# 36-303: Sampling, Surveys and Society

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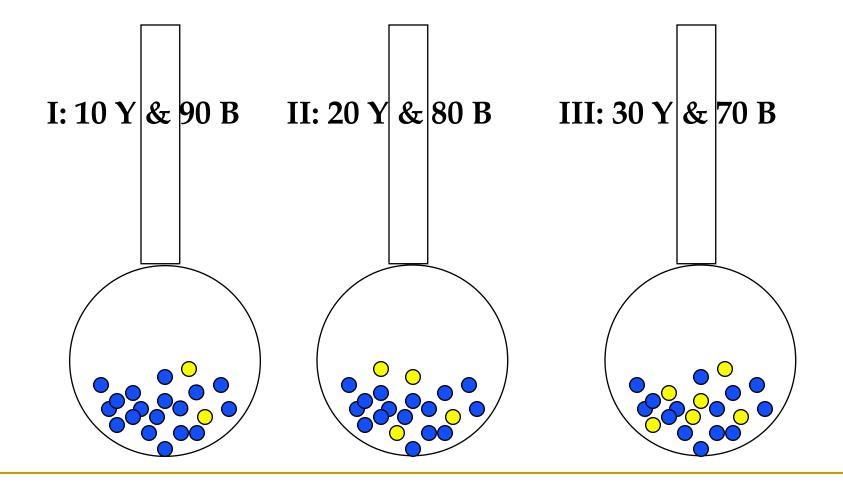
## Handouts

Lecture Notes (only!)

# Outline

- Results of our Survey Sampling Experiment
- Central Limit Theorem??
- Finite Population Correction

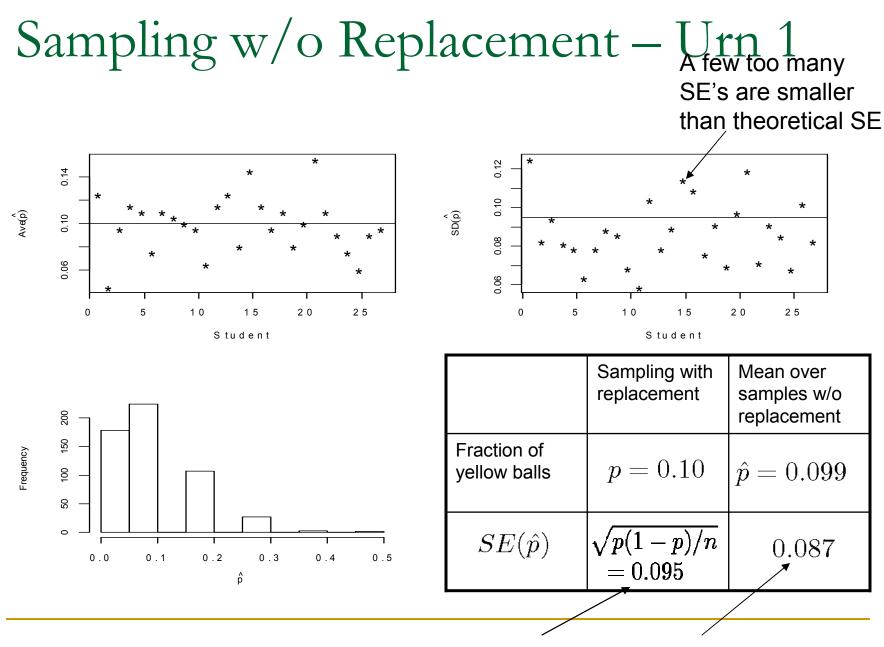
Last Time: Survey Sampling Experiment



#### Last Time: Survey Sampling Experiment

- Circulate all three urns
- Each student should mix the balls; then draw a sample and record # of yellows out of 10
  - Turn in a piece of paper with your name, and 3 neat columns of 20 results each (20 for each urn!)
- Of 39 students in class...
  - □ 27 did Urn 1 (10/90)
  - □ 24 did Urn 2 (20/80)
  - 25 did Urn 3 (30/70)

