

## THE CONSISTENCY OF NONLINEAR REGRESSION MINIMIZING THE $L_1$ -NORM

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We consider conditions in a nonlinear regression model for the consistency of the estimator obtained by minimizing the  $L_1$ -norm, i.e. the sum of absolute deviations.

**1. Introduction.** We consider the following nonlinear regression model for a univariate response  $y$

$$(1) \quad y_t = g(x_t, b_0) + e_t, \quad 1 \leq t \leq T,$$

where  $x_t \in \mathbb{R}^m$  denotes the  $t$ th fixed input vector,  $b_0 \in \mathbb{R}^p$  is the vector of parameters,  $e_t$  is the  $t$ th error residual and  $g$  is a function  $g: \mathbb{R}^{m+p} \rightarrow \mathbb{R}$ . We denote the vector of all responses by  $y = (y_1, \dots, y_T)'$ .

Given an observation  $y$ , any vector  $b$  minimizing

$$(2) \quad Q_T(y, b) = \frac{1}{T} \sum_{t=1}^T |y_t - g(x_t, b)|$$

will be called an  $L_1$ -norm estimate of  $b_0$  based on the observation vector  $y$ , and will be denoted by  $b(y)$ .

In Section 2, we establish conditions which will guarantee the existence and consistency of  $b(y)$ . Problems of  $L_1$ -norm estimation are discussed in Bassett and Koenker (1978) and Taylor (1974). In the former, the authors show that the relative efficiency of the  $L_1$ -norm estimator to the least squares estimator is the same as that of the sample median to the sample mean. Results for nonlinear least squares estimation are given by Malinvaud (1970) and Jennrich (1969). A relationship also exists between the problems treated in this paper and the so-called monotone median regression; see Cryer et al. (1972).

**2. Existence and consistency of  $L_1$ -norm estimator.** It is necessary to minimize the function (2) as a function of  $b$  over a given domain  $K \subset \mathbb{R}^p$ . In order to guarantee the existence of  $b(y)$  minimizing (2) we assume

A1:  $b_0 \in K \subset \mathbb{R}^p$  and  $K$  is compact

A2: For all  $t$ , the function  $g(x_t, b)$  is defined for all  $b \in K$  and continuous in  $K$  as a function of  $b$ .

Existence is then an immediate corollary of Lemma 2 of Jennrich (1969).

In addition to A1 and A2, we will assume

A3:  $b_0$  is an inner point of  $K$ .

Furthermore, by  $K_0 \subset K$ , we denote a closed set not containing  $b_0$ . Due to the fact that

$$(3) \quad Q_T(y, b(y)) - Q_T(y, b) \leq 0 \quad \text{for every } b \in K,$$

the (weak) consistency of  $b(y)$  ensues provided we can demonstrate that for any  $K_0$  and every  $b \in K_0$

$$(4) \quad \lim_{T \rightarrow \infty} P\{\inf_{b \in K_0} Q_T(y, b) - Q_T(y, b_0) > 0\} = 1.$$

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If we define  $h_t(b) = g(x_t, b_0) - g(x_t, b)$ , then by (1) and (2) we obtain.

$$(5) \quad Q_T(y, b) - Q_T(y, b_0) = \frac{1}{T} \sum_t \{ |e_t + h_t(b)| - |e_t| \}.$$

For the proof of (4) we require the following lemmas.

LEMMA 1. *If we assume*

A4:  $(e_t)$  is a sequence of independent random variables,

A5:  $\lim_{T \rightarrow \infty} \frac{1}{T^2} \sum_t \{h_t(b)\}^2 = 0$  for all  $b \in K$ , then we have

$$(6) \quad p \lim_{T \rightarrow \infty} \frac{1}{T} \sum_t [ |e_t + h_t(b)| - |e_t| - E \{ |e_t + h_t(b)| - |e_t| \} ] = 0.$$

PROOF. It is obvious that

$$|e_t + h_t(b)| - |e_t| \leq |h_t(b)|.$$

Therefore, the random variable in (6) has mean zero and variance bounded by

$$T^{-2} \sum_t h_t^2(b),$$

which converges to 0 as  $T \rightarrow \infty$  from the assumption. The lemma follows from Chebychev's inequality.  $\square$

LEMMA 2. *Denote the distribution function of  $e_t$  by  $F_t(x)$ . We assume*

A6:  $F_t(0) = \frac{1}{2}$ .

