

# Lecture 20 Dec 7, 2014

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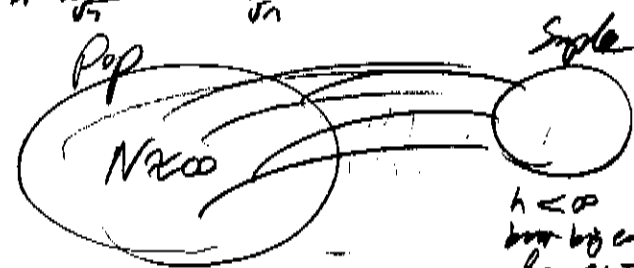
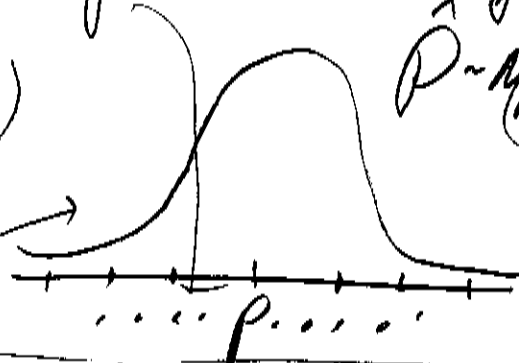
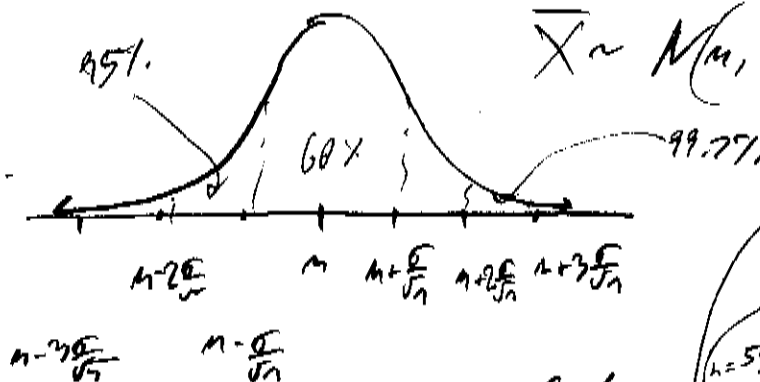
Recall

by CLT

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

possible version

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



$$T_n \sim \text{Binom}(n=594, p)$$

$$n=594 \text{ \# trials } X_n = 116 = \sum_{i=1}^n X_i$$

$$\hat{p} = \frac{\sum_{i=1}^n X_i}{n} = \frac{116}{594} = 0.1953$$

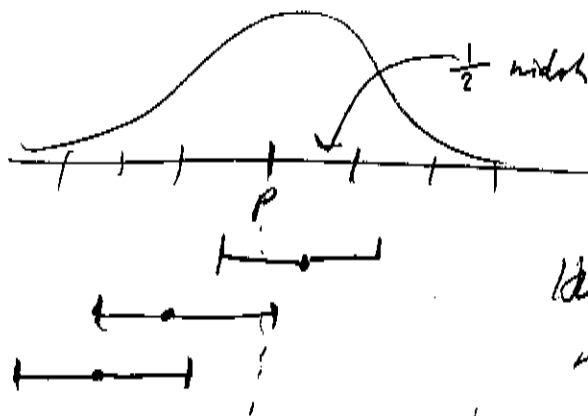
we don't know  $p$ , so we estimate  $\hat{p}$  to replace  $p$ !

get normal  $\rightarrow$  replace  $\hat{p}$  from  $\hat{p}$

Imagine if we construct intervals...

"margin of error"

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



- $P(|Z| > 3) \approx \frac{1}{1000}$
- $P(|Z| > 4) \approx \frac{1}{10,000}$
- $P(|Z| > 5) \approx \frac{1}{10 \text{ million}}$
- $P(|Z| > 6) \approx \frac{1}{1 \text{ billion}}$
- $P(|Z| > 7) \approx \frac{1}{3 \text{ trillion}}$
- $P(|Z| > 8) \approx \frac{1}{1 \text{ quadrillion}}$
- $P(|Z| > 9) \approx \text{R. tails... too small.}$

How often do I "ignore" the real  $p$  in this world?

this comes as long as  $\hat{p}$  is within one SD.

$$\Rightarrow P(\text{coverage}) = P(\hat{p} \in (p - \sqrt{\frac{p(1-p)}{n}}, p + \sqrt{\frac{p(1-p)}{n}})) = P\left(\frac{\hat{p} - p}{\sqrt{\frac{p(1-p)}{n}}} \in (-1, 1)\right) = P(|Z| \leq 1) = 68\%$$

68% is small.

