

# STOR 455

# **STATISTICAL METHODS I**

Jan Hannig

# Fundamental Concepts (Section 1.2)

- **Population**: the entire group of individuals that we want information about.
- **Sample**: a part of the population that we actually examine in order to gather information.
- **Sample size**: number of observations/individuals in a sample.
- **Statistical inference**: to make an inference about a population based on the information contained in a sample.

# Fundamental Concepts

- A *model* is mathematical description of the quantities of interest
  - Gaussian with unknown mean and variance
- A *parameter* is a value that describes the population. It's fixed but unknown in practice.
  - the mean and variance of the SAT score of all the students, who are about to take it.
- A *statistic* is a value that describes a sample. It's known once a sample is obtained.
  - The mean and variance SAT score of all the students, who are selected into the study.
  - A sample analogy of the parameter.
- **Statistics is a course about lots of statistics!!! 😊**

# Models (Section 1.4)

- There are many possible model distributions
  - Gaussian distribution
  - Binomial distribution
  - Poisson distribution
  - Gamma distribution
  - ...
- In this class we will almost exclusively use *Gaussian Distribution*

# Density Curve

- Define a probability density function  $f(x)$ .
- The curve that plots  $f(x)$  is called the corresponding density curve.
- $f(x)$  satisfies:
  - $f(x) \geq 0$ ;
  - The total area under the curve representing  $f(x)$  equals 1.
- Areas under the curve represent relative frequencies of observations

# Normal Distribution (Section 1.4)

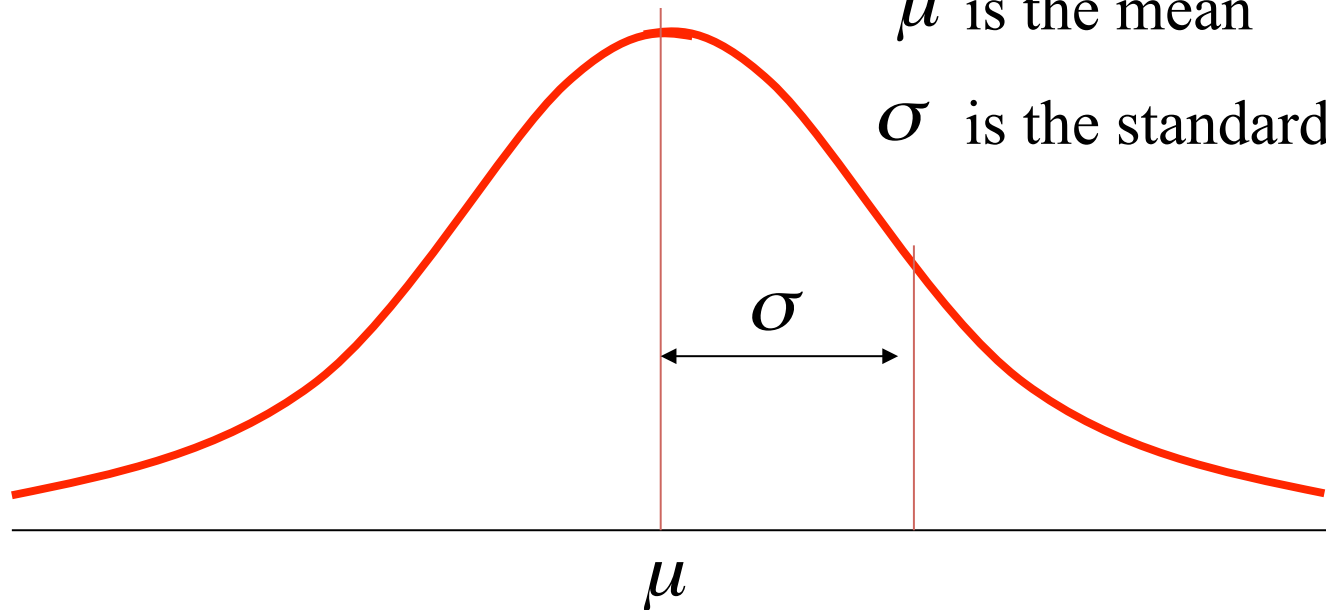
- Pictorially speaking, a Normal Distribution is a distribution that has a **symmetric, unimodal** and **bell-shaped** density curve.
- The mean and standard deviation completely specify the curve.
- The mean, median, and mode are the same.

