



# Log likelihood

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# **JIS**

Input estimation: 
$$\overline{R}_k = GP_{k|k-1}^x G^T + R$$

$$M = (I^T \overline{R}^{-1} I)^{-1} I^T \overline{R}$$

$$\boldsymbol{M}_{k} = (\boldsymbol{J}^{\mathrm{T}} \boldsymbol{\overline{R}}_{k}^{-1} \boldsymbol{J})^{-1} \boldsymbol{J}^{\mathrm{T}} \boldsymbol{\overline{R}}_{k}^{-1}$$

$$\hat{p}_{k|k} = M_k (y_k - G\hat{x}_{k|k-1})$$

$$P_{k|k}^p = (J^{\mathrm{T}} \overline{R}_k^{-1} J)^{-1}$$

Measurement update: 
$$K_k = P_{k|k-1}^x G^T \overline{R}_k^{-1}$$

$$\hat{x}_{k|k} = \hat{x}_{k|k-1} + K_k (y_k - G\hat{x}_{k|k-1} - J\hat{p}_k)$$

$$P_{k|k}^{x} = P_{k|k-1}^{x} - K_{k}(\overline{R}_{k} - JP_{k|k}^{p}J^{T})K_{k}^{T}$$

$$P_{k|k}^{xp} = -K_k J P_{k|k}^p$$

Time update: 
$$\overline{S}_k = AP_{x,k|k-1}G^T + S$$

$$\overline{K}_{k} = \overline{S}_{k} \overline{R}_{k}^{-1} + (B - \overline{S}_{k} \overline{R}_{k}^{-1} J) M_{k}$$

$$\hat{x}_{k+1|k} = A\hat{x}_{k|k-1} + \overline{K}_k (y_k - G\hat{x}_{k|k-1})$$

$$P_{k+1|k}^{x} = AP_{k|k-1}^{x}A^{\mathrm{T}} + Q - \overline{S}_{k}\overline{K}_{k}^{\mathrm{T}} - \overline{K}_{k}\overline{S}_{k}^{T} + \overline{K}_{k}\overline{R}_{k}\overline{K}_{k}^{\mathrm{T}}$$

# log(L)

#### Joint input and state estimation:

$$\begin{split} \hat{p}_{k|k} &= M_k (y_k - G\hat{x}_{k|k-1}) \\ \hat{x}_{k|k} &= \hat{x}_{k|k-1} + K_k (y_k - G\hat{x}_{k|k-1} - J\hat{p}_{k|k}) \\ \hat{x}_{k+1|k} &= A\hat{x}_{k|k-1} + \overline{K}_k (y_k - G\hat{x}_{k|k-1}) \end{split}$$

# $\Omega_k = \begin{bmatrix} G & J \end{bmatrix} \operatorname{cov} \left[ \begin{bmatrix} x_k - \hat{x}_{k|k-1} \\ p_k - \hat{p}_{k|k} \end{bmatrix} \right] \begin{bmatrix} G & J \end{bmatrix}^{\mathrm{T}} + R$

#### Residual sequence:

$$egin{aligned} \hat{m{i}}_{k} &= y_{k} - G\hat{m{x}}_{k|k-1} - J\hat{m{p}}_{k|k} = y_{k} - egin{bmatrix} G & J \end{bmatrix} egin{bmatrix} \hat{m{x}}_{k|k-1} \ \hat{m{p}}_{k|k} \end{bmatrix} = egin{bmatrix} G & J & I \end{bmatrix} egin{bmatrix} x_{k} - x_{k|k-1} \ p_{k} - \hat{m{p}}_{k|k} \ v_{k} \end{bmatrix} \end{aligned}$$

#### Log likelihood:

$$\log(L) = \sum_{k} \log(|\Omega_{k}|) + i_{k}^{\mathsf{T}} \Omega_{k}^{-1} i_{k}$$

$$\Omega_k = \begin{bmatrix} G & J & I \end{bmatrix} \operatorname{cov} \begin{bmatrix} x_k - \hat{x}_{k|k-1} \\ p_k - \hat{p}_{k|k} \\ v_k \end{bmatrix} \begin{bmatrix} G & J & I \end{bmatrix}^{\mathrm{T}}$$

