

EEG Signal Processing - Radia Khayfani - 99101579

1-

2)

$f_k = \frac{K}{N} f_s$. we assume power line noise at 50 Hz

$$\rightarrow 50 = \frac{K}{512} \times 256 \rightarrow \boxed{K = 100}$$

so if we have peak on $K = 100, 200, 300, \dots$

it means we have 50 Hz, 100 Hz, 150 Hz, ...

components in signal, which means we have power line noise.

b)

alpha band: 8 - 13 Hz

delta band: 0.5 - 4 Hz

8 Hz $\sim K = 16$, 13 Hz $\sim K = 26$, 0.5 Hz $\sim K = 1$

4 Hz $\sim K = 8$

$$\rightarrow E_{\alpha\text{1pha}} = \sum_{k=16}^{26} |x[k]|^2 = \sum_{k=16}^{26} 2^8 \left(\frac{1}{2}\right)^k$$

$$= 2^8 \sum_{k=16}^{26} \left(\frac{1}{2}\right)^k = 2^8 \times \frac{2^{-16}(2^{-11}-1)}{-\frac{1}{2}}$$

$$E_{\text{delta}} = \sum_{k=1}^8 2^8 \left(\frac{1}{2}\right)^k = 2^8 \times \frac{2^{-1}(2^{-8}-1)}{-\frac{1}{2}}$$

$$\rightarrow \frac{E_{\alpha\text{1pha}}}{E_{\text{delta}}} = \frac{2^{-16}(2^{-12}-1)}{2^{-1}(2^{-7}-1)} \approx \boxed{2^{-15}}$$

c)

$$f_x = \frac{k}{N} f_s \rightarrow \alpha\text{1pha Band} = 8\text{ Hz} - 13\text{ Hz} \sim k=32-52$$

$$\text{delta Band} = 0.5\text{ Hz} - 4\text{ Hz} \sim k=2-8$$

$$E_{\alpha\text{1pha}} = 2^8 \times \frac{2^{-32}(2^{-21}-1)}{-\frac{1}{2}}$$

$$E_{\text{delta}} = 2^8 \times \frac{2^{-2}(2^{-7}-1)}{-\frac{1}{2}} \rightarrow \frac{E_{\alpha\text{1pha}}}{E_{\text{delta}}} \approx \boxed{2^{-30}}$$

$$\rightarrow \boxed{\frac{E_{\alpha\text{1pha}}}{E_{\text{delta}}} \Big|_{N=1024} \approx \frac{1}{2} \frac{E_{\alpha\text{1pha}}}{E_{\text{delta}}} \Big|_{N=512}}$$

2-

$$f_k = \frac{K}{N} f_s$$

we want $f_{k+1} - f_k \leq 100 \text{ Hz}$

$$\rightarrow \left(\frac{k+1}{N} - \frac{k}{N} \right) f_s \leq 100 \rightarrow \boxed{\frac{f_s}{N} \leq 100}$$

because of nyquist rate : $f_s \geq 2 \text{ kHz}$

at least $\rightarrow \boxed{f_s = 2 \text{ kHz}} \rightarrow \frac{2K}{N} \leq 100 \rightarrow N \geq 20$

at least $\rightarrow \boxed{N = 20}$

\downarrow
length of signal $\geq 20 \times \frac{1}{2K}$ at least $\rightarrow \boxed{L = 0.01 \text{ s}}$

