CMSE 820 HW 11

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1 Question 1

Let Ω be a non-empty set and \mathcal{A} be a collection of events. Define

$$\mathcal{I}(\mathcal{A}) = \{ \mathcal{F} | \mathcal{F} \text{ is a } \sigma\text{-algebra and } \mathcal{F} \supseteq \mathcal{A} \}.$$

and

$$\sigma(\mathcal{A}) = \cap_{\mathcal{F} \in \mathcal{I}(\mathcal{A})} \mathcal{F}.$$

Prove that $\sigma(A)$ is a σ -algebra. Note that $\sigma(A)$ is called the σ -algebra generated by A.

Proof: Let $a \subseteq \mathcal{A}$ be any subset of \mathcal{A} . By the definition of $\sigma(\mathcal{A})$,

$$\forall \mathcal{F} \in \mathcal{I}. \ a \in \mathcal{F}.$$

Following from the facts that \mathcal{F} is a σ -algebra of \mathcal{A} , we have, for any sequence of subsets of \mathcal{A} , $\{a_i\}_{i=1}^{\infty}$,

$$[\forall \mathcal{F} \in \mathcal{I}, (\mathcal{A} \in \mathcal{F} \implies \emptyset \in \mathcal{F})] \implies \emptyset \in \sigma(\mathcal{A}),$$

$$\forall \mathcal{F} \in \mathcal{I}, a_i^c \in \mathcal{F} \implies a_i^c \in \sigma(\mathcal{A}),$$

$$\forall \mathcal{F} \in \mathcal{I}, \bigcup_{i=1}^{\infty} a_i \in \mathcal{F} \implies \bigcup_{i=1}^{\infty} a_i \in \sigma(\mathcal{A}).$$

 $\sigma(\mathcal{A})$ contains \emptyset , and is closed under taking complement and countable union. Thus, $\sigma(\mathcal{A})$ is a σ -algebra of \mathcal{A} .

2 Question 2

Let $\Omega = \mathbb{R}$. Define

$$\mathcal{A} = \{(a, b) | a, b \in \mathbb{R}, a \le b\}.$$

Then $\sigma(\mathcal{A}) = \mathcal{B}(\mathbb{R})$ is the Borel σ -algebra on \mathbb{R} . Let

$$\mathcal{D} = \{(-\infty, a) : a \in \mathbb{R}\}.$$

Prove that $\sigma(\mathcal{D}) = \mathcal{B}(\mathbb{R})$.

Proof:

To show that $\sigma(\mathcal{D}) \supseteq \sigma(\mathcal{A})$, it suffices to show that for any $a, b \in \mathbb{R}$ such that $a \leq b$,

$$(a,b) \in \sigma(\mathcal{D}).$$

Note that σ -algebra is closed under countable intersection (which amounts to the complement of the countable union of the complements). So

$$\bigcap_{n=1}^{\infty} (-\infty, a + \frac{1}{n}) \in \sigma(\mathcal{D}).$$

In fact, we claim that $\bigcap_{n=1}^{\infty}(-\infty, a + \frac{1}{n}) = (-\infty, a]$, because

$$\forall x \in (-\infty, a], x \in (-\infty, a + \frac{1}{n}) \implies (-\infty, a] \subseteq \bigcap_{n=1}^{\infty} (-\infty, a + \frac{1}{n}),$$

and

$$\forall x > a, \, \exists \, n \text{ such that } x \not\in (-\infty, a + \frac{1}{n}) \implies x \not\in \cap_{n=1}^{\infty} (-\infty, a + \frac{1}{n}).$$

Now we can simply express (a, b) as

$$(a,b) = (a,\infty) \cap (-\infty,b) = \left(\bigcap_{n=1}^{\infty} \left(-\infty,a+\frac{1}{n}\right)\right)^{c} \cap (-\infty,b) \in \sigma(\mathcal{D}).$$

So $\sigma(\mathcal{A}) \subseteq \sigma(\mathcal{D})$.

Next, we will need to show that $\sigma(A) \supseteq \sigma(D)$. That is, for any $a \in \mathbb{R}$, we need to show that $(-\infty, a) \in \sigma(A)$.

Consider a sequence of elements $\{A_n\}_{n=1}^{\infty}$ in $\sigma(A)$ defined by $A_n = (a-n, a)$. It is easy to see that

$$\bigcup_{n=1}^{\infty} A_n = \bigcup_{n=1}^{\infty} (a - n, a) = (-\infty, a) \subseteq \sigma(\mathcal{A}).$$

Thus, $\sigma(\mathcal{A}) = \sigma(\mathcal{D}) = \mathcal{B}(\mathbb{R})$.

