

## Chapter 17

### Inference about a Population Mean

## Conditions for Inference about a Mean

- ◆ Data are from a **SRS** of size  $n$ .
- ◆ Population has a **Normal distribution** with mean  $\mu$  and standard deviation  $\sigma$ .
- ◆ Both  $\mu$  and  $\sigma$  are usually unknown.
  - we use inference to estimate  $\mu$ .
  - **Problem:**  $\sigma$  unknown means we cannot use the  $z$  procedures previously learned.

## Standard Error

- ◆ When we do not know the population standard deviation  $\sigma$  (which is usually the case), we must estimate it with the sample standard deviation  $s$ .
- ◆ When the standard deviation of a statistic is estimated from data, the result is called the **standard error** of the statistic.
- ◆ The standard error of the sample mean  $\bar{x}$  is

$$\frac{s}{\sqrt{n}} \text{ where } s = \sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}}$$

## One-Sample $t$ Statistic

- ◆ When we estimate  $\sigma$  with  $s$ , our one-sample  $z$  statistic becomes a **one-sample  $t$  statistic**.

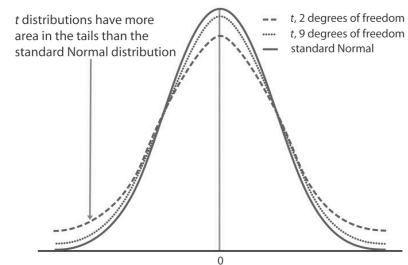
$$z = \frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}} \rightarrow t = \frac{\bar{x} - \mu_0}{s / \sqrt{n}}$$

- ◆ By changing the denominator to be the standard error, our statistic no longer follows a Normal distribution. The  $t$  test statistic follows a  **$t$  distribution with  $n - 1$  degrees of freedom**.

## The $t$ Distributions

- ◆ The  $t$  density curve is similar in shape to the standard Normal curve. They are both symmetric about 0 and bell-shaped.
- ◆ The spread of the  $t$  distributions is a bit greater than that of the standard Normal curve (i.e., the  $t$  curve is slightly “fatter”).
- ◆ As the degrees of freedom increase, the  $t$  density curve approaches the  $N(0, 1)$  curve more closely. This is because  $s$  estimates  $\sigma$  more accurately as the sample size increases.

## The $t$ Distributions



### Using Table C

- ◆ Table C on page 693 gives critical values having upper tail probability  $p$  along with corresponding confidence level  $C$ .
- ◆  $z^*$  values are also displayed at the bottom.

df	Upper tail probability $p$											
	.25	.20	.15	.10	.05	.025	.02	.01	.005	.0025	.001	.0005
1	1.000	1.376	1.963	3.078	6.314	12.71	15.89	31.82	63.66	127.3	318.3	636.6
2	0.816	1.061	1.386	1.886	2.920	4.303	4.849	6.965	9.925	14.09	22.33	31.60
3	0.765	0.978	1.250	1.638	2.353	3.182	3.462	4.541	5.841	7.453	10.21	12.92
4	0.741	0.941	1.190	1.533	2.132	2.776	2.999	3.747	4.604	5.598	7.173	8.610
5	0.727	0.920	1.156	1.476	2.015	2.571	2.757	3.395	4.032	4.773	5.893	6.869
6	0.718	0.906	1.134	1.440	1.943	2.447	2.612	3.143	3.707	4.317	5.208	5.959
7	0.711	0.896	1.119	1.415	1.895	2.355	2.517	2.998	3.499	4.029	4.785	5.408
8	0.706	0.889	1.108	1.397	1.860	2.306	2.449	2.896	3.355	3.833	4.501	5.041
9	0.703	0.883	1.100	1.383	1.833	2.262	2.398	2.821	3.250	3.690	4.297	4.781
10	0.700	0.879	1.093	1.372	1.812	2.228	2.359	2.764	3.169	3.581	4.144	4.587
$z^*$	0.674	0.841	1.036	1.282	1.645	1.960	2.054	2.326	2.576	2.807	3.091	3.291
	50%	60%	70%	80%	90%	95%	96%	98%	99%	99.5%	99.8%	99.9%

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### Using Table C

- ◆ Find the value  $t^*$  with probability 0.025 to its right under the  $t(6)$  density curve.

df	Upper tail probability $p$											
	.25	.20	.15	.10	.05	.025	.02	.01	.005	.0025	.001	.0005
1	1.000	1.376	1.963	3.078	6.314	12.71	15.89	31.82	63.66	127.3	318.3	636.6
2	0.816	1.061	1.386	1.886	2.920	4.303	4.849	6.965	9.925	14.09	22.33	31.60
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	50%	60%	70%	80%	90%	95%	96%	98%	99%	99.5%	99.8%	99.9%

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**$t^* = 2.365$**

## One-Sample $t$ Confidence Interval

Take an SRS of size  $n$  from a population with unknown mean  $\mu$  and unknown standard deviation  $\sigma$ . A level  $C$  confidence interval for  $\mu$  is:

$$\bar{x} \pm t^* \frac{s}{\sqrt{n}}$$

where  $t^*$  is the critical value for confidence level  $C$  from the  $t$  density curve with  $n - 1$  degrees of freedom.

