

36-303: Sampling, Surveys and Society

Statistics of Surveys III
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3 February 2011

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Handouts

- Lecture Notes (only!)

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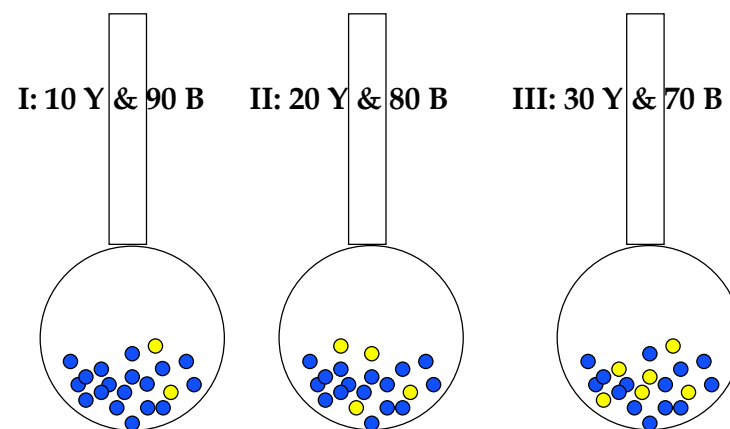
Outline

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

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Last Time: Survey Sampling Experiment



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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)

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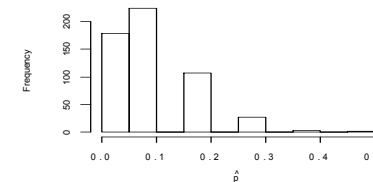
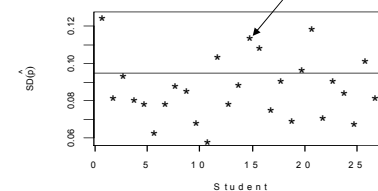
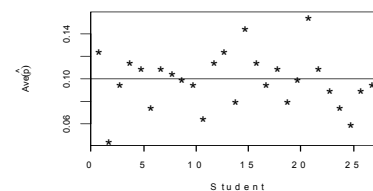
Urn 1	Urn 2	Urn 3
2	1	3
0	2	5
0	2	2
3	2	2
1	0	4
0	2	4
2	9	2
1	1	9
1	2	1
2	4	3
1	1	4
0	5	3
0	0	0
0	0	0

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Sampling w/o Replacement – Urn 1

A few too many SE's are smaller than theoretical SE



	Sampling with replacement	Mean over samples w/o replacement
Fraction of yellow balls	$p = 0.10$	$\hat{p} = 0.099$
$SE(\hat{p})$	$\sqrt{p(1-p)/n} = 0.095$	0.087

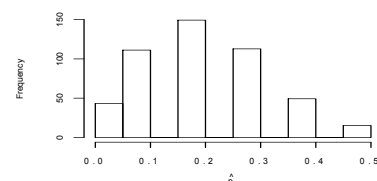
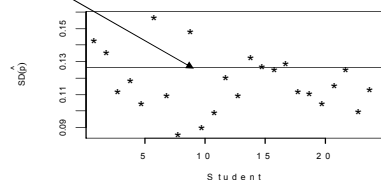
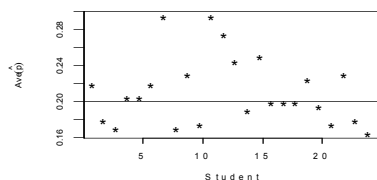
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Theoretical SE too big for our samples

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Sampling w/o Replacement – Urn 2

most sample SD's are below the theoretical SE!