3 -

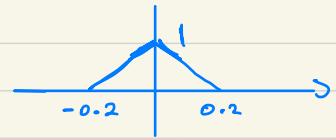
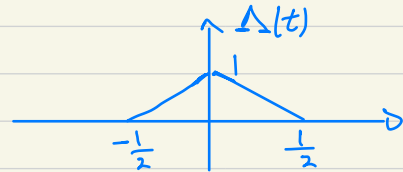
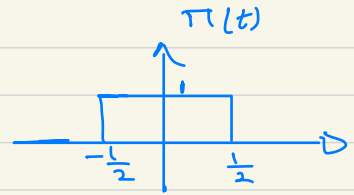
$$\Pi(t) \xrightarrow{F} \text{sinc}(f)$$

$$\Delta\left(\frac{t}{2}\right) = \Pi(t) * \Pi(t)$$

$$\rightarrow \Delta\left(\frac{t}{2}\right) \xrightarrow{F} \text{sinc}^2(f)$$

$$EOG(f) = \frac{1}{5} \text{sinc}^2\left(\frac{f}{5}\right)$$

$$\rightarrow EOG(f) = 0 \text{ for } |f| > 5\text{Hz}$$



So this noise can have effect
on beta and theta bands.

4 -

a)

Peak at 0 Hz : it's DC and it doesn't have any useful information. it could be happen cause to our measuring method or poor electrode connection.

Peak at 50 Hz : it's because of Power line noise .

Peak at 10 Hz : it can show activity of alpha band in brain. it can be happen by SSVEP too.

Peak at 22 Hz : it shows activity of β Band .

this band is active when person is awake and do something with open eyes.

I think this data is collected through an SSVEP

task. the person was looking at monitor and because of

that, we can observe Beta band. also we have a peak at 10 Hz caused by turning off and on of monitor.

b)

as we can see, we don't have so much information below

50 Hz, so we can use a lowpass filter with $f_c = 45 \text{ Hz}$

and then sample the signal with $f_s = 90 \text{ Hz}$.

with doing this we can remove 50 Hz noise and prevent aliasing too. most of the time default sampling

frequency of EEG cap is around 500 Hz, but we can

down sample the signal.

c)

we know if a signal is real then its DFT is symmetric conjugate.

$$\rightarrow X_1[500] = X_1[1500]^* = 2 - 3j$$

$$f_k = \frac{k}{N} f_s \rightarrow f_{500} = \frac{500}{2000} \times 700 = 175 \text{ Hz}$$

$$f_{1500} = \frac{1500}{2000} \times 700 = 525 \text{ Hz}$$

$$\rightarrow X_1(e^{j2\pi \times 175}) = 2 - 3j, \quad X_1(e^{j2\pi \times 525}) = 2 + 3j$$

d)

$$f_k = \frac{K}{N} f_s \rightarrow K = \frac{N f_k}{f_s}$$

$$\xrightarrow{12 \text{ Hz}} K = \frac{1200 \times 12}{500} = 28.8$$

$$\xrightarrow{22 \text{ Hz}} K = \frac{1200 \times 22}{500} = 52.8$$

$$\rightarrow [12 \text{ Hz}, 22 \text{ Hz}] \sim K = [28, 53]$$

5-

a)

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) dx dy = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \alpha xy dx dy = 1$$

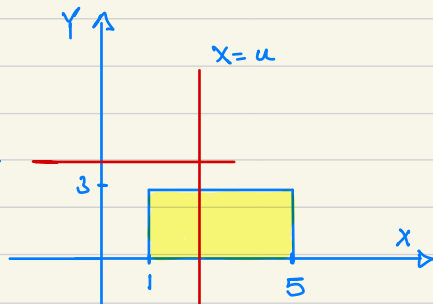
$$\rightarrow \int_0^3 y \int_1^5 x dx dy = \frac{1}{\alpha} \rightarrow 12 \frac{y^2}{2} \Big|_0^3 = \frac{1}{\alpha}$$

$$\rightarrow \boxed{\alpha = \frac{1}{54}} \rightarrow f_{XY}(x, y) = \begin{cases} \frac{xy}{54} & 1 < x < 5, 0 < y < 3 \\ 0 & \text{o.w.} \end{cases}$$

$$F_{XY}(u, v) = P(X \leq u, Y \leq v) = \int_{-\infty}^v \int_{-\infty}^u f_{XY}(u, v) du dv$$

