

Solutions.

1. a) The unit vector from A to the observer O is

$$e_A = \frac{1}{\|r_O - r_A\|} (r_O - r_A)$$

$$r_O = \begin{bmatrix} 0 \\ -2.3 \end{bmatrix}, \quad r_A = \begin{bmatrix} -1 \\ 0 \end{bmatrix}, \quad r_B = \begin{bmatrix} 0 \\ 1.5 \end{bmatrix}, \quad r_C = \begin{bmatrix} 1.3 \\ 0 \end{bmatrix}$$

$$e_A = \frac{1}{\sqrt{1 + 2.3^2}} \begin{bmatrix} 1 \\ -2.3 \end{bmatrix} = \begin{bmatrix} 0.3987 \\ -0.2917 \end{bmatrix}$$

Unit vector from B to O is $e_B = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$.

Unit vector from C to O is

$$e_C = \frac{1}{\|r_O - r_C\|} (r_O - r_C) = \frac{1}{\sqrt{1.3^2 + 2.3^2}} \begin{bmatrix} -1.3 \\ -2.3 \end{bmatrix} = \begin{bmatrix} -0.492 \\ -0.87 \end{bmatrix}$$

The measured range rates to points A & B are given by

$$\dot{r}_A = e_A^T v, \quad \dot{r}_B = e_B^T v, \quad \text{where } v \text{ is the velocity vector of } O.$$

Letting $A = \begin{bmatrix} e_A^T \\ e_B^T \end{bmatrix}_{2 \times 2}$ it follows that
$$v = A^{-1} \begin{bmatrix} \dot{r}_A \\ \dot{r}_B \end{bmatrix}$$

∴ Estimate of velocity

$$V = \begin{bmatrix} 0.3987 & -0.917 \\ 0 & -1 \end{bmatrix}^{-1} \begin{bmatrix} -4.58 \\ -5 \end{bmatrix}$$

$$= \begin{bmatrix} 0.012 \\ 5 \end{bmatrix} \text{ m/s.}$$

— 2½ mark.

The covariance matrix P_{AB} of the error vector $\begin{bmatrix} \delta p_A \\ \delta p_B \end{bmatrix}$ is the 1-2 submatrix of the covariance matrix of the error vector $[\delta p_A, \delta p_B, \delta p_C]^T$. That is,

$$P_{AB} = \begin{bmatrix} 0.2 & 0.1 \\ 0.1 & 0.3 \end{bmatrix}$$

∴ ~~Covariance matrix of the~~ The error in the velocity estimate is related to the measurement error

— by
$$\delta v = A \begin{bmatrix} \delta p_A \\ \delta p_B \end{bmatrix}$$

∴ Covariance matrix of the estimation error is

$$P_{vv} = A^{-1} \begin{bmatrix} 0.2 & 0.1 \\ 0.1 & 0.3 \end{bmatrix} A^{-T}$$

$$= \begin{bmatrix} 1.6914 & 0.4392 \\ 0.4392 & 0.3 \end{bmatrix}$$

— 1 mark.

The RMS value of the estimation error is

$$= \sqrt{\text{trace } P_{rr}} = \sqrt{1.9914} = 1.4112 \text{ m/s.} \quad \frac{1}{2} \text{ mark}$$

NOTE: If measurements A & C had to be used, to estimate velocity, then the estimate would be given

$$\text{by } v = \begin{bmatrix} e_A^T \\ e_C^T \end{bmatrix}^{-1} \begin{bmatrix} \dot{p}_A \\ \dot{p}_C \end{bmatrix} = \begin{bmatrix} 0.0055 \\ 4.9969 \end{bmatrix}.$$

The estimate ~~to~~ error covariance matrix would be

$$P_{rr} = \begin{bmatrix} e_A^T \\ e_C^T \end{bmatrix}^{-1} \begin{bmatrix} 0.2 & 0 \\ 0 & 0.25 \end{bmatrix} \begin{bmatrix} e_A^T \\ e_C^T \end{bmatrix}^{-T} = \begin{bmatrix} 0.5678 & 0.0091 \\ 0.0091 & 0.1374 \end{bmatrix}$$

resulting in a RMS error of $\sqrt{\text{trace } P_{rr}} = 0.8404 \text{ m/s.}$

If measurements B & C had to be used, then

$$\text{velocity estimate } v = \begin{bmatrix} e_B^T \\ e_C^T \end{bmatrix}^{-1} \begin{bmatrix} \dot{p}_B \\ \dot{p}_C \end{bmatrix} = \begin{bmatrix} 0.0 \\ +5.0 \end{bmatrix}$$

$$P_{rr} = \begin{bmatrix} e_B^T \\ e_C^T \end{bmatrix}^{-1} \begin{bmatrix} 0.3 & 0 \\ 0 & 0.25 \end{bmatrix} \begin{bmatrix} e_B^T \\ e_C^T \end{bmatrix}^{-T} = \begin{bmatrix} 1.9708 & -0.5305 \\ -0.5305 & 0.3 \end{bmatrix}$$

resulting in RMS error $\sqrt{\text{tr. } P_{rr}} = 1.5069.$

Note that the least RMS value results when the ~~most~~ worst sensor B is ignored. However, part (b) will show that BLUE gives least error compared to all the 3 combinations above.

