

Module 16: One-sample t-tests and Confidence Intervals

This module presents a useful statistical tool, the one-sample t-test and the confidence interval for the population mean.

The t-test and t Distribution Situation

What happens when we don't know the true value for the population standard deviation σ ? Suppose we have only the information from a random sample, $n = 5$, from the population of body weights, with

$$\begin{aligned}\bar{x} &= 153.0 \text{ lbs, and} \\ s &= 12.9 \text{ lbs.}\end{aligned}$$

When we had a value for the population parameter σ , we used the following formula:

$$C[\bar{x} - 1.96 \sigma/\sqrt{n} \leq \mu \leq \bar{x} + 1.96 \sigma/\sqrt{n}] = 0.95$$

We do have an estimate of the population standard deviation, σ , namely the sample standard deviation $s = 12.9$ lbs. Hence, it seems reasonable to think that we should be able to use this estimate in some way. It is also reasonable to think that, if we substitute s for σ , we are substituting a guess at the truth for the truth itself and we will probably have to pay a price for doing so. So, what is the price?

The essence of the situation is that, when we substitute a guess for the truth, we add noise to the system. The question then becomes one of characterizing this noise and taking it into account. Noise in this situation is equivalent to variability, so we are adding variability to the system. How much and exactly where?

If we use this estimate, then we must make appropriate adjustments to the formula to account for the variability of this estimate.

To properly account for this situation, we need to use a distribution different from the normal distribution. The appropriate distribution is the t distribution, which is very similar to the normal distribution for large sample sizes, but differs importantly for smaller samples, especially those with $n < 30$.

Confidence Interval for μ using s

The appropriate formula, when $\alpha = 0.05$, is

$$C[\bar{x} - t_{0.975}(n-1) (s/\sqrt{n}) \leq \mu \leq \bar{x} + t_{0.975}(n-1) (s/\sqrt{n})] = 0.95,$$

where $t_{0.975}(n-1)$ references the t distribution with $n-1$ degrees of freedom (df), specifically the point on that distribution below which lies 0.975 of the total area. For this situation, the correct number of degrees of freedom is one less than the sample size, i.e. $df = n - 1$.

Tables for the t Distribution

To obtain the values for the t distribution, see Table
Module 2: The t distribution.

For the above situation, with $n = 5$, $\bar{x} = 153.0$ lbs, and $s = 12.9$ lbs, we have $t_{.975}(n-1) = t_{.975}(4) = 2.776$ so that the interval becomes:

$$C[\bar{x} - t_{.975}(4)(s/\sqrt{n}) \leq \mu \leq \bar{x} + t_{.975}(4)(s/\sqrt{n})] = 0.95,$$