The height of a normal density curve at any point  $x$  is given by

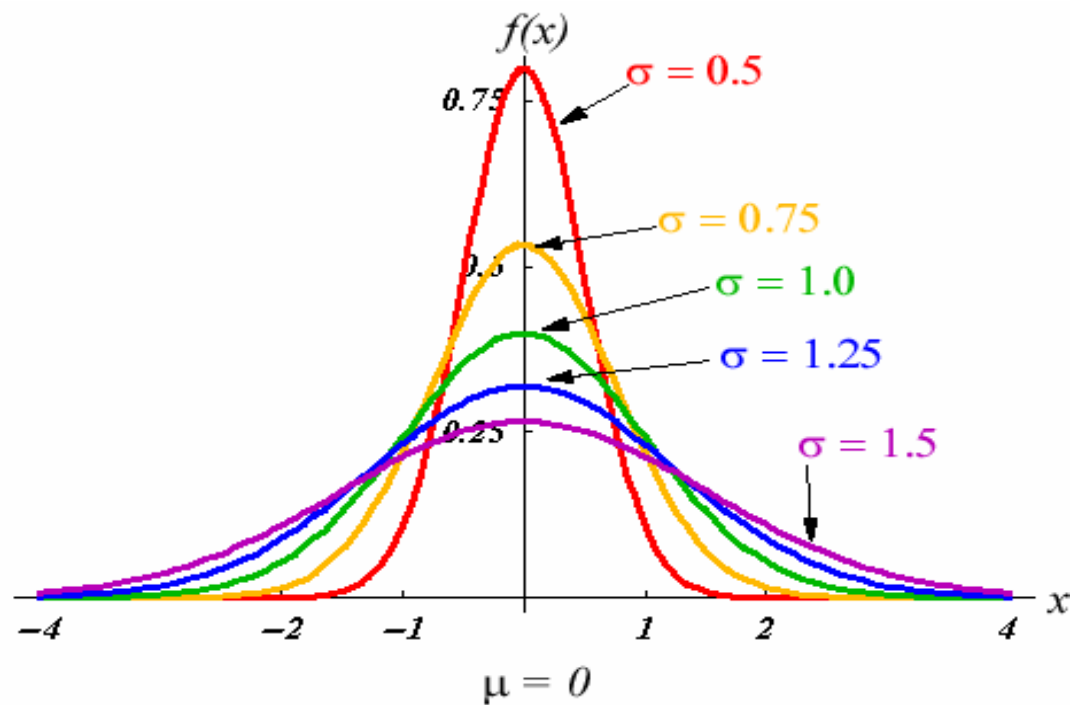
$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

