

线性模型中误差方差估计的指数收敛速度*

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考虑线性回归模型

$$y_j = x_j^\top \beta + \mu_j, \quad j = 1, 2, \dots \quad (1)$$

其中 $\{\mu_j\}$ 为独立的试验误差序列, 满足条件:

$$E\mu_j = 0, \quad \text{Var } \mu_j = \sigma^2, \quad 0 < \sigma^2 < \infty, \quad j = 1, 2, \dots \quad (2)$$

根据最小二乘法规定和模型自身特点, 通常, 在考虑极限性质时, 给 σ^2 以估计.

$$\hat{\sigma}_n^2 = \frac{1}{n-r} \left\{ \sum_{j=1}^n \mu_j^2 - \sum_{k=1}^r \left(\sum_{j=1}^n a_{kj} \mu_j \right)^2 \right\}. \quad (3)$$

其中 r 是一个与设计矩阵 $X_n = (x_1 | \dots | x_n)$ 有关的常正整数, 数列 $\{a_{kj}\}$ 满足等式 $\sum_{j=1}^n a_{nk} a_{nj} = \delta_{kn}$, 这里 δ_{kn} 为 Kronecker 符号, $n = 1, 2, \dots$. 本文旨在讨论 $\hat{\sigma}_n^2 \rightarrow \sigma^2$ 的指数收敛速度, 主要结果是:

定理 如果 μ_1, μ_2, \dots 独立, 满足 (2) 式, 对每个 $\eta > 0$, 存在仅依赖于 η 的常数 $M_\eta > 0$, $t_\eta > 0$, 使得对一切 $|t| \leq t_\eta$, 一切 n , 有

$$\prod_{j=1}^n E e^{t(\mu_j^2 - \sigma^2)} = E e^{t \sum_{j=1}^n (\mu_j^2 - \sigma^2)} \leq M_\eta e^{n\eta|t|} \quad (4)$$

又如果对一切 n, k, j , 存在常数 $0 < L < \infty$, 使得 $0 \leq a_{nk}^2 \leq \frac{L}{n}$. 则对任何充分小的 $\varepsilon > 0$, 存在仅依赖于 ε 的常数 $A = A(\varepsilon) > 0$, 和 $0 < \rho = \rho(\varepsilon) < 1$, 使得

$$P\{|\hat{\sigma}_n^2 - \sigma^2| \geq \varepsilon\} \leq A\rho^n. \quad (5)$$

为证定理, 先证一个引理.

引理 设 μ_1, μ_2, \dots 独立、对称, 满足条件 (2)、(4), 又存在 $0 < L < \infty$, 使 $0 \leq a_{nj}^2 \leq \frac{L}{n}$, 且 $\sum_{j=1}^n a_{nj}^2 = 1$, 则对任何充分小的 $\varepsilon > 0$ 和充分大的 n , 存在仅依赖于 ε 的常数 $A > 0$, $0 < \rho < 1$, 使得

$$P\left\{ \left| \left(\sum_{j=1}^n a_{nj} \mu_j \right)^2 - \sigma^2 \right| \geq \frac{n-r}{r} \varepsilon \right\} \leq A\rho^n, \quad (6)$$

证 易知, 只须证明

$$J_n = P\left\{ \left(\sum_{j=1}^n a_{nj} \mu_j \right)^2 \geq \frac{n-r}{r} \varepsilon \right\} \leq A\rho^n. \quad (7)$$

为方便计, 记 $a_{nj} = a_j$, 并设 $\sigma^2 = \frac{1}{L}$. 若不然, 可令 $\tilde{\mu}_j = \frac{\mu_j}{\sigma\sqrt{L}}$, 以 $\varepsilon_1 = \frac{\varepsilon}{\sigma^2 L}$ 代 ε , 由条件

(4) 中 $\eta > 0$ 的任意性, 知存在 $\tilde{M}_\eta > 0$, $\tilde{t}_\eta > 0$, 使当 $|t| \leq \tilde{t}_\eta$ 时对一切 n , 仍有 $\prod_{j=1}^n E e^{t(\tilde{\mu}_j^2 - \frac{1}{L})} = \prod_{j=1}^n E e^{\frac{t}{\sigma^2 L}(\mu_j^2 - \sigma^2)} \leq \tilde{M}_\eta e^{\frac{t}{\sigma^2 L} n \eta}$. 事实上, 只要令 $\tilde{\eta} = \sigma^2 L \eta$, 取 $\tilde{M}_\eta = M_\eta$, 并使 $\left| \frac{t}{\sigma^2 L} \right| \leq |\tilde{t}_\eta|$ 即可.

