# INEQUALITY FOR THE VARIANCE OF AN ASYMMETRIC LOSS

### NAOYA YAMAGUCHI, YUKA YAMAGUCHI AND MAIYA HORI

(Communicated by M. Krnić)

*Abstract.* We assume that the forecast error follows a probability distribution which is symmetric and monotonically non-increasing on non-negative real numbers, and if there is a mismatch between observed and predicted value, then we suffer a loss. Under the assumptions, we solve a minimization problem with an asymmetric loss function. In addition, we give an inequality for the variance of the loss.

### 1. Introduction

Let  $\hat{y}$  be a predicted value of an observed value y. In this paper, we make the assumptions (I) and (II):

- (I) The prediction error z := ŷ − y is the realized value of a random variable Z, whose probability density function f(z) satisfies f(x) = f(-x) for x ∈ ℝ and f(x) ≥ f(y) for 0 ≤ x ≤ y.
- (II) Let  $k_1, k_2 \in \mathbb{R}_{>0}$ . If there is a mismatch between y and  $\hat{y}$ , then we suffer a loss

$$L(z) := \begin{cases} k_1 z, & z \ge 0, \\ -k_2 z, & z < 0. \end{cases}$$

Under the assumptions (I) and (II), we solve the minimization problem for the expected value of L(Z + c):

 $C = \arg\min_{c} \{ \mathbf{E}[L(Z+c)] \}.$ 

In addition, we give the following theorem.

THEOREM 1. We have

$$\mathbf{V}[L(Z+C)] \leqslant \mathbf{V}[L(Z)],$$

where equality holds only when C = 0; that is, when  $k_1 = k_2$ .

Theorem 1 is obtained by the following lemma.

© CENN, Zagreb Paper JMI-17-58

*Mathematics subject classification* (2020): 62J10, 26D15, 39B72. *Keywords and phrases*: Inequality on variance, asymmetric loss.

LEMMA 2. Suppose that a probability density function f(t) is monotonically nonincreasing on  $\mathbb{R}_{\geq 0}$  and satisfies  $\int_0^{\infty} f(t)dt = \frac{1}{2}$ . Then, for any  $x \geq 0$ , we have

$$\alpha(x) := 4 \int_0^x f(t)dt \int_x^\infty t f(t)dt - \frac{x}{2} + 2x \left(\int_0^x f(t)dt\right)^2 \ge 0.$$

If f(t) is strictly decreasing, then  $\alpha(x) > 0$  holds for x > 0. Also,  $\alpha(x) = 0$  holds for  $x \ge 0$  if and only if f(t) equals to the probability density function of a continuous uniform distribution on  $\mathbb{R}_{\ge 0}$ .

These results are a generalization of the results of [5]. The paper [5] made the assumptions (I') and (II):

(I') The prediction error  $z := \hat{y} - y$  is the realized value of a random variable Z, whose probability density function is a generalized Gaussian distribution function (see, e.g., [1], [2], and [3]) with mean zero

$$f(z) := \frac{1}{2ab\Gamma(a)} \exp\left(-\left|\frac{z}{b}\right|^{\frac{1}{a}}\right),$$

where  $\Gamma(a)$  is the gamma function and a, b > 0.

Assumption (I) is weaker than (I'). Thus, we assume a more general situation than in [5]. In [5], under the assumptions (I') and (II), the minimization problem for the expected value of L(Z + c) is solved and the inequality  $V[L(Z + C)] \leq V[L(Z)]$  is obtained. This inequality is derived from the following inequality: For a, x > 0, we have

$$x^{a}\gamma(a,x)^{2} - x^{a}\Gamma(a)^{2} + 2\gamma(a,x)\Gamma(2a,x) > 0,$$

$$(1)$$

where

$$\Gamma(a) := \int_0^{+\infty} t^{a-1} e^{-t} dt, \quad \Gamma(a,x) := \int_x^{+\infty} t^{a-1} e^{-t} dt, \quad \gamma(a,x) := \int_0^x t^{a-1} e^{-t} dt.$$

Inequality (1) is the special case of Lemma 2 that f(z) is a generalized Gaussian distribution function.

Assumptions (I) and (II) have a background in the procurement from an electricity market. Suppose that we purchase electricity  $\hat{y}$  from an market, based on a forecast of the electricity y that will be needed. This situation makes the assumption (I). If  $\hat{y} - y > 0$ , then there is a waste of procurement fee proportional to  $\hat{y} - y$ . If  $y - \hat{y} > 0$ , then we are charged with a penalty proportional to  $y - \hat{y}$ . This situation makes the assumption (II). For details, see [4].

