

Instructions. Please work on this with your group and submit a neatly typed (or handwritten) file to Khoa: one submission per group. Use of other resources including books and the Web is not allowed. In case you can make partial progress on a problem, please write your approach clearly instead of giving no answer.

This HW. We primarily study tail bounds and sampling here.

0. **Required Reading.** Read <http://www.cs.cmu.edu/afs/cs/academic/class/15859-f04/www/scribes/lec8.pdf>. It gives random sampling as a powerful tool to even approximate #P-complete problems; in Section 8.3, it gives a “plus 1 / minus 1” trick very similar to Alon-Matias-Szegedy – and then discusses the limitations of this approach in the context there (approximating non-negative permanents) due to the high variance.

1. Prove that truncated inclusion-exclusion alternately lower- and upper- bounds $\Pr[E_1 \vee E_2 \vee \dots \vee E_n]$ where the E_i are arbitrary given events. (**Hint:** Use one of the two ways we discussed of viewing a joint probability distribution. The identity $\sum_{i=0}^r (-1)^i \binom{n}{i} = (-1)^r \binom{n-1}{r}$ (for integers $0 \leq r \leq n$, and with $\binom{n-1}{n}$ taken to be zero) could help.)

2. Consider a Poisson random variable X with mean λ . Use the exponential moment-generating function approach discussed for Chernoff (i.e., bounding $\mathbf{E}[e^{tX}]$ for an optimally-chosen t) to upper-bound $\Pr[X \geq \lambda(1 + \delta)]$ with $\delta > 0$. Do you get a familiar bound?

The next three problems study the following setup, where we have some random variables X_1, X_2, \dots, X_n whose sum X we want tail-bounds for, but where the X_i 's are not fully independent. We will assume that each X_i lies in $[0, 1]$ and that $\mathbf{E}[X] = \mu$. Importantly, the X_i 's have a dependency graph G – an undirected graph on the vertices $\{1, 2, \dots, n\}$ – where informally, an edge between i and j indicates that X_i and X_j are dependent, and the absence of such an edge means that X_i and X_j are independent. Formally, if S is any independent set in G , then the random variables $(X_i : i \in S)$ are mutually independent of each other. Let Δ denote the maximum degree of G .

3. Use the second-moment method to show that for any $a > 0$, $\Pr[|X - \mu| \geq a] \leq O(\mu\Delta/a^2)$.

4. Why stop at the second moment? In this problem and the next, we try for Chernoff-Hoeffding-type bounds, under the assumption that for some p , $\mathbf{E}[X_i] = p$ for all i . We will use the following useful result on graph coloring (lookup “graph coloring” if you need to): *the vertices of the graph G can be properly colored using $\Delta + 1$ colors in such a way that each color is used in total by either $\lfloor n/(\Delta + 1) \rfloor$ or $\lceil n/(\Delta + 1) \rceil$ vertices.*

Let $CH^+(\alpha, \beta)$ be the following Chernoff-Hoeffding upper-tail bound: suppose, for any m , any independent Y_1, Y_2, \dots, Y_m with $Y_i \in [0, 1]$, and $Y = \sum_i Y_i$ with $\mathbf{E}[Y] \geq \alpha$, we have $\Pr[Y \geq \alpha(1 + \beta)] \leq CH^+(\alpha, \beta)$ for $\beta > 0$.

Show that for any $\delta > 0$, $\Pr[X \geq \mu(1 + \delta)] \leq (\Delta + 1) \cdot CH^+(\lfloor n/(\Delta + 1) \rfloor \cdot p, \delta)$.

5. Again under the assumption that $\mathbf{E}[X_i] = p$ for all i , formulate and prove an analog of problem 4 for $\Pr[X \leq \mu(1 - \delta)]$, with, of course, $\delta > 0$ again.