

Sub-Gaussian r.v.'s Continued

Note Title

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Recall: sub-Gaussian properties:

① Tails: $P(|X| > t) \leq \exp\left(-\frac{t^2}{k_1^2}\right)$ $t \geq 0$

② Moments: $(\mathbb{E}|X|^p)^{1/p} \leq k_2 \sqrt{p}$ $p \geq 1$

③ Tails': $\mathbb{E} \exp\left(\frac{X^2}{k_3}\right) \leq e$.

④ MGF: If $\mathbb{E}X = 0$, then
 $\mathbb{E} \exp(tX) \leq e^{t^2 k_4}$ $t \in \mathbb{R}$

How to compare "size" of sub-Gaussian r.v.'s?

Def The sub-Gaussian norm of a r.v. X is the smallest k_2 in property ②:

$$\|X\|_{\psi_2} := \sup p^{-\frac{1}{2}} (\mathbb{E}|X|^p)^{\frac{1}{p}}$$

Note: X is sub-Gaussian $\Leftrightarrow \|X\|_{\psi_2} < \infty$

$\|X\|_{\psi_2}$ plays similar role as standard deviation for mean-zero r.v.'s.

Exercise: Show that $\|\cdot\|_{\psi_2}$ is indeed a norm, defining the space of sub-Gaussian r.v.'s.

Corollary

Let X be a sub-Gaussian r.v. Then

$$\textcircled{1} P(|X| > t) \leq \exp\left(-\frac{ct^2}{\|X\|_{\psi_2}^2}\right), \quad t \geq 0$$

$$\textcircled{2} (\mathbb{E}|X|^p)^{\frac{1}{p}} \leq \|X\|_{\psi_2} \quad p \geq 1$$

$$\textcircled{3} \mathbb{E} \exp\left(c \frac{X^2}{\|X\|_{\psi_2}^2}\right) \leq e$$

$\textcircled{4}$ If $\mathbb{E}X = 0$, then

$$\mathbb{E} \exp(tX) \leq \exp(ct^2 \|X\|_{\psi_2}^2) \quad t \in \mathbb{R}$$

C, c : abs. const's.

$\|X\|_{\psi_2}$ is the smallest number possible in these #'s (up to abs. const's).

Examples of sub-Gaussian r.v.'s:

• Gaussian: $g \sim N(0, 1)$, $\|g\|_{\psi_2} = C$

More generally, $g \sim N(0, \sigma^2)$, $\|g\|_{\psi_2} = C\sigma$

• Rademacher:

$$|X|=1 \Rightarrow \mathbb{E}|X|^p = 1 \Rightarrow \|X\|_{\psi_2} = 1$$

• More generally, for any bounded r.v.,

$$\|X\|_{\psi_2} \leq \|X\|_{\infty} \quad (\text{Recall: } |X| \leq \|X\|_{\infty} \text{ a.s.})$$

Not Sub-Gaussian:

- Exponential r.v.'s: $X \sim \text{Exp}(\lambda)$

$$\Rightarrow P(|X| > t) = \exp(-\lambda t)$$

exponential tail does not decay fast enough!

Others: Geometric, Cauchy, Poisson...

"Size" of sum of r.v.'s.

Note: If $g_i \sim N(0, \sigma_i^2)$ then

$$\sum_{i=1}^n g_i \sim N(0, \sum_{i=1}^n \sigma_i^2)$$

Lemma (sum of sub-Gaussian r.v.'s)

Let X_1, \dots, X_n be ind. sub-Gaussian r.v.'s with $\mathbb{E} X_i = 0$. Then $\sum_{i=1}^n X_i$ is sub-Gaussian and

$$\left\| \sum_{i=1}^n X_i \right\|_{\psi_2}^2 \leq C \sum_{i=1}^n \|X_i\|_{\psi_2}^2 \quad (*)$$

