



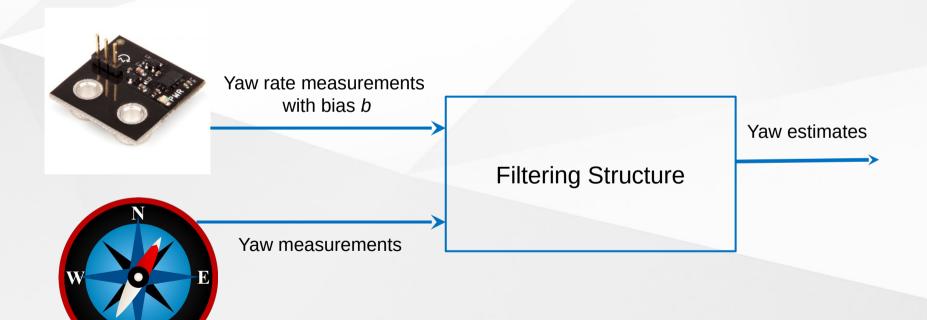


DSOR Group Meeting 27 November and 5 December 2020

Navigation: from concept to filter implementation (use-cases)

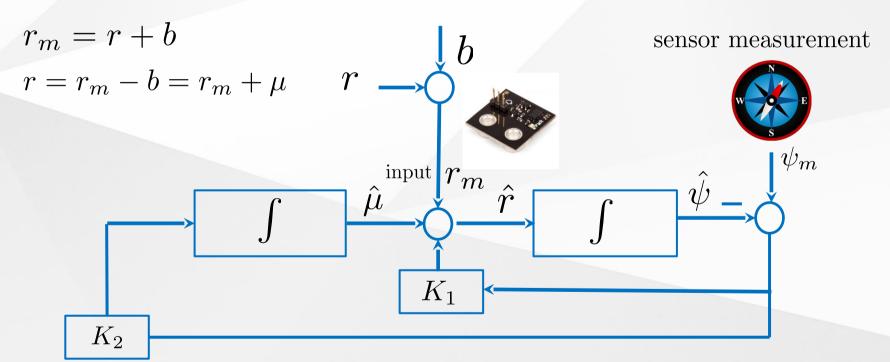


Yaw estimation from Rate Gyro and Compass Measurements





Complementary Filter Structure



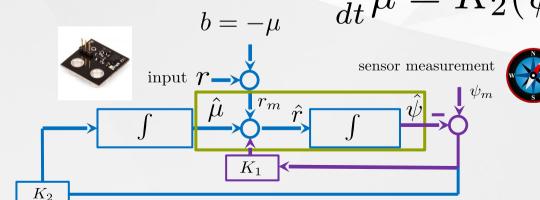


Complementary Filter Structure

Prediction

Correction

$$\frac{d}{dt}\hat{\psi} = \hat{\mu} + r_m + K_1(\psi_m - \hat{\psi})$$



 $\frac{d}{dt}\hat{\mu} = K_2(\psi_m - \hat{\psi})$

Underlying Design Model

$$\frac{d}{dt}\psi=r_m+\mu+\xi_1$$
 state noise $\frac{d}{dt}\mu=0+\xi_2$ measurement noise

