

$$W = \sum_i \xi_i \quad E\xi_i = 0 \quad \sum E\xi_i^2 = 1$$

$$P(a \leq W \leq b) \leq b-a+2\gamma, \quad \gamma = \sum E|\xi_i|^3$$

$$E(Wf(W)) = \sum E \int_{-\infty}^{\infty} f'(W^{(i)} + t) K_i(t) dt$$

idea:  $E(Wf(W)) = \sum E(\xi_i f(W)) = \sum E\xi_i (f(W) - f(W - \xi_i))$

$$= \sum E\xi_i \int_{-\infty}^{\infty} f'(W+t)(1(-\xi_i < t < 0) - 1(0 < t < \xi_i)) dt$$

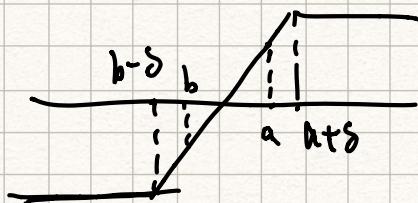
$$= \sum E \int_{-\infty}^{\infty} f'(W+t) \underbrace{\xi_i (1(\cdot) - 1(\cdot))}_{\sim} dt$$

hope it to be

$\hat{K}_i(t)$  'hat' means random variable

$1(a \leq W \leq b)$

$\geq 0$



构造函数

$$f(w) = \begin{cases} -\frac{b-a}{2}s - s, & w < a-s \\ w - \frac{a+b}{2}, & a-s \leq w \leq b+s \\ \frac{b-a}{2}s + s, & w > b+s \end{cases}$$

$$|f(w)| \leq \frac{b-a}{2} + s \Rightarrow |f(w)| \leq \frac{b-a}{2} + s$$

$$f(w) \geq 0$$

$$E Wf(W) \leq \left( \frac{b-a}{2} + s \right) E|W| \leq \frac{b-a}{2} + s$$

upper bound

$$\sum E \int_{-\infty}^{\infty} f'(W+t) \hat{K}_i(t) dt \geq \sum E 1(a \leq W \leq b) \int_{|t| \leq s} \hat{K}_i(t) dt$$

$$= \sum E \mathbb{1}(a \leq w \leq b) |\xi_i| \min(|\xi_i|, \delta)$$

$$= E \mathbb{1}(a \leq w \leq b) \underbrace{\sum |\xi_i| \min(|\xi_i|, \delta)}_{\eta_i}$$

$$= E \mathbb{1}(a \leq w \leq b) \sum (E \eta_i) + E \mathbb{1}(a \leq w \leq b) \sum (\eta_i - E \eta_i)$$

$$\geq P(a \leq w \leq b) \sum (E \eta_i) - E \left| \sum (\eta_i - E \eta_i) \right|$$

$$\leq \sqrt{\sum E \eta_i^2}$$

$$\leq \sqrt{\sum E \xi_i^2 \delta^2} = \delta$$

$$\therefore \frac{(b-a)}{2} + \delta \geq P(a \leq w \leq b) \sum E \eta_i - \delta \geq \frac{1}{2} P(a \leq w \leq b) - \delta$$

$\min(x, y) \geq x - \frac{x^2}{4y}$

$$E \eta_i \geq E \left( \xi_i^2 - \frac{|\xi_i|^3}{4\delta} \right)$$

$$\sum E \eta_i \geq \sum E \xi_i^2 - \frac{1}{4\delta} \sum E |\xi_i|^3 \quad \text{choose } \delta = \frac{1}{2} \sum E |\xi_i|^3$$

$$= 1 - \frac{1}{2} = \frac{1}{2}$$

$$\Rightarrow P(a \leq w \leq b) \leq b-a+4\delta$$

$$\begin{aligned} E(\xi_i^2 \mathbb{1}(|\xi_i| \leq \delta)) &= E \xi_i^2 - E \xi_i^2 \mathbb{1}(|\xi_i| > \delta) \\ &\geq E \xi_i^2 - E |\xi_i|^3 / \delta \end{aligned}$$

$$\textcircled{1} \quad |P(W \leq z) - \Phi(z)| \leq C \sum E |\xi_i|^3$$

$$\textcircled{2} \quad |P(W \leq z) - \Phi(z)| \leq C \sum E |\xi_i|^3 / (1 + |\zeta|)^3$$

non-uniform  
bayesian bound.

