Bounding the Variance of a Sub-Gaussian Random Variable

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Abstract

In this note, we will study the variance of sub-gaussian random variables.

Theorem 1. Let X be a zero-mean, σ^2 -subGaussian random variable. $\operatorname{Var}[X] \leq \sigma^2$.

Proof. By the subGaussian property, for any $\lambda \in \mathbb{R}$ we have $\mathbb{E}[e^{\lambda Y}] \leq e^{\lambda^2 \sigma^2/2}$. We can write

$$2e^{\frac{\lambda^2 \sigma^2}{2}} \ge \mathbb{E}[e^{\lambda Y}] + \mathbb{E}[e^{-\lambda Y}] \tag{1}$$

$$= 2 + 2\sum_{k=1}^{\infty} \lambda^{2k} \frac{\mathbb{E}[Y^{2k}]}{(2k)!}$$
(2)

$$\geq 2 + 2\lambda^2 \frac{\mathbb{E}[Y^2]}{2}.$$
(3)

Hence, for all $\lambda > 0$, we have $h(\lambda) = 1 + \frac{\lambda}{2}\mathbb{E}[Y^2] - e^{\lambda\sigma^2/2} \le 0$. We have h(0) = 0, hence h'(0) should be negative. Thus

$$h'(\lambda) = \frac{1}{2}\mathbb{E}[Y^2] - \frac{\sigma^2}{2}e^{\lambda\sigma^2/2}\Big|_{\lambda=0} \le 0 \implies \mathbb{E}[Y^2] \le \sigma^2, \tag{4}$$

which concludes the proof.

Remark 2. The equality $\mathbb{E}[Y^2] = \sigma^2$ might not hold even if σ is the smallest number satisfying $\mathbb{E}[e^{\lambda Y}] \leq e^{\lambda^2 \sigma^2/2}$. To see this, let X be the following discrete random variable

$$X = \begin{cases} +1, & p \\ 0, & 1-2p \\ -1, & p \end{cases}$$
(5)

It can easily be seen that $\mathbb{E}[X] = 0$ and $\mathbb{E}[X^2] = 2p$. Assume that the equality holds with $\sigma^2 = \operatorname{Var}[X] = 2p$.

$$g(\lambda) \triangleq \log \mathbb{E}[e^{\lambda X}] - \frac{\lambda^2 \sigma^2}{2} - \lambda \mu$$

= $\log(pe^{\lambda} + pe^{-\lambda} + 1 - 2p) - \lambda^2 p \le 0 \quad \forall \lambda \in \mathbb{R}.$

Now, let p = 0.1 and $\lambda = 1$, then

$$\log(0.1e^1 + 0.1e^{-1} + 1 - 0.2) - 1.1 \approx 0.00311 \ge 0$$

which is a contradiction.