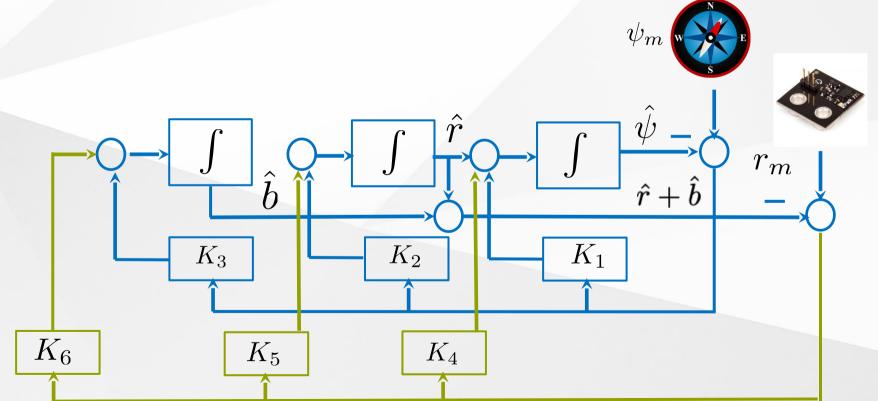


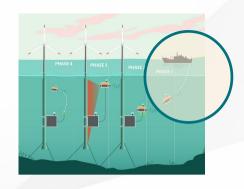
Example 1- a Sensor Fusion perspective

Underlying Design Model

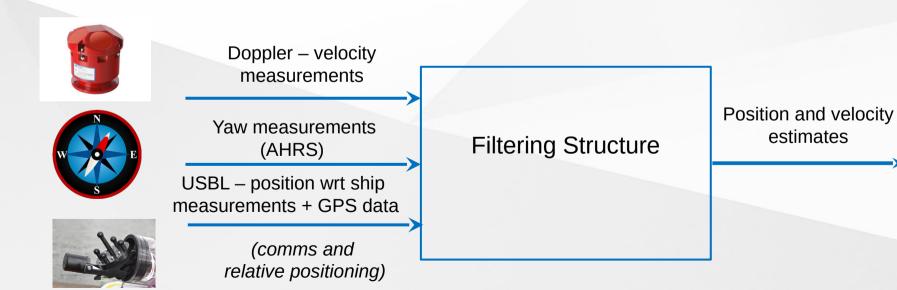


Example 1- a Sensor Fusion perspective





AUV-borne position estimation from Doppler, AHRS, inverted USBL, and GPS measurements (2D)



PHASE PHASE PHASE

 $p = (x, y)^T$

 $v = (\dot{x}, \dot{y})^T$

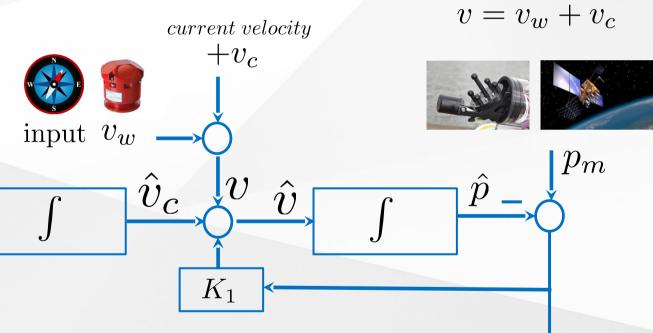
inertial velocity

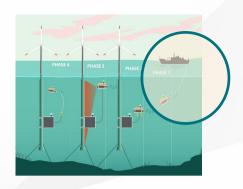
 K_2

inertial position

Example 2

Complementary Filter Structure $v_m = v_w = v - v_c$

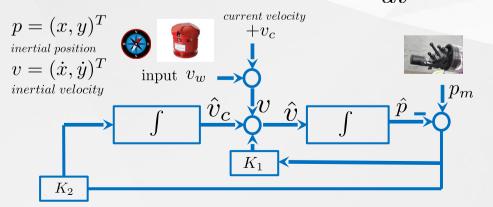




Complementary Filter Structure

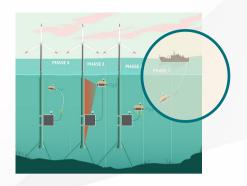
$$\frac{d}{dt}\hat{p} = \hat{v}_c + v_m + K_1(p_m - \hat{p})$$

$$\frac{d}{dt}\hat{v}_c = K_2(p_m - \hat{p})$$

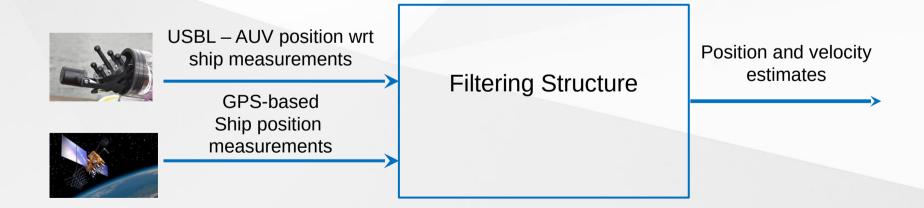


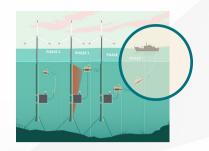
Underlying Design Model

$$\frac{d}{dt}p = v_c + v_w + \xi_1$$
 state noise $\frac{d}{dt}v_c = 0 + \xi_2$ measurement noise



