

### def 4.13 Characteristic Function

$$\phi_X(u) = E[e^{juX}] = \int_{-\infty}^{\infty} e^{juX} p_X(x) dx = \sum_{n=0}^{\infty} \frac{(ju)^n}{n!} E[X^n]$$

$$p_X(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-juX} \phi_X(u) du$$

### def 4.14 Moment Generating Function

$$\begin{aligned} \Psi_X(u) &= E[e^{uX}] = \sum_{n=0}^{\infty} \frac{u^n}{n!} E[X^n] \\ &= \phi_X(0) + \sum_{n=1}^{\infty} \frac{u^n}{n!} \phi_X^{(n)}(0) \\ &= 1 + \sum_{n=1}^{\infty} \frac{ju^n}{n!} E[X^n] \end{aligned}$$

$$\rightarrow \phi_X^{(k)}(0) = j^k E[X^k] \rightarrow j^{-k} \phi_X^{(k)}(0) = E[X^k] \rightarrow j^k \phi_X^{(k)}(0) = E[X^k]$$

### Gaussian Normal R.V.

$$p_X(x) = \frac{1}{\sqrt{2\pi}\sigma_X} \exp\left(-\frac{(x - \mu_X)^2}{2\sigma_X^2}\right)$$

### def 4.16 Expected value of function of two R.V.

$$E[g(X, Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) p_{XY}(x, y) dx dy$$

### def 4.17 Moments of function of two R.V.

$$E[X^m Y^n] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x^m y^n p_{XY}(x, y) dx dy$$

### def 4.18 Mean Cross Product (평균 외적)

$$E[XY] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy p_{XY}(x, y) dx dy$$

### def 4.19 Covariance (공분산)

$$\text{cov}[X, Y] = E[(X - \mu_X)(Y - \mu_Y)] = E[XY] - \mu_X \mu_Y$$

### def 4.20 Correlation Coefficient (상관계수)

$$\rho_{XY} = \frac{\text{cov}[X, Y]}{\sigma_X \sigma_Y}, |\rho_{XY}| \leq 1$$

### def 4.21 Characteristic function of two R.V. with a joint pdf

$$\phi_{XY}(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{j(ux+vy)} p_{XY}(x, y) dx dy$$

If two R.V. are independent,

$$\phi_{XY}(u, v) = \phi_X(u) \phi_Y(v)$$

### def 4.22 Conditional Expectation of R.V. X given that Y=y

$$E[X|Y=y] = \int_{-\infty}^{\infty} x p_{X|Y}(x|y) dx = \int_{-\infty}^{\infty} x \frac{p_{XY}(x, y)}{p_Y(y)} dx$$

### def 4.23 Expectation of R.V. Vector

If

$$\vec{X} = (X_1, X_2, \dots, X_n)^T$$

then mean vector

$$\vec{\mu}_{\vec{X}} = (\mu_1, \mu_2, \dots, \mu_n)^T$$

where

$$\begin{aligned} \mu_i &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} x_i p_{\vec{X}}(x_1, x_2, \dots, x_n) dx_1 dx_2 \dots dx_n \\ &= \int_{-\infty}^{\infty} x_i p_{X_i}(x) dx_i \text{ for } i = 1, 2, \dots, n \end{aligned}$$

### def 4.24 Covariance Matrix

$$C_{\vec{X}\vec{X}^T} = E[(\vec{X} - \vec{\mu}_{\vec{X}})(\vec{X} - \vec{\mu}_{\vec{X}})^T]$$

### def 4.25 Correlation Matrix

$$R_{\vec{X}\vec{X}^T} = E[(\vec{X})(\vec{X})^T]$$

### Given independent two R.V. X & Y, Z=X+Y?

$$p_Z(z) = \int_{-\infty}^{\infty} p_X(x) p_Y(z-x) dx = p_X(z) * p_Y(z)$$

- characteristic

$$E[Z] = E[X] + E[Y]$$

$$\sigma_Z^2 = \sigma_X^2 + \sigma_Y^2 \text{ (증명문제 시험나옴)}$$

### Central Limit Theorem

$$p_Z(z) = p_{X_1}(z) * p_2(z) * \dots * p_{X_n}(z)$$

$$\lim_{n \rightarrow \infty} p_Z(z) = \frac{1}{\sqrt{2\pi}\sigma_Z} \exp\left(-\frac{(z - \mu_Z)^2}{2\sigma_Z^2}\right)$$