$$f(w) = \begin{cases} e^{\lambda w} (w - (a - \delta)) & , a - \delta \leq w < b + \delta \\ e^{\lambda w} (b - a + 2\delta) & , w \geq b + \delta \end{cases}$$

$$f'(w) \geq e^{\lambda w} , a - \delta \leq w \leq b + \delta$$

$$E e^{\lambda w} \mathbf{1}(a \leq w \leq b) \geq e^{\lambda a} P( )$$

$$T = W + \Delta$$

$$W = \sum \xi_i \quad E \xi_i = 0$$

$$\Delta = \Delta(\xi_1, \dots, \xi_n)$$

$$|P(W + \Delta \geq z) - \Phi(z)|$$

$$\leq \sup_x |P(W \leq x) - \Phi(x)| + 2(E|\delta|^p)^{\frac{1}{p+1}}, p > 0$$

$$+ 4E|W\Delta|$$

- Shorack (2000). Probability for Statisticians.

$$P(W + \Delta \leq z) - P(W \leq z) \leq P(z < W \leq z - \Delta) \leq \cdots + E|W\Delta|$$

$$\uparrow \text{const.} \quad \uparrow \text{r.v.} \quad + \sum E|\xi_i (\Delta_{i,i})|$$

### Randomized Concentration Inequality

$$P(\Delta_1 \leq W \leq \Delta_2) \leq 4 \sum E|\xi_i|^3 + E|W(\Delta_2 - \Delta_1)|$$

$$+ \sum E|\xi_i(\Delta_1 - \Delta_{1,i})| + \sum E|\xi_i(\Delta_2 - \Delta_{2,i})|$$

$$\text{where } \Delta_{1,i} = \Delta_{1,i}(\xi_j, j \neq i)$$

$$\Delta_{2,i} = \Delta_{2,i}(\xi_j, j \neq i)$$

[ Pf. ]

$$f_{a,b}(w) = \begin{cases} -\frac{b-a}{2} - s, & w \leq a-s \\ w - \frac{a+b}{2}, & a-s \leq w \leq b+s \\ \frac{b-a}{2} + s, & w \geq b+s \end{cases}$$

Refer to  
Chapter 10

$$E W f_{\Delta_1, \Delta_2}(w) = \sum E \xi_i (f_{\Delta_1, \Delta_2}(w) - f_{\Delta_1, \Delta_2}(w - \xi_i))$$

$$+ \sum E_{\xi_i}^e \left( f_{\Delta_1 \Delta_2}(\omega_{\xi_i}) - f_{\Delta_1 i \Delta_2 i}(\omega_{\xi_i}) \right) \#]$$

↑ (trick)      ↑ independent

Note:  $|f_{a,b}(w) - f_{c,d}(w)| \leq |a-c| + |b-d|$

## U-Statistics

$$\text{Examples: } U_n = \frac{1}{\binom{n}{m}} \sum_{1 \leq i_1 < i_2 < \dots < i_m \leq n} h(X_{i_1}, \dots, X_{i_m})$$

$$\theta = Eh(X_1, \dots, X_m)$$

$$\underline{m = 2}$$

$$h(y, x) = h(x, y) \quad \theta = E h(x_1, x_2)$$

Korolyuk, Borouskikh (1994). Theory of U-Stat.

[Pf. W.L.O.G. Assume  $\theta = 0$

$$\sum_{1 \leq i < j \leq n} h(x_i, x_j) = \sum_{j=2}^n \sum_{i=1}^{j-1} [h(x_i, x_j) - E[h(x_i, x_j | \mathcal{F}_{j-1})] - g(x_j)]$$

constructed       $g(x_i)$

$$F_j = \sigma(X_1, \dots, X_j)$$

$$g(x) = Eh(x, X_2)$$

$$\sigma_1^2 = \text{Var}(g(x_1))$$

$$+ \sum_{j=2}^r \sum_{i=1}^{j-1} (g(x_i) + g(x_j))$$

$$= (n-1) \sum_{i=1}^n g(x_i)$$

$$\begin{aligned}
 U_n &= \frac{1}{\binom{n}{2}} \sum_{1 \leq i < j \leq n} h(x_i, x_j) \\
 &= \frac{1}{\binom{n}{2}} \left( (n-1) \sum g(x_i) + \sum_{j=2}^n \sum_{i=1}^{j-1} (h(x_i, x_j) - g(x_i) - g(x_j)) \right) \\
 &= \frac{2}{n} \sum g(x_i) + \frac{2}{n(n-1)} \sum_{j=2}^n \sum_{i=1}^{j-1} (\dots) \\
 \frac{\sqrt{n}}{2} \frac{U_n}{\sigma_1} &= \frac{\sum g(x_i)}{\sqrt{n} \sigma_1} + \frac{1}{\sqrt{n}(n-1)} \sum_{j=2}^n \sum_{i=1}^{j-1} (\dots) \\
 &= \sum \xi_i + \delta
 \end{aligned}$$

$$E(\delta^2) = O(\frac{1}{n})$$

$$E[(\delta - \delta_i)^2] = O(\frac{1}{n^2})$$

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