

המחלקה להנדסת חשמל ואלקטרוניקה

תאריך הבחינה: 02.02.2023

שעות הבחינה: 9:00-12:00

מבוא לאותות אקראים

מועד א'

ד"ר דימה בוחבסקי, מר טל פאר

תשפ"ב סטודנט א'

חומר עזר - דף נוסחים אישי (עמוד אחד), מחשבון
הוראות מיוחדות:

- ◻ השאלה כולל שאלות ללא בחירה, סך הכל של 110 נקודות.
- ◻ סעיפים הם בעלי ניקוד זהה, אלא אם צוין אחרת.
- ◻ יש לציין באופן מלא וברור את שלבי הפתרון. תשובה ללא הסבר לא תתקבלנה.
- ◻ במקומות בו נדרש חישוב ממספרי, יש קודם לרשום את הנוסחה, ורק לאחר מכן!
- ◻ יש לציין ייחדות למספרים, ובמידה וקייםות!
- ◻ כל הشرطוטים יהיו גדולים, ברורים, עם סימון ציריים!
- ◻ אין חובה להגיע לערך ממספרי של הפונקציה (x, Q) , במידה ומופיעה בתשובה.

השאלון כולל 12 דפים (כולל דף זה)

בהצלחה !

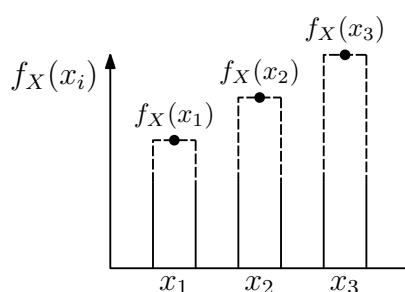
1 היסטוגרמה (30 נק')

נתונות תוצאות של ניסוי אקראי

$$[16, 98, 96, 49, 81, 14, 43, 92, 80, 96]$$

מעוניינים להציג את ה-PDF של התוצאות בצורה גרפית ע"י היסטוגרמה בעלת 3 נקודות חישוב, כמפורט באירור להלן (באירור מדובר בגדים **שרירותיים** של $f_X(x_1), f_X(x_2), f_X(x_3)$)

יש להשלים את הערכים של $x_1, x_2, x_3, f_X(x_1), f_X(x_2), f_X(x_3)$



2 קשר בין תהליכיים (80 נק')

נתון אות אקראי $w[n]$ בעל התפלגות כלשהי ובעל $E[w[n]] = \mu_w = \frac{1}{2}$, $R_w[k] = \exp\left(-\frac{|k|}{10}\right)$ (פרט לסעיף ד', שם ההגדרה היא אחרת).

(א) נתונים זוג אותן

$$(1) \quad x[n] = w[n] \cos(\omega_0 n)$$

$$y[n] = w[n] \cos([\omega_0 + \Delta\omega] n)$$

יש לחשב $E[y[n]]$, $\text{Var}[y[n]]$. האם אותן סטציאונריים במשותף?

(ב) נתונים זוג אותן, כאשר $w[n] \sim U[-\pi, \pi]$ בלווית לילי ב-

$$(2) \quad x[n] = w[n] \cos(\omega_0 n + \theta)$$

$$y[n] = w[n] \cos([\omega_0 + \Delta\omega] n + \theta)$$

יש לחשב $E[y[n]]$, $\text{Var}[y[n]]$. האם אותן סטציאונריים במשותף?

(ג) נתונים זוג אותן, כאשר $\omega \sim U[-1, 1]$, $\theta \sim U[-\pi, \pi]$ בלווית תלויים ביניהם ו- $w[n]$

$$(3) \quad x[n] = w[n] \cos(\omega_0 n + \theta)$$

$$y[n] = w[n] \cos([\omega_0 + \Delta\omega] n)$$

יש לחשב $E[y[n]]$. האם אותן סטציאונריים במשותף?

(ד) נתונים זוג אוטות, כאשר $\mathbf{w}[n] \sim U[-1, 1]$ הוא רעש לבן בעל התפלגות **אחדית**, $\theta \sim \text{Uniform}(-\pi, \pi)$. זוית θ בLATI תלויה ב- $\mathbf{w}[n]$.

$$(4) \quad \begin{aligned} \mathbf{x}[n] &= \mathbf{w}[n] \cos(\omega_0 n + \theta) \\ \mathbf{y}[n] &= \cos(\omega_0 n + \mathbf{w}[n]) \end{aligned}$$

יש לחשב $E[\mathbf{y}[n]]$. האם אוטות \mathbf{x}, \mathbf{y} הם סטציאונריים במשמעות?