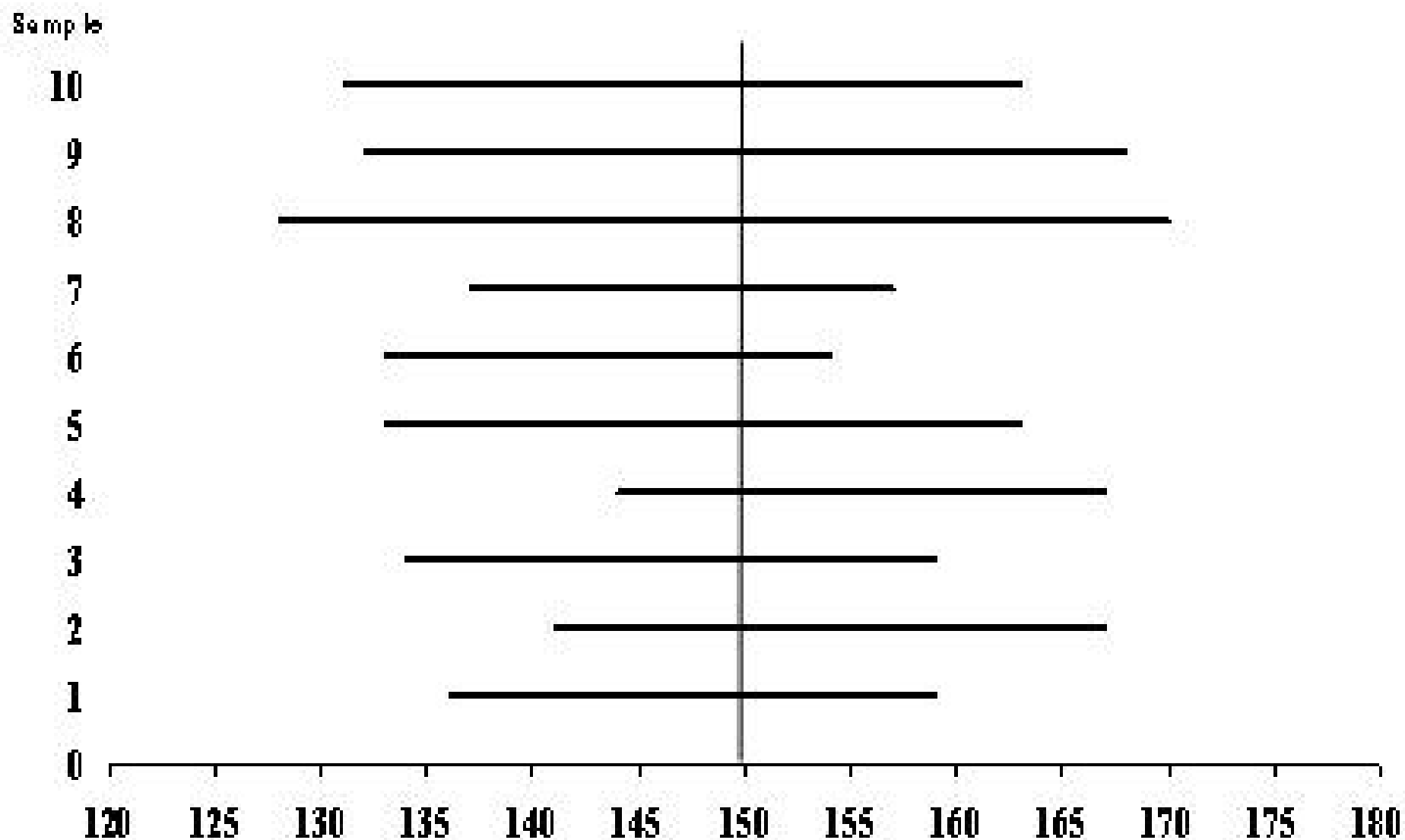
$$C[153.0 - 2.776(12.9/\sqrt{5}) \leq \mu \leq 153.0 + 2.776(12.9/\sqrt{5})] \\ = 0.95,$$

$$C[137.0 \leq \mu \leq 169.0] = 0.95.$$

Given this confidence interval, would you believe that the population mean for the population from which this sample was selected had the value $\mu = 170.0$ lbs?

