

# Topic12: Confidence Intervals

# Outline

## Topic12: Confidence Intervals

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## Example1: Birth Weight

It is known that the birthweight (in kgs) of babies born at term (37-41 weeks gestation) is  $W \sim N(\mu, 0.525^2)$ .

The following data are 8 term births:

```
x=c(2.853, 3.127, 3.159, 3.800, 2.656, 3.245, 3.510, 3.082)
```

- (i) What is the best estimate for  $\mu$ ?
- (ii) Find a 95% and 99% CI for  $\mu$ .



## Example2: Paint Primer Thickness

Assume that paint primer thickness can be modelled by  $X \sim N(\mu, \sigma^2)$ . In an ongoing process of quality control in an industrial system, the following first sample of values was obtained:

```
x=c(1.30,1.10,1.20,1.25,1.05,0.95,1.10,1.16,1.37,0.98)
```

- (i) What is a 95% CI for the primer thickness?
- (ii) The company advertises that the primer thickness is 1.25. What would you conclude?



## Example3: Concrete Tensile Strength

We are interested in the influence of the size of test specimens of concrete on the tensile strength. 8 concrete mixes were made, and from each mix 2 test specimens were prepared and tested, resulting in the following strengths (in  $kN/m^2$ ):

```
small=c(4404, 4236, 3788, 3475, 3418, 2262, 7415, 6993)
large=c(4140, 3984, 3842, 3053, 3145, 1813, 6867, 7091)
diff=large-small
diff

## [1] -264 -252    54 -422 -273 -449 -548    98
```

- (i) Find a 95% CI for the mean tensile strength of small specimens, assuming that the strengths can be modelled by  $N(\mu, 1000^2)$ .
- (ii) Find a 90% CI for the mean difference in tensile strengths, assuming that the differences can be modelled by  $N(\mu, \sigma^2)$ .



## Example4: Clinton vs Trump Polls

From a recent report on the [▶ USA Election 2016](#), we find the following quotes:

‘The tighter race in Florida showed Clinton edging Trump 45% to 42% among likely voters, with Johnson at 5% and Stein at 3%. That three-point lead was within the poll’s margin of error.’



'The NBC/WSJ/Marist poll Florida poll surveyed 700 likely voters between October 3-5 with a margin of error of plus or minus 3.7 percentage points.'

'The Pennsylvania poll surveyed 709 likely voters between October 3-6 with a margin of error of plus or minus 3.7 percentage points.'

A fuller report is found here with an interesting video overview:

► USA Election Polls

A random survey of 2000 voters found that 1165 were going to vote for Hilary Clinton.

- (i) Find a 95% CI for the proportion of voters  $p$  that will vote for Hilary.
- (ii) What is the 'margin of error'?
- (iii) What sample size is needed to give a 95% CI for  $p$  with width  $\pm 0.03$ ?

## Estimating Parameters

So far in Part3 we have been testing *hypotheses* about an unknown parameter. Now we want to *estimate* the unknown parameter.

If we can find a Pivot, then we can find:

- (1) A *Point Estimate* for the parameter.
- (2) A *Confidence Interval (CI)* for the parameter.



## Definition (Pivot)

A pivot is:

- ▶ a function (based on the data and parameters) which always has the same distribution regardless of the value of the parameter, for some statistical model.
- ▶ the random variable from which we construct CIs.
- ▶ often of the form  $\frac{\text{Estimate} - \text{Parameter}}{\text{Standard Error}} \sim \text{Distribution}$  where the *Standard Error* is the standard deviation of the *Estimate*.

It turns out that the Test Statistics, previously considered in Part 3, can be used as Pivots. Hence, we effectively rearrange the Pivot to get the CI.

# Overview of Confidence Intervals

## Definition (Confidence Interval)

A Confidence Interval is:

- ▶ a sequence of intervals which contain the unknown parameter  $(1 - \alpha)\%$  of the time, where  $\alpha$  is the confidence level, often  $\alpha = 0.05$ . ▶ Sequence of CIs
- ▶ of the form  $(Point\ Estimate \pm Critical\ Value \times Standard\ Error)$ .
- ▶ a set of possible hypotheses  $\{H_0 : \mu = \mu_0\}$ , which will be retained if  $\mu_0 \in CI$ .

A Confidence Interval is *not* an interval which contains the unknown parameter  $(1 - \alpha)\%$  of the time.