Random Processes – Formulas

1 Distributions

1.1 Continuous

	Notation	PDF	CDF	$E[X]$	$\text{Var}[X]$
Uniform	$U[a,b]$	$\begin{cases} \frac{1}{b-a} & a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$	$\begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & a \leq x \leq b \\ 1 & b < x \end{cases}$	$\frac{a+b}{2}$	$\frac{(b-a)^2}{12}$
Normal	$N(\mu, \sigma^2)$	$\frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right]$	$\Phi(x)$	μ	σ^2
Exponential	$Exp(\lambda)$	$\lambda \exp(-\lambda x), x \geq 0$	$1 - \exp(-\lambda x)$	$1/\lambda$	$1/\lambda^2$

1.1.1 Q-function

Given $Y \sim N(\mu, \sigma^2)$

$$\frac{Y - \mu}{\sigma} \sim N(0, 1) \quad (1)$$

$$p(Y > y) = Q\left(\frac{y - \mu}{\sigma}\right) \quad (2)$$

$$Q(x) = 1 - \Phi(x) \quad (3)$$

$$Q(-x) = 1 - Q(x) \quad (4)$$

$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^\infty \exp\left(-\frac{s^2}{2}\right) ds \quad (5)$$

1.2 Discrete

	Notation	PDF	CDF	$E[X]$	$\text{Var}[X]$
Bernoulli	$Ber(p)$	$\begin{cases} 1-p & k=0 \\ p & k=1 \end{cases}$	$\begin{cases} 0 & x < 0 \\ 1-p & 0 \leq x < 1 \\ 1 & 1 \leq x \end{cases}$	p	$p(1-p)$
Binomial	$Bin(n,p)$	$\binom{n}{k} p^k (1-p)^{n-k}$		np	$np(1-p)$

2 Random Variables

Definitions:

$$F_X(x) = p(X \leq x) \quad (6)$$

$$f_X(x) = \frac{\partial F_X(x)}{\partial x} \geq 0 \quad (7)$$

$$F_X(x) = \int_{-\infty}^x f_X(p) dp \quad (8)$$

$$p(a < X \leq b) = F_X(b) - F_X(a) \quad (9)$$

$$f_X(x) \geq 0 \quad (10)$$

$$\int_{-\infty}^{\infty} f_X(x) dx = 1 \quad (11)$$

$$p_X[x_k] = \Pr[X = x_k] \quad (12)$$

$$0 \leq p_X[x_i] \leq 1 \quad \forall i \quad (13)$$

$$\sum_i p_X[x_i] = 1 \quad (14)$$

$$F_X(x) = \Pr(X \leq x), \quad x \in \mathbb{R} \quad (15)$$

$$F_X(x) = \sum_{k: x_k \leq x} p_X[x_k] \quad (16)$$

Expectation:

$$E[X] = \begin{cases} \int_{-\infty}^{\infty} x f_X(x) dx \\ \sum_i x_i p_X[x_i] \end{cases} \quad (17a)$$

$$E[g(X)] = \begin{cases} \int_{-\infty}^{\infty} g(x) f_X(x) dx \\ \sum_i g(x_i) p_X[x_i] \end{cases} \quad (17b)$$

$$E[aX + b] = aE[X] + b \quad (17c)$$

Variance:

$$\begin{aligned} \text{Var}[X] &= E[(X - E[X])^2] \\ &= E[X^2] - E^2[X] \end{aligned} \quad (18a)$$

$$\text{Var}[aX + b] = a^2 \text{Var}[X] \quad (18b)$$

$$\text{Var}[b] = 0 \quad (18c)$$

2.1 Numerical calculation

$$E[X] = \frac{1}{N} \sum_{i=1}^N x_i \quad (19)$$

$$\text{Var}[X] = \frac{1}{N} \sum_{i=1}^N (x_i - E[X])^2 \quad (20)$$

2.2 Histogram

$$p_X[x_i] \approx \frac{n_i}{N} \quad i = 1, \dots, k \quad (21)$$

$$f_X(x_i) \approx \frac{n_i}{N} \cdot \frac{1}{\Delta x} \quad i = 1, \dots, k \quad (22)$$

3 Two Random Variables

3.1 Joint Distributions

Definitions:

$$F_{XY}(x, y) = p(X \leq x, Y \leq y) \quad (23a)$$

$$f_{XY}(x, y) = \frac{\partial^2 F_{XY}(x, y)}{\partial x \partial y} \geq 0 \quad (23b)$$

$$F_{XY}(x, y) = \int_{-\infty}^x \int_{-\infty}^y f_{XY}(s, p) dp ds \quad (23c)$$

$$p[x_j, y_k] = p(X = x_j, Y = y_k) \quad (24a)$$

$$F_{XY}(x, y) = p(X \leq x_j, Y \leq y_k) \quad (24b)$$

Expectation:

$$E[g(X, Y)] = \begin{cases} \iint g(x, y) f_{XY}(x, y) dx dy \\ \sum_i \sum_k g(x_i, y_k) p_{XY}[x_i, y_k] \end{cases} \quad (25a)$$

$$E[aX + bY] = aE[X] + bE[Y] \quad (25b)$$

For **independent** random variables:

$$f_{XY}(x, y) = f_X(x)f_Y(y) \quad (26a)$$

$$p_{XY}[x_k, y_j] = p_X[x_k]p_Y[y_j] \quad (26b)$$

$$F_{XY}(x, y) = F_X(x)F_Y(y) \quad (26c)$$

$$(26d)$$

Independent random variables properties:

$$E[XY] = E[X]E[Y] \quad (27a)$$

$$E[g_1(X)g_2(Y)] = E[g_1(X)]E[g_2(Y)] \quad (27b)$$

$$\text{Var}[aX + bY] = a^2 \text{Var}[X] + b^2 \text{Var}[Y] \quad (27c)$$

Marginal distribution:

$$f_X(x) = \int_{-\infty}^{\infty} f_{XY}(x, y) dy \quad (28a)$$

$$p_X[x_k] = \sum_j p_{XY}[x_k, y_j] \quad (28b)$$

$$F_X(x) = F_{XY}(x, \infty) \quad (28c)$$

$$F_Y(y) = F_{XY}(\infty, y) \quad (28d)$$

3.2 Correlation, Covariance & Correlation Coefficient

- For two jointly-distributed random variables X and Y , covariance is given by

$$\begin{aligned} \text{Cov}[X, Y] &= E[(X - E[X])(Y - E[Y])] \\ &= E[XY] - E[X]E[Y] \end{aligned} \quad (29)$$

