

Homework Two

EMALCA 2025: High-dimensional Statistics

Due in class on July 3 (Submit in groups of 2 or 3)

1. Let $A_j \in \mathbb{R}^{d \times d}$ be given deterministic matrices, for $j = 1 \dots, N$, and let $z \sim N(0, I_d)$. Set

$$U = \max_{j=1, \dots, N} \|A_j z\|_2.$$

- a) (**2 points**) For any fixed j show that $\|A_j z\|_2$ is sub-Gaussian and obtain a bound on its sub-Gaussian parameter (or norm).
- b) (**2 points**) Derive an upper bound on U that holds with high probability (for large N) and has logarithmic dependence on N .
- c) (**2 points**) Derive an upper bound on $\mathbb{E}(U)$.
- d) (**2 points**) Someone claims that U is sub-Gaussian. Prove or disprove

2. (**2 points**) **Exercise 2.15 from HDS.**

Exercise 2.15 (Concentration and kernel density estimation) Let $\{X_i\}_{i=1}^n$ be an i.i.d. sequence of random variables drawn from a density f on the real line. A standard estimate of f is the *kernel density estimate*

$$\widehat{f}_n(x) := \frac{1}{nh} \sum_{i=1}^n K\left(\frac{x - X_i}{h}\right),$$

where $K: \mathbb{R} \rightarrow [0, \infty)$ is a kernel function satisfying $\int_{-\infty}^{\infty} K(t) dt = 1$, and $h > 0$ is a bandwidth parameter. Suppose that we assess the quality of \widehat{f}_n using the L^1 -norm $\|\widehat{f}_n - f\|_1 := \int_{-\infty}^{\infty} |\widehat{f}_n(t) - f(t)| dt$. Prove that

$$\mathbb{P}[\|\widehat{f}_n - f\|_1 \geq \mathbb{E}[\|\widehat{f}_n - f\|_1] + \delta] \leq e^{-\frac{n\delta^2}{8}}.$$