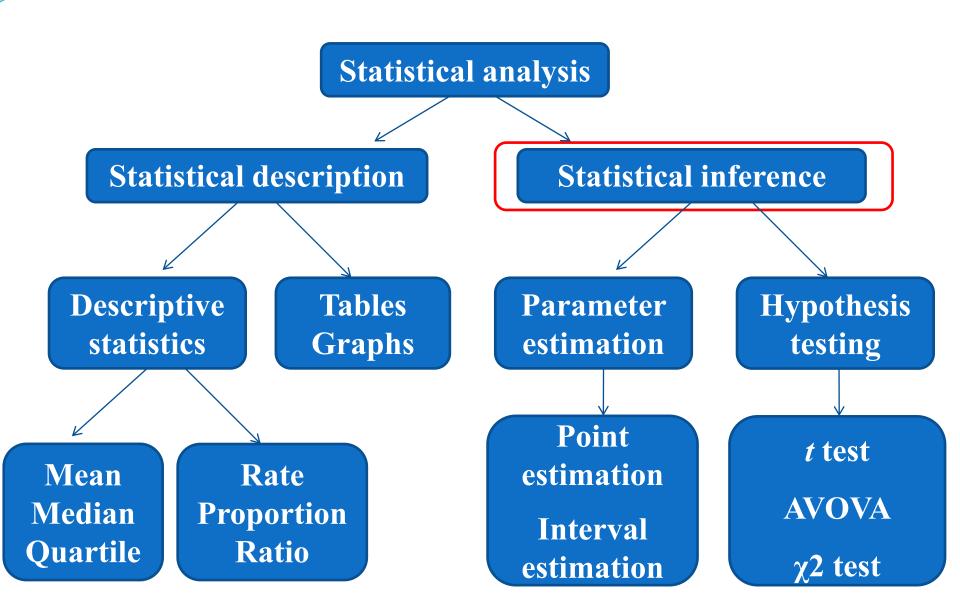


# Procedures of Statistical Analysis



## What is sampling distribution?

Distribution of sample statistics

## Main contents

1

 Sampling distribution of sample means (sampling error)

