

# Probability System

- A sample space  $S$
- A class  $\mathcal{E}$  of events that are subsets of  $S$
- A probability measure  $P(\cdot)$  for each event  $A$  in  $\mathcal{E}$ , which has the following properties
  - $P(S) = 1$
  - $0 \leq P(A) \leq 1$
  - If  $A \cap B = \emptyset$ ,  $P(A + B) = P(A) + P(B)$
- Example
  - $S = \{1, 2, 3, 4, 5, 6\}$
  - $\mathcal{E} = \{\emptyset, \{1\}, \{2\}, \{3\}, \{4\}, \{5\}, \{6\}, \{1, 2\}, \{1, 3\}, \dots, \{1, 2, 3\}, \dots, S\}$
  - $P(A) = |A|/6$
  - $A = \{1, 2\}, B = \{3, 4\}$
  - $P(A+B) = 2/3 = P(A) + P(B)$

- Joint events and joint probabilities

Two experiments

$A_i, i = 1, 2, \dots, n$

$B_j, j = 1, 2, \dots, m$

With properties

$0 \leq P(A_i, B_j) \leq 1$

$$\sum_{j=1}^m P(A_i, B_j) = P(A_i)$$

$$\sum_{i=1}^n P(A_i, B_j) = P(B_j)$$

- **Conditional probabilities**

One experiment, two events

$$P(A|B) = \frac{P(A, B)}{P(B)} = \frac{P(B|A)P(A)}{P(B)}$$

- **Statistical independence**

$$P(A|B) = P(A)$$

$$P(A, B) = P(A)P(B)$$

- **Examples**

$$A = \{1\}, B = \{1, 3, 5\}$$

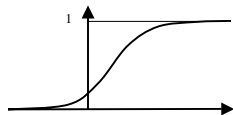
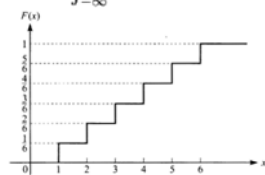
$$P(A|B) = 1/3.$$

## Probability Distribution One-Dimensional Random Variable

- **Cumulative distribution function (CDF)**

$$F(x) = P(X \leq x), \quad -\infty < x < \infty$$

$$F(x) = \int_{-\infty}^x p(u) du, \quad -\infty < x < \infty$$

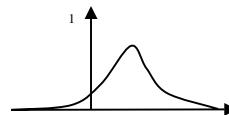


- **Probability density function (PDF)**

$$p(x) = \frac{dF(x)}{dx}, \quad -\infty < x < \infty$$

- **Examples**

$$p(x) = \sum_{i=1}^n P(X = x_i) \delta(x - x_i)$$



# Joint probability distributions

- Cumulative distribution function (CDF)

$$F(x_1, x_2) = P(X_1 \leq x_1, X_2 \leq x_2)$$

$$= \int_{-\infty}^{x_1} \int_{-\infty}^{x_2} p(u_1, u_2) du_1 du_2$$

- $N$ -dimensional CDF

$$F(x_1, x_2, \dots, x_n) = P(X_1 \leq x_1, X_2 \leq x_2, \dots, X_n \leq x_n)$$

$$= \int_{-\infty}^{x_1} \int_{-\infty}^{x_2} \dots \int_{-\infty}^{x_n} p(u_1, u_2, \dots, u_n) du_1 du_2 \dots du_n$$

- Independence

$$F(x_1, x_2, \dots, x_n) = F(x_1)F(x_2) \dots F(x_n)$$

$$p(x_1, x_2, \dots, x_n) = p(x_1)p(x_2) \dots p(x_n)$$

- Probability density function (PDF)

$$p(x_1, x_2) = \frac{\partial^2}{\partial x_1 \partial x_2} F(x_1, x_2)$$

- Marginal PDF

$$\int_{-\infty}^{\infty} p(x_1, x_2) dx_2 = p(x_1)$$

$$\int_{-\infty}^{\infty} p(x_1, x_2) dx_1 = p(x_2)$$

- Conditional PDF

$$p(x_1|x_2) = \frac{p(x_1, x_2)}{p(x_2)}$$

$$p(x_1, x_2) = p(x_1|x_2)p(x_2)$$

$$= p(x_2|x_1)p(x_1)$$

# Functions of Random Variables

- One Dimension

$$Y = g(X)$$

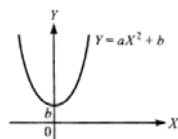
$$p_Y(y) = \sum_{i=1}^n \frac{p_X(x_i)}{|g'(x_i)|}$$

$$y = g(x_i), i = 1, \dots, n$$

- Example

$$Y = aX^2 + b, \quad a > 0$$

$$p_Y(y) = \frac{p_X[\sqrt{(y-b)/a}]}{2a\sqrt{(y-b)/a}} + \frac{p_X[-\sqrt{(y-b)/a}]}{2a\sqrt{(y-b)/a}}$$



