



12 - R991KJRDHAJS9SV



0406186522

FÖRSÄTTSLAD TENTAMEN/ EXAMINATION COVER

Jag intygar att mobiltelefon och annan otillåten elektronisk utrustning är avstängd och förvaras på anvisad plats. / I hereby confirm that mobile phones and other unauthorized electronic equipment is shut off and placed according to instructions

MARKERA MED "X" /
MARK WITH "X"



IFYLLES AV STUDENT OCH TENTAMENSVAKT /
TO BE FILLED IN BY THE STUDENT AND THE INVIGILATOR:

KURSKOD / COURSE CODE E Q 1 2 2 0		EFTERNAMN / FAMILY NAME Domur																	
KURSNAMN / COURSE NAME Signal Theory		FÖRNAMN / FIRST NAME Alexandre																	
PROVKOD / TEST CODE T E N 1		NAMNTECKNING / YOUR SIGNATURE A. Domur																	
TENTAMENSDATUM / EXAMINATION DATE Y/Y/Y/Y M/M D/D 2 0 1 6 - 1 2 - 2 1		PERSONNUMMER / PERSONAL NUMBER Y/Y/M/M/D/D 9 4 0 6 1 8 - T 1 5 7																	
PROGRAMKOD / PROGRAM CODE:	INLÄMNINGSTID / TIME SUBMITTED: 12:05	SIGNATUR TENTAMENSVAKT / SIGNATURE INVIGILATOR: H	ANTAL SIDOR / NO OF PAGES: 10																
MARKERA BEHANDLADE UPPGIFTER MED "X" OCH EJ BEHANDLADE UPPGIFTER MED "-." / MARK WITH "X" PROBLEMS SOLVED. MARK WITH "-." PROBLEMS NOT ATTEMPTED																			
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
X	X	X	X	X															

IFYLLES AV INSTITUTIONEN / TO BE FILLED IN BY THE DEPARTMENT:

BEDÖMNING / ASSESSMENT																			
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20

BONUSPOÄNG /
BONUS POINTS:

--	--	--

SLUTSUMMA /
FINAL POINTS:

--	--	--

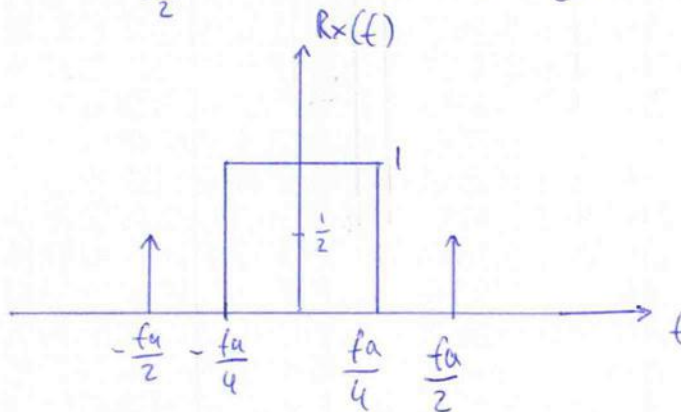
BETYG /
GRADE:

--

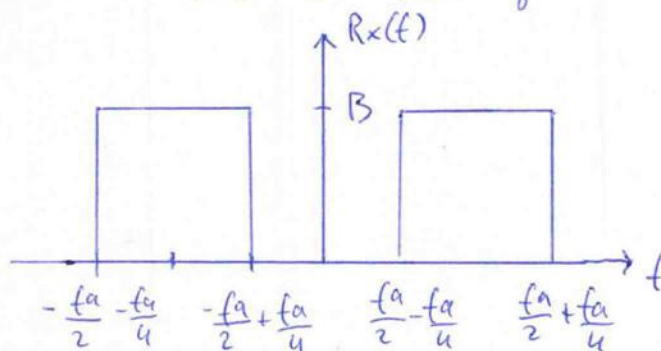
Godkänns av examinator /
approved by Examiner.....

