

**UCLA STAT 13**  
**Introduction to Statistical Methods for the  
 Life and Health Sciences**

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[http://www.stat.ucla.edu/~dinov/courses\\_students.html](http://www.stat.ucla.edu/~dinov/courses_students.html)

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**Chapter 5**  
**Sampling Distributions**

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**Sampling Distributions**

- **Definition:** *Sampling Variability* is the variability among random samples from the same population.
- A probability distribution that characterizes some aspect of sampling variability is called a sampling distribution.
  - tells us how close the resemblance between the sample and the population is likely to be.
- We typically construct a sampling distribution for a statistic.
  - Every statistics has a sampling distribution.

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**The Meta-Experiment**

- All the possible samples that might be drawn from the population (infinity repetitions).
  - In other words if we were to repeatedly take samples of the same size from the same population, over and over.

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**The Meta-Experiment**

- Meta-experiments are important because probability can be interpreted as the long run relative frequency of the occurrence of an event.
- Meta-experiments also let us visualize sampling distributions.
  - and therefore understand the variability among the many random samples of a meta-experiment.

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**Dichotomous Observations**

- Dichotomous - two outcomes
  - (yes or no, good or evil, etc...)
- We use the following notation for a dichotomous outcome
 

$P$	population proportion
$\hat{p}$	sample proportion
- The big question is how close is  $\hat{p}$  to  $P$ ?
- To determine this we need to examine the sampling distribution of  $\hat{p}$
- What we want to know is:
  - if we took many samples of size  $n$  and observed  $\hat{p}$  each time, how would those values of  $\hat{p}$  be distributed around  $p$ ?

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## Dichotomous Observations

**Example:** Suppose we would like to estimate the true proportion of male students at UCLA. We could take a random sample of 50 students and calculate the sample proportion of males.

- What is the correct notation for:
  - the true proportion of males?
  - the sample proportion of males?
- Suppose we repeat the experiment over and over. Would we get the same proportion of males for the second sample?

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## Reece's Pieces Experiment

**Example:** Suppose we would like to estimate the true proportion of orange reece's pieces in a bag. To investigate we will take a random sample of 10 reece's pieces and count the number of orange. Next we will make an approximation to a sampling distribution with our class results.

What you need to calculate:

- the number of orange
- the sample proportion of orange (number of orange/10)

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## An Application of a Sampling Distribution

**Example:** Mendel's pea experiment. Suppose a tall offspring is the event of interest and that the true proportion of tall peas (based on a 3:1 phenotypic ratio) is  $3/4$  or  $p = 0.75$ . If we were to randomly select samples with  $n = 10$  and  $p = 0.75$  we could create a probability distribution as follows:

	$\hat{p}$	Number Tall	Number Dwarf	Probability
	0.0	0	10	0.000
	0.1	1	9	0.000
	0.2	2	8	0.000
	0.3	3	7	0.003
	0.4	4	6	0.016
	0.5	5	5	0.058
	0.6	6	4	0.146
	0.7	7	3	0.250
	0.8	8	2	0.282
	0.9	9	1	0.188
	1.0	10	0	0.056

Lab\_Mendel\_Pea\_Experiment.html  
(work out in discussion/lab)

Validate using:  
[http://www.stat.ucla.edu/Applets/Normal\\_T\\_Ch2\\_F\\_Table.htm](http://www.stat.ucla.edu/Applets/Normal_T_Ch2_F_Table.htm)  
E.g.,  $B(n=10, p=0.75, a=6, b=6)=0.146$

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## An Application of a Sampling Distribution

- What is the probability that 5 are tall and 5 are dwarf?

$$\begin{aligned} P(5 \text{ tall and } 5 \text{ dwarf}) &= P(\hat{p} = 5/10) \\ &= P(\hat{p} = 0.5) \\ &= 0.058 \end{aligned}$$

	$\hat{p}$	Number Tall	Number Dwarf	Probability
	0.0	0	10	0.000
	0.1	1	9	0.000
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## An Application of a Sampling Distribution

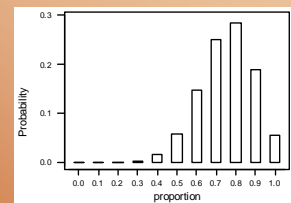
- If we think about this in terms of a meta-experiment and we sample 10 offspring over and over, about 5.8% of the  $\hat{p}$ 's will be 0.5.
  - This is the sampling distribution of sample proportion of tall offspring is the distribution of in repeated samples of size 10.
- If we take a random sample of size 10, what is the probability that six or more offspring are tall?

$$\begin{aligned} P(\hat{p} \geq 0.6) &= 0.146 + 0.250 + 0.282 + 0.188 + 0.056 \\ &= 0.922 \end{aligned}$$

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## An Application of a Sampling Distribution

- This table could also be represented as a histogram with probability on the y-axis and proportion on the x-axis.
  - easier to draw these by hand



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### Relationship to Statistical Inference

- We can also use our sampling distribution of  $\hat{p}$  to estimate how much sampling error there is within 5 percentage points of  $p$ . Because we knew  $p$  from the previous example ( $p=0.75$ ), we might want to estimate:

$$P(0.7 \leq \hat{p} \leq 0.8) = 0.250 + 0.282 = 0.532$$

There is a 53% chance that for a sample of size 10,  $\hat{p}$  will be within  $\pm 0.05$  of  $p$ .

- This seems a little crazy, why?

$\hat{p}$	Number Tall	Number Dwarf	Probability
0.0	0	10	0.000
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0.2	2	8	0.000
0.3	3	7	0.003
0.4	4	6	0.016
0.5	5	5	0.058
0.6	6	4	0.146
0.7	7	3	0.250
0.8	8	2	0.282
0.9	9	1	0.188
1.0	10	0	0.056

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### Relationship to Statistical Inference

- So far we have been using  $p$  to determine the sampling distribution of  $\hat{p}$ .
- Why sample for  $\hat{p}$  when we already know  $p$ ?
  - We don't need to know  $p$  to get a good estimate (this will come later).

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### Sample Size

- As  $n$  gets larger,  $\hat{p}$  will become a better estimate of  $p$ .
- Just to show...

N	$P(0.7 \leq \hat{p} \leq 0.8)$
10	0.53
20	0.56
50	0.673
100	0.798

\*These calculations were done using the SOCR binomial distribution Calculator.

[http://socr-stat.ucla.edu/Applets.dir/Normal\\_T\\_Chi2\\_F\\_Tables.htm](http://socr-stat.ucla.edu/Applets.dir/Normal_T_Chi2_F_Tables.htm)

E.g.,  $B(n=20, p=0.75, a=0.7 \times 20=14, b=0.8 \times 20=16)=0.5606$

**THE POINT:** A larger sample improves the chance that  $\hat{p}$  is close to  $p$ .

- Caution: this doesn't necessarily mean that the estimate will be closer to  $p$ , only that there is a better chance that it will be close to  $p$ .

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