

M-泛函的经验似然置信区间*

Empirical Likelihood Confidence Intervals for *M*-Functionals

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摘要 利用经验似然思想, 分别讨论不含附加信息和含附加信息时, *M*-泛函的经验似然置信区间, 并推广 Zhang Biao (1997) 在独立同分布下的结果.

关键词 *M*-泛函 经验似然 置信区间 附加信息

中图法分类号 O212.2

Abstract We employ the method of empirical likelihood to construct the confidence intervals for *M*-functionals in the presence of auxiliary information and without auxiliary information, and extend Zhang Biao's (1997) results.

Key words *M*-functionals, empirical likelihood, confidence intervals, auxiliary information

1 定义

***M*-泛函** X_1, \dots, X_n 为强平稳 O -混合(见文献 [1] 中定义)的随机变量, 分布函数为 F , *M*-泛函 $\theta(F)$ 定义为下面方程的根

$$E_{F^h}(X, \theta_0) = \int h(x, \theta_0) dF(x) = 0, X \sim F, x, \theta_0, h(x, \theta_0) \in R. \quad (1.1)$$

(a) 不带附加信息时的经验似然比统计量

$$E_{F^h}(X, \theta_0) = 0.$$

在约束条件下,

$$\sum_{i=1}^n p_i = 1, p_i \geq 0 \sum_{i=1}^n p_i h(X_i, \theta_0) = 0,$$

则经验似然比统计量为

$$l_1(\theta_0) = \sum_{i=1}^n \log [1 + \lambda h(X_i, \theta_0)], \quad (1.2)$$

其中 λ 满足

$$\sum_{i=1}^n \frac{h(X_i, \theta_0)}{1 + \lambda h(X_i, \theta_0)} = 0. \quad (1.3)$$

(b) 含附加信息时的经验似然比统计量

$$E_{Fg}(X) = 0, g(x) = (g_1(x), \dots, g_r(x))', \quad (1.4)$$

令 $h(x, \theta_0) = (g'(x), O(x, \theta_0))'$, 易知 $E_{Fh}(X, \theta_0) =$

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0,由 Zhang Biao(1997) 文献中, 可知经验似然比统计量为^[2]

$$l_2(\theta_0) = \sum_{i=1}^n \log [1 + \lambda' h(X_i, \theta_0)] - \sum_{i=1}^n \log [1 + Zg(X_i)], \quad (1.5)$$

其中 $\lambda = (\lambda_1, \dots, \lambda_{n+1})'$ 满足

$$\frac{1}{n} \sum_{i=1}^n \frac{h(X_i, \theta_0)}{1 + \lambda' h(X_i, \theta_0)} = 0. \quad (1.6)$$

$Z = (Z_1, \dots, Z_n)'$ 满足

$$\frac{1}{n} \sum_{i=1}^n \frac{g(X_i)}{1 + Zg(X_i)} = 0. \quad (1.7)$$

经验似然是由 Owen 引入的一种非参数推断方法^[3~5], Zhang Biao 已经讨论了在有附加信息时 *M*-泛函的经验似然置信区间^[2], 并假设 X_1, X_2, \dots, X_n 为独立同分布的随机变量, 本文假设 X_1, \dots, X_n 为强平稳 O -混合随机变量, 讨论有附加信息及不带附加信息时, *M*-泛函的经验似然置信区间. 本文仅给出似然比统计量的极限分布.

2 引理

引理 1 设 $\{X_i\}$ 为强平稳 O -混合随机变量序列, 满足 $\sum_{k=1}^{\infty} \mathbb{E}^{(k)} < \infty$, 假定 $E|X_1|^2 < \infty$, $EX_1 = 0$, 则 $A^2 = EX_1^2 + \sum_{i=1}^{\infty} E(X_1 X_{1+i})$ 收敛且有

$$\sup_{-\infty < x < \infty} |P\left\{\frac{n^{-\frac{1}{2}} \sum_{j=1}^n X_j}{A} < x\right\} - H(x)| \rightarrow 0,$$

a.s.

注 该引理为文献[6]中系4.7的特例.

引理2 Y_1, \dots, Y_n 为同分布的随机变量, $Z_n = \max_{1 \leq i \leq n} |Y_i|$, 若 $E|Y_1|^s < \infty$, 则 $Z_n = o_p(n^{1/s})$, a.s.

注 参见文献[7].

引理3 令 U_1, \dots, U_n 是一元 O-混合随机变量, $E|U_i|^r < \infty$, 对于 $i = 1, \dots, n$ 及某个 $r \geq 2$, 则

$$E\left|\sum_{i=1}^n U_i\right|^r \leq Cn^{r/2} \max_{1 \leq i \leq n} E|U_i|^r.$$

证明 参见文献[8]中定理2.3的证明.