Main covariance properties are:

$$\text{Cov}[X, X] = \text{Var}[X] \quad (30a)$$

$$\text{Cov}[X, Y] = \text{Cov}[Y, X] \quad (30b)$$

$$\text{Cov}[X, a] = 0 \quad (30c)$$

$$\text{Cov}[aX, bY] = ab \text{Cov}[X, Y] \quad (30d)$$

$$\text{Cov}[X, Y] = \text{Cov}[X + a, Y + b] \quad (30e)$$

$$\text{Var}[X \pm Y] = \text{Var}[X] + \text{Var}[Y] \pm 2 \text{Cov}[X, Y] \quad (30f)$$

- Correlation coefficient (also termed as Pearson product-moment correlation coefficient) is given by

$$\rho_{XY} = \frac{\text{Cov}[X, Y]}{\sqrt{\text{Var}[X] \text{Var}[Y]}} \quad (31)$$

such that $|\rho_{XY}| \leq 1$.

3.3 MMSE Linear Prediction

Mean square error (MSE) of predictor \hat{Y} is given by

$$mse = E[(Y - \hat{Y})^2] \quad (32)$$

Linear prediction of $\hat{Y} = ax + b$ for $X = x$ is

$$\hat{Y} = E[Y] + \frac{\text{Cov}[X, Y]}{\text{Var}[X]} (x - E[X]) \quad (33)$$

and

$$mse_{min} = E \left[(Y - (aX + b))^2 \right] = \text{Var}[Y] (1 - \rho_{XY}^2) \quad (34)$$

When X, Y are jointly Gaussian, this prediction is optimal among **all** possible predictors

3.4 Relations

- When X and Y are *orthogonal*, $E[XY] = 0$.
- When X and Y are *uncorrelated*, $\text{Cov}[X, Y] = \rho_{XY} = 0$.

- When X and Y are *independent*, they are also uncorrelated (see also Eqs. 26).
- When X and Y are *jointly Gaussian* and uncorrelated $\Rightarrow X$ and Y are independent.

3.5 Bi-variate Normal Distribution

Joint Gaussian distribution of X_1 and X_2

$$\begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \sim N \left(\begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}, C_X \right) \quad (35)$$

with covariance matrix

$$C_X = \begin{bmatrix} \text{Cov}[X_1, X_1] & \text{Cov}[X_1, X_2] \\ \text{Cov}[X_2, X_1] & \text{Cov}[X_2, X_2] \end{bmatrix} \quad (36)$$

Important properties:

- Sum of independent Gaussian variables is a Gaussian variable.
- Random vector $[X_1, \dots, X_n]$ is **jointly Gaussian** distributed, iff (if and only if) for all possible real vectors $\mathbf{a} = (a_1, \dots, a_n)^T$ linear combination $Y = a_1X_1 + \dots + a_nX_n$ is Gaussian distributed.
- If jointly distributed Gaussian random variables are *uncorrelated*, they are also *independent*

4 Random Processes – General Properties

- PDF & CDF

$$F_{\mathbf{x}}(x; t) = p(\mathbf{x}(t) \leq x) \quad (37a)$$

$$f_{\mathbf{x}}(x; t) = \frac{\partial}{\partial x} F_{\mathbf{x}}(x; t) \quad (37b)$$

$$p_{\mathbf{x}}[x_k; n] = p(\mathbf{x}[n] = x_k) \quad (37c)$$

- Average:

$$E[\mathbf{x}(t)] = \int_{-\infty}^{\infty} x f_{\mathbf{x}}(x; t) dx \quad (38a)$$

$$E[\mathbf{x}[n]] = \sum_i x_i p_{\mathbf{x}}[x_k; n] \quad (38b)$$

- Variance:

$$\text{Var}[\mathbf{x}(t)] = E[\mathbf{x}^2(t)] - E^2[\mathbf{x}(t)] = \sigma_{\mathbf{x}}(t) \quad (39a)$$

$$\text{Var}[\mathbf{x}[n]] = E[\mathbf{x}^2[n]] - E^2[\mathbf{x}[n]] = \sigma_{\mathbf{x}}[n] \quad (39b)$$

