# UC Berkeley Department of Electrical Engineering and Computer Sciences

## EE126: PROBABILITY AND RANDOM PROCESSES

## Discussion Session 2

Fall 2018

#### Problem 1. Linearity of Expectation, Independence

Let  $X_1, X_2, X_3$  be discrete independent random variables with mean 0. Find  $\mathbb{E}[(X_1 + X_2)(X_2 + X_3)(X_3 + X_1)]$ .

#### Problem 2. Poisson Packet Routing

Packets arriving at a switch are routed to either destination A (with probability p) or destination B (with probability 1-p). The destination of each packet is chosen independently of each other. In the time interval [0,1], the number of arriving packets is  $Pois(\lambda)$ .

- 1. Show that the number of packets routed to A is Poisson distributed. With what parameter?
- 2. Are the number of packets routed to A and to B independent?

### Problem 3. Poisson Merging

The Poisson distribution is used to model *rare events*, such as the number of customers who enter a store in the next hour. The theoretical justification for this modeling assumption is that the limit of the binomial distribution, as the number of trials n goes to  $\infty$  and the probability of success per trial p goes to 0, such that  $np \to \lambda > 0$ , is the Poisson distribution with mean  $\lambda$ .

Now, suppose we have two independent streams of rare events (for instance, the number of female customers and male customers entering a store), and we do not care to distinguish between the two types of rare events. Can the combined stream of events be modeled as a Poisson distribution?

Mathematically, let X and Y be independent Poisson random variables with means  $\lambda$  and  $\mu$  respectively. Prove that  $X + Y \sim \operatorname{Pois}(\lambda + \mu)$ . (This is known as **Poisson merging**.) Note that it is **not** sufficient to use linearity of expectation to say that X + Y has mean  $\lambda + \mu$ . You are asked to prove that the *distribution* of X + Y is Poisson.

*Note*: This property will be extensively used when we discuss Poisson processes.

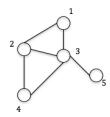


Figure 1: Friendship and clustering coefficient.

## Problem 4. [Extra] Clustering Coefficient

This problem will explore an important probabilistic concept of clustering that is widely used in machine learning applications today. Consider n students, where n is a positive integer. For each pair of students  $i, j \in \{1, ..., n\}$ ,  $i \neq j$ , they are friends with probability p, independently of other pairs. We assume that friendship is mutual. We can see that the friendship among the n students can be represented by an undirected graph G. Let N(i) be the number of friends of student i and T(i) be the number of triangles attached to student i. We define the **clustering coefficient** C(i) for student i as follows:

$$C(i) = \frac{T(i)}{\binom{N(i)}{2}}.$$

The clustering coefficient is not defined for the students who have no friends. An example is shown in Figure 1. Student 3 has 4 friends (1, 2, 4, 5) and there are two triangles attached to student 3, i.e., triangle 1-2-3 and triangle 2-3-4. Therefore  $C(3) = 2/\binom{4}{2} = 1/3$ .

Find  $\mathbb{E}[C(i) \mid N(i) \geq 2]$ .