(Ch 4.3)

The <u>normal distribution</u> is probably the most important distribution in all of probability and statistics.

Many populations have distributions that can be fit very closely by an appropriate normal (or Gaussian, bell) curve.

Examples: height, weight, and other physical characteristics, scores on various tests, etc.

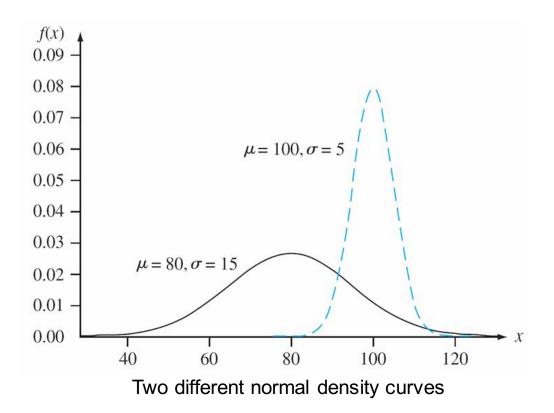
Definition

A continuous r.v. X is said to have a **normal distribution** with parameters μ and $\sigma > 0$ (or μ and σ^2), if the pdf of X is

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

The statement that X is normally distributed with parameters μ and σ^2 is often abbreviated $X \sim N(\mu, \sigma^2)$.

Figure below presents graphs of $f(x; \mu, \sigma)$ for several different (μ, σ) pairs.



 μ $\mu + \sigma$

Visualizing μ and σ for a normal distribution

The normal distribution with parameter values μ = 0 and σ = 1 is called the **standard normal distribution**.

A r.v. with this distribution is called a standard normal random variable and is denoted by Z. Its pdf is:

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

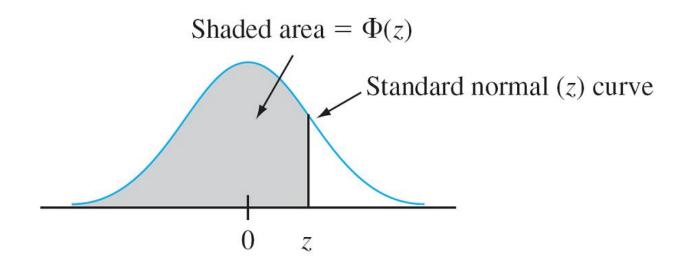
The normal distribution with parameter values μ = 0 and σ = 1 is called the **standard normal distribution**.

We use special notation to denote the cdf of the standard normal curve:

$$\Phi(z) = \int_{-\infty}^{z} f(y; 0, 1) \, dy$$

- The standard normal distribution <u>rarely</u> occurs naturally.
- Instead, it is a reference distribution from which information about other normal distributions can be obtained via a simple formula.
- These probabilities can then be found "normal tables".
- This can also be computed with a single command in R.

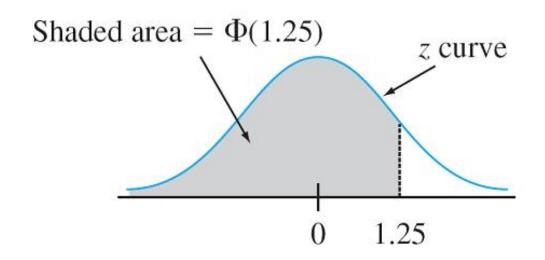
Figure below illustrates the probabilities found in a normal table (this can easily be found online):



 $P(Z \le 1.25) = \Phi(1.25)$, a probability that is tabulated in a normal table.

What is this probability?

The figure below illustrates this probability:



Example

- a) $P(Z \ge 1.25) = ?$
- b) Why does $P(Z \le -1.25) = P(Z >= 1.25)$? What is Φ (-1.25)?
- c) How do we calculate $P(-.38 \le Z \le 1.25)$?

Example

The 99th percentile of the standard normal distribution is that value of z such that the area under the z curve to the left of the value is 0.99.

Tables give for fixed z the area under the standard normal curve to the left of z, whereas now we have the area and want the value of z.

This is the "inverse" problem to $P(Z \le z) = ?$

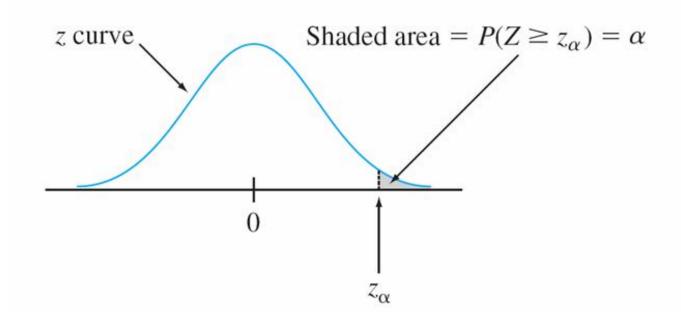
How can the table be used for this?

Notation: z_{α}

In statistical inference, we need the z values that give certain tail areas under the standard normal curve.

There, this notation will be standard:

 z_{α} will denote the z value for which α of the area under the z curve lies to the right of z_{α} .



z_{α} Notation for z Critical Values

For example, $z_{.10}$ captures upper-tail area .10, and $z_{.01}$ captures upper-tail area .01.

Since α of the area under the z curve lies to the right of z_{α} , $1 - \alpha$ of the area lies to its left.

Thus z_{α} is the 100(1 – α)th percentile of the standard normal distribution.

Similarly, what does $-z_{\alpha}$ mean?

Nonstandard Normal Distributions

When $X \sim N(\mu, \sigma^2)$, probabilities involving X are computed by "standardizing." The **standardized variable** is $(X - \mu)/\sigma$.

Subtracting μ shifts the mean from μ to zero, and then dividing by σ scales the variable so that the standard deviation is 1 rather than σ .

Proposition

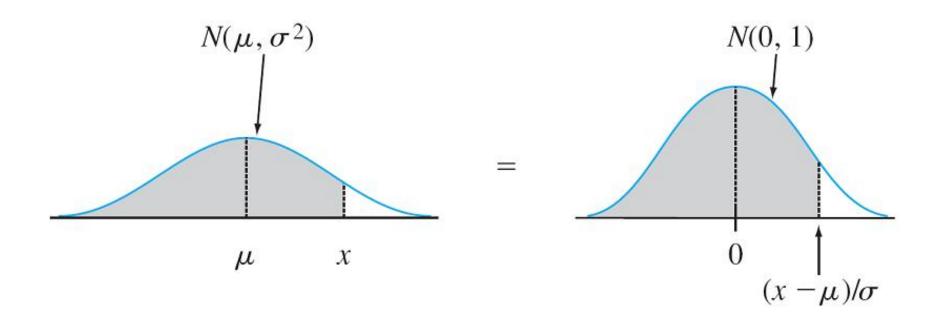
If X has a normal distribution with mean μ and standard deviation σ , then

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

is distributed standard normal.

Nonstandard Normal Distributions

Why do we standardize normal random variables?



Equality of nonstandard and standard normal curve areas

Example

The time that it takes a driver to react to the brake lights on a decelerating vehicle is critical in helping to avoid rear-end collisions.

The article "Fast-Rise Brake Lamp as a Collision-Prevention Device" (*Ergonomics*, 1993: 391–395) suggests that reaction time for an in-traffic response to a brake signal from standard brake lights can be modeled with a normal distribution having mean value 1.25 sec and standard deviation of .46 sec.

What is the probability that reaction time is between 1.00 sec and 1.75 sec?

We will revisit the normal distribution later on in this class to perform **statistical inference**.

THE NORMAL DISTRIBUTION IN R.