

# **CENG 222**

## **Statistical Methods for Computer Engineering**

### **Week 5**

#### Chapter 4

#### Continuous Distributions: Gamma and Normal Distributions, Central Limit Theorem

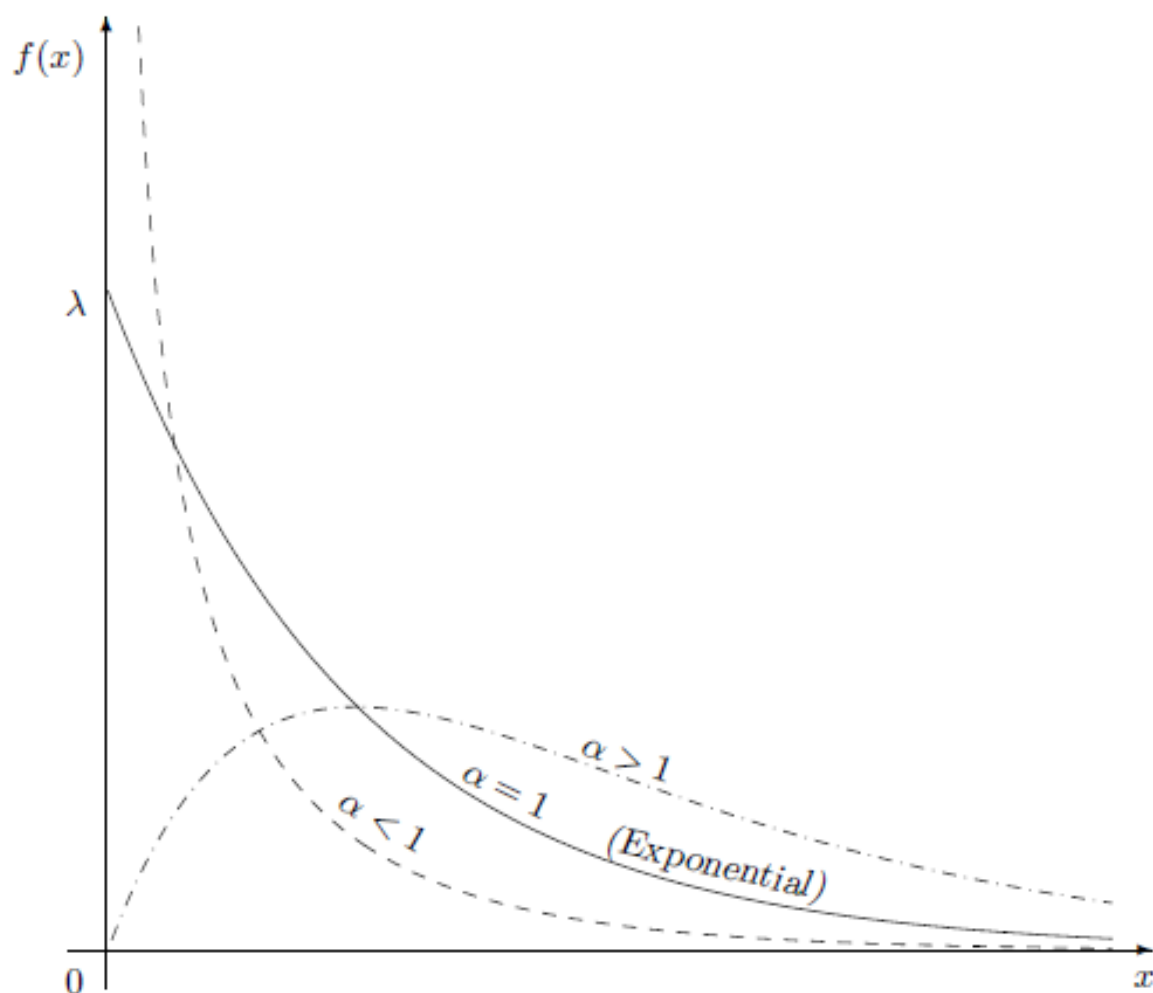
# Gamma distribution

- $X$  = the total time of observing  $\alpha$  rare and independent events each with exponential waiting times (with parameter  $\lambda$ )
  - i.e., it is the sum of  $\alpha$  exponential rvs
- Expectation and variance can be found using linearity of expectation.
  - $E(X) = \frac{\alpha}{\lambda}$ ,  $Var(X) = \frac{\alpha}{\lambda^2}$

## Gamma pdf

- $f_X(x) = \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda}, \quad x > 0$
- $\Gamma(\alpha) = (\alpha - 1)!$

