

ECE 520.651 Random Signal Analysis

Homework # 11

Due 9:00 AM on Thursday, November 29, 2007.

Review Chapter IV, Sections A through D, from Poor.

1. Solve problem **IV.F.13(b)–(c)** from Poor; simply state your answer to part **(a)** from Homework #10, without repeating your derivation.
2. Solve problem **IV.F.17** from Poor.
3. Solve problem **IV.F.18** from Poor.
4. Solve problem **IV.F.21** from Poor.
5. Solve problem **IV.F.24** from Poor. Compare your answers in part (a), and the mean squared-error $E_{\theta} [(\hat{\theta}_{\text{ML}} - \theta)^2]$, with the corresponding answers in example **IV.D.1**, and discuss the differences.

Review example **IV.B.3** and Chapter V, up to the statement of Proposition **V.B.1**.

6. Show that for jointly Gaussian random vectors \mathbf{X} and \mathbf{Y} , with

$$\mu_{\mathbf{X},\mathbf{Y}} = \begin{bmatrix} \mu_{\mathbf{X}} \\ \mu_{\mathbf{Y}} \end{bmatrix} \quad \text{and} \quad \Sigma_{\mathbf{X},\mathbf{Y}} = \begin{bmatrix} \Sigma_{\mathbf{X}\mathbf{X}} & \Sigma_{\mathbf{X}\mathbf{Y}} \\ \Sigma_{\mathbf{Y}\mathbf{X}} & \Sigma_{\mathbf{Y}\mathbf{Y}} \end{bmatrix},$$

the conditional density of \mathbf{X} given $\mathbf{Y} = \mathbf{y}$ is also a Gaussian pdf, with

$$\mu_{\mathbf{X}|\mathbf{Y}} = \mu_{\mathbf{X}} + \Sigma_{\mathbf{X}\mathbf{Y}}\Sigma_{\mathbf{Y}\mathbf{Y}}^{-1}(\mathbf{y} - \mu_{\mathbf{Y}}) \quad \text{and} \quad \Sigma_{\mathbf{X}|\mathbf{Y}} = \Sigma_{\mathbf{X}\mathbf{X}} - \Sigma_{\mathbf{X}\mathbf{Y}}\Sigma_{\mathbf{Y}\mathbf{Y}}^{-1}\Sigma_{\mathbf{Y}\mathbf{X}}.$$

Conclude that the MMSE estimate of \mathbf{X} based on \mathbf{Y} is the conditional mean

$$\hat{\mathbf{X}}(\mathbf{Y}) = \mu_{\mathbf{X}} + \Sigma_{\mathbf{X}\mathbf{Y}}\Sigma_{\mathbf{Y}\mathbf{Y}}^{-1}(\mathbf{Y} - \mu_{\mathbf{Y}}),$$

and the estimation error \mathbf{e} is a $\mathbf{0}$ -mean random vector with (unconditional) covariance

$$\Sigma_{\mathbf{e}\mathbf{e}} \equiv E[\mathbf{e}\mathbf{e}^T] = E[(\mathbf{X} - \hat{\mathbf{X}}(\mathbf{Y}))(\mathbf{X} - \hat{\mathbf{X}}(\mathbf{Y}))^T] = \Sigma_{\mathbf{X}\mathbf{X}} - \Sigma_{\mathbf{X}\mathbf{Y}}\Sigma_{\mathbf{Y}\mathbf{Y}}^{-1}\Sigma_{\mathbf{Y}\mathbf{X}}.$$