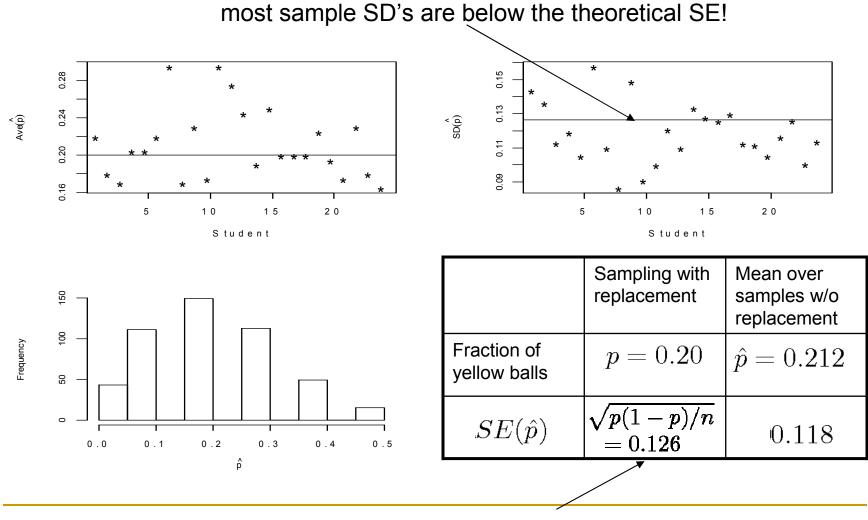
Urn 1 Urn 2 2 0 0 3 1 0 2 1 0 1 1 2	Urn 3 1 2 2 2 2 0 5 2 2 2 2 2 3 1	3 5 2 5 4 2 4 2 1 3 1 1 2	
2 1 0 1 0 0 0 0 0	1 4 1 5 0 2 3	3 3 4 3 2 3 0 3	



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Theoretical SE too big for our samples 6

# Sampling w/o Replacement – Urn 2

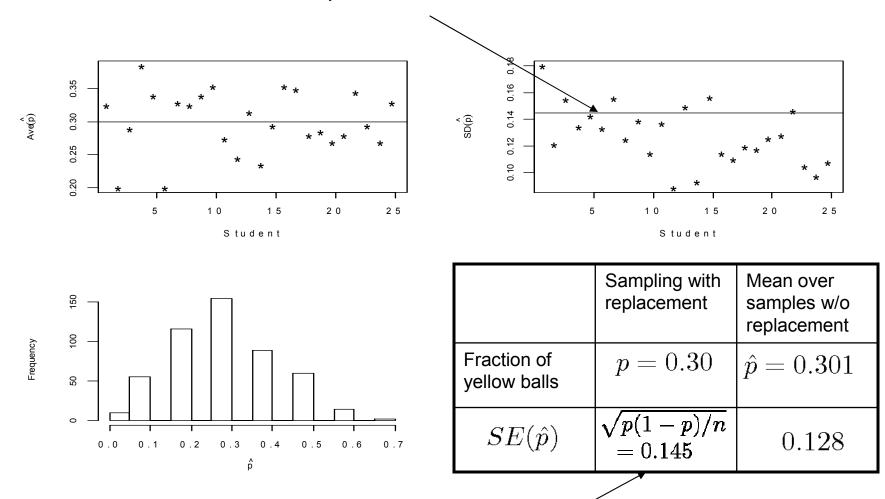


Theoretical SE too big again...

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# Sampling w/o Replacement – Urn 3

most sample SD's are below the theoretical SE!



Again, theoretical SE too big...

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Central Limit Theorem for Surveys?

For simple random sampling (SRS) <u>with</u> <u>replacement</u>,

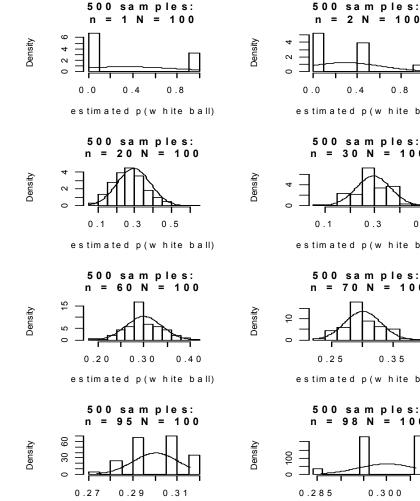
$$E[\overline{X}] = \mu$$
,  $Var(\overline{X}) = \frac{\sigma^2}{n}$ 

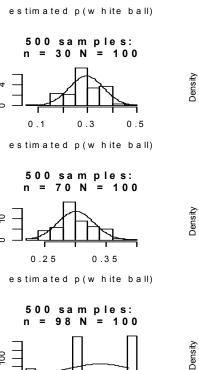
- The Central Limit Theorem then tells us  $\frac{\overline{X} \mu}{\sigma / \sqrt{n}} \sim N(0, 1)$
- $\sigma$  is the SD of X<sub>i</sub>;  $\sigma/\sqrt{n}$  is the SE of  $\overline{X}$
- But in survey sampling we sample <u>w/o replacement</u>!

# Central Limit Theorem for Surveys?

- We will look at 500 draws from Urn 3, at different sample sizes:
  - □ n=1, 2, 5, 10, 20, ..., 98, 99
  - N=100 always
- Compare histogram of  $\hat{p}$ 's with a normal curve with the same center and spread as the  $\hat{p}$ 's
- If CLT holds, histogram & curve will agree
   Agreement should get better as n gets larger!!