- Auto-correlation

$$R_{\mathbf{x}}(t_1, t_2) = E[\mathbf{x}(t_1)\mathbf{x}(t_2)] \quad (40a)$$

$$R_{\mathbf{x}}(t, t + \tau) = E[\mathbf{x}(t)\mathbf{x}(t + \tau)] \quad (40b)$$

$$R_{\mathbf{x}}(t_1, t_2) = R_{\mathbf{x}}(t_2, t_1) \quad (40c)$$

$$R_{\mathbf{x}}(t, t) = E[\mathbf{x}^2(t)] \quad (40d)$$

$$R_{\mathbf{x}}[n_1, n_2] = E[\mathbf{x}[n_1]\mathbf{x}[n_2]] \quad (40e)$$

$$R_{\mathbf{x}}[n, n] = E[\mathbf{x}^2[n]] \quad (40f)$$

- Auto-covariance

$$C_{\mathbf{x}}(t_1, t_2) = E \left[\{\mathbf{x}(t_1) - E[\mathbf{x}(t_1)]\} \{\mathbf{x}(t_2) - E[\mathbf{x}(t_2)]\} \right] \quad (41)$$

$$= R_{\mathbf{x}}(t_1, t_2) - E[\mathbf{x}(t_1)]E[\mathbf{x}(t_2)] \quad (42)$$

$$C_{\mathbf{x}}[n_1, n_2] = E \left[\{\mathbf{x}[n_1] - E[\mathbf{x}[n_1]]\} \{\mathbf{x}[n_2] - E[\mathbf{x}[n_2]]\} \right] \quad (43)$$

$$= R_{\mathbf{x}}[n_1, n_2] - E[\mathbf{x}[n_1]] E[\mathbf{x}[n_2]] \quad (44)$$

$$C_{\mathbf{x}}(t, t) = \text{Var}[\mathbf{x}(t)] \quad (45a)$$

$$C_{\mathbf{x}}[n, n] = \text{Var}[\mathbf{x}[n]] \quad (45b)$$

- Correlation Coefficient

$$\rho_{\mathbf{x}}(t_1, t_2) = \frac{C_{\mathbf{x}}(t_1, t_2)}{\sqrt{C_{\mathbf{x}}(t_1, t_1)C_{\mathbf{x}}(t_2, t_2)}} \quad (46a)$$

$$|\rho_{\mathbf{x}}(t_1, t_2)| \leq 1 \quad (46b)$$

- When $\mathbf{x}(t_1)$ and $\mathbf{x}(t_2)$ are *orthogonal*, $R_{\mathbf{x}}(t_1, t_2) = 0$.

- When $\mathbf{x}(t_1)$ and $\mathbf{x}(t_2)$ are *uncorrelated*, $C_{\mathbf{x}}(t_1, t_2) = \rho_{\mathbf{x}}(t_1, t_2) = 0$.

- When $\mathbf{x}(t_1)$ and $\mathbf{x}(t_2)$ are *independent*, $R_{\mathbf{x}}(t_1, t_2) = E[\mathbf{x}(t_1)]E[\mathbf{x}(t_2)]$.

5 Wide-Sense Stationary (WSS) Process

Definition:

$$E[\mathbf{x}(t)] = E[\mathbf{x}(0)] = \mu_{\mathbf{x}} = \text{const} \quad (47a)$$

$$R_{\mathbf{x}}(t_1, t_2) = R_{\mathbf{x}}(\tau = |t_2 - t_1|), \quad \forall t_1, t_2 \quad (47b)$$

$$E[\mathbf{x}[n]] = E[\mathbf{x}[0]] = \mu_{\mathbf{x}} = \text{const} \quad (47c)$$

$$R_{\mathbf{x}}[n_1, n_2] = R_{\mathbf{x}}(k = |n_2 - n_1|), \quad \forall n_1, n_2 \quad (47d)$$

- Auto-correlation

$$R_{\mathbf{x}}(\tau) = E[\mathbf{x}(t)\mathbf{x}(t + \tau)] \quad (48a)$$

$$R_{\mathbf{x}}[k] = E[\mathbf{x}[n]\mathbf{x}[n + k]] \quad (48b)$$

Properties:

$$R_{\mathbf{x}}(-\tau) = R_{\mathbf{x}}(\tau) \quad (49a)$$

$$R_{\mathbf{x}}(0) = E[|\mathbf{x}(0)|^2] = E[|\mathbf{x}(t)|^2] \quad (49b)$$

$$\text{Var}[\mathbf{x}(t)] = C_{\mathbf{x}}(0) = \sigma_{\mathbf{x}}^2 \quad (49c)$$

$$R_{\mathbf{x}}(0) \geq |R_{\mathbf{x}}(\tau)| \quad (49d)$$

- Auto-covariance

$$C_{\mathbf{x}}(\tau) = R_{\mathbf{x}}(\tau) - \mu_{\mathbf{x}}^2 \quad (50a)$$

$$C_{\mathbf{x}}[k] = R_{\mathbf{x}}[k] - \mu_{\mathbf{x}}^2 \quad (50b)$$

- Correlation Coefficient

$$\rho_{\mathbf{x}}(\tau) = \frac{C_{\mathbf{x}}(\tau)}{C_{\mathbf{x}}(0)} \quad (51a)$$

$$\rho_{\mathbf{x}}[k] = \frac{C_{\mathbf{x}}[k]}{C_{\mathbf{x}}[0]} \quad (51b)$$

5.1 Power Spectral Density (PSD)

$$S_{\mathbf{x}}(F) = \mathcal{F}\{R_{\mathbf{x}}(\tau)\} = \int_{-\infty}^{\infty} R_{\mathbf{x}}(\tau) \exp(-j2\pi F\tau) d\tau \quad (52a)$$

$$R_{\mathbf{x}}(\tau) = \mathcal{F}^{-1}\{S_{\mathbf{x}}(F)\} = \int_{-\infty}^{\infty} S_{\mathbf{x}}(f) \exp(j2\pi F\tau) dF \quad (52b)$$

$$S_{\mathbf{x}}(f) = \text{DTFT}\{R_{\mathbf{x}}[k]\} = \sum_{k=-\infty}^{\infty} R_{\mathbf{x}}[k] e^{-j2\pi fk} \quad (52c)$$