$$\operatorname{cov}\!\left(\!\!\left[\!\!\begin{array}{c} x_{k} - \hat{x}_{k|k-1} \\ p_{k} - \hat{p}_{k|k} \\ v_{k} \end{array}\!\!\right]\!\!\right) = \operatorname{E}\!\left[\!\!\left[\!\!\begin{array}{cccc} \tilde{x}_{k|k-1} \tilde{x}_{k|k-1}^{\mathrm{T}} & \tilde{x}_{k|k-1} \tilde{p}_{k|k}^{\mathrm{T}} & \tilde{x}_{k|k-1} v_{k}^{\mathrm{T}} \\ \tilde{p}_{k|k} \tilde{x}_{k|k-1}^{\mathrm{T}} & \tilde{p}_{k|k} \tilde{p}_{k|k}^{\mathrm{T}} & \tilde{p}_{k|k} v_{k}^{\mathrm{T}} \\ v_{k} \tilde{x}_{k|k-1}^{\mathrm{T}} & v_{k} \tilde{p}_{k|k}^{\mathrm{T}} & v_{k} v_{k}^{\mathrm{T}} \end{array}\!\!\right] = \!\!\left[\!\!\begin{array}{c} P_{k|k-1}^{x} & E[\tilde{x}_{k|k-1} \tilde{p}_{k|k}^{\mathrm{T}}] & 0 \\ E[\tilde{p}_{k|k} \tilde{x}_{k|k-1}^{\mathrm{T}}] & P_{k|k-1}^{p} & E[\tilde{p}_{k|k} v_{k}^{\mathrm{T}}] \\ 0 & E[v_{k} \tilde{p}_{k|k}^{\mathrm{T}}] & R \end{array}\!\!\right]$$

Two cross-covariance terms need be derived

## **Derivation covariance**

Rewrite measurement update for state:

$$\hat{x}_{k|k} = \hat{x}_{k|k-1} + K_k (y_k - G\hat{x}_{k|k-1}) - \hat{x}_{k|k-1} + K_k (y_k - G\hat{x}_{k|k-1}) - K_k J \hat{p}_k = \hat{x}_{k|k-1} + K_k (y_k - G\hat{x}_{k|k-1}) - K_k J M_k (y_k -$$

Derive a relation between the state prediction error and state filter error:

$$\begin{split} \tilde{x}_{k|k} &= x_k - \hat{x}_{k|k} = x_k - \hat{x}_{k|k-1} - K_k (I - JM_k) (Gx_k + Jp_k + v_k - G\hat{x}_{k|k-1}) \\ \tilde{x}_{k|k} &= \tilde{x}_{k|k-1} - K_k (I - JM_k) G\tilde{x}_{k|k-1} - K_k (I - JM_k) Gv_k - K_k (I - JM_k) Jp_k \end{split}$$

The matrix term in front of the force above is zero (Gillijns)

We then rearrange:

$$\begin{split} \tilde{x}_{k|k} &= (I - K_k (I - JM_k)G) \tilde{x}_{k|k-1} - K_k (I - JM_k) v_k \\ \tilde{x}_{k|k} &+ K_k (I - JM_k) v_k = (I - K_k (I - JM_k)G) \tilde{x}_{k|k-1} \end{split}$$

## **Derivation covariance**

Multiply the equation by the input error transposed:

$$\tilde{x}_{k|k} \tilde{p}_{k|k}^{T} + K_{k} (I - JM_{k}) v_{k} \tilde{p}_{k|k}^{T} = (I - K_{k} (I - JM_{k})G) \tilde{x}_{k|k-1} \tilde{p}_{k|k}^{T}$$

We use that (Gillijns):

$$\tilde{p}_{k|k} = -M_k (G\tilde{x}_{k|k-1} + v_k)$$

We find the first term of interest:

$$E\left[v_{k}\tilde{p}_{k|k}^{\mathrm{T}}\right] = -RM_{k}^{\mathrm{T}}$$

And the second term:

$$E\left[\tilde{x}_{k|k-1}\tilde{p}_{k|k}^{T}\right] = (I - K_{k}(I - JM_{k})G)^{-1} \left(P_{k|k}^{xp} - K_{k}(I - JM_{k})RM_{k}^{T}\right)$$

### Not used

#### Next we define:

$$\tilde{y}_k = y_k - G\hat{x}_{k|k-1} = Gx_k + Jp_k + v_k - G\hat{x}_{k|k-1} = G\tilde{x}_{k|k-1} + v_k + Jp_k = e_k + Jp_k, \quad e_k = G\tilde{x}_{k|k-1} + v_k$$