

Consequences :-

- Extremely useful & very necessary to appreciate.
- For practical applications, it is an important tool in inference.

Confidence Interval :-

- Already we have seen the 68, 98, 99 - rule for identifying with normal.
- $P(|Z| \leq 1.96) = 0.95$, $Z \sim N(0,1)$

Suppose $\sqrt{n}(\bar{X}_n - \mu) \stackrel{\text{"\approx"} }{\sim} Z$ for large enough n

Then $P\left(\left|\frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma}\right| \leq 1.96\right) \stackrel{\text{very close}}{\approx} 0.95$

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

$$\Leftrightarrow P\left(-\frac{\sigma}{\sqrt{n}} 1.96 + \bar{X}_n \leq \mu \leq \bar{X}_n + \frac{1.96\sigma}{\sqrt{n}}\right) = 0.95$$

true mean

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

So instead of giving a - point estimator -
 \bar{X}_n we are able to provide
 a Confidence - interval for μ

Questions :-

① How large should n be for
Central limit theorem to take shape?

i.e. when is $\sqrt{n} \left(\bar{X}_n - \mu \right) \xrightarrow{\text{d}} N(0, 1)$

② Is n the same no matter what
the distribution of X is ?

③ What does Confidence interval exactly
mean?

Does it mean : Given a realized interval

$$\left(\frac{-1.96\sigma}{\sqrt{n}} + \bar{X}_n, \frac{1.96\sigma}{\sqrt{n}} + \bar{X}_n \right) \text{ Then is}$$

95% chance μ is in it?

④ What if σ is not known?

⑤ Can we say something for n small?

Confidence Intervals

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 .

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.

Confidence Intervals

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)  
+ }
```

Three Confidence Intervals for Normal(0,1)

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

```
> y
```

```
[1] -0.29570433 0.09628847
```

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

```
> z
```

```
[1] -0.2396115 0.1523813
```

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

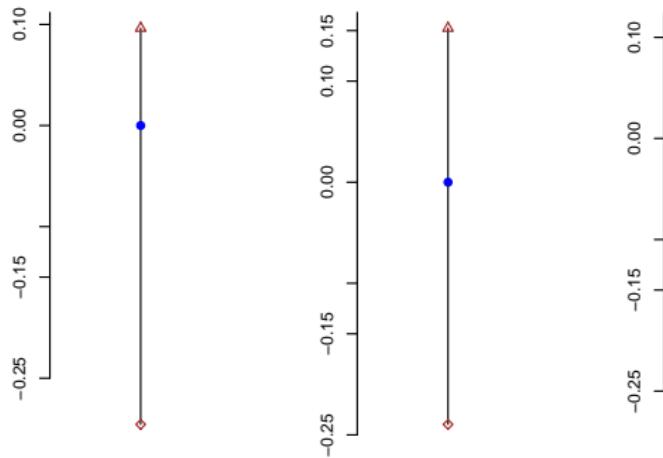
```
> w
```

```
[1] -0.2829300 0.1090628
```

Does 0 belong to all the three confidence intervals ?

Confidence Intervals Plots

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



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)
```

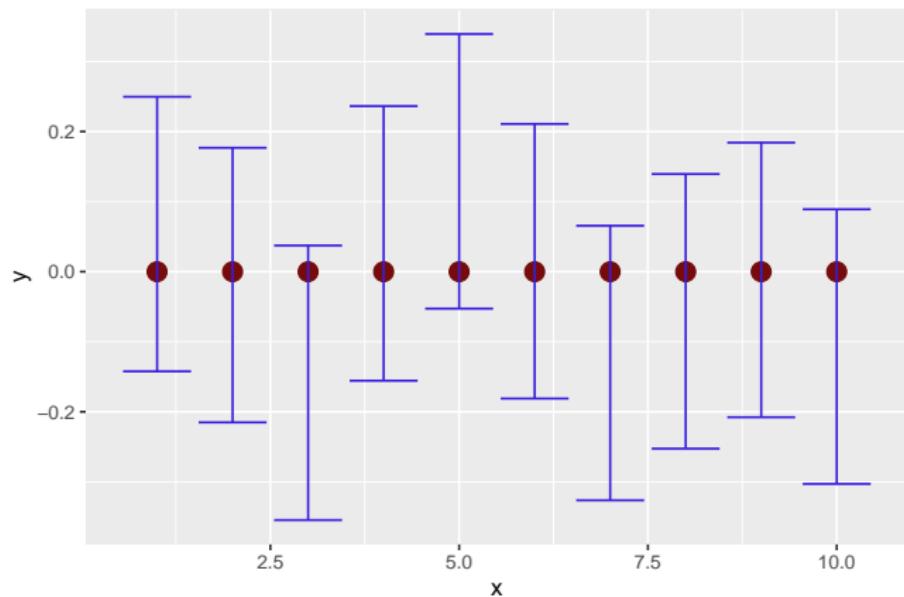
It is easy to check how many of them contain 0.

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

TRUEIN

TRUE

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.