### Transform of a R.V.

$$Y = f(X) = g^{-1}(X)$$

$$P_Y(Y \leq y) = P_X(X \leq x) = P_X(g(Y) \leq x)$$

$$p_Y(y) = p_X(g(y)) \left[ \frac{d}{dy} g(y) \right]$$

def 5.19 **Poisson Random Process**

$$P_X(i, t) = \frac{(\lambda t)^i}{i!} e^{-\lambda t} u(t)$$

$$p_X(i, t) = \sum_{r=0}^{\infty} \frac{(\lambda t)^r}{r!} e^{-\lambda t} \delta(i - r)$$

Ex 5.18 풀어보도록.

def 5.20 **Poisson Impulse**

$$d(t) = \sum_n \delta(t - t_n)$$

def 5.22 **Shot Noise**

$$S(t) = h(t) * D(t)$$

def 5.23 **Bernoulli Random Process**

+1과 -1(혹은 +1과 0)의 두 가지 값만을 가질 수 있는 이산 랜덤 시퀀스  $X(nT)$ 이다. +1의 확률은  $p$ 이며, -1의 확률은  $q = 1 - p$ 이다.

def 5.24 **Binary counting Process**

Bernoulli R.P. 에서 펄스의 발생을 계수하는 R.P. 펄스는 +1 혹은 0이다.  $X(i) = X_i$ 는  $i$ 번 째 시점에서 펄스가 발생되면 1, 그렇지 않으면 0이다.

$$C(n) = C_n = \sum_{i=1}^n X_i$$

if  $C(n) = k$

$$p(C_n = K) = \binom{n}{k} p^k q^{n-k}, \quad k = 0, 1, \dots, n$$

def 5.25 **Independent intervals (or Independent increment)**

Bernoulli R.P.에서 연속되는 값들의 차이가 상호독립 즉,  $(X(1) - X(0)), (X(2) - X(1)), \dots, (X(k) - X(k-1)), \dots, (X(n) - X(n-1))$  이 상호독립일 경우에, Independent intervals (or Independent increment)를 갖는다고 한다.

def 5.26 **Random Walk Process**

$$P_X(n, x) = \binom{n}{\frac{n+x}{2}} p^{\frac{n+x}{2}} q^{\frac{n-x}{2}}$$

$$p_X(n, x) = \sum_{i=-n}^n \binom{n}{\frac{n+i}{2}} p^{\frac{n+i}{2}} q^{\frac{n-i}{2}} \delta(x - i)$$

$$= \sum_{r=0}^n \binom{n}{r} p^r q^{n-r} \delta(x - 2r + n)$$

def 5.27 **Winner Random Process**

Random Walk P.R. 에서  $p$ 와  $q$ 의 확률이 같을 때.

def 5.28 **Markov Random Process**

$$P_S = [s_j(n) | s_a(n-1), s_b(n-1), s_c(n-1), \dots]$$

$$= P_S[s_j(n) | s_a(n-1)]$$

$$P_{ij} = P[s_j(n) | s_i(n-1)] \text{ for } 1 \leq i, j \leq k$$

def 5.29 **Markov Chain**

초기 상태와 유한하면서 셀수 있는 상태들 그리고 이와 관련된 전이 확률들을 갖는 Markov R.P.

def 5.30 **Transition Probability Matrix**

$$T = \begin{bmatrix} P_{11} & P_{12} & P_{13} \\ P_{21} & P_{22} & P_{23} \\ P_{31} & P_{32} & P_{33} \end{bmatrix}$$

$$\sum_j P_{ij} = 1, 0 \leq P_{ij} \leq 1$$

$i$ 번째 상태  $s$ 에서 시작하여,  $m$ 번 천이 과정 후,  $j$ 번째 상태  $s$ 에 있을 확률.

