

# Probability and Statistics with Programming

## Continuous Random Variables

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# Continuous Random Variables

- Concept of Continuous Random Variables
- PDF and CDF
- Expected Values
- Normal Distribution
- Uniform Distribution
- Exponential Distribution

# Continuous Random Variables

## Continuous RV: Example

If in the study of the ecology of a lake, we make depth measurements at randomly chosen locations, then  $X =$  the depth at such a location is a continuous rv. Here  $A$  is the minimum depth in the region being sampled, and  $B$  is the maximum depth. ■

If a chemical compound is randomly selected and its pH  $X$  is determined, then  $X$  is a continuous rv because any pH value between 0 and 14 is possible. If more is known about the compound selected for analysis, then the set of possible values might be a subinterval of  $[0, 14]$ , such as  $5.5 \leq x \leq 6.5$ , but  $X$  would still be continuous. ■

# Continuous Random Variables

## Probability density function (PDF)

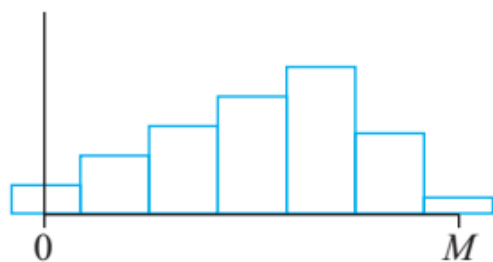
- Depth of a lake at a random location



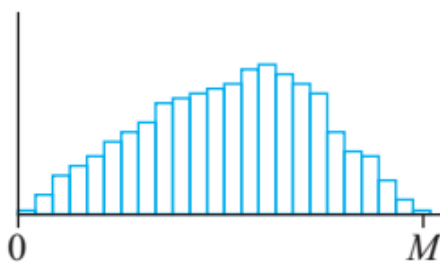
# Continuous Random Variables

## Probability density function (PDF)

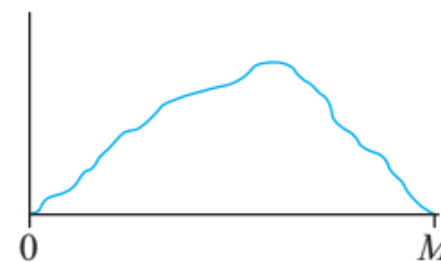
- Depth of a lake at a random location



(a)



(b)



(c)

# Continuous Random Variables

## Probability density function (PDF)

### DEFINITION

Let  $X$  be a continuous rv. Then a **probability distribution** or **probability density function** (pdf) of  $X$  is a function  $f(x)$  such that for any two numbers  $a$  and  $b$  with  $a \leq b$ ,

$$P(a \leq X \leq b) = \int_a^b f(x) dx$$

That is, the probability that  $X$  takes on a value in the interval  $[a, b]$  is the area above this interval and under the graph of the density function, as illustrated in Figure 4.2. The graph of  $f(x)$  is often referred to as the *density curve*.

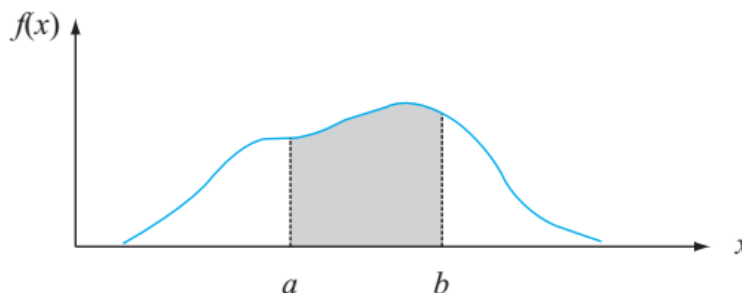


Figure 4.2  $P(a \leq X \leq b) =$  the area under the density curve between  $a$  and  $b$

# Continuous Random Variables

## Uniform Distribution

### DEFINITION

A continuous rv  $X$  is said to have a **uniform distribution** on the interval  $[A, B]$  if the pdf of  $X$  is

$$f(x, A, B) = \begin{cases} \frac{1}{B - A} & A \leq x \leq B \\ 0 & \text{otherwise} \end{cases}$$

# Continuous Random Variables

## Cumulative Distribution Function (CDF)

### DEFINITION

The **cumulative distribution function**  $F(x)$  for a continuous rv  $X$  is defined for every number  $x$  by

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(y) dy$$

For each  $x$ ,  $F(x)$  is the area under the density curve to the left of  $x$ . This is illustrated in Figure 4.5, where  $F(x)$  increases smoothly as  $x$  increases.

