

Supplementary Material for “A Probabilistic Approach for Predicting Vessel Motion”

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A. Units and range of measurements

TABLE I: Measurements

Measurement	Range	Unit
ϕ	$[-90^\circ, 90^\circ]$	decimal/DMS
λ	$[-180^\circ, 180^\circ]$	decimal/DMS
ψ	$[0^\circ, 359.9^\circ]$	degree
U	—	knot
χ	$[0^\circ, 359.9^\circ]$	degree

B. Parameters in Equations (8)—(11)

TABLE II: Parameters

Parameter	Value	Parameter	Value
a	6378137	d_1	$10^{-1}/\sqrt{60}$
e	0.08181919	d_2	$10^{-3}/\sqrt{60}$
h	0	K_p	1
T	107.3	K_d	20
K	0.185	Δ	500

C. Prediction metrics for each measurement

TABLE III: Comparative Experiment Results

Measurement	Proposed method			EKF		
	GE	GER	EAPI	GE	GER	EAPI
2	322.65	24.97	130.26	370.54	28.68	162.16
3	101.26	4.53	40.79	144.70	6.47	95.68
4	665.07	18.60	41.65	2278.62	63.73	183.51
Average	362.99	16.03	70.90	931.28	32.96	147.11

D. Approximation algorithm for Equations (4)—(7)

Algorithm 1 Gaussian Assumed Density Approximation for Sequential Bayesian Inference

Input: The prior distribution $p(\mathbf{x}_N | \mathbf{y}_{1:(N-1)}) = \mathcal{N}(\mathbf{m}_N^-, \mathbf{P}_N^-)$.

Procedure:

- 1) Calculate the mean and covariance of the posterior distribution $p(\mathbf{z}_N | \mathbf{y}_{1:N}) = \mathcal{N}(\mathbf{m}_N^*, \mathbf{P}_N^*)$:

$$\begin{aligned}
 \mathbf{Q}_N &= \mathbf{H}_x(\mathbf{m}_N^-) \mathbf{P}_N^- \mathbf{H}_x^T(\mathbf{m}_N^-) + \mathbf{R}_N, \\
 \mathbf{K}_N &= \mathbf{P}_N^- \mathbf{H}_x^T(\mathbf{m}_N^-) \mathbf{Q}_N^{-1}, \\
 \mathbf{m}_N &= \mathbf{m}_N^- + \mathbf{K}_N (\mathbf{y}_N - \mathbf{h}(\mathbf{m}_N^-)), \\
 \mathbf{P}_N &= \mathbf{P}_N^- - \mathbf{K}_N \mathbf{Q}_N \mathbf{K}_N^T, \\
 \mathbf{m}_N^* &= \begin{bmatrix} \mathbf{m}_N \\ \mathbf{m}_s \end{bmatrix}, \mathbf{P}_N^* = \begin{bmatrix} \mathbf{P}_N & \mathbf{0} \\ \mathbf{0} & \mathbf{P}_s \end{bmatrix}.
 \end{aligned} \tag{1}$$

- 2) Integrate the following equations up to time t_f :

$$\begin{aligned}
 \frac{d\mathbf{m}}{dt} &= \mathbf{g}(\mathbf{m}), \\
 \frac{d\mathbf{P}}{dt} &= \mathbf{P} \mathbf{G}_z^T(\mathbf{m}) + \mathbf{G}_z(\mathbf{m}) \mathbf{P} + \mathbf{L}(\mathbf{m}) \mathbf{L}^T(\mathbf{m}).
 \end{aligned} \tag{2}$$

where $\mathbf{G}_z(\mathbf{z})$ and $\mathbf{H}_x(\mathbf{x})$ represent the Jacobian matrices of $\mathbf{g}(\mathbf{z})$ and $\mathbf{h}(\mathbf{x})$ with respect to \mathbf{z} and \mathbf{x} , respectively. The initial conditions for (2) are \mathbf{m}_N^* and \mathbf{P}_N^* , respectively. The corresponding solutions are given by:

$$\mathbf{m}_f^* = \begin{bmatrix} \mathbf{m}_f^- \\ \mathbf{m}_s \end{bmatrix}, \mathbf{P}_f^* = \begin{bmatrix} \mathbf{P}_f^- & * \\ * & \mathbf{P}_s \end{bmatrix}. \tag{3}$$

Output: The prior distribution $p(\mathbf{x}_f | \mathbf{y}_{1:N}) = \mathcal{N}(\mathbf{m}_f^-, \mathbf{P}_f^-)$.