# Formulae for Confidence Intervals

Test	Parameter	Test Statistic and CI
Proportion	$p$	<div style="display: flex; align-items: center;"> <span style="border: 1px solid black; padding: 2px 10px; margin-right: 10px;">T</span> <math display="block">\frac{\hat{p} - p}{\sqrt{\frac{\hat{p}(1-\hat{p})}{n}}} \sim Z</math> </div> <div style="display: flex; align-items: center;"> <span style="border: 1px solid black; padding: 2px 10px; margin-right: 10px;">CI</span> <math display="block">\hat{p} \pm z^* \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}</math> <span style="margin-right: 10px;">Approximate</span> </div> <div style="display: flex; align-items: center;"> <span style="border: 1px solid black; padding: 2px 10px; margin-right: 10px;">CI</span> <math display="block">\hat{p} \pm z^* \frac{1}{2\sqrt{n}}</math> <span style="margin-right: 10px;">Conservative</span> </div>
1 sample $Z$	$\mu$	<div style="display: flex; align-items: center;"> <span style="border: 1px solid black; padding: 2px 10px; margin-right: 10px;">T</span> <math display="block">\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim Z</math> </div> <div style="display: flex; align-items: center;"> <span style="border: 1px solid black; padding: 2px 10px; margin-right: 10px;">CI</span> <math display="block">\bar{x} \pm z^* \frac{\sigma}{\sqrt{n}}</math> </div>
1 sample $T$	$\mu$	<div style="display: flex; align-items: center;"> <span style="border: 1px solid black; padding: 2px 10px; margin-right: 10px;">T</span> <math display="block">\frac{\bar{X} - \mu}{\frac{s}{\sqrt{n}}} \sim t_{n-1}</math> </div> <div style="display: flex; align-items: center;"> <span style="border: 1px solid black; padding: 2px 10px; margin-right: 10px;">CI</span> <math display="block">\bar{x} \pm t_{n-1}^* \frac{s}{\sqrt{n}}</math> </div>
2 sample $T$	$\mu_1 - \mu_2$	<div style="display: flex; align-items: center;"> <span style="border: 1px solid black; padding: 2px 10px; margin-right: 10px;">T</span> <math display="block">\frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim t_{n_1+n_2-2}</math> </div> <div style="display: flex; align-items: center;"> <span style="border: 1px solid black; padding: 2px 10px; margin-right: 10px;">CI</span> <math display="block">\bar{x}_1 - \bar{x}_2 \pm t_{n_1+n_2-2}^* s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}</math> </div>

## Confidence Interval for Mean based on Z Test Statistic

We will consider this example in detail, and then treat the other CIs by analogy.

Assume we have a sample  $x_1, x_2, \dots, x_n$  from a Normal population  $X \sim N(\mu, \sigma^2)$ , where  $\mu$  is unknown and  $\sigma^2$  is known.

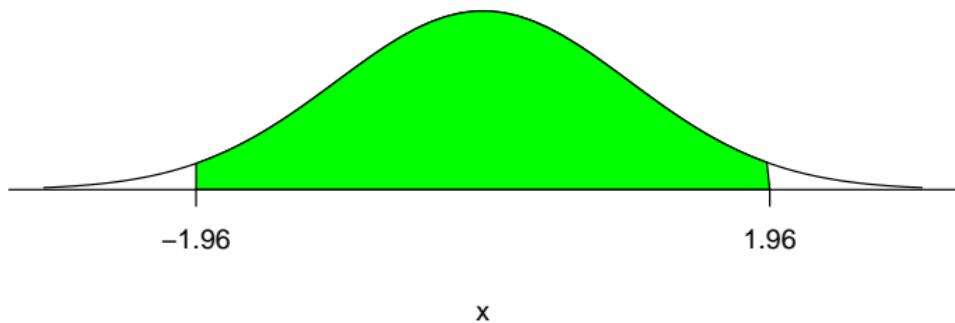
The best estimate of the population mean is the sample mean:

$$\hat{\mu} = \bar{x} = \frac{\sum_{i=1}^n x_i}{n}$$

As the sample mean will differ from sample to sample, we want to find a plausible set of values for  $\mu$  that incorporates this sample to sample variation.

Based on the 1 sample  $Z$  test statistic, we want to find a 95% CI.

(1) If  $P(-z^* \leq Z \leq z^*) = 0.95$ , we know that  $z^* = 1.96$ .



```
## [1] 1.959964
```

(2) Now substituting the pivot  $Z = \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}}$ , gives

$$P\left(-z^* \leq \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \leq z^*\right) = 0.95$$

Rearranging gives

$$P\left(-\bar{X} - z^* \frac{\sigma}{\sqrt{n}} \leq -\mu \leq -\bar{X} + z^* \frac{\sigma}{\sqrt{n}}\right) = 0.95$$

which simplifies to

$$P\left(\bar{X} - z^* \frac{\sigma}{\sqrt{n}} \leq \mu \leq \bar{X} + z^* \frac{\sigma}{\sqrt{n}}\right) = 0.95$$

(3) This is a random interval which covers  $\mu$  with probability 0.95.

$$\bar{X} \pm 1.96 \frac{\sigma}{\sqrt{n}}$$

(4) The observed value of the interval is called the 95% Confidence Interval (CI) for  $\mu$ .