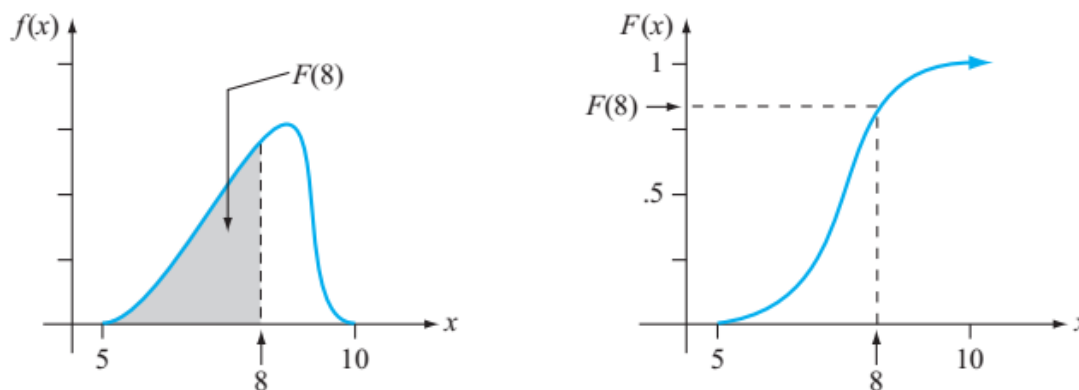


Figure 4.5 A pdf and associated cdf



# Continuous Random Variables

$f(X)$  from  $F(X)$

If  $X$  is a continuous rv with pdf  $f(x)$  and cdf  $F(x)$ , then at every  $x$  at which the derivative  $F'(x)$  exists,  $F'(x) = f(x)$ .

When  $X$  has a uniform distribution,  $F(x)$  is differentiable except at  $x = A$  and  $x = B$ , where the graph of  $F(x)$  has sharp corners. Since  $F(x) = 0$  for  $x < A$  and  $F(x) = 1$  for  $x > B$ ,  $F'(x) = 0 = f(x)$  for such  $x$ . For  $A < x < B$ ,

$$F'(x) = \frac{d}{dx} \left( \frac{x - A}{B - A} \right) = \frac{1}{B - A} = f(x)$$



# Continuous Random Variables

## Expected Value

### DEFINITION

The **expected** or **mean value** of a continuous rv  $X$  with pdf  $f(x)$  is

$$\mu_X = E(X) = \int_{-\infty}^{\infty} x \cdot f(x) dx$$

# Continuous Random Variables

## Example: PDF

The pdf of weekly gravel sales  $X$  was

$$f(x) = \begin{cases} \frac{3}{2}(1 - x^2) & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} E(X) &= \int_{-\infty}^{\infty} x \cdot f(x) \, dx = \int_0^1 x \cdot \frac{3}{2} (1 - x^2) \, dx \\ &= \frac{3}{2} \int_0^1 (x - x^3) \, dx = \frac{3}{2} \left( \frac{x^2}{2} - \frac{x^4}{4} \right) \bigg|_{x=0}^{x=1} = \frac{3}{8} \end{aligned}$$



# Continuous Random Variables

## Variance

### DEFINITION

The **variance** of a continuous random variable  $X$  with pdf  $f(x)$  and mean value  $\mu$  is

$$\sigma_X^2 = V(X) = \int_{-\infty}^{\infty} (x - \mu)^2 \cdot f(x) dx = E[(X - \mu)^2]$$

The **standard deviation** (SD) of  $X$  is  $\sigma_X = \sqrt{V(X)}$ .

