

Lecture 20 Prob 241 11/29/15 Recall

1

$X_1, \dots, X_n$  iid sampling with finite  $\mu, \sigma^2$   $Z \sim N(0,1), M_Z(t) = e^{t^2/2}$

$$C_n := \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \quad \left. \vphantom{\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}}} \right\} \text{the standardized } \bar{X}$$

Goal: what is the distribution of  $C_n$  as  $n$  gets big?

from last time

$$C_n = \dots = \frac{Z_1}{\sqrt{n}} + \dots + \frac{Z_n}{\sqrt{n}} \quad \text{st } Z_i := \frac{X_i - \mu}{\sigma} \left. \vphantom{\frac{X_i - \mu}{\sigma}} \right\} \text{the standard X r.v.}$$

$$M_{C_n}(t) = M_{\sqrt{\cdot}}(t) = \left( M_{\frac{Z}{\sqrt{n}}}(t) \right)^n = \left( M_{\frac{Z}{\sqrt{n}}}\left(\frac{t}{\sqrt{n}}\right) \right)^n =$$

$\uparrow$   
 $a = \frac{1}{\sqrt{n}}$

2

$$M_X(t) = 1 + \frac{t E(X)}{1!} + \frac{t^2 E(X^2)}{2!} + \frac{t^3 E(X^3)}{3!} + \dots$$

$$\left( M_2\left(\frac{t}{\sqrt{n}}\right) \right)^n = \left( 1 + \underbrace{\frac{t E(Z)}{\sqrt{n} 1!}}_{=0} + \underbrace{\frac{t^2 E(Z^2)}{(\sqrt{n})^2 2!}}_{=1} + \overbrace{\frac{t^3 E(Z^3)}{(\sqrt{n})^3 3!} + \frac{t^4 E(Z^4)}{(\sqrt{n})^4 4!} + \dots}^{o(n)} \right)^n$$

due to standard

$$= \left( 1 + \frac{t^2/2}{n} + o(n) \right)^n = \left( 1 + \frac{t^2/2}{n} + o\left(\frac{1}{n}\right) \right)^n$$

Now  $o(n) = o\left(\frac{1}{n}\right)$  which means  $\lim_{n \rightarrow \infty} \frac{o(n)}{\frac{1}{n}} = 0$

$o(n) = 1.1$

$$\lim_{n \rightarrow \infty} \frac{\frac{a}{n^{1.5}}}{\frac{1}{n}} + \frac{\frac{b}{n^2}}{\frac{1}{n}} + \dots = \frac{a}{\sqrt{n}} + \frac{b}{n} + \dots = 0 \checkmark$$

$n$	$\ln$
100	5.04
1000	4.48
106	3.99
109	3.08
1012	2.90 - 2.718

$\Rightarrow o(n)$  goes to 0 faster than  $\frac{1}{n}$

he can about  $\lim_{h \rightarrow \infty} C_h$

$$\lim_{h \rightarrow \infty} M_{C_h}(t) = \lim_{h \rightarrow \infty} \left( 1 + \frac{t^{3/2}}{h} + o\left(\frac{1}{h}\right) \right)^h$$

he knows  $e^x = \lim_{h \rightarrow \infty} \left( 1 + \frac{x}{h} \right)^h$

what is  $\lim_{h \rightarrow \infty} \left( 1 + \frac{x}{h} + o\left(\frac{1}{h}\right) \right)^h = e^x$  as well but it may converge slowly!!!!

does this grow quick enough to offset the limit?

turns out ... no so

$$\lim_{h \rightarrow \infty} \left( 1 + \frac{t^{3/2}}{h} + o\left(\frac{1}{h}\right) \right)^h = e^{t^{3/2}} \Rightarrow C_h \sim N(0, 1)$$

as  $h$  gets large.

Central Limit Theorem comes from B2A1!

So  $\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \overset{\text{approx}}{\sim} N(0, 1)$  if  $n$  is large ... How large? (see HW)

$$\Rightarrow Z = \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \Rightarrow \bar{X} = \frac{\sigma}{\sqrt{n}} Z + \mu \stackrel{\text{approx}}{\sim} N\left(\mu, \left(\frac{\sigma}{\sqrt{n}}\right)^2\right)$$