## 2. Proof of results

For  $c \in \mathbb{R}$ , let  $\operatorname{sgn}(c) := 1$   $(c \ge 0)$ ; -1 (c < 0). From  $\int_0^{\infty} f(z)dz = \frac{1}{2}$ , the expected value of L(Z+c) and  $L(Z+c)^2$  are as follows: For any  $c \in \mathbb{R}$ ,

$$\begin{split} \mathbf{E}[L(Z+c)] &= (k_1+k_2) \int_{|c|}^{\infty} zf(z)dz + \frac{c(k_1-k_2)}{2} + |c|(k_1+k_2) \int_{0}^{|c|} f(z)dz, \\ \mathbf{E}[L(Z+c)^2] &= (k_1^2+k_2^2) \int_{0}^{\infty} z^2 f(z)dz + \operatorname{sgn}(c)(k_1^2-k_2^2) \int_{0}^{|c|} z^2 f(z)dz \\ &\quad + 2c(k_1^2-k_2^2) \int_{|c|}^{\infty} zf(z)dz + \frac{c^2(k_1^2+k_2^2)}{2} + c|c|(k_1^2-k_2^2) \int_{0}^{|c|} f(z)dz. \end{split}$$

Therefore, the expected value and the variance of L(Z) are as follows:

$$\begin{split} \mathbf{E}[L(Z)] &= (k_1 + k_2) \int_0^\infty z f(z) dz, \\ \mathbf{V}[L(Z)] &= (k_1^2 + k_2^2) \int_0^\infty z^2 f(z) dz - (k_1 + k_2)^2 \left( \int_0^\infty z f(z) dz \right)^2. \end{split}$$

We determine the value c that gives the minimum value of E[L(Z+c)]. From

$$\frac{d}{dc} \operatorname{E}[L(Z+c)] = \frac{k_1 - k_2}{2} + \operatorname{sgn}(c)(k_1 + k_2) \int_0^{|c|} f(z) dz,$$
  
$$\frac{d^2}{dc^2} \operatorname{E}[L(Z+c)] = (k_1 + k_2) f(c) \ge 0,$$

we can see that E[L(Z+c)] has the minimum value at the zero point of  $\frac{d}{dc}E[L(Z+c)]$ . The zero point *C* satisfies the following equation:

$$\frac{k_1 - k_2}{2} + \operatorname{sgn}(C)(k_1 + k_2) \int_0^{|C|} f(z) dz = 0.$$

From this, C = 0 if and only if  $k_1 = k_2$ . Also, we have

$$\begin{split} \mathbf{E}[L(Z+C)] &= (k_1+k_2) \int_{|C|}^{\infty} zf(z)dz, \\ \mathbf{V}[L(Z+C)] &= (k_1^2+k_2^2) \int_0^{\infty} z^2 f(z)dz - 2(k_1+k_2)^2 \int_0^{|C|} f(z)dz \int_0^{|C|} z(z)dz \int_0^{|C|} z(z)dz \\ &- 4|C|(k_1+k_2)^2 \int_0^{|C|} f(z)dz \int_{|C|}^{\infty} zf(z)dz + \frac{C^2(k_1+k_2)^2}{4} \\ &- (k_1+k_2)^2 \left(\int_{|C|}^{\infty} zf(z)dz\right)^2 - C^2(k_1+k_2)^2 \left(\int_0^{|C|} f(z)dz\right)^2. \end{split}$$

Let

$$\beta(x) := -\left(\int_0^\infty zf(z)dz\right)^2 + 2\int_0^x f(z)dz\int_0^x z^2 f(z)dz + 4x\int_0^x f(z)dz\int_x^\infty zf(z)dz - \frac{x^2}{4} + \left(\int_x^\infty zf(z)dz\right)^2 + x^2\left(\int_0^x f(z)dz\right)^2.$$

Then,  $V[L(Z)] - V[L(Z+C)] = (k_1 + k_2)^2 \beta(C)$  holds. From  $\beta(0) = 0$  and

$$\frac{d}{dx}\beta(x) = 4\int_0^x f(z)dz \int_x^\infty zf(z)dz - \frac{x}{2} + 2x\left(\int_0^x f(z)dz\right)^2 + 2f(x)\int_0^x z^2 f(z)dz + 2xf(x)\int_x^\infty zf(z)dz,$$

if Lemma 2 is proved, then Theorem 1 is immediately obtained. We prove Lemma 2.

