

# TSDT14 Signal Theory

## Solutions to the exam 2019-10-23

Mikael Olofsson, mikael@isy.liu.se

**1**

These subtasks have to be treated correctly as partial fulfillment to pass the exam.

- a. A time-discrete stochastic process is an infinite sequence of stochastic variables.
- b. The PSD  $R_Y(f)$  of the output is given by

$$R_Y(f) = |H(f)|^2 R_X(f) = \begin{cases} 4R_0, & |f| < 1, \\ 0, & \text{elsewhere} \end{cases}$$

- c. According to Page 15 in the T & F booklet, we have

$$\begin{aligned} r_Y(\tau) &= 9r_X^2(0)r_X(\tau) + 6r_X^3(\tau) \\ &= 9\text{sinc}(\tau) + 6\text{sinc}^3(\tau), \end{aligned}$$

where we have used the relation  $\text{sinc}(0) = 1$ .

**2**

We have a WSS time-continuous Gaussian process,  $X(t)$ , with mean zero and ACF

$$r_X(\tau) = R_0 \text{sinc}^2(\tau/T),$$

which is the input to an ideal BP filter with cutoff frequencies  $1/2T$  and  $3/2T$ , with amplification 1 in the passband.

We want to determine the probability that the output  $Y(t)$  is larger than 1.

Using the symmetry relation (second identity in the table on page 18 in T&F), the fact that the ACF is an even function, and the Fourier transform of a triangle (third identity in the table on page 19 in T&F), we write the PSD of  $X(t)$  as

$$R_X(f) = \mathcal{F}\{r_X(\tau)\} = R_0 T \text{triangle}(fT),$$

This signal is the input to a BP filter with frequency response

$$H(f) = \begin{cases} 1, & \frac{1}{2T} < |f| < \frac{3}{2T}, \\ 0, & \text{elsewhere.} \end{cases}$$

According to the super formula on page 9 in T&F, the output of that filter,  $Y(t)$ , has PSD

$$R_Y(f) = |H(f)|^2 R_X(f) = \begin{cases} R_X(f), & \frac{1}{2T} < |f| < \frac{3}{2T}, \\ 0, & \text{elsewhere.} \end{cases}$$

We notice that  $R_X(f)$  is zero for  $|f| \geq 1/T$ . Thus, we can rewrite the output PSD as

$$\begin{aligned} R_Y(f) &= |H(f)|^2 R_X(f) \\ &= \begin{cases} R_0 T (1 - |f|T), & \frac{1}{2T} < |f| < \frac{1}{T}, \\ 0, & \text{elsewhere.} \end{cases} \end{aligned}$$

We observe that we have an LTI system whose input is Gaussian. Thus the output  $Y(t)$  is also Gaussian. To determine the probability that the output  $Y(t)$  is larger than 1, we need the mean and the variance of  $Y(t)$ . Again, according to page 9 in T&F, we have the output mean

$$m_Y = m_X H(0) = 0,$$

since  $m_X$  is zero. Then for the variance, we have

$$\begin{aligned} \sigma_Y^2 &= r_Y(0) - m_Y^2 = r_Y(0) \\ &= \int_{-\infty}^{\infty} R_X(f) df = 2 \cdot \frac{1}{2} \cdot \frac{R_0 T}{2} \cdot \left( \frac{1}{T} - \frac{1}{2T} \right) = \frac{R_0}{4}, \end{aligned}$$

and thus we have the output standard deviation

$$\sigma_Y = \frac{\sqrt{R_0}}{2}.$$

We notice that  $Y(t)/\sigma_Y$  is  $N(0, 1)$ . The probability is therefore given as

$$\begin{aligned} \Pr\{Y(t) > 1\} &= \Pr\left\{ \frac{Y(t)}{\sigma_Y} > \frac{1}{\sigma_Y} \right\} \\ &= Q\left( \frac{1}{\sigma_Y} \right) = Q\left( \frac{2}{\sqrt{R_0}} \right) \end{aligned}$$

**3**

We have a time-discrete process,  $X[n]$ , consisting of independent, identically distributed, binary variables, taking values  $\pm 1$ , and  $p$  is the probability that  $X[n]$  is 1.

- a. We want the mean and ACF of  $X[n]$ . The mean is given by

$$m_X[n] = p \cdot 1 + (1 - p) \cdot (-1) = 2p - 1.$$

The ACF  $r_X[n + k, n]$  for  $k = 0$  is given by

$$r_X[n, n] = E\{X^2[n]\} = E\{1\} = 1.$$

For  $k \neq 0$ , we find

$$r_X[n + k, n] = E\{X[n + k]X[n]\} = m_X^2 = (2p - 1)^2,$$

where we have used the independence of consecutive samples in the second equality. As we can see, neither the mean  $m_X[n]$  nor the ACF  $r_X[n + k, n]$  depends on  $n$ . We have a WSS process with mean and ACF

$$m_X = 2p - 1, \\ r_X[k] = \begin{cases} 1, & k = 0, \\ (2p - 1)^2, & k \neq 0. \end{cases}$$

