

DIS 14A

1 Working with the Law of Large Numbers

- (a) 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.
- (b) 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.
- (c) 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.
- (d) 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:

- (a) 100 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.
- (b) 100 tosses. Based on the first part, consider the inverse of the event “more than 60% heads” and the symmetry of heads and tails.
- (c) 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”).
- (d) 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 Uniform Probability Space

Let $\Omega = \{1, 2, 3, 4, 5, 6\}$ be a uniform probability space. Let also $X(\omega)$ and $Y(\omega)$, for $\omega \in \Omega$, be the random variables defined in the table:

Table 1: All the rows in the table correspond to random variables.

ω	1	2	3	4	5	6	$\mathbb{E}[\cdot]$
$X(\omega)$	0	0	1	1	2	2	
$Y(\omega)$	0	2	3	5	2	0	
$X^2(\omega)$							
$Y^2(\omega)$							
$XY(\omega)$							
$L[Y X](\omega)$							

- Fill in the blank entries of the table. In the column to the far right, fill in the expected value of the random variable.
- Are the variables correlated or uncorrelated? Are the variables independent or dependent?

(c) Calculate $\mathbb{E}[(Y - L[Y | X])^2]$.

Solution:

(a) See the following table:

ω	1	2	3	4	5	6	$\mathbb{E}[\cdot]$
$X(\omega)$	0	0	1	1	2	2	1
$Y(\omega)$	0	2	3	5	2	0	2
$X^2(\omega)$	0	0	1	1	4	4	5/3
$Y^2(\omega)$	0	4	9	25	4	0	7
$XY(\omega)$	0	0	3	5	4	0	2
$L[Y X](\omega)$	2	2	2	2	2	2	2

The third, fourth, and fifth rows can be calculated directly from the corresponding X and Y values. Recall that

$$L[Y | X] = \frac{\text{cov}(X, Y)}{\text{var}(X)}(X - \mathbb{E}(X)) + \mathbb{E}(Y).$$

But $\text{cov}(X, Y) = \mathbb{E}(XY) - \mathbb{E}(X)\mathbb{E}(Y) = 2 - (1)(2) = 0$, so $L[Y | X] = \mathbb{E}(Y) = 2$ for all ω .

(b) Since $\text{cov}(X, Y) = 0$, the variables are uncorrelated. But, we see that $\mathbb{P}(Y = 0) = 1/3$ and $\mathbb{P}(Y = 0 | X = 3) = 0$, so the two variables are not independent. Recall that independence implies uncorrelation, but the converse is not true.

(c)

$$\mathbb{E}[(Y - L[Y | X])^2] = \mathbb{E}[(Y - 2)^2] = \frac{4 + 0 + 1 + 9 + 0 + 4}{6} = 3$$

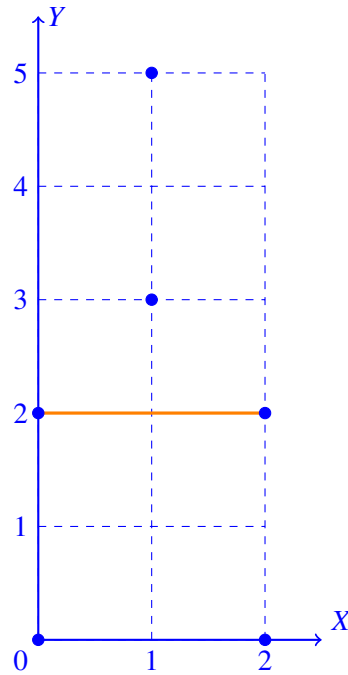


Figure 1: Visualization of regression. The circles are the (X, Y) points. The orange line is the LLSE.

3 LLSE

We have two bags of balls. The fractions of red balls and blue balls in bag A are $2/3$ and $1/3$ respectively. The fractions of red balls and blue balls in bag B are $1/2$ and $1/2$ respectively. Someone gives you one of the bags (unmarked) uniformly at random. You then draw 6 balls from that same bag with replacement. Let X_i be the indicator random variable that ball i is red. Now, let us define $X = \sum_{1 \leq i \leq 3} X_i$ and $Y = \sum_{4 \leq i \leq 6} X_i$. Find $L(Y | X)$. *Hint:* Recall that

$$L(Y | X) = \mathbb{E}(Y) + \frac{\text{cov}(X, Y)}{\text{var}(X)} (X - \mathbb{E}(X)).$$

Also remember that covariance is bilinear.

Solution:

Note that although the indicator random variables are not independent, we can still apply linearity of expectation. By symmetry, we also know that each indicator follows the same distribution.

Therefore:

$$\begin{aligned}\mathbb{E}[X] &= \mathbb{E}[Y] = 3 \cdot \mathbb{E}(X_1) = 3 \cdot \mathbb{P}(X_1 = 1) = 3 \cdot \left(\frac{1}{2} \cdot \frac{2}{3} + \frac{1}{2} \cdot \frac{1}{2} \right) = \frac{7}{4}. \\ \text{cov}(X, Y) &= \text{cov}\left(\sum_{1 \leq i \leq 3} X_i, \sum_{4 \leq j \leq 6} X_j \right) = 9 \cdot \text{cov}(X_1, X_4) \\ &= 9 \cdot (\mathbb{E}(X_1 X_4) - \mathbb{E}(X_1) \cdot \mathbb{E}(X_4)). \\ \mathbb{E}(X_1 X_4) - \mathbb{E}(X_1) \mathbb{E}(X_4) &= \mathbb{P}(X_1 = 1, X_4 = 1) - \mathbb{P}(X_1 = 1)^2 \\ &= \left[\frac{1}{2} \cdot \left(\frac{2}{3} \right)^2 + \frac{1}{2} \cdot \left(\frac{1}{2} \right)^2 \right] - \left[\frac{1}{2} \cdot \left(\frac{2}{3} \right) + \frac{1}{2} \cdot \left(\frac{1}{2} \right) \right]^2 = \frac{1}{144}. \\ \text{var}(X) &= \text{cov}\left(\sum_{1 \leq i \leq 3} X_i, \sum_{1 \leq j \leq 3} X_j \right) \\ &= 3 \cdot \text{var}(X_1) + 6 \cdot \text{cov}(X_1, X_2) = 3(\mathbb{E}(X_1^2) - \mathbb{E}(X_1)^2) + 6 \cdot \frac{1}{144} \\ &= 3 \left[\frac{7}{12} - \left(\frac{7}{12} \right)^2 \right] + 6 \cdot \frac{1}{144} = \frac{111}{144}.\end{aligned}$$

So,

$$L(Y | X) = \frac{7}{4} + \frac{9}{111} \left(X - \frac{7}{4} \right) = \frac{3}{37} X + \frac{119}{74}.$$