

A Note on Fuzzy Linear Regression Analysis of Fuzzy Valued Variables

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Abstract

In this note, we show that a linear regression model, using entropy and degree of nearness of fuzzy numbers, suggested by Wang and Li[FSS 36, 125-136] seems to be unreasonable by an example.

Key Words and Phrases: Estimation of parameter; Fuzzy linear regression.

1. Introduction

In general, there are two possible cases in systems in which human intelligence participates : (1) The relations of the variables are subject to fuzziness, and (2) the variables themselves are fuzzy. Wang and Li[2] constructed two different fuzzy linear regression models for the second case, based on possibility theory[3]. But, a linear regression model(\tilde{E} -D estimation method) using entropy and degree of nearness of fuzzy numbers does not make sense. We will show this by an example.

To begin with, we introduce the \tilde{E} -D estimation method. For convenience, we use the same notations as they used.

Let X_0, X_1, \dots, X_n be variables which take values in $F(R)$ (we call them fuzzy variables) and suppose there exists a linear relation among them:

$$X_0 = \beta_1 X_1 + \dots + \beta_{n-1} X_{n-1} + \beta_n X_n,$$

where $\beta_i, i = 1, \dots, n$, are unknown real coefficients, and

$$\begin{aligned} X_n(x) &= \begin{cases} 1, & x = 1, \\ 0 & x \neq 1 \end{cases} \\ &\equiv I_1. \end{aligned}$$

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Generally, let X_0, X_1, \dots, X_n be fuzzy variables and take values in $\mathcal{L}_{\bar{\mu}}$. We observe these fuzzy variables m times, and in the j -th observation ($j = 1, \dots, m$), we obtain the sample values x_{ijk} ($k = 1, \dots, s(i, j)$) of X_i ($i = 0, 1, \dots, n$). We call the data $\{x_{ijk} | i = 0, 1, \dots, n; j = 1, \dots, m; k = 1, \dots, s(i, j)\}$ initial data. Then we determine a $\bar{\mu}$ -type fuzzy number X_{ij}^* , the j -th fuzzy observed value of X_i , $i = 0, \dots, n, j = 1, \dots, m$, using the method of maximum μ/E estimation. Thus we obtain fuzzy observed values $(X_{0j}^*, X_{1j}^*, \dots, X_{nj}^*)$ of (X_0, X_1, \dots, X_n) , $j = 1, \dots, m$.

Let $\tilde{E} = \sum_{j=1}^m |E(X_{0j}^*) - E(\hat{X}_{0j}^*)|$, where $E(X_{0j}^*)$, $E(\hat{X}_{0j}^*)$ denote the entropy of the j -th fuzzy observed value, and the j -th fuzzy estimate value of X_0 , respectively. If β_1, \dots, β_n have a change, it can lead to a change of \tilde{E} and affect the degree of nearness between the fuzzy observed values and the fuzzy estimate values. Owing to these two aspects, we advance the following principle:

Choose $\beta_1^*, \dots, \beta_n^*$, such that

$$\sum_{j=1}^m |E(X_{0j}^*) - E(\hat{X}_{0j}^*)| = \min_{\beta_1, \dots, \beta_n} \sum_{j=1}^m |E(X_{0j}^*) - E(\bar{X}_{0j}^*)|$$

subject to

$$\bigwedge_{j=1}^m D(X_{0j}^*, \hat{X}_{0j}^*) \geq h, \quad j = 1, \dots, m,$$

where D is the degree of nearness of fuzzy numbers, h is some given standard of nearness ($h \in (0, 1]$), $\hat{X}_{0j}^* = \sum_{i=1}^n \beta_i^* X_{ij}^*$, $j = 1, \dots, m$.

$\beta_1^*, \dots, \beta_n^*$ are called \tilde{E} - D estimation of β_1, \dots, β_n . This method is called \tilde{E} - D estimation method.

Example. We define a fuzzy number H as follow:

$$H(x) = \begin{cases} 1 - \frac{1}{2}|x| & \text{if } x \in [-1, 1], \\ \frac{1}{2} & \text{if } |x| \in [1, 5], \\ 0 & \text{otherwise.} \end{cases}$$

Let $\mathcal{L}_{\bar{\mu}} = \mathcal{L}_H$ and consider the following model

$$X_0 = \beta_1 X_1 + \beta_2 X_2$$

with standard of nearness $h = 0.5$. Now we consider the data, $(X_{0j}^*, X_{ij}^*) = (H(10 - j, 1), H(j, 1))$, $j = 1, 2, \dots, 10$. Then we can easily see that $\beta_1^* = -1, \beta_2^* = 10$ is a reasonable possible coefficients to fit the data. But $\beta_1^* = 1, \beta_2^* = 0$ is another possible answer, since $E(X_{0j}^*) = E(\bar{X}_{0j}^*) = 5.5$ for all $j = 1, 2, \dots, 10$ and $D(X_{0j}^*, \bar{X}_{0j}^*) = 0.5$ for $j = 1, 2, \dots, 10$. As we can see from this fact, the slope of one possible answer is 1 and that of another possible answer is -1 . So this model seems to be unreasonable.

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