

SOLUTIONS TO EXAM FOR RANDOM PROCESSES, 7.5 ECTS

January 16, 2010, 9.00 – 13.00

Max number of points: 30.

Halmstad University grading bounds: 12p \Rightarrow grade 3, 18p \Rightarrow grade 4, 24p \Rightarrow grade 5.

Allowed aids: Summary of formulae attached to the exam, calculator and Math. Handbook: Beta.

Examiner: Eric Järpe (035-16 76 53, 0702-822 844).

1. Prove that $G = \mathcal{F}(g)$, $h(\tau) \equiv G(\tau) \Rightarrow H(f) \equiv g(-f)$ (where $H = \mathcal{F}(h)$ and $g = \mathcal{F}^{-1}(G)$). (4p)

Solution: (See the document linked from the course homepage for a proof of this result.) \square

2. Prove that if $\{X_t : t \in \mathbb{R}\}$ is differentiable, then $E(X'_t) = 0$. (3p)

Solution: (See page 100 in the course literature.) \square

3. Let $\{X_t : t = 1, 2, 3, \dots\}$ be defined by $X_t = \frac{N_t - \lambda t}{\sqrt{\lambda t}}$, $t = 1, 2, 3, \dots$ where $\{N_t : t \in \mathbb{R}\}$ is a Poisson process with intensity λ .

(a) Calculate $P(X_2 \leq 1)$ if $\lambda = 3$. (3p)

(b) Show that X_t is asymptotically standard normally distributed, i.e. give some arguments to why $\lim_{t \rightarrow \infty} P(X_t \leq x) = \Phi(x)$. (4p)

Solution:

(a) $P(X_2 \leq 1) = P\left(\frac{N_2 - 2 \cdot 3}{\sqrt{2 \cdot 3}} \leq 1\right) = P(N_2 \leq \sqrt{6} + 6) = P(N_2 \leq 8.449 \dots) \stackrel{N_2 \in Poi(6)}{=} P(N_2 \leq 8) = 0.847$ (from the table).

(b) $X_t = \frac{N_t - \lambda t}{\sqrt{\lambda t}} = \frac{\sum_{k=1}^t Z_k - \lambda t}{\sqrt{\lambda t}}$ where $\{Z_k\}$ is a sequence of independent variables, each distributed $Poi(\lambda)$. Thus $\{Z_k\}$ could be considered to be a sample of a variable $Z \in Poi(\lambda) \Rightarrow E(Z) = \lambda$ and $V(Z) = \lambda$. Therefore, according to the Central Limit Theorem (see the summary of formulae) we have that $P\left(\frac{\sqrt{t}}{\sigma}(\bar{Z} - \mu) \leq x\right) \rightarrow \Phi(x)$ as $t \rightarrow \infty$ where $\bar{Z} = \frac{1}{t} \sum_{k=1}^t Z_k$, $\mu = E(Z) = \lambda$, $\sigma = \sqrt{V(Z)} = \sqrt{\lambda}$ and $\Phi(x)$ is the standard normal distribution function. Rewriting this we get $P\left(\frac{\sqrt{t}}{\sigma}(\bar{Z} - \mu) \leq x\right) = P\left(\sqrt{\frac{t}{\lambda}}\left(\frac{1}{t} \sum_{k=1}^t Z_k - \lambda\right) \leq x\right) = P\left(\frac{1}{\sqrt{\lambda t}}(N_t - \lambda t) \leq x\right) = P(X_t \leq x)$. Thus X_t is asymptotically standard normally distributed. \square

4. Let $\{X_t : t \in \mathbb{Z}\}$ be an $AR(1)$ process with $a_1 = -\frac{2}{3}$ and $\epsilon_t \in N(0, \frac{1}{2})$ (i.e. $V(\epsilon_t) = \frac{1}{2}$). Calculate $E(X_t)$ and $V(X_t)$. (3p)

