

MA3K0 High-Dimensional Probability

Example Sheet 2

2020, term 2
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Students should hand in solutions to the questions by noon 12 pm Friday week 5 (07.02.2020) to the maths drop off box outside the undergraduate office.

Exercise 5: Khintchine inequality [30 points] Let X_1, \dots, X_N be independent sub-Gaussian random variables with zero mean and unit variance, and let $a \in \mathbb{R}^N$. Prove that for every $p \in [2, \infty)$ the following estimate holds

$$\|a\|_2 \leq \left\| \sum_{i=1}^N a_i X_i \right\|_{L^p} \leq CK \sqrt{p} \|a\|_2,$$

where $C > 0$ is an absolute constant and $K := \max_{1 \leq i \leq N} \{\|X_i\|_{\psi_2}\}$.

(Hint: Use the General Hoeffding inequality from the class: Let X_1, \dots, X_N be independent sub-Gaussian random variables with zero mean and unit variance, and let $a \in \mathbb{R}^N$. Then, for every $t \geq 0$, we have

$$\mathbb{P}\left(\left|\sum_{i=1}^N a_i X_i\right| \geq t\right) \leq 2 \exp\left(-\frac{ct^2}{K^2 \|a\|_2^2}\right),$$

where $K := \max_{1 \leq i \leq N} \{\|X_i\|_{\psi_2}\}$ and where $c > 0$ is an absolute constant.)

Exercise 6: Expectation of the Euclidean norm

Deduce from the 'Concentration of norm'-Theorem ¹ the following estimates for the expected Euclidean norm and the variance of a random vector $X = (X_1, \dots, X_n) \in \mathbb{R}^n$ with independent sub-Gaussian coordinates X_i that satisfy $\mathbb{E}[X_i^2] = 1, i = 1, \dots, n$.

(a) [10 points] **Expectation:**

$$\sqrt{n} - CK^2 \leq \mathbb{E}[\|X\|_2] \leq \sqrt{n} + CK^2,$$

where $K := \max_{1 \leq i \leq n} \{\|X_i\|_{\psi_2}\}$ and $C >$ an absolute constant.

(b) [10 points] **Variance:**

$$\text{var}(\|X\|_2) \leq \tilde{C}K^4,$$

where $K := \max_{1 \leq i \leq n} \{\|X_i\|_{\psi_2}\}$ and $\tilde{C} >$ an absolute constant.

¹Under the given assumptions the theorem states that

$$\left| \|X\|_2 - \sqrt{n} \right| \leq CK^2$$

with $C > 0$ an absolute constant and $K = \max_{1 \leq i \leq n} \{\|X_i\|_{\psi_2}\}$.

Exercise 7: Multivariate normal/Gaussian random variables Using the rotation invariance of the normal distribution deduce the following statements:

- (a) [10 points] Consider a n -dimensional Gaussian vector $Y \sim N(0, \mathbb{1}_n)$ and a fixed vector $u \in \mathbb{R}^n$ ($\mathbb{1}_n$ is the identity ($n \times n$)-matrix). Then

$$\langle Y, u \rangle \sim N(0, \|u\|_2^2),$$

where $\langle \cdot, \cdot \rangle$ is the Euclidean scalar product.

- (b) [15 points] Let X_1, \dots, X_n be independent Gaussian real-valued random variables $X_i \sim N(0, \sigma_i^2)$, Then

$$\sum_{i=1}^n X_i \sim N(0, \sigma^2) \quad \text{where } \sigma^2 = \sum_{i=1}^n \sigma_i^2.$$

(Hint: (b) is standard property of Gaussian random variables as discussed in the class, and for a proof use the convolution and the independence. It may be useful to do the proof for $n = 2$ first, and then argue with induction.)

Exercise 8: Normal and spherical distribution Write $Y \sim N(0, \mathbb{1}_n)$ as

$$Y = r\vec{Y},$$

where $r := \|Y\|_2$ is the length and $\vec{Y} = Y/\|Y\|_2$ is the direction of the n -dimensional random vector Y . Prove the following:

- (a) [15 points] The length r and the direction \vec{Y} are independent random variables.
 (b) [10 points] The direction \vec{Y} is uniformly distributed on the unit sphere S^{n-1} , where

$$S^{n-1} = \{x \in \mathbb{R}^n : \sum_{i=1}^n x_i^2 = 1\}.$$

hand-in times – assignment sheets:

Sheet	week	date	discussion in Support class
1	3	24.01.2020 at noon 12 pm	week 4
2	5	07.02.2020 at noon 12 pm	week 6
3	7	21.02.2020 at noon 12 pm	week 8
4	9	06.03.2020 at noon 12 pm	week 10