2

 Sampling distribution of sample rates

3

• t distribution

# 1. Sampling distribution of sample means

# Sampling study

parameter?  $\mu, \sigma, \pi$ 

population



statistic  $\bar{x}$ , s, p

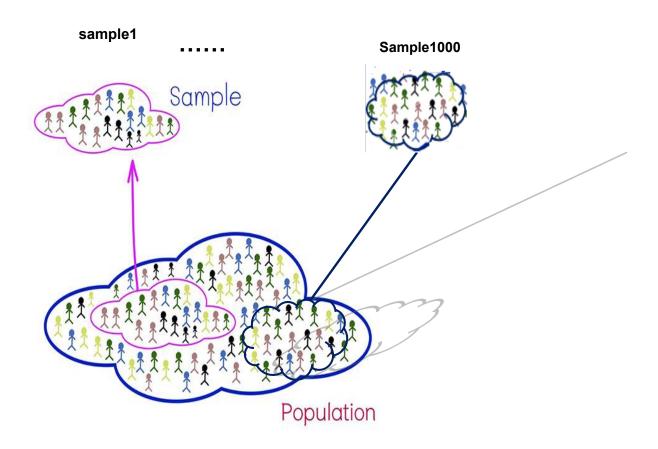
sample







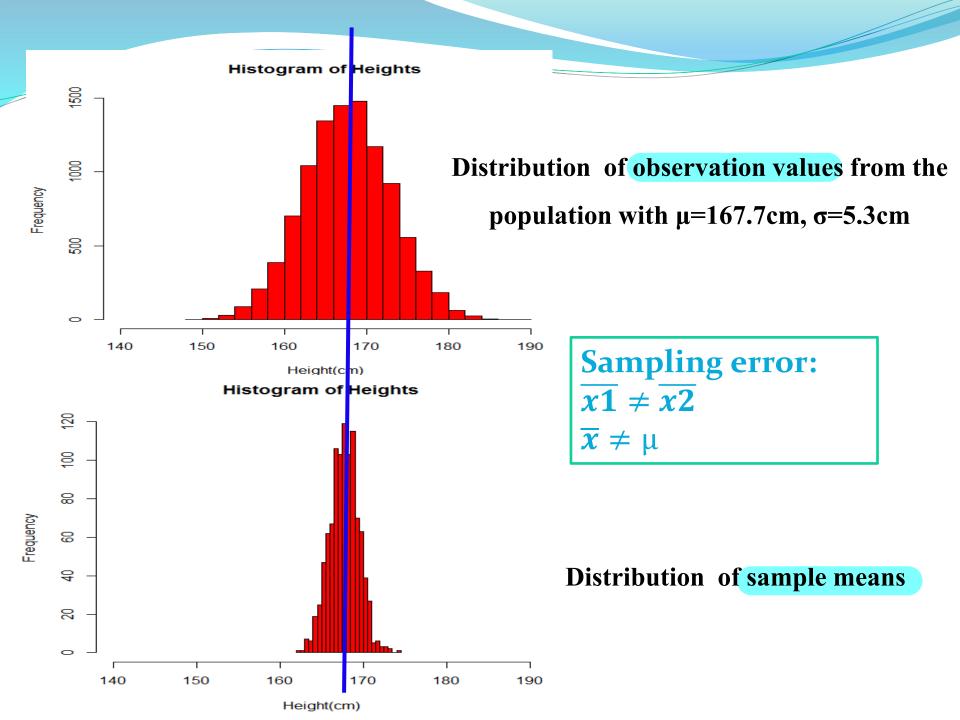
# Sampling error



Female height(cm): X~N(167,5.3)

A sampling error occurs when the sample used in the study is not representative of the whole population.

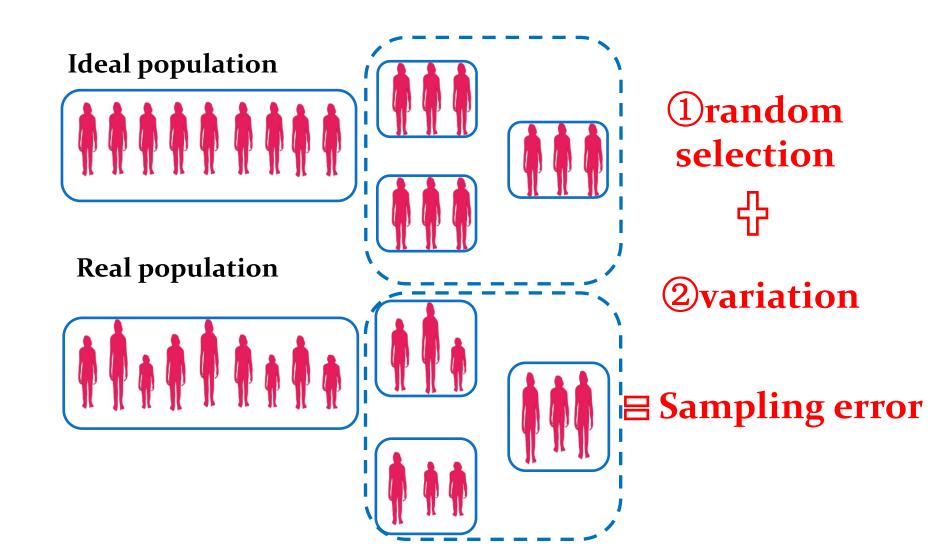
$$\bar{x} = 167.5$$
  $S_{\bar{x}} = 1.67$ 



#### Characteristics of the distribution of sample means:

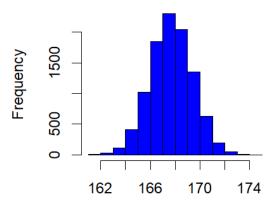
- The sample mean dose not necessarily equals the population mean  $\mu$ ;
- ✓ There are differences among sample means;
- ✓ The sample means are in the normal distribution;
- $\checkmark$  The mean of sample means is equal to μ;
- ✓ The deviation of sample mean is <u>smaller</u> than that of the original variable.