if $x < 1$ or $y < 0 \rightarrow F_{XY}(x, y) = \boxed{0}$

if $x > 5$ and $y > 3 \rightarrow F_{XY}(x, y) = \boxed{1}$



if $1 < x < 5$ and $0 < y < 3$:

$$F_{XY}(x, y) = \int_0^y \int_1^x \frac{uv}{54} du dv = \frac{1}{54} \times \frac{x^2 - 1}{2} \times \frac{y^2}{2} = \boxed{\frac{y^2(x^2 - 1)}{216}}$$

if $x > 5$ and $0 < y < 3$:

$$F_{XY}(x, y) = F_{XY}(5, y) = \frac{24 y^2}{216} = \boxed{\frac{y^2}{9}}$$

if $1 < x < 5$ and $y > 3$:

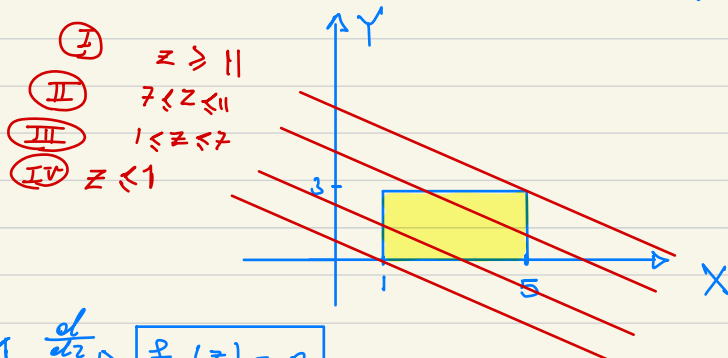
$$F_{XY}(x, y) = F_{XY}(x, 3) = \boxed{\frac{x^2 - 1}{24}}$$

$$F_{XY}(x, y) = \begin{cases} 0, & x < 1 \text{ or } y < 0 \\ 1, & x > 5 \text{ and } y > 3 \\ \frac{y^2(x^2 - 1)}{216}, & 1 < x < 5 \text{ and } 0 < y < 3 \\ \frac{y^2}{9}, & x > 5 \text{ and } 0 < y < 3 \\ \frac{x^2 - 1}{24}, & 1 < x < 5 \text{ and } y > 3 \end{cases}$$

b)

$$Z = X + 2Y$$

$$F_Z(z) = P(Z \leq z) = P(X + 2Y \leq z) = P(Y \leq -\frac{X}{2} + \frac{z}{2})$$



$$\text{I: } F_Z(z) = 1 \xrightarrow{\frac{d}{dz}} \boxed{f_Z(z) = 0}$$

$$\text{IV: } F_Z(z) = 0 \xrightarrow{\frac{d}{dz}} \boxed{f_Z(z) = 0}$$

$$\text{II: } F_Z(z) = 1 - \left(\int_{z-6}^5 \int_{-\frac{x}{2} + \frac{z}{2}}^3 \frac{x}{54} dy dx \right) = 1 - \frac{(z-1)^2 (z^2 + 22z - 3)}{5184}$$

$$\rightarrow \boxed{f_Z(z) = \frac{d}{dz} F_Z(z) = \frac{-z^3 + 183z - 682}{1296}}$$

$$\text{III: } F_Z(z) = \int_1^z \int_0^{-\frac{x}{2} + \frac{z}{2}} \frac{x}{54} dy dx = \frac{(z-1)^3 (z+3)}{5184}$$

$$\rightarrow \boxed{f_Z(z) = \frac{d}{dz} F_Z(z) = \frac{z^3 - 3z + 2}{1296}}$$

$$\rightarrow f_Z(z) = \begin{cases} 0, & z \geq 11 \\ \frac{-z^3 + 183z - 682}{1296}, & 7 \leq z \leq 11 \\ \frac{z^3 - 3z + 2}{1296}, & 1 \leq z \leq 7 \\ 0, & z \leq 1 \end{cases}$$

