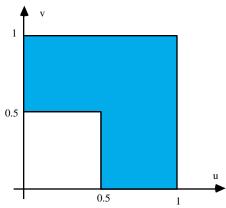
**Assigned:** Wednesday, April 30 Wednesday, May 7

**Reading:** Yates and Goodman: Chapters 5, 7 and 9.4–9.6

**Problems:** 

- 1. Let  $\mathbf{x}(\cdot)$  denote the characteristic function of a random variable  $\mathbf{X}$ .
- (a) What is the value of  $\mathbf{x}(0)$ ? Does it matter whether  $\mathbf{X}$  is discrete or continuous?
- (b) Assume that  $\mathbf{X}(\ )$  is differentiable at =0. What is the value of  $\frac{d}{d}$   $\mathbf{X}(\ )$  at =0?
- (c) What is the characteristic function of a Cauchy random variable? [Hint: the answer can be found in the inside front cover of the Kudeki-Munson Lecture Notes for ECE 210] Why can't the method of part (b) be used to compute the mean of a Cauchy random variable?
- 2. The random point (X,Y) is uniformly distributed on the shaded region shown below.
- (a) Find the marginal pdf  $f_{\mathbf{X}}(\mathbf{u})$  of the random variable  $\mathbf{X}$ .
- (b) Write down the marginal pdf  $f_{\mathbf{V}}(\mathbf{v})$  of the random variable Y from your answer to part (b).
- (c) Find  $P\{X < Y < 2X\}$ .
- (d) What is  $f_{\mathbf{X}|\mathbf{Y}}(\mathbf{u}|)$ , the conditional pdf of  $\mathbf{X}$  given that  $\mathbf{Y}=$ , if satisfies 0<<1/2? What is  $f_{\mathbf{X}|\mathbf{Y}}(\mathbf{u}|)$ , the conditional pdf of  $\mathbf{X}$  given that  $\mathbf{Y}=$ , if satisfies 1/2<<1? Now, apply the theorem of total probability to compute the unconditional pdf of  $\mathbf{X}$  from  $f_{\mathbf{X}|\mathbf{Y}}(\mathbf{u}|)$ . Do you get the same answer as in part (a)?



- 3. Let E[X] = 1, E[Y] = 4, var(X) = 4, var(Y) = 9, and var(Y) = 0.1
- (a) If  $\mathbf{Z} = 2(\mathbf{X} + \mathbf{Y})(\mathbf{X} \mathbf{Y})$ , what is  $\mathbf{E}[\mathbf{Z}]$ ?
- (b) If T = 2X + Y and U = 2X Y, what is cov(T, U)?
- (c) If  $\mathbf{W} = 3\mathbf{X} + \mathbf{Y} + 2$ , find  $\mathbf{E}[\mathbf{W}]$  and  $\mathbf{var}(\mathbf{W})$ .
- (d) If X and Y are jointly Gaussian random variables, and W is as defined in (c), what is  $P\{W>0\}$ ?
- **4.** This problem has three independent parts. Do not apply the numbers from one part to the others.
- (a) If  $var(\mathbf{X} + \mathbf{Y}) = 36$  and  $var(\mathbf{X} \mathbf{Y}) = 64$ , what is  $cov(\mathbf{X}, \mathbf{Y})$ ? If you are also told that  $var(\mathbf{X}) = 3 \cdot var(\mathbf{Y})$ , what is  $\mathbf{X} \cdot \mathbf{Y}$ ?
- (b) If var(X + Y) = var(X Y), are X and Y uncorrelated?
- (c) If var(X) = var(Y), are X and Y uncorrelated?
- 5. Consider the random point (X, Y) of Problem 2 above.
- (a) Compute E[X] and var(X).
- (b) Explain why the random variable  $\mathbf{Y}$  has the same mean and variance as  $\mathbf{X}$ .
- (c) Compute E[XY] and hence find cov(X,Y). should hold. *Is* the above equation satisfied by the numerical values you obtained?

- The conditional pdf of  $\mathbf{X}$  given  $\mathbf{Y} = \mathbf{w}$  was obtained in Problem 2 above, and it is easy to see that the conditional pdf of  $\mathbf{Y}$  given  $\mathbf{X} = \mathbf{i}$  is similar. Now, the **best** (least mean-square error) estimate of  $\mathbf{Y}$  given  $\mathbf{X} = \mathbf{i}$  is the mean of the conditional pdf of  $\mathbf{Y}$  given  $\mathbf{X} = \mathbf{i}$ . Thus, if  $\mathbf{X}$  has value 0.5, then  $\mathbf{\hat{Y}}$ , the best estimate of  $\mathbf{Y}$ , is 0.75 while if  $\mathbf{X}$  has value > 0.5, then  $\mathbf{\hat{Y}} = 0.5$ . Now, the **best linear** (least mean-square error) estimate of  $\mathbf{Y}$  (given that  $\mathbf{X}$  is known to have value  $\mathbf{Y}$  is  $\mathbf{Y} = \mathbf{Y} = \mathbf$
- Since the estimates  $\hat{\mathbf{Y}}$  and  $\hat{\mathbf{Y}}$  depend on the value of  $\mathbf{X}$ , they really are *functions* of  $\mathbf{X}$ , that is, they are *random variables* that can be expressed as  $\hat{\mathbf{Y}} = \begin{cases} 0.75, & 0 & \mathbf{X} & 0.5, \\ 0.5, & 0.5 < \mathbf{X} & 1 \end{cases}$  and  $\hat{\mathbf{Y}} = \mathbf{a} + \mathbf{b}\mathbf{X}$ . What are the average and the mean-square errors of each estimate? That is, what are the values of  $\mathbf{E}[(\mathbf{Y} \hat{\mathbf{Y}})]$ ,  $\mathbf{E}[(\mathbf{Y} \hat{\mathbf{Y}})]^2$ , and  $\mathbf{E}[(\mathbf{Y} \hat{\mathbf{Y}})^2]$ ?