### PROPOSITION

$$V(X) = E(X^2) - [E(X)]^2$$

# Continuous Random Variables

## Exercise: PDF

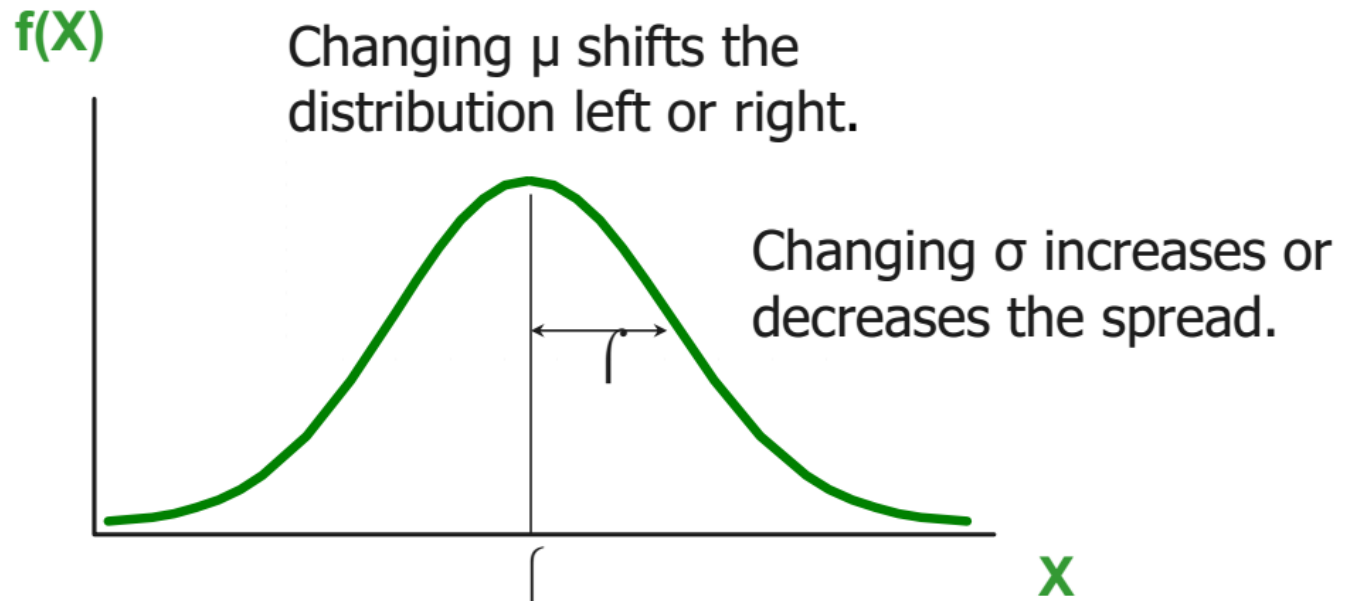
The current in a certain circuit as measured by an ammeter is a continuous random variable  $X$  with the following density function:

$$f(x) = \begin{cases} .075x + .2 & 3 \leq x \leq 5 \\ 0 & \text{otherwise} \end{cases}$$

- a.** Graph the pdf and verify that the total area under the density curve is indeed 1.
- b.** Calculate  $P(X \leq 4)$ . How does this probability compare to  $P(X < 4)$ ?
- c.** Calculate  $P(3.5 \leq X \leq 4.5)$  and also  $P(4.5 < X)$ .

# Continuous Random Variables

Normal Distribution (i.e., Gaussian Distribution)