$$P_{ij}(m) = P[s_j(n+m) | s_i(n)] = \sum_{r=1}^k P_{ir}(m-l) P_{rj}(l)$$

def 5.31 **Transient State (일시적 상태)**

$$P_{ij}(l) \neq 0 \text{ \& } P_{ji}(r) = 0 \text{ for } r = 0, 1, 2, \dots$$

def 5.32 **Recurrent State (재귀상태)**

모든 상태는 일시적 상태가 아니면 재귀상태이다.

def 5.33 **Periodic State (주기상태)**

$c, 2c, 3c, \dots$ 가 아닌 모든  $r$ 에 대해서

$$P_{ii}(r) = 0, \text{ for } c > 1$$

def 5.34 **Steady State (정상상태)**

$$\lim_{m \rightarrow \infty} P[s_j(m)] = P_j \text{ for } j = 1, 2, \dots, k$$

def 5.35 **Regular Chain (정규상태)**

전이행렬의 거듭제곱이 양의 원소만을 가질 경우에 마르코프 사슬을 정규사슬이라 한다. 이는 정규사슬이 주기 상태를 가지지 않음을 의미한다.

def 6.1 **Expected Value of a continuous time R.P.**

$$\begin{aligned} \mu_X(t) &= E[X(t)] \text{ for } -\infty < t < \infty \\ &= \int_{-\infty}^{\infty} x(t) p_X(x(t)) d(x(t)) \end{aligned}$$

def 6.2 **Autocorrelation function**

$$\begin{aligned} R_{XX}(t_1, t_2) &= E[X(t_1)X^*(t_2)] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x(t_1)x(t_2)p_X(x(t_1), x(t_2))dx(t_1) dx(t_2) \end{aligned}$$

def 6.3 **Autocovariance**

$$\begin{aligned} Cov_{XX}(t, t + \tau) &= E[(X(t) - \mu_X(t))(X(t + \tau) \\ &\quad - \mu_X(t + \tau))] \\ &= R_{XX}(t, t + \tau) - \mu_X(t)\mu_X(t + \tau) \\ &= \sigma_{XX}^2(t, t + \tau) \end{aligned}$$

def 6.4 **Autocorrelation Coefficient**

$$\rho_{XX}(t, t + \tau) = \frac{Cov_{XX}(t, t + \tau)}{\sqrt{Cov_{XX}(t, t)Cov_{XX}(t + \tau, t + \tau)}}$$

def 6.5 **Strict Sense Stationary R.P.**

$$\begin{aligned} \rho_X[X(t_1), X(t_2), \dots, X(t_n)] \\ = \rho_X[X(t_1 + \tau), X(t_2 + \tau), \dots, X(t_n + \tau)] \end{aligned}$$

def 6.6 **Wide Sense Stationary R.P.**

$$\begin{aligned} \mu_X(t) &= \mu_X \\ R_{XX}(t, t + \tau) &= R_{XX}(\tau) \end{aligned}$$

def 6.7 **Autocovariance for W.S.S. R.P.**

$$Cov_{XX}(\tau) = R_{XX}(\tau) - \mu_X^2 = \sigma_{XX}^2(\tau)$$

def 6.8 **Autocorrelation Coefficient for W.S.S R.P.**

$$\rho_{XX}(\tau) = \frac{\sigma_{XX}^2(\tau)}{\sqrt{\sigma_{XX}^2(0)\sigma_{XX}^2(0)}} = \frac{\sigma_{XX}^2(\tau)}{\sigma_{XX}^2(0)}$$

def 6.9 **Mean function for Random sequence**

$$\begin{aligned} \mu_X(n) &= E[X(n)] \text{ for } -\infty < n < \infty \\ &= \int_{-\infty}^{\infty} x(n)p_X(x(n))d(x(n)) \end{aligned}$$

def 6.10 **Autocorrelation function for R.S.**

$$\begin{aligned} R_{XX}(n, n + k) &= E[X(n)X(n + k)] = \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x(n)x(n + k)p_X(x(n), x(n + k))dx(n) dx(n + k) \end{aligned}$$

def 6.11 **Autocovariance for R.S.**

$$\begin{aligned} Cov_{XX}(n, n + k) \\ &= E[(X(n) - \mu_X(n))(X(n + k) - \mu_X(n + k))] \\ &= R_{XX}(n, n + k) - \mu_X(n)\mu_X(n + k) = \sigma_{XX}^2(n, n + k) \end{aligned}$$