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由 Чебышев 不等式, 对一切 $t > 0$, 有

$$\begin{aligned} J_n &\leq e^{t(1-\frac{n}{r})\epsilon} Ee^{t(\sum_{j=1}^n a_j \mu_j)^2} = e^{t(1-\frac{n}{r})\epsilon} \sum_{k=0}^{\infty} \frac{t^k E\left(\sum_{j=1}^n a_j \mu_j\right)^{2k}}{k!} \\ &= e^{t(1-\frac{n}{r})\epsilon} \left\{ \left(\sum_{0 < k < \frac{r}{2}n} + \sum_{\frac{r}{2}n < k < \frac{\ln 2}{2}n} + \sum_{k > \frac{\ln 2}{2}n} \right) \frac{t^k E\left(\sum_{j=1}^n a_j \mu_j\right)^{2k}}{k!} \right\} \\ &\triangleq J_{1n} + J_{2n} + J_{3n} \end{aligned} \quad (8)$$

由 μ_1, μ_2, \dots 独立、对称, 知 $E\left(\sum_{j=1}^n a_j \mu_j\right)^{2k}$ 中的非零项皆具形式 $\frac{(2k)!}{(2i_1)!(2i_2)!\dots(2i_l)!}$
 $E(a_{i_1} \mu_{j_1})^{2i_1} E(a_{i_2} \mu_{j_2})^{2i_2} \dots E(a_{i_l} \mu_{j_l})^{2i_l}$, 其中 $i_1 + \dots + i_l = k$. 由 Stirling 公式知

$$\begin{aligned} \frac{(2k)!}{(2i_1)!(2i_2)!\dots(2i_l)!} &\leq \frac{(2k)^{2k+\frac{1}{2}} i_1^{i_1+\frac{1}{2}} i_2^{i_2+\frac{1}{2}} \dots i_l^{i_l+\frac{1}{2}} e^{\frac{1}{12}[\frac{1}{24} + \frac{1}{24}]} }{k^{k+\frac{1}{2}} (2i_1)^{2i_1+\frac{1}{2}} (2i_2)^{2i_2+\frac{1}{2}} \dots (2i_l)^{2i_l+\frac{1}{2}}} \\ &\leq \frac{k^k}{i_1^{i_1} i_2^{i_2} \dots i_l^{i_l}} 2^{\frac{1}{2}(1-l)} e^{\frac{1}{12}\min(k,n)} \leq l^k 2^{\frac{1}{2}(1-l)} e^{\frac{1}{12}\min(k,n)} \end{aligned} \quad (9)$$

这里, 不等式 $\frac{k^k}{i_1^{i_1} i_2^{i_2} \dots i_l^{i_l}} \leq l^k$ 可由 $f(t) = tlnt$ 在 $(0, +\infty)$ 中的凸性得出. 又通过对 $g_1(l) = l^k 2^{\frac{1}{2}(1-l)}$ 求导, 知

$$l^k 2^{\frac{1}{2}(1-l)} \leq k^k 2^{\frac{1}{2}(1-k)} \quad (1 \leq l \leq k) \quad (10)$$

综合 (9)(10) 二式, 知当 $\frac{e}{r}n < k < \frac{\ln 2}{2}n$ 时, 有

$$E\left(\sum_{j=1}^n a_j \mu_j\right)^{2k} \leq \sqrt{2} \left(\frac{\ln 2}{2}\right)^{\frac{e}{r}n} \cdot n^k \left(\frac{e^{\frac{1}{12}}}{\sqrt{2}}\right)^k E\left(\sum_{j=1}^n a_j^2 \mu_j^2\right)^k \leq \sqrt{2} \left(\frac{\ln 2}{2}\right)^{\frac{e}{r}n} \cdot L^k E\left(\sum_{j=1}^n \mu_j^2\right)^k$$