*Proof of Lemma* 2. Take any  $x \ge 0$ . If f(x) = 0, then  $\alpha(x) = 0 - \frac{x}{2} + 2x \cdot \frac{1}{4} = 0$ . Below, we consider the case that f(x) > 0. Let  $\gamma := \int_0^x f(t)dt$ . For a function g = g(t) satisfying  $f(x) \ge g(t) \ge 0$  for  $x \le t$  and  $\gamma + \int_x^\infty g(t)dt = \frac{1}{2}$ , we define a functional S(g) by

$$S(g) := \int_x^\infty tg(t)dt$$

Regarding S(g) as a solid with the bottom surface area  $\int_x^{\infty} g(t)dt = \frac{1}{2} - \gamma$  (constant), we find that if we make g(t) as large as possible within the range where t is small, then S(g) become smaller. Thus, the function g that minimizes S(g) is g(t) = u(t) defined by

$$u(t) := \begin{cases} f(x), & x \leq t \leq x + \frac{1}{f(x)} \left(\frac{1}{2} - \gamma\right), \\ 0, & \text{otherwise.} \end{cases}$$

From

$$S(u) = \int_{x}^{\infty} tu(t)dt = x\left(\frac{1}{2} - \gamma\right) + \frac{1}{2f(x)}\left(\gamma^{2} - \gamma + \frac{1}{4}\right)$$

and  $\gamma \ge x f(x)$ , we have

$$\begin{aligned} \alpha(x) &\ge 4\gamma S(u) - \frac{x}{2} + 2x\gamma^2 \\ &= 4\gamma \left\{ x \left( \frac{1}{2} - \gamma \right) + \frac{1}{2f(x)} \left( \gamma^2 - \gamma + \frac{1}{4} \right) \right\} - \frac{x}{2} + 2x\gamma^2 \\ &\ge 2x\gamma - 4x\gamma^2 + 2x \left( \gamma^2 - \gamma + \frac{1}{4} \right) - \frac{x}{2} + 2x\gamma^2 \\ &= 0. \end{aligned}$$

912

Also, from this, if f(t) is strictly decreasing, then  $\alpha(x) > 0$  holds for x > 0. In addition, f(t) is the function of the form

$$f(t) = \begin{cases} \frac{1}{2a}, & 0 \leq t \leq a, \\ 0, & t > a \end{cases}$$

if and only if  $\alpha(x) = 0$  holds for  $x \ge 0$ .  $\Box$ 

#### REFERENCES

- ALEX DYTSO, RONIT BUSTIN, H. VINCENT POOR AND SHLOMO SHAMAI, Analytical properties of generalized Gaussian distributions, Journal of Statistical Distributions and Applications, 5 (1): 6, Dec 2018.
- [2] SARALEES NADARAJAH, A generalized normal distribution, Journal of Applied Statistics, 32 (7): 685–694, 2005.
- [3] TH. SUBBOTIN, On the law of frequency of error, Recueil Mathématique, 31: 296-301, 1923.
- [4] NAOYA YAMAGUCHI, MAIYA HORI AND YOSHINARI IDEGUCHI, Minimising the expectation value of the procurement cost in electricity markets based on the prediction error of energy consumption, Pac. J. Math. Ind., 10: Art 4, 16, 2018.
- [5] NAOYA YAMAGUCHI, YUKA YAMAGUCHI AND RYUEI NISHII, Minimizing the expected value of the asymmetric loss function and an inequality for the variance of the loss, Journal of Applied Statistics, 48 (13–15): 2348–2368, 2021, PMID: 35707067.

(Received December 31, 2022)

Naoya Yamaguchi Faculty of Education University of Miyazaki 1-1 Gakuen Kibanadai-nishi, Miyazaki 889-2192, Japan e-mail: n-yamaguchi@cc.miyazaki-u.ac.jp

Yuka Yamaguchi Faculty of Education University of Miyazaki 1-1 Gakuen Kibanadai-nishi, Miyazaki 889-2192, Japan e-mail: y-yamaguchi@cc.miyazaki-u.ac.jp

> Maiya Hori General Education Center Tottori University of Environmental Studies 1-1-1 Wakabadai-kita, Tottori, 689-1111 Japan e-mail: m-hori@kankyo-u.ac.jp