# sampling error can't be deleted

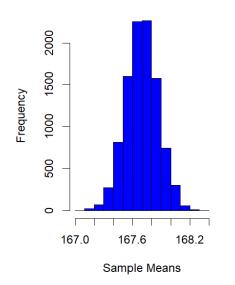


# The rule of sampling error

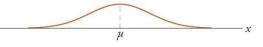
#### Sampling Distribution(n=10)



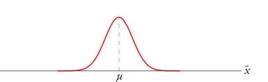
#### Sampling Distribution(n=1000)







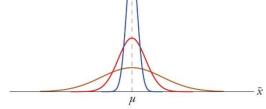
Sampling distribution of  $\overline{X}$  with n = 5



#### $E(\overline{X})=\mu$ ;

The distribution of sample means, is much narrower than the distribution of raw observations.

Distributions superimposed



# Prove

	Mean	Variance
Adding: T = X + Y	$\mu_T = \mu_X + \mu_Y$	$\sigma_T^2 = \sigma_X^2 + \sigma_Y^2$
Subtracting: $D = X - Y$	$\mu_D = \mu_X - \mu_Y$	$\sigma_D^2 = \sigma_X^2 + \sigma_Y^2$

$$\mu_{\overline{X}} = \mu_{\underbrace{(x_1 + x_2 + \dots x_n)}{n}}$$

$$= \frac{1}{n} (\mu_{x_1} + \mu_{x_2} + \dots + \mu_{x_n}) = \frac{1}{n} (\mu + \mu + \dots + \mu) = \mu$$

## Prove

if 
$$X \sim N(\mu_1, \sigma_1^2) Y \sim N(\mu_2, \sigma_2^2)$$

$$\sigma_{\bar{x}}^2 = \sigma_{\frac{(x_1 + x_2 + ... + x_n)}{n}}^2$$

$$= (\frac{1}{n})^2 (\sigma_{x_1}^2 + \sigma_{x_2}^2 + .... + \sigma_{x_n}^2)$$

$$= (\frac{1}{n})^2 (\sigma^2 + \sigma^2 + ...)$$

$$= \frac{\sigma^2}{n}$$

#### Standard Error of mean (SEM)

For a random variable  $x(\mu, \sigma^2)$ , the mean of sample mean is still  $\mu$ , and the standard error of sample mean is

$$\sigma_{\overline{x}} = \frac{\sigma}{\sqrt{n}}$$

For a sample to estimate

$$S_{\overline{x}} = \frac{S}{\sqrt{n}}$$

#### Question:

What are the differences between standard deviation and standard error of mean?



# Differences between standard deviation and standard error of mean

- > Standard error of mean reflects the variation of sample means and indicates the sampling error; standard deviation reflects the variation of individuals.
- > The sign for standard error of mean is:  $\sigma_{\bar{X}}$ ,  $S_{\bar{X}}$  and the sign for standard deviation is  $\sigma$ , S.
- > Standard error of mean can be calculated by  $\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$ ,  $S_{\bar{X}} = \frac{S}{\sqrt{n}}$
- > Standard error of mean can be decreased by increasing the sample size, but standard deviation can not be controlled.

#### **Definition of Sampling Error**

Due to the individual variations and random sampling, it results in:

- differences between sample statistics and population parameter;
- differences between sample statistic and sample statistic.

#### **Definition of Standard Error (SE)**

It is the standard deviation of sample statistics.

- **✓ Reflect the dispersion of sample statistics**
- ✓ Reflect the magnitude of sampling error

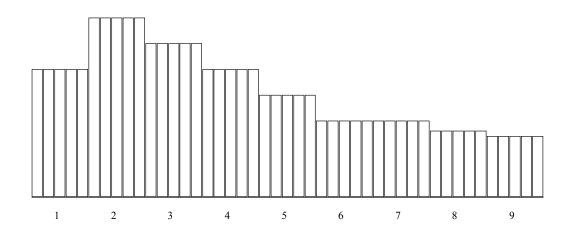
#### Definition of Standard Error of Mean (SEM)

It is the standard deviation of the sampling distribution of sample means.