所以

$$J_{2n} \leq \sqrt{2} e^{t(1-\frac{n}{r})\epsilon} \left(\frac{\ln 2}{2}\right)^{\frac{e}{r}n} \sum_{\frac{e}{r}n < k < \frac{\ln 2}{2}n} \frac{t^k L^k E\left(\sum_{j=1}^n \mu_j^2\right)^k}{k!} \leq \sqrt{2} e^{t(1-\frac{n}{r})\epsilon + nt} \left(\frac{\ln 2}{2}\right)^{\frac{e}{r}n} Ee^{L t \sum_{j=1}^n (\mu_j^2 - \frac{1}{L})}$$

命 $\eta = \frac{1}{L} > 0$, 取 $0 < t < \frac{t_\eta}{L}$ 足够小, 使 $0 < \rho_1 \triangleq e^{t(2-\frac{e}{r})} \left(\frac{\ln 2}{2}\right)^{\frac{e}{r}} < 1$, 由条件 (4) 知: 存在仅依赖于 ϵ 的常数 $A_1 > 0$, 使

$$J_{2n} \leq A_1 \rho_1^n \quad (11)$$

又通过对 $g_2(l) = \left(\frac{1}{n}\right)^{\frac{1+2n}{2}} 2^{\frac{1}{2}(n-l)}$ 取导数, 可以证得: 当 $k \geq \frac{\ln 2}{2}n$ 且 $1 \leq l \leq n$ 时, 有

$\left(\frac{1}{n}\right)^k 2^{\frac{n-l}{2}} \leq 1$ 记 $\rho_2 = \frac{e^{\frac{1}{12}}}{\sqrt{2}} < 1$, 我们就有

$$l^k 2^{\frac{1}{2}(1-l)} e^{\frac{1}{12}\min(n,k)} = \sqrt{2} n^k 2^{-\frac{n}{2}} e^{\frac{1}{12}\min(n,k)} \cdot \left(\frac{1}{n}\right)^k 2^{\frac{1}{2}(n-l)} \leq \sqrt{2} n^k \rho_2^n.$$

结合 (9) 式, 立刻得到

$$J_{3n} \leq e^{t(1-\frac{n}{r})\epsilon} \sqrt{2} \rho_2^n \sum_{k > \frac{\ln 2}{2}n} \frac{(nt)^k E\left(\sum_{j=1}^n a_j^2 \mu_j^2\right)^k}{k!} \leq$$

$$e^{t(1-\frac{n}{r})} \sqrt{2} \rho_2^n \sum_{k>\frac{\ln 2}{r}n} \frac{(Lt)^k E\left(\sum_{j=1}^n \mu_j^2\right)^k}{k!} \leq \sqrt{2} \rho_2^n e^{t(1-\frac{n}{r})\epsilon+nt} E e^{Lt \sum_{j=1}^n (\mu_j^2 - \frac{1}{L})}$$

仍命 $\eta = \frac{1}{L} > 0$, 取 $0 < t < \frac{t_0}{L}$ 足够小, 使 $0 < \rho_3 \leq e^{t(2-\frac{\epsilon}{r})} \rho_2 < 1$, 由条件 (4) 得

$$J_{3n} \leq \sqrt{2} M_\eta \rho_2^n e^{t(1-\frac{n}{r})\epsilon+2tn} \leq A_2 \rho_3^n \quad (12)$$

取定 $0 < t_0 < \frac{t_0}{L}$, 使其满足 J_{2n}, J_{3n} 中的要求, 我们来估计 J_{1n} . 取 θ : $0 < \theta < 1 - \frac{e^{\frac{1}{12}}}{\sqrt{2}}$, 取 $\eta_1 = \frac{1}{L} \theta t_0$, 由条件 (4) 知: 存在 $M_{\eta_1} > 0$, $t_{\eta_1} > 0$, 使当

$$0 < \frac{\epsilon L}{r} < t_{\eta_1} = \frac{t_\theta}{L} t_0 \quad (13)$$

时, 就有

$$E e^{\frac{\epsilon L}{r} \left(\sum_{j=1}^n (\mu_j^2 - \frac{1}{L}) \right)} \leq M_{\eta_1} e^{\frac{\epsilon L}{r} \eta_1 n} = M_\theta e^{\frac{\epsilon}{r} \theta n t_0} \quad (14)$$

由于可取 $\epsilon > 0$ 充分小, 所以 (13)(14) 的成立是没有问题的. 不难证明:

$$\lim_{n \rightarrow \infty} e^{-\frac{\epsilon}{r} n} \sum_{k>\left[\frac{\epsilon}{r} n\right]+1} \frac{\left(\frac{\epsilon}{r} n\right)^k}{k!} = \frac{1}{2}$$

