

# Kiefer, Wolfowitz: Stochastic Estimation Of The Maximum Of A Regression Function

## 1 Introduction

$Y \in \mathbb{R}$  is a random variable which depends on a parameter  $x \in \mathbb{R}$ . Let

$$M(x) = \mathbb{E}[Y(x)]$$

and assume  $\text{Var}(Y(x)) \leq S$  ( $\forall x \in \mathbb{R}$ ).

Close form of  $M(x)$  and distribution of  $Y$  is not known. The only way to calculate  $M(x)$  is to draw random samples from  $H(y|x)$  which is the cdf of  $Y$  given  $x$ .

This paper presents the following procedure to maximize  $M(x)$  via sampling  $Y(x)$ .

$$z_{n+1} = z_n + a_n \frac{y_{2n} - y_{2n-1}}{c_n}$$

where  $z_1$  is an arbitrary number,  $y_{2n}, y_{2n-1}$  are sampled from  $H(y|z_n + c_n), H(y|z_n - c_n)$  respectively.  $a_n, c_n$  are predetermined sequences of positive numbers.

It can be proved  $z_n$  converges to the optimal solution of maximizing  $M(x)$  in probability under regularity conditions on  $M(x), a_n, c_n$ .

## 2 Regularity conditions

### 2.1 Regularity conditions on $a_n, c_n$

- $\lim_{n \rightarrow \infty} c_n = 0$ .
- $\sum_{n=1}^{\infty} a_n = \infty$ .
- $\sum_{n=1}^{\infty} a_n c_n < \infty$ .
- $\sum_{n=1}^{\infty} \frac{a_n^2}{c_n^2} < \infty$ .

### 2.2 Regularity conditions on $M(x)$

- $M(x)$  is maximized at  $x = \theta$  and  $M(x)$  strictly increasing when  $x < \theta$ , strictly decreasing when  $x > \theta$ .
- There exist  $\beta > 0$  and  $B > 0$  such that

$$|x' - \theta| + |x'' - \theta| < \beta \text{ implies } |M(x') - M(x'')| < B |x' - x''|$$

- There exist  $\rho > 0$  and  $R > 0$  such that

$$|x' - x''| < \rho \text{ implies } |M(x') - M(x'')| < R$$

- For every  $\delta > 0$  there exists  $\pi(\delta) > 0$  such that

$$|z - \theta| > \delta \text{ implies } \inf_{\frac{1}{2}\delta > \epsilon > 0} \frac{|M(z + \epsilon) - M(z - \epsilon)|}{\epsilon} > \pi(\delta)$$

## 3 Proof of convergence

Prove  $z_n \xrightarrow{P} \theta$  when regularity conditions on  $a_n, c_n, M(x)$  hold.

$$\begin{aligned}
\mathbb{E}[(z_{n+1} - \theta)^2] &= \mathbb{E}[(z_n - \theta + a_n \frac{y_{2n} - y_{2n-1}}{c_n})^2] \\
&= \mathbb{E}[(z_n - \theta)^2] + 2 \frac{a_n}{c_n} \mathbb{E}[(z_n - \theta)(y_{2n} - y_{2n-1})] + \frac{a_n^2}{c_n^2} \mathbb{E}[(y_{2n} - y_{2n-1})^2] \\
&= \mathbb{E}[(z_1 - \theta)^2] + 2 \sum_{j=1}^n \frac{a_j}{c_j} \mathbb{E}[(z_j - \theta)(y_{2j} - y_{2j-1})] + \sum_{j=1}^n \frac{a_j^2}{c_j^2} \mathbb{E}[(y_{2j} - y_{2j-1})^2]
\end{aligned}$$

Check sign of  $\mathbb{E}[(z_n - \theta)(y_{2n} - y_{2n-1})]$ .