Properties:

$$S_{\mathbf{x}}(F) = S_{\mathbf{x}}(-F) \quad (53a)$$

$$S_{\mathbf{x}}(F) \geq 0, \quad \forall F \quad (53b)$$

$$S_{\mathbf{x}}(F) \in \mathbb{R} \quad (53c)$$

$$S_{\mathbf{x}}(f) = S_{\mathbf{x}}(-f) \quad (53d)$$

$$S_{\mathbf{x}}(f) \geq 0, \quad \forall f \quad (53e)$$

$$S_{\mathbf{x}}(f) \in \mathbb{R} \quad (53f)$$

$$S_{\mathbf{x}}(f) = S_{\mathbf{x}}(f + 1) \quad (53g)$$

Average power

$$P_{\mathbf{x}} = E[\mathbf{x}^2(t)] = R_{\mathbf{x}}(0) = \int_{-\infty}^{\infty} S_{\mathbf{x}}(F) dF \quad (54a)$$

$$P_{\mathbf{x}} = E[\mathbf{x}^2[n]] = R_{\mathbf{x}}[0] = \int_{-\frac{1}{2}}^{\frac{1}{2}} S_{\mathbf{x}}(f) df \quad (54b)$$

5.2 White Noise & White Gaussian Noise (WGN) Process

White noise process is SSS (WSS) process that is characterized by

$$R_{\mathbf{n}}(\tau) = \sigma^2 \delta(\tau) \quad (55a)$$

$$S_{\mathbf{n}}(F) = \sigma^2 \quad \forall F \quad (55b)$$

For WGN process, $\mathbf{n}(t) \sim N(0, \sigma^2)$,

$$R_{\mathbf{n}}(\tau) = \frac{N_0}{2} \delta(\tau) \quad (56a)$$

$$S_{\mathbf{n}}(F) = \frac{N_0}{2} \quad \forall F \quad (56b)$$

5.3 Relation Between Covariance Matrix & Auto-covariance

Given WSS process $\mathbf{x}(t)$, the corresponding correlation matrix of $\mathbf{X} = [\mathbf{x}(t_1), \dots, \mathbf{x}(t_N)]^T$ is given by

$$R_{\mathbf{X}} = E[\mathbf{X}\mathbf{X}^T] \quad (57)$$

$$R_{\mathbf{X}}(i, j) = E[X_i X_j] = R_{\mathbf{x}}(|t_i - t_j|) \quad (58)$$

6 Cross-Signal

- Cross-correlation

$$R_{\mathbf{xy}}(t_1, t_2) = E[\mathbf{x}(t_1)\mathbf{y}(t_2)] \quad (59)$$

- Cross-covariance

$$C_{\mathbf{xy}}(t_1, t_2) = R_{\mathbf{xy}}(t_1, t_2) - E[\mathbf{x}(t_1)]E[\mathbf{y}(t_2)] \quad (60)$$

- Correlation Coefficient

$$\rho_{\mathbf{xy}}(t_1, t_2) = \frac{C_{\mathbf{xy}}(t_1, t_2)}{\sqrt{C_{\mathbf{x}}(t_1, t_1)C_{\mathbf{y}}(t_2, t_2)}} \quad (61)$$

6.1 WSS Cross-signal

- $\mathbf{x}(t), \mathbf{y}(t)$ are jointly WSS, if $\mathbf{x}(t)$ and $\mathbf{y}(t)$ each of them is WSS and

$$R_{\mathbf{xy}}(\tau) = E[\mathbf{x}(t)\mathbf{y}(t + \tau)] \quad (62)$$

- When $\mathbf{x}(t)$ and $\mathbf{y}(t + \tau)$ are *uncorrelated jointly WSS*, $C_{\mathbf{xy}}(\tau) = 0$.

Properties

$$R_{\mathbf{xy}}(\tau) = R_{\mathbf{yx}}(-\tau) \quad (63a)$$

$$|R_{\mathbf{xy}}(\tau)| \leq \sqrt{R_{\mathbf{x}}(0)R_{\mathbf{y}}(0)} \quad (63b)$$

$$|R_{\mathbf{xy}}(\tau)| \leq \frac{1}{2} [R_{\mathbf{x}}(0) + R_{\mathbf{y}}(0)] \quad (63c)$$

- Cross-covariance

$$C_{\mathbf{xy}}(\tau) = R_{\mathbf{xy}}(\tau) - \mu_{\mathbf{x}}\mu_{\mathbf{y}} \quad (64)$$

- Cross-PSD

$$S_{\mathbf{xy}}(f) = \mathcal{F}\{R_{\mathbf{xy}}(\tau)\} \quad (65)$$

Properties

$$S_{\mathbf{xy}}(f) = S_{\mathbf{yx}}(-f) = S_{\mathbf{xy}}^*(-f) \quad (66)$$

Correlation coefficient

$$\rho_{\mathbf{xy}}(\tau) = \frac{C_{\mathbf{xy}}(\tau)}{C_{\mathbf{xy}}(0)} \quad (67)$$

