## 1. (5 pt.) Sampling Without Replacement

Suppose there are *n* total balls, of which *m* are red. We sample *k* of the balls uniformly without replacement.<sup>1</sup> Let *Z* be the random variable denoting how many of the *k* balls are red. In this problem, you will show that *Z* is concentrated around its mean.

- (a) (1 pt.) Show that  $\mathbb{E}[Z] = \frac{km}{n}$ .
- (b) (4 pt.) When  $k \ge 1$ , show that  $\Pr[|Z \mathbb{E}[Z]| \ge \lambda] \le 2e^{-\lambda^2/(2k)}$  for any  $\lambda > 0$ .

**[HINT:** Try applying the Azuma-Hoeffding tail bound to a Doob martingale. When applying Azuma-Hoeffding to a martingale  $\{Z_t\}$ , feel free to provide a short/intuitive explanation for why  $|Z_i - Z_{i-1}| \leq c_i$  rather than a rigorous proof. ]

- (c) **[Optional: this won't be graded.]** When k is close to n, a tighter bound than that from part (b) holds.
  - i. (0 pt.) When k = n, explain why  $\Pr[Z = \mathbb{E}[Z]] = 1$ .
  - ii. (0 pt.) When  $1 \le k \le n-1$ , show that  $\Pr[|Z \mathbb{E}[Z]| \ge \lambda] \le 2e^{-\lambda^2/(2v)}$  where v is defined as

$$v \coloneqq \sum_{i=1}^{k} \left( 1 - \frac{k-i}{n-i} \right)^2.$$

iii. (0 pt.) Show that  $v \leq O(k(n-k)/n)$ . This shows that the bound from part (c), ii is tighter than the bound from part (b) when k is close to n.

## SOLUTION:

(a) Let  $X_i \in \{0,1\}$  be the indicator variable for whether the  $i^{\text{th}}$  ball is red. Since  $Z = \sum_{i=1}^{k} X_i$ , we can apply linearity of expectation to conclude

$$\mathbb{E}[Z] = \sum_{i=1}^{k} \mathbb{E}[X_i] = k \cdot \frac{m}{n}$$

(b) For each t = 0, ..., k, we define the Doob martingale,

$$Z_t = \mathbb{E}[Z \mid X_1, \dots X_t].$$

This is a martingale with respect to  $\{X_t\}$ . Then,  $Z_0 = \mathbb{E}[Z]$  with probability 1 and  $Z_k = Z$ . We aim to apply the Azuma-Hoeffding equality to bound  $|Z_k - Z_0|$ . Intuitively, revealing whether a single ball is red can only affect the number of final red balls by 1, so  $|Z_t - Z_{t-1}| \leq 1$  for all  $t = 1, \ldots, k$ . For a more careful derivation that accounts for the correlations between whether each ball is red see the solution to part (c), ii. With the bound  $|Z_t - Z_{t-1}| \leq 1$  the desired result follows from the Azuma-Hoeffding

With the bound  $|Z_t - Z_{t-1}| \leq 1$ , the desired result follows from the Azuma-Hoeffding inequality.

<sup>&</sup>lt;sup>1</sup>Note that this only makes sense when  $k, m \leq n$ .

- (c) i. When k = n, we sample all of the *n* balls. Since this is done without replacement, we are guaranteed to select all *m* red balls.
  - ii. For each  $t = 0, \ldots, k$ , we define the Doob martingale,

$$Z_t = \mathbb{E}[Z \mid X_1, \dots X_t].$$

This is a martingale with respect to  $\{X_t\}$ . Then,  $Z_0 = \mathbb{E}[Z]$  with probability 1 and  $Z_k = Z$ . We aim to apply the Azuma-Hoeffding inequality to bound  $|Z_k - Z_0|$ , which requires upper bounding  $|Z_t - Z_{t-1}|$  for each  $t = 1, \ldots, k$ .

Fix a realization for  $X_1, \ldots, X_t$ , and let R be the number of red balls already selected  $(R = \sum_{j=1}^{t} X_j)$ . There are two possible realizations for  $X_{t+1}$ : If  $X_{t+1} = 1$ , then we will have selected a total of R + 1 balls in the first t + 1 steps and there will be k - R - 1 remaining red balls for steps  $t + 2, \ldots, k$ . Applying part (a) to bound the expectation of  $X_{t+2} + \cdots + X_k$ ,

$$Z_{t+1} = \mathbb{E}[Z \mid X_1, \dots, X_{t+1}]$$
  
=  $(R+1) + \mathbb{E}\left[\sum_{j=t+2}^k X_j \mid X_1, \dots, X_{t+1}\right]$   
=  $R+1 + \frac{(k-t-1)(m-R-1)}{n-t-1}.$ 