$\alpha$  does not need to be an integer

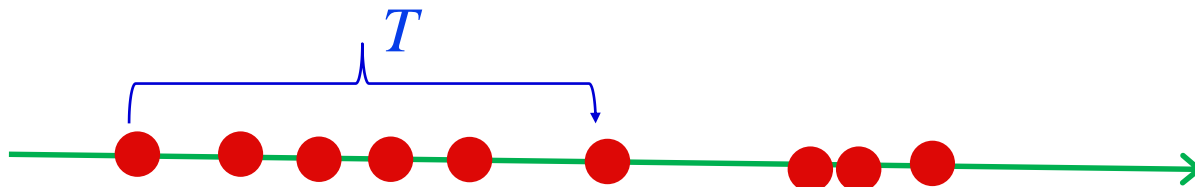


# Gamma distribution

- Is widely used to model random variables other than waiting times (since  $\alpha$  does not need to be an integer)
  - Amount of money spent
  - Amount of resources used (electricity, gas, etc.)

# Gamma-Poisson formula

- Rare events



- $T = \text{time of the } \alpha\text{th rare event} = \text{Gamma } (\alpha, \lambda)$ 
  - The event  $\{T > t\}$  means that fewer than  $\alpha$  events occur in  $t$  time.
  - Let  $X$  be a Poisson rv with parameter  $\lambda t$
  - $\{T > t\} = \{X < \alpha\}$  hence  $P(T > t) = P(X < \alpha)$
  - $\rightarrow P(T \leq t) = P(X \geq \alpha)$
  - $\rightarrow$  we can use the Poisson table for computation of Gamma probabilities (**Caution**:  $T$  is continuous,  $X$  is discrete)

## Example 4.9

- Lifetimes for computer chips have Gamma distribution with expectation  $\mu=12$  years and standard deviation  $\sigma=4$  years. What is the probability that such a chip has a lifetime between 8 and 10 years?
- Step 1: what are the parameters of this Gamma rv?

$$-\frac{\alpha}{\lambda} = 12, \frac{\alpha}{\lambda^2} = 16 \rightarrow \lambda = 12/16 = 0.75, \alpha = 12*0.75 = 9$$

## Example 4.9 continued

- Step 2: Compute the probability
  - $P(8 < T < 10) = F_T(10) - F_T(8)$
  - $F_T(10) = P(T \leq 10) = P(X_1 \geq 9)$  where  $X_1 = \text{Poisson}(7.5)$ 
    - $P(X_1 \geq 9) = 1 - F_{X_1}(8) = 0.338$
  - $F_T(8) = P(T \leq 8) = P(X_2 \geq 9)$  where  $X_2 = \text{Poisson}(6)$ 
    - $P(X_2 \geq 9) = 1 - F_{X_2}(8) = 0.153$
  - $P(8 < T < 10) = 0.338 - 0.153 = 0.185$