$$\bar{x} \pm 1.96 \frac{\sigma}{\sqrt{n}}$$

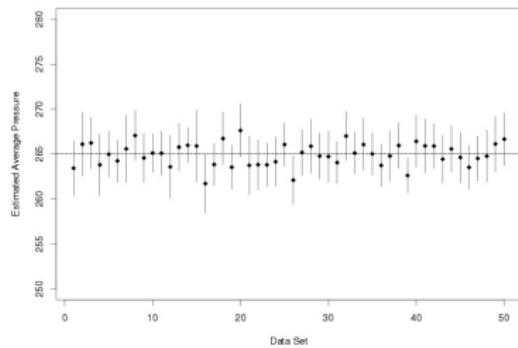
(5) More generally, the  $(1 - \alpha)\%$  Confidence Interval (CI) for  $\mu$  is:

$$\bar{x} \pm z^* \frac{\sigma}{\sqrt{n}}$$

where  $P(-z^* \leq Z \leq z^*) = 1 - \alpha$  or  $P(Z \geq z^*) = \frac{\alpha}{2}$ .

Note:

- (1) The CI refers to the proportion of CIs that will cover the true  $\mu$  if the above procedure is repeated for many samples of size  $n$ . We only observe one sample, so we don't know if it is one that contains  $\mu$ .



- (2) As we increase  $n$ , the CI gets narrower, and so  $\bar{x}$  is a better estimate for the long term estimate of  $\mu$ .

(3) As we increase the confidence level, the CI gets wider.

## Practise finding critical values

Find  $z^*$  for 70%, 80%, 90%, 95% and 99% CIs.

```
qnorm(0.85)  
## [1] 1.036433  
  
qnorm(0.9)  
## [1] 1.281552  
  
qnorm(0.95)  
## [1] 1.644854  
  
qnorm(0.975)  
## [1] 1.959964  
  
qnorm(0.995)  
## [1] 2.575829
```

## Example1: Birth Weight

We have  $W \sim N(\mu, 0.525^2)$  and the data:

```
x=c(2.853,3.127,3.159,3.800,2.656,3.245,3.510,3.082)
mean(x)
```

```
## [1] 3.179
```

(i) The best estimate for  $\mu$  is  $\bar{x} = 3.179$

(ii) A 95% CI for  $\mu$  is:

$$\bar{x} \pm 1.96 \frac{\sigma}{\sqrt{n}}$$

which is  $3.179 \pm 1.96 \frac{0.525}{\sqrt{8}}$ , giving  $(2.82, 3.54)$ .

```
qnorm(0.975)
```

```
## [1] 1.959964
```

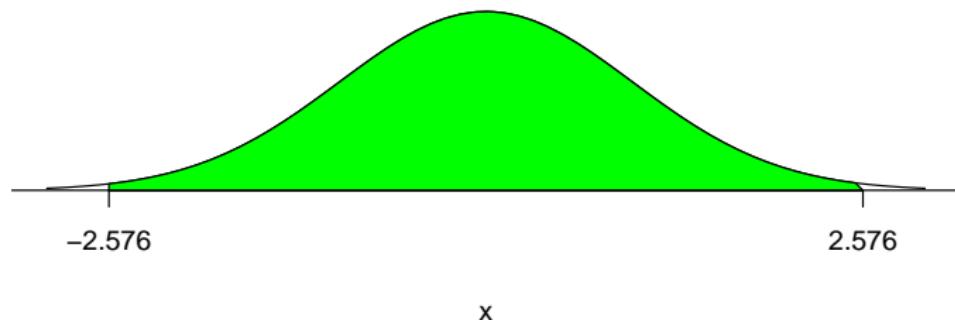
A 99% CI for  $\mu$  is:

$$\bar{x} \pm z^* \frac{\sigma}{\sqrt{n}}$$

which is  $3.179 \pm 2.576 \frac{0.525}{\sqrt{8}}$ , giving (2.70,3.66).

`qnorm(0.995)`

`## [1] 2.575829`



## Example2: Paint Primer

We have  $X \sim N(\mu, \sigma^2)$  and the data:

```
x=c(1.30,1.10,1.20,1.25,1.05,0.95,1.10,1.16,1.37,0.98)
mean(x)

## [1] 1.146

sd(x)

## [1] 0.1363166
```

(i) A 95% CI for  $\mu$  is:

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

which is  $1.146 \pm 2.262 \frac{0.136}{\sqrt{10}}$ , giving (1.05,1.24).

```
qt(0.975, 9)
```

```
## [1] 2.262157
```

(ii) The CI does not contain  $H_0 : \mu = 1.25$  hence the data provide evidence against the company's advertising.