$\mu$  is the mean

$\sigma$  is the standard deviation

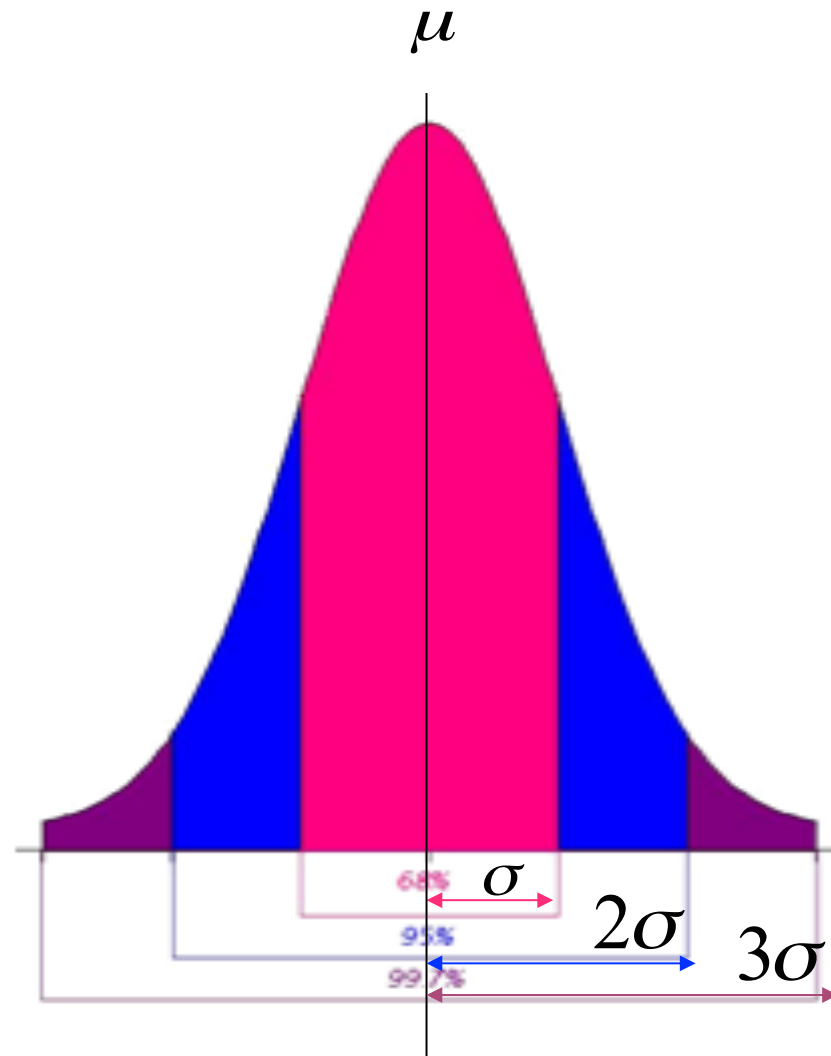


**Example:** The normal distribution is the most important distribution in Statistics. Typical normal curves with different sigma (standard deviation) values are shown below.





## The 68-95-99.7 Rule for General Normal Dist.



# Examples with approximate Normal distributions

- Heights
- Weights
- IQ scores
- Standardized test scores
- Body temperatures
- Repeated measurements of the same quantity
- ...

# Standardizing and z-Scores

- An observation  $x$  comes from a distribution with mean  $\mu$  and standard deviation  $\sigma$
- The standardized value of  $x$  is defined as

$$z = \frac{x - \mu}{\sigma},$$

which is also called a **z-score**.

- A z-score indicates how many standard deviations the original observation is away from the mean, and in which direction.
- Mean and S.D. of the distribution of  $z$ ?