Ship borne tracking of AUV position from USBL and GPS measurements (2D)





 $p = (x, y)^T$

 K_2

inertial position

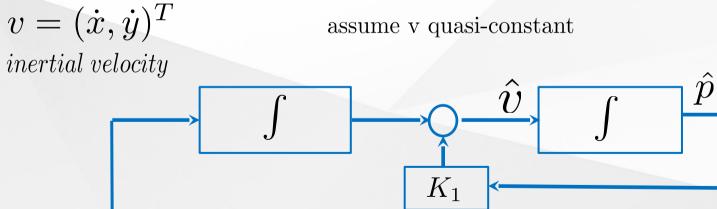
Example 3

 p_m obtained from USBL and GPS data

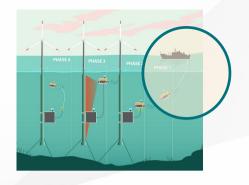




there is no direct or indirect measurement of v !



 p_m



$p = (x, y)^T$ inertial position

 $v = (\dot{x}, \dot{y})^T$

inertial velocity

there is no direct or indirect measurement of v!

Example 3

Underlying Design Model

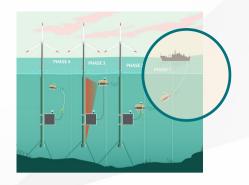
$$\frac{d}{dt}p = v + \xi_1 \leftarrow \underbrace{\frac{d}{dt}v} = 0 + \xi_2 \leftarrow \underbrace{\phantom{\frac{d}{dt}v}}$$

state noise

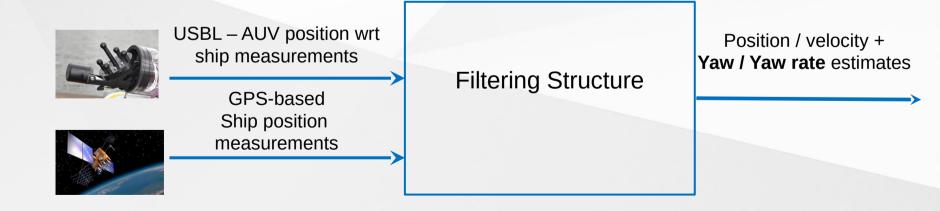
$$\frac{d}{dt}v = 0 + \xi_2$$
 \longleftarrow

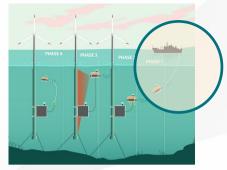
$$p_m = p + \eta$$
 — Measurement noise

 p_m obtained from USBL and GPS data



Ship borne tracking of AUV position from USBL and GPS measurements (2D)





$p = (x, y)^T$

inertial position

$$\psi - yaw \ angle$$

r-yaw rate

 $u-surge\ speed$

Example 4

Underlying Design Model (small sideslip angle) *

$$\frac{d}{dt}u = 0 + \xi_u \quad \longleftarrow \quad \frac{d}{dt}\psi = r + \xi_\psi \quad \longleftarrow \quad \text{st}$$

$$\frac{d}{dt}\psi = r + \xi_{\psi}$$
 state nois

$$\frac{d}{dt}r = 0 + \xi_r \longleftarrow$$

$$p_m = p + \eta$$
 — Me

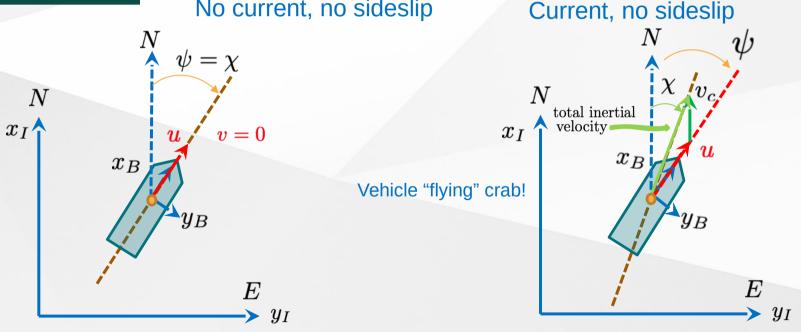
Measurement noise

^{*} A rigorous interpretation is to view u as the absolute value of the total velocity vector and ψ as the course angle

