

Q1) We need to prove that the Kalman Filter is a MSME estimator.

→ This essentially means that we need to prove that the Kalman Gain minimises the expectation of the squared error $E[(x - \hat{x})^2]$

$$\Rightarrow E[(x - \hat{x})^2] = \text{cov}[x_t - \hat{x}_{t|t}]$$

$$\Rightarrow \text{cov}[x_t - \hat{x}_{t|t-1} - K_t(x_t - C\hat{x}_{t|t-1})]$$

$$\Rightarrow \text{cov}[(I - K_t C_t)(x_t - \hat{x}_{t|t-1}) - K_t \eta_t]$$

$$\Rightarrow (I - K_t C_t) P_{t|t-1} (I - K_t C_t)^T + K_t R_t K_t^T$$

→ Minimising $E[(x - \hat{x})^2]$ by differentiating w.r.t K_t :

$$\Rightarrow \frac{\partial E[(x - \hat{x})^2]}{\partial K_t} = \frac{\partial \text{tr}(E[(x - \hat{x})^2])}{\partial K_t} = 0$$

$$\Rightarrow \text{tr}\left(\frac{\partial E[(x - \hat{x})^2]}{\partial K_t}\right) = 0$$

Upon deriving, we finally get:

$$\Rightarrow -2P_{t|t-1}^T C_t^T + 2K_t S_t = 0$$

$$\Rightarrow K_t S_t = P_{t|t-1}^T C_t^T$$

$$[K_t = P_{t|t-1}^T C_t^T S_t^{-1}]$$

* This is the gain of the Kalman filter. Hence, we can say that Kalman filter is a MSME estimator.

Q2) Polar to rectangular coordinates:

$$x = r \cos \theta$$

$$y = r \sin \theta$$

$$f = h(r, \theta)$$

$$r = \hat{r} + r_e$$

$$\theta = \hat{\theta} + \theta_e$$

$$\left(\bar{r} = 1m, \quad \bar{\theta} = \frac{\pi}{2} \right) \rightarrow p(r) = N(\bar{r}, \sigma_r^2)$$

$$p(\theta) = N(\bar{\theta}, \sigma_\theta^2)$$

This function is written as:

$$f = h(s)$$

$$\Rightarrow \begin{bmatrix} x \\ y \end{bmatrix} = r \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix} = \begin{bmatrix} r \cos \theta \\ r \sin \theta \end{bmatrix}$$

(i) Mean & covariance of f :

$$\mu = \begin{bmatrix} \hat{x} \\ \hat{y} \end{bmatrix} = E \left[\begin{bmatrix} x \\ y \end{bmatrix} \right] = \begin{bmatrix} E(x) \\ E(y) \end{bmatrix}$$

$$\mu = \begin{bmatrix} E(r \cos \theta) \\ E(r \sin \theta) \end{bmatrix} = \begin{bmatrix} E(r) \times E(\cos \theta) \\ E(r) \times E(\sin \theta) \end{bmatrix}$$

$$= \begin{bmatrix} \bar{r} e^{-\frac{\sigma_\theta^2}{2}} \cos \bar{\theta} \\ \bar{r} e^{-\frac{\sigma_\theta^2}{2}} \sin \bar{\theta} \end{bmatrix}$$

$$= \begin{bmatrix} \mu_x \\ \mu_y \end{bmatrix}$$

Substitute $\bar{r} = 1, \bar{\theta} = \frac{\pi}{2}$

$$\mu = \begin{bmatrix} 0 \\ e^{-\frac{\sigma_\theta^2}{2}} \end{bmatrix}$$

Covariance

$$\Rightarrow \Sigma = E[(\mathbf{f} - \mu)(\mathbf{f} - \mu)^T]$$

$$= E \left[\begin{bmatrix} x - \mu_x \\ y - \mu_y \end{bmatrix} \begin{bmatrix} x - \mu_x & y - \mu_y \end{bmatrix} \right]$$

$$= E \left[\begin{bmatrix} (x - \mu_x)^2 & (x - \mu_x)(y - \mu_y) \\ (x - \mu_x)(y - \mu_y) & (y - \mu_y)^2 \end{bmatrix} \right]$$

$$\text{Now } \Sigma = \begin{bmatrix} E[(x - \mu_x)^2] & E[(x - \mu_x)(y - \mu_y)] \\ E[(x - \mu_x)(y - \mu_y)] & E[(y - \mu_y)^2] \end{bmatrix}$$

Now, computing each element individually:

$$\Rightarrow E(x - \mu_x)^2 = E(x^2 + \mu_x^2 - 2x\mu_x)$$

$$= E(x^2) + E[\mu_x^2] - 2\mu_x E[x]$$

$$= E[r^2 \cos^2 \theta] + \mu_x^2 - 2\mu_x^2$$

$$= E[r^2] E[\cos^2 \theta] + \mu_x^2 - 2\mu_x^2$$

$$\cdot E[r^2] = E[(r + \mu - \mu)^2]$$

$$= E[(r + \mu)^2 + \mu^2 + 2(r + \mu)\mu]$$

$$= E[(r + \mu)^2 + \mu^2 - 2\mu^2 + 2r\mu]$$

$$= E[(r - \mu)^2] + E[\mu^2]$$

$$= \sigma_r^2 + \mu^2$$

$$= \sigma_r^2 + \bar{r}^2$$

$$\Rightarrow E(x - \mu_x)^2 = (\sigma_r^2 + \bar{r}^2) E\left(\frac{\cos 2\theta + 1}{2}\right) - \mu_x^2$$

$$= \frac{1}{2} (\sigma_r^2 + \bar{r}^2) (1 + \exp(-2\sigma_r^2) \cos 2\theta) - \mu_x^2$$

