Problem 1. Two hockey teams, A and B play a series of games, until one of the teams wins 4 games. Suppose team A has probability p of winning each game, and games are independent. Let X be the total number of games that are played.

- (a) Find the probability mass function of X.
- (b) What is the probability that team A wins the series conditioned on X = 4?
- (c) What is the probability that team A wins the series conditioned on X = 7? (Simplify your expressions as much as possible.)

Solution.

(a) Denote q = 1 - p to simplify some formulas.

$$p(4) = p^{4} + q^{4} \qquad p(6) = 10p^{4}q^{2} + 10p^{2}q^{4}$$

$$p(5) = 4p^{4}q + 4pq^{4} \qquad p(7) = 20p^{4}q^{3} + 20p^{3}q^{4} = 20p^{3}q^{3}.$$

In each case the first term is the probability that A wins and the second that B wins in that many games. For example, to find p(6), in order for team A to win in 6 games, they must lead 3-2 after 5 games $\binom{5}{2} = 10$ ways for this to happen) and then win the last game. Each such sequence has probability p^4q^2 .

- (b) This is $\frac{p^4}{p^4+q^4}$.
- (c) This is $\frac{20p^4q^3}{20p^3q^3} = p$. We can also see this directly, since if the series lasts 7 games, then after 6 games the score is 3-3, and the last game determines the winner.

Problem 2. A fair (6-sided) die is rolled four times.

- (a) Let Y denote the number of distinct results. Find the probability mass function and expectation of Y.
- (b) Let Z denote the minimal result out of the 4 throws. Find the probability mass function and expectation of Z.

Solution.

(a) We have

$$p(1) = \frac{6}{6^4} \qquad p(2) = \frac{14 \cdot \binom{6}{2}}{6^4} \qquad p(3) = \frac{\binom{6}{3} \cdot 3 \cdot 12}{6^4} \qquad p(4) = \frac{6 \cdot 5 \cdot 4 \cdot 3}{6^4}.$$

The cases 1 and 4 are easier. For k = 2, there are $\binom{6}{2}$ ways to choose which two values appear, and there are 14 sequences where the two specific values appear and no others $(2^4$ where no others appear, minus the two where only one of them appears). (The sequences are aaab,aaba,aabb,....) Similarly, if 3 values appear, there are $\binom{6}{2}$ ways to pick which three values appear, 3 ways to specify which of them appears twice, and 12 ways to arrange the throws with given outcomes $\{a, a, b, c\}$. (b) We have $p(k) = \frac{(7-k)^4 - (6-k)^4}{6^4}$ for k = 1, 2, ..., 6. This is since there are $(7-k)^4$ sequences where the

minimum is at least k, and we subtract $(6-k)^4$ where the minimum is at least k+1.

Problem 3. This problem investigates the similarity between the geometric and exponential random variables observed last week. Let Y be a geometric random variable with parameter p, so that Y represents the trial number of the first success in a sequence of independent Bernoulli trials. Suppose the trials occur at times $\delta, 2\delta, \ldots$, and that δ and p are both very small. Let $\lambda = p/\delta$. At time t, about t/δ trials have taken place.

(a) Compute P(Y > m), which represents the probability that no success has been observed by time $t = m\delta$.

- (b) Show that the probability that no success has been observed by time t converges to $e^{-\lambda t}$ as $p, \delta \to 0$ with $\lambda = p/\delta$ fixed.
- (c) Conclude that the time of the first success is approximately an exponential random variable with parameter λ .

Solution.

- (a) For integer m, $P(Y > m) = (1 p)^m$, since this is the probability of the first m experiments failing.
- (b) by time t, the number of attempts is $[t/\delta]$, (here [x] is the largest integer less than x). Therefore $P(Y > t) = (1-p)^{[t/\delta]} = e^{[t/\delta]\log(1-p)}$. As $p \to 0$ we have $\log(1-p) = -p + O(p^2)$, and therefore

$$\log P(Y > t) = -p[t/\delta] + O(p^2[t/\delta]).$$

The first term tends to $-\lambda t$ and the second to 0. Therefore $P(Y > t) \to e^{-\lambda t}$ as $p, \delta \to 0$.