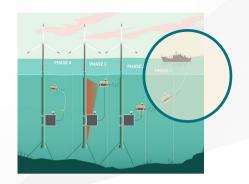
A word about notation

 ψ - yaw angle χ - course angle v_c - velocity of current





Vehicle along straight line



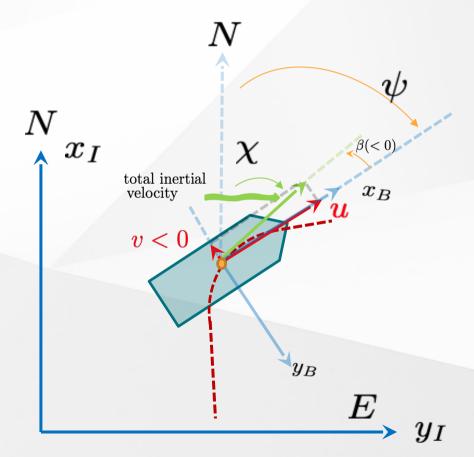
No current, with sideslip - vehicle moving along a curved line

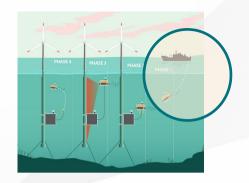
 ψ – yaw angle

 $\beta-sideslip\ angle$

 χ – course angle

A word about notation





Current, with sideslip

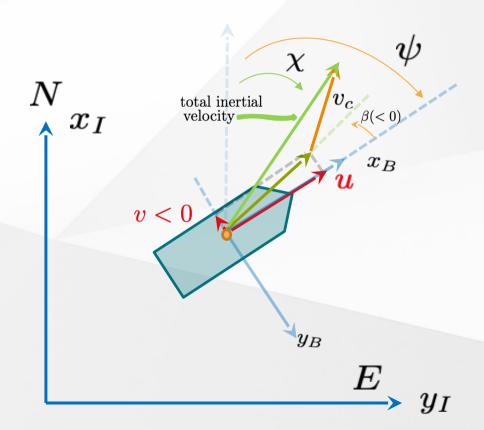
 ψ – yaw angle

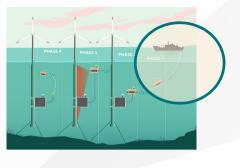
 $\beta-sideslip\ angle$

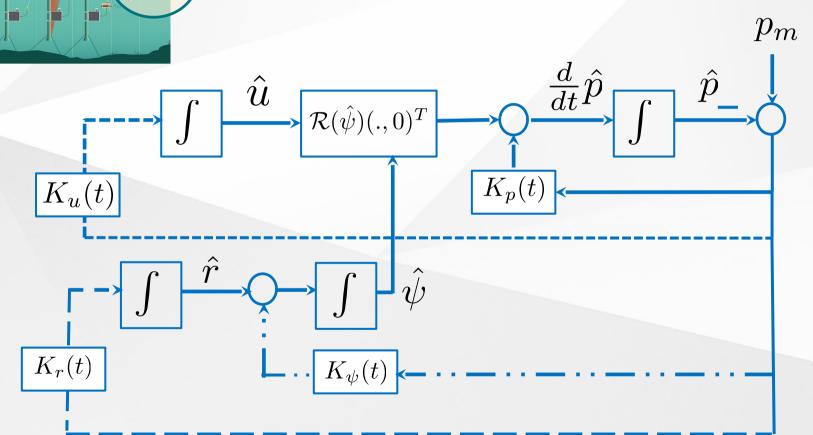
 χ – course angle

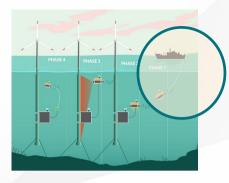
 v_c - velocity of current

A word about notation









AUV-borne position estimation from Doppler, AHRS, and measurements of ranges-only to known beacon (e.g. on docking station)