Ex 1

$$\begin{aligned}
 a) \quad R_x(f) &= \mathcal{F}\{r_x(\tau)\} = \mathcal{F}\left\{2B \frac{\sin(\frac{1}{2}\pi fa\tau)}{\pi\tau} \cos(\pi fa\tau)\right\} \\
 &= 2B \left(\mathcal{F}\left\{\frac{\sin(\pi \frac{fa}{2}\tau)}{\pi\tau}\right\} * \mathcal{F}\{\cos(2\pi \frac{fa}{2}\tau)\} \right) \\
 &= 2B \left(\text{rect}_{\frac{fa}{2}}(f) * \left(\frac{1}{2}(\delta(f - \frac{fa}{2}) + \delta(f + \frac{fa}{2}))\right) \right)
 \end{aligned}$$



By graphical convolution and scaling with a factor = 2B we get:

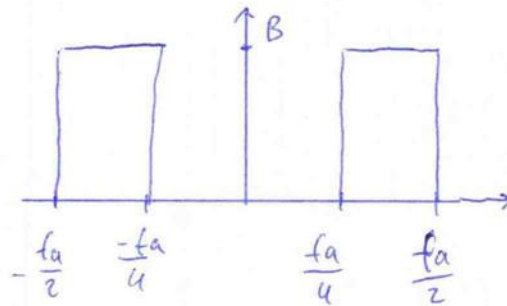


$$R_x(f) = \begin{cases} B & |fa - f| \leq \frac{fa}{4} \\ 0 & \text{otherwise} \end{cases}$$

I write on the
other side too
sorry :/

B) Since by the super formula we have: $R_Y(f) = |A(f)|^2 R_X(f)$
and $A(f)$ is simply a cut off of frequency $\frac{1}{2}f_a$ it is easy to see
by looking at the graph in a) that

$$R_Y(f) = \begin{cases} 1 & \text{if } -\frac{f_a}{4} \leq f \leq \frac{f_a}{4} \text{ or } \frac{f_a}{4} \leq f \leq \frac{3f_a}{4} \\ 0 & \text{otherwise} \end{cases}$$