$$\begin{aligned}
\mathbb{E}[(z_n - \theta)(y_{2n} - y_{2n-1})] &= \mathbb{E}[(z_n - \theta) \mathbb{E}[(y_{2n} - y_{2n-1})|z_n]] = \mathbb{E}[(z_n - \theta)(M(z_n + c_n) - M(z_n - c_n))] \\
&= \mathbb{E}\left[\frac{z_n - \theta}{2c_n} \cdot 2c_n \cdot (M(z_n + c_n) - M(z_n - c_n))\right]
\end{aligned}$$

$$|z_n - \theta| \geq c_n \Rightarrow (z_n - \theta)(M(z_n + c_n) - M(z_n - c_n)) \leq 0$$

Separate sign of  $\mathbb{E}[(z_n - \theta)(y_{2n} - y_{2n-1})]$ . Denote  $U_n(x) = (x - \theta)(M(x + c_n) - M(x - c_n))$ ,  $U_n^+(x) = \frac{1}{2}(U_n(x) + |U_n(x)|)$ ,  $U_n^-(x) = \frac{1}{2}(U_n(x) - |U_n(x)|)$ .

$$\mathbb{E}[(z_n - \theta)(y_{2n} - y_{2n-1})] = \mathbb{E}[U_n(z_n)] = \mathbb{E}[U_n^+(z_n) + U_n^-(z_n)]$$

Therefore

$$\begin{aligned}
\mathbb{E}[(z_{n+1} - \theta)^2] &= \mathbb{E}[(z_1 - \theta)^2] + 2 \sum_{j=1}^n \frac{a_j}{c_j} \mathbb{E}[U_j^+(z_j)] + 2 \sum_{j=1}^n \frac{a_j}{c_j} \mathbb{E}[U_j^-(z_j)] + \sum_{j=1}^n \frac{a_j^2}{c_j^2} \mathbb{E}[(y_{2j} - y_{2j-1})^2] \\
&\triangleq \mathbb{E}[(z_1 - \theta)^2] + 2 \sum_{j=1}^n \frac{a_j}{c_j} P_j + 2 \sum_{j=1}^n \frac{a_j}{c_j} N_j + \sum_{j=1}^n \frac{a_j^2}{c_j^2} e_j
\end{aligned} \tag{3.1}$$

$\sum_{j=1}^n \frac{a_j}{c_j} P_j$  and  $\sum_{j=1}^n \frac{a_j^2}{c_j^2} e_j$  converge  $\Rightarrow \lim_{n \rightarrow \infty} \mathbb{E}[(z_{n+1} - \theta)^2]$  exist. \(\blacksquare\)

Firstly, prove convergence of  $\sum_{j=1}^n \frac{a_j}{c_j} P_j$ .

$$U_n^+(z_n) > 0 \Leftrightarrow U_n(z_n) > 0 \Rightarrow |z_n - \theta| < c_n$$

When  $n$  is large enough,

$$|z_n + c_n - \theta| + |z_n - c_n - \theta| = 2c_n < \beta$$

\(\Downarrow\)

$$|M(z_n + c_n) - M(z_n - c_n)| < 2Bc_n$$

\(\Downarrow\)

$$\frac{a_n}{c_n} P_n \leq 2Ba_n c_n$$

Therefore  $\sum_{j=1}^n \frac{a_j}{c_j} P_j$  converge.

Secondly, prove convergence of  $\sum_{j=1}^n \frac{a_j^2}{c_j^2} e_j$ .

$$e_n = \mathbb{E}[(y_{2n} - y_{2n-1})^2] = \mathbb{E}\{\mathbb{E}[(y_{2n} - y_{2n-1})^2|z_n]\}$$

$$\begin{aligned}
\mathbb{E}[(y_{2n} - y_{2n-1})^2|z_n] &= \mathbb{E}[\{y_{2n} - M(z_n + c_n) - (y_{2n-1} - M(z_n - c_n)) + M(z_n + c_n) - M(z_n - c_n)\}^2|z_n] \\
&= \mathbb{E}[(y_{2n} - M(z_n + c_n))^2|z_n] + \mathbb{E}[(y_{2n-1} - M(z_n - c_n))^2|z_n] + [M(z_n + c_n) - M(z_n - c_n)]^2 \\
&\leq 2S + R^2 (\text{when } n \text{ is large enough})
\end{aligned}$$

Therefore

$$\frac{a_n^2}{c_n^2} e_n \leq (2S + R^2) \frac{a_n^2}{c_n^2}$$

which implies  $\sum_{j=1}^n \frac{a_j^2}{c_j^2} e_j$  converge.