# Effects of Standardizing

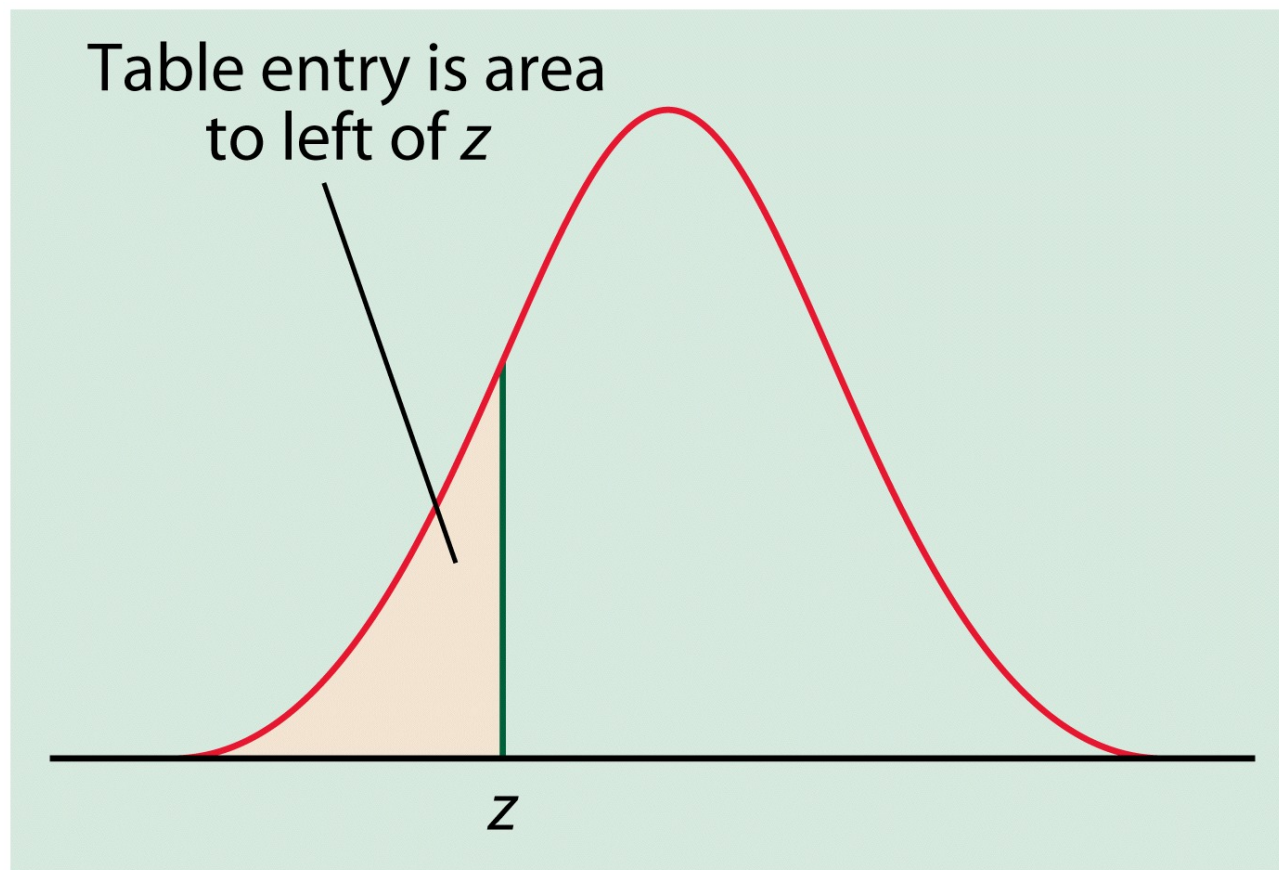
- Standardizing is a linear transformation. What are  $a$  and  $b$ ?
- Effects on shape, center and spread.
- The standardized values for any distribution always have mean  $0$  and standard deviation  $1$ .
- Linear transformation:  $normal$  into  $normal$ .

# The standard normal distribution

- The standard normal distribution is the normal dist. with mean 0 and standard deviation 1, denoted as  $N(0,1)$ .
- $N(0,1)$  can be treated as a baseline.
- Any normal distribution can be related to  $N(0,1)$  by a linear transformation.
- $Z: N(0,1)$ , what is the distribution for  $X=a+bZ$ ?