#### CLT ? Sampling without Replacement



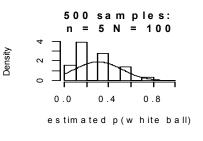


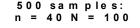
0.300

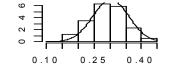
estimated p(w hite ball)

0.4

0.8







estimated p(w hite ball)

500 samples:

n = 80 N = 100

0.30

estimated p(w hite ball)

500 sam ples:

n = 99 N = 100

0.298

estimated p(w hite ball)

0.36

0.304

20

2

0

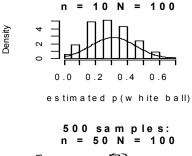
202

300

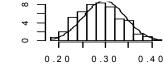
0

0.292

0.24



500 sam ples:

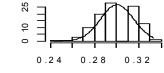


Density

Density

estimated p(w hite ball)





estimated p(w hite ball)

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estimated p(w hite ball)

Conclusions from the CLT Exploration (sampling w/o replacement)

- Small samples CLT hasn't kicked in yet
- For "moderate" samples, CLT seems to work
- Moderate means ... important to have n > 20 (or whatever rule of thumb), but also n/N has to be not close to 1
- CLT breaks when sample size is nearly whole population – then we are more certain about p, than CLT would have us believe

Finite Population Correction

The goal is to figure out what the right SE is

- Requires us to "think differently" about sampling
- Involves a little bit of summation notation tedium
  - Statistics is sometimes like that: we "pay for" good insights with the need for tedious calculation...

# Sampling from a Finite Population

- N = size of our fixed, finite population
- We want to measure Y. Y might be
  - □ income,
  - o 'did you cheat'
  - number of "free" PAT bus rides taken...
- For each person in the population, Y <u>is not random</u>, it is a fixed value: y<sub>1</sub>, y<sub>2</sub>, ..., y<sub>N</sub>
- What is random is whether the person gets in our sample or not:

$$Z_{i} = \begin{cases} 1, & \text{if } i \text{ is in our sample} \\ 0, & \text{if } i \text{ is not in our sample} \end{cases}$$

for i=1, 2, ..., N

The Z's are a "trick" for thinking about how sampling works...

- Population size N = 10
- Sample size n = 3
- y's are respondents' ages

Nonrandom Population y <sub>i</sub> 's	44	35	21	62	27	19	23	56	28	45
Random sampling indicators Z <sub>i</sub> 's	0	0	1	0	0	1	1	0	0	0
Random sample of Y <sub>i</sub> 's			21			19	23			

#### Example: Drawing Balls from an Urn

The colors of the 100 balls were not random. We could say

$$y_i = \begin{cases} 1, & \text{if ball is yellow} \\ 0, & \text{else} \end{cases}$$

What was random was which 10 balls were drawn:
 For 10 balls, Z<sub>i</sub> = 1, for the rest, Z<sub>i</sub>=0

We could write the fraction of yellows in the sample as

$$\hat{p} = \frac{1}{10} \sum_{i=1}^{10} y_i = \frac{1}{10} \sum_{i=1}^{100} Z_i y_i$$

# Sampling Without Replacement

- Population size N
- Sample of size n without replacement.
- What is P[Z<sub>i</sub>=1]?

$$P[Z_i = 1] = \frac{\#(\text{samples of size } n \text{ including } i)}{\#(\text{all possible samples of size } n)}$$

$$= \frac{\#(\text{put } i \text{ in sample}) \times \#(\text{samples of size } n - 1 \text{ from the remaining } N - 1)}{\#(\text{samples of size } n)}$$

$$= \frac{1 \times \binom{N-1}{n-1}}{\binom{N}{n}} = \frac{n}{N} \quad (\text{special case of hypergeometric distribution!})$$

# Sampling Without Replacement

$$E[\overline{Y}_{sample}] = E\left[\frac{1}{n}\sum_{i=1}^{n}Y_{i}\right] = E\left[\frac{1}{n}\sum_{i=1}^{N}Z_{i}y_{i}\right]$$
$$= \frac{1}{n}\sum_{i=1}^{N}y_{i}E[Z_{i}] = \frac{1}{n}\sum_{i=1}^{N}y_{i}\frac{n}{N}$$
$$= \frac{1}{N}\sum_{i=1}^{N}y_{i} = \overline{y}_{pop}$$

Sampling Without Replacement • But the Z's are not independent,  $E[Z_iZ_j] = P[Z_i = 1 \cap Z_j = 1]$   $= P[Z_j = 1|Z_i = 1]P[Z_i = 1]$   $= \left(\frac{n-1}{N-1}\right)\left(\frac{n}{N}\right)$ 