(c) Since the CDF of Y is 1 - P(Y > t) and tends to $1 - e^{-\lambda t}$, which is the CDF of the exponential, Y is approximately exponential. We will discuss convergence of distributions in detail later this term.

Problem 4. A binary message either 0 or 1 is transmitted by wire. However, data sent over the wire is subject to channel noise disturbance. If x is the value sent (either 0 or 1), then the value received at the other end is R = x + N, where N represents the noise. Assume that N is a normal random variable with mean $\mu = 0$ and variance $\sigma^2 = 0.04$. Assume that a message sent is equally likely to be 0 or 1. When the message is received the receiver decodes it according to the following rule: If $R \leq \frac{1}{2}$ she concludes the message is 0, and otherwise concludes it is 1. What is the probability that the message is received correctly?

Solution. If the message is 0 we need N < 1/2. If the message is 1 we need N > -0.5. The probability is the same in both cases, and is F(0.5), where F is the CDF Φ of the $N(0, 0.2^2)$ variable. Since $N(0, \sigma^2) = \sigma N(0, 1)$, we can write $F(x) = \Phi(x/0.2)$, where Φ is the standard normal CDF. The answer is therefore $\Phi(0.5/0.2) = \Phi(2.5) = 0.993...$ Note: $\Phi(x) = \frac{1+erf(x/\sqrt{2})}{2}$. This can be calculated in python as follows.

>>> import scipy.stats
>>> scipy.stats.norm.cdf(2.5)
0.99379033467422384
>>> from math import erf
>>> (1 + erf(2.5/sqrt(2)))/2
0.9937903346742238

Problem 5. The number of murders in Gotham on any given week is assumed to be Poisson with unknown mean λ , with different weeks independent.

- (a) We observe there were 2 murders one week. What is the value of λ for which this is most likely?
- (b) We observe for a second week, and there is 1 murder. What is the value of λ for which this pair of observations is most likely?
- (c) Generalize the above to observations a_1, \ldots, a_k over k weeks.

Solution.

- (a) The likelihood of observing 2 murders in a week is $e^{-\lambda}\frac{\lambda^2}{2}$. This tends to 0 at $0, \infty$, and is maximized when $\lambda = 2$.
- (b) The likelihood of the two observations is

$$L(\lambda) = e^{-\lambda} \frac{\lambda^2}{2} \cdot e^{-\lambda} \frac{\lambda}{1} = e^{-2\lambda} \frac{\lambda^3}{2}.$$

This is maximized at $\lambda = \frac{3}{2}$.

(c) If the observations are a_1, \ldots, a_k , then the likelyhood is

$$L(\lambda) = \prod_{i} e^{-\lambda} \frac{\lambda^{a_i}}{a_i!} = e^{-k\lambda} \frac{\lambda^S}{\prod_{i} a_i!},$$

where $S = \sum a_i$. This is maximized at $\lambda = \frac{1}{k} \sum a_i$, which is the average observation. Note that this is consistent with the idea of λ being the average rate of murders.

Problem 6. An airline books passengers for a flight on an airplane with 420 seats. From experience, the airline knows that each passenger has probability $p = \frac{1}{50}$ of missing the flight. Assume these events are independent. As such, the airline takes a risk and sells 430 tickets for the flight.

- (a) Using python, compute the probability that more than 420 passenger show up.
- (b) Use the Poisson approximation to compute an approximation to this probability.
- (c) Simulate the number of no-shows for an overbooked flight 50000 times. (You can use a function that returns a binomial random variable.) Plot a histogram of the fraction of times there were k no-shows, and the Poisson p.m.f. on the same graph.
- (d) Simulated the number of no-shows 50000 times, and define X_n = number of overfull flights in the first n simulated bookings. Then X_n/n is the running proportion of overbooked flights. Plot X_n/n . What happens to it as n gets large?

Note: When calculating X_n , do not simulate n new bookings for every n. Simulate 50000 flights, and then for every n, calculate X_n based on the first n of these.