引理4 令 $\{X_{nj}\}$ 是强平稳 O-混合三角阵列,

$$(1) \quad \left| \sum_{j=1}^{\infty} \frac{1}{j} O^2(j) \right| < \infty, \text{假设 } \{X_{nj}\} \in R, \text{并且}$$

$$(2) \quad EX_{n1} = 0, EX_{n1}^2 < \infty,$$

$$(3) \quad \text{对于任意的 } X > 0, \lim_{n \rightarrow \infty} nE[X_n^2 I(|X_{n1}| > X)] = 0,$$

$$\begin{aligned} & (3) \sup_{n \geq 1} nEX_{n1}^2 < \infty, A = \lim_{n \rightarrow \infty} \{nEX_{n1}^2 + \\ & 2\sum_{j=1}^{n-1} E(X_{n1}X_{n+j-1})\} \text{存在}, A > 0, \text{则有} \\ & \sum_{j=1}^n X_{nj} \xrightarrow{d} N(0, A). \end{aligned}$$

注 强平稳 O-混合三角阵列定义参考

Samur^[6],此引理为文献[6]中命题4.6的特例.

引理5^[9] 设 $\{U_i, i \geq 1\}$ 是一元 O-混合随机变量, $E|U_i|^p < \infty, i, j \geq 1, p, q > 1, 1/p + 1/q = 1$, 则

$$|EU_i U_{i+j} - EU_i U_{i+j}| \leq$$

$$2O^{1/p}(j) E^{1/p} |U_i|^p E^{1/q} |U_{i+j}|^q.$$

引理6 如果 $\sum_g = E[g(X)g'(X)]$ 是正定的, $\sum_{k=1}^{\infty} O^2(k) < \infty$, $E\|g(X)\|^4 < \infty$, 则 $Z = G^{-1}\bar{g} + o_p(n^{-1/2})$, 其中 $G = \frac{1}{n} \sum_{i=1}^n g(X_i)g'(X_i)$, $\bar{g} = \frac{1}{n} \sum_{i=1}^n g(X_i)$.

证明 令 $Z = \theta$, 其中 $d = \|Z\|$, $\|\theta\| = 1$, $G = \frac{1}{n} \sum_{i=1}^n g(X_i)g'(X_i)$, $\bar{g} = \frac{1}{n} \sum_{i=1}^n g(X_i)$, $0 = \|g_1(\theta)\| \geq |\theta' g_1(\theta)| = \frac{1}{n} |\theta' (\sum g(X_i) - \sum \frac{g(X_i)\theta' g(X_i)}{1+\theta' g(X_i)})| \geq \frac{d}{n} \theta \sum \frac{g(X_i)g'(X_i)}{1+\theta' g(X_i)} \theta - \frac{1}{n} |\sum_{j=1}^r e_j \sum g(X_i)| \geq \frac{\theta' \theta}{1+d} - \frac{1}{n} |\sum_{j=1}^r e_j \sum g(X_i)|$,

其中 $Z_n = \max_{1 \leq i \leq n} \|g(X_i)\|$, 其中 e_j 的第 j 个分量与 θ 的相应分量相同, 其余分量为 0.

由引理1, $\forall a \in R^r, a \neq 0$,

$$\sup_{-\infty < x < \infty} |P\left\{\frac{n^{-1/2} \sum a' g(X_i)}{A} < x\right\} -$$

$H(x)\right| \rightarrow 0, \text{a.s.}$

$$\text{故 } \frac{1}{n} \sum a' g(X_i) = O_p(1), \\ a' \bar{g} = O_p(n^{-1/2}), \quad (2.1)$$

$$\text{式中, } A^2 = E a' g(X_1) g'(X_1) a + \\ \sum_{i=1}^{\infty} E(a' g(X_1) g'(X_{1+i}) a) = a' [E g(X_1) g'(X_1) + \\ \sum_{i=1}^{\infty} E(g(X_1) g'(X_{1+i}))] a,$$

$$\text{记 } A_1 = E g(X_1) g'(X_1) + \sum_{i=1}^{\infty} E(g(X_1) g'(X_{1+i})),$$

$$\text{即 } \frac{\bar{n} a' \bar{g}}{a' A_1 a} \xrightarrow{d} N(0, 1), \\ \frac{\bar{n} \bar{g}}{a' A_1 a} \xrightarrow{d} N(0, A_1). \quad (2.2)$$

下证 $G = \sum_g + o_p(1)$.