3 Question 3

Pick $X \in [-1,1]^d$ uniformly at random. Prove that ||X|| is tightly concentrated around $\sqrt{d/3}$. Namely,

$$\Pr\left(\left|\|X\|^2 - \frac{d}{3}\right| > \epsilon d\right) \to 0,$$

as $d \to \infty$.

Proof: First of all, we compute a few moments of a uniform random variable $x \in [-1, 1]$,

$$\begin{split} & \mathbf{E}[x] &= 0, \\ & \mathbf{E}[x^2] &= \frac{1}{3}, \\ & \mathbf{E}[x^4] &= \int_{-1}^{1} \frac{1}{2} x^4 dx = \frac{1}{5}. \end{split}$$

Clearly, $||X||^2$ should concentrated around $\frac{d}{3}$, because $\mathrm{E}[||X||^2] = \mathrm{E}[\sum_{i=1}^d x_i^2] = \sum_{i=1}^d \mathrm{E}[x^2] = \frac{d}{3}$. We apply the Chebyshev's inequality to $||X||^2$,

$$\Pr\left(\left|\|X\|^2 - \frac{d}{3}\right| > \epsilon d\right) \le \frac{\mathrm{E}[(\|X\|^2 - d/3)^2]}{(\epsilon d)^2}.$$

We want to show that $E[(||X||^2 - d/3)^2]$ is finite.

$$E[(||X||^2 - d/3)^2] = E\left[\left[\sum_{i=1}^d (x_i^2 - \frac{1}{3})\right]^2\right]$$

$$= \sum_{i=1}^d E[(x_i^2 - 1/3)^2] + \sum_{i=1,j>i}^d E[(x_i^2 - 1/3)(x_j^2 - 1/3)]$$

$$= \sum_{i=1}^d E[(x^2 - 1/3)^2] = \frac{4d}{45}.$$

So the probability goes to 0 as d approaches ∞ at a rate of $4/45\epsilon^2 d$.

4 Question 4

4.1 a

A random variable X with mean $E[X] = \mu$ is sub-Gaussian if there is a positive number σ such that, for all $\lambda \in \mathbb{R}$,

$$E[e^{\lambda(X-\mu)}] \le e^{\sigma^2 \lambda^2/2}.$$

If X is a sub-Gaussian random variable with σ as its sub-Gaussian parameter, prove that

$$\Pr(|X - \mu| \ge t) \le 2e^{-\frac{t^2}{2\sigma^2}}$$
, for all $t \in \mathbb{R}$.

Proof: For any $t \in \mathbb{R}$, we apply the Cramer's inequality,

$$\begin{split} \Pr(|X - \mu| \geq t) &= 2\Pr[(X - \mu) \geq t] \leq 2\inf_{\lambda > 0} e^{-\lambda t} \mathrm{E}[e^{\lambda(X - \mu)}] \\ &\leq 2\inf_{\lambda > 0} \exp\left[\frac{\sigma^2}{2}\lambda^2 - t\lambda\right] \\ &\leq 2\exp\left(-\frac{t^2}{2\sigma^2}\right). \end{split}$$

4.2 b

Suppose that the random variables X_i , i=1,...,n are independent, and X_i has mean μ_i and sub-Gaussian parameter σ_i . Prove that for all $t \geq 0$ we have

$$\Pr\left[\sum_{i=1}^{n} (X_i - \mu_i) \ge t\right] \le \exp\left\{-\frac{t^2}{2\sum_{i=1}^{n} \sigma_i^2}\right\}.$$

Proof: For any $t \in \mathbb{R}$, we again apply the Cramer's inequality,

$$\Pr\left[\sum_{i=1}^{n} (X_i - \mu_i) \ge t\right] \le \inf_{\lambda > 0} e^{-\lambda t} \mathbb{E}\left[e^{\lambda \sum_{i=1}^{n} (X_i - \mu_i)}\right]$$

$$= \inf_{\lambda > 0} \prod_{i=1}^{n} e^{-\lambda t/n} \mathbb{E}\left[e^{\lambda (X_i - \mu_i)}\right]$$

$$\le \inf_{\lambda > 0} \prod_{i=1}^{n} \exp\left[\frac{\sigma_i^2}{2} \lambda^2 - \frac{t}{n} \lambda\right]$$

$$= \inf_{\lambda > 0} \exp\left[\frac{\sum_{i=1}^{n} \sigma_i^2}{2} \lambda^2 - t \lambda\right]$$

$$= \exp\left(-\frac{t^2}{2 \sum_{i=1}^{n} \sigma_i^2}\right).$$