6-

a)

$$X(t) = A \cos(\omega t + \theta)$$

∞ A and θ are two independent random variable

$$\rightarrow E[X(t)] = E[A] E[\cos(\omega t + \theta)]$$

$$E[A] = \int_0^{13} x \times \frac{1}{13} dx = \frac{13}{2}$$

$$E[\cos(\omega t + \theta)] = \int_0^{2\pi} \frac{1}{2\pi} \cos(\omega t + \theta) d\theta = \frac{1}{2\pi} \int_0^{2\pi} \cos(\omega t) \cos(\theta) - \sin(\omega t) \sin(\theta) d\theta$$

$$= \frac{1}{2\pi} \left(\cos(\omega t) \int_0^{2\pi} \cos(\theta) d\theta - \sin(\omega t) \int_0^{2\pi} \sin(\theta) d\theta \right) = 0$$

$$\rightarrow E[X(t)] = 0 \quad \textcircled{\text{I}}$$

$$R_X(t_1, t_2) = E[X(t_1) X(t_2)] = E[A^2 \cos(\omega t_1 + \theta) \cos(\omega t_2 + \theta)]$$

$$= E\left[A^2 \times \frac{1}{2} (\cos(\omega(t_1 + t_2) + 2\theta) + \cos(\omega(t_1 - t_2)))\right]$$

$$= \frac{1}{2} E[A^2] \left(E[\cos(\omega(t_1 + t_2) + 2\theta)] + E[\cos(\omega(t_1 - t_2))] \right)$$

it's a number $\rightarrow 0 = \cos(\omega(t_1 - t_2))$

$$E[A^2] = \int_0^{13} x^2 \times \frac{1}{13} dx = \frac{x^3}{33} \Big|_0^{13} = \frac{13^2}{3} \rightarrow R_X(t_1, t_2) = \frac{13^2}{6} \cos(\omega(t_1 - t_2)) \quad \textcircled{\text{II}}$$

I, II \rightarrow

E is constant and R_X only depends on $t_1 - t_2$

So this random process is WSS.

b)

$$\begin{aligned}
 \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^{+T} x(t) dt &= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^{+T} A \cos(\omega t + \theta) dt \\
 &= \lim_{T \rightarrow \infty} \frac{1}{2T} A \times \left(\cos(\theta) \int_{-T}^{+T} \cos(\omega t) dt - \sin(\theta) \int_{-T}^{+T} \sin(\omega t) dt \right) \\
 &= \lim_{T \rightarrow \infty} \frac{A}{2T} \times \left(\cos(\theta) (\sin(\omega T) - \sin(-\omega T)) \right) = \lim_{T \rightarrow \infty} \frac{A \cos(\theta) \sin(\omega T)}{T} \\
 &= \boxed{0}
 \end{aligned}$$

from part a we know: $E[x(t)] = 0$

$$\rightarrow \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^{+T} x(t) dt = E[x(t)] \rightarrow \text{it's mean ergodic.}$$

c)

$$\begin{aligned}
 \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^{+T} x(t) x(t-T) dt &= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^{+T} A^2 \cos(\omega t + \theta) \cos(\omega(t-T) + \theta) dt \\
 &= \lim_{T \rightarrow \infty} \frac{A^2}{2T} \left(\int_{-T}^{+T} \cos(2\omega t - \omega T + 2\theta) dt + \int_{-T}^{+T} \cos(\omega T) dt \right) \\
 &\quad \cos(2\theta - \omega T) \int_{-T}^{+T} \cos(2\omega t) dt + \sin(2\theta - \omega T) \int_{-T}^{+T} \sin(2\omega t) dt \\
 &= \cos(2\theta - \omega T) \times \frac{2}{2\omega} \times \sin(2\omega T) = \frac{\cos(2\theta - \omega T) \sin(2\omega T)}{\omega} \\
 &= \lim_{T \rightarrow \infty} \frac{A^2}{2T} \times \frac{\cos(2\theta - \omega T) \sin(2\omega T)}{\omega} + \lim_{T \rightarrow \infty} \frac{A^2}{2T} \times 2T \cos(\omega T) \\
 &= A^2 \cos(\omega T) \quad , \quad \text{from part a we know: } R_x(\tau) = \frac{A^2}{2} \cos(\omega \tau)
 \end{aligned}$$

\rightarrow it's not correlation ergodic.

