

2/8/21

Math 341/650

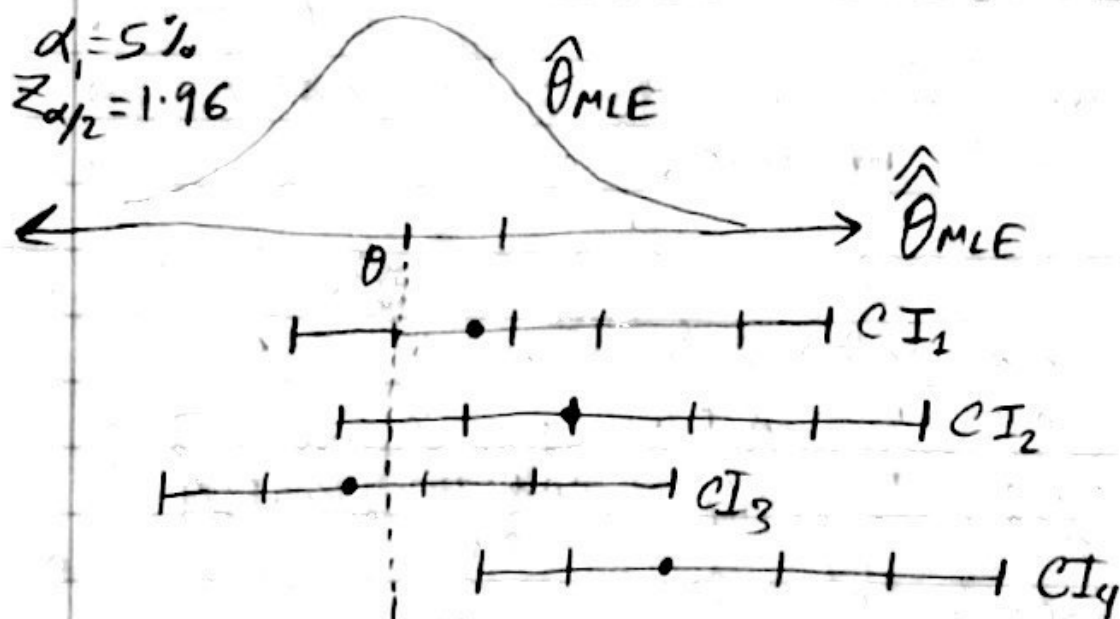
Lec 03

$\hat{\theta}_{MLE} = S(X_1, \dots, X_n)$ $\xrightarrow{\text{F: iid Bern}(\theta)}$
 some function
 of the data X_1, \dots, X_n

$X \in \mathbb{R}$ e.g. 0.1984

$\hat{\theta}_{MLE} = S(X_1, X_2, \dots, X_n) = \bar{X} \sim N(\theta, SE[\hat{\theta}_{MLE}])$
 the same function
 except of the rv's $\xleftarrow{\text{By property (2)}}$
 a rv, not a value
 rv's have distributions

We use this normally to create the confidence interval (CI). why do CI's work?



CI's have 95% probability (before they're realized) of capturing (i.e. including) the real value of θ .

Inference goal #3: "hypothesis testing" also called "theory testing"
Consider a situation where "I" am trying to convince "you" of something.

Scenario I: "I" declare aliens & UFO's exist.
If they don't exist "you" need to provide "me" "sufficient" evidence that Aliens & UFO's don't exist.

Scenario II: "I" will assume for the moment that Aliens & UFO's don't exist, H_0 and I will provide "you" sufficient evidence to the point that "you're" convinced that Aliens & UFO's exist, H_a

Scenario II is more convincing, and how science generally works. The theory "I" am trying to demonstrate is called the "alternative hypothesis" (H_a) since it's alternative to maybe business-as-usual. In scenario II, we assume the opposite of the theory which is called the "null hypothesis" (H_0). This is the "hypothesis Testing" procedure.

In our context, theories are phrases or mathematical statements about θ , the unknown parameter. We will study three types of

H_a 's:

$H_a: \theta \neq \theta_0$	vs.	$H_0: \theta = \theta_0$	two sided test / two tailed test
$H_a: \theta > \theta_0$	vs.	$H_0: \theta \leq \theta_0$	right sided test / right-tailed test
$H_a: \theta < \theta_0$	vs.	$H_0: \theta \geq \theta_0$	left sided test / left-tailed test.

There are two outcomes of a hypothesis test.