	Sampling with replacement	Mean over samples w/o replacement
Fraction of yellow balls	$p = 0.20$	$\hat{p} = 0.212$
$SE(\hat{p})$	$\sqrt{p(1-p)/n} = 0.126$	0.118

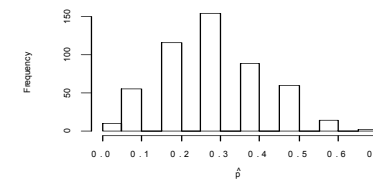
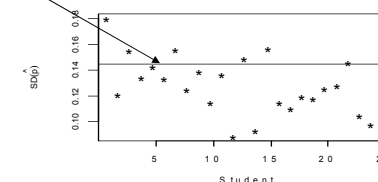
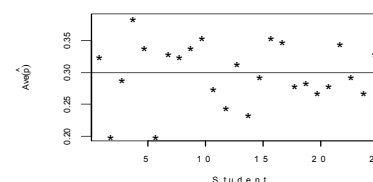
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Theoretical SE too big again...

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Sampling w/o Replacement – Urn 3

most sample SD's are below the theoretical SE!



	Sampling with replacement	Mean over samples w/o replacement
Fraction of yellow balls	$p = 0.30$	$\hat{p} = 0.301$
$SE(\hat{p})$	$\sqrt{p(1-p)/n} = 0.145$	0.128

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Again, theoretical SE too big...

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Central Limit Theorem for Surveys?

- For simple random sampling (SRS) with replacement,

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

- The Central Limit Theorem then tells us

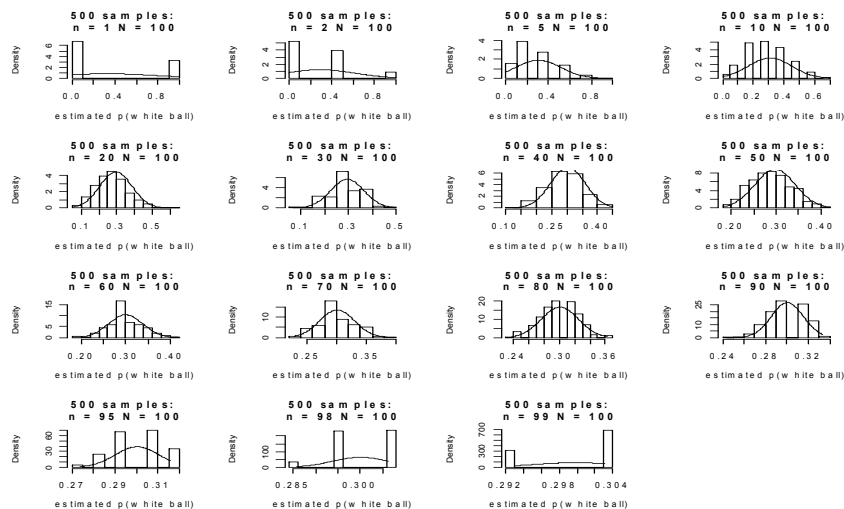
$$\frac{\bar{X} - \mu}{\sigma / \sqrt{n}} \sim N(0, 1)$$

- σ is the SD of X_i ; σ/\sqrt{n} is the SE of \bar{X}
- But in survey sampling we sample w/o replacement!

Central Limit Theorem for Surveys?

- We will look at 500 draws from Urn 3, at different sample sizes:
 - $n=1, 2, 5, 10, 20, \dots, 98, 99$
 - $N=100$ 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



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,
 - ‘did you cheat’
 - number of “free” PAT bus rides taken...
- For each person in the population, Y is not random, it is a fixed value: Y_1, Y_2, \dots, Y_N
- 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 $i=1, 2, \dots, N$

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_i ’s	44	35	21	62	27	19	23	56	28	45
Random sampling indicators Z_i ’s	0	0	1	0	0	1	1	0	0	0
Random sample of Y_i ’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, $Z_i = 1$, for the rest, $Z_i = 0$
 - 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 $P[Z_i=1]$?

$$\begin{aligned}
 P[Z_i = 1] &= \frac{\#(\text{samples of size } n \text{ including } i)}{\#(\text{all possible samples of size } n)} \\
 &= \frac{\#(\text{put } i \text{ in sample}) \times \#(\text{samples of size } n-1 \text{ from the remaining } N-1)}{\#(\text{samples of size } n)} \\
 &= \frac{1 \times \binom{N-1}{n-1}}{\binom{N}{n}} = \frac{n}{N} \quad (\text{special case of hypergeometric distribution!})
 \end{aligned}$$

