Lesson 8: Chapter 5 Section 2

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BSC Mathematics

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And we wanted to convert it to a probability distribution describing the random variable of the number of brown-eyed people in a sample of five people. Our random variable turned out to be as below.

This was a **binomial random variable**, and in this section we'll discuss the binomial random variable and how it relates to **population proportions**.

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$$\hat{p} = \frac{17}{50} = 0.34.$$

Notice, this is close to the population proportion p = 0.35, but some variation might be expected due to sampling.



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Sample proportions and sample counts (which are simply the number of successes observed in a sample) are frequently used statistics, so we should be interested in understanding their sampling distributions. To that end, let's talk about the **binomial random variable/binomial distribution**.

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Example

Does our previous example of counting the number of brown-eyed people in a sample of 5 people qualify as a binomial setting?



Definition (binomial distribution)

The distribution of the count X of successes in a binomial setting is called the **binomial distribution** with parameters n and p. The parameter n is the number of observations/trails while p is the probability of success in any single observation/trail. The possible values of X are the whole numbers 0 to n. We often denote this distribution of X as B(n, p).

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There is a formula that describes how to assign probabilities to each of these values of X. We'll see this later.

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The accuracy of this approximation improves as the size of the population increases relative to the size of the sample. As a rule, we will use the binomial sampling distribution for counts when the population is at least 20 times as large as the sample. You can relax this rule if the sampling is done with replacement.

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So, now that we know what a binomial distribution is and why it's useful, we need to figure out how to calculate binomial probabilities.

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 $\binom{n}{k} p^k q^{n-k}$.

Here, $\binom{n}{k}$ (read n choose k) simply denotes $\frac{n!}{(n-k)!k!}$. If you want to calculate $P(X \le k)$, you must calculate

$$\sum_{i=0}^{k} \binom{n}{i} p^{i} q^{n-i}.$$

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Example (binomial probability)

Calculate P(X < 2) for B(10, 0.25).

$$\sum_{i=0}^{1} {10 \choose i} (0.25)^{i} (0.75)^{10-i} =$$

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$${(1)(1)(0.75)^{10} + (10)(0.25)(0.75)^{9} = 0.244}.$$

TI-83/84, Minitab, and tables can also be used to compute these probabilities! Feel free to use these in addition to the formula.

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$$1 - \sum_{i=0}^{6} {12 \choose i} (0.25)^{i} (0.75)^{12-i} = 0.0143.$$

So, it would be very unusual for her to have missed 7 or more free throws, which means that 7 is an unusually high number of missed three throws (out of 12).

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Theorem (mean, variance, and standard deviation of B(n, p))

For a binomial $X \sim B(n, p)$ probability distribution, we have

$$\mu_X = np$$
 and $\sigma_X^2 = npq$,

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You can use this to figure out if a value is unusually high/low as well! Let's revisit our previous example.



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Well, $\mu_X = 12(0.25) = 3$ and $\sigma_X = \sqrt{12(0.25)(0.75)} = 1.5$. Thus, our range of usual values is [3 - 2(1.5), 3 + 2(1.5)] = [0, 6]. Notice, 7 is usually high in this setting as well!

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Well, $\mu_X=12(0.25)=3$ and $\sigma_X=\sqrt{12(0.25)(0.75)}=1.5$. Thus, our range of usual values is [3-2(1.5),3+2(1.5)]=[0,6]. Notice, 7 is usually high in this setting as well! This works because the binomial distribution is roughly symmetric, bell-shaped, and nearly normal!

Now that we've talked about sampling distributions for population counts, lets discuss sampling distributions of population proportions! These distributions are not binomial like population counts, though you can often restate questions involving proportions as questions involving counts. Often, however, we simply approximate these sampling distributions with a normal distribution!

Theorem (normal approximation for counts and proportions)

A SRS of size n from a large population, having proportion p successes, has sampling distributions for counts X and proportions $\hat{p} = \frac{X}{n}$ given by

$$N\left(np,\sqrt{np(1-p)}\right)$$
 and $N\left(p,\sqrt{\frac{p(1-p)}{n}}\right)$,

respectively. As a rule, we will only use this approximation for values of n and p that satisfy $np \ge 10$ and $n(1-p) \ge 10$.

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respectively. As a rule, we will only use this approximation for values of n and p that satisfy $np \ge 10$ and $n(1-p) \ge 10$.

We use the rule above because the binomial distribution is not approximately normal unless it has at least 10 successes and failures.

Example (bad apples)

An orchard's harvest typically has a proportion of bad apples of about 4%. An applesauce company will only accept a shipment of apples if the shipment contains less than 4.1% bad apples. It orders a shipment of 2000 apples. What's the chance that this shipment will be accepted?

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$$P(\hat{p} < 0.041) \approx P(Z < 0.23) = 0.5910.$$

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We must calculate P(X < 82), where X has approximately a normal distribution N(80, 8.76356).

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- $P(X < K) \approx P(X \le K 0.5)$

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The real probability P(X < 82) is 0.5746.



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Chapter 5 discusses the Poisson distribution - which is essentially the distribution which arises from the binomial setting when we assume that the number of trails in unlimited. We won't discuss this distribution in this course. Don't expect to see anything about it on the test.