

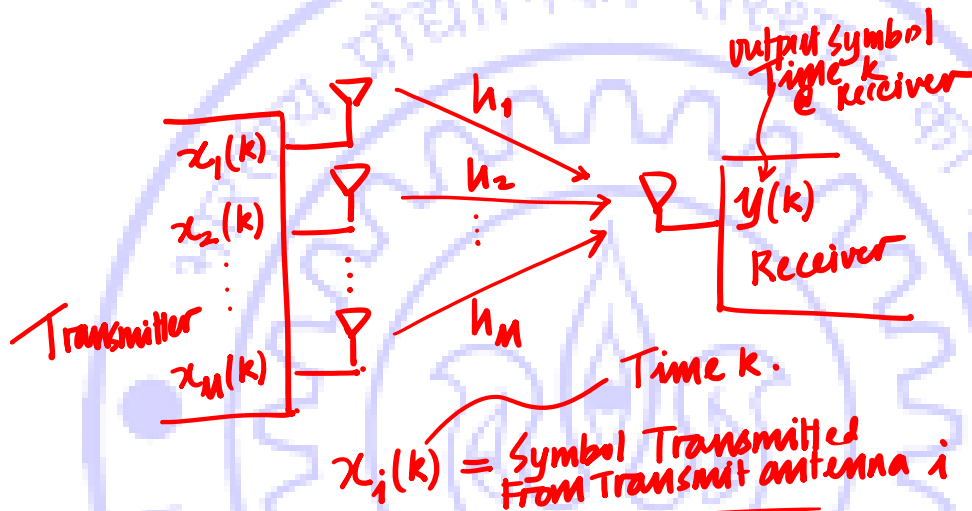
Live Interaction #1:

16th October 2023

E-masters Communication Systems

Estimation for Wireless

► Vector parameter estimation:



$$y(k) = x_1(k)h_1 + x_2(k)h_2 + \dots + x_M(k)h_M + v(k)$$

$$y(k) = \bar{\mathbf{x}}^T(k)\bar{\mathbf{h}} + v(k)$$

► How many pilot vectors? N

$$y(1) = \bar{\mathbf{x}}^T(1)\bar{\mathbf{h}} + v(1)$$

$$y(2) = \bar{\mathbf{x}}^T(2)\bar{\mathbf{h}} + v(2)$$

$$\vdots$$

$$y(N) = \bar{\mathbf{x}}^T(N)\bar{\mathbf{h}} + v(N)$$

► Net model can be expressed as:

$$\underbrace{\begin{bmatrix} y(1) \\ y(2) \\ \vdots \\ y(N) \end{bmatrix}}_{\substack{\bar{\mathbf{y}} \\ N \times 1}} = \underbrace{\begin{bmatrix} \bar{\mathbf{x}}^T(1) \\ \bar{\mathbf{x}}^T(2) \\ \vdots \\ \bar{\mathbf{x}}^T(N) \end{bmatrix}}_{\substack{\mathbf{X} \\ N \times M}} \underbrace{\bar{\mathbf{h}}}_{M \times 1} + \underbrace{\begin{bmatrix} v(1) \\ v(2) \\ \vdots \\ v(N) \end{bmatrix}}_{\substack{\bar{\mathbf{v}} \\ N \times 1}}$$

$$\bar{\mathbf{y}} = \mathbf{X}\bar{\mathbf{h}} + \bar{\mathbf{v}}$$

► $N \geq M \Rightarrow \mathbf{X}$ is a **tall-matrix**.

► Likelihood:

$$y(k) = \bar{\mathbf{x}}^T(k)\bar{\mathbf{h}} + v(k)$$

$$f_{Y(k)}(y(k)) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(y(k) - \bar{\mathbf{x}}^T(k)\bar{\mathbf{h}})^2}{2\sigma^2}}$$

$$p(\bar{\mathbf{y}}; \bar{\mathbf{h}}) = \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right)^N e^{-\frac{1}{2\sigma^2} \sum_{k=1}^N (y(k) - \bar{\mathbf{x}}^T(k)\bar{\mathbf{h}})^2}$$

$$= \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right)^N e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{y}} - \mathbf{X}\bar{\mathbf{h}}\|^2}$$

► Maximum likelihood reduces to

$$\underbrace{\min \|\bar{\mathbf{y}} - \mathbf{X}\bar{\mathbf{h}}\|^2}_{\text{Least squares}}$$

$$\hat{\mathbf{h}} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \bar{\mathbf{y}}$$