# Continuous Random Variables

## Normal Distribution (i.e., Gaussian Distribution)

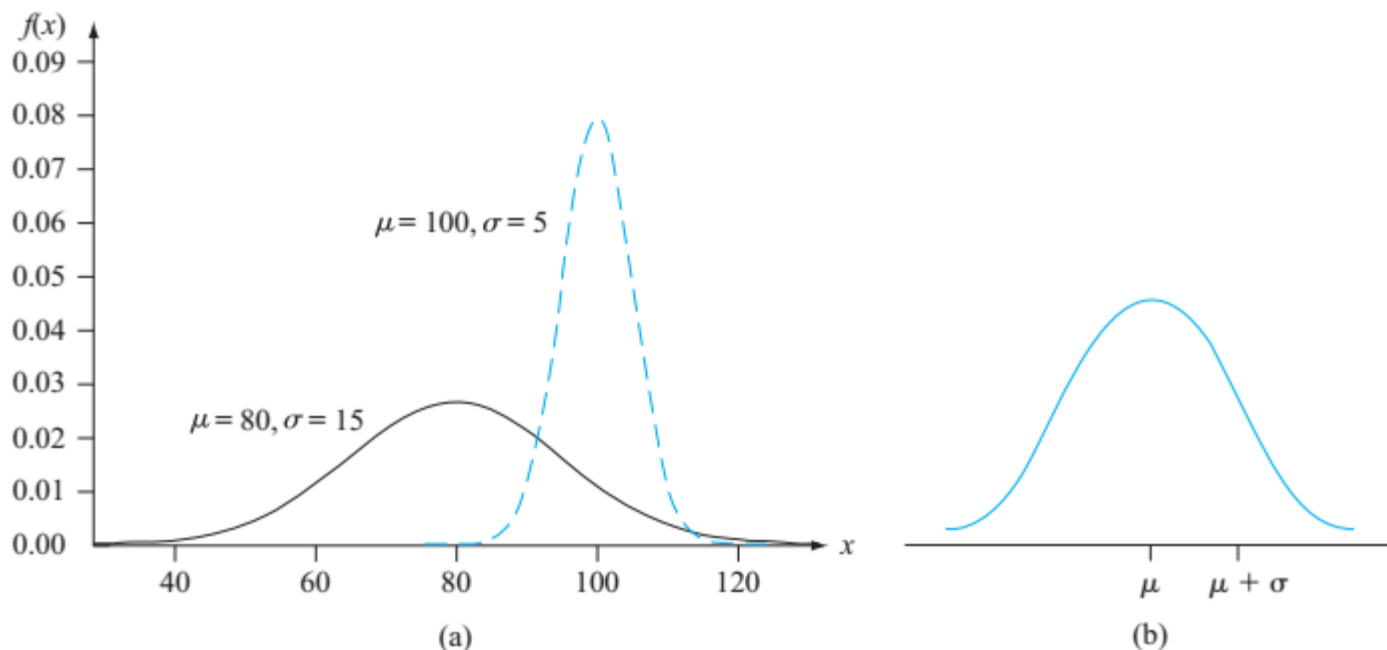
### DEFINITION

A continuous rv  $X$  is said to have a **normal distribution** with parameters  $\mu$  and  $\sigma$  (or  $\mu$  and  $\sigma^2$ ), where  $-\infty < \mu < \infty$  and  $0 < \sigma$ , if the pdf of  $X$  is

$$f(x, \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/(2\sigma^2)} \quad -\infty < x < \infty \quad (4.3)$$

# Continuous Random Variables

## Normal Distribution (i.e., Gaussian Distribution)



**Figure 4.13** (a) Two different normal density curves (b) Visualizing  $\mu$  and  $\sigma$  for a normal distribution



# Continuous Random Variables

Normal Distribution (i.e., Gaussian Distribution)

It's a probability function, so no matter what the values of  $\mu$  and  $\sigma$ , must integrate to 1!

$$\int_{-\infty}^{+\infty} \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx = 1$$

# Continuous Random Variables

Normal Distribution (i.e., Gaussian Distribution)

