

Computer Vision Group Prof. Daniel Cremers



Visual Navigation for Flying Robots

Exercise 2 – Kalman Filter

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Submissions

- 8 submissions
- Evaluation
 - A few teams had minor (sign) errors
 - Including our solution ;-)
 - Correct solution is in this file
- Nice team pictures!

Team Pictures



Team Brezel



Team Dragonsheep



Team Crash Pilots



Team Red One



Team Roter Baron



Team Beer



Team Weissbier



Team Weisswurst

- Local coordinates
- Global coordinates

Robot is located somewhere in space



Robot is located somewhere in space

$$X = \begin{pmatrix} R & \mathbf{t} \\ \mathbf{0} & 1 \end{pmatrix} = \begin{pmatrix} \cos\psi & -\sin\psi & x \\ \sin\psi & \cos\psi & y \\ 0 & 0 & 1 \end{pmatrix} \in \operatorname{SE}(2) \subset \mathbb{R}^{3x3}$$

Robot is located at x=0.7, y=0.5, yaw=45deg

$$X = \begin{pmatrix} \cos 45 & -\sin 45 & 0.7 \\ \sin 45 & \cos 45 & 0.5 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0.71 & -0.71 & 0.7 \\ 0.71 & 0.71 & 0.5 \\ 0 & 0 & 1 \end{pmatrix}$$



Vector Transformation

- Robot is located at x=0.7, y=0.5, yaw=45deg
- Robot observes an object with its sensors
- Observation happens in the local frame



What does the robot see?

Robot observes marker in its local frame



What does the robot see?

Robot observes marker in its local frame



What does the robot see?

• Robot observes marker at $\mathbf{z} = (0.4m \ 0.0m \ -45^{\circ})^{\top}$



Where is the marker located in global coordinates?

- Robot observes marker at $\mathbf{z} = (0.4m \ 0.0m)^{\top}$
- Current (estimated) robot pose is

$$X = \begin{pmatrix} \cos 45 & -\sin 45 & 0.7 \\ \sin 45 & \cos 45 & 0.5 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0.71 & -0.71 & 0.7 \\ 0.71 & 0.71 & 0.5 \\ 0 & 0 & 1 \end{pmatrix}$$

Convert from local frame to global frame

$$\tilde{\mathbf{z}}_W = X\tilde{\mathbf{z}} = \begin{pmatrix} 1.27\\ 1.07\\ 1 \end{pmatrix}$$

How to deal with orientations?

- Assume we observe the marker at -45deg
- Mathematical convention:
 - Positive angles \rightarrow counter clock-wise
 - Negative angles \rightarrow clock-wise y_W y_W ψ yaw (heading) $\mathbf{z} = (0.4m \ 0.0m \ -45^\circ)^\top$

Transformation Matrix

Turn this into a transformation matrix

$$Z = \begin{pmatrix} \cos -45 & -\sin -45 & 0.4 \\ \sin -45 & \cos -45 & 0.0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0.71 & 0.71 & 0.4 \\ -0.71 & 0.71 & 0.0 \\ 0 & 0 & 1 \end{pmatrix}$$

Concatenate robot pose and marker obs.

$$Z_W = XZ = \begin{pmatrix} 1 & 0 & 1.27 \\ 0 & 1 & 1.07 \\ 0 & 0 & 1 \end{pmatrix}$$

Convert back to x/y/yaw

$$\mathbf{z}_W = (1.27 \ 1.07 \ 0)$$

Visual Navigation for Flying Robots

 For the Kalman filter, we need to define the observation function

$$h(\mathbf{x}) = \mathbf{z}_{\text{pred}}$$

which computes the expected marker pose z_{pred} given the current estimate of the state x (and given the fixed, global marker pose z_{global})

This means that we need to transform the global marker pose into the ego-centric (local) frame of the robot!

- We have: $\mathbf{x} = (x_x \ y_x \ \psi_x)^\top, \mathbf{z}_{\text{global}} = (x_g \ y_g \ \psi_g)^\top$
- We want: $\mathbf{z}_{\text{pred}} = (x_p \ y_p \ \psi_p)^{\top}$

Compute h(x) in two steps:

- **1.** Compute angle $\psi_p = \psi_g \psi_x$
- **2.** Compute translation x_p, y_p

Local to Global Transform

• Convert robot pose $\mathbf{x} = (x_x \ y_x \ \psi_x)^{\top}$ into a homogeneous transformation matrix

$$X = \begin{pmatrix} R & \mathbf{t} \\ \mathbf{0} & 1 \end{pmatrix} = \begin{pmatrix} \cos \psi_x & -\sin \psi_x & x_x \\ \sin \psi_x & \cos \psi_x & y_x \\ 0 & 0 & 1 \end{pmatrix}$$