Define the function

$$(7) \quad G_t(z) = \begin{cases} z\{1 - 2F_t(-z)\} - 2 \int_{(0,-z]} x dF_t(x) & \text{for } z \leq 0 \\ z\{1 - 2F_t(-z)\} + 2 \int_{(-z,0]} x dF_t(x) & \text{for } z > 0. \end{cases}$$

Then, by integration of the piecewise linear function  $|e + h_t(b)| - |e|$ , we derive

$$\frac{1}{T} \sum_{t=1}^T E \{ |e_t + h_t(b)| - |e_t| \} = \frac{1}{T} \sum_{t=1}^T G_t(h_t(b)),$$

and using A6

$$\begin{aligned} \frac{1}{T} \sum_t E \{ |e_t + h_t(b)| - |e_t| \} &= 2 \frac{1}{T_{h_t}} \sum_{(b) \leq 0} \int_{(0,-h_t(b))} \{ |h_t(b)| - x \} dF_t(x) \\ &\quad + 2 \frac{1}{T_{h_t}} \sum_{(b) > 0} \int_{(-h_t(b),0)} \{ |h_t(b)| + x \} dF_t(x). \end{aligned}$$

If we restrict the domains of integration to  $(0, -h_t(b)/2]$ , resp. to  $(-h_t(b)/2, 0]$ , and again use A6, we obtain

$$\begin{aligned} &\frac{1}{T} \sum_t E \{ |e_t + h_t(b)| - |e_t| \} \\ &\geq \frac{1}{T_{h_t}} \sum_{(b) \leq 0} |h_t(b)| \left\{ F_t(-h_t(b)/2) - \frac{1}{2} \right\} + \frac{1}{T_{h_t}} \sum_{(b) > 0} |h_t(b)| \left\{ \frac{1}{2} - F_t(-h_t(b)/2) \right\} \\ &\geq \frac{1}{T} \sum_{t=1}^T |h_t(b)| \min \left\{ F_t(|h_t(b)|/2) - \frac{1}{2}, \frac{1}{2} - F_t(-|h_t(b)|/2) \right\}. \quad \square \end{aligned}$$

LEMMA 3. For all  $b, b^* \in K$  we have

$$\frac{1}{T} \sum_t \{ |e_t + h_t(b)| - |e_t + h_t(b^*)| \} \leq \frac{1}{T} \sum_t |g(x_t, b) - g(x_t, b^*)|$$

due to the inequality

$$|e_t + h_t(b)| - |e_t + h_t(b^*)| \leq |h_t(b) - h_t(b^*)| = |g(x_t, b) - g(x_t, b^*)|. \quad \square$$

We shall now prove the consistency of  $b(y)$  under the following assumptions, the first five of which imply A1 through A6.

B1:  $b_0$  is an inner point of a compact set  $K \subset \mathbb{R}^p$ .

B2: For all  $t$ , the function  $g(x_t, b)$  is defined for  $b \in K$  and is continuous in  $K$ .

B3:  $(e_t)$  is a sequence of independent random variables.

B4: The distribution function of  $e_t$  is denoted by  $F_t(x)$  and  $F_t(0) = 1/2$ .

B5: The function

$$\frac{1}{T} \sum_{t=1}^T \{g(x_t, b) - g(x_t, b^*)\}^2$$

is continuous in  $b \in K$  for every  $b^* \in K$  and uniformly for  $T$ .

B6: If we define  $h_t(b) = g(x_t, b) - g(x_t, b_0)$  and if  $K_0 \subset K$  is a closed set not containing  $b_0$ , then there exist numbers  $\varepsilon > 0$  and  $T_0$  such that for all  $T \geq T_0$

$$\inf_{K_0} \frac{1}{T} \sum_{t=1}^T |h_t(b)| \min\{F_t(|h_t(b)|/2) - \frac{1}{2}, \frac{1}{2} - F_t(-|h_t(b)|/2)\} \geq \varepsilon.$$

**THEOREM.** For the nonlinear regression model (1), under assumptions B1 through B6, the estimator  $b(y)$ , minimizing the  $L_1$ -norm of the errors, is weakly consistent.