Solution: For the $AR(1)$ process with $a_1 = -\frac{2}{3}$ we have $X_t - \frac{2}{3}X_{t-1} = \epsilon_t \Rightarrow X_t = \frac{2}{3}X_{t-1} + \epsilon_t$. Therefore $m = E(X_t) = E(\frac{2}{3}X_{t-1} + \epsilon_t) = \frac{2}{3}E(X_t) + E(\epsilon_t) = \frac{2}{3}m + 0 \Rightarrow E(X_t) = 0$ and $\sigma^2 = V(X_t) = C(\frac{2}{3}X_{t-1} + \epsilon_t, \frac{2}{3}X_{t-1} + \epsilon_t) = (\frac{2}{3})^2V(X_{t-1}) + 2 \cdot \frac{2}{3}C(X_{t-1}, \epsilon_t) + V(\epsilon_t) = \frac{4}{9}\sigma^2 + 0 + \frac{1}{2} \Rightarrow \sigma^2 = \frac{1/2}{1-4/9} = 0.9$. \square

5. The process $\{Y_t : t \in \mathbb{R}\}$ is weakly stationary with spectral density $R_Y(f) = |f|e^{-|f|}$. Now, when sampling Y_t one wants to choose the sampling distance such that the boundary value of the spectral density of the sampled process, $\{Z_t\}$, is not greater than 1, i.e. if one samples at $0, \pm d, \pm 2d, \pm 3d, \dots$ the number d is to be chosen such that $R_Z(\frac{1}{2d}) \leq 1$. What value should d have? (4p)

Solution: Since $R_Z(f)$ is non-zero only for $-\frac{1}{2d} < f \leq \frac{1}{2d}$, we have that

$$\begin{aligned}
R_Z(f) &= \sum_{k=-\infty}^{\infty} R_Y(f + \frac{k}{d}) \\
&= \sum_{k=-\infty}^{\infty} |f + \frac{k}{d}| e^{-|f+k/d|} \\
&= \sum_{k=-\infty}^{-1} -(f + \frac{k}{d}) e^{f+k/d} + |f + \frac{0}{d}| e^{-|f+0/d|} + \sum_{k=1}^{\infty} (f + \frac{k}{d}) e^{-(f+k/d)} \\
&= e^f \left(-f \sum_{k=1}^{\infty} e^{-k/d} + \sum_{k=1}^{\infty} \frac{k}{d} e^{-k/d} \right) + |f| e^{-|f|} + e^{-f} \left(f \sum_{k=1}^{\infty} e^{-k/d} + \sum_{k=1}^{\infty} \frac{k}{d} e^{-k/d} \right) \\
&= \frac{1}{d} (e^f + e^{-f}) \sum_{k=1}^{\infty} k (e^{-1/d})^k + |f| e^{-|f|} - f (e^f - e^{-f}) \sum_{k=1}^{\infty} (e^{-1/d})^k \\
&= \frac{1}{e^{1/d} - 1} \left(\frac{2 \cosh f}{d(1 - e^{-1/d})} - 2 \sinh f \right) + |f| e^{-|f|}
\end{aligned}$$

where $2 \cosh f = e^f + e^{-f}$, $2 \sinh f = e^f - e^{-f}$ and the last step of the calculation is due to the geometric series $\sum_{k=1}^{\infty} a^k = \frac{a}{1-a}$ and its derivative $\sum_{k=1}^{\infty} k a^k = a \sum_{k=1}^{\infty} k a^{k-1} = a \frac{d}{da} \sum_{k=1}^{\infty} a^k = a \frac{d}{da} \left(\frac{a}{1-a} \right) = \frac{a}{(1-a)^2}$.