b). To use BLUE, we need to relate measurements to the ~~quantity~~ quantity that we wish to estimate & the measurement error. In this case,

$$y = Hx + Sw, \text{ where.}$$

$$y = \begin{bmatrix} p_A \\ f_A \\ f_C \end{bmatrix} = \begin{bmatrix} -4.58 \\ -5 \\ -4.35 \end{bmatrix} \text{ is the measurement vector.}$$

$$H = \begin{bmatrix} e_A^T \\ e_B^T \\ e_C^T \end{bmatrix} = \begin{bmatrix} 0.3987 & -0.917 \\ 0 & -1 \\ -0.492 & -0.87 \end{bmatrix} \quad \left. \begin{array}{l} \rightarrow \\ \rightarrow \end{array} \right\} \begin{array}{l} 2\frac{1}{2} \\ \text{mark} \end{array}$$

~~x~~ v is the velocity vector to be estimated.

$\Delta Sw = [\delta p_A \ \delta f_A \ \delta f_C]$ is the measurement error.

The covariance matrix of Sw is given to be

$$P_{sw} = \begin{bmatrix} 0.2 & 0.1 & 0 \\ 0.1 & 0.2 & 0 \\ 0 & 0 & 0.25 \end{bmatrix}$$

∴ The best linear unbiased estimate is given by

$$\begin{aligned} \hat{v} &= (H^T P_{sw}^{-1} H)^{-1} H^T P_{sw}^{-1} y \\ &= \begin{bmatrix} 1.3213 & -0.3748 & -0.9618 \\ -0.4663 & -0.2436 & -0.3779 \end{bmatrix} y = \begin{bmatrix} 0.0066 \\ 4.9977 \end{bmatrix} \quad \left. \begin{array}{l} \rightarrow \\ \rightarrow \end{array} \right\} \begin{array}{l} 1 \\ \text{mark} \end{array} \end{aligned}$$

The covariance matrix of the estimate error is (2)

$$\begin{aligned} P_{\hat{x}\hat{x}} &= (H^T P_{ww}^{-1} H)^{-1} \\ &= \begin{bmatrix} 0.5235 & -0.0197 \\ -0.0197 & 0.1197 \end{bmatrix} \quad - \text{1 mark.} \end{aligned}$$

The RMS value of the estimation error is

$$\sqrt{\text{trace } P_{\hat{x}\hat{x}}} = \sqrt{0.6432} = 0.8020 \quad - \frac{1}{2} \text{ mark.}$$

2. The measurements are $x_1 = 806.63 \text{ m.}$

$$x_2 = 807.23 \text{ m.}$$

The standard deviations are $\sigma_1 = 0.321 \text{ m.}$

$$\sigma_2 = 0.334 \text{ m.}$$

The errors are uncorrelated. $\therefore \sigma_{12} = 0.$

The minimum variance estimate is given by

$$\hat{x} = \frac{\sigma_2^2}{\sigma_1^2 + \sigma_2^2} x_1 + \frac{\sigma_1^2}{\sigma_1^2 + \sigma_2^2} x_2 = 806.91 \text{ m.} \quad - 2 \text{ marks.}$$

The standard deviation of the estimation error is

$$\sigma = \sqrt{\frac{\sigma_1^2 \sigma_2^2}{\sigma_1^2 + \sigma_2^2}} = 0.231 \text{ m.} \quad - \text{1 mark.}$$

3 a The errors in ~~the~~ ~~the~~ the differences of bearings are related to the errors in the bearing measurements by

$$\delta\theta_{AB} = \delta\theta_A - \delta\theta_B, \quad \delta\theta_{BC} = \delta\theta_B - \delta\theta_C.$$

∴ Since all bearing measurements ^{errors} have the same mean, $\delta\theta_{AB}$ & $\delta\theta_{BC}$ have zero means. $\frac{1}{2}$ mark

