

Lecture 8

Normal Random Variables

Text: Chapter 4

STAT 8010 Statistical Methods I
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Whitney Huang
Clemson University



Notes

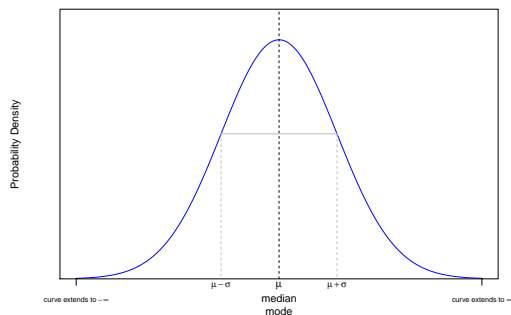
Agenda

- 1 Normal Density Curves
- 2 Standard Normal
- 3 Sums of Normal Random Variables



Notes

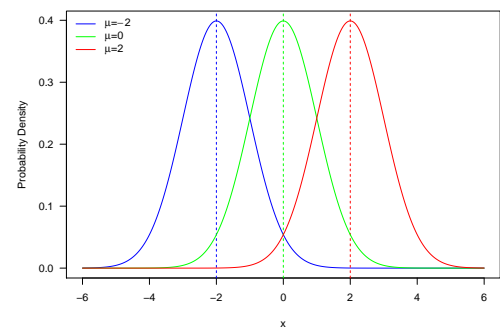
Probability Density Curve for Normal Random Variable



Notes

Normal Density Curves

Different μ but same σ^2



Normal Random Variables

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Normal Density Curves

Standard Normal

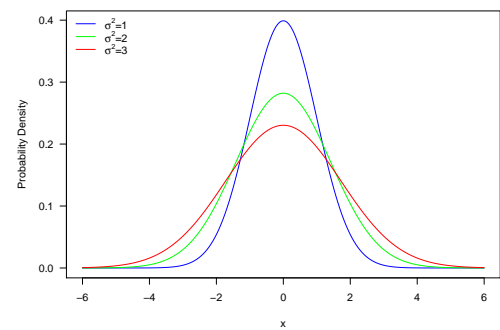
Sums of Normal Random Variables

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Notes

Normal Density Curves Cont'd

Same μ but different σ^2



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Normal Density Curves

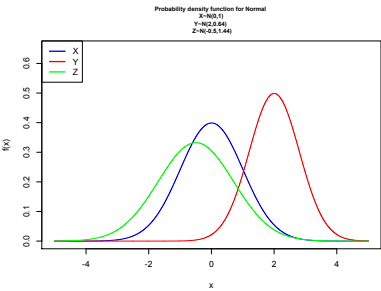
Standard Normal

Sums of Normal Random Variables

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Notes

Normal Density Curves



- The parameter μ determines the center of the distribution
- The parameter σ^2 determines the spread of the distribution
- Also called bell-shaped distribution

Normal Random Variables

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Normal Density Curves

Standard Normal

Sums of Normal Random Variables

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Notes

Characteristics of Normal Random Variables

Let X be a Normal r.v.

- The support for X : $(-\infty, \infty)$
- Parameters: μ : mean and σ^2 : variance
- The probability density function (pdf): $\frac{1}{\sqrt{2\pi\sigma^2}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}$ for $-\infty < x < \infty$
- The cumulative distribution function (cdf): No explicit form, look at the value $\Phi(\frac{x-\mu}{\sigma})$ for $-\infty < x < \infty$ from standard normal table
- The expected value: $\mathbb{E}[X] = \mu$
- The variance: $Var(X) = \sigma^2$

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Sums of Normal Random Variables

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Notes

Standard Normal $Z \sim N(\mu = 0, \sigma^2 = 1)$

- Normal random variable X with mean μ and standard deviation σ can convert to standard normal Z by the following :
$$Z = \frac{X - \mu}{\sigma}$$
- The cdf of the standard normal, denoted by $\Phi(z)$, can be found from the standard normal table
- The probability $\mathbb{P}(a \leq X \leq b)$ where $X \sim N(\mu, \sigma^2)$ can be compute

$$\begin{aligned} \mathbb{P}(a \leq X \leq b) &= \mathbb{P}\left(\frac{a - \mu}{\sigma} \leq Z \leq \frac{b - \mu}{\sigma}\right) \\ &= \Phi\left(\frac{b - \mu}{\sigma}\right) - \Phi\left(\frac{a - \mu}{\sigma}\right) \end{aligned}$$