- Coherence

$$\gamma_{\mathbf{xy}}(f) = \frac{S_{\mathbf{xy}}(f)}{\sqrt{S_{\mathbf{x}}(f)S_{\mathbf{y}}(f)}} \quad (68)$$

7 LTI and WSS Random Process

Output of LTI system with impulse response $h(t)$ and random process $x(t)$,

$$y(t) = x(t) * h(t) \quad (69)$$

Average

$$m_{\mathbf{y}} = m_{\mathbf{x}} \int_{-\infty}^{\infty} h(s)ds = m_{\mathbf{x}} H(F=0) \quad (70)$$

Cross-correlation & cross-covariance:

$$R_{\mathbf{xy}}(\tau) = R_{\mathbf{x}}(\tau) * h(\tau) \quad (71a)$$

$$C_{\mathbf{xy}}(\tau) = C_{\mathbf{x}}(\tau) * h(\tau) \quad (71b)$$

$$R_{\mathbf{yx}}(\tau) = R_{\mathbf{x}}(\tau) * h(-\tau) \quad (71c)$$

$$C_{\mathbf{yx}}(\tau) = C_{\mathbf{x}}(\tau) * h(-\tau) \quad (71d)$$

$$R_{\mathbf{y}}(\tau) = R_{\mathbf{x}}(\tau) * h(\tau) * h(-\tau) \quad (71e)$$

$$C_{\mathbf{y}}(\tau) = C_{\mathbf{x}}(\tau) * h(\tau) * h(-\tau) \quad (71f)$$

<p>Power-Spectral Density (PSD) & Cross-PSD: Given frequency response $H(F) = \mathcal{F}\{h(\tau)\}, H^*(F) = \mathcal{F}\{h(-\tau)\}$</p>	$S_{\mathbf{y}\mathbf{z}}(z) = S_{\mathbf{x}}(z)H_1(1/z)H_2(z)$ (77c)
$S_{\mathbf{xy}}(F) = S_{\mathbf{x}}(F)H(F)$ (72a)	Power of the process:
$S_{\mathbf{yx}}(F) = S_{\mathbf{x}}(F)H^*(F)$ (72b)	$P_x = R_{\mathbf{x}}[0] = \int_{-1/2}^{1/2} S_{\mathbf{x}}(f) df$ (78a)
$S_{\mathbf{y}}(F) = S_{\mathbf{x}}(F) H(F) ^2$ (72c)	$P_y = R_{\mathbf{y}}[0] = \int_{-1/2}^{1/2} S_{\mathbf{x}}(f) H(f) ^2 df$ (78b)
Power of the process:	Average
$P_x = R_{\mathbf{x}}(0) = \int_{-\infty}^{\infty} S_{\mathbf{x}}(F) dF$ (73a)	$\mu_{\mathbf{y}} = \mu_{\mathbf{x}} \sum_m h[m]$ (79)
$P_y = R_{\mathbf{y}}(0) = \int_{-\infty}^{\infty} S_{\mathbf{x}}(F) H(F) ^2 dF$ (73b)	For white Gaussian noise input
Same process passes two different systems	$\text{Var}[\mathbf{y}[n]] = \text{Var}[\mathbf{x}[n]] \sum_m h^2[m]$ (80)
$R_{\mathbf{yz}}(\tau) = R_{\mathbf{x}}(\tau) * h_1(-\tau) * h_2(\tau)$ (74) $S_{\mathbf{yz}}(F) = S_{\mathbf{x}}(F)H_1^*(F)H_2(F)$ (75)	<h2>7.2 Gaussian Process</h2> <p>A Gaussian process $\mathbf{x}(t)$ a random process that for $\forall k > 0$ and for all times t_1, \dots, t_k, the set of random variable $\mathbf{x}(t_1), \dots, \mathbf{x}(t_k)$ is jointly Gaussian.</p> <p>Properties:</p> <ul style="list-style-type: none"> • WSS Gaussian process is SSS. • Gaussian process $\mathbf{x}(t)$ that passes through LTI system, $\mathbf{y}(t) = h(t) * \mathbf{x}(t)$, is also Gaussian process that may be described by the change of expectation and auto-correlation,
$H(z) = \mathcal{Z}\{h[n]\} = \frac{B(z)}{A(z)}$ $\mathcal{Z}\{h[n] * h[-n]\} = \frac{B(z)B(z^{-1})}{A(z)A(z^{-1})}$ $S_{\mathbf{x}}(z) = \mathcal{Z}\{R_{\mathbf{x}}[n]\}$ $h[k] * h[-k] = \sum_m h[m]h[m+k]$	$E[\mathbf{y}(t)] = E[\mathbf{x}(t)] \int_{-\infty}^{\infty} h(s)ds$ (81a) $= E[\mathbf{x}(t)]H(0), \quad H(F) = \mathcal{F}\{h(t)\}$ $C_{\mathbf{y}}(\tau) = C_{\mathbf{x}}(\tau) * h(\tau) * h(-\tau)$ (81b)
PSD	<ul style="list-style-type: none"> • The resulting autocorrelation may be used for producing the correspondent covariance matrix $C_{\mathbf{Y}}$ of a multivariate Gaussian $\mathbf{Y} = [\mathbf{y}(t_1), \dots, \mathbf{y}(t_N)]^T$
$S_{\mathbf{xy}}(z) = S_{\mathbf{x}}(z)H(z)$ (76a) $S_{\mathbf{yx}}(z) = S_{\mathbf{x}}(z)H(z^{-1})$ (76b) $S_{\mathbf{y}}(z) = S_{\mathbf{x}}(z)H(z)H(z^{-1})$ (76c)	
Two different systems	
$R_{\mathbf{yz}}[k] = R_{\mathbf{x}}[k] * h_1[-k] * h_2[k]$ (77a) $S_{\mathbf{yz}}(f) = S_{\mathbf{x}}(f)H_1^*(f)H_2(f)$ (77b)	