$$G - \sum_g = \frac{1}{n} \sum_{i=1}^n [g(X_i)g'(X_i) - \\ E g(X_i)g'(X_i)],$$

$$E(G - \sum_g)^2 = \frac{1}{n^2} E \left[\sum_{i=1}^n (g(X_i)g'(X_i) - \right.$$

$$E g_k(X_i)g_l(X_i))^2 \leq \frac{C}{n^2} n E(g_k(X_1)g_l(X_1) -$$

$$E g_k(X_1)g_l(X_1))^2 \leq \frac{C}{n} E(g_k(X_1)g_l(X_1))^2 \leq$$

$$\frac{C}{n} [E_{g_k}^4(X_1) + E_{g_l}^4(X_1)] \rightarrow 0.$$

$$\text{故 } G = \sum_g + o_p(1). \quad (2.3)$$

$$\text{故 } \theta' \theta = \theta \sum_g \theta + o_p(1) \geq \frac{e}{p} + o_p(1).$$

由引理2, $Z_n = o_p(n^{1/2})$.

$$\frac{d}{1 + dZ_n} = O_p(n^{-1/2})$$

$$d = \frac{O_p(n^{-1/2})}{1 - Z_n O_p(n^{-1/2})} = \frac{O_p(n^{-1/2})}{1 + o_p(1)} = O_p(n^{-1/2}).$$

$$\text{即 } d = \|Z\| = O_p(n^{-1/2}).$$

令 $V_i = Z g(X_i)$, 则

$$\begin{aligned} \max_{1 \leq i \leq n} |V_i| &= \max_{1 \leq i \leq n} |Z g(X_i)| \leq \\ &\|Z\| \max_{1 \leq i \leq n} \|g(X_i)\| = O_p(n^{-1/2}) o_p(n^{1/2}) = \\ &o_p(1). \end{aligned}$$

$$g_1(Z) = \frac{1}{n} \sum_{i=1}^n \frac{g(X_i)}{1 + Z g(X_i)} = 0,$$

$$Z = G^{-1}\bar{g} + G^{-1} \frac{1}{n} \sum_{i=1}^n g(X_i) V_i^2 / (1 + V_i).$$

$$\frac{1}{n} \|\sum_{i=1}^n g(X_i) V_i^2 / (1 + V_i)\| \leq \frac{1}{n} \sum_{i=1}^n \|g(X_i) V_i\| /$$

$$(1 + V_i) \leq \frac{1}{n} \sum_{i=1}^n \|g(X_i)\|^3 \|Z\|^2 (1 + V_i)^{-1} =$$

$$o_p(n^{1/2}) O_p(n^{-1}) O_p(1) = o_p(n^{-1/2}).$$

$$\text{故 } Z = G^{-1}\bar{g} + o_p(n^{-1/2}).$$

引理7 假设 $\theta_0 = \theta(F)$ 满足(1.1)式, 存在且唯一, $j(x, \theta)$ 关于 x 可测, $\sum_h = E[h(X, \theta_0)h'(X, \theta_0)]$

是正定 $\sum_{k=1}^{\infty} \mathcal{O}^2(k) < \infty$, $\mathcal{Q}(1) < 1, E\|h(X, \theta_0)\|^4 < \infty$, $\frac{\partial h(x, \theta)}{\partial \theta}$ 在 θ_0 点连续, $|\frac{\partial j(x, \theta)}{\partial \theta}| \leq M(x)$, $|\theta - \theta_0| < X, EM^4(X) < \infty$, 则对于 $\theta = \theta_0 + O_p(n^{-1/2})$, $\lambda = H^{-1}\bar{h} + o_p(n^{-1/2})$, 其中 $H = \frac{1}{n} \sum_{i=1}^n h(X_i, \theta)h'(X_i, \theta)$, $\bar{h} = \frac{1}{n} \sum_{i=1}^n h(X_i, \theta)$.

证明 令 $\lambda = dU$, 其中 $d = \|\lambda\|$, $\|U\| = 1$. $\theta = \|g_2(dU)\| \geq |\mathbf{U}'g_2(dU)| = \frac{1}{n} |\mathbf{U} \sum h(X_i, \theta) - \sum \frac{h(X_i, \theta) \mathbf{U}'h(X_i, \theta)}{dU' h(X_i, \theta)}| \geq \frac{d}{n} U$.

$$\sum \frac{h(X_i, \theta) h'(X_i, \theta)}{1 + dU' h(X_i, \theta)} \mathbf{U} - \frac{1}{n} |\sum_{j=1}^{r+1} e_j \sum h(X_i, \theta)| \geq \frac{dU' H \mathbf{U}}{1 + dZ_i} - \frac{1}{n} |\sum_{j=1}^{r+1} e_j \sum h(X_i, \theta)|,$$