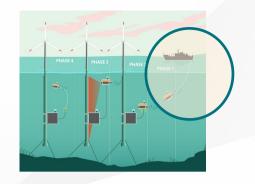


Doppler – velocity measurements

Yaw measurements (AHRS)

Range measuring device – measurements of range wrt fixed beacon Filtering Structure

Position and velocity estimates



$p = (x, y)^T$

inertial position

$$v = (\dot{x}, \dot{y})^T$$

inertial velocity

 $p_b - beacon\ position$

 r_m - range measurement

Example 5

Complementary Filter Structure

Underlying Design Model

current velocity

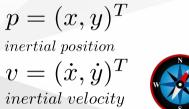
Doppler measurement
$$v_m = v_w = v - v_c$$

$$v = v_w + v_c$$

$$\frac{d}{dt}p = v_c + v_w + \xi_1 \leftarrow$$
 state noise
$$\frac{d}{dt}v_c = 0 + \xi_2 \leftarrow$$

$$r_m = ||p - p_b|| + \eta \leftarrow$$
 measurement noise

Complementary Filter Structure $v_m = v_w = v - v_c$ $v = v_w + v_c$







input v_w

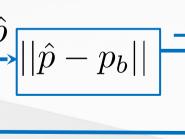




current velocity

 $+v_c$





 $K_2(t)$



$$\tilde{y} = z_k - C_k \hat{x}_{k|k-1}$$

$$S_k = C_k P_{k|k-1} C_k^T + R_k$$

$$K_k = P_{k|k-1} C_k^T S_k^{-1}$$

$$\hat{x}_{k|k} = \hat{x}_{k|k-1} + K_k \tilde{y}_k$$

$$P_{k|k-1} = (1 - k_k C_k) P_{k|k-1}$$

where

- $\tilde{y} = \text{Innovation Factor Vector}$
- $z_k = \text{Measurement Vector}$
- C_k = Observation Matrix
- R_k = Covariance of the Measurement Noise
- $S_k = \text{Innovation Matrix}$
- $K_k = \text{Kalman Gain Matrix}$

$$\begin{split} \hat{x}_{k|k-1} &= A_k \hat{x}_{k-1|k-1} + B_k u_k + w_k \\ P_{k|k-1} &= A_k P_{k-1|k-1} A_k^T + Q_k \end{split}$$

where

- A_k = State Transition Matrix
- B_k = Control-Input Model
- $u_k = \text{Input Vector}$
- $w_k = \text{Zero-mean Gaussian Process Noise}$
- Q_l = Covariance of the Process Noise

$$x = \begin{bmatrix} x \\ y \\ \dot{x} \\ \dot{y} \\ \dot{x} \\ \dot{y} \\ \dot{x} \\ \dot{x} \\ \dot{x} \\ \dot{x} \\ \dot{x} \end{bmatrix} \quad A_k = \begin{bmatrix} 1 & 0 & \delta t & 0 & 0.5\delta t^2 & 0 & 0 & 0 \\ 0 & 1 & 0 & \delta t & 0 & 0.5\delta t^2 & 0 & 0 \\ 0 & 0 & 1 & 0 & \delta t & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & \delta t & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & \delta t & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \quad B_k = [0]_{8 \times 8}$$

where

• $\delta t = \text{predict_period}$

Example 1: Yaw estimation from Rate Gyro and Compass

Measurements

Kalman Filter with Yaw and Yaw Rate as measurement vector

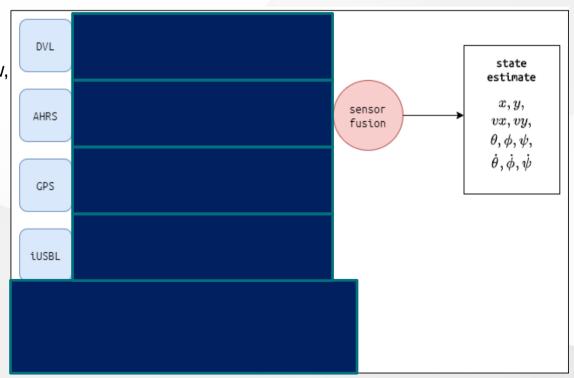
$$\begin{split} \tilde{y} &= z_k - C_k \hat{x}_{k|k-1} \\ S_k &= C_k P_{k|k-1} C_k^T + R_k \\ K_k &= P_{k|k-1} C_k^T S_k^{-1} \\ \hat{x}_{k|k} &= \hat{x}_{k|k-1} + K_k \tilde{y}_k \\ P_{k|k-1} &= (1 - k_k C_k) P_{k|k-1} \end{split}$$

