STAT234: Lecture 5 - Hypothesis Testing

Kushal K. Dey

▶ Random Experiment- States, Events

- ▶ Random Experiment- States, Events
- Random Variables functions on states

- ▶ Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials

- ▶ Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials
- Probability table for Random variable

- ▶ Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials
- Probability table for Random variable
- Expectation

- ▶ Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials
- ▶ Probability table for Random variable
- Expectation
- Variance

- Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials
- Probability table for Random variable
- Expectation
- Variance
- Moment generating function:

- Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials
- Probability table for Random variable
- Expectation
- Variance
- Moment generating function:
- Sums of random variables

- Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials
- Probability table for Random variable
- Expectation
- Variance
- Moment generating function:
- Sums of random variables
- Expectation and variance of sums of random variables

- Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials
- Probability table for Random variable
- Expectation
- Variance
- Moment generating function:
- Sums of random variables
- Expectation and variance of sums of random variables
- Sampling proportion

- Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials
- Probability table for Random variable
- Expectation
- Variance
- Moment generating function:
- Sums of random variables
- Expectation and variance of sums of random variables
- Sampling proportion
- Generalization to continuous distribution

- Random Experiment- States, Events
- Random Variables functions on states
- Probability proportion of occurrence under infinitely many trials
- Probability table for Random variable
- Expectation
- Variance
- Moment generating function:
- Sums of random variables
- Expectation and variance of sums of random variables
- Sampling proportion
- Generalization to continuous distribution
- ▶ Bernoulli, Binomial and Normal Distribution

Where is the Statistics in all this?

How can I use these concepts?

A Romantic Experiment!

Most people are right-handed and even the right eye is dominant for most people. Molecular biologists have suggested that late-stage human embryos tend to turn their heads to the right. German biopsychologist Onur Güntürkün conjectured that this tendency to turn to the right manifests itself in other ways as well, so he studied kissing couples to see if both people tended to lean to their right more often than to their left (and if so, how strong the tendency is). He and his researchers observed couples from age 13 to 70 in public places such as airports, train stations, beaches, and parks in the United States, Germany, and Turkey. They were careful not to include couples who were holding objects such as luggage that might have affected which direction they turned. In total, 124 kissing pairs were observed with 80 couples leaning right (Nature, 2003).

Assume that the probability of leaning to right while kissing is p

Assume that the probability of leaning to right while kissing is p

This means, if they had taken all possible couples and recorded them kissing, and noted the proportion, it would have been p.

Assume that the probability of leaning to right while kissing is p

This means, if they had taken all possible couples and recorded them kissing, and noted the proportion, it would have been p.

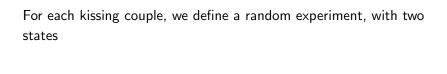
But thats damn difficult !!

Assume that the probability of leaning to right while kissing is p

This means, if they had taken all possible couples and recorded them kissing, and noted the proportion, it would have been p.

But thats damn difficult !!

Researchers however want to know about p, so they sample 124 couples.



S: {Kiss leaning right, Kiss not leaning right}

S: {Kiss leaning right, Kiss not leaning right}

define a random variable X,

$$X = 1$$
 Kiss leaning right (1)
= 0 Kiss not leaning right (2)

S: {Kiss leaning right, Kiss not leaning right}

define a random variable X,

$$X = 1$$
 Kiss leaning right (1)

$$= 0$$
 Kiss not leaning right (2)

(3)

$$Pr[X = 1] := Pr[Kiss leaning right] = p$$

 $\mathcal{S}: \{\textit{Kiss leaning right}, \textit{Kiss not leaning right}\}$

define a random variable X,

$$X = 1$$
 Kiss leaning right (1)
= 0 Kiss not leaning right (2)
(3)

$$Pr[X = 1] := Pr[Kiss leaning right] = p$$

We do not know p !! Week #5: Page 7 of 38

This is same as saying, they observed realizations of 124 random variables X_1, X_2, \dots, X_{124} .

This is same as saying, they observed realizations of 124 random variables X_1, X_2, \dots, X_{124} .

And they observed 80 of them gave 1 and the rest 48 gave 0.

This is same as saying, they observed realizations of 124 random variables X_1, X_2, \dots, X_{124} .

And they observed 80 of them gave 1 and the rest 48 gave 0.

What is 80?

$$\sum_{i=1}^{n} X_i = 80$$

$$\hat{p} = \frac{1}{n} \sum_{i=1}^{n} X_i = \frac{80}{124} = \frac{5}{8}$$

 \hat{p} is called the sampling proportion- here proportion of kissing right observed.

▶ Choosing between two statements about a parameter.

- ▶ Choosing between two statements about a parameter.
- ▶ Is the true probability p= 0.5?

- ▶ Choosing between two statements about a parameter.
- ▶ Is the true probability p=0.5?
- $H_0: \hat{p} = 0.5 \text{ vs } H_1: \hat{p} \neq 0.5 ?$

- ▶ Choosing between two statements about a parameter.
- ▶ Is the true probability p=0.5?
- ► $H_0: \hat{p} = 0.5 \text{ vs } H_1: \hat{p} \neq 0.5$? What's wrong?

