

Recall :-

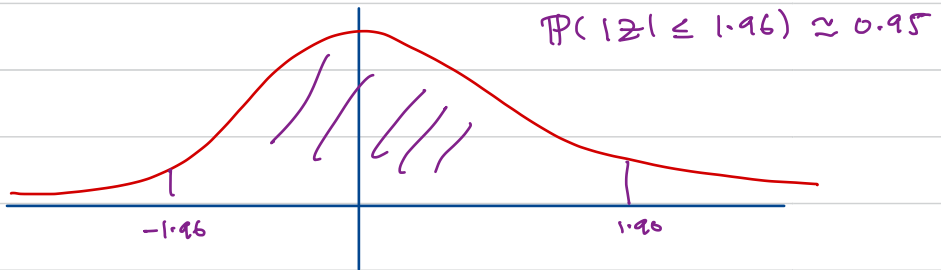
Confidence Interval

X_1, X_2, \dots, X_n - μ - mean

(known) σ^2 - variance

$$\frac{\sqrt{n} (\bar{X} - \mu)}{\sigma} \equiv \text{Central limit Theorem} \equiv \text{Normal-Z (0,1)}$$

$$\mathbb{P} \left(\frac{\sqrt{n} (\bar{X} - \mu)}{\sigma} \leq z \right) \equiv \mathbb{P} (Z \leq z)$$



$$\bar{X} = \frac{X_1 + \dots + X_n}{n}$$

$$\left| \frac{\sqrt{n} (\bar{X} - \mu)}{\sigma} \right| \equiv \text{Central limit Theorem} \equiv \text{Normal-Z (0,1)}$$

$$\left| \frac{\sqrt{n} (\bar{X} - \mu)}{\sigma} \right| \leq 1.96 \quad (\Rightarrow) \quad -1.96 \leq \frac{\sqrt{n} (\bar{X} - \mu)}{\sigma} \leq 1.96$$

$$(\Rightarrow) (-1.96)\sigma \leq \sqrt{n} (\bar{X} - \mu) \leq (1.96)\sigma$$

$$(\Rightarrow) \bar{X} - \frac{1.96\sigma}{\sqrt{n}} \leq \mu \leq \bar{X} + \frac{1.96\sigma}{\sqrt{n}}$$

Confidence Intervals

Sample
 x_1, x_2, \dots, x_n from
population

• Assume $CG \& X$
Unknown $E(X) = \mu$
 $Var(X) = \sigma^2$ - known

$$\text{Compute } \therefore \bar{X} = \frac{\sum_{i=1}^n x_i}{n}$$

Using the Central Limit Theorem for large n we have

$$P\left(\left| \frac{\sqrt{n}(\bar{X} - \mu)}{\sigma} \right| \leq 1.96\right) \approx 0.95$$

which is the same as saying

$$P\left(\mu \in \left(-\frac{1.96\sigma}{\sqrt{n}} + \bar{X}, \frac{1.96\sigma}{\sqrt{n}} + \bar{X}\right)\right) \approx 0.95$$

The interval $\left(-\frac{1.96\sigma}{\sqrt{n}} + \bar{X}, \frac{1.96\sigma}{\sqrt{n}} + \bar{X}\right)$ is called the 95% confidence interval for μ .

↳ dependent on sample.
and is valid if σ is known.

Confidence Intervals - Important - Precise meaning.

$$\mathbb{P}\left(\frac{|\bar{X} - \mu|}{\sigma} \leq 1.96\right) \approx 0.95$$

↑ interpretation

95% confidence interval for μ is $\left(-\frac{1.96\sigma}{\sqrt{n}} + \bar{X}, \frac{1.96\sigma}{\sqrt{n}} + \bar{X}\right)$

Meaning: for n large if we did m (large) repeated trials and computed the above interval for each trial then true mean would belong to approximately 95% of m intervals calculated.

