

Mathematical Foundations of Deep Learning (11.80020)

Assignment 0 (Voluntary exercises)

Reminder. For a real random variable X we call (if existent)

$$\varphi_X(\lambda) := \log \mathbb{E}[e^{\lambda X}] \quad \text{and} \quad \tilde{\varphi}_X(\lambda) := \log \mathbb{E}[e^{\lambda(X - \mathbb{E}X)}]$$

the *logarithmic moment generating function* or shortly *log-moment generating function* and the *centered logarithmic moment generating function* or shortly *centered log-moment generating function*, respectively. Note that $\varphi_X, \tilde{\varphi}_X: \mathbb{R} \rightarrow \mathbb{R}_{\geq 0} \cup \{+\infty\}$ and we denote their *domains* by

$$\text{dom}(\varphi_X) := \{\lambda \in \mathbb{R} : \varphi_X(\lambda) < \infty\} = \text{dom}(\tilde{\varphi}_X) := \{\lambda \in \mathbb{R} : \tilde{\varphi}_X(\lambda) < \infty\}.$$

We call

$$\varphi_X^*(\lambda) := \sup_{\lambda > 0} \{\lambda t - \varphi_X(\lambda)\} \quad \text{and} \quad \tilde{\varphi}_X^*(\lambda) := \sup_{\lambda > 0} \{\lambda t - \tilde{\varphi}_X(\lambda)\}$$

the *Cramer transform* of φ_X and $\tilde{\varphi}_X$, respectively. We call

$$\varphi_X^*(\lambda) := \sup_{\lambda \in \mathbb{R}} \{\lambda t - \varphi_X(\lambda)\} \quad \text{and} \quad \tilde{\varphi}_X^*(\lambda) := \sup_{\lambda \in \mathbb{R}} \{\lambda t - \tilde{\varphi}_X(\lambda)\}$$

the *Legendre transform* (or *Legendre-Fenchel transform* or *convex conjugate*) of φ_X and $\tilde{\varphi}_X$, respectively. We call a real random variable *sub-Gaussian* with parameter σ^2 (or shortly σ^2 -sub-Gaussian) if $\tilde{\varphi}_X(\lambda) \leq \frac{\sigma^2 \lambda^2}{2}$ for all $\lambda \in \mathbb{R}$.

Finally, recall *Hölder's inequality*. For this consider $p, q \in [1, \infty]$ such that $\frac{1}{p} + \frac{1}{q} = 1$, where $\frac{1}{\infty} := 0$. For two real random variables X and Y it holds that

$$\mathbb{E}[|XY|] \leq \mathbb{E}[|X|^p]^{\frac{1}{p}} \cdot \mathbb{E}[|Y|^q]^{\frac{1}{q}}.$$

Q1. (Logarithmic moment generating function of a Gaussian) Compute the centered and non centered log-moment generating function and their Legendre transforms of a Gaussian random variable $X \sim \mathcal{N}(\mu, \sigma^2)$ with mean μ and variance σ^2 .

Q2. (Stability of sub-Gaussianity) Let X_1 and X_2 be sub-Gaussian with parameters σ_1^2 and σ_2^2 , respectively.

- (a) Show that αX_1 is sub-Gaussian with parameter $\alpha^2 \sigma_1^2$ for a constant $\alpha \in \mathbb{R}$.
- (b) If X_1 and X_2 are independent, show that $X_1 + X_2$ is sub-Gaussian with parameter $\sigma_1^2 + \sigma_2^2$.
- (c) Show that in general (without assuming independence), the random variable $X_1 + X_2$ is sub-Gaussian with parameter $2(\sigma_1^2 + \sigma_2^2)$. Next, show that $X_1 + X_2$ is sub-Gaussian with parameter $(\sigma_1 + \sigma_2)^2$, which improves the result. *Hint:* Use Cauchy-Schwarz and Hölder's inequality respectively.

Q3. (Properties of logarithmic moment generating functions) For this exercise you can assume $\mathbb{E}X = 0$, but the statements hold in general.

- (a) *Convexity.* Show that φ_X and $\tilde{\varphi}_X$ are convex functions. *Hint:* Hölder's inequality
- (b) *Semi-continuity.* Show that φ_X and $\tilde{\varphi}_X$ lower semi-continuous, i.e., if $\lambda_n \rightarrow \lambda$ for $n \rightarrow \infty$ then

$$\liminf_{n \rightarrow \infty} \varphi_X(\lambda_n) \geq \varphi_X(\lambda) \quad \text{and} \quad \liminf_{n \rightarrow \infty} \tilde{\varphi}_X(\lambda_n) \geq \tilde{\varphi}_X(\lambda).$$

Hint: Fatou's lemma

- (c) *Smoothness.* Show that φ_X and $\tilde{\varphi}_X$ are smooth, i.e., infinitely many times continuously differentiable, function on their domains. *Hint:* It suffices to show that e^{φ_X} and $e^{\tilde{\varphi}_X}$ are smooth for which you can use the dominated convergence theorem.
- (d) *Derivatives.* Show that

$$\varphi'_X(\lambda) = \frac{\mathbb{E}[Xe^{\lambda X}]}{M(\lambda)} \quad \text{and} \quad \tilde{\varphi}'_X(\lambda) = \frac{\mathbb{E}[Xe^{\lambda(X-\mathbb{E}X)}]}{\tilde{M}(\lambda)},$$

where $M(\lambda) := \mathbb{E}[e^{\lambda X}] = e^{\varphi_X(\lambda)}$ and $\tilde{M}(\lambda) := \mathbb{E}[e^{\lambda(X-\mathbb{E}X)}] = e^{\tilde{\varphi}_X(\lambda)}$ denote the exponential and centered exponential moments of X .

- (e) *Existence of moments.* Assume that $\varphi_X(\lambda) < \infty$ or $\tilde{\varphi}_X(\lambda) < \infty$ for all $\lambda \in (-\varepsilon, \varepsilon)$ for some $\varepsilon > 0$. Show that all moments exist, i.e., $\mathbb{E}[|X|^k] < \infty$ for all $k \in \mathbb{N}$.
- (f) *Cramer transform equals Legendre transform.* Show that

$$\tilde{\varphi}_X^*(\lambda) = \sup_{\lambda > 0} \{\lambda t - \tilde{\varphi}_X(\lambda)\} = \sup_{\lambda \in \mathbb{R}} \{\lambda t - \tilde{\varphi}_X(\lambda)\} = \tilde{\varphi}_X^*(\lambda) \quad \text{for all } t \geq 0.$$