

## Estimation of the Actual Working Time by Interval Linear Regression Models with Constraint Conditions

S.G. Hwang\* and Y.J. Seo\*\*

제약부 구간 선형 회귀모델에 의한 실동시간의 견적

황승국\* · 서유진\*\*

### Abstract

The actual working time of jobs, in general, is different to the standard time of jobs. In this paper, in order to analyze the actual working time of each job in production, we use the total production amount and the necessary total working time. The method which analyzes the actual working time is as follows. In this paper, we propose the interval regression analysis for obtaining an interval linear regression model with constraint conditions with respect to interval parameters. The merits of this method are the following: 1) it is easy to obtain an interval linear model by solving a LP problem to which the formulation of proposed regression analysis is reduced, 2) it is easy to add constraint conditions about interval parameters, which are a sort of expert knowledge. As an application, within a case which has given certain data, the actual working time of jobs and the number of workers in a future plan are estimated through the real data obtained from the operation of processing line in a heavy industry company. It results from the proposed method that the actual working time and the number of workers can be estimated as intervals by the interval regression model.

---

\* Department of Industrial Engineering, University of Osaka Prefecture, Japan

\*\* Department of Industrial Engineering, University of Kyung Nam

## 1. Introduction

The objective of time study is to determine reliable standards for all works[1]. Time study is used to measure work, and its result is called the standard time for the work[2]. The standard time is the time being necessary to bring a work to completion using standard job methods and conditions[3]. However, in actual, the standard time of jobs is not equal to the actual working time of jobs by means of variation of working environment and conditions etc..

In order to analyze the actual working time, we use the following novel linear regression analysis. As the method of regression analysis based on the concept of possibility[4], the possibilistic linear regression model with fuzzy parameters, has been proposed by Tanaka et al.[5] and discussed in detail in references[6-13]. In the conventional regression analysis, deviations between the observed values and the estimated values are supposed to be due to observation errors. We assume, on the contrary, that these deviations depend on the fuzziness of the system structure in possibilistic regression models. Thus, the parameters in the regression models are assumed to be intervals. Tanaka et al.[14] has proposed the interval linear regression analysis with interval parameters instead

of fuzzy parameters[15] and has been researched[16].

Chisman[17] has applied LP to establish the standard time. The objective is to find the standard time which minimizes the sum of the absolute deviation between the actual total time for a job and calculated total time for that job.

In this paper, we propose the regression method for obtaining an interval linear regression model with constraint conditions with respect to interval parameters, which are a sort of expert knowledge. In actual problems, information about the parameters of the models are often obtained in a vague expression. Thus, this expert knowledge with respect to parameters is transferred to interval constraint conditions within which the estimated parameters will be found as intervals. Then, we try to obtain the interval parameters which are compatible with these constraint conditions. We obtain the upper and lower limits of the actual working time instead of a real value. In order to do this, let us define an interval linear regression model as follows:

$$Y = A_1x_1 + \dots + A_nx_n \dots\dots\dots (1)$$

where  $A_1$  is an interval parameter representing the actual working time of  $i$ th work. The rough knowledge about the actual working time of work are obtained from the factory manager so as to enable us to estimate a reasonable interval. This is

used as constraint conditions of the LP problem. Since we introduce the interval constraint conditions proposed by the expert, we obtain a more actual and reasonable analysis than before. Using the estimated actual working time by an established model, we estimate the number of workers being necessary in a week. Also, it is able to deal with the change of the amount of work.

The merits of this method are the following. First, it is easy to obtain the interval linear model by solving a LP problem to which the regression problem is reduced. Second, it is easy to add constraint conditions about interval parameters, which are a sort of expert knowledge.

As an application, within a case which has given certain data, we estimate the actual working time of jobs using the data obtained from the operation of a processing line from January to April 1987 in a heavy industry company.

Furthermore, the number of workers in a future plan are estimated by using the estimated actual working time. All results are obtained as intervals which present our partial ignorance in a vague situation. This method reflects our partial ignorance through intervals and is fitting to analyze vague phenomena like the operation of a processing line[18].