By calculating $R_Z(\frac{1}{2d})$, for some values of d , one finds that $d \leq 0.43992$ will give a boundary value of the spectral density below 1. \square

6. Let $\{\xi_t : t \in \mathbb{R}\}$ be weakly stationary process with cvf $r(\tau) = \frac{\cos \tau}{1+\tau^2}$.

(a) Prove that $\{\xi_t\}$ is differentiable in squared mean. (3p)

(b) Calculate $P(\xi_t^2 \leq 2)$ assuming that $\{\xi_t\}$ is Gaussian with $m_\xi = 1$. (3p)

(c) Assume the input $\{\xi_t\}$ is filtered using a filter with impulse response $h(t) = \frac{1}{\pi}\delta_{-\pi}(t) + \frac{1}{\pi}\delta_\pi(t)$. Determine the cvf of the output signal. (4p)

Solution:

(a) $r(\tau) = \frac{\cos \tau}{1+\tau^2} \Rightarrow r'(\tau) = \frac{(\cos \tau) \cdot 2\tau - (-\sin \tau)(1+\tau^2)}{(1+\tau^2)^2} = \frac{2\tau \cos \tau + (1+\tau^2)\sin \tau}{(1+\tau^2)^2} \Rightarrow$
 $r''(\tau) = \dots = \frac{8\tau^2 \cos \tau}{(1+\tau^2)^3} + \frac{4\tau \sin \tau}{(1+\tau^2)^2} + \frac{(3+\tau^2)\cos \tau}{1+\tau^2}$ which is continuous for all $\tau \in \mathbb{R}$,
i.e. r is twice differentiable, i.e. $\{\xi_t\}$ is differentiable in squared mean.

(b) $P(\xi_t^2 \leq 2) = P(-\sqrt{2} \leq \xi_t \leq \sqrt{2}) = \underbrace{P(\xi_t \leq \sqrt{2})}_I - \underbrace{P(\xi_t \leq -\sqrt{2})}_{II}$.

Here $r(\tau) = \frac{\cos \tau}{1+\tau^2} \Rightarrow V(\xi_t) = r(0) = 1 \Rightarrow I = P(\xi_t \leq \sqrt{2}) = P\left(\frac{\xi_t - 1}{1} \leq \frac{\sqrt{2} - 1}{1}\right) = \Phi(\sqrt{2} - 1) = 0.6591$ and $II = P(\xi_t \leq -\sqrt{2}) = \Phi(-\sqrt{2} - 1) = 1 - \Phi(\sqrt{2} + 1) = 0.008$. Thus $P(\xi_t^2 \leq 2) = 0.6591 - 0.008 = 0.6511$.

(c) $r_Y(\tau) \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(u)h(v)r_\xi(\tau + u - v) du dv =$
 $= \int_{-\infty}^{\infty} \left(\frac{1}{\pi}\delta_{-\pi}(u) + \frac{1}{\pi}\delta_\pi(u) \right) \left(\frac{1}{\pi} \int_{-\infty}^{\infty} \delta_{-\pi}(v) \frac{\cos(\tau+u-v)}{1+(\tau+u-v)^2} dv + \frac{1}{\pi} \int_{-\infty}^{\infty} \delta_\pi(v) \frac{\cos(\tau+u-v)}{1+(\tau+u-v)^2} dv \right) du =$
 $\frac{1}{\pi} \left(\frac{1}{\pi} \cdot \frac{\cos(\tau-\pi+\pi)}{1+(\tau-\pi+\pi)^2} + \frac{1}{\pi} \cdot \frac{\cos(\tau-\pi-\pi)}{1+(\tau-\pi-\pi)^2} \right) + \frac{1}{\pi} \left(\frac{1}{\pi} \cdot \frac{\cos(\tau+\pi+\pi)}{1+(\tau+\pi+\pi)^2} + \frac{1}{\pi} \cdot \frac{\cos(\tau+\pi-\pi)}{1+(\tau+\pi-\pi)^2} \right) =$
 $\frac{\cos \tau}{\pi^2} \left(\frac{2}{1+\tau^2} + \frac{1}{1+(\tau+2\pi)^2} + \frac{1}{1+(\tau-2\pi)^2} \right). \quad \square$