## Example3: Concrete Tensile Strength

(i)

```
small=c(4404, 4326, 3788, 3475, 3418, 2262, 7415, 6993)
mean(small)

## [1] 4510.125

sd(small)

## [1] 1792.36
```

Assuming that the strengths can be modelled by  $N(\mu, 1000^2)$ , a 95% CI for the mean tensile strength of small specimens is:

$$\bar{x} \pm 1.96 \frac{\sigma}{\sqrt{n}}$$

which is

$$4510.125 \pm 1.96 * 1000/\sqrt{8}$$

which gives

$$(3817, 5203).$$

(ii)

```
small=c(4404,4326,3788,3475,3418,2262,7415,6993)
large=c(4140,3984,3842,3053,3145,1813,6867,7091)
diff=large-small
mean(diff)

## [1] -268.25

sd(diff)

## [1] 232.3893
```

Assuming that the differences can be modelled by  $N(\mu, \sigma^2)$ , a 90% CI for the mean difference in tensile strengths is:

$$\bar{x} \pm t_7^* \frac{s}{\sqrt{n}}$$

which is  $-268.25 \pm 1.895 * 232.39/\sqrt{8}$  which gives (113,424).

```
qt(0.95,7)
```

```
## [1] 1.894579
```

## Example4: Clinton vs Trump Polls

Given a random survey of 2000 voters found that 1165 were going to vote for Clinton:

(i) A 95% approximate CI for the proportion of voters  $p$  that will vote for Hillary is:

$$\hat{p} \pm Z^* \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$$

Given  $\hat{p} = 1165/2000 = 0.5825$ , we get

$$0.5825 \pm 1.96 \sqrt{\frac{0.5825 * (1 - 0.5825)}{2000}}$$

which gives

$$(0.56, 0.60)$$

A 95% conservative CI for the proportion of voters  $p$  that will vote for Hillary is:

$$\hat{p} \pm Z^* \frac{1}{2\sqrt{n}}$$

which is

$$0.5825 \pm 1.96 \frac{1}{2\sqrt{2000}}$$

which gives

$$(0.56, 0.60).$$

(ii) The 'margin of error' is  $1.96 \frac{1}{2\sqrt{2000}}$  which is 0.02191347 or approximately 2%.

(iii) To give a 95% CI for  $p$  with width  $\pm 0.03$  (ie margin of error 3%):

We solve

$$1.96 \frac{1}{2\sqrt{n}} = 0.03$$

which gives

$$n = \left( \frac{1.96}{2(0.03)} \right)^2$$

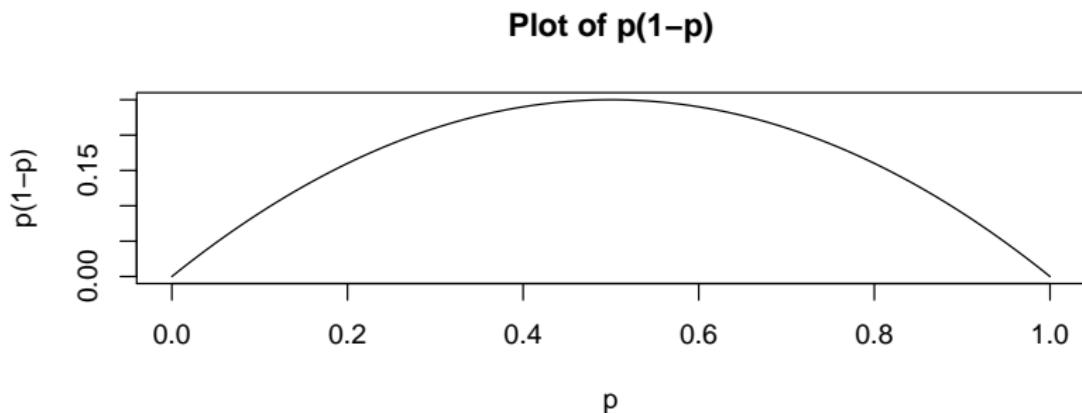
resulting in  $n = 1067$ , or approximately 1000.

Note that  $1.96 \frac{1}{2\sqrt{n}} \approx \frac{1}{\sqrt{n}}$ , so the margin of error is approximately  $\frac{1}{\sqrt{n}}$ , which clearly decreases for larger  $n$ .

## Justifying the Conservative CI for Proportion

The Standard Error in the Approximate CI is  $\sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$ .

As  $0 \leq \hat{p} \leq 1$ , then  $0 \leq \hat{p}(1 - \hat{p}) \leq \frac{1}{4}$ , as the maximum of the function occurs when  $\hat{p} = \frac{1}{2}$ .



Hence, the maximum that the SE can be is  $\sqrt{\frac{\frac{1}{4}}{n}} = \frac{1}{2\sqrt{n}}$  which is what we use in the conservative CI. That is, we choose the largest possible CI.