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Imagine you construct a more general interval.

$$\left[ \hat{p} \pm z_{\frac{\alpha}{2}} \sqrt{\hat{p} \frac{1-\hat{p}}{n}} \right] \rightarrow P(p \in \text{lower bound})$$

Coverage?  $P\left(\hat{p} \in \left[p - z_{\frac{\alpha}{2}} \sqrt{\frac{p(1-p)}{n}}, p + z_{\frac{\alpha}{2}} \sqrt{\frac{p(1-p)}{n}}\right]\right)$   
 $= P\left(\frac{\hat{p} - p}{\sqrt{\frac{p(1-p)}{n}}} \in [-z_{\frac{\alpha}{2}}, z_{\frac{\alpha}{2}}]\right) = P(|Z| \leq z_{\frac{\alpha}{2}})$

if  $z_{\frac{\alpha}{2}} = 2 \Rightarrow$  Coverage is 95%.

if  $z_{\frac{\alpha}{2}} = 3 \Rightarrow$  Coverage is 99.7%.

Select a coverage you are comfortable with  
 $1 - \alpha$  where " $\alpha$ " is the prob of  
no coverage

Solve  $1 - \Phi(z_{\frac{\alpha}{2}}) = \frac{\alpha}{2}$  ✓

Look  $\frac{\alpha}{2}$  in  
left or  $\frac{\alpha}{2}$   
on right

e.g.  $\alpha = 5\% \Rightarrow 1 - \alpha = 95\%$   
Coverage

Need 2.5% in the right  
tail. So calculate

$$2.5\% = \int_{z_{2.5\%}}^{\infty} f(z) dz \quad \text{--- PDF st. norm ---} = 2$$

- 2.5%

e.g.  $\alpha = 1\% \Rightarrow 1 - \alpha = 99\%$  Coverage Need 0.5% on right tail.

$$0.5\% = \int_{z_{0.5\%}}^{\infty} f(z) dz \quad \text{---} = 2.56$$

Define

usually we say  $\sqrt{\frac{p(1-p)}{n}}$ . Assume

$$\sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \approx \sqrt{\frac{p(1-p)}{n}}$$

$$C_{p, 1-\alpha} := \left[ \hat{p} \pm z_{\frac{\alpha}{2}} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \right]$$

Ph.D. thesis has been written about this, but will assume it's golden...

"  
to be a  $(1-\alpha)$  confidence interval for the parameter  $p$ , the "true" binomial proportion.

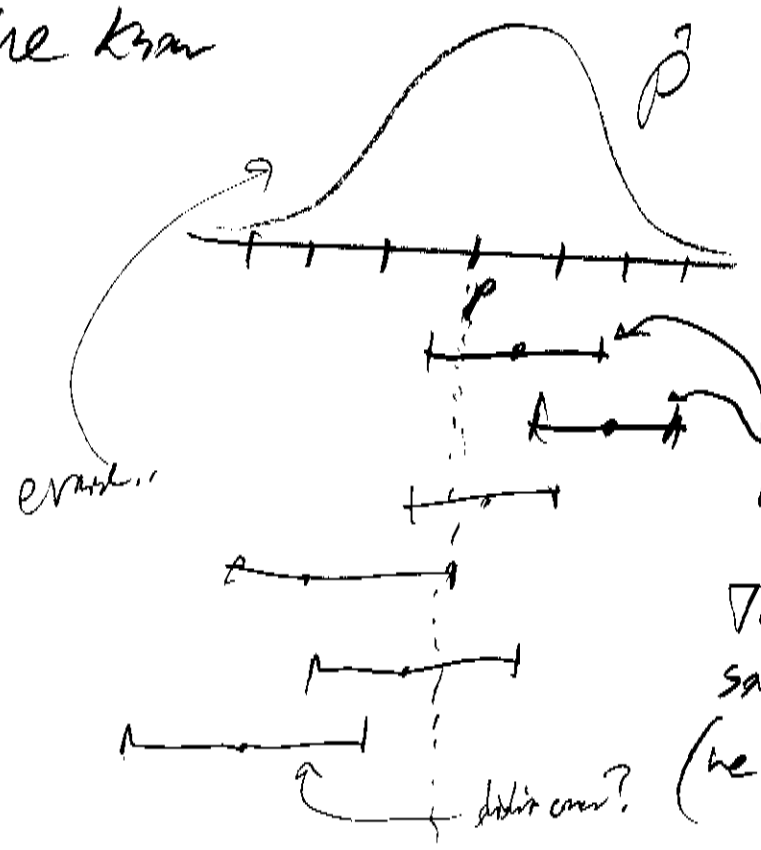
Why do we care? There is a true  $p$  parameter.