- ▶ Choosing between two statements about a parameter.
- ▶ Is the true probability p= 0.5?
- ► $H_0: \hat{p} = 0.5 \text{ vs } H_1: \hat{p} \neq 0.5$? What's wrong?
- $H_0: p = 0.5 \text{ vs } H_1: p \neq 0.5$

- ▶ Choosing between two statements about a parameter.
- ▶ Is the true probability p= 0.5?
- ► $H_0: \hat{p} = 0.5 \text{ vs } H_1: \hat{p} \neq 0.5$? What's wrong?
- ► $H_0: p = 0.5 \text{ vs } H_1: p \neq 0.5$
- ► Two-sided hypothesis. Why?

- ▶ Choosing between two statements about a parameter.
- ▶ Is the true probability p= 0.5?
- ► $H_0: \hat{p} = 0.5 \text{ vs } H_1: \hat{p} \neq 0.5$? What's wrong?
- ► $H_0: p = 0.5 \text{ vs } H_1: p \neq 0.5$
- ► Two-sided hypothesis. Why?

Null: Usually denoted by H_0 :- the null hypothesis refers to a general statement or default position that there is no relationship between two measured phenomena.

Null: Usually denoted by H_0 :- the null hypothesis refers to a general statement or default position that there is no relationship between two measured phenomena.

Example: There is no connection between kissing and direction of leaning.

Null: Usually denoted by H_0 :- the null hypothesis refers to a general statement or default position that there is no relationship between two measured phenomena.

Example: There is no connection between kissing and direction of leaning.



Alternate: Usually denoted by H_1 :- The alternative hypothesis is the hypothesis that is contrary to the null hypothesis. It is usually taken to be that the observations are the result of a real effect.

Alternate: Usually denoted by H_1 :- The alternative hypothesis is the hypothesis that is contrary to the null hypothesis. It is usually taken to be that the observations are the result of a real effect.

Example: There is a connection between kissing and direction of leaning.

Alternate: Usually denoted by H_1 :- The alternative hypothesis is the hypothesis that is contrary to the null hypothesis. It is usually taken to be that the observations are the result of a real effect.

Example: There is a connection between kissing and direction of leaning.



$$\hat{p} = 80/124 = 0.645$$
. Is $p = 0.645$?

- $\hat{p} = 80/124 = 0.645$. Is p = 0.645?
- ▶ What are the possible values of *p* here?

- $\hat{p} = 80/124 = 0.645$. Is p = 0.645?
- ▶ What are the possible values of *p* here?
- ► Fix one possible *p*. What is the 'p-value' for the observed count 80?

- $\hat{p} = 80/124 = 0.645$. Is p = 0.645?
- ▶ What are the possible values of *p* here?
- ► Fix one possible *p*. What is the 'p-value' for the observed count 80?
- p-value = P(observing as 'extreme' as 80).

- $\hat{p} = 80/124 = 0.645$. Is p = 0.645?
- ▶ What are the possible values of *p* here?
- ► Fix one possible *p*. What is the 'p-value' for the observed count 80?
- p-value = P(observing as 'extreme' as 80).

So, under the null H_0 , no relation between kissing and leaning direction.

Mathematically

So, under the null H_0 , no relation between kissing and leaning direction.

Mathematically

$$p = 0.5$$

Now, if the null hypothesis was true,

$$X = 1$$
 with prob 0.5 (4)

$$= 0 \quad with \ prob \ 0.5 \tag{5}$$

(6)

So, under the null H_0 , no relation between kissing and leaning direction.

Mathematically

$$p = 0.5$$

Now, if the null hypothesis was true,

$$X = 1$$
 with prob 0.5 (4)

$$= 0 \quad with \ prob \ 0.5 \tag{5}$$

(6)

and we have observed X_1, X_2, \dots, X_{124} .

So, under the null H_0 , no relation between kissing and leaning direction.

Mathematically

$$p = 0.5$$

Now, if the null hypothesis was true,

$$X = 1$$
 with prob 0.5 (4)
= 0 with prob 0.5 (5)

and we have observed X_1 , X_2 , \cdots , X_{124} .

What is the distribution of 124 P 19 of 38

$$\sum_{i=1}^{124} X_i \sim Bin(124, 0.5)$$

Can you compute $Pr[\sum_{i=1}^{124} X_i = 80]$ under this probability table.

$$\sum_{i=1}^{124} X_i \sim Bin(124, 0.5)$$

Can you compute $Pr[\sum_{i=1}^{124} X_i = 80]$ under this probability table.

Yes!!

$$\sum_{i=1}^{124} X_i \sim Bin(124, 0.5)$$

Can you compute $Pr[\sum_{i=1}^{124} X_i = 80]$ under this probability table.