def 6.12 **Autocorrelation Coefficient for R.S.**

$$\rho_{XX}(n, n + k) = \frac{Cov_{XX}(n, n + k)}{\sqrt{Cov_{XX}(n, n)Cov_{XX}(n + k, n + k)}}$$

def 6.13 **Strict Sense Stationary R.S.**

$$\begin{aligned} \rho_X[X(n_1), X(n_2), \dots, X(n_n)] \\ = \rho_X[X(n_1 + k), X(n_2 + k), \dots, X(n_n + k)] \end{aligned}$$

def 6.14 **Wide Sense Stationary R.S.**

$$\begin{aligned} \mu_X(n) &= \mu_X \\ R_{XX}(n, n + k) &= R_{XX}(k) \end{aligned}$$

def 6.15 **Autocovariance function for W.S.S. R.S.**

$$Cov_{XX}(k) = R_{XX}(k) - \mu_X^2 = \sigma_{XX}^2(k)$$

def 6.16 **Autocorrelation Coefficient for W.S.S. R.S.**

$$\rho_{XX}(k) = \frac{\sigma_{XX}^2(k)}{\sqrt{\sigma_{XX}^2(0)\sigma_{XX}^2(0)}} = \frac{\sigma_{XX}^2(k)}{\sigma_{XX}^2(0)}$$

def 6.17 **Crosscorrelation function**

$$\begin{aligned} R_{XY}(t_1, t_2) &= E[X(t_1)Y(t_2)] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x(t_1)y(t_2)p_X(x(t_1), y(t_2))dx(t_1) dy(t_2) \end{aligned}$$

$$R_{XY}(\tau) = R_{YX}(-\tau)$$

$$R_{XX}(\tau) = R_{XX}(-\tau)$$

$$|R_{XY}(\tau)| \leq \frac{1}{2}(R_{XX}(0) + R_{YY}(0))$$

$$|R_{XY}(\tau)|^2 \leq R_{XX}(0)R_{YY}(0)$$

def 6.18 **Crosscovariance**

$$\begin{aligned} Cov_{XX}(t, t + \tau) &= E[(X(t) - \mu_X(t))(Y(t + \tau) - \mu_Y(t + \tau))] \\ &= R_{XY}(t, t + \tau) - \mu_X(t)\mu_Y(t + \tau) \\ &= \sigma_{XY}^2(t, t + \tau) \end{aligned}$$

def 6.19 **Crosscorrelation Coefficient**

$$\rho_{XY}(t, t + \tau) = \frac{Cov_{XY}(t, t + \tau)}{\sqrt{Cov_{XX}(t, t)Cov_{YY}(t + \tau, t + \tau)}}$$

def 6.25 **If two R.P. are independent**

$$P_{XY}(x(t), y(t)) = P_X(x(t))P_Y(y(t))$$

def 6.26 **If two R.P. are orthogonal**

$$R_{XY}(t, t + \tau) = E[X(t)Y(t + \tau)] = 0$$

def 6.27 **If two R.P. are uncorrelated**

$$\begin{aligned} Cov_{XY}(t, t + \tau) &= R_{XY}(t, t + \tau) - \mu_X(t)\mu_Y(t + \tau) = 0 \\ \rightarrow R_{XY}(t, t + \tau) &= \mu_X(t)\mu_Y(t + \tau) \end{aligned}$$

def 6.28 **Power Spectral Density Function**

$$S_{XX}(\omega) = S_{XX}(2\pi f) = \int_{-\infty}^{\infty} R_{XX}(\tau) e^{-j\omega\tau} d\tau$$

$$R_{XX}(\tau) = \int_{-\infty}^{\infty} S_{XX}(f) e^{j\omega\tau} df = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{XX}(\omega) e^{j\omega\tau} d\omega$$

**Average power of the R.P.**

$$R_{XX}(0) = \int_{-\infty}^{\infty} S_{XX}(f) e^{j\omega\tau} df \Big|_{\tau=0} = \int_{-\infty}^{\infty} S_{XX}(f) df \geq 0$$