其中 e_j 的第 j 个分量与 U 的相应分量相同, 其余为 0, $Z_i = \max_{1 \leq i \leq n} \|h(X_i, \theta)\|$. $\forall a \neq 0, a \in R^{n-1}$, 令 $X_{ni} = \frac{1}{n} a'(h(X_i, \theta) - Eh(X_i, \theta))$.

$$(1) EX_{n1} = 0,$$

$$EX_{n1}^2 = \frac{1}{n} E[a'(h(X_1, \theta) - Eh(X_1, \theta))]^2 \leq \frac{1}{n} \|a\|^2 E\|h(X_1, \theta)\|^2 \leq \frac{1}{n} \|a\|^2 [E(g_1^2(X_1) + \dots + g_r^2(X_1) + h^2(X_1, \theta)) + CEM^2(X_1) + CEM(X_1)] < \infty.$$

$$(2) \forall X > 0,$$

$$nE[X_{n1}^2 I(|X_{n1}| > X)] \leq n \frac{1}{X} EX_{n1}^4 =$$

$$\frac{1}{nX} E[a'(h(X_1, \theta) - Eh(X_1, \theta))]^4.$$

$$I = \frac{1}{n} E[a'(h(X_1, \theta) - Eh(X_1, \theta))]^4 \leq \frac{1}{n} \|a\|^4 E\|h(X_1, \theta)\|^4 \leq \frac{C}{n} \left(\sum_{i=1}^r E g_i^4(X_1) + Eh^4(X_1, \theta) \right) \rightarrow 0.$$

$$(3) nEX_{n1}^2 = E[a'(h(X_1, \theta) - Eh(X_1, \theta))]^2 = a'E(h(X_1, \theta) - Eh(X_1, \theta))(h(X_1, \theta) - Eh(X_1, \theta))'a \rightarrow a'Eh(X_1, \theta_0)h'(X_1, \theta_0)a.$$

$$\sum_{j=1}^{n-1} E(X_{n1} X_{n, j+1}) = \sum_{j=1}^{n-1} a'E(h(X_1, \theta) - Eh(X_1, \theta))(h(X_{1+j}, \theta) - Eh(X_{1+j}, \theta))'a.$$

$$\text{令 } a_j = a'Eh(X_1, \theta_0)h'(X_{1+j}, \theta_0)a, a_{nj} = a'E(h(X_1, \theta) - Eh(X_1, \theta))(h(X_{1+j}, \theta) - Eh(X_{1+j}, \theta))'a.$$

先证 $\sum_{j=1}^{n-1} |a_j|$ 收敛.

$$\sum_{j=1}^{n-1} |a_j| \leq C \sum_{j=1}^{n-1} \mathcal{O}^2(j),$$

故 $\sum_{j=1}^{\infty} a_j$ 收敛, 即 $\sum_{j=1}^{\infty} a'Eh(X_1, \theta_0)h'(X_{1+j}, \theta_0)a$ 收敛.

由引理 5,

$$\sum_{j=1}^{n-1} |a_{nj}| \leq \sum_{j=1}^{n-1} \mathcal{O}^2(j),$$

$\forall X > 0, \exists N, N_1$, 使得当 $n > \max(N, N_1)$ 时,

$$\sum_{N+1}^{\infty} \mathcal{O}^2(j) < X, |a_{nj} - a_j| \leq X$$

$$|\sum_{i=N+1}^n a_{ni}| \leq \sum_{i=N+1}^n |a_{ni}| \leq X,$$

$$|\sum_{i=N+1}^n a_i| \leq X,$$

$$|\sum_{i=1}^n a_{ni} - \sum_{i=1}^n a_i| \leq |\sum_{i=1}^N a_{ni} - \sum_{i=1}^N a_i| +$$

$$|\sum_{i=N+1}^n a_{ni} - \sum_{i=N+1}^n a_i| \leq \sum_{i=1}^n |a_{ni} - a_i| + 2X \leq (N+2)X$$

$$\lim_{n \rightarrow \infty} \sum_{i=1}^n a_{ni} = \sum_{i=1}^{\infty} a_i. \quad (2.4)$$

故 $A = a'[Eh(X_1, \theta_0)h'(X_1, \theta_0) + \sum_{j=1}^{\infty} Eh(X_{1+j}, \theta_0)h'(X_{1+j}, \theta_0)]a$, 且 $A > 0$.