where

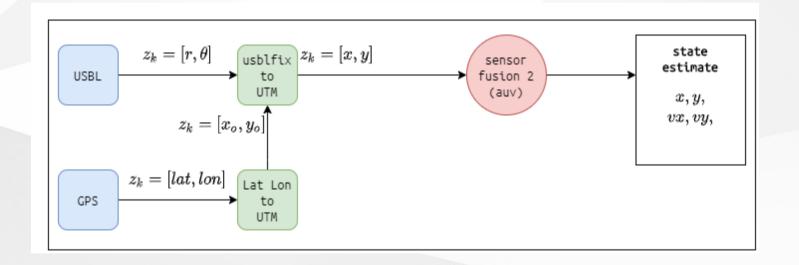
- $\tilde{y} = \text{Innovation Factor Vector}$
- $z_k = \text{Measurement Vector}$
- $C_k = \text{Observation Matrix}$
- R_k = Covariance of the Measurement Noise
- $S_k = \text{Innovation Matrix}$
- $K_k = \text{Kalman Gain Matrix}$

Example 2: AUV-borne position estimation from Doppler, AHRS, inverted USBL, and GPS measurements (2D)

- Doppler updates velocity
- AHRS updates roll, pitch, yaw, roll rate, pitch rate, yaw rate
- Inverted USBL updates position
- GPS updates position (at surface)

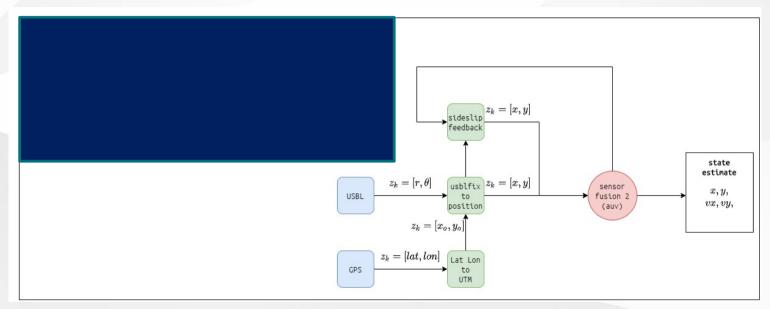


Example 3: Ship borne tracking of AUV position from USBL and GPS measurements (2D)



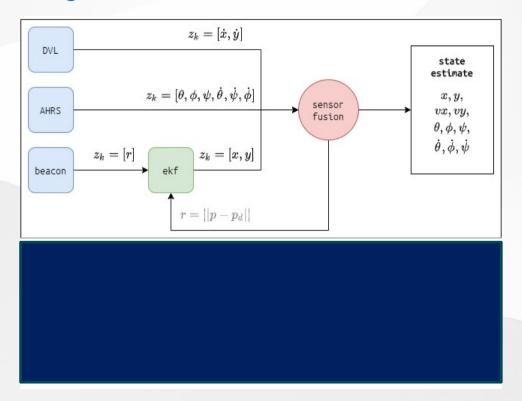
 Absolute Position of the Ship (GPS) + Relative Position of AUV (USBL) = Absolute position of the AUV

Example 4: Ship borne tracking of AUV position from USBL and GPS measurements (2D)



- Relative Position of AUV (USBL) + Absolute Position of the Ship (GPS) = Absolute position of the AUV
- Sideslip angle feedback provides position updates using "predict" stage

Example 5: AUV-borne position estimation from Doppler, AHRS, and measurements of ranges-only to known beacon (e.g. on docking station)



• Ekf to incorporate range as position updates







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