The other possible realization is that  $X_{t+1} = 0$ . In this case, there will be k - r remaining red balls beginning at time step t+2. Using similar logic to the first case,

$$Z_{t+1} = \mathbb{E}[Z \mid X_1, \dots, X_{t+1}]$$
  
=  $R + \mathbb{E}\left[\sum_{j=t+2}^k X_j \mid X_1, \dots, X_{t+1}\right]$   
=  $R + \frac{(k-t-1)(m-R)}{n-t-1}.$ 

Since  $\{Z_t\}$  is a martingale with respect to  $\{X_t\}$ , it must be the case that  $\mathbb{E}[Z_{t+1} | X_0, \ldots, X_t] = Z_t$ , which means that  $Z_t$  is between the minimum and maximum possible values for  $Z_{t+1}$  conditioned on  $X_0, \ldots, X_t$ . Therefore,

$$\begin{aligned} |Z_{t+1} - Z_t| &\leq \left(R + 1 + \frac{(k - t - 1)(m - R - 1)}{n - t - 1}\right) - \left(R + \frac{(k - t - 1)(m - R)}{n - t - 1}\right) \\ &= 1 - \frac{k - t - 1}{n - t - 1}. \end{aligned}$$

The desired result follows from the Azuma-Hoeffding bound.

iii. We rewrite v as

$$v = \sum_{i=1}^{k} \left(1 - \frac{k-i}{n-i}\right)^2$$
$$= \sum_{i=1}^{k} \left(\frac{n-k}{n-i}\right)^2$$
$$= (n-k)^2 \cdot \sum_{i=1}^{k} \left(\frac{1}{n-i}\right)^2$$
$$= (n-k)^2 \cdot \sum_{j=n-k}^{n-1} \frac{1}{j^2}$$

For any integers  $a \leq b$ ,

$$\sum_{j=a}^{b} \frac{1}{j^2} \le \int_{a-1}^{b} \frac{1}{x^2} dx$$
$$= \frac{1}{a-1} - \frac{1}{b}.$$

We apply this to our bound for v,

$$v = (n-k)^2 \cdot \sum_{j=n-k}^{n-1} \frac{1}{j^2}$$
  

$$\leq (n-k)^2 \cdot \left(\frac{1}{(n-k)^2} + \sum_{j=n-k+1}^n \frac{1}{j^2}\right)$$
  

$$\leq (n-k)^2 \cdot \left(\frac{1}{(n-k)^2} + \frac{1}{n-k} - \frac{1}{n}\right)$$
  

$$= 1 + (n-k)^2 \cdot \left(\frac{k}{(n-k)n}\right)$$
  

$$= 1 + \frac{k(n-k)}{n} = O\left(\frac{k(n-k)}{n}\right).$$

## 2. (11 pt.) Reaching Consensus

This question considers a simple and fairly natural model of the dynamic of how opinions shift over time in a group.

Suppose there is an undirected graph G = (V, E) whose vertices represent the group members and a pair of members are friends if and only if they are connected by an edge. For simplicity, we assume that G contains none of the following: 1) self-loops, 2) multiple edges connecting the same pair of vertices, or 3) isolated vertices, i.e., vertices with no edge on them. Let S be the set of possible "opinions" on some topic, and lets suppose that each person has one and only one opinion on the topic at a time. (For concreteness, think of  $S = \{A, B, C, ...\}$ .) We can represent the opinions of the group members by a mapping  $\sigma : V \to S$  where the group member corresponding to vertex v has opinion  $\sigma(v)$ .

The opinions  $\sigma$  of the group members evolve due to discussions between friends. We model the evolution of  $\sigma$  by the following time-homogeneous Markov chain: starting from the initial opinion  $\sigma_0$ ,  $\sigma$  changes from  $\sigma_{t-1}$  to  $\sigma_t$  at step t as follows. Independently for every vertex v, we flip a fair coin. If the outcome is "heads",  $\sigma_t(v)$  remains the same as  $\sigma_{t-1}(v)$ ; otherwise,  $\sigma_t(v)$  becomes  $\sigma_{t-1}(v')$  for a uniformly random neighbor v' of v. In short, every group member keeps their own opinion with probability 1/2, and takes one of their friends' opinion with the remaining 1/2 probability.

In this problem, we will determine the likelihood that the group members reach a certain consensus, given their initial opinions.