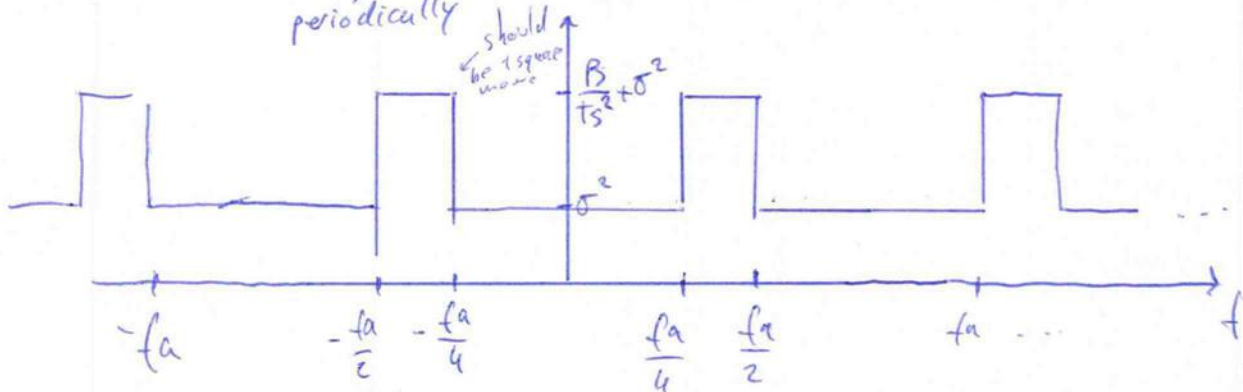


~~NA~~ $\lambda \geq 1$

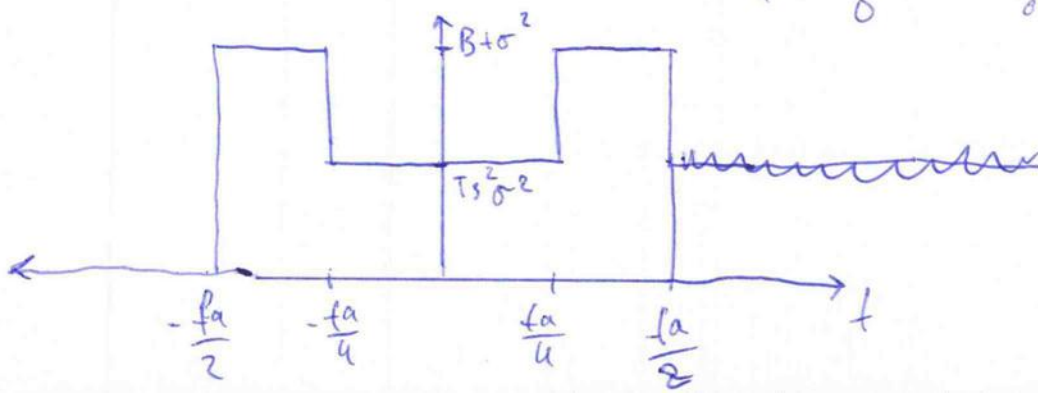
First we take ~~NA~~ which means that the Nyquist theorem
is satisfied. Since $w(n)$ is independent of $x(t)$ and $|P(f)|=1$
and the PAM is an LTI system
we have ~~$R_Z = R_Y$~~

$$E[w(n)w(m)] = \delta_{nm} \quad \text{after sampling}$$

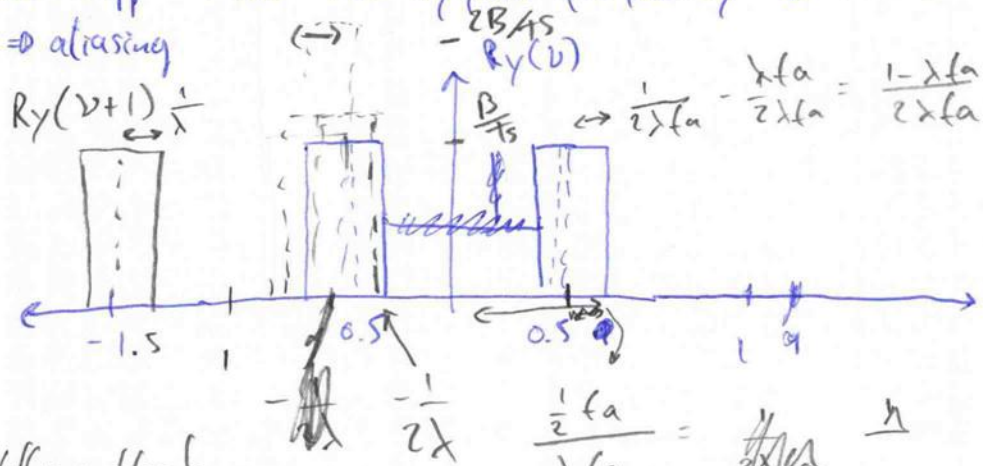
$$R_Z(f) = \frac{1}{T_s^2} R_Y(f) + \sigma^2$$



$$R_X(f) \text{ will be } = |C(f)|^2 \cdot R_Z(f) = \begin{cases} T_s^2 \left(\frac{1}{T_s^2} R_Y(f) + \sigma^2 \right) & \text{if } |f| \leq \frac{1}{2}f_a \\ 0 & \text{otherwise} \end{cases}$$