所以当 n 充分大, 就有 $\sum_{k>\frac{\epsilon}{r} n} \frac{\left(\frac{\epsilon}{r} n\right)^k}{k!} \geq \frac{1}{4} e^{\frac{\epsilon}{r} n}$. 于是

$$\begin{aligned} E\left(\sum_{j=1}^n a_j^2 \mu_j^2\right)^{\left[\frac{\epsilon}{r} n\right]} \frac{e^{\frac{\epsilon}{r} n}}{4} &\leq E\left(\sum_{j=1}^n a_j^2 \mu_j^2\right)^{\left[\frac{\epsilon}{r} n\right]} \sum_{k>\frac{\epsilon}{r} n} \frac{\left(\frac{\epsilon}{r} n\right)^k}{k!} \\ &\leq \sum_{k>\frac{\epsilon}{r} n} \frac{\left(\frac{\epsilon}{r} n\right)^k}{k!} \left\{ 1 + E\left(\sum_{j=1}^n a_j^2 \mu_j^2\right)^k \right\} \leq e^{\frac{\epsilon}{r} n} + E e^{\frac{\epsilon L}{r} \left(\sum_{j=1}^n (\mu_j^2 - \frac{1}{L}) \right)}, \end{aligned}$$

$$\text{从而 } E\left(\sum_{j=1}^n a_j^2 \mu_j^2\right)^{\left[\frac{\epsilon}{r} n\right]} \leq 4 + 4 E e^{\frac{\epsilon L}{r} \left(\sum_{j=1}^n (\mu_j^2 - \frac{1}{L}) \right)} \leq 4 + 4 M_\theta e^{\frac{\epsilon}{r} \theta t_0 n}. \quad (15)$$

于是, 当 $1 \leq k \leq \frac{\epsilon}{r} n$ 时, 由 (9)(10) 可知

$$\begin{aligned} E\left(\sum_{j=1}^n a_j \mu_j\right)^{2k} &\leq e^{\frac{1}{12} k^2 2^{\frac{1}{2}(1-k)}} E\left(\sum_{j=1}^n a_j^2 \mu_j^2\right)^k \leq \sqrt{2} \left(\frac{e^{\frac{1}{12}} \epsilon}{\sqrt{2} r} \right)^n n^k E\left(\sum_{j=1}^n a_j^2 \mu_j^2\right)^k \\ &\leq \sqrt{2} \left(\frac{e^{\frac{1}{12}} \epsilon}{\sqrt{2} r} \right)^k n^k \left\{ 1 + E\left(\sum_{j=1}^n a_j^2 \mu_j^2\right)^k \right\} \leq \sqrt{2} \left(\frac{e^{\frac{1}{12}} \epsilon}{\sqrt{2} r} \right)^k n^k \left\{ 5 + 4 M_\theta e^{\frac{\epsilon}{r} \theta t_0 n} \right\}, \end{aligned}$$

所以

$$\begin{aligned} J_{1n} &\leq \sqrt{2} e^{t(1-\frac{n}{r})} \left\{ 5 + 4 M_\theta e^{\frac{\epsilon}{r} \theta t_0 n} \right\} \sum_{k=0}^{\left[\frac{\epsilon}{r} n\right]} \frac{1}{k!} \left(\frac{e^{\frac{1}{12}} \epsilon n t}{\sqrt{2} r} \right)^k \\ &\leq \sqrt{2} \left\{ 5 + 4 M_\theta e^{\frac{\epsilon}{r} \theta t_0 n} \right\} e^{t(1-\frac{n}{r}) \epsilon + t n \frac{e^{\frac{1}{12}} \epsilon}{\sqrt{2} r}}. \end{aligned}$$

取 $t = t_0$, 注意到 θ 的取法, 立知 $J_{1n} \leq A_3 \rho_3^n$. 由此式结合 (11)(12) 知引理获证. 下面来

证定理本身。

众所周知,若随机变量 X 有有限方差,则其中位数 $m(X)$ 满足 $E(X) - \sqrt{2\text{Var}(X)} \leq m(X) \leq E(X) + \sqrt{2\text{Var}(X)}$ 。我们有

$$\left. \begin{aligned} E\left(\sum_{j=1}^n a_{nkj}\mu_j\right) &= \sum_{j=1}^n a_{nkj}E\mu_j = 0 \\ \text{Var}\left(\sum_{j=1}^n a_{nkj}\mu_j\right) &= \sum_{j=1}^n a_{nkj}^2\text{Var}(\mu_j) = \sigma^2 \end{aligned} \right\} \quad \text{对一切 } n, k,$$