7.3 Linear Prediction

Given N samples of process $\mathbf{x}[n]$, and predictor

$$\hat{\mathbf{x}}[n+1] = \sum_{i=1}^N a_i \mathbf{x}[n-i+1], \quad (82)$$

the mean-square error is given by

$$\begin{aligned} mse &= E \left[(\mathbf{x}[n+1] - \hat{\mathbf{x}}[n+1])^2 \right] \\ &= E \left[(\mathbf{x}[n+1] - a_0 \mathbf{x}[n] - a_1 \mathbf{x}[n-1] - \dots - a_N \mathbf{x}[n-N])^2 \right] \end{aligned} \quad (83)$$

and the values of a_i are given by a solution of

$$\begin{bmatrix} R_{\mathbf{x}}[0] & R_{\mathbf{x}}[1] & \cdots & R_{\mathbf{x}}[N-1] \\ R_{\mathbf{x}}[1] & R_{\mathbf{x}}[0] & \cdots & R_{\mathbf{x}}[N-2] \\ \vdots & \vdots & \ddots & \vdots \\ R_{\mathbf{x}}[N-1] & R_{\mathbf{x}}[N-2] & \cdots & R_{\mathbf{x}}[0] \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_N \end{bmatrix} = \begin{bmatrix} R_{\mathbf{x}}[1] \\ R_{\mathbf{x}}[2] \\ \vdots \\ R_{\mathbf{x}}[N] \end{bmatrix} \quad (84)$$

and the resulting minimum MSE is

$$mse_{min} = R_{\mathbf{x}}[0] - \sum_{i=1}^N a_i R_{\mathbf{x}}[i] \quad (85)$$

7.4 Match Filter

The goal of filter $h(t)$ is to provide maximum SNR at time $t = t_0$ for *deterministic* signal $x(t)$ and noise $n(t)$.

$$H(f) = \alpha \frac{X^*(f)}{S_n(f)} e^{-j2\pi f t_0} \quad (86)$$

$$y(t) = \frac{1}{\alpha} R_v(t - t_0) \quad (87)$$

For white noise, $n(t)$, with $S_N(f) = N_0/2$, the filter

is given by

$$H_{mf}(f) = X^*(f) e^{-j2\pi f t_0} \longleftrightarrow h_{mf}(t) = x(t_0 - t) \quad (88)$$

and the resulting maximum SNR is given by

$$SNR_{max} = \frac{2}{N_0} \int_{-\infty}^{\infty} |X(f)|^2 df = \frac{2E_x}{N_0} \quad (89)$$

8 Different Supplementary Formulas

8.1 Derivatives

$$\begin{aligned} \frac{d}{dx} x^n &= nx^{n-1} \\ \frac{d}{dx} \exp[f(x)] &= \exp[f(x)] \frac{d}{dx} f(x) \end{aligned}$$

8.2 Integrals

$$\begin{aligned} \int x^n dx &= \frac{1}{n+1} x^{n+1}, \quad n \neq -1 \\ \int \exp(ax) dx &= \frac{1}{a} \exp(ax) \\ \int x \exp(ax) dx &= \exp(ax) \left[\frac{x}{a} - \frac{1}{a^2} \right] \\ \int x^2 \exp(ax) dx &= \exp(ax) \left[\frac{x^2}{a} - \frac{2x}{a^2} + \frac{2}{a^3} \right] \\ \int \frac{1}{a^2 + x^2} dx &= \frac{1}{a} \tan^{-1} \frac{x}{a} \\ \int \cos(ax + b) dx &= \frac{\sin(ax + b)}{a} \end{aligned}$$

8.2.2 Definite

$$\begin{aligned} \int_0^\infty \exp(-a^2 x^2) dx &= \frac{\sqrt{\pi}}{2a} \\ \int_0^\infty x^2 \exp(-a^2 x^2) dx &= \frac{\sqrt{\pi}}{4a^3} \\ \int_{-\infty}^\infty \delta(x) dx &= 1 \\ \int_{-\infty}^\infty f(x) \delta(x - a) dx &= f(a) \end{aligned}$$

8.3 Fourier Transform

8.3.1 Properties

$$\begin{aligned} \frac{d^n}{dt^n} g(t) &\xleftrightarrow{\mathcal{F}} (j2\pi F)^n G(F) \\ g(-t) &\xleftrightarrow{\mathcal{F}} G^*(F) \\ g(t - t_0) &\xleftrightarrow{\mathcal{F}} G(F) e^{-j2\pi F t_0} \\ g(t) e^{j2\pi f_0 t} &\xleftrightarrow{\mathcal{F}} G(F - F_0) \end{aligned}$$