**PROOF.** We have to prove inequality (4). We take an arbitrary  $b^* \in K_0 \subset K$ , where  $K_0$  is a closed set not containing  $b_0$ . Under B6, according to Lemma 2, there exist an  $\varepsilon > 0$  and a  $T_0$ , such that for all  $T \geq T_0$

$$\frac{1}{T} \sum_t E \{ |e_t + h_t(b^*)| - |e_t| \} \geq \varepsilon.$$

Due to B5 and Lemma 1

$$\lim_{T \rightarrow \infty} P \left[ \frac{1}{T} \sum_t \{ |e_t + h_t(b^*)| - |e_t| \} \geq \frac{\varepsilon}{2} \right] = 1.$$

Finally, using this fact with B5, Lemma 3 and the Cauchy-Schwarz inequality for all  $b \in V$ , where  $V$  is a suitable neighborhood of  $b^*$ , we obtain

$$(8) \quad \lim_{T \rightarrow \infty} P \left[ \inf_V \frac{1}{T} \sum_t \{ |e_t + h_t(b)| - |e_t| \} \geq \frac{\varepsilon}{4} \right] = 1.$$

Accordingly, a neighborhood  $V$  is associated with every  $b^* \in K_0$ . Because of the compactness of  $K_0$ , there exists a finite set of such neighborhoods  $(V_1, V_2, \dots, V_n)$  covering  $K_0$ . Therefore we have proved (8) with  $V$  replaced by  $K_0$ . The proof is thus complete.

**3. Discussion of assumptions.** Assumptions B1, B2 and B3 are also made in nonlinear least squares estimation. See Malinvaud (1970) and Jennrich (1969). The normalisation  $F_t(0) = 1/2$  (characteristic also in the monotone regression, see Cryer et al, 1972) occurs in place of the least squares regression normalisation  $Ee_t = 0$ . The Assumption B5 corresponds with Assumption 4 in Malinvaud's (1970) paper. Of paramount interest is Assumption B6. As we will demonstrate, this basically is an identifiability assumption. We obtain  $G_t(0) = 0$ . If a  $b^*$  exists, such that  $g(x_t, b^*) = g(x_t, b_0)$  for all  $t$ , then B6 is not fulfilled and  $b_0$  is not identified. In the following we seek conditions which will guarantee B6.

LEMMA 4. Assumption B6 is fulfilled if the following two assumptions hold.

B7: For every closed set  $K_0$  not containing  $b_0$ , there exist numbers  $c > 0$ ,  $d > 0$  and  $T_0 > 0$  such that for all  $b \in K_0$  and all  $T \geq T_0$

$$|\{t | t \leq T, |h_t(b)| \geq c\}|/T \geq d > 0.$$

B8: For every  $c > 0$ , there exists a real number  $f > 0$ , such that for all  $t$

$$\min[F_t(c) - 1/2, 1/2 - F_t(-c)] \geq f > 0.$$

PROOF. Given a closed set  $K_0$  not containing  $b_0$ , due to B7 we obtain numbers  $c > 0$ ,  $d > 0$  and  $T_0$ ; further according to B8 we have a  $f > 0$  such that for all  $T \geq T_0$

$$\begin{aligned} \inf_{K_0} \frac{1}{T} \sum_{t=1}^T |h_t(b)| \min\{F_t(|h_t(b)|/2) - \frac{1}{2}, \frac{1}{2} - F_t(-|h_t(b)|/2)\} \\ \geq \inf_{K_0} \frac{1}{T} \sum_{|h_t(b)| \geq c} |h_t(b)| f \geq \text{cdf} > 0. \end{aligned}$$

Therefore, Lemma 4 is proved if  $\epsilon$  is replaced by  $\text{cdf} > 0$ .

Assumption B7 occurs in place of the least squares regression assumption that, for every closed set  $K_0 \subset K$  not containing  $b_0$ , there exist numbers  $\epsilon > 0$  and  $T_0$  such that for  $T \geq T_0$

$$\inf_{K_0} \frac{1}{T} \sum_t \{h_t(b)\}^2 \geq \epsilon.$$

It can easily be shown that if B6 is fulfilled, then due to the fact that  $\min\{F_t(c) - 1/2, 1/2 - F_t(-c)\}$  is bounded above, for every closed set  $K_0 \subset K$  not containing  $b_0$  there exist numbers  $\epsilon > 0$  and  $T_0$  such that for  $T \geq T_0$

$$\inf_{K_0} \frac{1}{T} \sum_t |h_t(b)| \geq \epsilon.$$

In the linear case and for  $m = 1$ , i.e. when

$$g(x_t, b) = b_1 + b_2 x_t,$$

a sufficient condition for B7 is that for any non-empty interval  $I$  of arbitrarily small length

$$\lim_T \inf |\{t | t \leq T, x_t \text{ outside of } I\}|/T > 0.$$

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