

NPTEL MOOC Estimation: Assignment #2

1. Consider MMSE estimation for the wireless sensor network (WSN) scenario as described in lectures with each observation $y(k) = h + v(k)$, for $1 \leq k \leq N$, i.e. number of observations N and IID Gaussian noise samples of variance σ^2 . What is the expression for the MMSE estimate \hat{h} of the unknown parameter h

a.
$$\hat{h} = \frac{\frac{(1^T y)/N + \frac{\mu_h}{\sigma_h^2}}{\frac{1}{\sigma^2/N} + \frac{1}{\sigma_h^2}}}$$

b.
$$\hat{h} = \frac{\frac{(1^T y)/N + \frac{\mu_h}{\sigma_h^2}}{\frac{1}{\sigma^2/N} + \frac{1}{\sigma_h^2}}}$$

c.
$$\hat{h} = \frac{\frac{(1^T y)/N + \frac{\mu_h}{\sigma_h^2}}{\frac{1}{\sigma^2} + \frac{1}{\sigma_h^2}}}$$

d.
$$\hat{h} = \frac{\frac{(1^T y)/N + \frac{\mu_h}{\sigma_h^2}}{\frac{1}{\sigma^2} + \frac{1}{\sigma_h^2}}}$$

Ans b

2. Consider MMSE estimation for the wireless sensor network (WSN) scenario as described in lectures with each observation $y(k) = h + v(k)$, for $1 \leq k \leq N$, i.e. number of observations N and IID Gaussian noise samples of dB variance σ^2 . As the number of observations $N \rightarrow \infty$, the MMSE estimate \hat{h} of the unknown parameter h tends to,

- 0
- 1
- ML Estimate
- Prior Mean

Ans c

3. Consider MMSE estimation for the wireless sensor network (WSN) scenario as described in lectures with each observation $y(k) = h + v(k)$, for $1 \leq k \leq 5$, i.e. number of observations $N = 5$ and IID Gaussian noise samples of dB variance $\sigma^2 = -3$ dB i.e. $10 \log_{10} \sigma^2 = -3$. Let the observations be $y(1) = 1, y(2) = 1, y(3) = 2, y(4) = 3/2, y(5) = 5/2$. Let $\mu_h = 1/2$ and $\sigma_h^2 = 1/4$. What is the maximum likelihood estimate \hat{h} of the unknown parameter h

- 1.6
- 1
- 1.5
- 2

Ans a

4. Consider MMSE estimation for the wireless sensor network (WSN) scenario as described in lectures with each observation $y(k) = h + v(k)$, for $1 \leq k \leq 5$, i.e. number of observations $N = 5$ and IID Gaussian noise samples of dB variance $\sigma^2 = -3$ dB i.e. $10\log_{10} \sigma^2 = -3$ dB. Let the observations be $y(1) = 1, y(2) = 1, y(3) = 2, y(4) = 3/2, y(5) = 5/2$. Let $\mu_h = 1/2$ and $\sigma_h^2 = 1/4$. What is the MMSE estimate \hat{h} of the unknown parameter h
- 10/7
 - 8/7
 - 5/7
 - 9/7

Ans d

5. Consider MMSE estimation for the wireless sensor network (WSN) scenario as described in lectures with each observation $y(k) = h + v(k)$, for $1 \leq k \leq 5$, i.e. number of observations $N = 5$ and IID Gaussian noise samples of dB variance $\sigma^2 = -3$ dB i.e. $10\log_{10} \sigma^2 = -3$ dB. Let the observations be $y(1) = 1, y(2) = 1, y(3) = 2, y(4) = 3/2, y(5) = 5/2$. Let $\mu_h = 1/2$ and $\sigma_h^2 = 1/4$. What is the posterior probability density function of the unknown parameter h
- Gaussian
 - Exponential
 - Rayleigh
 - Uniform

Ans a

6. Given the fading channel estimation problem where the output symbol $y(k)$ is $y(k) = hx(k) + v(k)$, with $h, x(k), v(k)$ denoting the *real* channel coefficient, pilot symbol and noise sample respectively. Let $\mathbf{x} = [x(1), x(2), \dots, x(N)]^T$ denote the pilot vector of transmitted pilot symbols and $\mathbf{y} = [y(1), y(2), \dots, y(N)]^T$ denote the corresponding received symbol vector. Let $v(k)$ be IID Gaussian noise with zero-mean and variance σ^2 . Let μ_h, σ_h^2 denote the prior mean, variance of the parameter h . The maximum likelihood estimate of the channel coefficient h is,

- $\hat{h} = \frac{\mathbf{x}^T \mathbf{y}}{\|\mathbf{x}\|}$
- $\hat{h} = \frac{\mathbf{y}^T \mathbf{y}}{\|\mathbf{x}\|^2}$
- $\hat{h} = \frac{\mathbf{x}^T \mathbf{y}}{\|\mathbf{x}\|^2}$
- $\hat{h} = \frac{\mathbf{y}^T \mathbf{y}}{\|\mathbf{x}\|}$