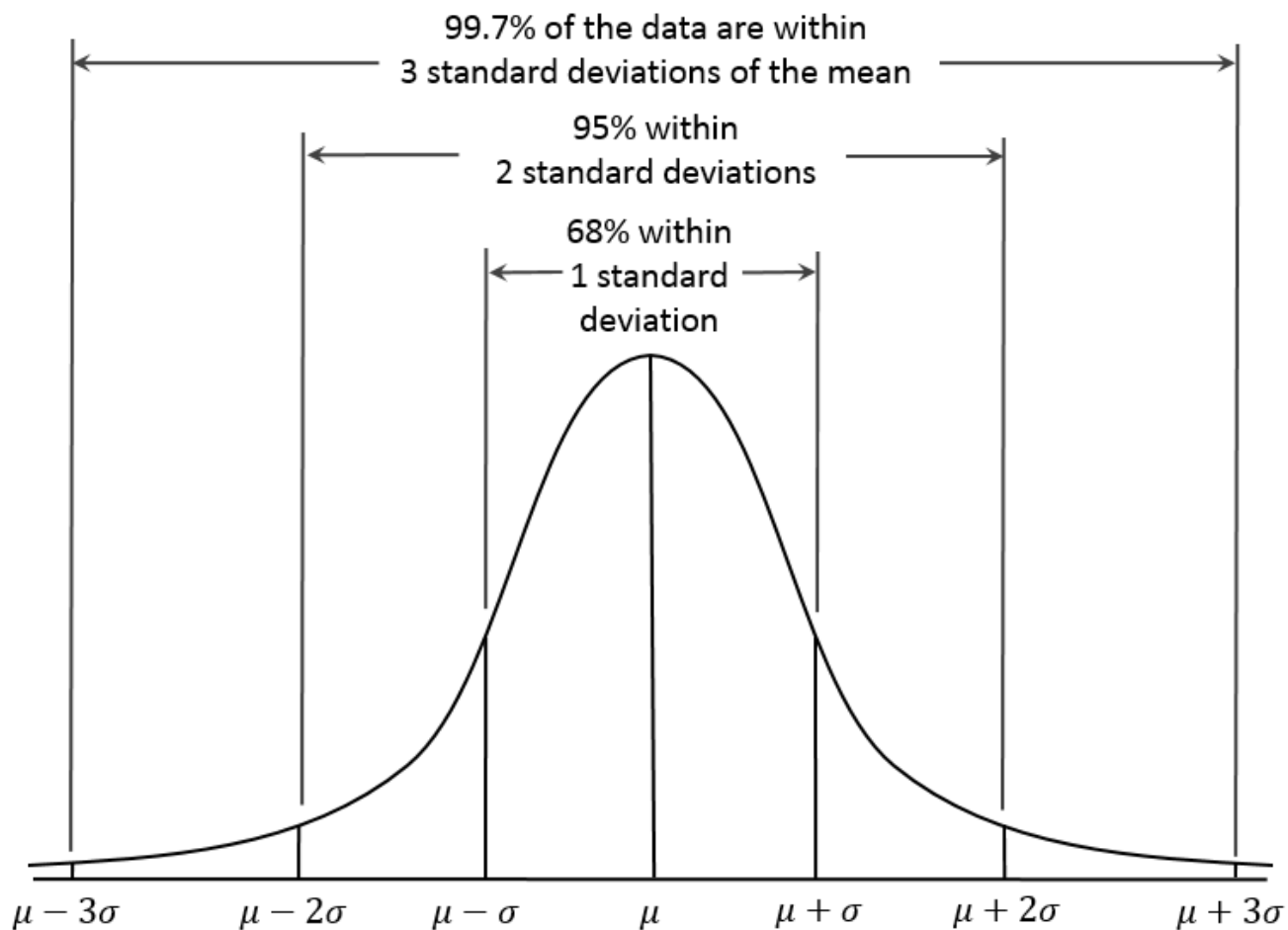
$$E(X)=\mu = \int_{-\infty}^{+\infty} x \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

$$\text{Var}(X)=\sigma^2 = \int_{-\infty}^{+\infty} x^2 \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx - \mu^2$$

$$\text{Standard Deviation}(X)=\sigma$$

# Continuous Random Variables

## Normal Distribution (i.e., Gaussian Distribution)



# Continuous Random Variables

Normal Distribution (i.e., Gaussian Distribution)

$$\int_{\mu-\sigma}^{\mu+\sigma} \frac{1}{\sigma\sqrt{2\pi}} \bullet e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx = .68$$

$$\int_{\mu-2\sigma}^{\mu+2\sigma} \frac{1}{\sigma\sqrt{2\pi}} \bullet e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx = .95$$