Moreover, since $\delta\theta_A, \delta\theta_B, \delta\theta_C$ are uncorrelated, $\frac{1}{2}$ mark

$$\sigma_{AB}^2 = E(\delta\theta_{AB}^2) = E[(\delta\theta_A - \delta\theta_B)^2]$$

$$= \sigma_A^2 + \sigma_B^2 = 2\sigma^2. \quad \frac{1}{2} \text{ mark}$$

$$\sigma_{BC}^2 = E(\delta\theta_{BC}^2) = E[(\delta\theta_B - \delta\theta_C)^2]$$

$$= \sigma_B^2 + \sigma_C^2 = 2\sigma^2. \quad \frac{1}{2} \text{ mark}$$

$$E(\delta\theta_{AB} \delta\theta_{BC}) = E(\delta\theta_A \delta\theta_B - \delta\theta_A \delta\theta_C - \delta\theta_B^2 + \delta\theta_B \delta\theta_C)$$

$$= -\sigma^2. \quad \frac{1}{2} \text{ mark}$$

∴ If $\delta\theta = \begin{bmatrix} \delta\theta_{AB} \\ \delta\theta_{BC} \end{bmatrix}$ ~~is the~~, then

$$P_{\theta\theta} = \begin{bmatrix} 2\sigma^2 & -\sigma^2 \\ -\sigma^2 & 2\sigma^2 \end{bmatrix} = \begin{bmatrix} 0.18 & -0.09 \\ -0.09 & 0.18 \end{bmatrix}. \quad \frac{1}{2} \text{ mark}$$

b) ~~The measure~~ The positions of the landmarks are

$$r_A = \begin{bmatrix} x_A \\ y_A \end{bmatrix} = \begin{bmatrix} 0 \\ 2.6 \end{bmatrix}, r_B = \begin{bmatrix} x_B \\ y_B \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, r_C = \begin{bmatrix} x_C \\ y_C \end{bmatrix} = \begin{bmatrix} 2.26 \\ 0 \end{bmatrix}$$

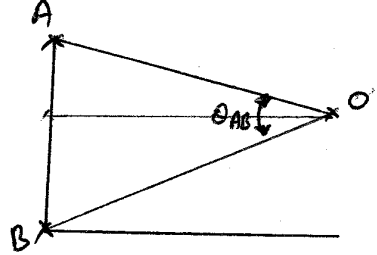
The estimated position of the observer is

$$r = \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 2.54 \\ 1.34 \end{bmatrix}$$

The measurement θ_{AB} is related to the position estimate

by

$$\theta_{AB} = \tan^{-1} \frac{(y-y_B)}{(x-x_B)} + \tan^{-1} \frac{(y_A-y)}{(x-x_A)}$$



$$= 54.2^\circ \quad (\text{numerical value is not required}) \quad | \text{mark.}$$

The measurement error $\delta\theta_{AB}$ & the estimate

error $\delta r = \begin{bmatrix} \delta x \\ \delta y \end{bmatrix}$ are related (to 1st order) by

$$\delta\theta_{AB} = \frac{\partial\theta_{AB}}{\partial x} \delta x + \frac{\partial\theta_{AB}}{\partial y} \delta y$$

$$= \frac{1}{\|r-r_B\|^2} \left[-(y-y_B) \delta x + (x-x_B) \delta y \right]$$

$$+ \frac{1}{\|r-r_A\|^2} \left[-(y_A-y) \delta x - (x-x_A) \delta y \right] \quad | \text{mark}$$

Substituting values yields

$$\|r-r_B\|^2 = \sqrt{2.54^2 + 1.34^2} = 8.2472$$

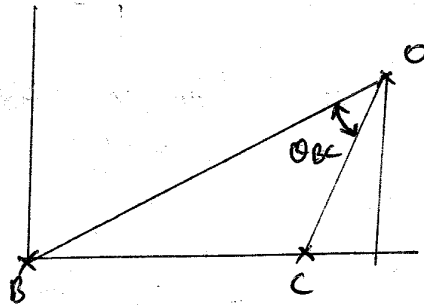
$$\|r-r_A\|^2 = 2.54^2 + (1.34-2.6)^2 = 8.0392$$

$$\therefore \delta \theta_{AB} = -0.319 \delta x - 0.08 \delta y.$$

Similarly, the measurement θ_{BC} is related to estimate r by

$$\theta_{BC} = \tan^{-1} \left(\frac{x-x_B}{y-y_B} \right) - \tan^{-1} \left(\frac{x-x_C}{y-y_C} \right) \quad \text{--- 1 mark.}$$

$$= 50.38^\circ \quad (\text{not required for the answer}).$$



$$\therefore \delta \theta_{BC} = \frac{\partial \theta_{BC}}{\partial x} \delta x + \frac{\partial \theta_{BC}}{\partial y} \delta y.$$

$$= \left[\frac{(y-y_B)}{\|r-r_B\|^2} - \frac{(y-y_C)}{\|r-r_C\|^2} \right] \delta x$$

$$+ \left[-\frac{(x-x_B)}{\|r-r_B\|^2} + \frac{(x-x_C)}{\|r-r_C\|^2} \right] \delta y \quad \text{--- 1 mark.}$$

