

Sub-Gaussian $\mathbb{E} e^{-\alpha(W-\mu)} \leq e^{\alpha^2 \sigma^2/2}$

$$\rightarrow \mathbb{P}(W-\mu \geq t) \leq e^{-t^2/(2\sigma^2)}$$

Linear combinations of sub-Gaussian r.v.s are sub-Gaussian.

Prop 12 Let $(W_i)_{i=1}^n$ be a sequence of independent mean-zero sub-Gaussian r.v.s with respective parameters $(\sigma_i)_{i=1}^n$ and let $\gamma \in \mathbb{R}^n$. Then $\gamma^T W$ is sub-Gaussian with parameter $(\sum_{i=1}^n \sigma_i^2 \gamma_i^2)^{1/2}$.

Proof $\mathbb{E} \exp(\alpha \sum_{i=1}^n \gamma_i W_i) = \mathbb{E} \prod_{i=1}^n \exp(\alpha \gamma_i W_i)$
 $= \prod_{i=1}^n \mathbb{E} \exp(\alpha \gamma_i W_i)$
 $\leq \prod_{i=1}^n \exp(\alpha^2 \gamma_i^2 \sigma_i^2 / 2)$
 $= \exp(\alpha^2 \sum_{i=1}^n \gamma_i^2 \sigma_i^2 / 2) \quad \square$

General version of the probability bound required for theorem 9.

Lemma 13 Suppose $(\varepsilon_i)_{i=1}^n$ are independent mean-zero sub-Gaussian r.v.s with parameters $\sigma > 0$. (Note this includes $\varepsilon \sim N_n(0, \sigma^2 I)$). Let $\lambda = A \sigma \sqrt{\frac{\log p}{n}}$. Then $\mathbb{P}(\|X^T \varepsilon\|_\infty / n \leq \lambda) \geq 1 - 2p^{-(A^2/2)-1}$.

Proof $\mathbb{P}(\|X^T \varepsilon\|_\infty / n > \lambda) \leq \sum_{j=1}^p \mathbb{P}(|X_j^T \varepsilon| / n > \lambda) \quad (*)$

But $\pm X_j^T \varepsilon / n$ are sub-Gaussian with parameter $(\sigma^2 \sum_{i=1}^n (\frac{X_{ij}}{n})^2)^{1/2} = \sigma \sqrt{n}$.

Thus by prop 10, $(*) \leq 2p e^{-\lambda^2 n / (2\sigma^2)}$

$$= 2p \exp(-A^2 \log p / 2) = 2p^{1-A^2/2} \quad \square$$

The following will be helpful for later results

Def 4 (Bernstein's condition)

We say a r.v. W satisfies Bernstein's condition with parameter (σ, b) where $\sigma, b > 0$ if

$$\mathbb{E} |W - \mathbb{E} W|^k \leq \frac{1}{2} k! \sigma^2 b^{k-2} \quad \text{for } k = 2, 3, \dots$$

Prop 14 (Bernstein's inequality) Let W_1, \dots, W_n be independent r.v.s with

$\mathbb{E} W_i = \mu$. Suppose each W_i satisfies Bernstein's condition with parameter (σ, b) .

Then

$$\mathbb{E} e^{\alpha(W_i - \mu)} \leq \exp\left(\frac{\alpha^2 \sigma^2 / 2}{1 - b|\alpha|}\right) \text{ for all } |\alpha| < \frac{1}{b}$$

$$\mathbb{P}\left(\frac{1}{n} \sum_{i=1}^n W_i - \mu \geq t\right) \leq \exp\left(-\frac{n t^2}{2(\sigma^2 + b t)}\right) \quad \forall t \geq 0$$

Proof Fix i and let $W = W_i$ and let $|\alpha| < \frac{1}{b}$.