In conclusion,  $\lim_{n \rightarrow \infty} \mathbb{E}[(z_{n+1} - \theta)^2]$  exist.

Next show there exist a subsequence  $z_{n_j}$  ( $n_1 < n_2 < n_3 < \dots$ ) of  $z_n$ ,  $z_{n_j} \xrightarrow{P} \theta$  ( $j \rightarrow \infty$ ). Let

$$K_n = \left| \frac{M(z_n + c_n) - M(z_n - c_n)}{c_n} \right|$$

Then

$$\mathbb{E}[K_n | z_n - \theta |] = \frac{P_n - N_n}{c_n}$$

Due to the convergence of  $\sum_{j=1}^n a_j \frac{P_j - N_j}{c_j}$  and divergence of  $\sum_{j=1}^n a_j$ ,

$$\liminf_{n \rightarrow \infty} \frac{P_n - N_n}{c_n} = 0$$

By the definition of limit inferior, denote  $H_n = \inf_{k > n} \frac{P_k - N_k}{c_k}$ , then  $\lim_{n \rightarrow \infty} H_n = \liminf_{n \rightarrow \infty} \frac{P_n - N_n}{c_n}$ .

Obviously,  $\lim_{n \rightarrow \infty} H_n \geq 0$  or  $\lim_{n \rightarrow \infty} H_n = \infty$ .

Assume  $\lim_{n \rightarrow \infty} H_n = \mu > 0$

$$\text{take } \epsilon = \frac{\mu}{2}, \text{ then } \exists N, \text{ when } n \geq N, H_n \geq \mu - \epsilon = \frac{\mu}{2}$$

which is

$$\frac{P_n - N_n}{c_n} > \frac{\mu}{2} (\forall n > N)$$

then

$$\sum_{j=N}^{\infty} a_j \frac{P_j - N_j}{c_j} \geq \frac{\mu}{2} \sum_{j=N}^{\infty} a_n$$

diverge, which is a contradiction.

Therefore there exists a sequence of positive integers  $n_1 < n_2 < n_3 < \dots$  such that

$$\lim_{j \rightarrow \infty} E[K_{n_j} | z_{n_j} - \theta |] = 0$$

Moreover,  $z_{n_j} \xrightarrow{P} \theta$  ( $j \rightarrow \infty$ ).

Assume  $z_{n_j} \not\xrightarrow{P} \theta$ . Then

$$\exists \epsilon, \eta > \boxed{\epsilon, \eta} \quad \mathbb{P}(|z_{n_j} - \theta| > \eta) \geq \epsilon$$

Let  $j$  be large enough,  $c_{n_j} \leq \frac{\eta}{2}$ . By regularity conditions on  $M(x)$ ,

$$K_{n_j} \geq \inf_{\frac{\eta}{2} > \epsilon > 0} \frac{|M(z_{n_j} + \epsilon) - M(z_{n_j} - \epsilon)|}{\epsilon} > \pi(\eta) \text{ when } |z_{n_j} - \theta| > \eta \text{ hold.}$$

$$E[K_{n_j} | z_{n_j} - \theta |] \geq E[K_{n_j} | z_{n_j} - \theta | I_{\{|z_{n_j} - \theta| > \eta\}}] > \eta \pi(\eta) \mathbb{E}[I_{\{|z_{n_j} - \theta| > \eta\}}] \geq \eta \pi(\eta) \epsilon > 0$$

Which is a contradiction.