7-

a)

$$\text{Cum}(Y) = (-j)^k \frac{d^k \ln \Phi_Y(\omega)}{d\omega^k} \Big|_{\omega=0}$$

$$\begin{aligned}\Phi_Y(\omega) &= E[e^{j\omega Y}] = E[e^{j\omega(X+C)}] = e^{j\omega C} E[e^{j\omega X}] \\ &= e^{j\omega C} \Phi_X(\omega)\end{aligned}$$

$$\rightarrow \text{Cum}(Y) = (-j)^k \frac{d^k \ln [e^{j\omega C} \Phi_X(\omega)]}{d\omega^k} \Big|_{\omega=0}$$

$$= \text{Cum}(X) + (-j)^k \frac{d^k \ln e^{j\omega C}}{d\omega^k} \Big|_{\omega=0} = \text{Cum}(X) + (-j)^k \frac{d^k j\omega C}{d\omega^k} \Big|_{\omega=0}$$

$$\rightarrow \text{Cum}(Y) = \begin{cases} \text{Cum}(X) + C, & k=1 \\ \text{Cum}(X), & k>1 \end{cases}$$

b)

the purpose of this part is to find $C_n(Y)$ for

$n \geq 2$. as I mentioned in part a $C_n(Y)$ for $n \geq 2$

is equal to $C_n(X)$

c)

$$\text{cum}(X, Y) = -\mathcal{J} \frac{\partial}{\partial \omega_X \partial \omega_Y} \ln \Phi_{XY}(\omega_X, \omega_Y) \Big|_{\omega_X = \omega_Y = 0}$$

$$\Phi_{X_1+Y_1, X_2+Y_2, \dots, X_m+Y_m}(\omega_1, \omega_2, \dots, \omega_m) = E \left[e^{(X_1+Y_1)\omega_1 + \dots + (X_m+Y_m)\omega_m} \right]$$

cause to independency between $\{X_1, X_2, \dots, X_m\}$ and $\{Y_1, Y_2, \dots, Y_m\}$

we can write:

$$= E \left[e^{X_1\omega_1 + \dots + X_m\omega_m} \right] \cdot E \left[e^{Y_1\omega_1 + \dots + Y_m\omega_m} \right]$$

$$= \Phi_{X_1, X_2, \dots, X_m} \cdot \Phi_{Y_1, Y_2, \dots, Y_m}$$

$$\rightarrow \text{cum}(X_1+Y_1, \dots, X_m+Y_m) = -\mathcal{J} \frac{\partial}{\partial \omega_1 \dots \partial \omega_m} \ln \Phi_{X_1+Y_1, \dots, X_m+Y_m}(\omega_1, \dots, \omega_m) \Big|_{\omega_1 = \dots = \omega_m = 0}$$

$$= -\mathcal{J} \frac{\partial}{\partial \omega_1 \dots \partial \omega_m} \left(\ln \Phi_{X_1, \dots, X_m} + \ln \Phi_{Y_1, \dots, Y_m} \right) \Big|_{\omega_1 = \dots = \omega_m = 0}$$

$$= \boxed{\text{cum}(X_1, X_2, \dots, X_m) + \text{cum}(Y_1, Y_2, \dots, Y_m)}$$