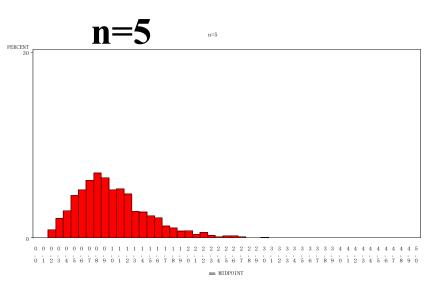
- **✓** Reflect the dispersion of sample means
- ✓ Reflect the magnitude of sampling error

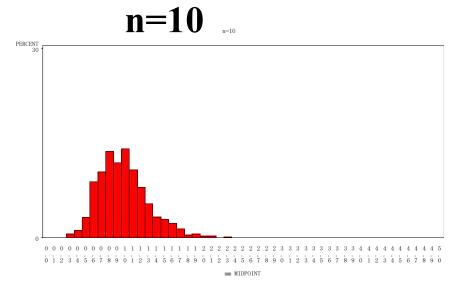
$$\sigma_{\overline{x}} = \frac{\sigma}{\sqrt{n}}$$
  $S_{\overline{x}} = \frac{S}{\sqrt{n}}$ 

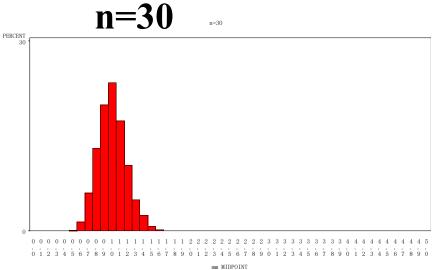
### The population is not the normal distribution?

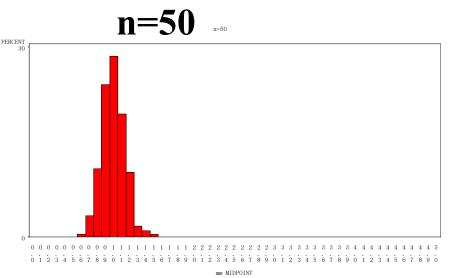


### Computer simulation results









#### Sampling distribution of sample means

1. 
$$\mu_{\bar{X}} = \mu_{X}$$

$$2. \ \sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$$

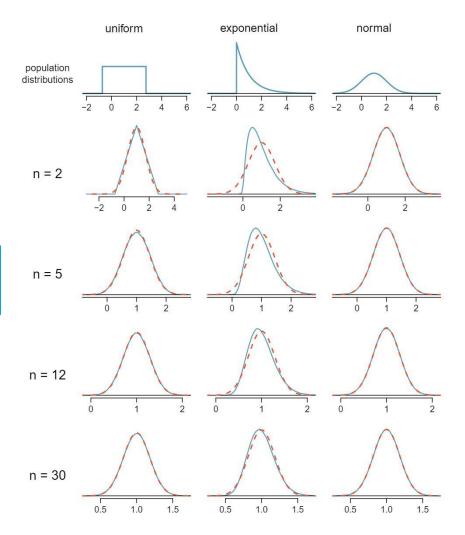
3. If X is normal, X is normal. If X is nonnormal,

X is approximately normally distributed for sufficiently large sample size.