- This interval is exact when the population distribution is Normal and approximate for large  $n$  in other cases.

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## Case Study American Adult Heights



A study of 7 American adults from an SRS yields an average height of  $\bar{x} = 67.2$  inches and a standard deviation of  $s = 3.9$  inches. A 95% confidence interval for the average height of all American adults ( $\mu$ ) is:

$$\begin{aligned} \bar{x} \pm t^* \frac{s}{\sqrt{n}} &= 67.2 \pm 2.447 \frac{3.9}{\sqrt{7}} = 67.2 \pm 3.606 \\ &= 63.594 \text{ to } 70.806 \end{aligned}$$

"We are 95% confident that the average height of all American adults is between 63.594 and 70.806 inches."

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## One-Sample $t$ Test

Like the confidence interval, the  $t$  test is close in form to the  $z$  test learned earlier. When estimating  $\sigma$  with  $s$ , the test statistic becomes:

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$$

where  $t$  follows the  $t$  density curve with  $n - 1$  degrees of freedom, and the  $P$ -value of  $t$  is determined from that curve.

- The  $P$ -value is exact when the population distribution is Normal and approximate for large  $n$  in other cases.

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## $P$ -value for Testing Means


- ◆  $H_a: \mu > \mu_0$ 
  - ✦  $P$ -value is the probability of getting a value as large or larger than the observed test statistic ( $t$ ) value.
- ◆  $H_a: \mu < \mu_0$ 
  - ✦  $P$ -value is the probability of getting a value as small or smaller than the observed test statistic ( $t$ ) value.
- ◆  $H_a: \mu \neq \mu_0$ 
  - ✦  $P$ -value is *two times* the probability of getting a value as large or larger than the absolute value of the observed test statistic ( $t$ ) value.


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**THE ONE-SAMPLE  $t$  TEST**  
 Draw an SRS of size  $n$  from a population having unknown mean  $\mu$ . To test the hypothesis  $H_0: \mu = \mu_0$  based on an SRS of size  $n$ , compute the one-sample  $t$  statistic:

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$$

In terms of a variable  $T$  having the  $t(n-1)$  distribution, the  $P$ -value for a test of  $H_0$  against

$H_a: \mu > \mu_0$  is  $P(T \geq t)$  

$H_a: \mu < \mu_0$  is  $P(T \leq t)$  

$H_a: \mu \neq \mu_0$  is  $2P(T \geq |t|)$  

These  $P$ -values are exact if the population distribution is Normal and are approximately correct for large  $n$  in other cases.

## Case Study Sweetening Colas (Ch. 14)



Cola makers test new recipes for loss of sweetness during storage. Trained tasters rate the sweetness before and after storage. Here are the sweetness losses (sweetness before storage minus sweetness after storage) found by 10 tasters for a new cola recipe:

2.0 0.4 0.7 2.0 -0.4 2.2 -1.3 1.2 1.1 2.3

Are these data good evidence that the cola lost sweetness during storage?

## Case Study Sweetening Colas



It is reasonable to regard these 10 carefully trained tasters as an SRS from the population of all trained tasters.

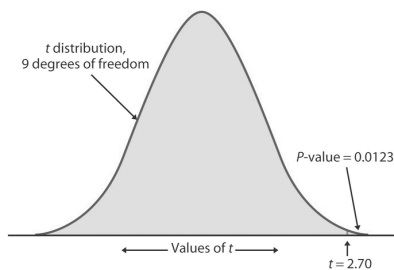
While we cannot judge Normality from just 10 observations, the data shows no outliers, clusters, or extreme skewness. Thus,  $P$ -values for the  $t$  test will be reasonably accurate.

## Case Study



- Hypotheses:**  $H_0: \mu = 0$   
 $H_a: \mu > 0$
- Test Statistic:**  $(df = 10 - 1 = 9)$   $t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} = \frac{1.02 - 0}{1.196/\sqrt{10}} = 2.70$
- P-value:**  $P\text{-value} = P(T > 2.70) = 0.0123$  (using a computer)  
 $P$ -value is between 0.01 and 0.02 since  $t = 2.70$  is between  $t^* = 2.398$  ( $p = 0.02$ ) and  $t^* = 2.821$  ( $p = 0.01$ ) (Table C)
- Conclusion:** Since the  $P$ -value is smaller than  $\alpha = 0.02$ , there is quite strong evidence that the new cola loses sweetness on average during storage at room temperature.

## Case Study Sweetening Colas



## Matched Pairs $t$ Procedures

- To compare two treatments, subjects are matched in pairs and each treatment is given to one subject in each pair.
- Before-and-after observations on the same subjects also calls for using **matched pairs**.
- To compare the responses to the two treatments in a matched pairs design, apply the one-sample  $t$  procedures to the observed **differences** (one treatment observation minus the other).
- The parameter  $\mu$  is the mean difference in the responses to the two treatments within matched pairs of subjects in the entire population.