Yes!!

$$Pr\left[\sum_{i=1}^{124} X_i = 80\right] = {124 \choose 80} 0.5^{80} 0.5^{(124-80)}$$

We can also compute

$$Pr\left[\sum_{i=1}^{124} X_i >= 80\right] = \sum_{y=80}^{124} {124 \choose y} 0.5^y 0.5^{(124-y)}$$

Week #5: Page 14 of 38

What is the probability of observing something as extreme as the observed value.

What is the probability of observing something as extreme as the observed value.

One extreme : $\sum_{i=1}^{124} X_i >= 80$

What is the probability of observing something as extreme as the observed value.

One extreme :
$$\sum_{i=1}^{124} X_i >= 80$$

What if $\sum_{i=1}^{124} X_i$ is very small.

What is the probability of observing something as extreme as the observed value.

One extreme : $\sum_{i=1}^{124} X_i >= 80$

What if $\sum_{i=1}^{124} X_i$ is very small.

$$E(X) = 124 \times 0.5 = 62$$

So, 44 is as far from 62 as 80 is.

What is the probability of observing something as extreme as the observed value.

One extreme : $\sum_{i=1}^{124} X_i >= 80$

What if $\sum_{i=1}^{124} X_i$ is very small.

$$E(X) = 124 \times 0.5 = 62$$

So, 44 is as far from 62 as 80 is.

So, probability of observing something as extreme as observed value should be

pvalue:
$$Pr\left[\sum_{i=1}^{124} X_i >= 80 \atop \text{Week } \#5: \text{ Page} \atop 15 \text{ of } 38 \right] + Pr\left[\sum_{i=1}^{124} X_i <= 44\right]$$

```
upper_tail_p <- pbinom(80,124,0.5,lower.tail=FALSE) + dbinom(80,
lower_tail_p <- pbinom(44,124,0.5,lower.tail=TRUE)
pvalue <- upper_tail_p + lower_tail_p
pvalue</pre>
```

[1] 0.001565

By convention in statistics, we usually say if *pvalue* is less than 0.05, then the null hypothesis is rejected.

```
upper_tail_p <- pbinom(80,124,0.5,lower.tail=FALSE) + dbinom(80,
lower_tail_p <- pbinom(44,124,0.5,lower.tail=TRUE)
pvalue <- upper_tail_p + lower_tail_p
pvalue</pre>
```

[1] 0.001565

By convention in statistics, we usually say if *pvalue* is less than 0.05, then the null hypothesis is rejected.

So, what do we do here?

```
upper_tail_p <- pbinom(80,124,0.5,lower.tail=FALSE) + dbinom(80,
lower_tail_p <- pbinom(44,124,0.5,lower.tail=TRUE)
pvalue <- upper_tail_p + lower_tail_p
pvalue</pre>
```

[1] 0.001565

By convention in statistics, we usually say if *pvalue* is less than 0.05, then the null hypothesis is rejected.

So, what do we do here?

We reject the null or we believe the researcher.

Normal Approximation

But we know that for large n, if

$$\sum_{i=1}^n X_i \sim Bin(n,p)$$

then one can use a normal approximation

$$\sum_{i=1}^{n} X_{i} \sim N(np, np(1-p))$$

Here if we assume n = 124, which is a fair assumption,

$$\sum_{i=1}^{n} X_i \sim N(124 \times 0.5, 124 \times 0.5 \times 0.5)$$

$$\sum_{i=1}^n X_i \sim N(62,31)$$

p-value Normal

under normal approximation, Continuity correction

$$p - value : Pr\left[\sum_{i=1}^{124} X_i > 79.5\right] + Pr\left[\sum_{i=1}^{124} X_i < 44.5\right]$$

Assume $Y = \sum_{i=1}^{124} X_i$,

$$Pr[Y > 79.5] := P(\frac{Y - 62}{\sqrt{32}} > \frac{79.5 - 62}{\sqrt{31}})$$

p-value Normal

under normal approximation, Continuity correction

$$p - value : Pr\left[\sum_{i=1}^{124} X_i > 79.5\right] + Pr\left[\sum_{i=1}^{124} X_i < 44.5\right]$$

Assume $Y = \sum_{i=1}^{124} X_i$,

$$Pr[Y > 79.5] := P(\frac{Y - 62}{\sqrt{32}} > \frac{79.5 - 62}{\sqrt{31}}) = P(z > 3.14) = 0.0008$$

$$Pr[Y < 44.5] := P(\frac{Y - 62}{\sqrt{32}} < \frac{44.5 - 62}{\sqrt{31}})$$

p-value Normal

under normal approximation, Continuity correction

$$p - value : Pr\left[\sum_{i=1}^{124} X_i > 79.5\right] + Pr\left[\sum_{i=1}^{124} X_i < 44.5\right]$$

Assume $Y = \sum_{i=1}^{124} X_i$,

$$Pr[Y > 79.5] := P(\frac{Y - 62}{\sqrt{32}} > \frac{79.5 - 62}{\sqrt{31}}) = P(z > 3.14) = 0.0008$$

$$Pr[Y < 44.5] := P(\frac{Y - 62}{\sqrt{32}} < \frac{44.5 - 62}{\sqrt{31}}) = P(z < -3.14) = 0.0008$$

$$pvalue: 0.0008 + 0.0008 = 0.0016$$

Week #5: Page 18 of 38

Demo of hypothesis testing

We can see a demo of what we discussed so far here:

Demo of hypothesis testing

We can see a demo of what we discussed so far here:

http:

//www.rossmanchance.com/applets/OneProp/OneProp.htm

Okay so now we know that our null did not work. $p \neq 0.5$.