► Properties of estimator:

$$E\{\hat{\mathbf{h}}\} = \bar{\mathbf{h}}$$

$$E\{(\hat{\mathbf{h}} - \bar{\mathbf{h}})(\hat{\mathbf{h}} - \bar{\mathbf{h}})^T\} = \sigma^2 (\mathbf{X}^T \mathbf{X})^{-1}$$

$$\begin{aligned}
\text{MSE} &= \sum_{i=1}^M E \{ (\hat{h}_i - h_i)^2 \} \\
&= \text{Tr} \left\{ E \{ (\hat{\mathbf{h}} - \bar{\mathbf{h}})(\hat{\mathbf{h}} - \bar{\mathbf{h}})^T \} \right\} \\
&= \text{Tr} \left\{ E \left\{ \begin{bmatrix} \hat{h}_1 - h_1 \\ \hat{h}_2 - h_2 \\ \vdots \\ \hat{h}_M - h_M \end{bmatrix} [\hat{h}_1 - h_1 \quad \hat{h}_2 - h_2 \quad \dots \quad \hat{h}_M - h_M] \right\} \right\} \\
&\quad E \{ (\hat{\mathbf{h}} - \bar{\mathbf{h}})(\hat{\mathbf{h}} - \bar{\mathbf{h}})^T \} = \sigma^2 (\mathbf{X}^T \mathbf{X})^{-1} \\
\text{MSE} &= E \left\{ \text{Tr} \{ (\hat{\mathbf{h}} - \bar{\mathbf{h}})(\hat{\mathbf{h}} - \bar{\mathbf{h}})^T \} \right\} \\
&= \text{Tr} \left\{ E \{ (\hat{\mathbf{h}} - \bar{\mathbf{h}})(\hat{\mathbf{h}} - \bar{\mathbf{h}})^T \} \right\} \\
&= \text{Tr} \{ \sigma^2 (\mathbf{X}^T \mathbf{X})^{-1} \} = \sigma^2 \text{Tr} \{ (\mathbf{X}^T \mathbf{X})^{-1} \}
\end{aligned}$$

► Example:

$$\mathbf{X} = \begin{bmatrix} 1 & 1 \\ -1 & -1 \\ -1 & 1 \\ 1 & -1 \end{bmatrix}, \bar{\mathbf{y}} = \begin{bmatrix} -2 \\ 1 \\ 3 \\ -2 \end{bmatrix}, \sigma^2 = \frac{1}{2}$$

► Find ML estimate, error covariance, MSE.

$$\hat{\mathbf{h}} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \bar{\mathbf{y}}$$

$$\begin{aligned}
\mathbf{X}^T \mathbf{X} &= \begin{bmatrix} 1 & -1 & -1 & 1 \\ 1 & -1 & 1 & -1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -1 & -1 \\ -1 & 1 \\ 1 & -1 \end{bmatrix} \\
&= 4\mathbf{I}
\end{aligned}$$

$$\begin{aligned}
 (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T &= \frac{1}{4} \mathbf{I} \begin{bmatrix} 1 & -1 & -1 & 1 \\ 1 & -1 & 1 & -1 \end{bmatrix} \\
 &= \frac{1}{4} \begin{bmatrix} 1 & -1 & -1 & 1 \\ 1 & -1 & 1 & -1 \end{bmatrix} \\
 \hat{\mathbf{h}} &= (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \bar{\mathbf{y}}
 \end{aligned}$$

$$= \frac{1}{4} \begin{bmatrix} 1 & -1 & -1 & 1 \\ 1 & -1 & 1 & -1 \end{bmatrix} \begin{bmatrix} -2 \\ 1 \\ 3 \\ -2 \end{bmatrix}$$

$$\hat{\mathbf{h}} = \frac{1}{4} \begin{bmatrix} -8 \\ 2 \end{bmatrix} = \begin{bmatrix} -2 \\ 1/2 \end{bmatrix}$$

► Error covariance

$$\sigma^2 (\mathbf{X}^T \mathbf{X})^{-1} = \frac{1}{2} \times \frac{1}{4} \mathbf{I} = \frac{1}{8} \mathbf{I} = \begin{bmatrix} \frac{1}{8} & 0 \\ 0 & \frac{1}{8} \end{bmatrix}$$

► MSE

$$\text{MSE} = \text{Tr} \left\{ \begin{bmatrix} \frac{1}{8} & 0 \\ 0 & \frac{1}{8} \end{bmatrix} \right\} = \frac{1}{4}$$

- Assignment #3 deadline: 21st October Saturday 11:59PM.
- Live interaction 26th October 9-10 PM.
- Assignment #4 deadline: 28th October Saturday 11:59 AM.

- ▶ **Assignment #3, #4 discussion: 28th October Saturday 3-4 PM.**
- ▶ **Quiz #2 28th October Saturday 6-6:45 PM.**