- b. Now we want to determine if the process  $Y[n] = X[n]X[n - 1]$  is WSS. The mean is given by

$$m_Y[n] = E\{X[n]X[n - 1]\} = r_X[1] = (2p - 1)^2,$$

which does not depend on  $n$ . The ACF  $r_Y[n + k, n]$  for  $k = 0$  is given by

$$r_Y[n, n] = E\{X^2[n]X^2[n - 1]\} = E\{1\} = 1.$$

For  $k = 1$ , we find

$$r_Y[n + 1, n] = E\{X[n + 1]X[n]X[n]X[n - 1]\} \\ = E\{X[n + 1]X[n - 1]\} \\ = r_X[2] = (2p - 1)^2.$$

Similarly for  $k = -1$ , we find

$$r_Y[n - 1, n] = r_X[2] = (2p - 1)^2.$$

For  $k > 1$ , we find

$$r_Y[n + k, n] = E\{X[n + k]X[n + k - 1]X[n]X[n - 1]\} \\ = m_X^4 = (2p - 1)^4,$$

where we have used the independence of consecutive samples in the second equality. Similarly for  $k < -1$ , we find

$$r_Y[n + k, n] = m_X^4 = (2p - 1)^4,$$

We notice that neither the mean  $m_Y[n]$  nor the ACF  $r_Y[n + k, n]$  depends on  $n$ . Thus, we have a WSS process with mean and ACF

$$m_Y = (2p - 1)^2, \\ r_Y[k] = \begin{cases} 1, & k = 0, \\ (2p - 1)^2, & |k| = 1, \\ (2p - 1)^4, & |k| > 1. \end{cases}$$

So yes, the process  $Y[n]$  is WSS.

#### 4

We have PAM with pulse-shape  $p(t)$ . The output is constant  $Y[n]$  in an interval of duration  $T$ . Identifying in the expression

$$Z(t) = \sum_{n=-\infty}^{\infty} Y[n]p(t - nT - \Psi),$$

where  $\Psi$  is uniformly distributed over  $[0, T)$ , gives us

$$p(t) = \begin{cases} 1, & T_0 \leq t < T + T_0, \\ 0, & \text{elsewhere,} \end{cases}$$

for some constant  $T_0$ . The Tables and Formulas booklet, page 18, gives us

$$|P(f)|^2 = T^2 \text{sinc}^2(fT).$$

Let  $H[\theta]$  be the frequency response of the filter that generates  $Y[n]$  from  $X[n]$ . This gives us

$$H[\theta] = 1 - e^{-j2\pi\theta} = e^{-j\pi\theta} (e^{j\pi\theta} - e^{-j\pi\theta}),$$

which in turn gives us

$$|H[\theta]|^2 = |e^{j\pi\theta} - e^{-j\pi\theta}|^2 = 4 \sin^2(\pi\theta) = 2[1 - \cos(2\pi\theta)].$$

We get the PSD of the input by determining the Fourier transform of its ACF:

$$R_X[\theta] = \frac{1}{2}e^{j2\pi\theta} + 1 + \frac{1}{2}e^{-j2\pi\theta} = 1 + \cos(2\pi\theta).$$

LTI-filtering (Tables and Formulas, p. 9) and PAM (Tables and Formulas, p. 12) gives us

$$R_Z(f) = \\ = \frac{1}{T}|P(f)|^2 \cdot |H[fT]|^2 \cdot R_X[fT] \\ = T \cdot \text{sinc}^2(fT) \cdot 2[1 - \cos(2\pi fT)][1 + \cos(2\pi fT)] \\ = T \cdot \text{sinc}^2(fT) \cdot 2[1 - \cos^2(2\pi fT)] \\ = T \cdot \text{sinc}^2(fT) \cdot 2 \sin^2(2\pi fT),$$

which is our answer.

#### 5

We are given the definition of the cross correlation function (CCF) of two jointly WSS complex processes,  $X(t)$  and  $Y(t)$ ,

$$r_{X,Y}(\tau) = E\{X(t + \tau)Y^*(t)\},$$

and the definition of the pseudo-CCF of those processes,

$$\tilde{r}_{X,Y}(\tau) = E\{X(t + \tau)Y(t)\}.$$

We want to determine demands on the CCF and/or pseudo-CCF of  $X(t)$  and  $Y(t)$  to guarantee that

$$Z(t) = X(t) + Y(t)$$

is circular if both  $X(t)$  and  $Y(t)$  are circular.

