

CS 174 Homework Assignment 3 (due Wednesday, Feb. 20)

1. Choose N coins from an infinite collection of coins, where N is a Poisson random variable with parameter λ . Each coin has a probability p of coming up heads (i.e., the same value for all coins). Toss each of the N coins once. Show that the total number of heads is a Poisson random variable with parameter λp .
2. Let X and Y be independent Poisson random variables with parameters λ and μ respectively. Let $Z = X + Y$. What is the conditional probability distribution $P(X|Z = n)$?
3. Let X be a Poisson random variable with parameter λ . Compute the mode of the Poisson distribution (the point of maximal probability). When is the mean equal to the mode? Is it possible to have $P(X = 1) > P(X = 0)$?
4. Let X_1, X_2, \dots, X_n be independent random variables, and let X_i be Bernoulli with parameter p_i . Show that $Y = \sum_{i=1}^n X_i$ has mean and variance given by:

$$E[Y] = \sum_{i=1}^n p_i \quad \text{Var}[Y] = \sum_{i=1}^n p_i(1 - p_i).$$

Show that, for $E[Y]$ fixed, $\text{Var}[Y]$ is a maximum when $p_1 = p_2 = \dots = p_n$. That is to say, the variation in the sum is greatest when individuals are most alike. Is this contrary to intuition?

5. Consider a random graph $G = (V, E)$ on n vertices, where edges are chosen independently with probability $P(n)$. Derive an upper bound on the probability that such a random graph G has no triangles.
6. This problem will help you to understand the inclusion-exclusion principle (page 1 of Note 3). Consider once again the balls-and-bins experiment, in which we toss m labeled balls into n labeled bins. We are going to use inclusion-exclusion to figure out precisely the probability that there are no empty bins.
 - (a) Let E_i be the event that bin i is empty, and for any given set of k bins i_1, i_2, \dots, i_k , let $p_{i_1 i_2 \dots i_k}$ denote the probability that all these bins are empty, i.e., $p_{i_1 i_2 \dots i_k} = \Pr[E_{i_1} \wedge E_{i_2} \wedge \dots \wedge E_{i_k}]$. Show that $p_{i_1 i_2 \dots i_k} = (1 - \frac{k}{n})^m$.
 - (b) As in Note 3, define $S_k = \sum_{i_1, i_2, \dots, i_k} p_{i_1 i_2 \dots i_k}$. Show that $S_k = \binom{n}{k} (1 - \frac{k}{n})^m$.
 - (c) Now let q denote the probability that there are no empty bins. Deduce from Theorem 1 of Note 3 that

$$q = 1 - \binom{n}{1} \left(1 - \frac{1}{n}\right)^m + \binom{n}{2} \left(1 - \frac{2}{n}\right)^m - \binom{n}{3} \left(1 - \frac{3}{n}\right)^m + \dots \pm \binom{n}{n-1} \left(1 - \frac{n-1}{n}\right)^m.$$

- (d) Now let us consider the special case $m = n \ln n + cn$. (Recall from our discussion of the coupon collecting problem in Note 1 that this is around the threshold required for no empty bins.) Show that, in this case, for any fixed value of k ,

$$S_k = \binom{n}{k} \left(1 - \frac{k}{n}\right)^m \sim \frac{1}{k!} e^{-kc}.$$

Hint: If k is fixed, then it should be clear that $\binom{n}{k} \sim \frac{n^k}{k!}$. And you already know how to handle things like $\left(1 - \frac{k}{n}\right)^m$.

- (e) Deduce that when $m = n \ln n + cn$, for any constant c ,

$$\Pr[\text{there are no empty bins}] \sim e^{-e^{-c}} \quad \text{as } n \rightarrow \infty.$$

- (f) Recall that we proved on page 3 of Note 1 that $E[X] \sim e^{-c}$, where the r.v. X denotes the number of empty bins. Applying Markov's inequality immediately gives us the bound $\Pr[X \geq 1] \leq e^{-c}$, and hence

$$\Pr[\text{there are no empty bins}] \geq 1 - e^{-c} \quad \text{as } n \rightarrow \infty.$$

The exact asymptotic value we have just proved in part (e) (with quite a bit more work!) is stronger than this bound, especially when c is small. To see this, take $c = 0$ (i.e., we are throwing exactly $n \ln n$ balls) and show that the above bound from Note 1 is useless while the exact value computed in part (e) gives a significant probability of finding no empty bins. What happens for large values of c ? How about negative values?