# **Central Limit Theorem**

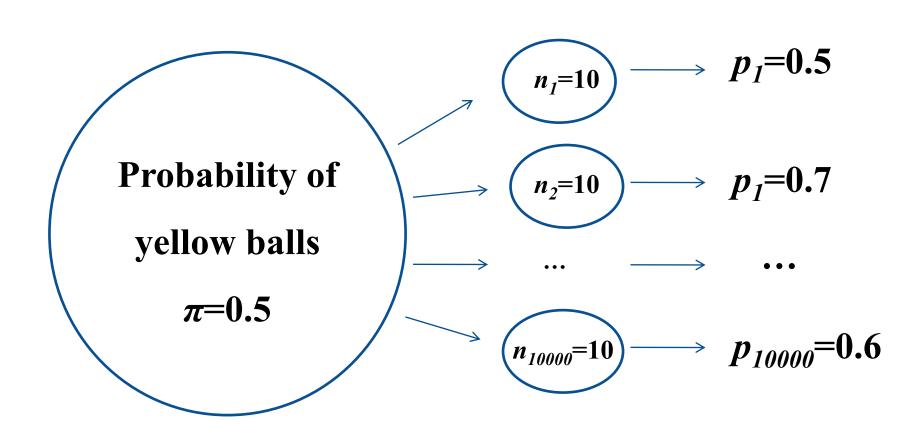
n is large enough

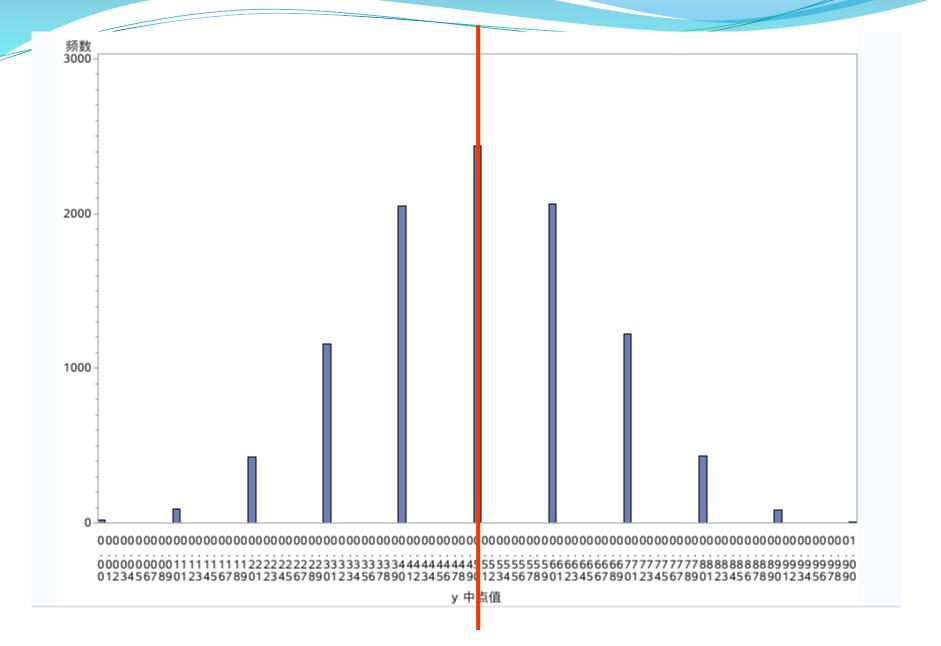
n>30



# 2. Sampling distribution of sample rates

## Example 2





## Sampling distribution of sample rates

- ✓ The mean of sample rates is equal to the population rate:  $\mu_n = \pi$
- ✓ The population standard deviation of sample rates (namely SER) is:

$$\sigma_p = \sqrt{\frac{\pi(1-\pi)}{n}}$$

✓ In practice, the population rate is unknown, so the sample rate is often used instead:

$$S_p = \sqrt{\frac{p(1-p)}{n}}$$

#### Definition of Standard Error of Rate (SER)

It is the standard deviation of the sampling distribution of sample rates.

- **✓** Reflect the dispersion of sample rates
- **✓** Reflect the sampling error of sample rates

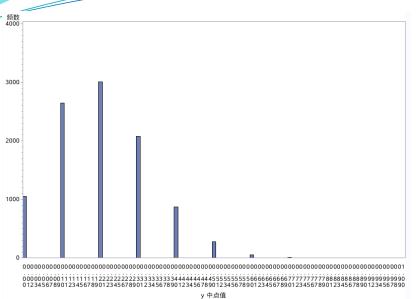
The population rate is not equal to 0.5  $(\pi \neq 0.5)$ ?

$$\pi = 0.8$$
 or  $\pi = 0.2$ 

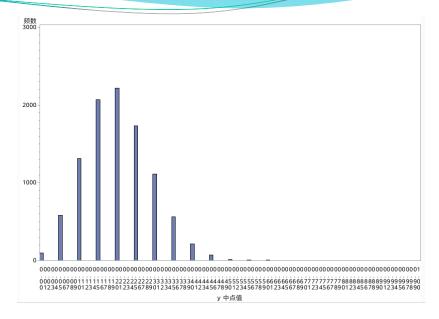
What is the sampling distribution of sample rates in these cases?