Okay so now we know that our null did not work. $p \neq 0.5$.

What is *p* then?

Okay so now we know that our null did not work. $p \neq 0.5$.

What is *p* then?

We will not know the exact p...

Okay so now we know that our null did not work. $p \neq 0.5$.

What is p then?

We will not know the exact p...

But can we give a probable range of values for p?

Okay so now we know that our null did not work. $p \neq 0.5$.

What is p then?

We will not know the exact p...

But can we give a probable range of values for p?

Yes we can !!

Okay so now we know that our null did not work. $p \neq 0.5$.

What is p then?

We will not know the exact p...

But can we give a probable range of values for p?

Yes we can !!

This is where confidence intervals come in

 \hat{p} was found to be 0.64. Is p = 0.64 ?

 \hat{p} was found to be 0.64. Is p = 0.64 ?

No, but 0.64 may not be too far from the truth right?

 \hat{p} was found to be 0.64. Is p = 0.64 ?

No, but 0.64 may not be too far from the truth right?

if I ask for an interval for possible values of p, what would you choose?

 \hat{p} was found to be 0.64. Is p = 0.64 ?

No, but 0.64 may not be too far from the truth right?

if I ask for an interval for possible values of p, what would you choose?

what if I say $p \in \{0.60, 0.70\}$?

 \hat{p} was found to be 0.64. Is p = 0.64 ?

No, but 0.64 may not be too far from the truth right?

if I ask for an interval for possible values of p, what would you choose?

what if I say $p \in \{0.60, 0.70\}$?

How confident am I about it? Can I give a measure

From the previous classes,

$$\hat{p} = \frac{1}{124} \sum_{i=1}^{124} X_i \sim N\left(p, \frac{p(1-p)}{n}\right)$$

The observed value of $\hat{p} = 0.64$.

From the previous classes,

$$\hat{p} = \frac{1}{124} \sum_{i=1}^{124} X_i \sim N\left(p, \frac{p(1-p)}{n}\right)$$

The observed value of $\hat{p} = 0.64$.

What does it mean by $p \in \{0.60, 0.70\}$?

From the previous classes,

$$\hat{p} = \frac{1}{124} \sum_{i=1}^{124} X_i \sim N\left(p, \frac{p(1-p)}{n}\right)$$

The observed value of $\hat{p} = 0.64$.

What does it mean by $p \in \{0.60, 0.70\}$?

This can be written as $p > \hat{p} - 0.04$ and $p < \hat{p} + 0.06$.

From the previous classes,

$$\hat{p} = \frac{1}{124} \sum_{i=1}^{124} X_i \sim N\left(p, \frac{p(1-p)}{n}\right)$$

The observed value of $\hat{p} = 0.64$.

What does it mean by $p \in \{0.60, 0.70\}$?

This can be written as $p > \hat{p} - 0.04$ and $p < \hat{p} + 0.06$.

This means $\hat{p} - p < 0.04$ and $\hat{p} - p > -0.06$

From the previous classes,

$$\hat{p} = \frac{1}{124} \sum_{i=1}^{124} X_i \sim N\left(p, \frac{p(1-p)}{n}\right)$$

The observed value of $\hat{p} = 0.64$.

What does it mean by $p \in \{0.60, 0.70\}$?

This can be written as $p > \hat{p} - 0.04$ and $p < \hat{p} + 0.06$.

This means $\hat{p} - p < 0.04$ and $\hat{p} - p > -0.06$

But \hat{p} is actually a random variable $\frac{1}{124} \sum_{i=1}^{124} X_i$

If we had repeated this experiment infinitely many times, what proportion of cases, would we have got

If we had repeated this experiment infinitely many times, what proportion of cases, would we have got

$$-0.06 < \hat{p} - p < 0.04$$

If we had repeated this experiment infinitely many times, what proportion of cases, would we have got

$$-0.06 < \hat{p} - p < 0.04$$

This is given by

$$Pr[-0.06 < \hat{p} - p < 0.04]$$

If we had repeated this experiment infinitely many times, what proportion of cases, would we have got

$$-0.06 < \hat{p} - p < 0.04$$

This is given by

$$Pr[-0.06 < \hat{p} - p < 0.04]$$

$$\hat{p} \sim N\left(p, \frac{p(1-p)}{n}\right)$$

If we had repeated this experiment infinitely many times, what proportion of cases, would we have got