由引理 4, $\sum_{j=1}^n X_{nj} \xrightarrow{d} N(0, A)$.

$$\frac{1}{n} \sum_{j=1}^n a'(h(X_j, \theta) - Eh(X_j, \theta)) \xrightarrow{d} N(0, A).$$

$$\frac{1}{n} a' \bar{h} - \frac{1}{n} a' Eh(X_1, \theta) = O_p(1).$$

$$a' \bar{h} = O_p(n^{-1/2}).$$

$$\frac{1}{n} (\bar{h} - Eh) \xrightarrow{d} N(0, A_2).$$

$$A_2 = Eh(X_1, \theta_0)h'(X_1, \theta_0) +$$

$$\sum_{j=1}^{\infty} Eh(X_1, \theta_0)h'(X_{1+j}, \theta_0).$$

$$\bar{h} \xrightarrow{d} N(b, A_2), b = (0, \mathbf{e}_n \mathbf{W}'). \quad (2.5)$$

下证 $H = Eh + o_p(1)$.

$$H - EH = \frac{1}{n} \sum_{i=1}^n [h(X_i, \theta)h'(X_i, \theta) - Eh(X_i, \theta)h'(X_i, \theta)].$$

$$E(H_k - EH_k)^2 = \frac{1}{n^2} E \sum_{i=1}^n (h_k(X_i, \theta)h_l(X_i, \theta) -$$

$$Eh_k(X_i, \theta)h_l(X_i, \theta))^2 \leq \frac{C}{n^2} n E(h_k(X_1, \theta)h_l(X_1, \theta))^2$$

$$= \frac{C}{n} E(h_k(X_1, \theta)h_l(X_1, \theta))^2 \leq \frac{C}{n} [Eh_k^4(X_1, \theta) + Eh_l^4(X_1, \theta)] \rightarrow 0.$$

$$\text{故 } H = EH + o_p(1).$$

$$\text{又 } H = \sum h + o(1).$$

$$\text{故 } H = \sum h + o_p(1).$$

$$UH_0 \geq \mathbf{e}_n + o_p(1), Z_n = \max_{1 \leq i \leq n} \|h(X_i, \theta)\| = o_p(n^{1/4}).$$

$$\frac{d}{1 + dZ_n} = O_p(n^{-1/2}), d = \|Z\| = O_p(n^{-1/2}).$$

同理, 令 $V = Zh(X_i, \theta)$, 则

$$\max_{1 \leq i \leq n} |V_i| = o_p(1),$$

$$Z = H^{-1} \bar{h} + H^{-1} \sum_{i=1}^n h(X_i, \theta) V_i^2 / (1 + V_i).$$

引理 8 如果 A 是 $(p+q) \times (p+q)$ 满秩方阵, $A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}$, A_{11} 是 $p \times p$ 满秩方阵, A_{22} 是 $q \times q$ 满秩方阵, 则 $(A^{-1} - B_1)A$ 为秩 q 的幂等阵, 其中 $B_1 = \begin{pmatrix} A_{11}^{-1} & O \\ O & O \end{pmatrix}$.

证明 $(A^{-1} - B_1)A = I - B_1 A$.

$$B_1 A = \begin{pmatrix} I_{p \times p} & A_{11}^{-1} A_{12} \\ O & O \end{pmatrix},$$

$$(B_1 A)^2 = \begin{pmatrix} I_{p \times p} & A_{11}^{-1} A_{12} \\ O & O \end{pmatrix} \begin{pmatrix} I_{p \times p} & A_{11}^{-1} A_{12} \\ O & O \end{pmatrix} = \begin{pmatrix} I_{p \times p} & A_{11}^{-1} A_{12} \\ O & O \end{pmatrix} = B_1 A,$$

故 $B_1 A$ 为幂等阵.

$$[(A^{-1} - B_1)A]^2 = (I - B_1 A)(I - B_1 A) = I - 2B_1 A + (B_1 A)^2 = I - B_1 A = (A^{-1} - B_1)A.$$

则 $(A^{-1} - B_1)A$ 为幂等阵.

$$I - B_1 A = \begin{pmatrix} O_{p \times p} & A_{11}^{-1} A_{12} \\ O & I_{q \times q} \end{pmatrix},$$

秩为 q .

3 定理

定理 1 假设 $\theta_0 = \theta(F)$ 满足 (1.1) 式, 存在且唯一, $h(x, \theta)$ 关于 x 可测, 进一步假设 $\lambda_F(\theta_0)$ 有限且非 0, $h(x, \theta)$ 在 θ_0 点关于 x 一致连续, $\frac{\partial h(x, \theta)}{\partial \theta}$ 在 θ_0 连续, $|\frac{\partial h(x, \theta)}{\partial \theta}| \leq M(x)$, 在 θ_0 邻域内, $EM(X)^4 < \infty$, 对于正数 d , 常数 W , 有

$$\lim_{n \rightarrow \infty} P(l_1(\theta) \leq d) = P(\frac{A}{\epsilon_h^2} i_{(1)}^2 (\epsilon_h^2 W) \leq d),$$