9-

$$a) A = X \Lambda X^{-1} = \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 2 & -1 \end{bmatrix} = \begin{bmatrix} 3 & -1 \\ 2 & 0 \end{bmatrix}$$

$$b) A^3 = \begin{bmatrix} 3 & -1 \\ 2 & 0 \end{bmatrix} \begin{bmatrix} 3 & -1 \\ 2 & 0 \end{bmatrix} \begin{bmatrix} 3 & -1 \\ 2 & 0 \end{bmatrix} = \begin{bmatrix} 7 & -3 \\ 6 & 2 \end{bmatrix} \begin{bmatrix} 3 & -1 \\ 2 & 0 \end{bmatrix} = \begin{bmatrix} 15 & -7 \\ 14 & -6 \end{bmatrix}$$

c)

$$X \Lambda^3 X^{-1} = \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 8 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 2 & -1 \end{bmatrix} = \begin{bmatrix} 1 & 8 \\ 2 & 8 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 2 & -1 \end{bmatrix} = \begin{bmatrix} 15 & -7 \\ 14 & -6 \end{bmatrix}$$

d)

$$A = X \Lambda X^{-1} \rightarrow A^n = \underbrace{(X \Lambda X^{-1})(X \Lambda X^{-1})(X \Lambda X^{-1}) \dots (X \Lambda X^{-1})}_m$$

$$\rightarrow A^n = X \Lambda^n X^{-1}$$

10.

$$R_X = E[XX^H] = \begin{bmatrix} \overline{X_1 X_1^*} & \overline{X_1 X_2^*} & \overline{X_1 X_3^*} \\ \overline{X_2 X_1^*} & \overline{X_2 X_2^*} & \overline{X_2 X_3^*} \\ \overline{X_3 X_1^*} & \overline{X_3 X_2^*} & \overline{X_3 X_3^*} \end{bmatrix} \xrightarrow{\text{real variables}} \begin{bmatrix} \overline{X_1^2} & \overline{X_1 X_2} & \overline{X_1 X_3} \\ \overline{X_2 X_1} & \overline{X_2^2} & \overline{X_2 X_3} \\ \overline{X_3 X_1} & \overline{X_3 X_2} & \overline{X_3^2} \end{bmatrix}$$

$$E[X_1^2] = \text{Var}(X_1) + E[X_1]^2 \underset{\text{independence}}{=} 1 + 256 = 257$$

$$E[X_1 X_2] \overset{\uparrow}{=} E[X_1] E[X_2] = 256$$

$$\text{Like above: } E[X_1 X_3] = E[X_2 X_3] = 256$$

$$E[X_2^2] = 258, E[X_3^2] = 259$$

$$\rightarrow R_X = \begin{bmatrix} 257 & 256 & 256 \\ 256 & 258 & 256 \\ 256 & 256 & 259 \end{bmatrix}$$

$$C_X = E[(X - m_X)(X - m_X)^H] = \begin{bmatrix} \sigma_{X_1}^2 & \sigma_{X_1 X_2} & \sigma_{X_1 X_3} \\ \sigma_{X_2 X_1} & \sigma_{X_2}^2 & \sigma_{X_2 X_3} \\ \sigma_{X_3 X_1} & \sigma_{X_3 X_2} & \sigma_{X_3}^2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

11-

a) $X(t) = e^{w(t)}$, $w(t) \sim N(0, t)$

$$E[X(t)] = E[e^{w(t)}]$$

in each t it's equal to $\overset{\text{moment generating function}}{M_w(1)}$

if $X \sim N(\mu, \sigma^2) \rightarrow M_X(t) = E[e^{xt}] = \int e^{xt} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$

$z = \frac{x-\mu}{\sigma} \rightarrow M_X(t) = e^{\mu t} \int e^{z\sigma t} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2}z^2} \times \sigma dz$

$$= e^{\mu t} \int e^{z\sigma t} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz = e^{\mu t} e^{\frac{\sigma^2 t^2}{2}}$$

$$\rightarrow M_w(1) = e^{\mu} e^{\frac{\sigma^2}{2}} = e^{\frac{t^2}{2}}$$

$$\rightarrow E[e^{w(t)}] = e^{\frac{t^2}{2}}$$

b)

$$\text{var}(X(t)) = \text{var}(e^{w(t)}) = E[e^{2w(t)}] - E[e^{w(t)}]^2$$

$$E[e^{2w(t)}] = M_w(2) = e^{2t^2} \rightarrow \text{var}(X(t)) = e^{2t^2} - e^{\frac{t^2}{2}} = e^{\frac{t^2}{2}} \left(e^{\frac{3t^2}{2}} - 1 \right)$$

