

# EE5603:Conceention Inequalities

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## 1 CONVERGENCE

### 1.1 Definitions

**Review: •Hoeffding's lemma:** For a random variable  $x$  with  $E[Xi] = 0$  and  $a \leq x \leq b$ ,

$$E[e^{sx}] \leq e^{s^2 \frac{(b-a)^2}{8}} \quad (1.1)$$

### Recall

$$L(h) = -hp + \log[(1-p) + p.e^h] \text{ where} \quad (1.2)$$

$$h = s.(b-a), p = \frac{-a}{(b-a)} \quad (1.3)$$

$$L''(h) \leq \frac{1}{4} \text{ for any } h. \quad (1.4)$$

**•Hoeffding's inequality:** if  $s_n \sum_{i=1}^n x_i$  where  $x_i$ 's are independent RVs with  $ai \leq x_i \leq bi$  then

$$\Pr s_n - ES_n \leq t \leq \exp\left[\frac{-2t^2}{\sum_{i=1}^n (bi - ai)^2}\right], \quad (1.5)$$

$$\Pr s_n - ES_n \leq -t \leq \exp\left[\frac{-2t^2}{\sum_{i=1}^n (bi - ai)^2}\right] \quad (1.6)$$

**Taylor's theorem:** if  $f(x)$  is a continuous function in the bound interval  $[a, b]$  and has  $f'(x)$  and  $f''(x)$  define in this interval then

$$f(h) = f(0).h^0 + f'(0).\frac{h^1}{1!} + f''(v).\frac{h^2}{2!} \quad (1.7)$$

sub- gaussian RV: A real valued RV  $x$  is sub-gaussian if there exists a  $\sigma$  such that

$$E[e^{\lambda x}] \leq \exp\left(\frac{\lambda^2 \sigma^2}{2}\right) \text{ for any } \lambda \quad (1.8)$$

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**• Observation about the Hoeffding's inequality:**  
The bound does not involve the variance of the R.V .

**• Question:** can we find a tighter bound when the RV has low variance.? Yes, the Bennett's inequality.

## 2 BENNETT'S

let,

$$X_1, \dots, X_n \quad (2.1)$$

be independent RVS with finite variance and  $x_i \leq b$  for  $b > 0$  almost surely for  $i \leq n$ .

let

$$\vartheta = \sum_{i=1}^n E[x_i^2]. \text{ if } s = \sum_{i=1}^n (x_i - E x_i), \text{ then} \quad (2.2)$$

$$\log[Ee^{\lambda s}] \leq n \cdot \log\left[1 + \frac{\vartheta}{nb^2} \phi(\lambda b)\right] \leq \quad (2.3)$$

where

$$\phi(u) = e^u - u - 1 \quad u \in \mathbb{R} \quad (2.4)$$

**Proof:** • let us assume  $b=1$ .

• show that  $u^1 \cdot \phi(u)$  is a non decreasing function.

$$g(u) = \frac{\phi(u)}{u^2} \Rightarrow g^1(u) = \frac{e^u(u-2) + (u+2)}{u^3} \quad (2.5)$$

$$(\lambda x_i)^2 \phi(\lambda x_i) \leq \lambda^2 \phi(\lambda) \quad (2.6)$$

$$\Rightarrow \phi(\lambda x_i) \leq x_i^2 \phi(\lambda) \quad (2.7)$$

$$\text{i.e. } [e^{\lambda x_i} - \lambda x_i - 1 \leq x_i^2 (e^\lambda - \lambda - 1)] \quad (2.8)$$

$$E[e^{\lambda x_i - \lambda x_i - 1}] \leq \phi(\lambda). E[x_i^2] \quad (2.9)$$

$$E[e^{\lambda x_i}] \leq (E[\lambda X_i] + 1 + E[x_i^2] \cdot \phi(\lambda)) \quad (2.10)$$

$$\sum_{i=1}^n \log E[e^{\lambda x_i}] \leq \sum_{i=1}^n \log(E[\lambda x_i]) + 1 + E[XI^2] \cdot \phi(X) \quad (2.11)$$

$$\log[\prod_{i=1}^n E[e^{\lambda \cdot x_i}]] \leq \longrightarrow 1 \quad (2.12)$$

$$\log[\prod_{i=1}^n E[e^{\lambda x_i - E x_i}] + \sum_{i=1}^n \lambda \cdot E[x_i]] \leq \sum_{i=1}^n \log(E[\lambda x_i] + E[x_i^2]) \phi(\lambda) \quad (2.13)$$

$$\psi(\lambda) \text{ where } \psi_s(\lambda) = \log [E[e^{\lambda s}]]$$

$$\psi_s(\lambda) \leq \sum_{i=1}^n \log[E(\lambda x_i) + 1 + E[x_i^2]] \phi(\lambda) \quad (2.14)$$

$$-\sum_{i=1}^n \cdot E x_i. \quad (2.15)$$

$$\leq n \sum_{i=1}^n \frac{\log}{n} [] - \sum_{i=1}^n \lambda E[x_i] \quad (2.16)$$