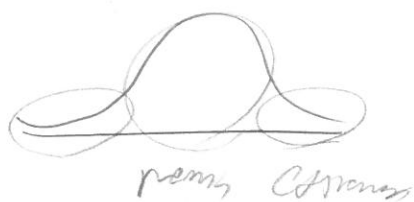
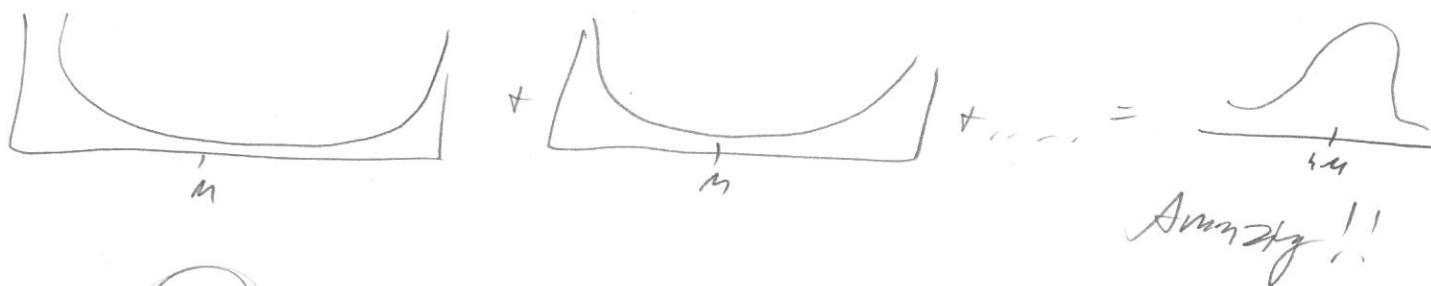
$\bar{X}$  is approx Norm distr. if  $n$  is large

$\Rightarrow$  beginning of Central Limit Theorem

$$T_n = \sqrt{n}(\bar{X} - \mu) = \sigma Z \stackrel{\text{approx}}{\sim} N(0, \sigma^2)$$

$T$  is also approx norm. distr. if  $n$  is large

Imagine  $X \sim$  Bernoulli



$$\frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

$e^{-x^2}$  makes small tails

It is the normal before between means & extremes



already it seems!

Ex Random walk 100 steps. What's the prob you're more than 10 steps away from origin?  $X_1, \dots, X_{100} \sim \text{Random}$   $\mu = 0, \sigma = 1$   $T \stackrel{\text{approx}}{\sim} N(\mu_n, (\sigma\sqrt{n})^2) = N(0, 10^2)$

Ex  $X_1, \dots, X_{10} \stackrel{\text{i.i.d.}}{\sim} \text{Geom}(\frac{1}{2})$   $P(H > 10) = 2P(T > 10) = 2P(Z > \frac{1}{\sqrt{2}}) = 2 \cdot .16 = \boxed{.32}$

What's the prob or avg I wait more than 2.75 flips

$\frac{1}{p} = \frac{1}{1/2} = 2$   
 $\mu = \frac{1}{2}, \sigma = \sqrt{\frac{1-p}{p^2}} = \sqrt{\frac{1-1/2}{(1/2)^2}} = \sqrt{2} = 1.414$   $\frac{\sigma}{\sqrt{n}} = \frac{1.414}{\sqrt{30}} = .258$

$\bar{X} \stackrel{\text{approx}}{\sim} N(2, .258)$   $P(\bar{X} > 2.75) = P\left(\frac{\bar{X} - 2}{.258} > \frac{2.75 - 2}{.258}\right) = P(Z > 3) = \boxed{.0015\%}$

Ex  $X_1, \dots, X_n \stackrel{\text{i.i.d.}}{\sim} \text{Bern}(p)$ . What is the prob  $\bar{X} >$  or  $<$  some  $\gamma$ ?

$\mu = p$   $\sigma = \sqrt{p(1-p)} \Rightarrow \frac{\sigma}{\sqrt{n}} = \sqrt{\frac{p(1-p)}{n}}$

$\bar{X} \stackrel{\text{approx}}{\sim} N\left(p, \left(\sqrt{\frac{p(1-p)}{n}}\right)^2\right)$

e.g. Shoppers are late 2% of the time. Over 10,000 orders. What's the prob more than 3% late on avg?