Ten Samples from $N(150,10)$ $n = 5$

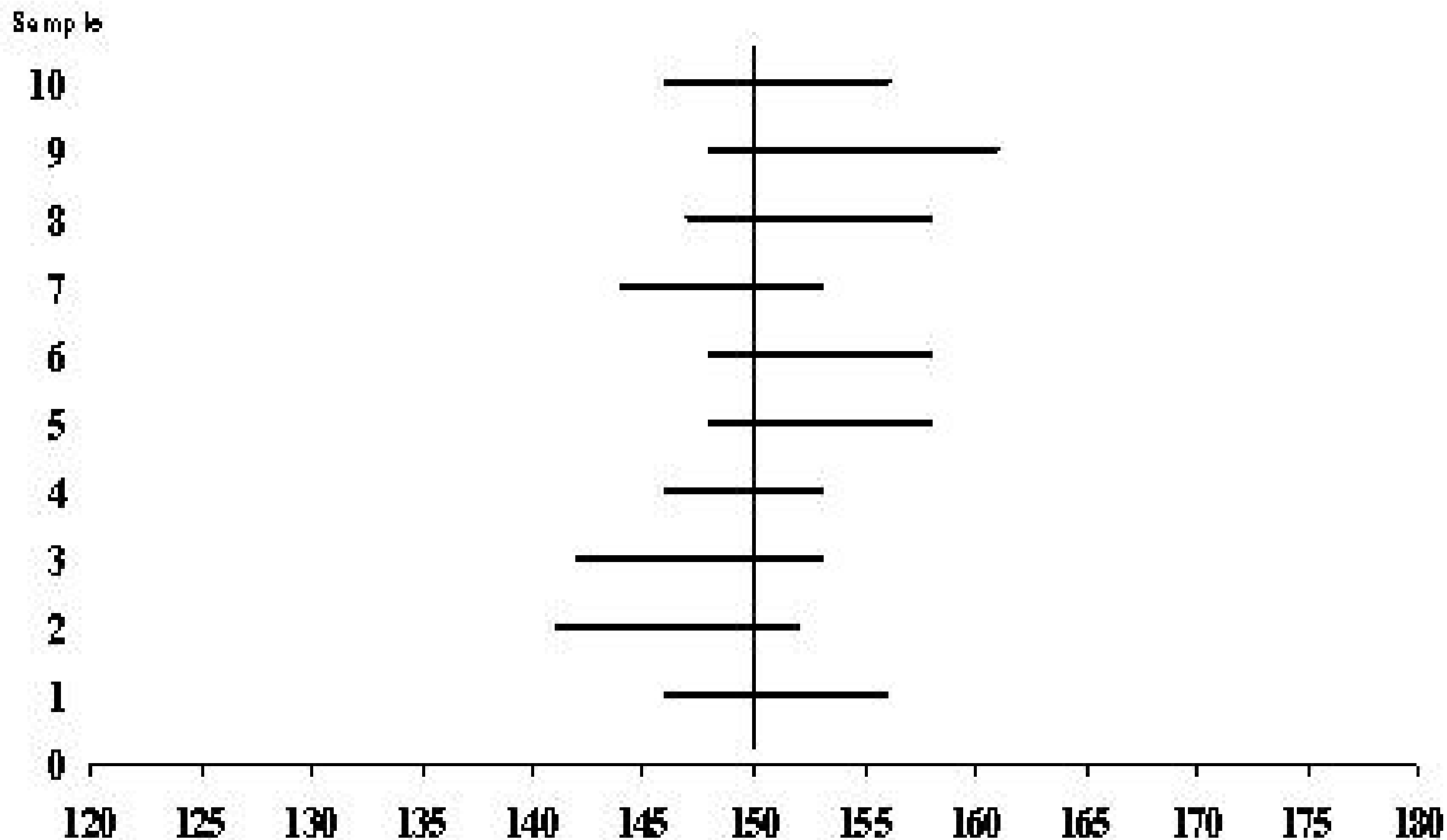
					Confidence Interval					
					Normal Distribution			t Distribution ($n-1, \alpha=2.77\%$)		
Sample	n	Mean	s^2	s	LL	UL	Length	LL	UL	Length
1	5	147.47	88.14	9.39	138.66	156.19	17.53	135.77	159.08	23.31
2	5	157.98	117.91	10.86	145.21	162.75	17.53	140.97	167.46	26.49
3	5	146.50	103.66	10.18	137.74	155.27	17.53	133.86	159.14	25.28
4	5	155.53	91.99	9.59	146.76	164.29	17.53	143.62	167.43	23.81
5	5	147.87	149.65	12.23	139.10	156.63	17.53	132.68	163.05	30.37
6	5	143.60	66.76	8.17	134.84	152.37	17.53	133.46	153.75	20.29
7	5	146.87	64.23	8.01	138.11	155.64	17.53	136.92	156.82	19.90
8	5	149.19	287.88	16.96	140.47	157.96	17.53	128.39	170.00	41.61
9	5	157.05	207.28	14.39	141.28	158.82	17.53	132.48	167.62	35.14
10	5	146.92	173.36	13.17	138.16	156.69	17.53	130.68	163.27	32.59



$$C[\bar{x} - t_{0.975}(4) s / \sqrt{n} \leq \mu \leq \bar{x} + t_{0.975}(4) s / \sqrt{n}] = 95\%, n = 5$$