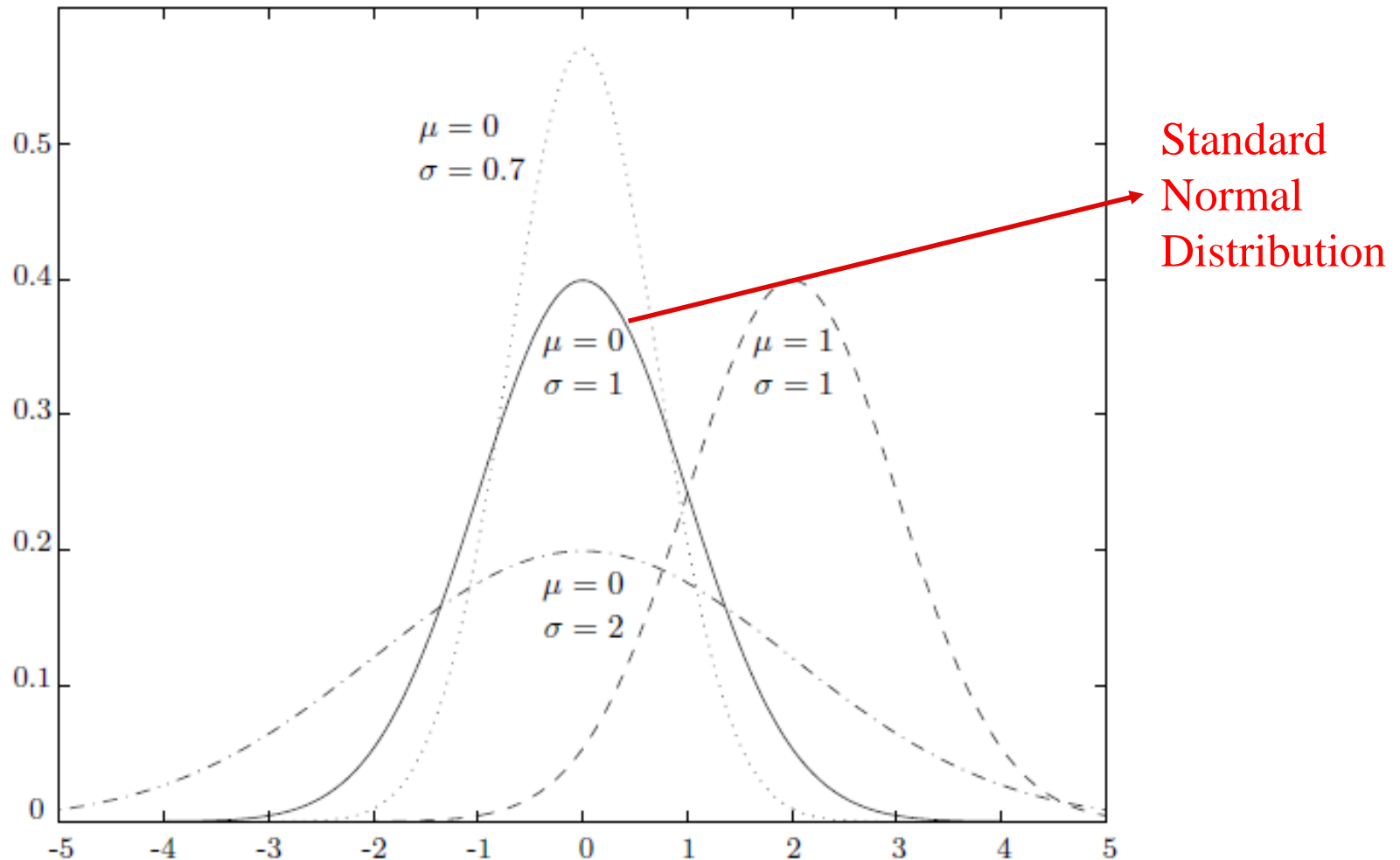


# Normal (Gaussian) distribution

- A good model for physical variables like weight, height, temperature, etc.
- Sums and averages of arbitrarily distributed rvs are also normally distributed (Central Limit Theorem)
  - Thus, very popular for modelling errors
- Normal pdf:
  - $f_X(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp \left\{ \frac{-(x-\mu)^2}{2\sigma^2} \right\} , \quad -\infty < x < +\infty$

# Normal distribution

- The mean and the std. dev. are also called *location* and *scale* parameters.



# Standard Normal Distribution

- Any non-standard Normal rv  $X$  with  $\text{Normal}(\mu, \sigma)$  can be standardized as follows:
  - $Z = \text{Normal}(0,1) = \frac{X-\mu}{\sigma}$
  - and vice versa:  $X = \mu + \sigma Z$
  - $\rightarrow$  we only need the Standard Normal Distribution table only
- Example 4.11 – computing non-standard probabilities using the standard normal table
- Example 4.12 – solving inverse problems

# Central Limit Theorem

- Let  $X_1, \dots, X_n$  be random variables from **any** distribution with  $\mu = \mathbf{E}(X_i)$  and  $\sigma^2 = \mathbf{Var}(X_i)$  ( $n$  rvs from the same distribution)

As  $n \rightarrow \infty$ ,

$$\frac{(X_1 + \dots + X_n) - n\mu}{\sigma\sqrt{n}} \rightarrow \text{Normal}(0,1)$$

$$\rightarrow \mathbf{P} \left( \frac{(X_1 + \dots + X_n) - n\mu}{\sigma\sqrt{n}} \leq x \right) \rightarrow F_{\text{Normal}(0,1)}(x)$$

Examples:

$\text{Binomial}(n, p) \approx \text{Normal}(\mu, \sigma)$  for large  $n$

$\text{Gamma}(\alpha, \lambda) \approx \text{Normal}(\mu, \sigma)$  for large  $\alpha$

# Central Limit Theorem

- Example 4.13
- Example 4.14

# Normal Approximation to Binomial

- $\text{Binomial}(n, p) \approx \text{Normal}(\mu = np, \sigma = \sqrt{np(1 - p)})$
- We need continuity correction
  - $P(X = x) = 0$  for a continuous variable  $X$
  - If we want to find  $f_B(b)$  for a Binomial variable  $B$ 
    - $f_B(b) = P(B = b) = P(b - 0.5 < B < b + 0.5)$
    - We expand the interval for the discrete variable 0.5 units in each direction and use the Normal approximation to compute the probability of an interval, not the probability of a point.
- Example 4.15