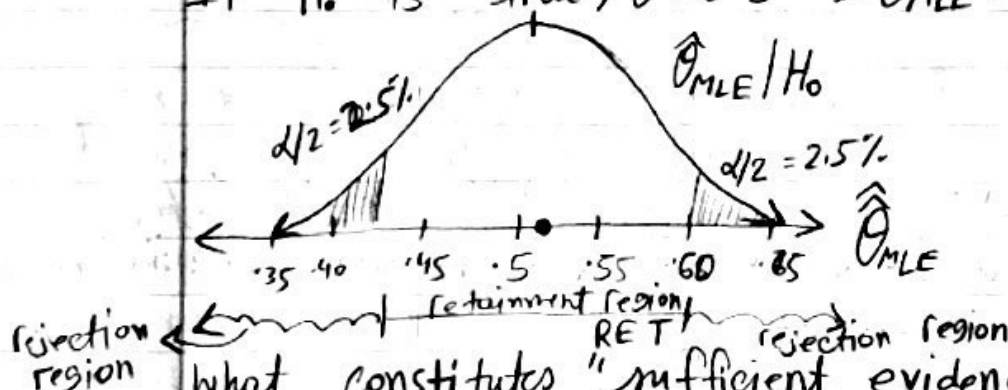
- (A) You were not shown sufficient evidence of H_a . Thus you "fail to reject H_0 " or "retain H_0 ".
 or (B) You were indeed shown sufficient evidence of H_a . Thus you "reject H_0 ", or "accept H_a ".

Imagine you're flipping a coin n times, counting # of heads; $F: \text{iid Bern}(\theta)$. You want to prove the coin is unfair.

$$H_a: \theta \neq 0.5$$

$$H_0: \theta = 0.5$$

If H_0 is true, $\theta = 0.5 \Rightarrow \hat{\theta}_{MLE} \sim N(\bar{x}, \text{SE}[\hat{\theta}_{MLE}]^2)$



What constitutes "sufficient evidence". It's a probability of rejecting when H_0 is true. (α) Everyone is different.

If $\alpha = 5\%$ in a 2-tailed test, we put $1/2$ the probability in each tail. 5% is the most common scientific standard. Thus, we retain H_0 for the most non-weird 95% of the $\hat{\theta}$ and reject H_0 for the 5% most weird $\hat{\theta}$.

→ Retainment region

$$RET = [\theta_0 + z_{\alpha/2} SE[\hat{\theta}_{MLE}]] = [0.5 \pm 1.96 \sqrt{\frac{0.5(1-0.5)}{100}}]$$

i.e. if $\bar{X} = \frac{61}{100} = 0.61, \alpha = 5\% \Rightarrow [0.402, 0.590]$

$\hat{\theta}_{MLE} = 0.61 \notin RET \Rightarrow$ Reject H_0 & conclude the coin is unfair.

e.g. if $\bar{X} = \frac{59}{100} = 0.59$

$\hat{\theta}_{MLE} = 0.59 \notin RET \Rightarrow$ fail to reject H_0 and conclude there's not enough evidence of coin being unfair.

We've covered the "frequentist" approach to statistical inference. But there are problems with it

① $\mathcal{T} = \text{iid Bern}(\theta), X = \langle 0, 0, 0 \rangle$

$\hat{\theta} = \bar{X} = 0$, is that a good point estimate?

NO! You shouldn't be able to say something is absolutely impossible after $n=3$ trials.

$$CI_{\theta, 1-\alpha} = [\theta \pm 1.96 \sqrt{\frac{\theta(1-\theta)}{3}}] = \{0\}$$

Is this a good confidence set? NO. This isn't a good set of "reasonable values".

② What if you have prior knowledge that θ was restricted to e.g. $[0.1, 0.2]$ and not the full $(0, 1)$. You can't "enter that into" your inference.

③ Consider the frequentist interpretation of a CI:
① Before you do the experiment, you have a 95% probability of capturing θ . But that doesn't tell you anything about after your experiment. After

Your experiment you have an interval e.g. $[0.37, 0.45]$ and you can't say:

$$P(\theta \in [0.37, 0.45]) = 0.95$$

no Randomness!!

@95% of CI's will cover θ . But again, I only make one!!! so this interpretation doesn't help me!

In conclusion, any specific CI means Nothing!

- ④ Hypothesis tests result in a binary outcome: either you reject H_0 or you fail to reject H_0 . What if you want to know?

$$P(H_0|x) \text{ or } P(H_a|x) ?$$

You cannot!! one thing you can do is:

$$P\text{-val} := P(\text{seeing } \hat{\theta} \text{ or more extreme} | H_0) \neq P(H_0|x)$$

- ⑤ $X: \text{iid Bern}(\theta), X = \langle 0, 1, 0 \rangle \Rightarrow \hat{\theta}_{MLE} = 0.33$

$$CI_{\theta, 95\%} = [0.33 \pm 1.96 \sqrt{\frac{0.33(0.67)}{3}}] = [-0.20, 0.87]$$

Is this a reasonable confidence set? No, it's outside of the legal parameter space which is $(0, 1)$.

$$\alpha = 0.05 = 5\%,$$

$$H_0: \theta = 0.5 \Rightarrow RET = [0.5 \pm 1.96 \sqrt{\frac{0.5(0.5)}{3}}] = [-0.066, 1.066]$$

Is this a good hypothesis test? NO b/c you NEVER Reject!!

The problem in ⑤ is b/c the asymptotic normality of the MLE doesn't "kick in" until n is large (MLE property 2 is not true yet).

we will solve all these problems with Bayesian Inference. Unfortunately, we will get other problems instead. It's a trade off and a personal decision.