Ten Samples from $N(150,10)$ $n = 20$

Sample	n	Mean	s^2	s	Confidence Interval					
					Normal Distribution			t Distribution ($t_{0.05, 19} = 1.729$)		
					LL	UL	Length	LL	UL	Length
1	20	150.86	100.96	10.05	146.48	155.24	8.77	146.16	155.96	9.39
2	20	146.88	122.70	11.08	142.50	151.27	8.77	141.71	152.06	10.35
3	20	147.65	119.51	10.93	143.27	152.03	8.77	142.54	152.76	10.22
4	20	149.37	61.07	7.15	144.99	153.75	8.77	146.03	152.71	6.68
5	20	153.30	109.54	10.47	148.92	157.69	8.77	148.41	158.19	9.78
6	20	152.83	111.96	10.58	148.45	157.21	8.77	147.89	157.77	9.89
7	20	148.62	91.94	9.59	144.24	153.01	8.77	144.14	153.10	8.96
8	20	152.16	140.83	11.87	147.77	156.54	8.77	146.61	157.70	11.09
9	20	154.40	179.56	13.40	150.02	158.79	8.77	148.14	160.67	12.52
10	20	151.43	115.85	10.76	147.04	155.81	8.77	146.40	156.46	10.06



$$C[\bar{x} - t_{0.975}(19) s/\sqrt{n} \leq \mu \leq \bar{x} + t_{0.975}(19) s/\sqrt{n}] = 95\%, n = 20$$

Ten Samples from $N(150,10)$ $n = 50$

Sample	n	Mean	s^2	s	Confidence Interval					
					Normal Distribution			t Distribution ($t_{0.05,49} = 2.01$)		
					LL	UL	Length	LL	UL	Length
1	50	148.79	121.96	11.00	146.02	151.57	5.54	145.67	151.92	6.25
2	50	151.43	83.83	9.16	147.66	153.20	5.54	147.83	153.03	5.21
3	50	151.86	108.58	10.42	148.09	153.63	5.54	147.90	153.82	5.92
4	50	152.92	144.88	12.04	150.15	155.69	5.54	149.50	156.34	6.84
5	50	149.68	104.21	10.21	146.91	152.45	5.54	146.78	152.58	5.80
6	50	151.62	101.25	10.01	147.85	153.39	5.54	147.78	153.47	5.69
7	50	150.57	74.04	8.61	147.79	153.34	5.54	148.12	153.01	4.89
8	50	149.75	97.23	9.86	146.98	152.53	5.54	146.95	152.56	5.61
9	50	151.68	61.77	7.86	147.91	153.45	5.54	148.44	152.91	4.47
10	50	151.99	110.58	10.52	148.22	153.77	5.54	148.00	153.98	5.98

Sample

10

9

8

7

6

5

4

3

2

1

0

120 125 130 135 140 145 150 155 160 165 170 175 180

$$C[\bar{x} - t_{0.975}(49) s/\sqrt{n} \leq \mu \leq \bar{x} + t_{0.975}(49) s/\sqrt{n}] = 95\%, n = 50$$

Hypothesis Testing: $\bar{x} = 153.0$ lbs, $s = 12.9$ lbs

A random sample of $n = 5$ measurements of weights from a population provides a sample mean of $\bar{x} = 153.0$ lbs and a **sample standard deviation of $s = 12.9$ lbs**. Is it likely that the population mean has the value $\mu = 170$ lbs?

1. *The hypothesis:* $H_0: \mu = 170$ versus $H_1: \mu \neq 170$
2. *The assumptions:* Random sample from a normal distribution
3. *The α -level:* $\alpha = 0.05$

4. The test statistic:

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

5. The critical region: Reject $H_0: \mu = 170$ if the value calculated for t is not between $\pm t_{0.975}(4) = 2.776$

6. The result:

$$t = \frac{153.0 - 170.0}{12.9/\sqrt{5}} = \frac{-17.0}{5.77} = -2.95$$

7. The conclusion: Reject $H_0: \mu = 170$ since the value calculated for t is not between ± 2.776

This test was performed under the assumption that $\mu = 170$. Our conclusion is that our sample mean $\bar{x} = 153.0$ is so far away from $\mu = 170$ that we find it hard to believe that $\mu = 170$. That is, our observed value for the sample mean of $\bar{x} = 153.0$ is too rare for us to believe that $\mu = 170$.

How rare is $\bar{x} = 153.0$ under the assumption that $\mu = 170$?

$P < 0.05$ or $0.02 < P < 0.05$