Finally use the definition of converge in probability to prove  $z_n \xrightarrow{P} \theta$  which is to prove

$$\forall \epsilon > 0, \exists N, \forall n > N \mathbb{P}[|z_n - \theta| > \eta] \leq \epsilon$$

$$\begin{aligned} \mathbb{P}[|z_n - \theta| > \eta] &= \mathbb{P}[|z_n - \theta| > \eta | |z_{N_0} - \theta| < s] \mathbb{P}[|z_{N_0} - \theta| < s] + \mathbb{P}[|z_n - \theta| > \eta | |z_{N_0} - \theta| \geq s] \mathbb{P}[|z_{N_0} - \theta| \geq s] \\ &\leq \mathbb{P}[|z_n - \theta| > \eta | |z_{N_0} - \theta| < s] + \mathbb{P}[|z_{N_0} - \theta| \geq s] \end{aligned}$$

Firstly, let  $s$  be a positive number which is undetermined. Due to  $z_{n_j} \xrightarrow{P} \theta$ , choose  $N_0$  which relies on  $s, \epsilon$  to make  $\mathbb{P}[|z_{N_0} - \theta| \geq s] \leq \frac{\epsilon}{2}$ . Note that  $N_0$  is not determined totally in this step.

Next apply markov inequality to connect  $\mathbb{P}[|z_n - \theta| > \eta | |z_{N_0} - \theta| < s]$  with second moment which has been explored previously.

$$\mathbb{P}[|z_n - \theta| > \eta | |z_{N_0} - \theta| < s] \leq \frac{\mathbb{E}[(z_n - \theta)^2 | |z_{N_0} - \theta| < s]}{\eta^2}$$

$$\begin{aligned} \mathbb{E}[(z_n - \theta)^2 | z_{N_0} = z] &= \mathbb{E}[(z_{n-1} - \theta + \frac{a_{n-1}(y_{2n-2} - y_{2n-3})}{c_{n-1}})^2 | z_{N_0} = z] \\ &= \mathbb{E}[(z_{n-1} - \theta)^2 | z_{N_0} = z] + 2 \frac{a_{n-1}}{c_{n-1}} \mathbb{E}[(z_{n-1} - \theta)(y_{2n-2} - y_{2n-3}) | z_{N_0} = z] \\ &\quad + \frac{a_{n-1}^2}{c_{n-1}^2} \mathbb{E}[(y_{2n-2} - y_{2n-3})^2 | z_{N_0} = z] \\ &= (z - \theta)^2 + 2 \sum_{j=N_0}^{n-1} \frac{a_j}{c_j} \mathbb{E}[(z_j - \theta)(y_{2j} - y_{2j-1}) | z_{N_0} = z] + \sum_{j=N_0}^{n-1} \frac{a_j^2}{c_j^2} \mathbb{E}[(y_{2j} - y_{2j-1})^2 | z_{N_0} = z] \\ &= (z - \theta)^2 + 2 \sum_{j=N_0}^{n-1} \frac{a_j}{c_j} \mathbb{E}[U_j(z_j) | z_{N_0} = z] + \sum_{j=N_0}^{n-1} \frac{a_j^2}{c_j^2} \mathbb{E}[(y_{2j} - y_{2j-1})^2 | z_{N_0} = z] \\ &\leq (z - \theta)^2 + 4B \sum_{j=N_0}^{\infty} a_j c_j + (2S + R^2) \sum_{j=N_0}^{\infty} \frac{a_j^2}{c_j^2} \text{(when } N_0 \text{ is large enough)} \\ &< (z - \theta)^2 + s \text{(Let } N_0 \text{ be larger to make } 4B \sum_{j=N_0}^{\infty} a_j c_j < \frac{s}{2}, (2S + R^2) \sum_{j=N_0}^{\infty} \frac{a_j^2}{c_j^2} < \frac{s}{2}) \end{aligned}$$

Therefore

$$\mathbb{E}[(z_n - \theta)^2 | |z_{N_0} - \theta| < s] = \mathbb{E}[\mathbb{E}[(z_n - \theta)^2 | z_{N_0}] | |z_{N_0} - \theta| < s] \leq \mathbb{E}[(z_{N_0} - \theta)^2 + s | |z_{N_0} - \theta| < s] < s^2 + s$$

Choose  $s$  to make  $\frac{s^2+s}{\eta^2} \leq \frac{\epsilon}{2}$ .

Combine with  $\mathbb{P}[|z_{N_0} - \theta| \geq s] \leq \frac{\epsilon}{2}$ , then proof is done.