Sampling Without Replacement

- The Z_i 's are Bernoulli's with

$$E[Z_i] = \frac{n}{N}, \quad \text{Var}(Z_i) = \frac{n}{N} \left(1 - \frac{n}{N}\right)$$

- Therefore

$$\begin{aligned}
 E[\bar{Y}_{\text{sample}}] &= E\left[\frac{1}{n} \sum_{i=1}^n Y_i\right] = E\left[\frac{1}{n} \sum_{i=1}^N Z_i y_i\right] \\
 &= \frac{1}{n} \sum_{i=1}^N y_i E[Z_i] = \frac{1}{n} \sum_{i=1}^N y_i \frac{n}{N} \\
 &= \frac{1}{N} \sum_{i=1}^N y_i = \bar{y}_{\text{pop}}
 \end{aligned}$$

Sampling Without Replacement

- But the Z_i 's are not independent,

$$\begin{aligned}
 E[Z_i Z_j] &= P[Z_i = 1 \cap Z_j = 1] \\
 &= P[Z_j = 1 | Z_i = 1] P[Z_i = 1] \\
 &= \left(\frac{n-1}{N-1}\right) \left(\frac{n}{N}\right)
 \end{aligned}$$

- We can calculate the covariance

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

- So having i "in" makes j a little less likely...

Sampling Without Replacement

$$\begin{aligned}
 \text{Var}(\bar{Y}_{\text{sample}}) &= \text{Var}\left(\frac{1}{n} \sum_{i=1}^n Y_i\right) = \text{Var}\left(\frac{1}{n} \sum_{i=1}^N Z_i y_i\right) \\
 &= \frac{1}{n^2} \left[\sum_{i=1}^N y_i^2 \text{Var}(Z_i) + \sum \sum_{i \neq j} y_i y_j \text{Cov}(Z_i, Z_j) \right] \\
 &= \frac{1}{n^2} \left[\left(\frac{n}{N}\right) \left(1 - \frac{n}{N}\right) \sum_{i=1}^N y_i^2 - \frac{1}{N-1} \left(1 - \frac{n}{N}\right) \left(\frac{n}{N}\right) \sum \sum_{i \neq j} y_i y_j \right] \\
 &= \frac{1}{n^2} \left(\frac{n}{N}\right) \left(1 - \frac{n}{N}\right) \left[\sum_{i=1}^N y_i^2 - \frac{1}{N-1} \sum \sum_{i \neq j} y_i y_j \right] \\
 &= \dots = \left(1 - \frac{n}{N}\right) \frac{S_{\text{pop}}^2}{n}
 \end{aligned}$$

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

The Finite Population Correction (FPC)

- We have seen that for SRS without replacement

$$E[\bar{Y}_{samp}] = \bar{y}_{pop} \quad (\bar{Y}_{samp} \text{ is unbiased})$$

$$Var(\bar{Y}_{samp}) = (1-f)S_{pop}^2/n, \quad f = n/N$$

- The quantity $(1-f)$ is called the *finite population correction (fpc)*.
 - When $n/N \approx 0$, $(1-f) \approx 1$, so “With or without replacement doesn’t matter for small SRS’s!”
 - As $n/N \rightarrow 1$, $(1-f) \rightarrow 0$ and $SE(\bar{y}_{samp}) \rightarrow 0$. “We don’t need statistical estimates for a true census!”

FPC, continued

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

$$Var(\bar{Y}_{samp}) \approx (1-f)s^2/n,$$

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y}_{samp})^2$$

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

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \hat{p})^2 = \frac{n}{n-1} \hat{p}(1-\hat{p})$$

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 w/o 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 with replacement were too large.

Comparing SE’s

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