$$-0.06 < \hat{p} - p < 0.04$$

This is given by

$$Pr[-0.06 < \hat{p} - p < 0.04]$$

$$\hat{p} \sim N\left(p, \frac{p(1-p)}{n}\right)$$

$$\sqrt{n}rac{\hat{
ho}-
ho}{\sqrt{
ho(1-
ho)}}\sim N(0,1)$$

$$\sqrt{n} \frac{\hat{p} - p}{\sqrt{p(1-p)}} \sim N(0,1)$$

$$Pr\left[-\sqrt{n}\frac{0.06}{\sqrt{p(1-p)}} < \sqrt{n}\frac{(\hat{p}-p)}{\sqrt{p(1-p)}} < \sqrt{n}\frac{0.04}{\sqrt{p(1-p)}}\right]$$

$$Pr(Z < \sqrt{n}\frac{0.04}{\sqrt{p(1-p)}}) - Pr(Z < -\sqrt{n}\frac{0.06}{\sqrt{p(1-p)}})$$

I can calculate that from normal table had I known p, but I dont.

$$\sqrt{n} \frac{\hat{p} - p}{\sqrt{p(1-p)}} \sim N(0,1)$$

$$Pr\left[-\sqrt{n}\frac{0.06}{\sqrt{p(1-p)}} < \sqrt{n}\frac{(\hat{p}-p)}{\sqrt{p(1-p)}} < \sqrt{n}\frac{0.04}{\sqrt{p(1-p)}}\right]$$

$$Pr(Z < \sqrt{n}\frac{0.04}{\sqrt{p(1-p)}}) - Pr(Z < -\sqrt{n}\frac{0.06}{\sqrt{p(1-p)}})$$

I can calculate that from normal table had I known p, but I dont.

What do I do?

$$\sqrt{n} \frac{\hat{p} - p}{\sqrt{p(1-p)}} \sim N(0,1)$$

$$Pr\left[-\sqrt{n}\frac{0.06}{\sqrt{p(1-p)}} < \sqrt{n}\frac{(\hat{p}-p)}{\sqrt{p(1-p)}} < \sqrt{n}\frac{0.04}{\sqrt{p(1-p)}}\right]$$

$$Pr(Z < \sqrt{n}\frac{0.04}{\sqrt{p(1-p)}}) - Pr(Z < -\sqrt{n}\frac{0.06}{\sqrt{p(1-p)}})$$

I can calculate that from normal table had I known p, but I dont.

What do I do?

An approximation would be

$$Pr\left[-\sqrt{n}\frac{0.06}{\sqrt{\hat{p}(1-\hat{p})}} < Z < \sqrt{n}\frac{0.04}{\sqrt{\hat{p}(1-\hat{p})}}\right]$$

$$Pr[-1.414214 < Z < 0.942809] \approx 0.7484612$$

This means that

$$Pr[-0.06 < \hat{p} - p < 0.04] \approx 0.7484612$$

$$Pr[\hat{p} - 0.04$$

$$Pr\left[\hat{p} - 0.04$$

$$Pr[\hat{p} - 0.04$$

This means if we repeat the experiment of sampling 124 kissing couples infinitely many times and recorded the proportion of kisses by leaning right, $\hat{\rho}$, then around 75% times, we will have

$$p \in [\hat{p} - 0.04, \hat{p} + 0.06]$$

$$Pr[\hat{p} - 0.04$$

This means if we repeat the experiment of sampling 124 kissing couples infinitely many times and recorded the proportion of kisses by leaning right, $\hat{\rho}$, then around 75% times, we will have

$$p \in [\hat{p} - 0.04, \hat{p} + 0.06]$$

What is random here?

$$Pr[\hat{p} - 0.04$$

This means if we repeat the experiment of sampling 124 kissing couples infinitely many times and recorded the proportion of kisses by leaning right, $\hat{\rho}$, then around 75% times, we will have

$$p \in [\hat{p} - 0.04, \hat{p} + 0.06]$$

What is random here?

The interval $[\hat{p} - 0.04, \hat{p} + 0.06]$, because \hat{p} is random.

$$Pr[\hat{p} - 0.04$$

This means if we repeat the experiment of sampling 124 kissing couples infinitely many times and recorded the proportion of kisses by leaning right, $\hat{\rho}$, then around 75% times, we will have

$$p \in [\hat{p} - 0.04, \hat{p} + 0.06]$$

What is random here?

The interval $[\hat{p} - 0.04, \hat{p} + 0.06]$, because \hat{p} is random.

Here the observed realization is $\hat{p} = 0.64$. So, for the given sample, the confidence interval is equal to

CI for
$$p:[0.64-0.04,0.64+0.06]=[0.60,0.70]$$

Here the observed realization is $\hat{p} = 0.64$. So, for the given sample, the confidence interval is equal to

CI for
$$p:[0.64-0.04,0.64+0.06]=[0.60,0.70]$$

.

Still confused?

Here the observed realization is $\hat{p} = 0.64$. So, for the given sample, the confidence interval is equal to

CI for
$$p:[0.64-0.04,0.64+0.06]=[0.60,0.70]$$

.