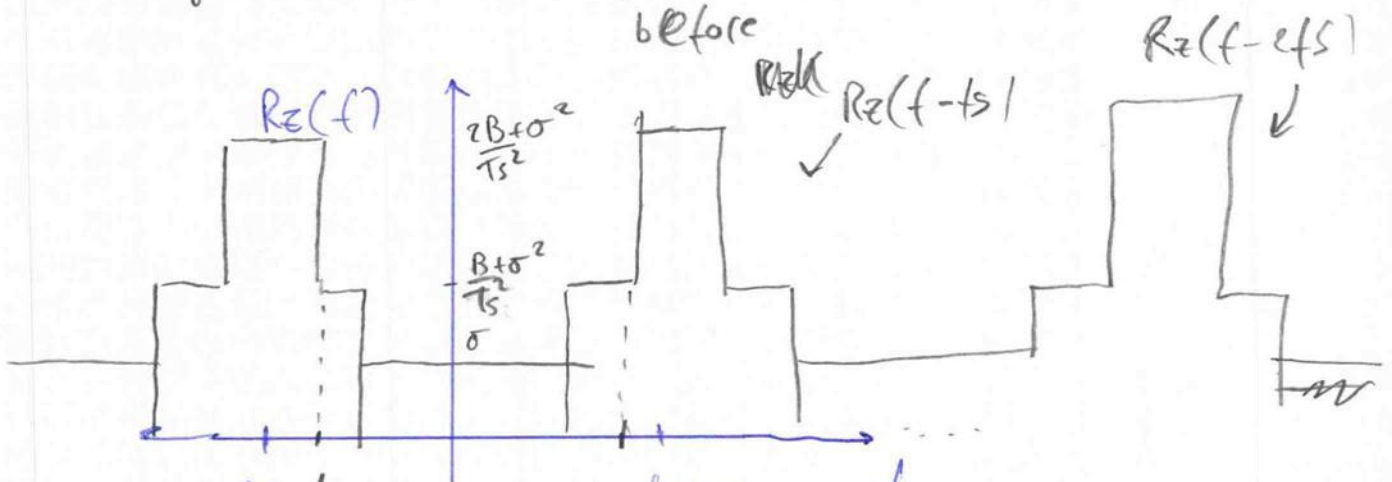


If now we take $\frac{3}{4} \lambda_s$ while sampling, we will have an overlap since the Nyquist frequency theorem is not satisfied \Rightarrow aliasing

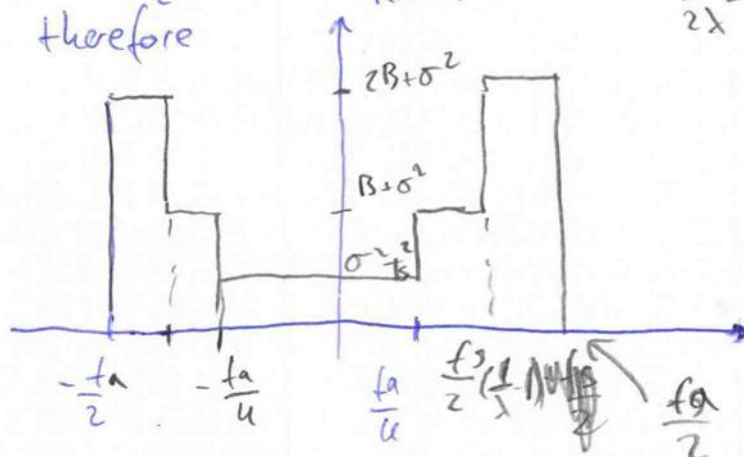


it follows that,

following the same reasoning as before



and therefore



P

c) for $\lambda > 1$ we have $P_e = 2 \cdot \left(\frac{f_a}{2} - \frac{f_a}{4} \right) \sigma^2 + \left(\frac{f_a}{4} - \left(-\frac{f_a}{4} \right) \right) T_s^2 \sigma^2$

$$= \frac{f_a}{2} \sigma^2 + \frac{f_a}{2} \sigma^2 T_s^2 = \frac{f_a}{2} \sigma^2 (1 + T_s^2)$$

$$P_s = 2 \cdot \left(\frac{f_a}{2} - \frac{f_a}{4} \right) B = \frac{f_a}{2} B$$

$$SNR = \frac{\frac{f_a}{2} B}{\frac{f_a}{2} \sigma^2 (1 + T_s^2)} = \frac{B}{\sigma^2 (1 + T_s^2)}$$

for $\lambda < 1$ we have $P_e = \frac{f_a}{2} \sigma^2 (1 + T_s^2) = \text{same as before}$

$$\frac{f_s}{2} \left(\frac{1}{\lambda} - 1 \right) = \frac{\lambda f_a}{2\lambda} - \frac{\lambda f_a}{2} = \frac{f_a}{2} - \frac{\lambda f_a}{2} \leftarrow \text{which is not correct since when we pick } \lambda = 1 \text{ we have } 0$$

$$P_s = 2 \left(\frac{f_a}{2} - \frac{f_s}{2} \left(\frac{1}{\lambda} - 1 \right) \right) 2B + 2 \cdot \left(\frac{f_s}{2} \left(\frac{1}{\lambda} - 1 \right) \right) B$$

$$= \dots = \text{thus } \lambda f_a 2B + (1 - \lambda) f_s B \Rightarrow SNR = \frac{P_s(\lambda)}{P_e}$$

something like that

d) If we see the SNR as ~~SNR~~ ~~SNR(λ)~~ SNR(λ) then we simply have to find when