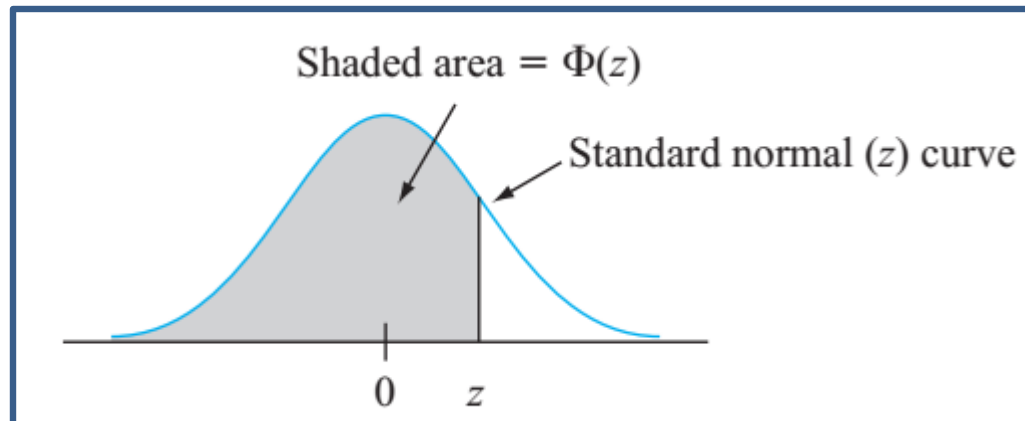
$$\int_{\mu-3\sigma}^{\mu+3\sigma} \frac{1}{\sigma\sqrt{2\pi}} \bullet e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx = .997$$

# Continuous Random Variables

## Standard Normal Distribution

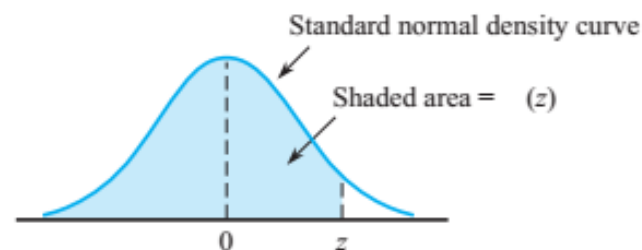
The normal distribution with parameter values  $\mu = 0$  and  $\sigma = 1$  is called the **standard normal distribution**. A random variable having a standard normal distribution is called a **standard normal random variable** and will be denoted by  $Z$ . The pdf of  $Z$  is