We can calculate the covariance

$$Cov(Z_i, Z_j) = E[Z_j Z_j] - E[Z_i] E[Z_j]$$
$$= \left(\frac{n-1}{N-1}\right) \left(\frac{n}{N}\right) - \left(\frac{n}{N}\right)^2$$
$$= -\frac{1}{N-1} \left(1 - \frac{n}{N}\right) \left(\frac{n}{N}\right)$$

So having i "in" makes j a little less likely…

Sampling Without Replacement  

$$Var(\overline{Y}_{sample}) = Var(\frac{1}{n}\sum_{i=1}^{n}Y_{i}) = Var(\frac{1}{n}\sum_{i=1}^{N}Z_{i}y_{i})$$

$$= \frac{1}{n^{2}}\left[\sum_{i=1}^{N}y_{i}^{2}Var(Z_{i}) + \sum\sum_{i\neq j}y_{i}y_{j}Cov(Z_{i}, Z_{j})\right]$$

$$= \frac{1}{n^{2}}\left[\left(\frac{n}{N}\right)\left(1 - \frac{n}{N}\right)\sum_{i=1}^{N}y_{i}^{2} - \frac{1}{N-1}\left(1 - \frac{n}{N}\right)\left(\frac{n}{N}\right)\sum\sum_{i\neq j}y_{i}y_{j}\right]$$

$$= \frac{1}{n^{2}}\left(\frac{n}{N}\right)\left(1 - \frac{n}{N}\right)\left[\sum_{i=1}^{N}y_{i}^{2} - \frac{1}{N-1}\sum_{i\neq j}y_{i}y_{j}\right]$$

$$= \cdots = \left(1 - \frac{n}{N}\right)\frac{S_{pop}^{2}}{n}$$

where  $S_{pop}^2 = \sum_{1}^{N} (y_i - \overline{y}_{pop})^2 / (N - 1)$ , the population variance.

#### The Finite Population Correction (FPC)

# We have seen that for SRS without replacement

 $E[\overline{Y}_{samp}] = \overline{y}_{pop} \qquad (\overline{Y}_{samp} \text{ is unbiased})$  $Var(\overline{Y}_{samp}) = (1-f)S_{pop}^2/n, \quad f = n/N$ 

# The quantity (1-f) is called the *finite* population correction (fpc).

- □ When  $n/N \approx 0$ , (1-f)  $\approx 1$ , so <u>"With or without</u> replacement doesn't matter for small SRS's!"
- As n/N -> 1, (1-f) -> 0 and SE(ȳ<sub>samp</sub>) ->0. <u>"We</u> don't need statistical estimates for a true census!"

# FPC, continued

In practice we replace  $S_{pop}^2$  with  $s_{samp}^2$ 

$$Var(\overline{Y}_{samp}) \approx (1-f)s^2/n,$$
  
$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \overline{y}_{samp})^2$$

• When  $y_i = 0$  (blue ball) or 1 (yellow ball), one can show, since  $\overline{y}_{samp} = \hat{p}$ 

$$s^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (y_{i} - \hat{p})^{2} = \frac{n}{n-1} \hat{p}(1-\hat{p})$$
  
and so  
$$Var(\hat{p}) \approx (1-f) \frac{1}{n-1} \hat{p}(1-\hat{p})$$

Returning to our Sampling Experiment...

 The SE under SRS <u>w/o</u> replacement should have been

$$SE(\hat{p}) = (1-f)\hat{p}(1-\hat{p})/(n-1)$$

rather than

$$SE(\hat{p}) = \hat{p}(1-\hat{p})/(n-1)$$

This is why, in our urn survey experiment, we saw that estimated SE's from SRS <u>with</u> replacement were <u>too large</u>.

# Comparing SE's

- Urn 1: 10/90
  - With replacement SE = sqrt(0.1\*(1-0.1)/10) = 0.95
  - □ "Without replacement" SE = (1-10/100)\*(0.95) = 0.86
  - Average SE in class samples
     = 0.87
- Urn 2: 20/80
  - "With replacement"  $SE = sqrt(0.2^{(1-0.2)}) = 0.126$
  - □ "Without replacement" SE = (1-10/100)\*(0.126) = 0.113
  - Average SE in class samples
     = 0.118
- Urn 3: 30/70
  - With replacement SE = sqrt(0.3\*(1-0.3)/10) = 0.145
  - □ "Without replacement" SE = (1 -10/100)\*(0.145) = 0.131
  - Average SE in class samples
     = 0.128

# Review

- Results of our Survey Sampling Experiment
- Central Limit Theorem??
- Finite Population Correction
- FOR NEXT WEEK: Groves, Ch's 7 & 8
  - BJ will start on survey questionnaires next Tues
- Turn in next week:
  - Tue: I.3 (BJ will send email about this)
    Tue: HW03