

Taylor series expansion of $g(Y_i; \hat{\theta})$ around θ

$$g(Y_i; \hat{\theta}) = g(Y_i; \theta) + (\hat{\theta} - \theta)^T g_{\theta}(Y_i; \theta) + \frac{1}{2} (\hat{\theta} - \theta) g_{\theta\theta}(Y_i; \theta^*) (\hat{\theta} - \theta)$$

where θ^* is between θ and $\hat{\theta}$

This implies that

$$T_n = \frac{1}{n} \sum_{i=1}^n g(Y_i; \hat{\theta})$$

$$= \frac{1}{n} \sum_{i=1}^n g(Y_i; \theta) + (\hat{\theta} - \theta)^T \frac{1}{n} \sum_{i=1}^n g_{\theta}(Y_i; \theta)$$

$$+ \frac{1}{2} (\hat{\theta} - \theta)^T \frac{1}{n} \sum_{i=1}^n g_{\theta\theta}(Y_i; \theta_i^*) (\hat{\theta} - \theta)$$

(Replace $\hat{\theta} - \theta$ by $\frac{1}{n} \sum h(Y_i) + R_n$ in 2nd summand)

$$= \frac{1}{n} \sum g(Y_i; \theta) \quad (A1)$$

$$+ \left(\frac{1}{n} \sum_{i=1}^n h(Y_i) + R_n \right) \left(\frac{1}{n} \sum g_{\theta}(Y_i; \theta) \right) \quad (A2)$$

$$+ \frac{1}{2} (\hat{\theta} - \theta)^T \frac{1}{n} \sum g_{\theta\theta}(Y_i; \theta_i^*) (\hat{\theta} - \theta) \quad (A3)$$

$$= A1 + A2 + A3$$

when we examine next each term in turn. We start with $A3$ and slowly move towards $A1$.

$$A_3 = \underbrace{\frac{1}{2}(\hat{\theta} - \theta)^T}_{O_p(n^{-1/2})} \underbrace{\left(\frac{1}{n} \sum_{i=1}^n \frac{g(Y_i; \theta_i^*)}{\theta_i^{*T}} \right)}_{b \times b \text{ matrix}} \underbrace{(\hat{\theta} - \theta)}_{O_p(n^{-1/2})}$$

the term in the middle has the summand with the property:

$$\left| \frac{g_{\theta_i \theta_i'}(Y_i, \theta_i^*)}{\theta_i^{*T}} \right| \leq M(Y_i) \quad \forall \theta_i^* \text{ around } \theta$$

with $EM(Y_i) < \infty$.

$$\left| \frac{1}{n} \sum \frac{g_{\theta_i \theta_i'}(Y_i, \theta_i^*)}{\theta_i^{*T}} \right| \leq \frac{1}{n} \sum M(Y_i) = O_p(1) \text{ since } EM(Y_i) < \infty$$

Thus $A_3 = O_p(n^{-1/2}) O_p(1) O_p(n^{-1/2}) = O_p(n^{-1/2})$

$$A_2 = \left(\frac{1}{n} \sum h(Y_i) + R_n \right) \left(\frac{1}{n} \sum g_{\theta}(Y_i; \theta) \right)$$

$$= \frac{1}{n} \sum h(Y_i) \cdot \frac{1}{n} \sum g_{\theta}(Y_i; \theta) + \underbrace{R_n}_{O_p(n^{-1/2})} \cdot \underbrace{\frac{1}{n} \sum g_{\theta}(Y_i; \theta)}_{O_p(1) \text{ since } E g_{\theta}(Y_i; \theta) < \infty}$$

$$= \left(\frac{1}{n} \sum h(Y_i) \right) \cdot \left(\frac{1}{n} \sum g_{\theta}(Y_i; \theta) \right) + O_p(n^{-1/2})$$

write this as $E g_{\theta}(Y_i; \theta) + \left\{ \frac{1}{n} \sum g_{\theta}(Y_i; \theta) - E g_{\theta}(Y_i; \theta) \right\}$

$$= \frac{1}{n} \sum h(Y_i) \cdot E[g_\theta(Y_i; \theta)]$$

$$+ \underbrace{\left(\frac{1}{n} \sum h(Y_i) \right)}_{O_p(n^{-1/2})} \underbrace{\left(\frac{1}{n} \sum g_\theta(Y_i; \theta) - E g_\theta(Y_i; \theta) \right)}_{O_p(1) \text{ from WLLN}} + O_p(n^{-1/2})$$

Thus

$$A_2 = \frac{1}{n} \sum h(Y_i) E[g_\theta(Y_i; \theta)] + O_p(n^{-1/2})$$

Since $\frac{1}{n} \sum g_\theta(Y_i; \theta) \xrightarrow{P} E[g_\theta(Y_i; \theta)]$ WLLN we can represent $(*)$

$$T_n = E[g_\theta(Y_i; \theta)]$$

$$+ \frac{1}{n} \sum_{i=1}^n \left\{ \underbrace{g_\theta(Y_i; \theta) - E[g_\theta(Y_i; \theta)]}_{h_T(Y_i)} + h(Y_i) E[g_\theta(Y_i; \theta)] \right\}$$

$$+ R_{2n} \quad \text{where } R_{2n} = O_p(n^{-1/2}).$$

$$\text{Since } E \left[g_\theta(Y_i; \theta) - E[g_\theta(Y_i; \theta)] + h(Y_i) E[g_\theta(Y_i; \theta)] \right]$$

$$= \underbrace{E[g_\theta(Y_i; \theta)] - E[g_\theta(Y_i; \theta)]}_{=0} + \underbrace{E[h(Y_i)]}_{=0} E[g_\theta(Y_i; \theta)] = 0$$

\Rightarrow the above representation $(*)$ is the Bahadur repes of T_n //