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Normal Density Curves

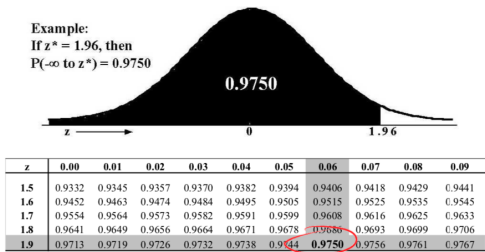
Standard Normal

Sums of Normal Random Variables

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Notes

Standard Normal (Z) Table



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Normal Density Curves

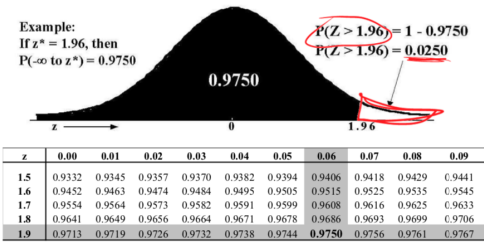
Standard Normal

Sums of Normal Random Variables

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Notes

Standard Normal (Z) Table Cont'd



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Normal Density Curves

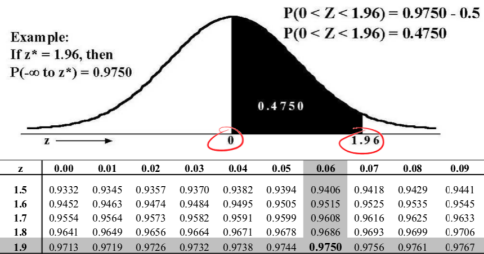
Standard Normal

Sums of Normal Random Variables

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Notes

Standard Normal (Z) Table Cont'd



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Normal Density Curves

Standard Normal

Sums of Normal Random Variables

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Notes

Properties of Φ

- $\Phi(0) = .50 \Rightarrow$ Mean and Median (50_{th} percentile) for standard normal are both 0
- $\Phi(-z) = 1 - \Phi(z)$
- $P(Z > z) = 1 - \Phi(z) = \Phi(-z)$

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Sums of Normal Random Variables

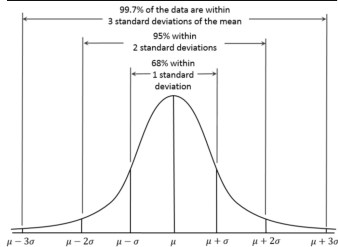
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Notes

The Empirical Rules

The Empirical Rules provide a quick way to approximate certain probabilities for the Normal Distribution as the following table:

Interval	Percentage with interval
$\mu \pm \sigma$	68%
$\mu \pm 2\sigma$	95%
$\mu \pm 3\sigma$	99.7%



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Sums of Normal Random Variables

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Notes

Example

Let us examine Z . Find the following probabilities with respect to Z :

- 1 Z is at most -1.75
- 2 Z is between -2 and 2 inclusive
- 3 Z is less than $.5$

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Sums of Normal Random Variables

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Notes

Example Cont'd

Solution.

- 1 $\mathbb{P}(Z \leq -1.75) = \Phi(-1.75) = .0401$
- 2 $\mathbb{P}(-2 \leq Z \leq 2) = \Phi(2) - \Phi(-2) = .9772 - .0228 = .9544$
- 3 $\mathbb{P}(Z < .5) = \Phi(.5) = .6915$

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Sums of Normal Random Variables

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Notes

Example

Suppose a STAT-8020 exam score follows a normal distribution with mean 78 and variance 36. Let X to denote the exam score, answer the following questions:

- 1 What is the probability that a randomly chosen test taker got a score greater than 84?
- 2 Suppose the passing score for this exam is 75. What is the probability that a randomly chosen test taker got a score greater than 84 given that she/he pass the exam?
- 3 Using the empirical rule to find the 84th percentile.

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Example

Find the following percentile with respect to Z

- 1 10th percentile
- 2 55th percentile
- 3 90th percentile

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Sums of Normal Random Variables

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Notes

Example Cont'd

Solution.

- 1 $Z_{10} = -1.28$
- 2 $Z_{55} = 0.13$
- 3 $Z_{90} = 1.28$

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Sums of Normal Random Variables

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Notes

Example

Let X be Normal with a mean of 20 and a variance of 49. Find the following probabilities and percentile:

- 1 X is between 15 and 23
- 2 X is more than 30
- 3 X is more than 12 knowing it is less than 20
- 4 What is the value that is smaller than 20% of the distribution?

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Sums of Normal Random Variables

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Notes

Example Cont'd

Solution.

- 1 $\mathbb{P}(15 \leq X \leq 23) = \Phi(\frac{15-20}{7}) - \Phi(\frac{23-20}{7}) = \Phi(0.43) - \Phi(-0.71) = .6664 - .2389 = .4275$
- 2 $\mathbb{P}(X > 30) = 1 - \mathbb{P}(X \leq 30) = 1 - \Phi(\frac{30-20}{7}) = 1 - .9236 = .0764$
- 3 $\mathbb{P}(X > 12 | X < 20) = \frac{\mathbb{P}(12 < X < 20)}{\mathbb{P}(X < 20)} = \frac{\Phi(0) - \Phi(-1.14)}{\Phi(0)} = .7458$
- 4 $Z_{.80} = 0.84 \Rightarrow X_{.80} = \mu + Z_{.80} \times \sigma = 20 + 0.84 \times \sqrt{49} = 25.88$

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Standard Normal

Sums of Normal Random Variables

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Notes

Sums of Normal Random Variables

If X_i $1 \leq i \leq n$ are independent normal random variables with mean μ_i , are variance σ_i^2 , respectively.

- Let $S_n = \sum_{i=1}^n X_i$ then $S_n \sim N(\sum_{i=1}^n \mu_i, \sum_{i=1}^n \sigma_i^2)$
- This can be applied for any integer n

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Sums of Normal Random Variables

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Notes

Example

Let $X_1, X_2,$ and X_3 be mutually independent, Normal random variables. Let their means and standard deviations be $3k$ and k for $k = 1, 2,$ and 3 respectively. Find the following distributions:

1 $\sum_{i=1}^3 X_i$

2 $X_1 + 2X_2 - 3X_3$

3 $X_1 + 5X_3$

Notes

Example Cont'd

Solution.

1 $\sum_{i=1}^3 X_i \sim N(\mu = 3+6+9 = 15, \sigma^2 = 1^2+2^2+3^2 = 14)$

2 $X_1 + 2X_2 - 3X_3 \sim N(\mu = 3 + 12 - 27 = -12, \sigma^2 = 1^2 + 4 \times 2^2 + 9 \times 3^2 = 98)$

3 $X_1 + 5X_3 \sim N(\mu = 3+45 = 48, \sigma^2 = 1^2+25 \times 3^2 = 226)$

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