$$\frac{\partial SNR(\lambda)}{\partial \lambda} = 0$$

and verify that the solution λ_s verify

$$\frac{\partial^2 SNR(\lambda)}{\partial \lambda^2} < 0$$

But I don't have any time left

$$T \sim \text{Exp}(\lambda)$$

$$f_T(t) = \lambda e^{-\lambda t} \quad t \geq 0$$

$$\begin{aligned} \text{a) } E[T] &= \int_{-\infty}^{\infty} t f_T(t) dt = \int_0^{\infty} t \lambda e^{-\lambda t} dt = -\frac{t}{e^{-\lambda t}} \Big|_0^{\infty} + \left[-\frac{1}{\lambda} e^{-\lambda t} \right]_0^{\infty} \\ &= 0 + \left[0 - \left(-\frac{1}{\lambda} \right) \right] = \frac{1}{\lambda} \end{aligned}$$

$$\begin{aligned} \text{b) We have to derive } P_r(T \leq 5 | T \geq 5) &= \frac{P_r(T \leq 5 \cap T \geq 5)}{P_r(T \geq 5)} \\ &= \frac{\int_5^5 \lambda e^{-\lambda t} dt}{\int_5^{\infty} \lambda e^{-\lambda t} dt} = \frac{[-e^{-\lambda t}]_5^5}{[-e^{-\lambda t}]_5^{\infty}} = \frac{e^{-\lambda 5} - e^{-\lambda 5}}{e^{-\lambda 5}} = 1 - e^{-\lambda(5-5)} \\ &= 1 - e^{-\lambda 0} = 1 - 1 = 0 \end{aligned}$$

\Rightarrow Probability of waiting 5_0 more minutes follows $\text{Exp}(\lambda)$

since $\frac{\partial}{\partial 5_0} (1 - e^{-\lambda 5_0}) = \lambda e^{-\lambda 5_0}$ so indep. of 5_s

$$\text{c) } Z = X + Y \quad f_Z(z) = \int_x f_{X,Y}(z-x, x)$$

$$Y(t) \geq 0 \Rightarrow X(t) \leq 0$$

$$f_{X(t)Y(t)}(x,y) = f_{X|Y}(x|y) \cdot f(y) = f_{X|Y}(x|y) f(y) + f_{X|Y}(x,y) f(y)$$

