

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

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 |

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 $p(\mathbf{z}_N | \mathbf{y}_{1:N}) = \mathcal{N}(\mathbf{m}_N^*, \mathbf{P}_N^*)$:

$$\mathbf{Q}_N = \mathbf{H}_{\mathbf{x}}(\mathbf{m}_N^-) \mathbf{P}_N^- \mathbf{H}_{\mathbf{x}}^T(\mathbf{m}_N^-) + \mathbf{R}_N,$$

$$\mathbf{K}_N = \mathbf{P}_N^- \mathbf{H}_{\mathbf{x}}^T(\mathbf{m}_N^-) \mathbf{Q}_N^{-1},$$

$$\mathbf{m}_N = \mathbf{m}_N^- + \mathbf{K}_N (\mathbf{y}_N - \mathbf{h}(\mathbf{m}_N^-)),$$

$$\mathbf{P}_N = \mathbf{P}_N^- - \mathbf{K}_N \mathbf{Q}_N \mathbf{K}_N^T, \quad (1)$$

$$\mathbf{m}_N^* = \begin{bmatrix} \mathbf{m}_N \\ \mathbf{m}_s \end{bmatrix}, \mathbf{P}_N^* = \begin{bmatrix} \mathbf{P}_N & \mathbf{0} \\ \mathbf{0} & \mathbf{P}_s \end{bmatrix}.$$

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

$$\frac{d\mathbf{m}}{dt} = \mathbf{g}(\mathbf{m}), \quad (2)$$

$$\frac{d\mathbf{P}}{dt} = \mathbf{P} \mathbf{G}_{\mathbf{z}}^T(\mathbf{m}) + \mathbf{G}_{\mathbf{z}}(\mathbf{m}) \mathbf{P} + \mathbf{L}(\mathbf{m}) \mathbf{L}^T(\mathbf{m}).$$

where $\mathbf{G}_{\mathbf{z}}(\mathbf{z})$ and $\mathbf{H}_{\mathbf{x}}(\mathbf{x})$ 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:

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

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