

### Problem 1

We know that independent random variables are uncorrelated. Argue that uncorrelated jointly Gaussian random variables are independent.

Hint: do this for two random variables first. For  $n$  random variables, you might find it easier to use the characteristic function.

**Solution:** Let  $\bar{Z} = (Z_1, \dots, Z_n)^T$  be the  $n$  uncorrelated jointly Gaussian random variables. Let  $K$  be the covariance matrix of  $Z$ . ■

### Problem 2

- (i) \* Let  $X$  and  $Y$  be independent random variables.  $X_1 \sim N(0, 1)$ ; and  $Y = +1$  with probability  $p$  and  $Y = -1$  with probability  $1 - p$ . We define  $X_2 = YX_1$ . Is  $X_2$  Gaussian? Are  $X_1, X_2$  jointly Gaussian? Justify your answers.

[See Example 3.3.4 from [G] for a solution]

- (ii) Repeat (i) if  $X_1 \sim N(m, 1)$  and  $m > 0$

**Solution:** ■

### Problem 3 [G] Exercise 3.8

- (a) Let  $[K] = \begin{bmatrix} 0.75 & 0.25 \\ 0.25 & 0.75 \end{bmatrix}$ . Show that 1 and  $\frac{1}{2}$  are eigenvalues of  $[K]$  and find the normalized eigenvectors. Express  $[K]$  as  $[Q\Lambda Q^{-1}]$ , where  $[\Lambda]$  is diagonal and  $[Q]$  is orthonormal.
- (b) Let  $[K'] = \alpha[K]$  for real  $\alpha \neq 0$ . Find the eigenvalues and eigenvectors of  $[K']$ . Don not use brute force - think!
- (c) Find the eigenvalues and eigenvectors of  $[K^m]$ , where  $[K^m]$  is the  $m$ th power of  $[K]$ .

**Solution:** ■

### Problem 4

We derived the p.d.f. of a jointly Gaussian random vector  $X = AW$ , where  $A$  is an  $n \times n$  matrix. We used the fact  $A$  is invertible. How would you precisely describe the distribution of  $X$  if  $A$  is not invertible? Describe the underlying geometry of the distribution of  $X$ . Use the following  $A$  as an example:

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 1 & 1 & 1 \\ 2 & 3 & 4 \end{pmatrix}$$

**Solution:** ■

### Problem 5 [G] Problem 3.9

Let  $X$  and  $Y$  be jointly Gaussian with means  $m_X, m_Y$ , variances  $\sigma_X^2, \sigma_Y^2$ , and normalized covariance  $\rho$ . Find the conditional density  $f_{X|Y}(x | y)$ .

**Solution:**

■

In the next two problems we will use a common model for communication systems. The transmitted signal  $\vec{X}$  is a Gaussian random vector of size  $m$  (vector since there are several, say  $m$ , transmit antennas and each component of the vector stands for the input to a separate antenna). The signal goes over a linear and additive Gaussian noise channel and is picked up by a receiver which also has  $n$  antennas. The received vector of length  $n$  has the form.

$$\vec{Y} = H\vec{X} + \vec{Z}, \quad (1)$$

where  $H$  is a constant  $n \times m$  vector and  $\vec{Z}$  is a Gaussian random vector of size  $n$  and independent of  $\vec{X}$ .

### Problem 6

Let us first consider the simpler case of  $m = 1$  and  $n = 2$ . So  $X$  is a scalar random variable. Let  $X$  have the standard normal distribution  $N(0, 1)$ . The received signals are

$$Y_i = h_i X + Z_i, \quad i = 1, 2,$$

where  $Z_i \sim N(0, \sigma^2)$  are i.i.d and independent of  $X$ . And  $h_i$ 's are constants which represent the channel "gains" from the transmit antenna to the receive antennas.

- (a) Find the conditional joint distribution of  $Y_1, Y_2$  conditioned on  $X = x$ .
- (b) Find the conditional joint distribution of  $X$  conditioned on  $Y_1 = y_1, Y_2 = y_2$ .
- (c) Using (b), what is your estimate of the transmitted signal  $X$  if you are told that the receive antennas observed  $Y_1 = y_1, Y_2 = y_2$ . **Interpret your results.** Does your answer make intuitive sense? What happens to the estimate when the noise variance  $\sigma^2$  becomes small? or large?

**Solution:**

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### Problem 7

Now consider the general model in (1) for general  $n, m$ . Let  $\vec{X} \sim N(\vec{0}, K_X)$ ,  $\vec{Z} \sim N(\vec{0}, K_Z)$  and  $\vec{Z}$  is independent of  $\vec{X}$ .

- (a) Show that  $\vec{U} = (\vec{X}, \vec{Y})$  is jointly Gaussian. You may use any of the equivalent definitions we saw in class
- (b) Find a simple condition on  $H, K_X, K_Z$  so that  $K_U$  is invertible.
- (c) What is the conditional distribution of the input  $\vec{X}$  given the output  $\vec{Y} = \vec{y}$ .

**Solution:**

■