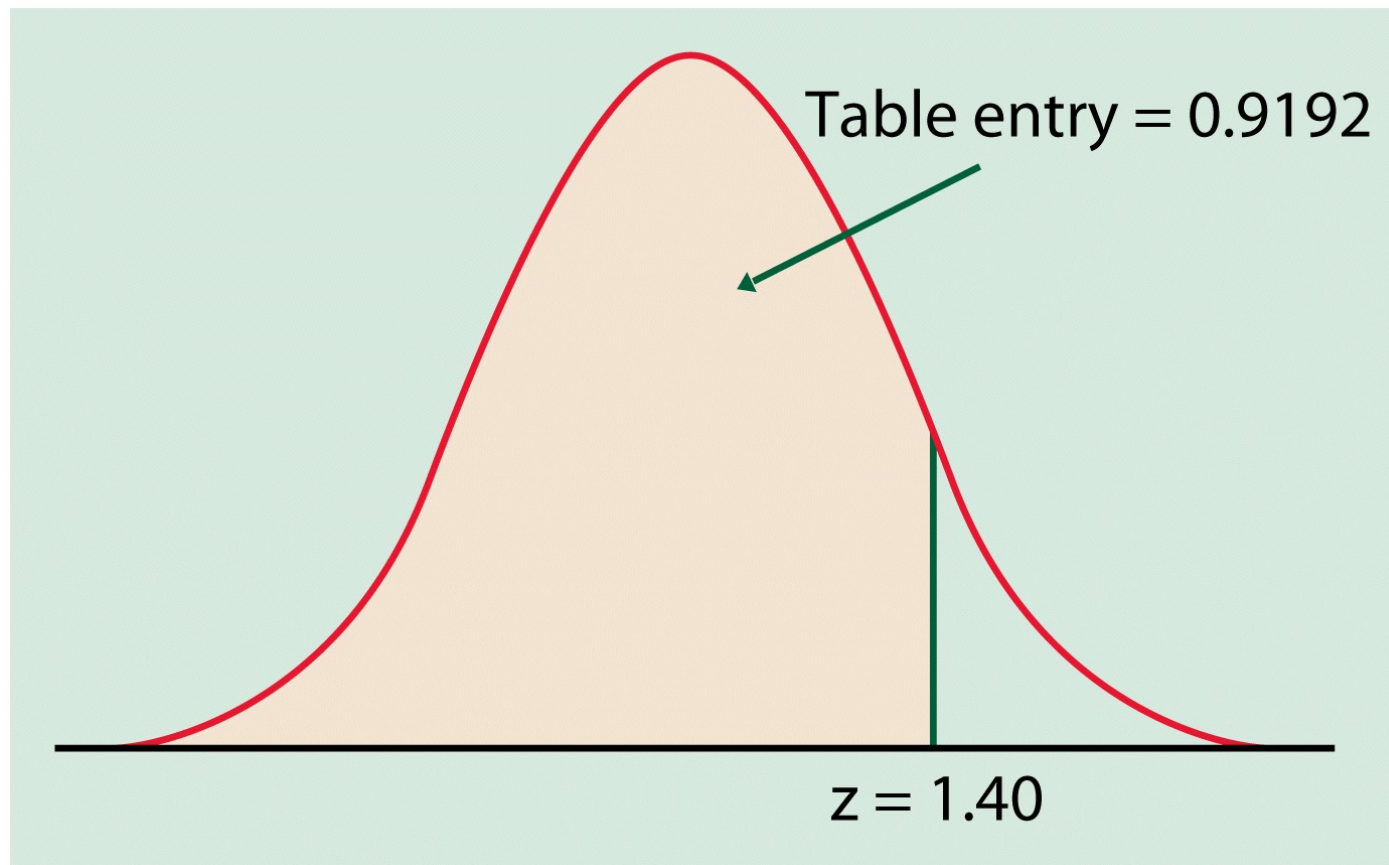
# Table: The Standard Normal Table

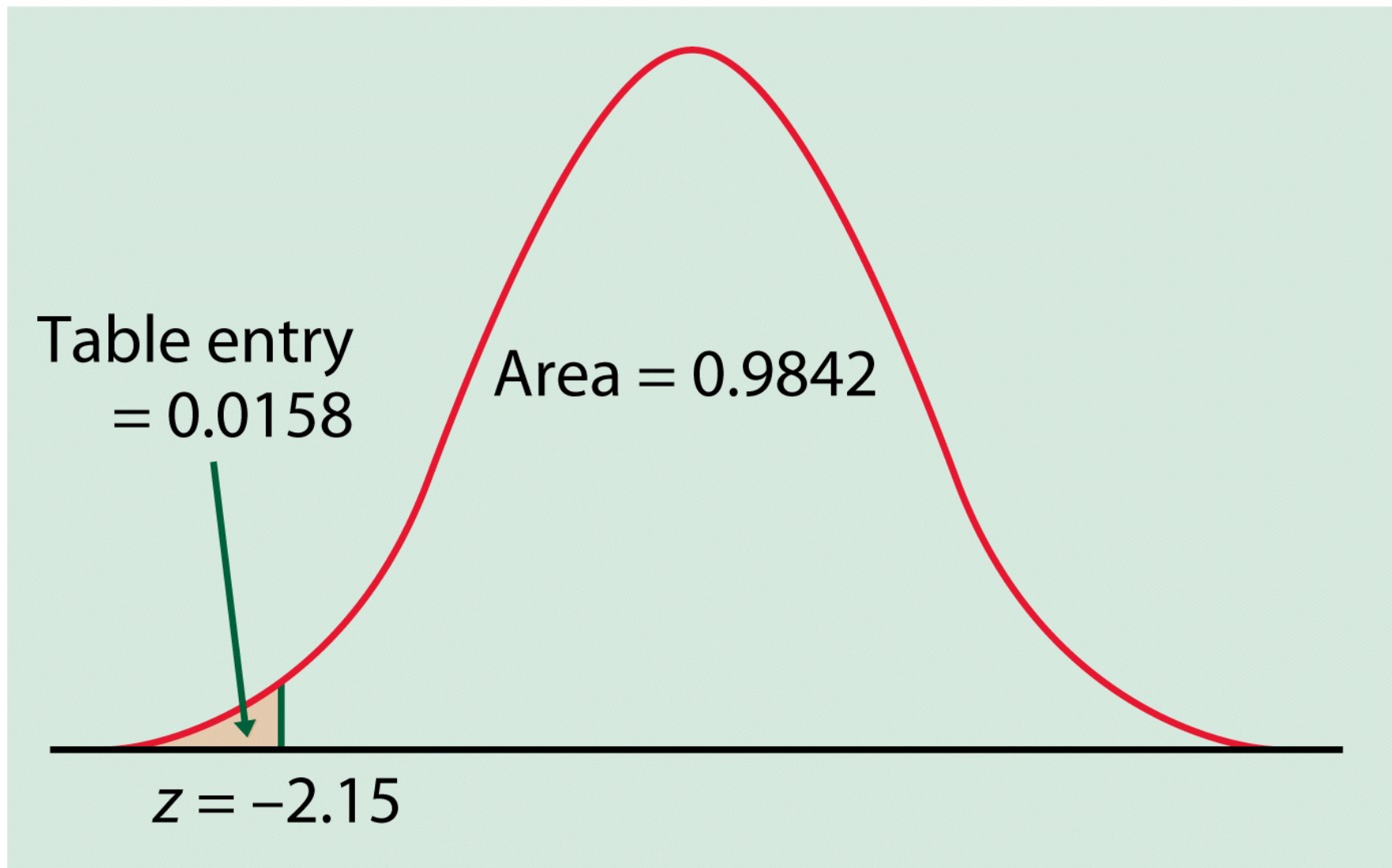
- **Table A** is a table of areas under the standard normal density curve. The table entry for each value  $z$  is the area under the curve to the left of  $z$ .



