Module 9.4: Indicators MCIT Online - CIT592 - Professor Val Tannen

LECTURE NOTES



Indicator random variables

Let A be an event in a probability space (Ω, Pr) . The **indicator** random variable of the event A, notation I_A , is defined by

$$I_A(w) = \begin{cases} 1 & \text{if } w \in A \\ 0 & \text{if } w \notin A \end{cases}$$

Note that I_A is a Bernoulli random variable with success probability $\Pr[I_A = 1] = \Pr[A]$.

As we have shown before, its expectation is $E[I_A] = Pr[I_A = 1] = Pr[A]$.

Number of heads in *n* coin flips I

Problem. We flip a biased coin n times with heads probability p. Let H be the r.v. that returns the number of heads observed. Compute E[H].

Answer (first attempt). We will try to use the formula for expectation.

The outcomes are 2^n sequences of length n of H's and T's. An outcome with k H's has probability p^kq^{n-k} where q=1-p.

There are $\binom{n}{k}$ outcomes with k H's. Therefore the probability of the event "k heads observed" is $\binom{n}{k}p^kq^{n-k}$.

Using the formula for expectation:

$$E[H] = \sum_{k=0}^{n} k \cdot Pr[H = k] = \sum_{k=0}^{n} k {n \choose k} p^{k} q^{n-k}$$



Number of heads in *n* coin flips II

Answer (second attempt). We are going to use **linearity of expectation**.

In the probability space of the n flips that we just saw, let H_k be the event "the k'th flip is H" for $k=1,\ldots,n$ and let I_k be the **indicator** random variable of the event H_k .

Clearly, $H = I_1 + \cdots + I_n$.

By linearity of expectation $E[H] = E[I_1] + \cdots + E[I_n]$.

We established earlier that $E[I_k] = Pr[H_k]$.

Since the flips are independent $Pr[H_k] = p$.

In conclusion $E[H] = p + \cdots + p = n \cdot p$.

Interestingly,

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

Balls to bins "on average"

Problem. We throw k balls into n bins. What is the number of balls that end up in Bin 1, "on average"?

Answer. We want E[X] where X is the random variable that returns the number of balls that end up in Bin 1.

We can express X as $X = I_1 + \cdots + I_k$.

where I_i is the **indicator** r.v. of the event L_i = "ball i ends up in Bin 1".

Recall from the discussion that we introduced the model that $Pr[L_i] = 1/n$.

Therefore $E[I_i] = Pr[L_i] = 1/n$.

Using **linearity of expectation** we obtain E[X] = (1/n) + ... + (1/n) = k/n.

ACTIVITY: Balls in bins and biased coins

The balls into bins problem we just saw can be seen as a particular case of the biased coin one that precedes it.

Indeed for each ball throw consider only two outcomes: (1) falls in Bin 1, and (2) does not fall in Bin 1. Since each of the n bins is equally likely to receive the ball, outcome (1) has probability 1/n (while outcome (2) has probability (n-1)/n).

Therefore each ball throw is like flipping biased coin with 1/n probability of showing heads.

We can then apply the calculation of the expected number of heads in multiple flips of biased coin. Here we have k throws so the expected number of balls in Bin 1 will be (k)(1/n) = k/n, the same answer.