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Substituting values,

$$\bar{r} = 1, \bar{\theta} = \frac{\pi}{2}$$

$$\Rightarrow \frac{1}{2} (\sigma_r^2 + 1) (1 - e^{-2\sigma_\theta^2})$$

$$\Rightarrow E[(x - \mu_x)(y - \mu_y)] = E[xy - x\mu_y - y\mu_x + \mu_x\mu_y]$$

$$\Rightarrow E(xy) - E[x\mu_y] - E[y\mu_x] + E[\mu_x\mu_y]$$

$$\Rightarrow E[r^2] \cdot E[\sin\theta\cos\theta] - \mu_x\mu_y$$

$$\Rightarrow (\bar{r}^2 + \sigma_r^2) E\left[\frac{\sin(2\theta)}{2}\right] - \mu_x\mu_y$$

$$\Rightarrow \frac{1}{2} (\sigma_r^2 + \bar{r}^2) e^{-2\sigma_\theta^2} \sin(2\theta) - \mu_x\mu_y$$

Substituting values,

$$E[(x - \mu_x)(y - \mu_y)] = 0$$

Now, $E[y - \mu_y]^2 = E[y^2 + \mu_y^2 - 2y\mu_y]$

$$= E[r^2 \sin^2\theta] - \mu_y^2$$

$$= E(r^2) E(\sin^2\theta) - \mu_y^2$$

$$= (\sigma_r^2 + \bar{r}^2) \cdot E\left[\frac{1 - \cos 2\theta}{2}\right] - \mu_y^2$$

$$= \frac{1}{2} (\sigma_r^2 + \bar{r}^2) (1 - \exp(-2\sigma_\theta^2) \cos 2\theta) - \mu_y^2$$

Substituting values, we get

$$\Rightarrow \frac{1}{2} (\sigma_r^2 + 1) (1 + e^{-2\sigma_\theta^2}) - e^{-\frac{\sigma_\theta^2}{2}}$$

$$\therefore \mu = \begin{bmatrix} 0 \\ e^{-\frac{\sigma_\theta^2}{2}} \end{bmatrix}$$

Finally,

$$\text{covariance} = \begin{bmatrix} \frac{1}{2} (\sigma_x^2 + 1) (1 - e^{-2\sigma_\theta^2}) & 0 \\ 0 & \frac{1}{2} (\sigma_x^2 + 1) (1 + e^{-2\sigma_\theta^2}) - e^{-\frac{\sigma_\theta^2}{2}} \end{bmatrix}$$

(ii) Mean & Covariance using first order linearization

Using Taylor series:

$$f = h(s_0) + h'(s_0)(s - s_0)$$

$$f = \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \bar{r} \cos \bar{\theta} \\ \bar{r} \sin \bar{\theta} \end{bmatrix} + \begin{bmatrix} \cos \bar{\theta} & -\bar{r} \sin \bar{\theta} \\ \sin \bar{\theta} & \bar{r} \cos \bar{\theta} \end{bmatrix} \begin{bmatrix} r - \bar{r} \\ \theta - \bar{\theta} \end{bmatrix}$$

 $h'(s) \rightarrow$ Jacobian of f at s_0 .

$$\mu = E \begin{bmatrix} \bar{r} \cos \bar{\theta} \\ \bar{r} \sin \bar{\theta} \end{bmatrix} + E \begin{bmatrix} \cos \bar{\theta} & -\bar{r} \sin \bar{\theta} \\ \sin \bar{\theta} & \bar{r} \cos \bar{\theta} \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\mu = \begin{bmatrix} \bar{r} \cos \bar{\theta} \\ \bar{r} \sin \bar{\theta} \end{bmatrix}$$

Substituting values, $\mu = \begin{bmatrix} \theta \\ 1 \end{bmatrix}$

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$$\text{Covariance} = E[(y - \mu)(y - \mu)^T]$$

$$= E \left[h'(z_0) \begin{bmatrix} x - \bar{x} \\ \theta - \bar{\theta} \end{bmatrix} \begin{bmatrix} x - \bar{x} & \theta - \bar{\theta} \end{bmatrix} h^T \right]$$

$$= \begin{bmatrix} \cos \bar{\theta} & -\bar{x} \sin \bar{\theta} \\ \sin \bar{\theta} & \bar{x} \cos \bar{\theta} \end{bmatrix} \begin{bmatrix} \sigma_x^2 & 0 \\ 0 & \sigma_\theta^2 \end{bmatrix} \begin{bmatrix} \cos \bar{\theta} & \bar{x} \sin \bar{\theta} \\ -\bar{x} \sin \bar{\theta} & \bar{x} \cos \bar{\theta} \end{bmatrix}$$

Substituting values,

$$\Sigma = \begin{bmatrix} \sigma_\theta^2 & 0 \\ 0 & \sigma_x^2 \end{bmatrix}$$