Since both  $X(t)$  and  $Y(t)$  are supposed to be circular, we have

$$\tilde{r}_X(\tau) = \mathbb{E}\{X(t+\tau)X(t)\} = 0,$$

$$\tilde{r}_Y(\tau) = \mathbb{E}\{Y(t+\tau)Y(t)\} = 0.$$

We now want to study the pseudo-ACF of  $Z(t)$  and determine under what circumstances it is zero. We have

$$\begin{aligned} \tilde{r}_Z(\tau) &= \mathbb{E}\{Z(t+\tau)Z(t)\} \\ &= \mathbb{E}\{(X(t+\tau) + Y(t+\tau))(X(t) + Y(t))\} \\ &= \mathbb{E}\{X(t+\tau)X(t)\} + \mathbb{E}\{X(t+\tau)Y(t)\} \\ &\quad + \mathbb{E}\{Y(t+\tau)X(t)\} + \mathbb{E}\{Y(t+\tau)Y(t)\} \\ &= \tilde{r}_X(\tau) + \tilde{r}_{X,Y}(\tau) + \tilde{r}_{Y,X}(\tau) + \tilde{r}_Y(\tau) \\ &= \tilde{r}_{X,Y}(\tau) + \tilde{r}_{Y,X}(\tau), \end{aligned}$$

where we have used that  $X(t)$  and  $Y(t)$  are circular in the last equality. Now,  $Z(t)$  is circular if  $\tilde{r}_Z(\tau)$  is zero. Thus, we have to demand

$$\tilde{r}_{X,Y}(\tau) = -\tilde{r}_{Y,X}(\tau).$$

Two questions remain:

- Are there pairs of processes for which this relation holds?
- Are there pairs of processes for which this relation does not hold?

The answer to both those questions is yes.

- An example of two jointly WSS circular processes for which the result holds is  $Y(t) = X(t)$ . Then  $\tilde{r}_{X,Y}(\tau)$  and  $\tilde{r}_{Y,X}(\tau)$  are both equal to  $\tilde{r}_X(\tau)$ , which is zero since  $X(t)$  is circular, and the relation holds trivially.
- An example of two jointly WSS circular processes for which the result does not hold is  $Y(t) = X^*(t)$ . Then we have  $\tilde{r}_{X,Y}(\tau) = r_X(\tau)$  and  $\tilde{r}_{Y,X}(\tau) = r_X^*(\tau)$ , which in all interesting cases have a non-zero real part. Thus, the relation does not hold.

In other words, it makes sense to check if the relation holds.

## 6

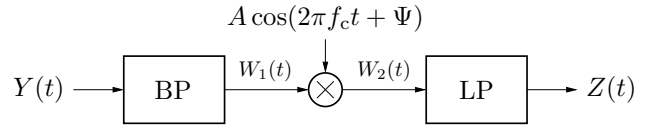
Given situation:  $M(t)$  is a WSS strictly bandlimited baseband message with bandwidth  $W$ . This message is modulated using AM-SC, i.e. the sent signal is

$$X(t) = M(t) \cos(2\pi f_c t + \Psi),$$

where in turn  $\Psi$  is uniformly distributed over  $[0, 2\pi)$  and where the carrier frequency fulfills  $f_c > W$ . The channel adds WGN  $N(t)$  which means that the received signal  $Y(t)$  is given by

$$Y(t) = X(t) + N(t).$$

$M(t)$ ,  $\Psi$  and  $N(t)$  are assumed to be mutually independent. The receiver is given in the figure below, with a non-zero  $A$ .



The two filters are ideal, with cut-off frequencies chosen so that the useful signal passes the filter unchanged. We assume full synchronization, i.e.  $\Psi$  is completely known at the receiver. Finally, we are supposed to show that the resulting SNR at the output of the receiver does not depend on  $A$ .

The short solution is the following, which uses the fact that everything is linear here. Let the output be

$$Z(t) = M_3(t) + N_3(t),$$

where  $M_3(t)$  is the useful part of the output, and where  $N_3(t)$  is the remaining noise in the output, for the case where we have  $A = 1$ . The SNR of the output is then given by

$$\text{SNR} = \frac{P_{M_3}}{P_{N_3}}.$$

Now, consider an arbitrary non-zero  $A$ . Then the output is given by

$$Z(t) = AM_3(t) + AN_3(t),$$

and the power of the useful part of  $Z(t)$  is the power of  $AM_3(t)$ , which is  $A^2 P_{M_3}$ . Similarly, the power of the noise in  $Z(t)$  is the power of  $AN_3(t)$ , which is  $A^2 P_{N_3}$ . The resulting SNR is now

$$\text{SNR} = \frac{A^2 P_{M_3}}{A^2 P_{N_3}} = \frac{P_{M_3}}{P_{N_3}},$$

which does not depend on  $A$ . However, for this to make any sense, we have to demand that  $A$  is non-zero as the problem formulation states.

**Note:** An unnecessarily long solution would be to define useful signals and noise for each signal in the figure, i.e.

$$W_1(t) = M_1(t) + N_1(t),$$

$$W_2(t) = M_2(t) + N_2(t),$$

$$Z(t) = M_3(t) + N_3(t),$$

where  $M_k(t)$  are useful signals and  $N_k(t)$  are noise in the three positions, and express signals and PSDs and finally get exact expressions of  $P_{M_3}$  and  $P_{N_3}$  above, which would both contain  $A^2$  as a factor. However, the essential part of such a solution is actually the short solution given above.