12-

if Y_1 and Y_2 are uncorrelated $\rightarrow E[Y_1, Y_2] = E[Y_1] E[Y_2]$

$$E[Y_1] = E[\cos(\alpha)X_1 + \sin(\alpha)X_2] = \cos\alpha E[X_1] + \sin\alpha E[X_2]$$

$$= 0 \rightarrow E[Y_1, Y_2] = 0$$

$$E[Y_1, Y_2] = E[-\cos\alpha \sin\alpha X_1^2 + \sin\alpha \cos\alpha X_2^2 + (\cos^2\alpha - \sin^2\alpha)X_1X_2]$$

$$= E[\sin\alpha \cos\alpha X_2^2] - E[\cos\alpha \sin\alpha X_1^2] + E[\cos 2\alpha X_1X_2]$$

$$= \sin\alpha \cos\alpha (E[X_2^2] - E[X_1^2]) + \cos 2\alpha E[X_1X_2]$$

$$= \sin\alpha \cos\alpha (\sigma_2^2 - \sigma_1^2) + \cos 2\alpha \rho \sigma_1 \sigma_2$$

$$= \frac{\sin 2\alpha}{2} (\sigma_2^2 - \sigma_1^2) + \cos 2\alpha \rho \sigma_1 \sigma_2 = 0$$

$$\div \cos 2\alpha \rightarrow \frac{1}{2} \tan(2\alpha) (\sigma_2^2 - \sigma_1^2) + \rho \sigma_1 \sigma_2 = 0$$

$$\rightarrow \tan(2\alpha) = \frac{-2\rho\sigma_1\sigma_2}{\sigma_2^2 - \sigma_1^2} = \frac{2\rho\sigma_1\sigma_2}{\sigma_1^2 - \sigma_2^2}$$

$$\rightarrow \boxed{\alpha = \frac{1}{2} \tan^{-1} \left(\frac{2\rho\sigma_1\sigma_2}{\sigma_1^2 - \sigma_2^2} \right)}$$

13 -

$$\begin{aligned} R_{yy}(t) &= E[Y(t)Y(t-\tau)] = E[(X(at+b) + X(at-b))(X(at-a\tau+b) + X(at-a\tau-b))] \\ &= E[X(at+b)X(at-a\tau+b)] + E[X(at+b)X(at-a\tau-b)] \\ &\quad + E[X(at-b)X(at-a\tau+b)] + E[X(at-b)X(at-a\tau-b)] \\ &= R_{XX}(a\tau) + R_{XX}(a\tau+2b) + R_{XX}(a\tau-2b) + R_{XX}(-a\tau) \\ &= \boxed{2R_{XX}(a\tau) + R_{XX}(a\tau+2b) + R_{XX}(a\tau-2b)} \end{aligned}$$

14 -

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

$$E[Z(t)] = E[X(t)] + E[Y(t)] = m_X + m_Y \rightarrow \text{Constant}$$

$$E[Z(t_1)Z(t_2)] = E[(X(t_1) + Y(t_1))(X(t_2) + Y(t_2))]$$

$$= \underbrace{E[X(t_1)X(t_2)]}_{R_{XX}(\tau)} + \underbrace{E[X(t_1)Y(t_2)]}_{R_{XY}(\tau)} + \underbrace{E[Y(t_1)X(t_2)]}_{R_{YX}(\tau)} + \underbrace{E[Y(t_1)Y(t_2)]}_{R_{YY}(\tau)}$$

→ $R_Z(t_1, t_2)$ only depends on $\tau = t_1 - t_2$

→ $Z(t)$ is WSS