Confidence Intervals : $\sigma^2 = 1.$

The below is code for finding the confidence interval for a data x .

```
> cifn = function(x, alpha=0.95){  
+ z = qnorm( (1-alpha)/2, lower.tail=FALSE)  
+ sdx = sqrt(1/length(x))  
+ c(mean(x) - z*sdx, mean(x) + z*sdx)  
+ }
```

$$\underbrace{\alpha=0.95}_{\text{}} \left(\bar{x} - \frac{1.96}{\sqrt{n}}, \bar{x} + \frac{1.96}{\sqrt{n}} \right)$$

Three Confidence Intervals for Normal(0,1)

```
> x1 = rnorm(100,0,1);y = cifn(x1)
```

```
> y
```

```
[1] -0.35705304  0.03493976
```

```
> x2 = rnorm(100,0,1);z = cifn(x2)
```

```
> z
```

```
[1] -0.2832489  0.1087439
```

```
> x3 = rnorm(100,0,1);w = cifn(x3)
```

```
> w
```

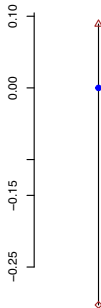
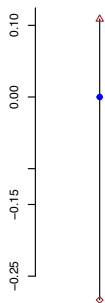
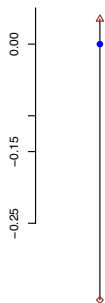
```
[1] -0.30294682  0.08904598
```

Does 0 belong to all the three confidence intervals ?

Confidence Intervals Plots

$rnorm(100, 0, 1) \equiv \text{data}$

The below is a plot of the three confidence intervals computed in the previous slide.



True mean
 $\mu = 0$

Contained in
all 3
c. intervals

Confidence Intervals : 10 Trials

We generate 10 trials of 100 samples from $\text{Normal}(0,1)$ and compute the confidence intervals using the function defined earlier.

```
> normaldata = replicate(10, rnorm(100,0,1),  
+ simplify=FALSE)  
> cidata = sapply(normaldata, cifn)
```

It is easy to check how many of them contain 0.

```
> TRUEIN = cidata[1,]*cidata[2,]<0  
> table(TRUEIN)
```

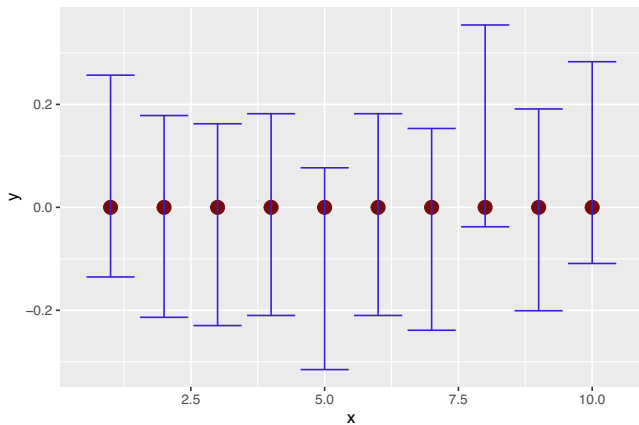
```
TRUEIN
```

```
TRUE
```

```
10
```

$$\begin{aligned} (a, b) \ni 0 \\ \Updownarrow \\ a \cdot b < 0 \end{aligned}$$

Confidence Intervals : 10 Trials



Confidence Intervals: 40 Trials

We generate 10 trials of 100 samples from $\text{Normal}(0,1)$ and compute the confidence intervals using the function defined earlier.

```
> normaldata = replicate(40, rnorm(100,0,1),  
+ simplify=FALSE)  
> cidata = sapply(normaldata, cifn)
```

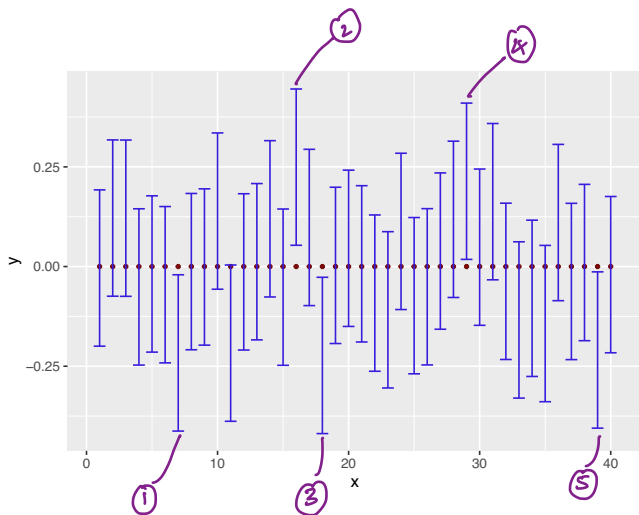
It is easy to check how many of them contain 0.

```
> TRUEIN = cidata[1,]*cidata[2,]<0  
> table(TRUEIN)
```

```
TRUEIN
```

```
FALSE  TRUE  
     5    35
```