We do not know it! We use  $\hat{p}$  to estimate  $p$ .

$E[\hat{p}] = p$ . So on average we are on the right track.

but  $SE[\hat{p}] = \sqrt{\frac{p(1-p)}{n}}$  which means  $\hat{p} \neq p$  exactly due to variance.

We know



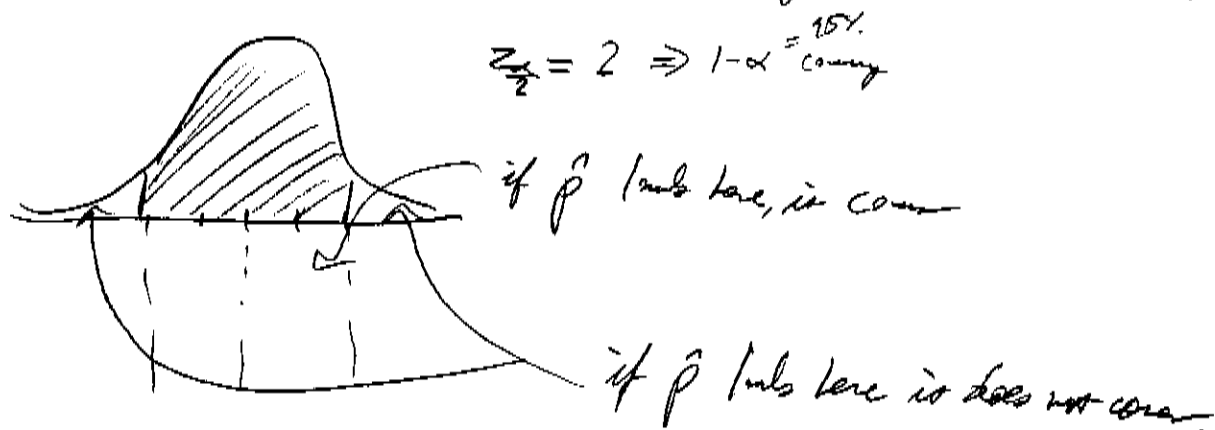
but  $\hat{p}$  can come from anywhere on this distribution!

And we don't know  $p$ . All we see is

we don't even see (---)!

This is the blessing (we know something about  $p$ ) and the curse (we may not cover  $p$ )!

Remember Coverage =  $P(p \in \text{conf. int.})$ . Can you see it from this picture?



## Interpretations

① If many samples are taken, a  $1 - \alpha$  two-sided CI will "capture" the true pop. proportion  $p$ ,  $1 - \alpha$  of the time. "Repeated experiments"

② Before the experiment, the  $1 - \alpha$  two-sided CI has a  $1 - \alpha$  prob of capturing the true pop. proportion.

## Problem

① It does not mean for a given sample that there is a  $1 - \alpha$  prob it covers  $p$ . Imagine  $P(p \in [0.41, 0.59])$ .  $p$  is a constant. The prob statement must be made about a r.v.! Does not make sense. It's degenerate, either 0 or 1.

$\Rightarrow$  CIs do not give you what you really want!!!

$1 - \alpha$  Confidence  $\neq 1 - \alpha$  probability!! "Confidence" is a technical term as it only means one of two things...

Get a 95% CI for  $p$ , the true prop. of blue M&M's.

$$\begin{aligned}
 CI_{p, 95\%} &= \left[ \hat{p} \pm z_{0.5\%} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \right] \quad z_{0.5\%} = 2 \\
 &= \left[ 0.1953 \pm 2 \cdot \sqrt{\frac{0.1953(1-0.1953)}{594}} \right] \\
 &= [0.1628, 0.2278]
 \end{aligned}$$

Interpret! If many <sup>samples</sup> experiments were made, this interval captures  $p$  95% of the time

Get a 99% CI for  $p$ , the true prop. of blue M&M's.

$$\begin{aligned}
 CI_{p, 99\%} &= \left[ \hat{p} \pm z_{0.5\%} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \right] \quad z_{0.5\%} = 2.56 \\
 &= \left[ 0.1953 \pm 2.56 \sqrt{\frac{0.1953(1-0.1953)}{594}} \right] \\
 &= [0.1537, 0.2369]
 \end{aligned}$$

Interpret!

Idea of CI's: don't know  $p$ , make inference for  $p$  using  $\hat{p}$ . Now... what if you hypothesize  $p$  is a certain value, and you want to see if your sample is consistent with the expectation?