

# CSE400: Fundamentals of Probability in Computing

## Lecture 8: Gaussian, Uniform, Exponential, and Gamma Random Variables

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## 1 Continuous Random Variables: Moments

### 1.1 Central Moments

The  $n$ -th order central moment is defined as the expectation of the random variable centered around its mean.

$$[cite_{start}] E[(x - \mu_X)^n] = \int_{-\infty}^{\infty} (x - \mu_X)^n f_X(x) dx \quad [cite: 20, 51] \quad (1)$$

**Specific Moments:**

- [cite<sub>start</sub>] • **Zeroth Moment** ( $n = 0$ ):  $E[(x - \mu_X)^0] = E[1] = 1$  [cite: 55]. [cite<sub>start</sub>]
- **First Moment** ( $n = 1$ ):  $E[(x - \mu_X)] = E[x] - \mu_x = 0$  [cite: 56, 58].
- **Second Moment** ( $n = 2$ ): Defined as the Variance ( $\sigma_X^2$ ).

$$[cite_{start}] \sigma_X^2 = E[(x - \mu_X)^2] = E[x^2] - \mu_X^2 \quad [cite: 59, 61] \quad (2)$$

### 1.2 Higher Order Moments (Shape)

- **Skewness** ( $n = 3$ ): A measure of the symmetry of the Probability Density Function (PDF).

$$[cite_{start}] C_s = \frac{E[(X - \mu_X)^3]}{\sigma_X^3} \quad [cite: 81] \quad (3)$$

- [cite<sub>start</sub>] – If  $C_s$  is positive (+), the PDF is Right Skewed [cite: 82]. [cite<sub>start</sub>]  
– If  $C_s$  is negative (-), the PDF is Left Skewed [cite: 83].
- **Kurtosis** ( $n = 4$ ): A measure of the "tailedness" or peak of the distribution.

$$[cite_{start}] C_k = \frac{E[(X - \mu_X)^4]}{\sigma^4} \quad [cite: 86, 341] \quad (4)$$

[cite<sub>start</sub>] A large value indicates the random variable will have a large peak near the mean [cite: 88, 89].

## 2 Theorems on Expectation

### 2.1 Linearity of Expectation

For any constants  $a$  and  $b$ :

$$[cite_{start}] E[ax + b] = aE[x] + b \quad [cite: 94] \quad (5)$$

## 2.2 Sum of Functions

For a sum of several functions  $g(x) = g_1(x) + g_2(x) + \dots + g_N(x)$ :

$$[\text{cite}_s\text{tart}] E \left[ \sum_{k=1}^N g_k(x) \right] = \sum_{k=1}^N E[g_k(x)] \quad [\text{cite: 97}] \quad (6)$$

## 3 Gaussian Random Variable

### 3.1 Definition

A Gaussian random variable  $X \sim \mathcal{N}(m, \sigma^2)$  is defined by the PDF:

$$[\text{cite}_s\text{tart}] f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-m)^2}{2\sigma^2}\right) \quad [\text{cite: 119}] \quad (7)$$

Where  $m$  is the mean and  $\sigma^2$  is the variance.  $[\text{cite}_s\text{tart}] If m=0 and \sigma^2 = 1$ , it is a Standard Normal Distribution[cite: 121].

### 3.2 Standard Functions

Several standard integrals are used to evaluate Gaussian probabilities:

- **Error Function:**  $[\text{cite}_s\text{tart}] \text{erf}(x) = 2 \frac{1}{\sqrt{\pi}} \int_0^x \exp(-t^2) dt$  [cite: 166].
- **Complementary Error Function:**  $[\text{cite}_s\text{tart}] \text{erfc}(x) = 1 - \text{erf}(x) = 2 \frac{1}{\sqrt{\pi}} \int_x^\infty \exp(-t^2) dt$  [cite: 167].
- **$\Phi$ -function (CDF of Standard Normal):** Represents the area under the left tail.

$$[\text{cite}_s\text{tart}] \Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x \exp\left(-\frac{t^2}{2}\right) dt \quad [\text{cite: 175, 250}] \quad (8)$$

- **Q-function (Gaussian Tail Function):** Represents the area under the right tail.

$$[\text{cite}_s\text{tart}] Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^\infty \exp\left(-\frac{t^2}{2}\right) dt \quad [\text{cite: 176, 254}] \quad (9)$$

### 3.3 Key Identities and Relations

[\text{cite}\_s\text{tart}]

- $Q(x) = 1 - \Phi(x)$  [cite: 214].
- $\Phi(-x) = 1 - \Phi(x)$  [cite: 305].
- The CDF of a general Gaussian variable can be expressed as:

$$[\text{cite}_s\text{tart}] F_X(x) = \Phi\left(\frac{x-m}{\sigma}\right) = 1 - Q\left(\frac{x-m}{\sigma}\right) \quad [\text{cite: 211, 214}] \quad (10)$$

- Tail probabilities for a general Gaussian variable:

$$[\text{cite}_s\text{tart}] Pr(X > x) = Q\left(\frac{x-m}{\sigma}\right) \quad [\text{cite: 205}] \quad (11)$$

## 4 Worked Example 1: Gaussian Probabilities

**Problem:** A random variable  $X$  has a PDF given by:

$$f_X(x) = \frac{1}{\sqrt{8\pi}} \exp\left(-\frac{(x+3)^2}{8}\right)$$

[cite\_start] Find probabilities in terms of  $Q$ -functions [cite : 262].