由于 A 未知, 所以此结论并不理想, 需要对它改进去掉其中的参数, 为此对其进行分组, 令

$$p' = [n^\top], 0 < \mathbb{T} \leq 1/3, q' = [\frac{n}{p}],$$

$$k_i = \frac{1}{p} \sum_{j=1}^{p'} h(X_{(i-1)p+j}), i = 1, \dots, q',$$

则分组后经验似然比统计量

$$\lim_{n \rightarrow \infty} P(\hat{l}_1(\theta) \leq d) = P(i_{(1)}^2 (\epsilon_h^2 W) \leq d),$$

其中 $\theta = \theta_0 + \frac{\epsilon_h}{n \lambda_F(\theta_0)} W$, $\epsilon_h^2 = E l^2(X, \theta_0)$, $\lambda_F(\theta_0) = E[\partial h(X, \theta)/\partial \theta]_{\theta=\theta_0}]$.

定理 2 假设 $\theta_0 = \theta(F)$ 满足 (1.1) 式, 存在且唯一, $h(x, \theta)$ 关于 x 可测, 进一步假设 $\lambda_F(\theta_0)$ 有限且非 0, $h(x, \theta)$ 在 θ_0 点关于 x 一致连续, $\frac{\partial h(x, \theta)}{\partial \theta}$ 在 θ_0 连续,

$|\frac{\partial h(x, \theta)}{\partial \theta}| \leq M(x)$, 在 θ_0 邻域内, $EM(X)^4 < \infty$, 对

于正数 d , 常数 W , 有

$$\lim_{n \rightarrow \infty} P(l_2(\theta) \leq d) = P(Z' B Z \leq d),$$

其中 $\theta = \theta_0 + \frac{\epsilon_h}{n \lambda_F(\theta_0)} W$, $Z \sim N(a, A_2)$, $a = (0', \epsilon_h W'), B = \sum_h^{-1} - \begin{pmatrix} \sum_g^{-1} & O \\ O & 0 \end{pmatrix}$.

由于我们不知道 A_2 , 所以使用分组经验似然来克服一般经验似然之不足, 令

$$p = [n^\top], 0 < \mathbb{T} \leq 1/3, q = [\frac{n}{p}],$$

$$k_{1i} = \frac{1}{p} \sum_{j=1}^p g(X_{(i-1)p+j}), i = 1, \dots, q,$$

$$k_{2i} = \frac{1}{p} \sum_{j=1}^p h(X_{(i-1)p+j}, \theta_0), i = 1, \dots, q,$$

则似然比统计量为

$$\hat{l}_2(\theta_0) = \sum_{i=1}^q \log(1 + \lambda' k_2) - \sum_{i=1}^q \log(1 + Z k_{1i}). \quad (3.1)$$

其中 λ 满足

$$\frac{p}{q} \sum_{j=1}^q \frac{k_{2j}}{1 + \lambda' k_2} = 0, \quad (3.2)$$

Z 满足

$$\frac{p}{q} \sum_{j=1}^q \frac{k_{1j}}{1 + Z k_{1j}} = 0. \quad (3.3)$$

定理 3 假设同定理 2,

$$\lim_{n \rightarrow \infty} P(\hat{l}_2(\theta) \leq d) = P(i_{(1)}^2(d) \leq d),$$

其中

$$d = \frac{\epsilon_h^2 W}{\epsilon_h^2 + \sum_{j=1}^{\infty} E_h(X_j, \theta_0) h(X_{1+j}, \theta_0) - B_2 A_1^{-1} B_2},$$

$$B_2 = E_h(X, \theta_0), g'(X_1) + \sum_{j=1}^{\infty} E_h(X_{1+j}, \theta_0), g'(X_1).$$

4 定理证明

定理 1 证明 令 $h = \frac{1}{n} \sum_{i=1}^n h^2(X_i, \theta)$, $\bar{h}(\theta) = \frac{1}{n} \sum_{i=1}^n h(X_i, \theta)$, $k_i = \max_{1 \leq i \leq n} |h(X_i, \theta)|$, $|h(X_i, \theta)| = |\bar{h}(X_i, \theta_0) + O(n^{-1/2}) \frac{\partial h(X_i, Y)}{\partial \theta}|$,

故

$$\max_{1 \leq i \leq n} |h(X_i, \theta)| \leq |\bar{h}(X_i, \theta_0)| +$$

$$O(n^{-1/2}) \max_{1 \leq i \leq n} |M(X_i)|,$$

因 $E l^2(X, \theta_0) < \infty$, 故

$$\max_{1 \leq i \leq n} |h(X_i, \theta_0)| = o_p(n^{1/2}).$$

又 $EM^4(X_i) < \infty$,

$$O(n^{-1/2}) \max_{1 \leq i \leq n} |M(X_i)| = o_p(n^{-1/4}),$$

$$\text{故 } \max_{1 \leq i \leq n} |\mathbf{h}(X_i, \theta)| = o_p(n^{1/2}). \quad (4.1)$$