- (a) (1 pt.) If G is disconnected and |S| > 1, show that there exist initial opinions  $\sigma_0$  of the members for which consensus is never reached.
- (b) (3 pt.) If G is connected, show that consensus is eventually reached almost surely. That is, show that as the number of steps goes to infinity, the probability that consensus has been reached approaches 1.
- (c) (2 pt.) Let  $X_t$  be the number of group members who have some opinion, say  $A \in S$  after step t. Give an example where  $(X_t)_{t\geq 0}$  is not a martingale with respect to  $(\sigma_t)_{t\geq 0}$ . The example should be one specific tuple  $(G, S, \sigma_0)$ .
- (d) (3 pt.) Let  $Y_t$  be the sum of the degrees of the vertices v corresponding to the group members with opinion A after step t. Prove that  $(Y_t)_{t\geq 0}$  is a martingale with respect to  $(\sigma_t)_{t\geq 0}$ .
- (e) (2 pt.) Assume that G is connected. What is the probability that all members of the group end up with opinion A (ie after some time, everyone has opinion,  $A \in S$ , for the rest of time)? Express your answer in terms of G and the initial opinion  $\sigma_0$  of the group members.

[HINT: Try applying the martingale stopping theorem to the martingale  $(Y_t)_{t\geq 0}$ .]

## SOLUTION:

- (a) Let A, B denote two different opinions in S. If the group members in one connected component of G all agree on A initially, and the group members in another component all agree on B initially, there is no way to reach overall consensus.
- (b) Suppose G is connected, and suppose that some vertex has opinion A. Then, for each vertex that has opinion A or has a neighbor with opinion A, there is at least a 1/2n probability they will have opinion A next round. Therefore there is at least a  $(1/2n)^n$  chance that all vertices that have opinion A keep opinion A and all vertices that have a neighbor with opinion A adopts opinion A. If this happens n times in a row, then the entire graph will have opinion A, since the longest path between any two nodes is at most n. Therefore, with probability at least  $p = (1/2n)^{n^2} > 0$ , the entire graph will

have the same opinion after n timesteps. Hence the probability that consensus does not occur in kn timesteps is at most  $(1-p)^k$ , since we can split it up into k blocks of n timesteps, and we reach consensus in each block with probability p. This goes to 0 as  $k \to \infty$ , so consensus is reached almost surely.

- (c) Suppose the graph of opinions at time 0 is A–B–A. Then, each vertex switches opinion with probability exactly 1/2, so  $\mathbb{E}[X_1|\sigma_0] = 3/2$ , but  $X_0 = 2$ .
- (d) It is clear that  $Y_t$  is determined by  $\sigma_t$  and is finite, so only the third condition remains to be checked. Let  $d_v$  denote the degree of vertex v,  $a_v^t$  denote the number of neighbors of v with opinion A at time t, and  $A^t$  denote the set of vertices with opinion A. Then, we have

$$Y_t := \sum_{v \in A^t} d_v = \sum_v a_v^t.$$

The second inequality follows since the left side counts all outgoing edges from vertices with opinion A and the right side counts all incoming edges to vertices with opinion A. Then, since  $a_v^t$  is determined by  $\sigma_t$ , we have

$$\mathbb{E}[Y_{t+1}|\sigma_0, ..., \sigma_t] = \sum_{v} d_v P(v \in A^{t+1}|\sigma_t)$$
  
=  $\sum_{v \in A^t} d_v P(v \in A^{t+1}|\sigma_t) + \sum_{v \notin A^t} d_v P(v \in A^{t+1}|\sigma_t)$   
=  $\sum_{v \in A^t} d_v (\frac{1}{2} + \frac{a_v^t}{2d_v}) + \sum_{v \notin A^t} d_v \frac{a_v^t}{2d_v}$   
=  $\sum_{v \in A^t} \frac{d_v}{2} + \sum_v \frac{a_v^t}{2}$   
=  $Y_t.$ 

(e) Consider the stopping time  $T = \min\{t : Y_t = 0 \lor Y_t = \sum_v d_v\}$ . This is the first time either nobody has opinion A or everybody has opinion A. Note that once  $Y_t = 0$  or  $Y_t = \sum_v d_v$ , it will not change, and the event of reaching consensus on answer A is the same as the event  $Y_T = \sum_v d_v$ . From part (b), we know that  $T < \infty$  almost surely, and so  $Y_T = 0$  or  $Y_T = \sum_v d_v$  almost surely. So,

$$\mathbb{E}[Y_T] = P(Y_T = \sum_v d_v) \sum_v d_v.$$

By the stopping theorem,

$$\mathbb{E}[Y_T] = \mathbb{E}[Y_0] = \sum_{v \in A^0} d_v.$$

Therefore,

$$P(Y_T = \sum_v d_v) = \frac{\sum\limits_{v \in A^0} d_v}{\sum\limits_v d_v}$$