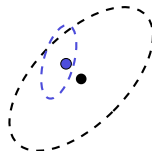
Large Kalman Gain

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$$\delta z = z_k - H_k m_k^-$$



The Kalman Gain

$$K_k = P_k^- H_k^T W_k^{-1}$$

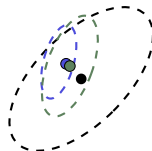
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$$\delta z = z_k - H_k m_k^-$$



The Kalman Gain

$$K_k = P_k^- H_k^T W_k^{-1}$$

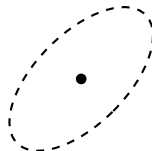
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Small Kalman Gain

Large Residual

Small Residual

$$\delta z = z_k - H_k m_k^-$$



The Kalman Gain

$$K_k = P_k^- H_k^T W_k^{-1}$$

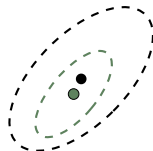
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Large Residual

Small Residual

$$\delta z = z_k - H_k m_k^-$$



The Kalman Gain

$$K_k = P_k^- H_k^T W_k^{-1}$$

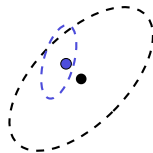
Large Kalman Gain

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Large Residual

Small Residual

$$\delta z = z_k - H_k m_k^-$$



The Kalman Gain

$$K_k = P_k^- H_k^T W_k^{-1}$$

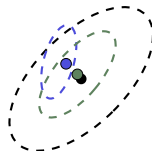
Large Kalman Gain

Small Kalman Gain

Large Residual

Small Residual

$$\delta z = z_k - H_k m_k^-$$



- Truth

$$\Delta x = \dot{x}_{k-1} \Delta t \quad \Delta \dot{x} = 0$$

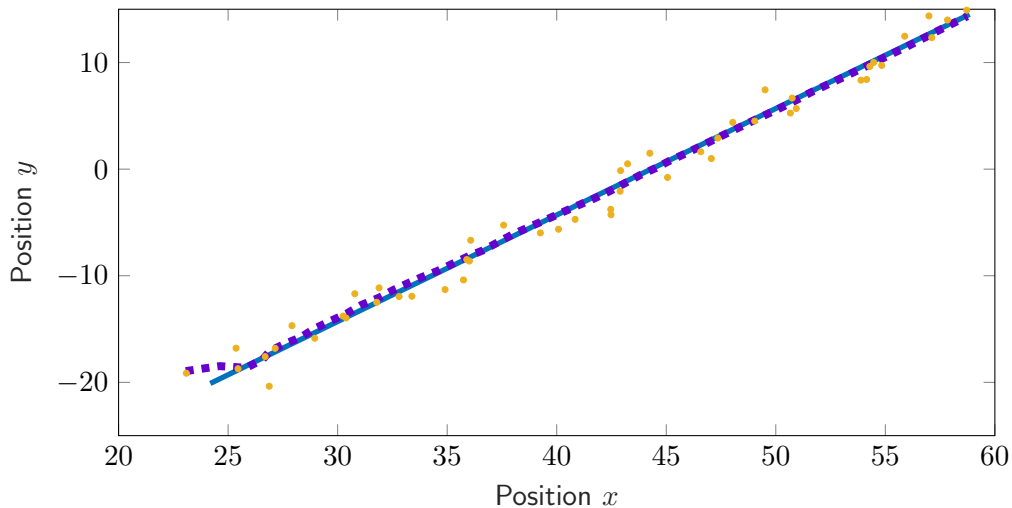
$$\Delta y = \dot{y}_{k-1} \Delta t \quad \Delta \dot{y} = 0$$