## 2. Interval Linear Regression Models with Constraint Conditions

Let us define an interval linear regression model

$$Y = A_1x_1 + \dots + A_mx_m \dots\dots\dots (2)$$

where  $A_j$  is an interval parameter. Using the interval parameter  $A_j$  with the center  $\alpha_j$  and the spread  $c_j$ , denoted as  $A_j = (\alpha_j, c_j) = \{a \mid \alpha_j - c_j \leq a \leq \alpha_j + c_j\}$ , Eq.(2) is transformed from interval arithmetic[19,20] as follows:

$$Y = (\alpha x, c \mid x \mid) \dots\dots\dots (3)$$

where  $\alpha = (\alpha_1, \dots, \alpha_n)$ ,  $x = (x_1, \dots, x_n)^t$ ,  $c = (c_1, \dots, c_n)$ ,  $\mid x \mid = (\mid x_1 \mid, \dots, \mid x_n \mid)^t$ .

Let us define inclusion relation between intervals as follows:

$$\begin{matrix} A_i \subset A_j \\ \uparrow \downarrow \\ \alpha_j - c_j \leq \alpha_i - c_i, \quad \alpha_j + c_j \geq \alpha_i + c_i \dots\dots\dots (4) \end{matrix}$$

Now, we assume that we can obtain observation data,  $(y_i, x_i)$ ,  $i=1, \dots, n$ . An interval linear regression problem data is formulated under the following assumptions.

1) Output value  $y_i$  is included in the estimated interval  $Y(x_i)$

$$Y(x_i) = A_1x_{i1} + \dots + A_nx_{in} \supset y_i, \quad i=1, \dots, n, \dots\dots\dots (5)$$

Using Eq.(4), this condition is transformed as follows:

$$\begin{aligned}
 y_i &\leq \alpha x_i + c |x_i|, \quad i=1, \dots, n, \\
 y_i &\geq \alpha x_i - c |x_i|, \quad i=1, \dots, n, \dots \dots \dots (6)
 \end{aligned}$$

2) Let us introduce the following performance index

$$J(c) = \sum_i c |x_i| \dots \dots \dots (7)$$

where  $c |x_i|$  is the spread of the estimated interval  $Y(x_i)$ . Thus,  $J(c)$  is the total amount of the spread of  $Y(x_i)$  which corresponds to the total amount of errors in the conventional regression model.

From the above assumptions, the problem is formulated as the following linear programming(LP) problem.

[Problem LP 1]

$$\min_{\alpha, c} J(c) = \sum_i c |x_i|,$$

subject to (6) and  $c \geq 0$ .

In many actual cases, we have to obtain beforehand the rough information about parameters from the expert, which serves constraint conditions. This can be transferred into the following inclusion relation:

$$A_j \subset B_j, \quad j=1, \dots, n \dots \dots \dots (8)$$

where  $B_j(\beta_j, d_j)$  is an interval given by the expert. Eq.(8) leads to

$$\begin{aligned}
 \beta_j - d_j &\leq \alpha_j - c_j, \quad j=1, \dots, n \\
 \beta_j + d_j &\geq \alpha_j + c_j, \quad j=1, \dots, n \dots \dots \dots (9)
 \end{aligned}$$

where  $\beta_j - d_j$  and  $\beta_j + d_j$  are the lower value and the upper value of the interval  $B_j$ , respectively. The linear regression problem

with constraint conditions is then reduced to the following LP problem based on the constraint conditions of Eqs.(6) and (9).

[Problem LP 2]

$$\min_{\alpha, c} J(c) = \sum_i c |x_i|,$$

subject to (6), (9) and  $c \geq 0$ .

### 3. Application to Measurement of Processing Work

The object of this section is to analyze the actual working time by the proposed interval regression model.