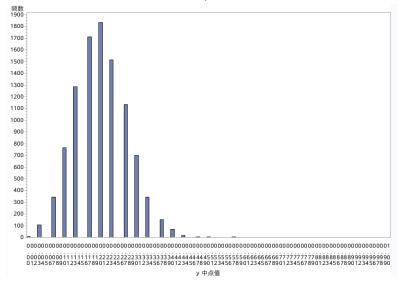
$$\pi = 0.2, n = 10$$



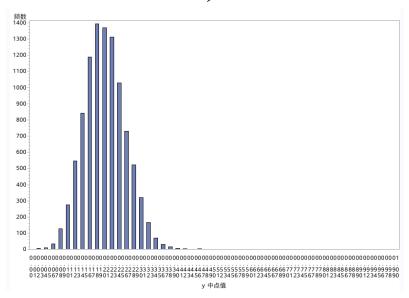
$$\pi$$
=0.2,  $n$ =20



$$\pi$$
=0.2,  $n$ =30



 $\pi$ =0.2, n=50



#### Sampling distribution of sample rates

1. 
$$\mu_p = \pi$$

$$2. \sigma_p = \sqrt{\frac{\pi(1-\pi)}{n}}$$

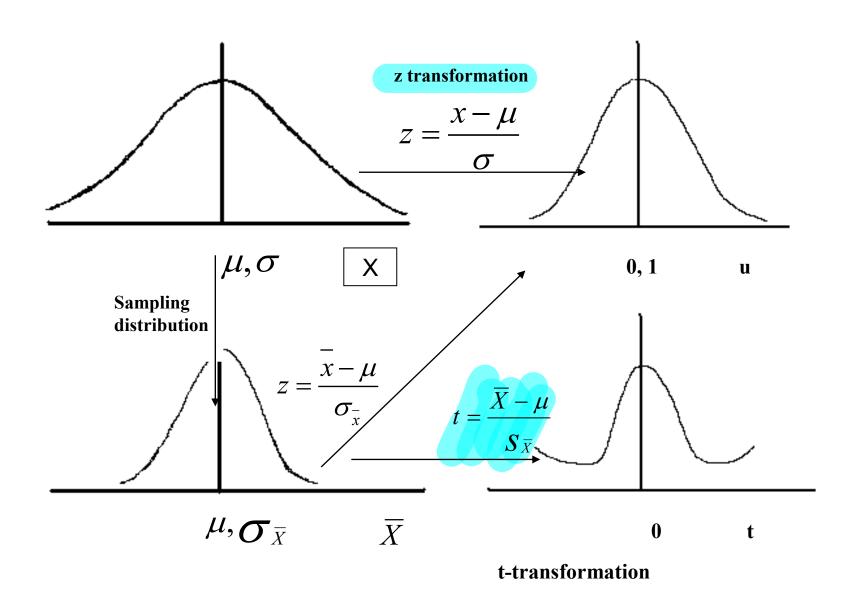
3. If  $\pi$  is 0.5, p is normal. If  $\pi$  is not 0.5,

p is approximately normally distributed

when  $n\pi \ge 5$ ,  $n(1-\pi) \ge 5$ .

# 3. t distribution

#### t-transformation



#### Who introduced t distribution?

William Sealy Gosset



Student in 1908 beer brewery in Gunnies

t distribution, was

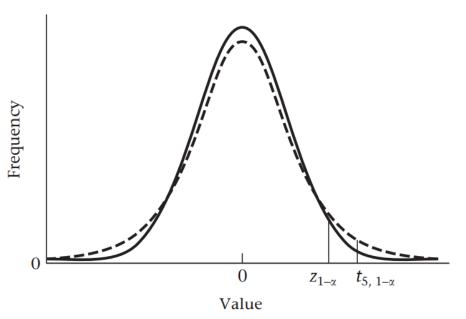
discovered by W. S.

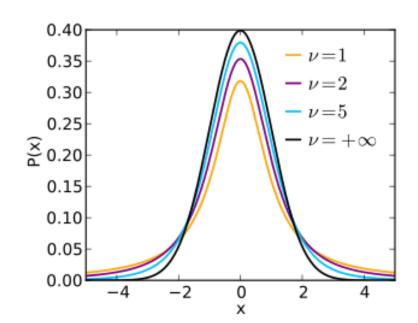
Gossett [1876-1937].

$$f(t) = \frac{\Gamma[(\nu+1)/2]}{\sqrt{\pi\nu}\Gamma(\nu/2)} (1 + \frac{t^2}{\nu})^{-\frac{(\nu+1)}{2}}$$