Confidence Intervals: 40 trials Plot



Confidence Intervals : 100 Trials

We generate 100 trials of 100 samples from $\text{Normal}(0,1)$ and compute the confidence intervals using the function defined earlier.

```
> normaldata = replicate(100, rnorm(100,0,1),  
+ simplify=FALSE)  
> cidata = sapply(normaldata, cifn)
```

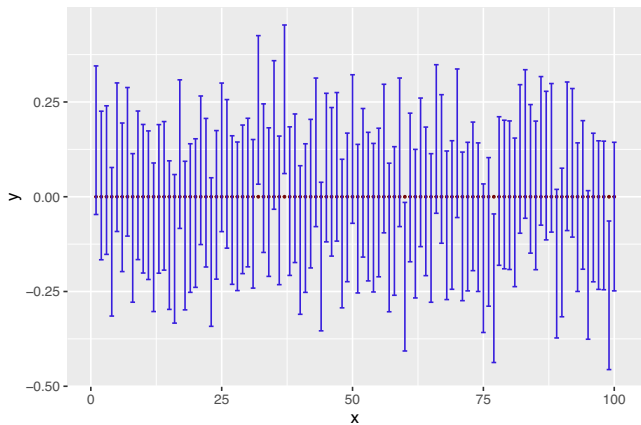
It is easy to check how many of them contain 0.

```
> TRUEIN = cidata[1,]*cidata[2,]<0  
> table(TRUEIN)
```

```
TRUEIN
```

```
FALSE  TRUE  
     5    95
```

Confidence Intervals : 100 Trials



Confidence Intervals : 1000 Trials

We generate 1000 trials of 100 samples from $\text{Normal}(0,1)$ and compute the confidence intervals using the function defined earlier.

```
> normaldata = replicate(1000, rnorm(100,0,1),  
+ simplify=FALSE)  
> cidata = sapply(normaldata, cifn)
```

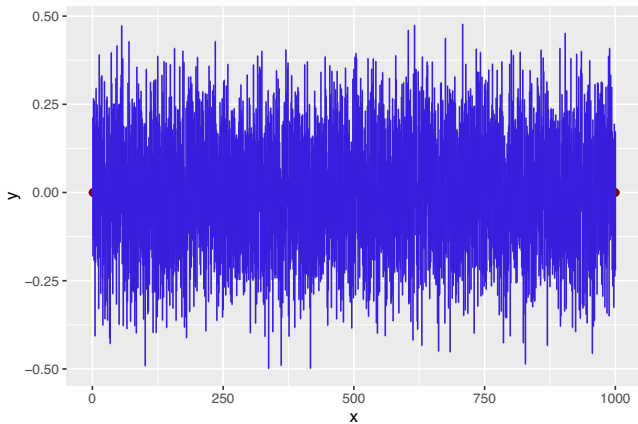
It is easy to check how many of them contain 0.

```
> TRUEIN = cidata[1,]*cidata[2,]<0  
> table(TRUEIN)
```

TRUEIN

FALSE	TRUE
54	946

Confidence Intervals : 1000 Trials



Confidence Intervals

95% confidence interval for μ is $\left(-\frac{1.96\sigma}{\sqrt{n}} + \bar{X}, \frac{1.96\sigma}{\sqrt{n}} + \bar{X}\right)$

Meaning: for n large if we did m (large) repeated trials and computed the above interval for each trial then true mean would belong to approximately 95% of m intervals calculated.

Thus numerically the above meaning seems to hold for a Normal population.