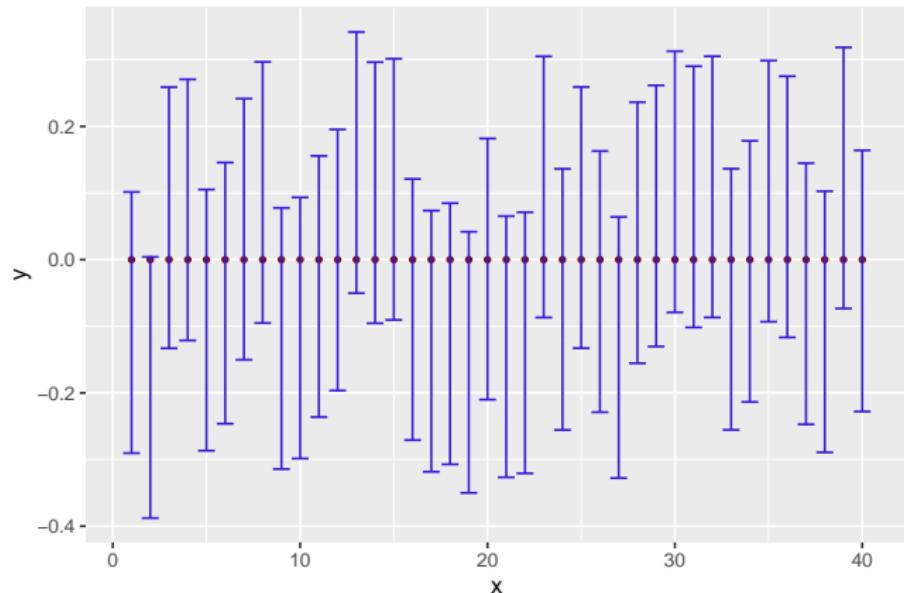
```
> TRUEIN = cidata[1,]*cidata[2,]<0  
> table(TRUEIN)
```

TRUEIN

TRUE

40

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.

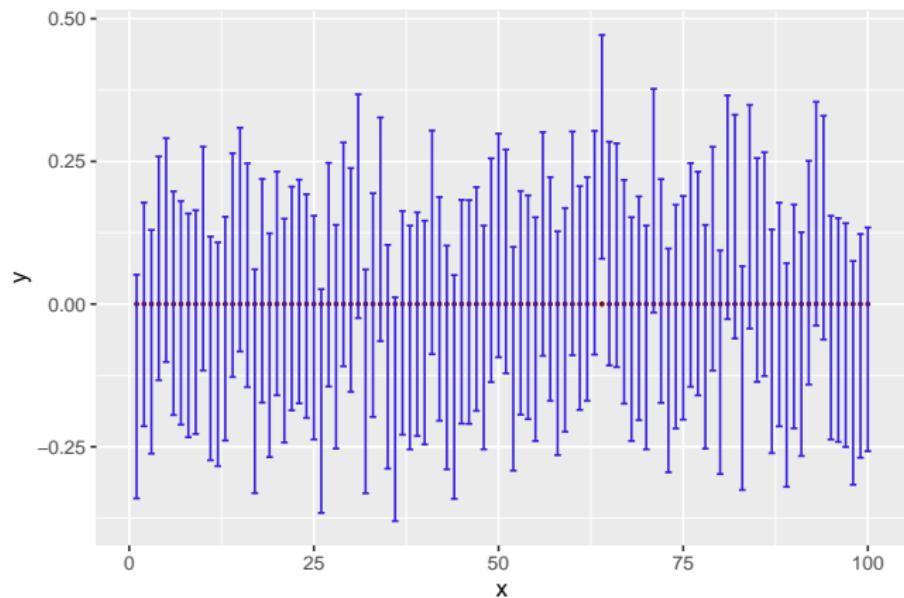
```
> TRUEIN = cidata[1,]*cidata[2,]<0  
> table(TRUEIN)
```

TRUEIN

FALSE TRUE

1 99

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.

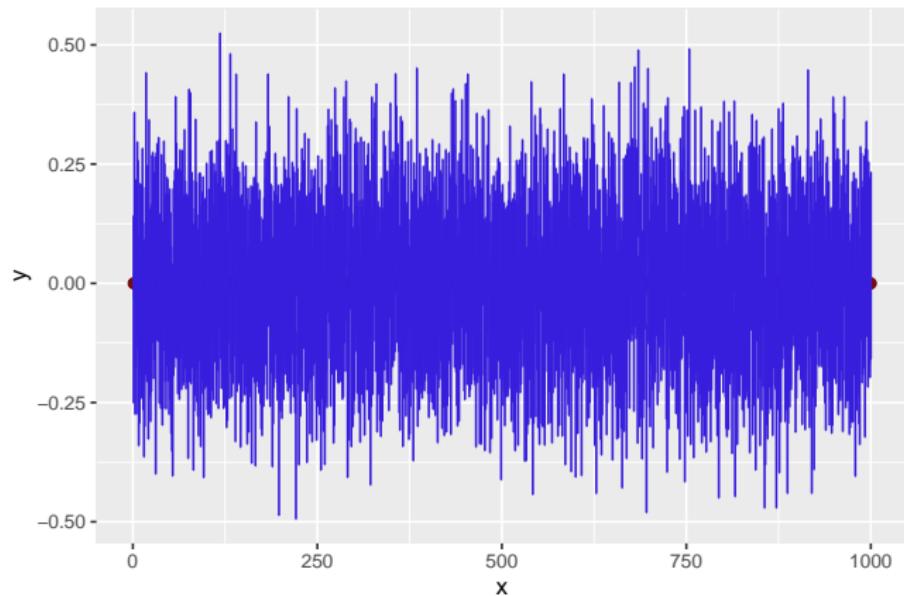
```
> TRUEIN = cidata[1,]*cidata[2,]<0  
> table(TRUEIN)
```

TRUEIN

FALSE TRUE

51 949

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.