Still confused?

Following 4 statements:- Correct or incorrect ?

Following 4 statements:- Correct or incorrect?

► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7).

Following 4 statements:- Correct or incorrect?

► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7). Incorrect.

Following 4 statements:- Correct or incorrect?

► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?

Following 4 statements:- Correct or incorrect?

- ► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ In 75% of all possible samples, the true *p* lies in the interval (0.6, 0.7).

Following 4 statements:- Correct or incorrect?

- ► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ In 75% of all possible samples, the true *p* lies in the interval (0.6, 0.7). Incorrect.

- ► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ In 75% of all possible samples, the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?

- ► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ In 75% of all possible samples, the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ► There is 95% probability that the true p lies in the random interval $(\hat{p} 0.04, \hat{p} + 0.06)$.

- ► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ In 75% of all possible samples, the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ There is 95% probability that the true p lies in the random interval ($\hat{p} 0.04, \hat{p} + 0.06$). Correct.

- ► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ In 75% of all possible samples, the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ There is 95% probability that the true p lies in the random interval ($\hat{p} 0.04, \hat{p} + 0.06$). Correct.
- If we repeatedly draw samples and calculate confidence intervals using this procedure, 75 % of these intervals $(\hat{p} 0.04, \hat{p} + 0.06)$ will cover the true p.

- ► There is a 75 % probability that the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ In 75% of all possible samples, the true *p* lies in the interval (0.6, 0.7). Incorrect. Why?
- ▶ There is 95% probability that the true p lies in the random interval ($\hat{p} 0.04, \hat{p} + 0.06$). Correct.
- If we repeatedly draw samples and calculate confidence intervals using this procedure, 75 % of these intervals $(\hat{p} 0.04, \hat{p} + 0.06)$ will cover the true p. Correct.

I will start with a normal table and try to figure out an interval in the standard normal table that contains 95% of the area

$$Pr[-1.96 < Z < 1.96] = 0.95$$

I will start with a normal table and try to figure out an interval in the standard normal table that contains 95% of the area

$$Pr[-1.96 < Z < 1.96] = 0.95$$

Now backtrack,

$$Pr\left[-1.96 < \sqrt{n} \frac{(\hat{p}-p)}{\sqrt{p(1-p)}} < 1.96\right] = 0.95$$

I will start with a normal table and try to figure out an interval in the standard normal table that contains 95% of the area

$$Pr[-1.96 < Z < 1.96] = 0.95$$

Now backtrack,

$$Pr\left[-1.96 < \sqrt{n} \frac{(\hat{p} - p)}{\sqrt{p(1-p)}} < 1.96\right] = 0.95$$

Or approximately,

$$Pr\left[-1.96 < \sqrt{n} \frac{(\hat{p} - p)}{\sqrt{\hat{p}(1 - \hat{p})}} < 1.96\right] = 0.95$$

Since we do not know p, we can try to approximate by

$$Pr\left[\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}$$

Since we do not know p, we can try to approximate by

$$Pr\left[\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}$$

So, the confidence interval that will contain p 95% times is

$$p \in \left[\hat{p} - 1.96 \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}, \hat{p} + 1.96 \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}\right]$$

Since we do not know p, we can try to approximate by

$$Pr\left[\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}$$

So, the confidence interval that will contain p 95% times is

$$p \in \left[\hat{p} - 1.96 \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}, \hat{p} + 1.96 \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}\right]$$

This confidence interval is called Wald's 95% interval.

Since we do not know p, we can try to approximate by

$$Pr\left[\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}$$

So, the confidence interval that will contain p 95% times is

$$p \in \left[\hat{p} - 1.96 \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}, \hat{p} + 1.96 \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}\right]$$

This confidence interval is called Wald's 95% interval.

What is the 95% confidence interval here?

Since we do not know p, we can try to approximate by

$$Pr\left[\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}$$

So, the confidence interval that will contain p 95% times is

$$ho \in \left[\hat{
ho} - 1.96 rac{\sqrt{\hat{
ho}(1-\hat{
ho})}}{\sqrt{n}}, \hat{
ho} + 1.96 rac{\sqrt{\hat{
ho}(1-\hat{
ho})}}{\sqrt{n}}
ight]$$

This confidence interval is called Wald's 95% interval.

What is the 95% confidence interval here?

Following 4 statements:- Correct or incorrect?

▶ There is a 95 % probability that the true p lies in the interval (0.56, 0.72).

Following 4 statements:- Correct or incorrect?

► There is a 95 % probability that the true *p* lies in the interval (0.56, 0.72). Incorrect.

Following 4 statements:- Correct or incorrect?

► There is a 95 % probability that the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?

- ► There is a 95 % probability that the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- ▶ In 95% of all possible samples, the true p lies in the interval (0.56, 0.72).

- ► There is a 95 % probability that the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- ▶ In 95% of all possible samples, the true *p* lies in the interval (0.56, 0.72). Incorrect.