- Initial Estimate

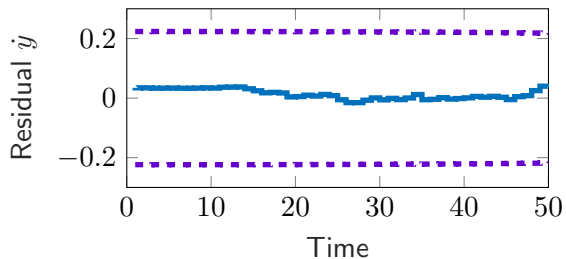
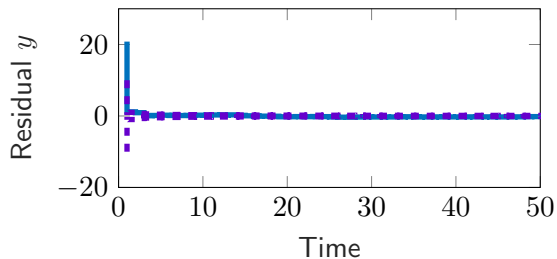
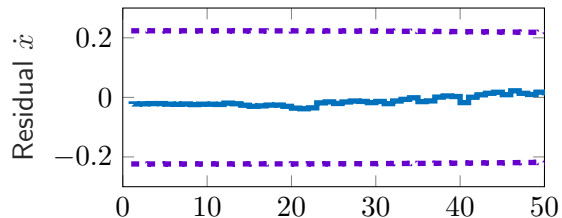
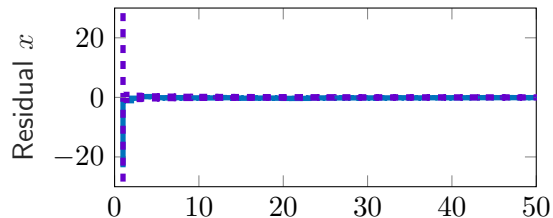
$$\mathbf{m}_0 = \begin{bmatrix} 0 \\ 0 \\ 100 \cos(\pi/4) \\ 100 \sin(\pi/4) \end{bmatrix} \quad \mathbf{P}_0 = \text{diag} \begin{bmatrix} 800 \\ 500 \\ 0.05 \\ 0.05 \end{bmatrix}$$

- No process error
- Directly measure x and y
- Measurement error covariance $\mathbf{R} = \mathbf{I}_2$

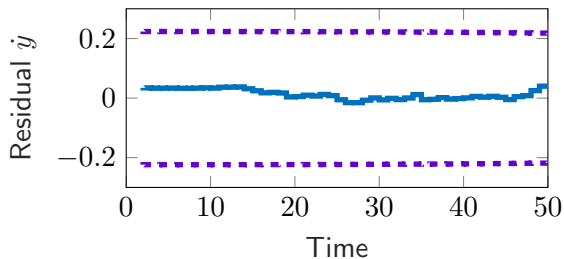
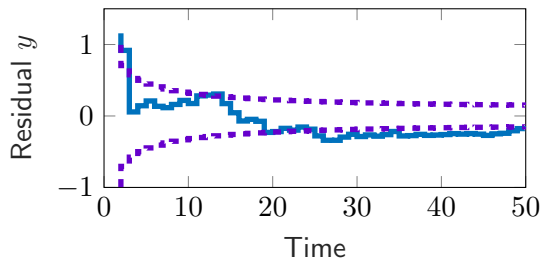
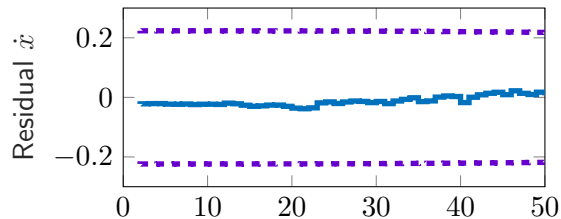
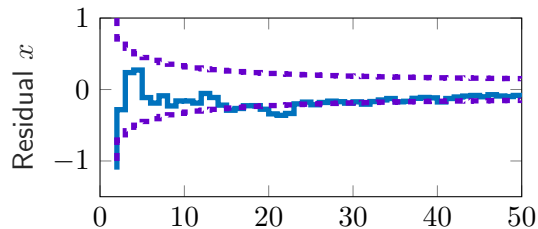
Raw Results



Residuals



Trimmed Residuals



Questions