#### 3.1 The Actual Working Time

Let us estimate the actual working time of the processing work. The input-output data are shown in Table 1. Table 2 represents the products, amount of products, the

Table 1. Input-output data

Output	Input							
$y_1$	$x_{11}$	..	..	..	..	..	..	$x_{1n}$
$y_1$	$x_{11}$	..	..	..	..	..	..	$x_{1n}$
.	.							.
.	.							.
.	.							.
$y_N$	$x_{N1}$	..	..	..	..	..	..	$x_{Nn}$

Table 2. Input-output data of processing work

unit: /week

Week no.	Working day	Total working time (y)	Workers	Products					
				A x <sub>1</sub>	B x <sub>2</sub>	C x <sub>3</sub>	D x <sub>4</sub>	E x <sub>5</sub>	F x <sub>6</sub>
1	6	11514	193	1.3	2.9	2.6	2.3	1.8	2.1
2	6	11765	189	1.2	3.0	2.9	2.4	1.6	2.4
3	6	11434	184	1.3	2.7	2.3	2.2	1.9	1.8
4	5	9931	189	1.2	2.4	2.2	2.1	1.7	1.7
5	6	11493	189	4.4	3.1	3.2	2.5	2.3	2.2
6	6	11029	189	4.2	2.7	3.4	2.7	2.1	2.4
7	6	11546	191	4.3	3.0	2.9	2.1	2.1	2.3
8	6	11462	190	4.6	3.1	4.0	2.7	2.8	2.1
9	6	12395	197	3.7	3.7	2.6	3.2	3.3	2.6
10	5	10572	199	2.9	2.9	2.4	2.7	2.5	2.2
11	6	12168	199	3.3	3.6	2.9	3.5	3.6	2.3
12	6	12279	197	3.6	3.9	2.3	3.1	3.0	2.9
13	6	11462	196	3.2	3.8	3.1	3.1	2.8	2.5
14	6	11450	198	3.1	3.6	3.1	3.2	2.7	2.7
15	6	11718	198	3.4	4.1	3.3	2.9	3.0	2.3
16	6	11786	199	3.3	3.8	3.2	2.9	2.7	2.5
17	6	9921	199	1.9	3.0	2.5	3.3	4.0	2.7
18	6	10083	200	2.1	3.1	2.2	3.3	3.7	2.7
19	6	10029	200	1.9	3.1	2.6	3.2	4.1	2.8
20	6	9931	200	3.1	2.8	2.8	3.7	4.5	2.7
21	6	11369	200	2.0	3.7	3.1	3.0	3.9	2.5

number of workers, total working time and total working days per week by a processing work team in 21 weeks ranging from January to April 1987 in a heavy industry. The data from the first week to the 16th week are used to obtain the model, five weeks' data ranging from the 17th week to the 21th week are used as data to check the efficiency of the model.

The total actual working time  $Y(x_i)$  is

estimated by

$$Y(x_i) = A_1x_{i1} + \dots + A_6x_{i6} \dots\dots\dots (10)$$

where the amount of production of products A, B, C, D, E and F are denoted by  $x_1, x_2, x_3, x_4, x_5$  and  $x_6$ , respectively. The information for setting the interval parameter  $B_j$  from the factory manager is given in Table 3.

Next, the estimate of the actual working time is obtained by solving LP 2, using

**Table 3. Constraint conditions of interval parameter**

Products	unit: hours	
	Lower value	Upper value
A	$\alpha_1 - c_1 \geq 443$	$\alpha_1 + c_1 \leq 1700$
B	$\alpha_2 - c_2 \geq 1039$	$\alpha_2 + c_2 \leq 1725$
C	$\alpha_3 - c_3 \geq 799$	$\alpha_3 + c_3 \leq 1554$
D	$\alpha_4 - c_4 \geq 275$	$\alpha_4 + c_4 \leq 400$
E	$\alpha_5 - c_5 \geq 106$	$\alpha_5 + c_5 \leq 337$
F	$\alpha_6 - c_6 \geq 150$	$\alpha_6 + c_6 \leq 354$

input-output data in Table 2 and constraint conditions about the interval parameters in Table 3. Using the estimated interval parameters, Eq.(10) is transformed as follow:

$$\begin{aligned}
 Y(x_i) &= (623, 180)x_{11} + (1429, 296)x_{12} + \\
 &\quad (1177, 378)x_{13} + (400, 0)x_{14} + \\
 &\quad (328, 9)x_{15} + (354, 0)x_{16} \\
 &= [443, 803]x_{11} + [1133, 1725]x_{12} + \\
 &\quad [799, 1554]x_{13} + [400, 400]x_{14} + \\
 &\quad [319, 337]x_{15} + [354, 354]x_{16}
 \end{aligned}$$

We try to examine the inclusion relation between these estimated interval parameters and constraint conditions about the interval parameter in Table 3. The actual working time appears in the left side of the constraint conditions. Thus, we can see that product A is located to the left side, but the allowance of spread is largely to the right side. Same as above, product B is placed to the right side, product C will become completely coincide with the constraint conditions, and products D, E and F

are shown in the right side. In particular, the spread is 0 in products D and F. It doesn't mean that the spread of products D and F is not needed at all, but that the relation between products D and F and the total actual working time is crisply approximated. By Eq.(11), the estimated interval value of processing work is denoted in Fig. 1. The checking data, not used in establishing the model, are from the 17th week to the 21th week. From this figure, we can see that the total actual working time and estimated actual working time are very reasonable. Only in the estimated working time of 20th week, the estimated actual working time is not included in the constraint conditions.

### 3.2 The Estimation of the Number of Workers

From establishing the actual working time of a work, the number of workers, that is, the fixed personnel, is determined. The fixed personnel means the reasonable personnel. The determination of the fixed personnel means to determine the necessary personnel in each work's accomplishment based on constant conditions[21]. The determination of the number of workers is done. That is, when the number of products is given as  $x_1, \dots, x_n$  produced by a work team, and the interval actual working times  $A_1, \dots, A_n$  are already obtained,