下证 $\mathbf{h} = E\mathbf{h} + o_p(1)$.

$$\mathbf{h} - E\mathbf{h} = \frac{1}{n} \sum_{i=1}^n [\mathbf{h}^2(X_i, \theta) - E\mathbf{h}^2(X_i, \theta)],$$

$$E(\mathbf{h} - E\mathbf{h})^2 = \frac{1}{n^2} E \sum_{i=1}^n (\mathbf{h}^2(X_i, \theta) - E\mathbf{h}^2(X_i, \theta))^2,$$

$$E(\mathbf{h} - E\mathbf{h}) \leq \frac{C_n}{n^2} E(\mathbf{h}^2(X_1, \theta) - E\mathbf{h}^2(X_1, \theta)) \leq$$

$$\frac{C}{n} E\mathbf{h}^2(X_1, \theta) \rightarrow 0,$$

$$\text{故 } \mathbf{h} = E\mathbf{h} + o_p(1).$$

而 $E\mathbf{h} = E\mathbf{h}^2(X_1, \theta_0) + o(1)$, 故

$$\mathbf{h} = E\mathbf{h}^2(X_1, \theta_0) + o_p(1). \quad (4.2)$$

$$\text{令 } X_{ni} = \frac{1}{n} (\mathbf{h}(X_i, \theta) - E\mathbf{h}(X_i, \theta)),$$

$$(1) EX_{ni} = 0, EX_{ni}^2 = \frac{1}{n} E(\mathbf{h}(X_1, \theta) - E\mathbf{h}(X_1, \theta))^2 \leq \frac{1}{n} E\mathbf{h}^2(X_1, \theta) < \infty.$$

$$(2) \forall X > 0,$$

$$nE[X_{ni}^2 I\{|X_{ni}| > X\}] \leq \frac{n}{X} EX_{ni}^4 = \frac{1}{X} \frac{1}{n} E(\mathbf{h}(X_i, \theta) - E\mathbf{h}(X_1, \theta))^4,$$

$$\text{记 } I = \frac{1}{n} E(\mathbf{h}(X_1, \theta) - E\mathbf{h}(X_1, \theta))^4 \leq \frac{1}{n} E\mathbf{h}^4(X_1, \theta) + \frac{1}{n} [E\mathbf{h}(X_1, \theta)]^4 \rightarrow 0.$$

$$(3) nEX_{ni}^2 = E[\mathbf{h}(X_1, \theta) - E\mathbf{h}(X_i, \theta)]^2 \rightarrow E\mathbf{h}^2(X_1, \theta_0).$$

$$\sum_{j=1}^{n-1} E(X_{nj} X_{n+j-1}) = \sum_{j=1}^{n-1} [E\mathbf{h}(X_1, \theta) \mathbf{h}(X_{n+j}, \theta) - E\mathbf{h}(X_1, \theta) E\mathbf{h}(X_{n+j}, \theta)] \rightarrow$$

$$\sum_{j=1}^{\infty} E\mathbf{h}(X_1, \theta_0) \mathbf{h}(X_{n+j}, \theta_0), \text{故}$$

$$A = E\mathbf{h}^2(X_1, \theta_0) + \sum_{j=1}^{\infty} E\mathbf{h}(X_1, \theta) \mathbf{h}(X_{n+j}, \theta_0),$$

$$A > 0.$$

$$\text{由引理 } 4, \sum_{j=1}^n X_{nj} \rightarrow dN(0, A).$$

$$\text{即 } -\frac{1}{n} \sum_{j=1}^n [\mathbf{h}(X_j, \theta) - E\mathbf{h}(X_j, \theta)] \rightarrow dN(0, A).$$

$$\overline{n} \bar{\mathbf{h}} - \overline{n} E\mathbf{h}(X, \theta) \rightarrow dN(0, A).$$

$$\text{故 } \overline{n} \bar{\mathbf{h}} = O_p(1),$$

$$\bar{\mathbf{h}} = O_p(n^{-1/2}),$$

$$\overline{n} \bar{\mathbf{h}} \rightarrow dN(b, A), \quad (4.3)$$

其中 $b = \mathbf{e}_h^T \bar{\mathbf{W}}$, 故

$$l_1(\theta) = ng^{-1}\bar{\mathbf{h}}^2 + o_p(1) \rightarrow \frac{A}{\mathbf{e}_h^2} \bar{\mathbf{l}}_{(1)}^2 (\mathbf{e}_h^T \bar{\mathbf{W}}).$$

证毕.