is negative
↓
neg

y is positive
↓
neg

$$f_{Y|X}(y|x) \cdot f(x) +$$

neg
↓

$$\mathbb{E}[W(n)U(n)] =$$

$$R(v) \downarrow r(v) = \sigma^2 +$$

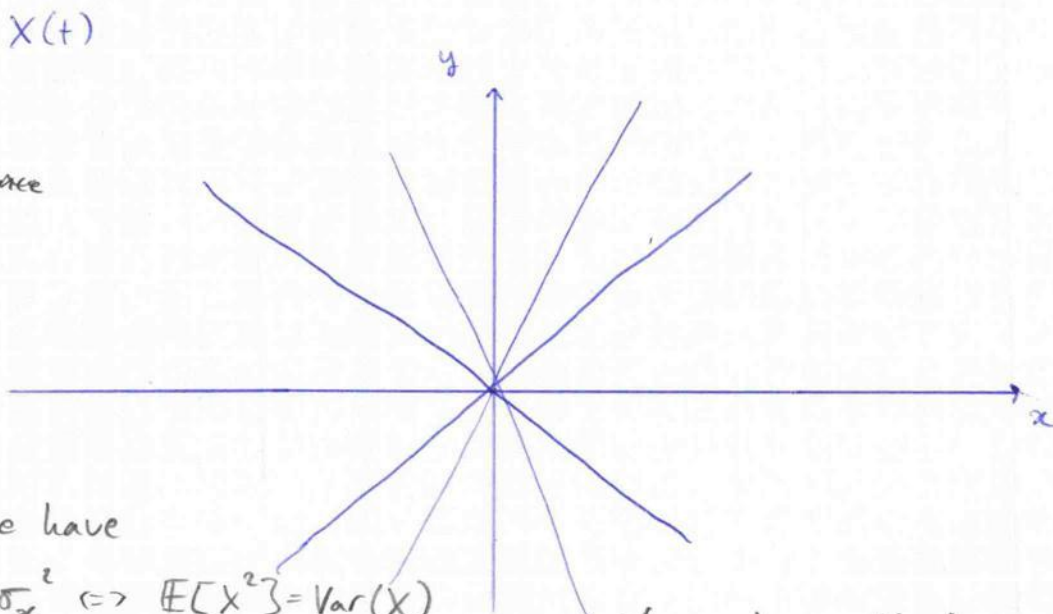
$$\mathbb{E}[V(t) \cdot \mathbb{E}$$

$$\mathbb{E}C$$

Ex 3

a)

Since



Since we have

$$R_X(f) = \sigma_x^2 \Leftrightarrow E[X^2] = \text{Var}(X)$$

$$m_X = 0 \quad \text{same for } m_Y$$

$$\text{and } f_{X,Y}(x,y) = \frac{f(x|y)}{f(y)} = \frac{f(y|x)}{f(x)}$$

It is

if we know that

~~y is negative~~

then the probability will lie on the positive side for $f(x)$

b)

$$r_{XY}(\tau) = E[X(t+\tau)Y(t)] = E[X(t+\tau)]E[Y(t)]$$

$$= 0 \quad \forall \tau \quad \text{since } E[X(t)] = 0 \text{ and } E[Y(t)] = 0$$

and the fact that they are uncorrelated

$$r_{XY}(0) = E[X(t)Y(t)]$$

c)

Yes Since the addition of two gaussian variable is still gaussian

and if we scale a Gaussian variable it will remain Gaussian

d)

b) Yes, since the PSD estimator is only interested of the value ~~for which~~ $\hat{r}_x(1)$ $\hat{r}_x(0)$ and with an AR(1) process we fix a condition on how 2 samples that differ from one time unit we can fix a strong constraint involving α and σ^2

c)

d) We have with that

$$\underbrace{\begin{pmatrix} \hat{r}_x(0) & \hat{r}_x(1) & \dots & \hat{r}_x(P-1) \\ \hat{r}_x(1) & \hat{r}_x(0) & & \\ \vdots & & \ddots & \\ \hat{r}_x(P-1) & & & \hat{r}_x(0) \end{pmatrix}}_{R_{\hat{x}}} \underbrace{\begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_P \end{pmatrix}}_a = - \underbrace{\begin{pmatrix} \hat{r}_x(1) \\ \vdots \\ \hat{r}_x(P) \end{pmatrix}}_{-r_{\hat{x}}}$$

We have to solve:

$a =$

$$a = -R_{\hat{x}}^{-1} r_{\hat{x}}$$



Alexandre Dumer

940618-T157

3 4

Family name, first name

Personal Registration Number

Programme

Sheet no.

Problem no.

e) It will happen that we will have a worse approximation ~~a~~ of the process, since the coefficients ~~will~~ will try to match the greater values

f)

$$Z_{n,i} = Y_{1,i} + \dots + Y_{n,i} \quad Y_{n,i} \text{ has zero mean}$$

Ex 5

Let $Z_{n,m}$ be a random seq $m=1,2,\dots$

$$a) \textcircled{1} E[Z_{n,i}] = E[Y_{1,i} + \dots + Y_{n,i}] = E[Y_{1,i}] + \dots + E[Y_{n,i}] = 0 \quad \forall i$$

$$\textcircled{2} MMSE[Z_{n,m}] = \hat{Z}_{n,m} = Z_{n,m-1} = E[Z_{n,m} | Z_{n,m-1}, \dots, Z_{n,1}]$$

Proof:

$$E[Z_{n,m} | Z_{n,m-1}, \dots, Z_{n,1}] = E[Y_{1,m} + \dots + Y_{n,m} | Y_{1,m-1} + \dots + Y_{n,m-1}, Y_{1,1} + \dots + Y_{n,1}]$$

$$E[\sum_{i=1}^{n,m-1} Z_{n,i}] \rightarrow \text{all are independent}$$

and mean 0 so it is equivalent for

$\hat{Z}_{n,m}$ to pick on of the possible $Z_{n,i}$ $\forall i < m$
So we will pick $Z_{n,m-1}$ hence it is martingal

b) We follow the same steps a)