8.3.2 Transform pairs

$$\begin{aligned}
u(t) &\xleftrightarrow{\mathcal{F}} \frac{1}{2} \left(\frac{1}{j\pi F} + \delta(F) \right) \\
\exp(-at)u(t) &\xleftrightarrow{\mathcal{F}} \frac{1}{a + j2\pi F} \\
t \exp(-at)u(t) &\xleftrightarrow{\mathcal{F}} \frac{1}{(a + j2\pi F)^2} \\
\exp(-a|t|) &\xleftrightarrow{\mathcal{F}} \frac{2a}{a^2 + 4\pi^2 F^2} \\
\exp(-at^2) &\xleftrightarrow{\mathcal{F}} \sqrt{\frac{\pi}{a}} \exp\left(-\frac{(\pi F)^2}{a}\right) \\
\cos(2\pi f_a t) &\xleftrightarrow{\mathcal{F}} \frac{1}{2} \left[\delta(F - F_a) + \delta(F + F_a) \right] \\
\sin(2\pi f_a t) &\xleftrightarrow{\mathcal{F}} \frac{1}{2j} \left[\delta(F - F_a) - \delta(F + F_a) \right] \\
u(t+a) - u(t-a) &\xleftrightarrow{\mathcal{F}} \text{sinc}(2\pi Fa) \quad \text{pulse in time} \\
\text{sinc}(2\pi Fa) &\xleftrightarrow{\mathcal{F}} u(F+a) - u(F-a)
\end{aligned}$$

8.4 Convolution

$$x(t) * y(t) = \int_{-\infty}^{\infty} f(s)g(t-s)ds$$

$$\begin{aligned}
x(t) * y(t) &\xleftrightarrow{\mathcal{F}} X(F)Y(F) \\
\delta(t) * y(t) &= y(t)
\end{aligned}$$

8.5 Trigonometry

$$\begin{aligned}
\sin^2(\alpha) &= \frac{1}{2} (1 - \cos(2\alpha)) \\
\cos^2(\alpha) &= \frac{1}{2} (1 + \cos(2\alpha)) \\
\cos(\alpha) \cos(\beta) &= \frac{1}{2} [\cos(\alpha + \beta) + \cos(\alpha - \beta)] \\
\sin(\alpha) \sin(\beta) &= \frac{1}{2} [\cos(\alpha - \beta) - \cos(\alpha + \beta)] \\
\sin(\alpha) \cos(\beta) &= \frac{1}{2} [\sin(\alpha - \beta) + \sin((\alpha + \beta))]
\end{aligned}$$

8.6 Matrices

$$\mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix},$$

$$\det[\mathbf{A}] = ad - bc$$

$$\mathbf{A}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

9 Discrete-Time

Series sum

$$\begin{aligned}
\sum_{n=0}^{N-1} r^n &= \frac{1 - r^N}{1 - r} \\
\sum_{n=N_1}^{N_2-1} r^n &= \frac{r^{N_1} - r^{N_2}}{1 - r} \quad N_1 \leq N_2 \\
\sum_{n=0}^{\infty} r^n &= \frac{1}{1 - r} \quad |r| < 1 \\
\sum_{n=N_1}^{\infty} r^n &= \frac{1}{1 - r^{N_1}} \quad |r| < 1 \\
\sum_{n=0}^{\infty} n r^n &= \frac{r}{(1 - r)^2} \quad |r| < 1
\end{aligned}$$

9.1 Z-transforms

$$X(z) = \sum_{k=-\infty}^{\infty} x[k]z^{-k}$$

Signal	Z transform	ROC
$\delta[n]$	1	\mathbb{C}
$u[n]$	$\frac{1}{1-z^{-1}}$	$ z > 1$
$-u[-n-1]$	$\frac{1}{1-z^{-1}}$	$ z < 1$
$\delta[n-m]$	z^{-m}	$\mathbb{C} - \{0\}$ if $m > 0$, $\mathbb{C} - \{\infty\}$ if $m < 0$
$a^n u[n]$	$\frac{1}{1-az^{-1}}$	$ z > a$
$-a^n u[-n-1]$	$\frac{1}{1-az^{-1}}$	$ z < a$

Property	Discrete Signal	Z transform	ROC
Linearity	$a_1 x_1[n] + a_2 x_2[n]$	$a_1 X_1(z) + a_2 X_2(z)$	includes $R_1 \cap R_2$
Time shift	$x[n-n_0]$	$z^{-n_0} X(z)$	R
Frequency scaling	$z_0^n x[n]$	$X\left(\frac{z}{z_0}\right)$	$ z_0 R$
Time reversal	$x[-n]$	$X(z^{-1})$	R^{-1} if $m < 0$
Convolution	$(x_1 * x_2)[n]$	$X_1(z)X_2(z)$	$R_1 \cap R_2$ (or possibly more)
Accumulation	$\sum_{k=-\infty}^n x[k]$	$\frac{X(z)}{1-z^{-1}}$	$R \cap \{ z > 1\}$

9.2 DTFT

$$X(f) = X(z) \Big|_{z=e^{j\omega}} = \sum_{n=-\infty}^{\infty} x[n] e^{-jn\omega}$$

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega}) e^{jn\omega} d\omega.$$