### Case Study Air Pollution



Pollution index measurements were recorded for two areas of a city on each of 8 days.

Area A	Area B	A - B
2.92	1.84	1.08
1.88	0.95	0.93
5.35	4.26	1.09
3.81	3.18	0.63
4.69	3.44	1.25
4.86	3.69	1.17
5.81	4.95	0.86
5.55	4.47	1.08

Are the average pollution levels the same for the two areas of the city?

### Case Study Air Pollution



It is reasonable to regard these 8 measurement pairs as an SRS from the population of all paired measurements.

While we cannot judge Normality from just 8 observations, the data shows no outliers, clusters, or extreme skewness. Thus,  $P$ -values for the  $t$  test will be reasonably accurate.

These 8 differences have  $\bar{X} = 1.0113$  and  $s = 0.1960$ .

### Case Study



1. **Hypotheses:**

$$H_0: \mu = 0$$

$$H_a: \mu \neq 0$$

2. **Test Statistic:**  
 $(df = 8 - 1 = 7) \quad t = \frac{\bar{x} - \mu_0}{s / \sqrt{n}} = \frac{1.0113 - 0}{0.1960 / \sqrt{8}} = 14.594$

3. **P-value:**  
 $P\text{-value} = 2P(T > 14.594) = 0.0000017$  (using a computer)  
 $P\text{-value}$  is smaller than  $2(0.0005) = 0.0010$  since  $t = 14.594$  is greater than  $t^* = 5.041$  (upper tail area = 0.0005) (Table C)

4. **Conclusion:**  
 Since the  $P$ -value is smaller than  $\alpha = 0.001$ , there is very strong evidence that the mean pollution levels are different for the two areas of the city.

### Case Study Air Pollution



Find a 95% confidence interval to estimate the difference in pollution indexes ( $A - B$ ) between the two areas of the city. ( $df = 8 - 1 = 7$  for  $t^*$ )

$$\bar{x} \pm t^* \frac{s}{\sqrt{n}} = 1.0113 \pm 2.365 \frac{0.1960}{\sqrt{8}} = 1.0113 \pm 0.1639$$

$$= 0.8474 \text{ to } 1.1752$$

We are 95% confident that the pollution index in area A exceeds that of area B by an average of 0.8474 to 1.1752 index points.

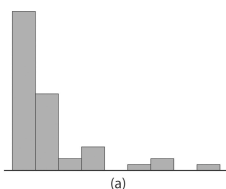
## Robustness of $t$ Procedures

- ◆ The  $t$  confidence interval and test are exactly correct when the distribution of the population is exactly normal.
- ◆ No real data are exactly Normal.
- ◆ The usefulness of the  $t$  procedures in practice therefore depends on how strongly they are affected by lack of Normality.
- ◆ A confidence interval or significance test is called **robust** if the confidence level or  $P$ -value does not change very much when the conditions for use of the procedure are violated.

## Using the $t$ Procedures

- ◆ Except in the case of small samples, the assumption that the data are an SRS from the population of interest is more important than the assumption that the population distribution is Normal.
- ◆ **Sample size less than 15:** Use  $t$  procedures if the data appear close to Normal (symmetric, single peak, no outliers). If the data are skewed or if outliers are present, do not use  $t$ .
- ◆ **Sample size at least 15:** The  $t$  procedures can be used except in the presence of outliers or strong skewness in the data.
- ◆ **Large samples:** The  $t$  procedures can be used even for clearly skewed distributions when the sample is large, roughly  $n \geq 40$ .

### Can we use $t$ ?



(a)

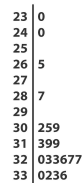
- ◆ This histogram shows the percent of each state's residents who are Hispanic.
- ◆ **Cannot use  $t$ .** We have a population, not an SRS.

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### Can we use $t$ ?



(b)

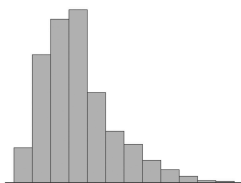
- ◆ This stemplot shows the force required to pull apart 20 pieces of Douglas fir.
- ◆ **Cannot use  $t$ .** The data are strongly skewed to the left, so we cannot trust the  $t$  procedures for  $n = 20$ .

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### Can we use $t$ ?



(c)

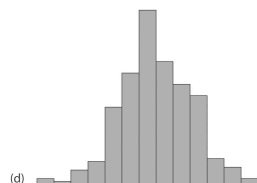
- ◆ This histogram shows the distribution of word lengths in Shakespeare's plays.
- ◆ **Can use  $t$ .** The data is skewed right, but there are no outliers. We can use the  $t$  procedures since  $n \geq 40$ .

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### Can we use $t$ ?



(d)

- ◆ This histogram shows the heights of college students.
- ◆ **Can use  $t$ .** The distribution is close to Normal, so we can trust the  $t$  procedures for any sample size.

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