$\bar{X} \sim \left(.02, \left(\sqrt{\frac{.02 \cdot .98}{10,000}} = .0014\right)^2\right)$   $P(\bar{X} > 3\%) = P\left(Z > \frac{.03 - .02}{.0014}\right) = P(Z > 7.14) \approx 0$   $1.312 \times 10^{-12}$

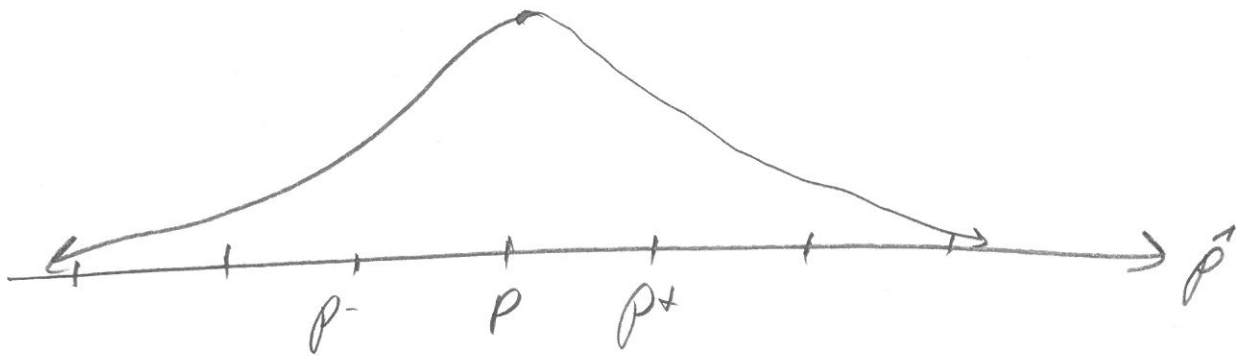
$\rightarrow \bar{X}$  is special. It's the avg # of successes.

It gets a special notation:

$\hat{p}$ ,  $\hat{p} := \frac{\# \text{ successes}}{n} = \bar{X}$

Ex: Who likes mushrooms?

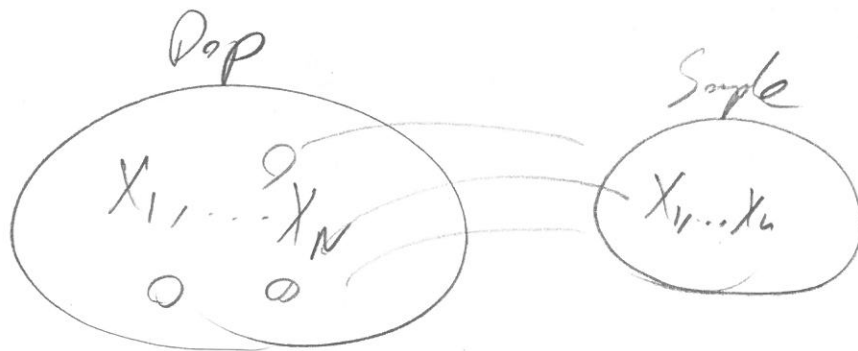
6



$$f(\hat{p}) = ?$$

$\hat{p}$  folks agree here...

ProbT  
Stat. ↓



$|N| = |M|$  infinite

$|n| \ll \infty$   
↑ sample size

Our focus... of more the...

↓

$X_1, \dots, X_n \stackrel{iid}{\sim} \text{Bern}(p)$

if  $n$  is big enough,  
CLT kicks in  
(see HW)

$\hat{p} \sim N(\dots)$

$\hat{p}$  is a random

Representative sample?

Sampling theory? Biased samples.

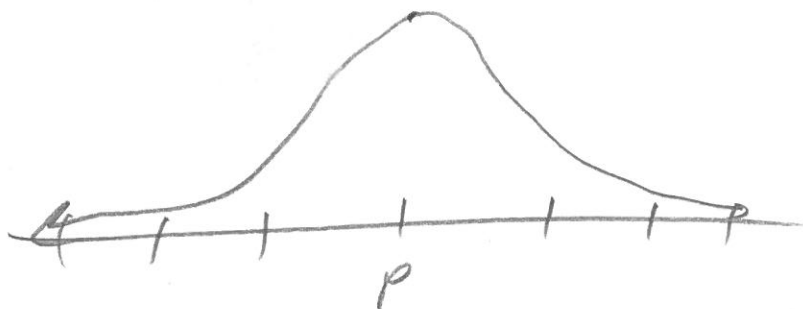
Goal: Use  $\hat{p}$  to infer/guess  $p$ , the parameter.

What are <sup>pop</sup> parameters?? Single values that we don't know.  
 Why? we don't know the pop. we can't see the whole pop.

Best guess  $\hat{p} \approx p$  via  $\bar{X} \approx \mu$  by LLN

↑  
 "point estimate" best estimate of a single pt.

What about an "interval estimate"? An estimate for  $p$  over an interval?  
 Imagine if I did the following:



$\hat{p}$

$\hat{p}$

$\hat{p}$

by def. margin of error

$$\left[ \hat{p} - \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}, \hat{p} + \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \right] = \left[ \underset{\substack{\text{pt est.}}}{\hat{p}} \pm \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \right]$$