

Homework #7

Due 11/17

1. The joint PDF of X and Y is given as:

$$f_{XY}(x,y) = \begin{cases} 3e^{-(x+2y)} & 0 \leq x \leq \infty, 0 \leq y \leq x \\ 0 & \text{otherwise} \end{cases}$$

Give the expression for the MSE of the optimal nonlinear estimator. Note that optimal nonlinear estimator also obeys an orthogonality principle: the error is orthogonal to any nonlinear function of the data. Please show that the MSE is expressed as $MSE = E\{Y^2\} - E\{\widehat{Y_{NL}}^2\}$. Here $\widehat{Y_{NL}}$ denotes the nonlinear estimator.

You can use the same procedure I used when deriving the MSE for the linear estimator, and then use iterated expectation (condition on X) on the term $E\{Y \widehat{Y_{NL}}\}$. You may leave your answer in terms of integrals, but each term in your answer need only be a single integral involving a marginal PDF as opposed to a double integral involving the joint PDF.

2. An elevator has a capacity of maximum 19 adults. Suppose each adult's weight has a mean of 170 lbs with a standard deviation of 15 lbs. Give an approximation of the probability that the total weight of the people, if there are a maximum number of them, is greater than 3340 lbs. Why are these people so heavy, I wonder too.

Hint: Note that the total weight is a sum of 19 iid RVs. Central Limit theorem (which we haven't seen yet, don't panic) says that such a sum will have a Gaussian distribution.

3. We have seen that given a vector X of N jointly Gaussian random variables, any linear transformation to a set of M ($M \leq N$) new variables, Y , will produce jointly Gaussian random variables.

My random wish is to create a set of N jointly Gaussian random variables, Y , with a specified covariance matrix, C . I could start with a set of uncorrelated Gaussian random variables using a typical Gaussian random number generator, and then perform a linear transformation to produce a new set of Gaussian random variables with my desired covariance matrix. I heard that typical Gaussian random number generators create random variables with a zero mean and unit variance. But, I need your help. How should I select the transformation to produce my desired covariance matrix? Since you are at it, I also want a specific mean for my new random variable.

If it would help you to help me, I can give some numbers. My desired random variable has a mean vector of $\mu = [1, 0, 3, -2]$ and desired covariance matrix of

$$C = \begin{bmatrix} 30 & -10 & -20 & 4 \\ -10 & 30 & 4 & -20 \\ -20 & 4 & 30 & -10 \\ 4 & -20 & -10 & 30 \end{bmatrix}$$

4. Now I am just dreaming the opposite of the problem you have just solved. That is given a vector Y with a mean μ and covariance matrix C , how can I create a vector of zero-mean, unit-variance and uncorrelated random variables? It is like washing random vector's sins away. Tell me about the transformation I need for this.
5. Suppose $X = [X_1, X_2, X_3, X_4]$ is a Gaussian random vector. I take the X_i to be uncorrelated with equal variance so that the covariance matrix is $C_{XX} = \sigma^2 I$. Here I is denoting the identity matrix and, σ^2 must be the individual variances. Please let $Z = X_1X_2 + X_3X_4$. (Oops that's nonlinear!) Find the PDF of Z .
Hint: This is one where characteristic function gets handy.

1. Orthogonality Principle:

$$E\{[y - \hat{y}_{NL}] G(x)\} = 0$$

$$f_x(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_0^x 3e^{-(x+2y)} dy = \frac{3}{2} e^{-x} (1 - e^{-2x}) u(x)$$

$$f_{y|x}(y|x) = \frac{f(x, y)}{f_x(x)} = \frac{3e^{-(x+2y)}}{\frac{3}{2} e^{-x} (1 - e^{-2x})} = \frac{2e^{-2y}}{1 - e^{-2x}}$$

2. mean = 14.170
= 3230

variance = 14.152 = 4275

$P(W > 3340) =$

$$1 - \Phi\left(\frac{3340 - 3230}{\sqrt{4275}}\right) = 1 - \Phi(1.682) = 0.04633$$