$$f(z, 0, 1) = \frac{1}{\sqrt{2\pi}} e^{-z^2/2} \quad -\infty < z < \infty$$



**Table A.3 Standard Normal Curve Areas**

$$(z) = P(Z \leq z)$$



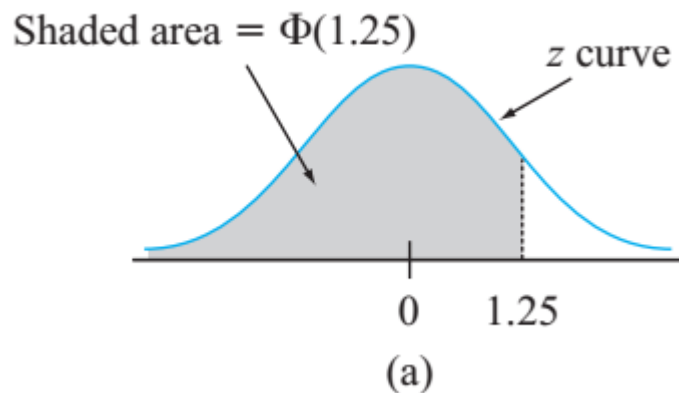
$z$	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
-3.4	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0002
-3.3	.0005	.0005	.0005	.0004	.0004	.0004	.0004	.0004	.0004	.0003
-3.2	.0007	.0007	.0006	.0006	.0006	.0006	.0006	.0005	.0005	.0005
-3.1	.0010	.0009	.0009	.0009	.0008	.0008	.0008	.0008	.0007	.0007
-3.0	.0013	.0013	.0013	.0012	.0012	.0011	.0011	.0011	.0010	.0010
-2.9	.0019	.0018	.0017	.0017	.0016	.0016	.0015	.0015	.0014	.0014
-2.8	.0026	.0025	.0024	.0023	.0023	.0022	.0021	.0021	.0020	.0019
-2.7	.0035	.0034	.0033	.0032	.0031	.0030	.0029	.0028	.0027	.0026
-2.6	.0047	.0045	.0044	.0043	.0041	.0040	.0039	.0038	.0037	.0036
-2.5	.0062	.0060	.0059	.0057	.0055	.0054	.0052	.0051	.0049	.0038
-2.4	.0082	.0080	.0078	.0075	.0073	.0071	.0069	.0068	.0066	.0064
-2.3	.0107	.0104	.0102	.0099	.0096	.0094	.0091	.0089	.0087	.0084
-2.2	.0139	.0136	.0132	.0129	.0125	.0122	.0119	.0116	.0113	.0110
-2.1	.0179	.0174	.0170	.0166	.0162	.0158	.0154	.0150	.0146	.0143
-2.0	.0228	.0222	.0217	.0212	.0207	.0202	.0197	.0192	.0188	.0183
-1.9	.0287	.0281	.0274	.0268	.0262	.0256	.0250	.0244	.0239	.0233
-1.8	.0359	.0352	.0344	.0336	.0329	.0322	.0314	.0307	.0301	.0294
-1.7	.0446	.0436	.0427	.0418	.0409	.0401	.0392	.0384	.0375	.0367
-1.6	.0548	.0537	.0526	.0516	.0505	.0495	.0485	.0475	.0465	.0455
-1.5	.0668	.0655	.0643	.0630	.0618	.0606	.0594	.0582	.0571	.0559
-1.4	.0808	.0793	.0778	.0764	.0749	.0735	.0722	.0708	.0694	.0681
-1.3	.0968	.0951	.0934	.0918	.0901	.0885	.0869	.0853	.0838	.0823
-1.2	.1151	.1131	.1112	.1093	.1075	.1056	.1038	.1020	.1003	.0985
-1.1	.1357	.1335	.1314	.1292	.1271	.1251	.1230	.1210	.1190	.1170
-1.0	.1587	.1562	.1539	.1515	.1492	.1469	.1446	.1423	.1401	.1379

[illegible]

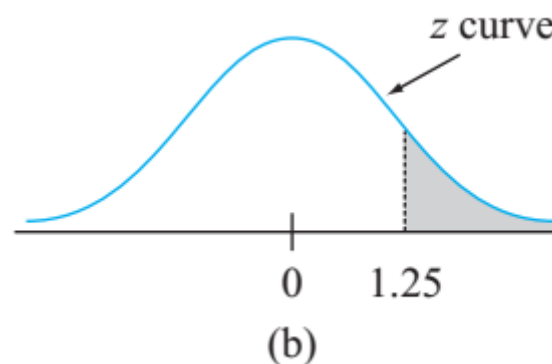
# Continuous Random Variables

## Example: Standard Normal Distribution

Let's determine the following standard normal probabilities: (a)  $P(Z \leq 1.25)$ , (b)  $P(Z > 1.25)$ , (c)  $P(Z \leq -1.25)$ , and (d)  $P(-.38 \leq Z \leq 1.25)$ .



