

## EE5603: Concentration Inequalities, Fall 2022 (12)

Indian Institute of Technology Hyderabad  
HW 0, Assigned: Sunday 21.08.2022. 90 points.

**Due: Friday 26.08.2022 at 11:59 pm.**

1. Find the moment generation function of  $X \sim \mathcal{N}(0, \sigma^2)$ . (5)
2. Use the Markov inequality to prove the following (with appropriate assumptions):
  - (a) The Chebyshev inequality. (5)
  - (b) The Chernoff bound. (5)
3.
  - (a) Give an example where the Markov inequality is tight. (5)
  - (b) Give an example where the Chernoff bound is tighter than the Chebyshev inequality. (5)
4.
  - (a) Use the basic inequalities to prove the weak law of large numbers (LLN). (5)
  - (b) Illustrate the efficacy of the LLN with a Python program using at least four different types of distributions. (5)
5. Definition: A real valued random variable  $X$  is said to be  $\sigma^2$ -sub-Gaussian if there exists a positive number  $\sigma$  such that  $\mathbb{E}[e^{\lambda X}] \leq e^{\frac{\sigma^2 \lambda^2}{2}}$  for every  $\lambda \in \mathbb{R}$ . Show that the following random variables are  $\sigma^2$ -sub-Gaussian and find  $\sigma$  for each of them.
  - (a) A Rademacher random variable. (5)
  - (b) A Gaussian random variable with zero mean and variance  $\sigma^2$ . (5)
  - (c) A random variable that is zero mean and bounded in the interval  $[a, b]$ . (5)
6. Prove that for a  $\sigma^2$ -sub-Gaussian random variable with mean  $\mu$ ,  $P[|X - \mu| \geq t] \leq \exp(-\frac{t^2}{2\sigma^2})$  for all  $t > 0$ . (5)
7. If  $X_i$  are independent, mean-zero,  $\sigma_i^2$ -sub-Gaussian random variables, show that  $\sum_{i=1}^n X_i$  is  $\sum_{i=1}^n \sigma_i^2$ -sub-Gaussian. (5)
8. If  $X_i$  are bounded random variable (in  $[a_i, b_i]$ ), show that  $P(\frac{1}{n} \sum_{i=1}^n (X_i - \mathbb{E}[X_i]) \geq t) \leq \exp\left(-\frac{2n^2 t^2}{\sum_{i=1}^n (b_i - a_i)^2}\right)$  and  $P(\frac{1}{n} \sum_{i=1}^n (X_i - \mathbb{E}[X_i]) \leq -t) \leq \exp\left(-\frac{2n^2 t^2}{\sum_{i=1}^n (b_i - a_i)^2}\right)$  for all  $t > 0$ . (5)
9. Definition: A random variable  $X$  with mean  $\mu$  is sub-exponential if there are non-negative parameters  $(\nu, b)$  such that  $\mathbb{E}[e^{\lambda(X-\mu)}] \leq e^{\frac{\nu^2 \lambda^2}{2}}$  for all  $|\lambda| < \frac{1}{b}$ . Show that:
  - (a) The exponential random variable with parameter  $\lambda$  is sub-exponential. (5)
  - (b) The  $\chi^2$ -random variable is sub-exponential. (5)
10. Suppose that  $X$  is a  $(\nu, b)$ -sub-exponential random variable with mean  $\mu$ , derive the tail bound  $P[X \geq \mu + t]$  for all  $t > 0$ . (5)
11. If  $X_i$  are independent,  $(\nu, b)$ -sub-exponential random variables, then  $\sum_{i=1}^n X_i$  is  $(\sum_{i=1}^n \nu_i, b_*)$ -sub-exponential where  $b_* = \max_i b_i$ . (5)
12. Bernstein bound: If  $X$  is a random variable with mean  $\mu$  and variance  $\sigma^2$  and satisfies the condition  $|\mathbb{E}[(X - \mu)^k]| \leq \frac{1}{2} k! \sigma^2 b^{k-2}$ , it is said to satisfy the Bernstein condition with parameter  $b$ . For such a random variable show that  $\mathbb{E}[e^{\lambda(X-\mu)}] \leq e^{\frac{\lambda^2 \sigma^2 / 2}{1 - b|\lambda|}}$  for all  $|\lambda| < \frac{1}{b}$ . (5)