定理 2 证明

$$l_2(\theta) = \sum_{i=1}^n \log [1 + \lambda' h(X_i, \theta)] - \sum_{i=1}^n \log [1 +$$

$$Zg(X_i)] = n\bar{h}' \bar{H}^{-1} \bar{h} - n\bar{g}' G^{-1} \bar{g} + o_p(1) =$$

$$= (\overline{n} \bar{b})' B (\overline{n} \bar{b}) + o_p(1) \rightarrow Z' B Z.$$

$$\text{其中 } B = \sum_h^{-1} - \begin{pmatrix} \sum_g^{-1} \\ O \\ 0 \end{pmatrix}, \bar{h}(\theta) =$$

$$\sum_{i=1}^n h(X_i, \theta), \bar{b} = (\bar{g}', \bar{h}(\theta_0) + \lambda_F(\theta_0)(\theta - \theta_0))'$$

$$\bar{h} = \bar{b} + o_p(n^{-1/2}), Z \sim Z(a, A_2), a = (0', \mathbf{e}_h^T \bar{\mathbf{W}}').$$

证毕.

定理 3 证明 仿定理 2 证明, 知

$$\overline{pq} \bar{k}_i \rightarrow dN(0, A_1), \quad (4.4)$$

$$a' \bar{k}_i = O_p((pq)^{-1/2}), \quad (4.5)$$

$$Z_q = \max_{1 \leq j \leq q} \|k_{ij}\| = o_p((q/p)^{1/2}). \quad (4.6)$$

$$\text{其中 } G_{(1)} = \frac{p}{q} \sum_{j=1}^q k_{ij} k_{ij}' , \bar{k}_i = \bar{g}, G_{(1)} - EG_{(1)} =$$

$$\frac{p}{q} \sum_{j=1}^q (k_{ij} k_{ij}' - E k_{ij} k_{ij}'),$$

$$E[(G_{(1)} - EG_{(1)})_{kl}]^2 \leq C \frac{p^2}{q} q \max_{1 \leq j \leq q} E(k_{ij,k} k_{ij,l} -$$

$$E k_{ij,k} k_{ij,l})^2 \leq C \frac{p^2}{q} E(k_{ij,k} k_{ij,l})^2 \leq C \frac{p^2}{q} (E k_{ij,k}^4 +$$

$$E k_{ij,l}^4) \leq \frac{C}{q} E g_k^4(X_1) \leq \frac{C}{q} \rightarrow 0.$$

$$\text{故 } G_{(1)} = EG_{(1)} + pq \bar{k}_2.$$

$$\text{而 } EG_{(1)} = \frac{p}{q} \sum_{i=1}^q E k_{li} k_{li}' = p E k_{l1} k_{l1}' =$$

$$\frac{1}{p} E \left(\sum_{j=1}^p g(X_j) \sum_{j=1}^p g'(X_j) \right),$$

$$EG_{(1)} = A_1 + o(1).$$

$$\text{同理可证 } EH_{(1)} = A_2 + o(1). \quad (4.7)$$

$$\mathcal{L}_2(\theta) = pq \bar{k}_2' A_2^{-1} \bar{k}_2 - pq \bar{k}_1' A_1^{-1} \bar{k}_1 + o_p(1) =$$

$$pq \bar{k}_2' B' \bar{k}_2 + o_p(1) = (\overline{pq} \bar{b})' B' (\overline{pq} \bar{b}) + o_p(1) \rightarrow Z' B' Z = \mathbf{i}_1^2(\mathbf{d}^2),$$

$$\text{其中 } B' = A_2^{-1} - \begin{pmatrix} A_1^{-1} & O \\ O & 0 \end{pmatrix}, Z \sim N(a, A_2),$$

$$\mathbf{d}^2 = a' B' a =$$

$$\frac{\mathbf{e}_h^T \bar{\mathbf{W}}}{\mathbf{e}_h^2 + \sum_{j=1}^{\infty} E \mathbf{h}(X_1, \theta_0) \mathbf{h}(X_{n+j}, \theta_0) - B_2 A_1^{-1} B_2'}$$

$$\frac{\mathbf{e}_h^T \bar{\mathbf{W}}}{\mathbf{e}_h^2 + \sum_{j=1}^{\infty} E \mathbf{h}(X_1, \theta_0) \mathbf{h}(X_{n+j}, \theta_0)},$$

其中

$$B_2 = E h(X_1, \theta_0) g'(X_1) + \sum_{j=1}^{\infty} E h(X_{1+j}, \\ \theta_0) g'(X_1),$$

$$B'_2 = Eg(X_1) h(X_1, \theta_0) + \sum_{j=1}^{\infty} Eg(X_1) h(X_{1+j}, \\ \theta_0).$$

证毕.

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