$$\mathbb{E} e^{\alpha(W - \mu)} = 1 + \alpha \mathbb{E}(W - \mu) + \sum_{k=2}^{\infty} \frac{\alpha^k \mathbb{E}(W - \mu)^k}{k!} \quad (\text{by dominated convergence})$$

$$\leq 1 + \sum_{k=2}^{\infty} |\alpha|^k \frac{\mathbb{E}|W - \mu|^k}{k!}$$

$$\leq 1 + \frac{\sigma^2 \alpha^2}{2} \sum_{k=2}^{\infty} |\alpha|^{k-2} b^{k-2} = 1 + \frac{\sigma^2 \alpha^2}{2} \frac{1}{1 - |\alpha|b}$$

$$\stackrel{1 + u \leq e^u}{\leq} \exp\left(\frac{\sigma^2 \alpha^2 / 2}{1 - |\alpha|b}\right)$$

For the probability bd

$$\mathbb{E} \exp\left(\sum_{i=1}^n \alpha(W_i - \mu) / n\right) = \prod_{i=1}^n \mathbb{E} \exp(\alpha(W_i - \mu) / n)$$

$$\leq \exp\left(n \frac{(\alpha/n)^2 \sigma^2 / 2}{1 - b|\alpha/n|}\right) \quad \left|\frac{\alpha}{n}\right| < \frac{1}{b}$$

Chernoff bd

$$\mathbb{P}\left(\frac{1}{n} \sum W_i - \mu \geq t\right) \leq e^{-\alpha t} \exp\left(n \frac{(\alpha/n)^2 \sigma^2 / 2}{1 - b|\alpha/n|}\right) \quad \left|\frac{\alpha}{n}\right| < \frac{1}{b}$$

Setting $\frac{\alpha}{n} = \frac{t}{bt + \sigma^2} \in [0, \frac{1}{b})$ gives the result. \square

Lemma 15 Let W, Z be mean-zero sub-G v.v.'s with parameters σ_W and σ_Z resp.

Then WZ satisfies Bernstein's condition with param $(8\sigma_W\sigma_Z, 4\sigma_W\sigma_Z)$.

Proof Note that $W^{2k} = \int_0^\infty \mathbb{1}_{\{x < W^{2k}\}} dx$

$$\mathbb{E} W^{2k} = \int_0^\infty \mathbb{P}(W^{2k} > x) dx$$

$$\text{subst } t^{2k} = x \quad 2k t^{2k-1} dt = dx$$

$$= 2k \int_0^\infty t^{2k-1} \mathbb{P}(|W| > t) dt$$

$$\leq 4k \int_0^\infty t^{2k-1} \exp\left(-\frac{t^2}{2\sigma_W^2}\right) dt$$

$$\text{subst } x = \frac{t^2}{2\sigma_W^2}, \quad \sigma_W^2 dx = t dt$$

$$= 4k\sigma_w^2 \int_0^\infty (2\sigma_w^2 x)^{k-1} e^{-x} dx = 2^{k+1} \sigma_w^{2k} k \underbrace{\int_0^\infty x^{k-1} e^{-x} dx}_{\Gamma(k) = (k-1)!}$$

$$= 2^{k+1} \sigma_w^{2k} k!$$

For any r.v. Y , $E|Y - EY|^k = 2^k E|\frac{1}{2}Y - \frac{1}{2}EY|^k$

$$\leq 2^{k+1} (E|Y|^k + |EY|^k) \text{ by Jensen's inequality}$$

$$\leq 2^k E|Y|^k \text{ applied to } t \mapsto |t|^k$$

Thus $E|WZ - E WZ|^k \leq 2^k E|WZ|^k \leq 2^k (E W^{2k})^{1/2} (E Z^{2k})^{1/2} \text{ by C-1}$

$$\leq 2^{2k+1} \sigma_w^k \sigma_z^k k! = \frac{k!}{2} (8\sigma_w \sigma_z)^2 (4\sigma_w \sigma_z)^{k-2} \quad \square$$