$$P(Z \leq 1.25) = .8944.$$



$$P(Z > 1.25) = .1056.$$



# Continuous Random Variables

## Example: Standard Normal Distribution

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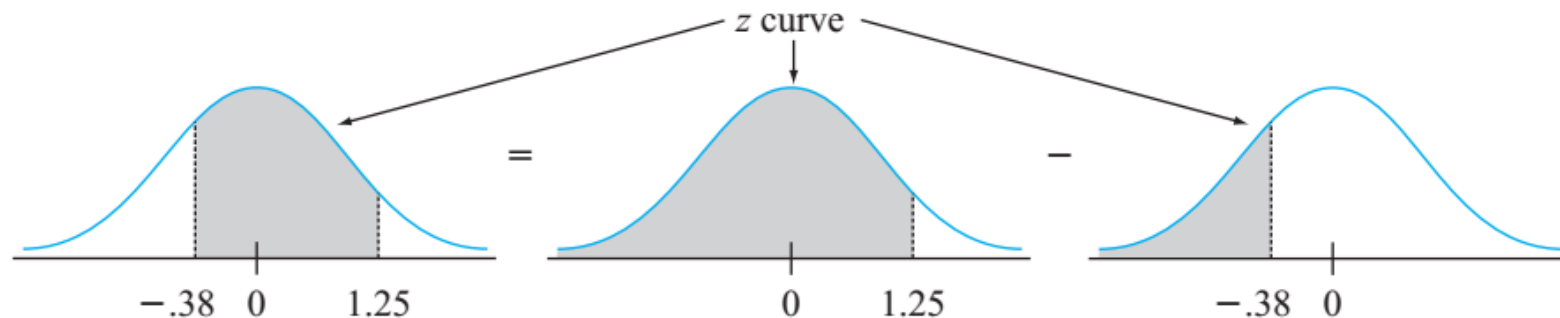


Figure 4.16  $P(-.38 \leq Z \leq 1.25)$  as the difference between two cumulative areas

# Continuous Random Variables

## Non-Standard Normal Distribution

If  $X$  has a normal distribution with mean  $\mu$  and standard deviation  $\sigma$ , then

$$Z = \frac{X - \mu}{\sigma}$$

has a standard normal distribution. Thus

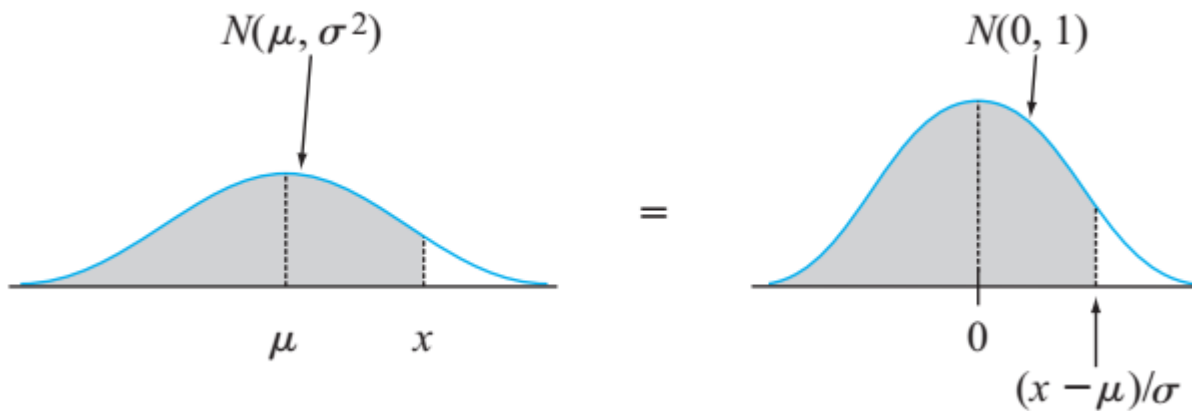
$$\begin{aligned} P(a \leq X \leq b) &= P\left(\frac{a - \mu}{\sigma} \leq Z \leq \frac{b - \mu}{\sigma}\right) \\ &= \Phi\left(\frac{b - \mu}{\sigma}\right) - \Phi\left(\frac{a - \mu}{\sigma}\right) \end{aligned}$$

$$P(X \leq a) = \Phi\left(\frac{a - \mu}{\sigma}\right) \quad P(X \geq b) = 1 - \Phi\left(\frac{b - \mu}{\sigma}\right)$$

# Continuous Random Variables

## Non-Standard Normal Distribution

$$P(Z \leq z) = P(X \leq \sigma Z + \mu) = \int_{-\infty}^{\sigma z + \mu} f(x; \mu, \sigma) dx$$



# Q&A