**Solution:**

1. **Identify parameters:** Comparing to the standard form  $\frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-m)^2}{2\sigma^2}\right)$ , we get: [cite\_start] $m = -3$ ,  $\sigma^2 = 4 \Rightarrow \sigma = 2$  [cite: 287]

2. **Calculate  $Pr(X \leq 0)$ :** [cite\_start] $Pr(X \leq 0) = F_X(0) = \Phi\left(\frac{0-(-3)}{2}\right) = \Phi(1.5) = 1 - Q(1.5)$  [cite: 297]

3. **Calculate  $Pr(X > 4)$ :** [cite\_start] $Pr(X > 4) = Q\left(\frac{4-(-3)}{2}\right) = Q(3.5)$  [cite: 298]

4. **Calculate  $Pr(|X + 3| < 2)$ :**

$$|X + 3| < 2 \Rightarrow -2 < X + 3 < 2 \Rightarrow -5 < X < -1 \quad [\text{cite: 303}]$$

$$Pr(-5 < X < -1) = F_X(-1) - F_X(-5) = \Phi(1) - \Phi(-1)$$

Using  $\Phi(-1) = 1 - \Phi(1)$ : [cite\_start] $= \Phi(1) - (1 - \Phi(1)) = 2\Phi(1) - 1 = 1 - 2Q(1)$  [cite: 307]

5. **Calculate  $Pr(|X - 2| > 1)$ :**

$$|X - 2| > 1 \Rightarrow X < 1 \text{ or } X > 3 \quad [\text{cite: 308}]$$

$$Pr = F_X(1) + Pr(X > 3)$$

$$F_X(1) = \Phi\left(\frac{1-(-3)}{2}\right) = \Phi(2) = 1 - Q(2)$$

$$Pr(X > 3) = Q\left(\frac{3-(-3)}{2}\right) = Q(3)$$

[cite\_start] Total =  $1 - Q(2) + Q(3)$  [cite: 309]

## 5 Worked Example 2: CDF Analysis

**Problem:** The CDF of a random variable is given by:

$$F_X(x) = \begin{cases} 0, & x < 0 \\ x^2, & 0 \leq x \leq 1 \\ 1, & x > 1 \end{cases}$$

[cite\_start] Find the Mean, Variance, Skewness, and Kurtosis [cite : 322, 323].

**Solution:**

1. **Find PDF:** [cite\_start] $f_X(x) = \frac{d}{dx} F_X(x) = 2x$ ,  $0 \leq x \leq 1$  [cite: 333]

2. **Calculate Mean ( $\mu_X$ ):** [cite\_start] $\mu_X = \int_0^1 x(2x)dx = \int_0^1 2x^2 dx = \left[\frac{2x^3}{3}\right]_0^1 = \frac{2}{3}$  [cite: 333]

3. **Calculate Variance ( $\sigma_X^2$ ):**

$$\sigma_X^2 = \int_0^1 x^2(2x)dx - \mu_X^2 = \int_0^1 2x^3 dx - \left(\frac{2}{3}\right)^2$$

$$[\text{cite_start}] = \left[\frac{2x^4}{4}\right]_0^1 - \frac{4}{9} = \frac{1}{2} - \frac{4}{9} = \frac{9-8}{18} = \frac{1}{18} \quad [\text{cite: 340}]$$

4. **Skewness and Kurtosis:** [cite\_start] $C_s = \frac{E[(X-\mu)^3]}{\sigma^3} = -\frac{2\sqrt{2}}{5}$  (Distribution is left-skewed) [cite: 341, 343] [cite\_start] $C_k = \frac{E[(X-\mu)^4]}{\sigma^4} = \frac{12}{5}$  (Platykurtic) [cite: 341, 344]

## 6 Applications

Gaussian Random Variables are used to model:

[cite<sub>start</sub>]

- **Thermal Noise:** Noise voltage in electronic circuits[cite: 314]. [cite<sub>start</sub>]
- **Measurement Error:** Sensor readings where Measured Value = True Value + Gaussian noise[cite: 315]. [cite<sub>start</sub>]
- **Jitter:** Packet delay variation in communication networks[cite: 316]. [cite<sub>start</sub>]
- **Density Estimation:** Using sample statistics (histogram) to estimate the distribution parameters  $\hat{\mu}$  and  $\hat{\sigma}$ [cite: 397, 404].