- ► There is a 95 % probability that the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- ▶ In 95% of all possible samples, the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?

- ► There is a 95 % probability that the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- ▶ In 95% of all possible samples, the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- There is 95% probability that the true p lies in the random interval $\left(\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}},\hat{p}+1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}\right)$.

- ► There is a 95 % probability that the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- ▶ In 95% of all possible samples, the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- There is 95% probability that the true p lies in the random interval $\left(\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}},\hat{p}+1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}\right)$. Correct.

- ► There is a 95 % probability that the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- ▶ In 95% of all possible samples, the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- There is 95% probability that the true p lies in the random interval $\left(\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}},\hat{p}+1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}\right)$. Correct.
- If we repeatedly draw samples and calculate confidence intervals using this procedure, 95 % of these intervals $\left(\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}},\hat{p}+1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}\right) \text{ will cover the true } p.$

- ► There is a 95 % probability that the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- ▶ In 95% of all possible samples, the true *p* lies in the interval (0.56, 0.72). Incorrect. Why?
- There is 95% probability that the true p lies in the random interval $\left(\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}},\hat{p}+1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}\right)$. Correct.
- If we repeatedly draw samples and calculate confidence intervals using this procedure, 95 % of these intervals $\left(\hat{p}-1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}},\hat{p}+1.96\frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}\right) \text{ will cover the true } p.$ Correct.

- Suppose we take a random sample of size n = 124 from a population and calculate a 95% confidence interval for parameter p
- We do not know whether this single interval contains p or not.

- Suppose we take a random sample of size n = 124 from a population and calculate a 95% confidence interval for parameter p
- We do not know whether this single interval contains p or not.
- We do know that in 95% of all possible samples of size n = 124, the interval that we construct by this method (possibly different intervals everytime, why?)

- Suppose we take a random sample of size n = 124 from a population and calculate a 95% confidence interval for parameter p
- We do not know whether this single interval contains p or not.
- We do know that in 95% of all possible samples of size n = 124, the interval that we construct by this method (possibly different intervals everytime, why?) will include p.

- Suppose we take a random sample of size n = 124 from a population and calculate a 95% confidence interval for parameter p
- We do not know whether this single interval contains p or not.
- We do know that in 95% of all possible samples of size n = 124, the interval that we construct by this method (possibly different intervals everytime, why?) will include p.

Suppose we repeat the following procedure multiple times:

- Draw a random sample of size n
- ► Calculate a 95% confidence interval for the sample

- Suppose we take a random sample of size n = 124 from a population and calculate a 95% confidence interval for parameter p
- We do not know whether this single interval contains p or not.
- We do know that in 95% of all possible samples of size n = 124, the interval that we construct by this method (possibly different intervals everytime, why?) will include p.

Suppose we repeat the following procedure multiple times:

- Draw a random sample of size n
- ► Calculate a 95% confidence interval for the sample

95% of the intervals thus constructed will cover the true (unknown) population mean/parameter p.

▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*?

▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*? Is it 50-50?

▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*? Is it 50-50?

▶ You: 95 % confident, Sir !

- ▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*? Is it 50-50?
- ▶ You: 95 % confident, Sir!
- Boss: Good job! But I need you to be 99 % sure before I take this to the board.

- ▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*? Is it 50-50?
- ▶ You: 95 % confident, Sir!
- Boss: Good job! But I need you to be 99 % sure before I take this to the board.
- You: Give me a minute.

- ▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*? Is it 50-50?
- ▶ You: 95 % confident, Sir!
- Boss: Good job! But I need you to be 99 % sure before I take this to the board.
- You: Give me a minute. I need to make the interval narrower/ wider.

- ▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*? Is it 50-50?
- ▶ You: 95 % confident, Sir!
- Boss: Good job! But I need you to be 99 % sure before I take this to the board.
- You: Give me a minute. I need to make the interval narrower/ wider.
- Narrower or wider?

A conversation between you and your boss

- ▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*? Is it 50-50?
- ▶ You: 95 % confident, Sir!
- Boss: Good job! But I need you to be 99 % sure before I take this to the board.
- You: Give me a minute. I need to make the interval narrower/ wider.
- Narrower or wider? By 'trial and error' (0.53, 0.74) is the 99% Confidence interval

A conversation between you and your boss

- ▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*? Is it 50-50?
- ▶ You: 95 % confident, Sir!
- Boss: Good job! But I need you to be 99 % sure before I take this to the board.
- You: Give me a minute. I need to make the interval narrower/ wider.
- ► Narrower or wider? By 'trial and error' (0.53, 0.74) is the 99% Confidence interval
- Wald's way: $\hat{p} \pm 2.5 \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}} \approx 99 \% \text{ CI}$

A conversation between you and your boss

- ▶ Boss: How confident are you that this interval (0.56, 0.72) contains the true *p*? Is it 50-50?
- ▶ You: 95 % confident, Sir!
- Boss: Good job! But I need you to be 99 % sure before I take this to the board.
- You: Give me a minute. I need to make the interval narrower/ wider.
- ► Narrower or wider? By 'trial and error' (0.53, 0.74) is the 99% Confidence interval
- Wald's way: $\hat{p} \pm 2.5 \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}} \approx 99 \% \text{ CI}$