# Table: The Standard Normal Table

- **Table A** can be used to find the proportion of observations of a variable which fall to the left of a specific value  $z$  if the variable follows a normal distribution.







# Example

- According to well-documented norms, the distribution of gestation time is approximately normal with mean 266 days and SD 16 days.
- What percent of babies have a gestation time greater or equal to 310 days (10 months and 5 days)?
- What about within a week of the due date?

# Inverse problem

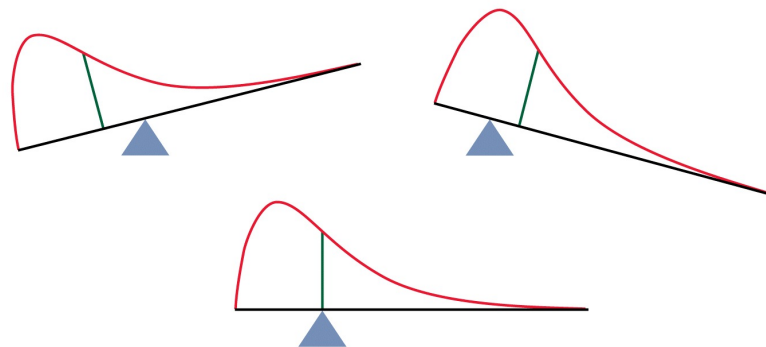
- Scores on the SAT verbal test in recent years follow approximately the  $N(505, 110)$  distribution.
- How high must a student score in order to be placed in the top 10% of all students taking the SAT?

# Parameters (Section 1.5)

- In general parameters are numerical summaries of the population.
- They are usually denoted by Greek letters, e.g.,  $\mu$ ,  $\sigma$ ,  $\rho$ ,  $\xi$ ,  $\theta$ ,  $\eta$ .
- Describe both univariate and multivariate populations

# Parameters

- **Mean**
  - If the population consists of equally likely numbers  $(y_1, \dots, y_N)$  then  $\mu = \frac{1}{N} \sum_{i=1}^N Y_i$  [Notice the upper case]
  - If not equally likely  $\mu = \sum_{i=1}^N Y_i p_i$  where  $p_i$  is the probability. (Explain on a picture)



