

**Problem Set 2**

CS265, Winter 2025

Due: 1/31 (Friday) at 11:59pm on Gradescope

Please follow the homework policies on the course website.

**1. (11 pt.) [Graph Coloring]**

The vertices of a simple graph (a graph with no self loops or multiple edges)  $G = (V, E)$  are each independently assigned one of three colors: red, green, or blue, chosen uniformly at random.

- (a) **(3 pt.)** Let  $n = |V|$  be the number of vertices in the graph. Show that the probability that more than half of the vertices are red is  $\exp(-\Omega(n))$ .
- (b) **(4 pt.)** Let  $m = |E|$  be the number of edges in the graph. Show that the probability that more than half of the edges are monochromatic (i.e., both endpoints have the same color) is  $O(1/m)$ .
- (c) **(4 pt.)** Suppose  $G$  is a complete graph (that is, all of the  $\binom{n}{2}$  possible edges are in the graph). Show that the probability that more than half of the edges are monochromatic is  $\exp(-\Omega(\sqrt{m}))$ .
- (d) **(0 pt.) [Optional: This won't be graded.]** Improve the bound from part (b).

**2. (12 pt.) [Tightness of Markov's and Chebyshev's Inequalities]**

- (a) **(4 pt.)** Show that Markov's inequality is tight. Specifically, for each value  $c > 1$ , describe a distribution  $D_c$  supported on non-negative real numbers such that if the random variable  $X$  is drawn according to  $D_c$  then
  - (1)  $\mathbb{E}[X] > 0$ , and
  - (2)  $\Pr[X \geq c\mathbb{E}[X]] = 1/c$ .
- (b) **(4 pt.)** Show that Chebyshev's inequality is tight. Specifically, for each value  $c > 1$ , describe a distribution  $D_c$  supported on real numbers such that if the random variable  $X$  is drawn according to  $D_c$  then
  - (1)  $\mathbb{E}[X] = 0$  and  $\text{Var}[X] = 1$ , and
  - (2)  $\Pr[|X - \mathbb{E}[X]| \geq c\sqrt{\text{Var}[X]}] = 1/c^2$ .
- (c) **(4 pt.) [One-sided version of Chebyshev's Inequality]** Consider the following statement: Let  $X$  be a random variable with  $\text{Var}[X] = 1$ . For all  $t \geq 0$ ,

$$\Pr[X - \mathbb{E}[X] \geq t] \leq \frac{1}{1+t^2}.$$

- i. Prove this statement.
- ii. Show that this is tight: For any  $t \geq 0$ , come up with a random variable  $X$  with distribution  $D_t$  and variance 1 for which  $\Pr[X - \mathbb{E}[X] \geq t] = \frac{1}{1+t^2}$ .

3. (11 pt.) [Concentration without Independence]

A computer system has  $n$  different failure modes, each of which happens with a small probability. Fortunately, the system is designed to be robust in the following sense: As long as less than half of the failure modes occur, things are fine; otherwise, a large-scale crash will happen. We want to make sure that the probability of crashing is small enough.

To model the above scenario, we define  $n$  Bernoulli random variables  $X_1, \dots, X_n$ . Each  $X_i$  is the indicator of the  $i$ -th failure mode, i.e.,  $X_i = 1$  if failure  $i$  occurs and  $X_i = 0$  otherwise. Our goal is to upper bound the probability  $\Pr[\sum_{i=1}^n X_i \geq n/2]$ .

- (a) (2 pt.) Let's first assume that the  $n$  failure events are independent and the probability of each failure is at most  $1/3$ . Formally, we have:

**Assumption 1.**  $\Pr[X_i = 1] \leq 1/3$  for every  $i \in [n]$  and  $X_1, \dots, X_n$  are independent.

Prove that under Assumption 1, for some constant  $C > 0$  that does not depend on  $n$ ,

$$\Pr\left[\sum_{i=1}^n X_i \geq n/2\right] \leq e^{-Cn}. \quad (1)$$

Thus, the probability of a crash is exponentially small in  $n$ .