Proof: Use MGF:

$$\mathbb{E} \exp\left(t \sum_{i=1}^n X_i\right) = \prod \mathbb{E} \exp(t X_i)$$

$$\leq \prod \exp\left(C t^2 \|X_i\|_{\psi_2}^2\right)$$

$$= \exp\left(t^2 C \underbrace{\sum_{i=1}^n \|X_i\|_{\psi_2}^2}_{\uparrow}\right)$$

K_4^2 from property
(4)

Note: K_4 is equivalent to $\|\cdot\|_{\psi_2}$,

$$\text{i.e. } \left\| \sum_{i=1}^n X_i \right\|_{\psi_2} \leq C K_4 = C \sqrt{\sum_{i=1}^n \|X_i\|_{\psi_2}^2}$$

QED

Thm: Hoeffding \neq for sub-Gauss. r.v.'s

Let X_i be mean-zero, sub-Gauss. r.v.'s. Then for all $\{a_1, \dots, a_n\} \in \mathbb{R}^n$,

$$P\left(\left| \sum_{i=1}^n a_i X_i \right| \geq t\right) \leq \exp\left(\frac{-t^2}{C \sum_{i=1}^n a_i^2 \|X_i\|_{\psi_2}^2}\right)$$

Proof: Exercise. (~2 lines)

Q: Agrees (roughly) with CLT?

Yes, if $\sqrt{\text{Var}(X_i)} \approx \|X_i\|_{\psi_2}$

Q: When is this not true?

Comparison of $\sqrt{\text{Var}(\sum_i X_i)}$, $\|\sum_i X_i\|_{\psi_2}$

given by:

Thm Khintchine \neq . Let X_1, \dots, X_n

be ind., sub-Gauss. r.v.'s, w/ $E X_i = 0$, $\text{Var}(X_i) = 1$.
Then $\forall \{a_1, \dots, a_n\} \in \mathbb{R}^n$, $p \geq 2$,

$$\sqrt{\sum_{i=1}^n a_i^2} \leq \left(E \left| \sum_{i=1}^n a_i X_i \right|^p \right)^{1/p} \leq C \sqrt{\sum_{i=1}^n a_i^2} \|X_i\|_{\psi_2} \sqrt{p}$$

Proof: Lower bound:

$$\left(E \left| \sum_{i=1}^n a_i X_i \right|^p \right)^{1/p} = \left(E \left| \sum_{i=1}^n a_i X_i \right|^2 \right)^{\frac{p}{2} \cdot \frac{1}{p}}$$

$$\leq \sqrt{E \left| \sum_{i=1}^n a_i X_i \right|^2} \quad (\text{by Jensen's } \neq)$$

$$= \sqrt{\text{Var}(\sum_i a_i X_i)}$$

$$= \sqrt{\sum_i a_i^2 \text{Var}(X_i)}$$

$$= \sqrt{\sum_{i=1}^n a_i^2} \quad \checkmark$$

Upper bound: Exercise.

Q: What if $\mathbb{E}X_i \neq 0$?

Remark: (Centering) We typically assume $\mathbb{E}X = 0$ (X_i is centered).

If not, we may replace X
w/
 $Y := X - \mathbb{E}X$.

Note: $\mathbb{E}Y = 0$

Further, if X is sub-Gaussian, then so is Y and

$$\|Y\|_{\psi_2} \leq 2\|X\|_{\psi_2}$$

$$\begin{aligned} \text{Proof: } \|Y\|_{\psi_2} &:= \|X - \mathbb{E}X\|_{\psi_2} \\ &\leq \|X\|_{\psi_2} + \|\mathbb{E}X\|_{\psi_2} \end{aligned}$$

$\mathbb{E}X$ is a const, so

$$\|\mathbb{E}X\|_{\psi_2} = |\mathbb{E}X| \leq \mathbb{E}|X| \leq \|X\|_{\psi_2}$$

↑
Jensen's

QED