

# EE5603:Conceention Inequalities

J. Balasubramaniam<sup>†</sup> and G V V Sharma<sup>\*</sup>

## 1 CONVERGENCE

### 1.1 Definitions

show that

## 2 REVIEW

$$u_i - L_i \leq c_i, u_i = \sup_{\lambda' \in x} E[g(x^n)|x^{1=1}, \lambda'] - E[g(x^{1=1})] \quad (2.1)$$

$$L_i = \inf_{\lambda \in x} E[g(x^n)|x^{1=1}, x] - E[g(x^n)|x^{1=1}] \quad (2.2)$$

$$u_i - L_i = \sup_{\lambda \in x} \sup_{\lambda' \in x} E[g(x^n)|x^{1=1} - E[g(x^n)|x^{1=1}, x'] \quad (2.3)$$

## 3 EFROM-STEIN INEQUALITY

let

$$X_1, \dots, X_n \quad (3.1)$$

be independent RVs, let  $f : x^n \rightarrow \mathbb{R}$  be a square integrable,  $z = f(x_1, \dots, x_n)$ .

$$\text{Var}(z) \leq \sum_{i=1}^n E[(z - E^i(z))^2] \quad (3.2)$$

$$\bullet E_i(z) = E[f(x_1, \dots, x_n)|x^i]; E_o = E \quad (3.3)$$

$$\bullet E^i(z) = \int_{\lambda_i \in x} f(x_1, \dots, x_i, \dots, x_n) dp(x_i) \quad (3.4)$$

$$\bullet \text{if } \Delta_i = E_i(z) - E_{i-1}(z), \sum_{i=1}^n \Delta_i = z - E(z). \rightarrow 1 \quad (3.5)$$

$$\bullet \text{Var}(z) = E[(z - E(z))^2] (\text{from defn}) \quad (3.6)$$

$$= E[(\sum_{i=1}^n \Delta_i)^2] (\text{from 1}) \quad (3.7)$$

$$= E[\sum_{i=1}^n \Delta_i^2 + 2\sum_{j>i} \Delta_i \Delta_j] \quad (3.8)$$

$$E[\sum_{i=1}^n \Delta_i^2] + 2\sum_{j>i} E[\Delta_i \Delta_j] \quad (3.9)$$

**Claim:** if  $j > i, E_i \Delta_i = 0$

$$E_i \Delta_j = E_i[E_j(z) - E_{j-1}(z)] \quad (3.10)$$

$$= E[E_j(z) - E_{j-1}(z)|x^i] \quad (3.11)$$

$$= E_j(z) - E_i(z) \leftarrow (\text{since } j > i) \quad (3.12)$$

$$= 0 \quad (3.13)$$

if  $E_i \Delta_j = 0, j > i$ , then  $2\sum_{j>i} E[\Delta_i \Delta_j] = ? = 0$

$$\Rightarrow \text{Var}(z) = E(\sum_{i=1}^n \Delta_i^2) \rightarrow 2 \quad (3.14)$$

• Recall.  $E_i[E^i(z)] = E_{i-1}(z)$

$$E^i(z) = \int_{\lambda_i \in x} f(x_1, \dots, x_{i-1}, x_i, x_{i+1}, \dots, x_n) dp(x_i) \quad (3.15)$$

$$E^i(z) = \int_{\lambda_{i+1} \in x^{n-1}} f(x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n) d.p(x_{i+1}^n) \quad (3.16)$$

$$\Rightarrow E_i[E^i(z)] = \int_{x_{i+1} \in x^{n-i}} \int_{x_i \in \mathcal{X}} f(x_1, \dots, x_i, \dots, x_{i-1}, \dots, x_n). dp(x_i) \quad (3.17)$$

$$= \int_{x_i \in x^{n-(i-1)}} f(x_1, \dots, x_{i-1}, x_i, \dots, x_n). dp(x_i^n) \quad (3.18)$$

$$= E_{i-1}(z) \rightarrow 3 \quad (3.19)$$

$$\Delta_i = E_i(z) - E_{i-1}(z) \quad (3.20)$$

$$\Delta_i = E_i[z - E^i(z)]. \quad (3.21)$$

$$\Delta_i^2 = [E_i[z - E^i(z)]]^2. \quad (3.22)$$

<sup>†</sup> The author is with the Department of Mathematics, IIT Hyderabad. <sup>\*</sup> The author is with the Department of Electrical Engineering, IIT, Hyderabad 502285 India e-mail: {jbalagadepall}@iith.ac.in. All material in the manuscript is released under GNU GPL. Free to use for all.

**jensen's inequality** if  $g$  is a convex function and  $x$  is a random variable then

$$E[f(x)] \geq f(Ex) \quad (3.23)$$

Apply this result to  $\Delta_i$

$$E[\Delta_i^2] \geq (E[\Delta_i])^2 \text{ or } \quad (3.24)$$

$$(E[\Delta_i])^2 \leq E[\Delta_i^2] \rightarrow 4 \quad (3.25)$$

Recall  $\text{Var} = E[\sum_{i=1}^n \Delta_i^2]$