故有 $|m\left(\sum_{j=1}^n a_{nkj}\mu_j\right)| \leq \sqrt{2\sigma}$ 。记 μ_j 的对称化随机变量为 $\tilde{\mu}_j$, 由引理及弱对称不等式知:

$$P\left\{\left|\sum_{j=1}^n a_{nkj}\mu_j - m\left(\sum_{j=1}^n a_{nkj}\mu_j\right)\right| \geq \sqrt{\sigma^2 + \frac{n-r}{r}\varepsilon}\right\} \leq 2P\left\{\left|\sum_{j=1}^n a_{nkj}\tilde{\mu}_j\right| \geq \sqrt{\sigma^2 + \frac{n-r}{r}\varepsilon}\right\} \leq A\rho^n.$$

$$\text{注意到 } m\left(\sum_{j=1}^n a_{nkj}\mu_j\right) \leq \sqrt{2\sigma}, \text{ 即知 } P\left\{\left|\sum_{j=1}^n a_{nkj}\mu_j\right| \geq \sqrt{\sigma^2 + \frac{n-r}{r}\varepsilon}\right\} \leq A\rho^n. \quad (16)$$

我们有 $\hat{\sigma}_n^2 - \sigma^2 = \frac{1}{n-r} \left\{ \sum_{j=1}^n (\mu_j^2 - \sigma^2) - \sum_{k=1}^r \left[\left(\sum_{j=1}^n a_{nkj}\mu_j \right)^2 - \sigma^2 \right] \right\}$ 。由(16)式可知

$$\begin{aligned} P\left\{\frac{1}{n-r} \left| \sum_{k=1}^r \left[\left(\sum_{j=1}^n a_{nkj}\mu_j \right)^2 - \sigma^2 \right] \right| \geq \varepsilon\right\} \\ \leq \sum_{k=1}^r P\left\{\left| \left(\sum_{j=1}^n a_{nkj}\mu_j \right)^2 - \sigma^2 \right| \geq \frac{n-r}{r}\varepsilon\right\} \leq A\rho^n. \end{aligned} \quad (17)$$

取 $0 < \eta < \varepsilon$, $0 < t < t_\eta$, 由条件(4)知

$$P\left\{\frac{1}{n-r} \sum_{j=1}^n (\mu_j^2 - \sigma^2) \geq \varepsilon\right\} \leq e^{-\frac{(n-r)t\varepsilon}{r}} Ee^{\frac{t}{r} \sum_{j=1}^n (\mu_j^2 - \sigma^2)} \leq M_\eta e^{rt\varepsilon} e^{-t(\eta-\varepsilon)n} \triangleq c_1 \tilde{\rho}_1^n. \quad (18)$$

同理可知

$$P\left\{\frac{1}{n-r} \sum_{j=1}^n (\mu_j^2 - \sigma^2) \leq -\varepsilon\right\} \leq c_2 \tilde{\rho}_2^n. \quad (19)$$

综合(17)–(19)知定理获证。

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参 考 文 献

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Exponential Convergence Rates of Error-variance Estimates in Linear Models

Sn Chun

Abstract

Suppose given a linear model $y_i = x_i' \beta + \mu_i$, $i = 1, 2, \dots$. The random errors all have a mean zero and unknown variance σ^2 , $0 < \sigma^2 < \infty$. Let $\hat{\sigma}_n^2$ be the estimate of σ^2 based on the residual sum of squares and calculated from (x_i, y_i) , $i = 1, \dots, n$. In this paper we show that if μ_1, μ_2, \dots are independent but not necessarily identically distributed, and some further conditions on $\{\mu_j\}$ and $(x_1 | \dots | x_n)$ are satisfied, then for any $\varepsilon > 0$ there exist constant ρ_ε , $0 < \rho_\varepsilon < 1$, Such that

$$P(|\hat{\sigma}_n^2 - \sigma^2| \geq \varepsilon) = O(\rho_\varepsilon^n).$$