- **N-Dimension**

$$Y_i = g_i(X_1, X_2, \dots, X_n), \quad i = 1, 2, \dots, n$$

$$X_i = g_i^{-1}(Y_1, Y_2, \dots, Y_n), \quad i = 1, 2, \dots, n$$

$$p_Y(y_1, y_2, \dots, y_n) = p_X(x_1 = g_1^{-1}, x_2 = g_2^{-1}, \dots, x_n = g_n^{-1}) |J|$$

$$J = \begin{vmatrix} \frac{\partial g_1^{-1}}{\partial y_1} & \frac{\partial g_2^{-1}}{\partial y_2} & \dots & \frac{\partial g_n^{-1}}{\partial y_1} \\ \vdots & \vdots & & \vdots \\ \frac{\partial g_1^{-1}}{\partial y_n} & \frac{\partial g_2^{-1}}{\partial y_n} & \dots & \frac{\partial g_n^{-1}}{\partial y_n} \end{vmatrix}$$

- **Examples**

$$Y_i = \sum_{j=1}^n a_{ij} X_j, \quad i = 1, 2, \dots, n$$

$$\mathbf{Y} = \mathbf{A}\mathbf{X}$$

$$X_i = \sum_{j=1}^n b_{ij} Y_j, \quad i = 1, 2, \dots, n$$

$$p_Y(y_1, y_2, \dots, y_n) = p_X\left(x_1 = \sum_{j=1}^n b_{1j} y_j, x_2 = \sum_{j=1}^n b_{2j} y_j, \dots, x_n = \sum_{j=1}^n b_{nj} y_j\right) \frac{1}{|\det \mathbf{A}|}$$

## Average/expectation

- Mean

$$E(X) = m_x = \int_{-\infty}^{\infty} xp(x) dx$$

- nth moment

$$E(X^n) = \int_{-\infty}^{\infty} x^n p(x) dx$$

- nth central moment

$$E[(X - m_x)^n] = \int_{-\infty}^{\infty} (x - m_x)^n p(x) dx$$

- Variance

$$\sigma_x^2 = \int_{-\infty}^{\infty} (x - m_x)^2 p(x) dx$$

$$\sigma_x^2 = E(X^2) - [E(X)]^2$$

$$= E(X^2) - m_x^2$$

- Covariance matrix

$$\mu_{ij} = E[(X_i - m_i)(X_j - m_j)]$$

$$= E(X_i X_j) - m_i m_j$$

- Example

$$p_X(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$

$$E[x^n] = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} x^n e^{-x^2/2} dx$$

$$= \begin{cases} 1 \cdot 3 \cdot 5 \cdot \dots \cdot (n-1) & n \text{ even} \\ 0 & n \text{ odd} \end{cases}$$

$$Y = \sigma X + \mu$$

$$p_Y(y) = \frac{1}{\sqrt{2\pi}} \exp\left[-\frac{(y - \mu)^2}{2\sigma^2}\right]$$

$$E[Y] = \mu$$

$$\text{var}[Y] = E[(y - \mu)^2] = \sigma^2$$

# Characteristic Functions

- Fourier transform of PDF

$$E(e^{jvX}) \equiv \psi(jv) = \int_{-\infty}^{\infty} e^{jvx} p(x) dx$$

$$p(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \psi(jv) e^{-jvx} dv$$

- Moment vs. characteristic functions

$$E(X) = m_x = -j \left. \frac{d\psi(jv)}{dv} \right|_{v=0}$$

$$E(X^n) = (-j)^n \left. \frac{d^n \psi(jv)}{dv^n} \right|_{v=0}$$

- Sum of random variables

$$Y = \sum_{i=1}^n X_i$$

$$\psi_Y(jv) = E(e^{jvY})$$

$$= E \left[ \exp \left( jv \sum_{i=1}^n X_i \right) \right]$$

$$= E \left[ \prod_{i=1}^n (e^{jvX_i}) \right]$$

$$= \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} \left( \prod_{i=1}^n e^{jvx_i} \right) p(x_1, x_2, \dots, x_n) dx_1 dx_2 \cdots dx_n$$

- Independence

$$p(x_1, x_2, \dots, x_n) = p(x_1)p(x_2) \cdots p(x_n)$$

$$\psi_Y(jv) = \prod_{i=1}^n \psi_{X_i}(jv)$$

- Identical independent distributed

$$\psi_Y(jv) = [\psi_X(jv)]^n$$

# Characteristic Functions

- $N$ -Dimensional

$$\begin{aligned}\psi(jv_1, jv_2, \dots, jv_n) &= E\left[\exp\left(j\sum_{i=1}^n v_i X_i\right)\right] \\ &= \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} \exp\left(j\sum_{i=1}^n v_i x_i\right) p(x_1, x_2, \dots, x_n) dx_1 dx_2 \dots dx_n\end{aligned}$$

- Example

$$\psi(jv_1, jv_2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{j(v_1 x_1 + v_2 x_2)} p(x_1, x_2) dx_1 dx_2$$

$$E(X_1 X_2) = -\left. \frac{\partial^2 \psi(jv_1, jv_2)}{\partial v_1 \partial v_2} \right|_{v_1=v_2=0}$$

## Distribution Derived from Gaussian Distribution $X$ is Gaussian distributed

- Chi-square,  $\chi^2$  (Gamma,  $\Gamma$ ) distribution  $n$  degrees of freedom
  
- Noncentral  $\chi^2$  distribution with  $n$  degrees of freedom

- Rayleigh Distribution
- Rice Distribution
- Nakagami  $m$ -distribution