Substituting values yields $\|r-r_C\|^2 = (2.54-2.20)^2 + (1.34)^2 = 1.874$

$$\therefore \delta \theta_{BC} = -0.552 \delta x - 0.158 \delta y$$

$$\text{Thus } \delta \theta = \begin{bmatrix} \delta \theta_{AB} \\ \delta \theta_{BC} \end{bmatrix} = A \delta r, \quad \text{where}$$

$$A = \begin{bmatrix} -0.319 & -0.08 \\ -0.552 & -0.158 \end{bmatrix} \quad \& \quad \delta r = \begin{bmatrix} \delta x \\ \delta y \end{bmatrix}. \quad \text{--- 1 mark}$$

$$\therefore \Sigma_r = A^{-1} \delta \Phi.$$

(3)

To find A^{-1} , we compute $\det A = 0.046$

$$\therefore A^{-1} = \begin{bmatrix} \frac{-0.158}{0.046} & \frac{0.008}{0.046} \\ \frac{0.552}{0.046} & \frac{-0.319}{0.046} \end{bmatrix} = \begin{bmatrix} -3.43 & 0.174 \\ 12 & -6.93 \end{bmatrix}$$

$$\begin{aligned} \therefore P_{rr} &= A^{-1} P_{\theta\theta} A^{-T} = \begin{bmatrix} -3.43 & 0.174 \\ 12 & -6.93 \end{bmatrix} \begin{bmatrix} 0.18 & -0.09 \\ -0.09 & 0.18 \end{bmatrix} \begin{bmatrix} -3.43 & 12 \\ 0.174 & -6.93 \end{bmatrix} \\ &= \begin{bmatrix} 2.2379 & -9.9739 \\ -9.9739 & 49.886 \end{bmatrix}. \quad \text{— 2 marks.} \end{aligned}$$

Thus the covariance matrix of the error in the position estimate is P_{rr} as above.

4. The mean of x_1 is

$$\begin{aligned} E(x_1) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x_1 f_{x_1, x_2}(x_1, x_2) dx_1 dx_2 = \int_0^1 \int_0^1 x_1 dx_1 dx_2 \\ &= \int_0^1 \left. \frac{x_1^2}{2} \right|_0^1 dx_2 = \frac{1}{2}. \end{aligned}$$

Similarly, $E(x_2) = \frac{1}{2}$.

$$\therefore \bar{x} = E(x) = \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}. \quad \text{1/2 mark.}$$

The variance of x_1 is

$$\begin{aligned}\sigma_{x_1}^2 &= E(x_1 - \bar{x}_1)^2 = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x_1 - 1/2)^2 f_{x_1, x_2}(x_1, x_2) dx_1 dx_2 \\ &= \int_0^1 \int_0^1 (x_1 - 1/2)^2 dx_1 dx_2 = \frac{1}{3} (x_1 - 1/2)^3 \Big|_0^1 \\ &= \frac{1}{3} \left[\frac{1}{8} - \left(-\frac{1}{8}\right) \right] = \frac{1}{12}.\end{aligned}$$

Similarly $\sigma_{x_2}^2 = E(x_2 - \bar{x}_2)^2 = \frac{1}{12}$. | 1/2 mark

$$\begin{aligned}\bar{\rho}_{x_1 x_2} &= E[(x_1 - \bar{x}_1)(x_2 - \bar{x}_2)] = \int_0^1 \int_0^1 (x_1 - 1/2)(x_2 - 1/2) dx_1 dx_2 \\ &= \left[\int_0^1 (x_2 - 1/2) dx_2 \right] \left(\frac{x_1^2}{2} - \frac{x_1}{2} \right) \Big|_0^1 = 0. \quad | \text{mark}.\end{aligned}$$

∴ Covariance matrix $\rho_{x_1} = \begin{bmatrix} \frac{1}{12} & 0 \\ 0 & \frac{1}{12} \end{bmatrix}$.

5). Taking Cuddalore & Port Blair to be lying on the same latitude, the rhumbline distance between the two is

$$\begin{aligned}R_{\text{RHUMB}} &= R \cos \phi |\lambda_B - \lambda_A| \\ &= 6378 \times \cos 11.7^\circ \times (92.5 - 79.77) \times \frac{\pi}{180} \text{ km.} \\ &= 1387.62 \text{ km.} \quad | \text{1/2 mark}.\end{aligned}$$

The great circle distance between the two parts is given by

$$R_{GC} = R \cos^{-1} \left[\cos^2 \phi \cos(\lambda_A - \lambda_B) + \sin^2 \phi \right]$$

$$= 6378 \times \frac{\pi}{180} \times \cos^{-1} \left[\cos^2 11.7 \cos(92.5 - 79.77) + \sin^2 11.7 \right] \text{ km.}$$

$$= 1387.5 \text{ km.}$$

1/2 mark.