Ans c

7. Given the fading channel estimation problem where the output symbol $y(k)$ is $y(k) = hx(k) + v(k)$, with h , $x(k)$, $v(k)$ denoting the *real* channel coefficient, pilot symbol and noise sample respectively. Let $\mathbf{x} = [x(1), x(2), \dots, x(N)]^T$ denote the pilot vector of transmitted pilot symbols and $\mathbf{y} = [y(1), y(2), \dots, y(N)]^T$ denote the corresponding received symbol vector. Let $v(k)$ be IID Gaussian noise with zero-mean and variance σ^2 . Let μ_h , σ_h^2 denote the prior mean, variance of the parameter h . The covariance matrix R_{yy} of the output vector \mathbf{y} is,

- a. $R_{yy} = \sigma^2 \mathbf{xx}^T + \sigma_h^2 I$
- b. $R_{yy} = \sigma_h^2 \mathbf{xx}^T + \sigma^2 I$
- c. $R_{yy} = \mathbf{xx}^T + \sigma^2 I$
- d. $R_{yy} = \sigma_h^2 \mathbf{xx}^T + I$

Ans b

8. Given the fading channel estimation problem where the output symbol $y(k)$ is $y(k) = hx(k) + v(k)$, with h , $x(k)$, $v(k)$ denoting the *real* channel coefficient, pilot symbol and noise sample respectively. Let $\mathbf{x} = [x(1), x(2), \dots, x(N)]^T$ denote the pilot vector of transmitted pilot symbols and $\mathbf{y} = [y(1), y(2), \dots, y(N)]^T$ denote the corresponding received symbol vector. Let $v(k)$ be IID Gaussian noise with zero-mean and variance σ^2 . Let μ_h , σ_h^2 denote the prior mean, variance of the parameter h . The MMSE estimate of the channel coefficient h is,

- a. $\hat{h} = \frac{\frac{(\mathbf{x}^T \mathbf{y}) / \|\mathbf{x}\|^2 + \frac{\mu_h}{\sigma_h^2}}{\frac{1}{\sigma^2} + \frac{1}{\sigma_h^2}}}{\frac{1}{\sigma^2} + \frac{1}{\sigma_h^2}}$
- b. $\hat{h} = \frac{\frac{(\mathbf{x}^T \mathbf{y}) / \|\mathbf{x}\|^2 + \frac{\mu_h}{\sigma_h^2}}{\frac{1}{\sigma^2 / \|\mathbf{x}\|^2} + \frac{1}{\sigma_h^2}}}{\frac{1}{\sigma^2 / \|\mathbf{x}\|^2} + \frac{1}{\sigma_h^2}}$
- c. $\hat{h} = \frac{\frac{(\mathbf{x}^T \mathbf{y}) / \|\mathbf{x}\|^2 + \frac{\mu_h}{\sigma_h^2}}{\frac{1}{\sigma^2 / \|\mathbf{x}\|} + \frac{1}{\sigma_h^2}}}{\frac{1}{\sigma^2 / \|\mathbf{x}\|} + \frac{1}{\sigma_h^2}}$
- d. $\hat{h} = \frac{\frac{(\mathbf{x}^T \mathbf{y}) / \|\mathbf{x}\|^2 + \frac{\mu_h}{\sigma_h^2}}{\frac{1}{\sigma^2 / \|\mathbf{x}\|^2} + \frac{1}{\sigma_h^2}}}{\frac{1}{\sigma^2 / \|\mathbf{x}\|^2} + \frac{1}{\sigma_h^2}}$

Ans d

9. Consider the fading channel estimation problem where the output symbol $y(k)$ is $y(k) = hx(k) + v(k)$, with h , $x(k)$, $v(k)$ denoting the *real* channel coefficient, pilot symbol and noise sample respectively. Let $\mathbf{x} = \begin{bmatrix} 1 & 2 & 1 & 3 \\ 2 & & & 2 \end{bmatrix}^T$ denote the pilot vector of transmitted

pilot symbols and $\mathbf{y} = [-2 \ 2 \ -1 \ 1]^T$ denote the corresponding received symbol vector. Let $v(k)$ be IID Gaussian noise with zero-mean and dB variance $\sigma^2 = 3$ dB. Let $\mu_h = 1$, $\sigma_h^2 = 1/2$ denote the prior mean, variance of the parameter h . The MMSE estimate of the channel coefficient h is,

- a. 15/23
- b. 20/23
- c. 21/23
- d. 17/23

Ans a

10. Consider the fading channel estimation problem where the output symbol $y(k)$ is $y(k) = hx(k) + v(k)$, with h , $x(k)$, $v(k)$ denoting the *real* channel coefficient, pilot symbol and noise sample respectively. Let $\mathbf{x} = \left[\frac{1}{2} \ 2 \ 1 \ \frac{3}{2} \right]^T$ denote the pilot vector of transmitted pilot symbols and $\mathbf{y} = [-2 \ 2 \ -1 \ 1]^T$ denote the corresponding received symbol vector. Let $v(k)$ be IID Gaussian noise with zero-mean and variance σ^2 . Let $\mu_h = 1$, $\sigma_h^2 = 1/2$ denote the prior mean, variance of the parameter h . As the noise variance $\sigma^2 \rightarrow \infty$, the MMSE estimate of the channel coefficient h becomes,

- a. 15/23
- b. 7/4
- c. 1
- d. 0

Ans c