

1 Working with the Law of Large Numbers

- A fair coin is tossed and you win a prize if there are more than 60% heads. Which is better: 10 tosses or 100 tosses? Explain.
- A fair coin is tossed and you win a prize if there are more than 40% heads. Which is better: 10 tosses or 100 tosses? Explain.
- A coin is tossed and you win a prize if there are between 40% and 60% heads. Which is better: 10 tosses or 100 tosses? Explain.
- A coin is tossed and you win a prize if there are exactly 50% heads. Which is better: 10 tosses or 100 tosses? Explain.

Solution:

- 10 tosses. By LLN, the sample mean should have higher probability to be close to the population mean as n increases. Therefore the average proportion of coins that are heads should be closer to 0.50, and has a lower chance of being greater than 0.60 if there are 100 tosses compared with 10 tosses.
- 100 tosses. Based on the first part, consider the inverse of the event “more than 60% heads” and the symmetry of heads and tails.
- 100 tosses. Based on the first part, consider the union of the events “more than 60% heads” and “more than 60% tails” (“less than 40% heads”).
- 10 tosses. Compare the probability of getting equal number of heads and tails between $2n$ and $2n + 2$ tosses.

$$\begin{aligned}
 \mathbb{P}[n \text{ heads in } 2n \text{ tosses}] &= \binom{2n}{n} \frac{1}{2^{2n}} \\
 \mathbb{P}[n+1 \text{ heads in } 2n+2 \text{ tosses}] &= \binom{2n+2}{n+1} \frac{1}{2^{2n+2}} = \frac{(2n+2)!}{(n+1)!(n+1)!} \cdot \frac{1}{2^{2n+2}} \\
 &= \frac{(2n+2)(2n+1)2n!}{(n+1)(n+1)n!n!} \cdot \frac{1}{2^{2n+2}} \\
 &= \frac{2n+2}{n+1} \cdot \frac{2n+1}{n+1} \binom{2n}{n} \cdot \frac{1}{2^{2n+2}} < \left(\frac{2n+2}{n+1}\right)^2 \binom{2n}{n} \cdot \frac{1}{2^{2n+2}} \\
 &= 4 \binom{2n}{n} \cdot \frac{1}{2^{2n+2}} = \binom{2n}{n} \frac{1}{2^{2n}} = \mathbb{P}[n \text{ heads in } 2n \text{ tosses}]
 \end{aligned}$$

The larger n is, the less probability we'll get 50% heads. □

Note: By Stirling's approximation, $\binom{2n}{n}2^{-2n}$ is roughly $(\pi n)^{-1/2}$ for large n .

2 Markov's Inequality and Chebyshev's Inequality

A random variable X has variance $\text{var}(X) = 9$ and expectation $\mathbb{E}[X] = 2$. Furthermore, the value of X is never greater than 10. Given this information, provide either a proof or a counterexample for the following statements.

- (a) $\mathbb{E}[X^2] = 13$.
- (b) $\mathbb{P}[X \leq 1] \leq 8/9$.
- (c) $\mathbb{P}[X \geq 6] \leq 9/16$.
- (d) $\mathbb{P}[X \geq 6] \leq 9/32$.

Solution:

(a) TRUE. Since $9 = \text{var}(X) = \mathbb{E}[X^2] - \mathbb{E}[X]^2 = \mathbb{E}[X^2] - 2^2$, we have $\mathbb{E}[X^2] = 9 + 4 = 13$.

(b) TRUE. Let $Y = 10 - X$. Since X is never exceeds 10, Y is a non-negative random variable. By Markov's inequality,

$$\mathbb{P}[10 - X \geq a] = \mathbb{P}[Y \geq a] \leq \frac{\mathbb{E}[Y]}{a} = \frac{\mathbb{E}[10 - X]}{a} = \frac{8}{a}.$$

Setting $a = 9$, we get $\mathbb{P}[X \leq 1] = \mathbb{P}[10 - X \geq 9] \leq 8/9$.

(c) TRUE. Chebyshev's inequality says $\mathbb{P}[|X - \mathbb{E}[X]| \geq a] \leq \text{var}(X)/a^2$. If we set $a = 4$, we have

$$\mathbb{P}[|X - 2| \geq 4] \leq \frac{9}{16}.$$

Now we observe that $\mathbb{P}[X \geq 6] \leq \mathbb{P}[|X - 2| \geq 4]$, because the event $X \geq 6$ is a subset of the event $|X - 2| \geq 4$.

(d) FALSE. Construct a random variable X that satisfies the conditions in the question but does not have an equal chance of being less than -2 or greater than 6. A simple example would be a random variable that takes on 2 values, where $\mathbb{P}[X = a] = p, \mathbb{P}[X = b] = 1 - p$. The expectation must be 2, so we have $pa + (1 - p)b = 2$. The variance is 9, so $\mathbb{E}[X^2] = 13$ and $pa^2 + (1 - p)b^2 = 13$. Solving for a and b . One example is $\mathbb{P}[X = 0] = 9/13, \mathbb{P}[X = 13/2] = 4/13$.

3 Vegas

On the planet Vegas, everyone carries a coin. Many people are honest and carry a fair coin (heads on one side and tails on the other), but a fraction p of them cheat and carry a trick coin with heads on both sides. You want to estimate p with the following experiment: you pick a random sample of n people and ask each one to flip his or her coin. Assume that each person is independently likely to carry a fair or a trick coin.

1. Given the results of your experiment, how should you estimate p ?
2. How many people do you need to ask to be 95% sure that your answer is off by at most 0.05?

Solution:

1. We want to construct an estimate \hat{p} such that $\mathbb{E}[\hat{p}] = p$. Then, if we have a large enough sample, we'd expect to get a good estimate of p . Let X_i be the indicator that the i th person's coin flips to a heads. What we observe is the fraction of people whose coin is heads. In other words, we measure $q = \frac{1}{n} \sum_{i=1}^n X_i$. How can we use this observation to construct \hat{p} ?

First,

$$\mathbb{E}[q] = \frac{1}{n} \sum_{i=1}^n \mathbb{E}[X_i] = \mathbb{E}[X_i] = p \cdot 1 + (1 - p) \cdot \frac{1}{2},$$

where the last equality follows from total probability. Solving for p , we find that

$$p = 2\mathbb{E}[q] - 1 = \mathbb{E}[2q - 1].$$

Thus, our estimator \hat{p} should be $2q - 1$.

2. We want to find n such that $P[|\hat{p} - p| \leq 0.05] > 0.95$. Another way to state this is that we want

$$P[|\hat{p} - p| > 0.05] \leq 0.05.$$

Notice that $\mathbb{E}[\hat{p}] = p$ by construction, so we can immediately apply Chebyshev's inequality on \hat{p} . What we get is:

$$P[|\hat{p} - p| > 0.05] \leq \frac{\text{var}[\hat{p}]}{0.05^2} \leq 0.05$$

So, we want n such that $\text{var}[\hat{p}] \leq 0.05^3$.

$$\text{var}[\hat{p}] = \text{var}[2q - 1] = 4 \text{var}[q] = \frac{4}{n^2} \text{var}\left[\sum_{i=1}^n X_i\right] = \frac{4}{n} \text{var}[X_1].$$

But X_i is an indicator (Bernoulli variable), so its variance is bounded by $\frac{1}{4}$. Therefore we have

$$\text{var}[\hat{p}] \leq \frac{4}{n} \frac{1}{4} = \frac{1}{n}.$$

So, we choose n such that $\frac{1}{n} \leq 0.05^3$, so $n \geq \frac{1}{0.05^3} = 8000$.