 Right-multiplication with a vector transforms from local coordinates to global coordinates

$$\tilde{\mathbf{t}}_{\text{global}} = X \tilde{\mathbf{t}}_{\text{local}}$$

Global to Local Transform

■ We have the marker pose in global coordinates and we want to transform it into local coordinates → We need the inverse

$$X^{-1}\tilde{\mathbf{t}}_{\text{global}} = \tilde{\mathbf{t}}_{\text{local}}$$

There is an efficient way to compute the inverse of transformation matrices

$$\begin{pmatrix} R & \mathbf{t} \\ \mathbf{0}^{\top} & 1 \end{pmatrix}^{-1} = \begin{pmatrix} R^{\top} & -R^{\top}\mathbf{t} \\ \mathbf{0}^{\top} & 1 \end{pmatrix}$$

This gives

$$X^{-1} = \begin{pmatrix} R^{\top} & -R^{\top} \mathbf{t} \\ \mathbf{0} & 1 \end{pmatrix}$$
$$= \begin{pmatrix} \cos \psi_x & \sin \psi_x & -x_x \cos \psi_x - y_x \sin \psi_x \\ -\sin \psi_x & \cos \psi_x & x_x \sin \psi_x - y_x \cos \psi_x \\ 0 & 0 & 1 \end{pmatrix}$$

- We have the global marker position $\mathbf{t}_{\text{global}} = (x_g \ y_g)^{\top}$
- We want the marker position in the local frame

$$\begin{aligned} \tilde{\mathbf{t}}_{\text{local}} &= X^{-1} \tilde{\mathbf{t}}_{\text{global}} \\ &= \begin{pmatrix} \cos \psi_x & \sin \psi_x & -x_x \cos \psi_x - y_x \sin \psi_x \\ -\sin \psi_x & \cos \psi_x & x_x \sin \psi_x - y_x \cos \psi_x \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_g \\ y_g \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} (x_g - x_x) \cos \psi_x + (y_g - y_x) \sin \psi_x \\ -(x_g - x_x) \sin \psi_x + (y_g - y_x) \cos \psi_x \\ 1 \end{pmatrix} \end{aligned}$$

Finally, we get

$$h(\mathbf{x}) = \begin{pmatrix} (x_g - x_x)\cos\psi_x + (y_g - y_x)\sin\psi_x \\ -(x_g - x_x)\sin\psi_x + (y_g - y_x)\cos\psi_x \\ \psi_g - \psi_x \end{pmatrix}$$

Jacobian of the Observation Function

 Now derive the observation function with respect to all components of its argument

$$H = \frac{\partial h(\mathbf{x})}{\partial \mathbf{x}} = \begin{pmatrix} \frac{\partial h(\mathbf{x})}{\partial x_x} & \frac{\partial h(\mathbf{x})}{\partial y_x} & \frac{\partial h(\mathbf{x})}{\partial \psi_x} \end{pmatrix}$$
$$= \begin{pmatrix} -\cos\psi_x & -\sin\psi_x & -(x_g - x_x)\sin\psi_x + (y_g - y_x)\cos\psi_x \\ \sin\psi_x & -\cos\psi_x & -(x_g - x_x)\cos\psi_x - (y_g - y_x)\sin\psi_x \\ 0 & 0 & -1 \end{pmatrix}$$

That's it!

Extended Kalman Filter

- For each time step, do
- 1. Apply motion model

$$\bar{\mu}_t = g(\mu_{t-1}, u_t)$$

$$\bar{\Sigma}_t = G_t \Sigma G_t^\top + Q \quad \text{with } G_t = \frac{\partial g(\mu_{t-1}, u_t)}{\partial x_{t-1}}$$

2. Apply sensor model

$$\mu_t = \bar{\mu}_t + K_t(z_t - h(\bar{\mu}_t))$$

$$\Sigma_t = (I - K_t H_t) \bar{\Sigma}_t$$

with $K_t = \bar{\Sigma}_t H_t^{\top} (H_t \bar{\Sigma}_t H_t^{\top} + R)^{-1}$ and $H_t = \frac{\partial h(\bar{\mu}_t)}{\partial x_t}$

Sheet 2

- Fly the robot with a joystick
- Record own bag files
- You need WLAN
- Down-looking camera
- Markers on the floor, known positions
- Goal: Estimate robot position

Sheet 2

1. Fly the robot!

- Using a joystick
- Record your own bag file
- 2. Kalman Filter Prediction step
 - Will provide C++ framework
 - Run on recorded bag files (or online)
 - Change parameters and see what happens
- 3. Kalman Filter Correction step
 - Specify observation function and compute Jacobian
 - Implement it

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Sheet 3

- Markers on the floor, known positions
- Given: Full EKF code
- Goal: Implement P control
 - **1**. Try it on a recorded bag file and check steering commands in RVIZ
 - 2. Try it on real quadrocopter