Applet

http://www.rossmanchance.com/applets/ConfSim.html

Is the CI unique?

if we follow normal distribution table, you can check

$$Pr[-1.15 < Z < 1.15] = 0.75$$

This would lead to p belonging to the interval

$$ho \in \left[\hat{
ho} - 1.15 rac{\sqrt{\hat{
ho}(1-\hat{
ho})}}{\sqrt{n}}, \hat{
ho} + 1.15 rac{\sqrt{\hat{
ho}(1-\hat{
ho})}}{\sqrt{n}}
ight]$$

around 75% times. This reduces to the CI (0.59, 0.689) as a 75% confidence interval for p given that $\hat{p} = 0.64$.

But we already saw another 75% confidence interval (0.6, 0.7), which is not symmetric about 0.64.

Suppose we have data X_1, X_2, \cdots, X_n independent variables coming from an distribution with mean μ and population variance σ

Suppose we have data X_1, X_2, \dots, X_n independent variables coming from an distribution with mean μ and population variance σ By Central Limit theorem, if n is large, we can write

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

Suppose we have data X_1, X_2, \dots, X_n independent variables coming from an distribution with mean μ and population variance σ By Central Limit theorem, if n is large, we can write

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

In general, CI:-

estimate \pm margin of error

A
$$(1-lpha)$$
 CI for μ is
$$\bar{\mathbf{x}}\pm\mathbf{z}^*\frac{\sigma}{\sqrt{n}}$$

A
$$(1-lpha)$$
 CI for μ is $ar x\pm z^*rac{\sigma}{\sqrt n}$

Here z^* is the **critical value**, selected so that a standard Normal density has area $(1 - \alpha)$ between $-z^*$ and z^* .

A
$$(1-lpha)$$
 CI for μ is
$$\bar{\mathbf{x}}\pm\mathbf{z}^*\frac{\sigma}{\sqrt{n}}$$

Here z^* is the **critical value**, selected so that a standard Normal density has area $(1 - \alpha)$ between $-z^*$ and z^* .

The quantity $z^*\sigma/\sqrt{n}$, then, is the **margin of error**.

A
$$(1-lpha)$$
 CI for μ is
$$\bar{\mathbf{x}}\pm\mathbf{z}^*\frac{\sigma}{\sqrt{n}}$$

Here z^* is the **critical value**, selected so that a standard Normal density has area $(1 - \alpha)$ between $-z^*$ and z^* .

The quantity $z^*\sigma/\sqrt{n}$, then, is the **margin of error**. In general, CI :- estimate \pm margin of error

A
$$(1-lpha)$$
 CI for μ is
$$\bar{\mathbf{x}}\pm\mathbf{z}^*\frac{\sigma}{\sqrt{n}}$$

Here z^* is the **critical value**, selected so that a standard Normal density has area $(1 - \alpha)$ between $-z^*$ and z^* .

The quantity $z^*\sigma/\sqrt{n}$, then, is the **margin of error**. In general, CI :- estimate \pm margin of error

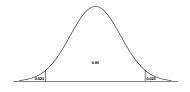
If the population distribution is normal, the interval is exact. Otherwise, it is approximately correct for large n.

Finding z^*

For a given confidence level $(1 - \alpha)$, how do we find z^* ? Let $Z \sim N(0, 1)$:

Finding z*

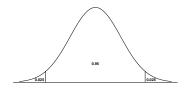
For a given confidence level $(1 - \alpha)$, how do we find z^* ? Let $Z \sim N(0, 1)$:



$$P(-z^* \le Z \le z^*) = (1 - \alpha) \iff P(Z < -z^*) = \frac{\alpha}{2}$$

Finding z^*

For a given confidence level $(1 - \alpha)$, how do we find z^* ? Let $Z \sim N(0, 1)$:

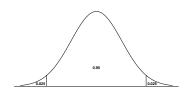


$$P(-z^* \le Z \le z^*) = (1-\alpha) \iff P(Z < -z^*) = \frac{\alpha}{2}$$

Thus, for a given confidence level $(1 - \alpha)$, $z^* = z_{\alpha/2}(\text{upper } \alpha/2 \text{ critical point})$, we can look up the corresponding z^* value on the Normal table.

Finding z^*

For a given confidence level $(1 - \alpha)$, how do we find z^* ? Let $Z \sim N(0, 1)$:



$$P(-z^* \le Z \le z^*) = (1 - \alpha) \iff P(Z < -z^*) = \frac{\alpha}{2}$$

Thus, for a given confidence level $(1 - \alpha)$, $z^* = z_{\alpha/2}(\text{upper } \alpha/2 \text{ critical point})$, we can look up the corresponding z^* value on the Normal table.

Common z^* values:

Confidence				
z^*	Week #5	1,645 _f	₃₈ 1.96	2.576