[**HINT:** Feel free to use (without proof) any of the Chernoff bounds in lecture note #5 (including Theorem 2 and Corollaries 5 and 6) and also the inequality  $\frac{e^\delta}{(1+\delta)^{1+\delta}} \leq e^{-\delta^2/3}$  for  $\delta \in [0, 1]$ . ]

- (b) (1 pt.) Now we decide that Assumption 1 is too unrealistic, since many of the failure modes are known to be strongly correlated. Show that only assuming  $\Pr[X_i = 1] \leq 1/3$  (and not the independence), the probability of crashing can be as large as  $\Omega(1)$ . In particular, prove that for any  $n \geq 2$ , there exist random variables  $X_1, \dots, X_n$  that satisfy: (1)  $\Pr[X_i = 1] \leq 1/3$  for every  $i \in [n]$ ; (2)  $\Pr[\sum_{i=1}^n X_i \geq n/2] \geq 1/3$ .
- (c) (2 pt.) Let's try the following relaxation of Assumption 1, which states that the probability for  $k$  different failures to occur simultaneously is exponentially small in  $k$ :

**Assumption 2.** For any  $S \subseteq [n]$ ,  $\Pr[X_i = 1 \text{ for all } i \in S] \leq (1/3)^{|S|}$ .

Show that Assumption 2 is strictly weaker than Assumption 1 by proving:

- i. Assumption 1 implies Assumption 2; and
- ii. the implication on the other direction does not hold, i.e., there exist some  $n$  and  $X_1, \dots, X_n$  that satisfy Assumption 2 but not Assumption 1.

[**HINT:** For (ii), there exists a counterexample for  $n = 2$ . ]

- (d) (6 pt.) Prove that under Assumption 2, inequality (1) holds for some constant  $C > 0$ . In your proof, you can appeal to the proof of the Chernoff bounds from lecture videos/notes if you need to write it out verbatim at some point. For example, if you manage to upper bound  $\Pr[\sum_{i=1}^n X_i \geq n/2]$  by an expression involving the moment-generating function of some random variable  $Y$  that is the sum of  $n$  independent Bernoulli random variables, you can simply say that "the rest of the proof is exactly the proof of Theorem 2 from Lecture #5".

[**HINT:** Consider independent Bernoulli random variables  $Y_1, \dots, Y_n$  with  $\Pr[Y_i = 1] = 1/3$  for each  $i \in [n]$ . For distinct indices  $i, j, \ell \in [n]$ , does  $\mathbb{E}[X_i X_j X_\ell] \leq \mathbb{E}[Y_i Y_j Y_\ell]$  hold? Can you extend your proof of the inequality to the case with repeating indices? ]

[**HINT:** Let  $X = \sum_{i=1}^n X_i$  and  $Y = \sum_{i=1}^n Y_i$ . What can we say about  $\mathbb{E}[X^k]$  and  $\mathbb{E}[Y^k]$  for integer  $k \geq 0$ ? Considering the identity  $e^z = \sum_{k=0}^{+\infty} \frac{z^k}{k!}$ , what can we say about  $\mathbb{E}[e^{tX}]$  and  $\mathbb{E}[e^{tY}]$  for any  $t > 0$ ? ]

- (e) **(0 pt.) [Optional: this won't be graded.]** Can you construct counterexamples for Part 3b that satisfy *pairwise independence* but have a crashing probability of  $\Omega(1/n)$ ? Formally, prove that there exists  $C > 0$  such that for any  $n \geq 2$ , there exist  $X_1, \dots, X_n$  that satisfy: (1)  $\Pr[X_i = 1] \leq 1/3$ ; (2)  $X_i$  and  $X_j$  are independent for distinct  $i, j \in [n]$ ; (3)  $\Pr[\sum_{i=1}^n X_i \geq n/2] \geq C/n$ .

[**NOTE:** This shows that unlike Chebyshev's inequality, Chernoff bounds do not hold if we only assume pairwise independence. ]

[**HINT:** Recall pairwise independent hash functions if you have seen them before. You can use the Bertrand-Chebyshev theorem, which states that for any integer  $n \geq 1$ , there exists a prime number  $p$  with  $n < p < 2n$ . ]