Student's t-distribution

#### Student t distribution





= N(0, 1) distribution

 $---=t_5$  distribution

#### Comparison of the 97.5th percentile of the t distribution and the normal distribution

d	t <sub>d,.975</sub>	Z <sub>.975</sub>	d	t <sub>d,975</sub>	<b>Z</b> <sub>.975</sub>
4	2.776	1.960	60	2.000	1.960
9	2.262	1.960	∞	1.960	1.960
29	2.045	1.960			

#### Characteristics of t distribution curve

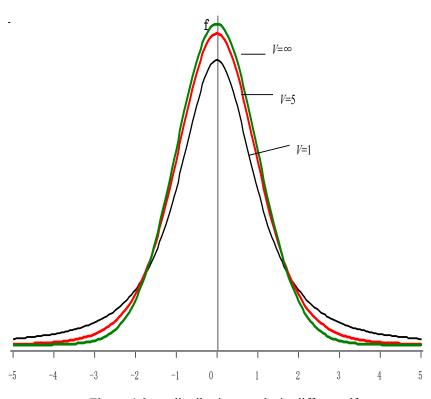
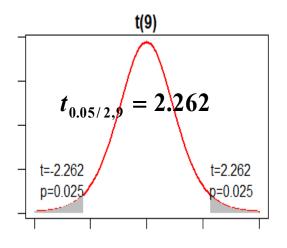


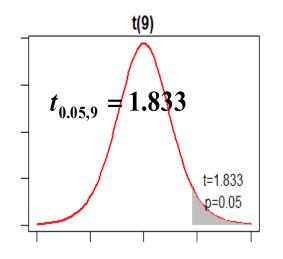
Figure 4-3 t-distribution graphs in different df

- ■It is symmetrical by y axis and has one apex;
- ■Only one parameter,
  v( degree of freedom, v=n-1)
  determines the shape of t
  distribution.
- ■The total area under the t distribution equals to 1.
- ■When v approaches ∞, t distribution approaches standard normal distribution

# t statistic table

df	P				
ν	ı-tail:	0.05	0.025	0.005	0.0005
	2-tail:	0.1	0.05	0.01	0.001
1		6.314	12.706	63.657	636.619
2		2.920	4.303	9.925	31.599
3		2.353	3.182	5.841	12.924
•••••					
9		1.833	2.262	3.250	4.781
•••••					
$\infty$		1.645	1.960	2.576	3.291





# Summary

✓ Sampling distribution of sample means

1. 
$$\mu_{ar{\mathbf{X}}} = \mu_X$$

1. 
$$\mu_{\bar{X}} = \mu_X$$
2.  $\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$ 

3. If X is normal, X is normal. If X is nonnormal, X is approximately normally distributed for sufficiently large sample size.

# Summary

#### $\checkmark$ t distribution

$$Z = \frac{\overline{X} - \mu}{\sigma_{\overline{X}}}$$

$$\downarrow$$

$$t = \frac{\overline{X} - \mu}{S_{\overline{X}}}$$

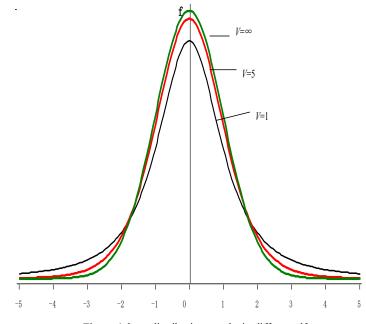
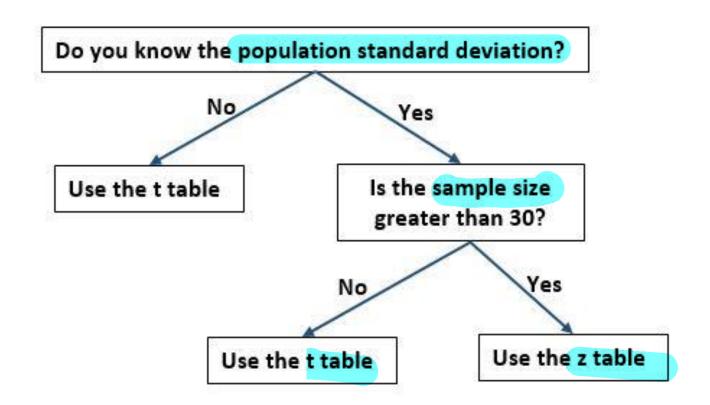


Figure 4-3 t-distribution graphs in different df



# Thank you!