# Parameters

- **Standard Deviation**

- If the population consists of equally likely numbers  $(Y_1, \dots, Y_N)$  then

$$\sigma = \sqrt{\frac{1}{N} \sum_{i=1}^N (y_i - \mu)^2}$$

- If not equally likely

$$\sigma = \sqrt{\sum_{i=1}^N (Y_i - \mu)^2 p_i}$$

- Variance = (standard deviation)<sup>2</sup>

# Multivariate Distributions

- Schematic representation

	<b><math>k</math> Measurements on Each Item</b>			
<b>Items</b>	<b>1</b>	<b>2</b>	<b>...</b>	<b><math>k</math></b>
1	$X_{11}$	$X_{12}$	...	$X_{1k}$
2	$X_{21}$	$X_{22}$	...	$X_{2k}$
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$I$	$X_{I1}$	$X_{I2}$	...	$X_{Ik}$
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$N$	$X_{N1}$	$X_{N2}$	...	$X_{Nk}$
<b>Mean</b>	$\mu_1$	$\mu_2$	...	$\mu_k$
<b>Standard deviation</b>	$\sigma_1$	$\sigma_2$	...	$\sigma_k$

- Each measurement has its own mean and s.d.
- Each pair of measurements has a correlation

# Multivariate Distributions

- Linear relationship between measurements are described by correlations.

- If items are equally likely

$$\rho_{Y,X} = \frac{\sum_{I=1}^N (Y_I - \mu_Y)(X_I - \mu_X)}{\sqrt{\left[\sum_{I=1}^N (Y_I - \mu_Y)^2\right] \left[\sum_{I=1}^N (X_I - \mu_X)^2\right]}}$$

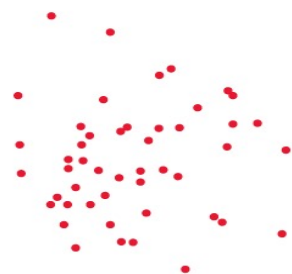
- Items are not equally likely

$$\rho_{Y,X} = \frac{\sum_{i=1}^N (Y_i - \mu_Y)(X_i - \mu_X)p_i}{\sigma_X \sigma_Y}$$

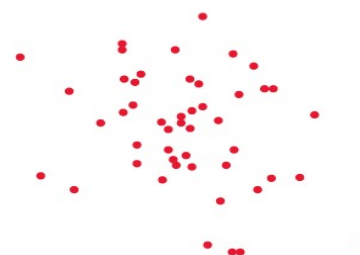
- Meaning

- Always  $-1 \leq \rho \leq 1$ , if independent  $\rho=0$

# Meaning of correlation



Correlation  $r = 0$



Correlation  $r = -0.3$



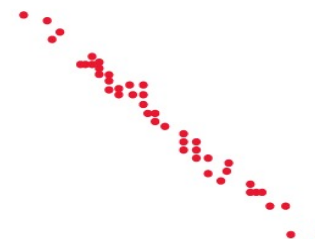
Correlation  $r = 0.5$



Correlation  $r = -0.7$



Correlation  $r = 0.9$



Correlation  $r = -0.99$



# Samples (Section 1.6)

- Populations are often unavailable to study directly.
- Solution:
  - Select a subsets of items ( $y_1, \dots, y_n$ ) [Notice the lower case]
  - If the selection is done through a well defined random procedure, a *measure of uncertainty* can be calculated.
- Possible selection procedures
  - *Simple random sampling*
  - Other methods: stratified sampling, probability proportional to size sampling, ...

# Simple random sampling (SRS)

- In SRS, all the samples with the same size are equally likely to be chosen.
  - Eliminate bias in a sample.
- To conduct a SRS
  - give each unit a number
  - randomly select the sample numbers. Use a random digits table, or a software package.
- Suppose we have 100 students, need to choose 4 to answer a set of questions.
  - randomly select four number between 00 and 99
- If there are  $N$  units in the population and we want to select  $n$  we have possible samples  $H = \binom{N}{n} = \frac{N \times (N - 1) \times \cdots \times (N - n + 1)}{1 \times 2 \times \cdots \times n}$

# Inferential Procedures

- Once the sample is selected, inference about the sample is performed
- Types of inference
  - Point Estimation
  - Confidence Intervals
  - Hypothesis Testing