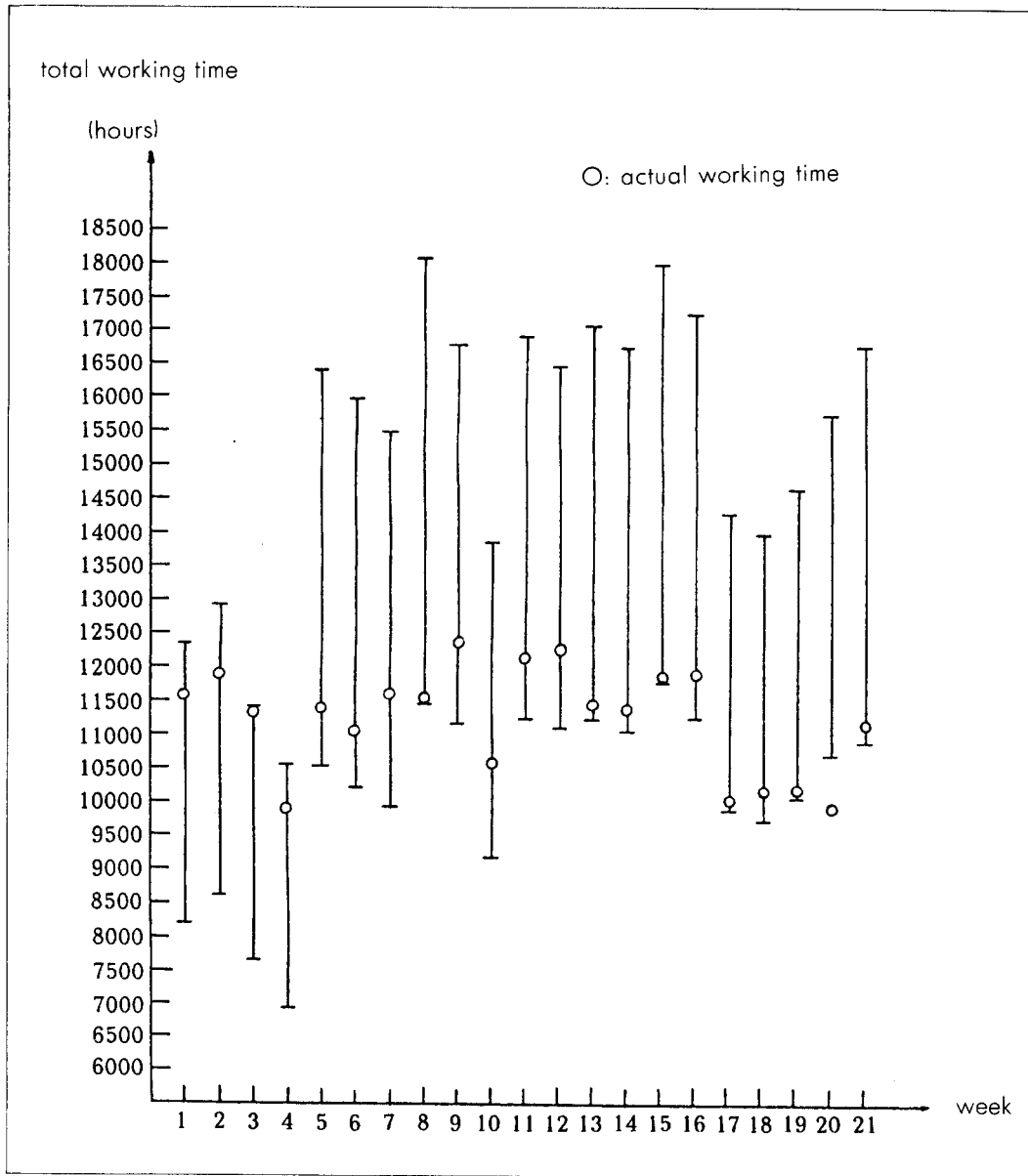


Fig. 1 The results of a model in a processing work.

then the number of necessary workers is determined as follows:

$$N = \frac{(\sum_j A_j x_j)}{T} \dots\dots\dots (12)$$

where T denotes the total actual working time per worker within a fixed period. In many cases, T should be represented by an interval because of its fuzziness. Therefore,

the number of workers is also obtained as an interval with a spread, because Eq. (10) is given by the interval actual working time and the interval working time. Eq.(12) is computed using the following interval arithmetic. That is, in general, when  $0 \notin V$ , it is expressed as follows:

$$\frac{U}{V} = \frac{(u, d)}{(v, k)} = \frac{[u-d, u+d]}{[v-k, v+k]} = \left[ \frac{u-d}{v+k}, \frac{u+d}{v-k} \right]$$

..... (13)

From the result of the computation of Eq. (12) and Table 4, we can obtain the number of necessary workers  $N(x_i)$  in the  $i$ th week.

$$N(x_i) = \{ (623, 180)x_{i1} + (1429, 296)x_{i2} + (1177, 378)x_{i3} + (400, 0)x_{i4} + (328, 9)x_{i5} + (354, 0)x_{i6} \} / T_i$$

$$= \{ [443, 803]x_{i1} + [1133, 1725]x_{i2} + [799, 1554]x_{i3} + [400, 400]x_{i4} + [319, 337]x_{i5} + [354, 354]x_{i6} \} / T_i \quad (14)$$

Table 4. Estimated interval paramter

unit: hours			
Products	Inerval parameter	Center	Spread
A	A <sub>1</sub>	623	180
B	A <sub>2</sub>	1429	296
C	A <sub>3</sub>	1177	378
D	A <sub>4</sub>	400	0
E	A <sub>5</sub>	328	9
F	A <sub>6</sub>	354	0

where  $T_i$  denotes the total actual working time for the  $i$ th week per worker. Fig. 2 represents the relation between the number of workers obtained by Eq.(14) and the number of actual workers. From this figure, we can see that the model of Eq.(14) adopts the checking data(from the week 17 to week 21). The number of estimated workers of 20th week does not include the number of actual workers. The number of

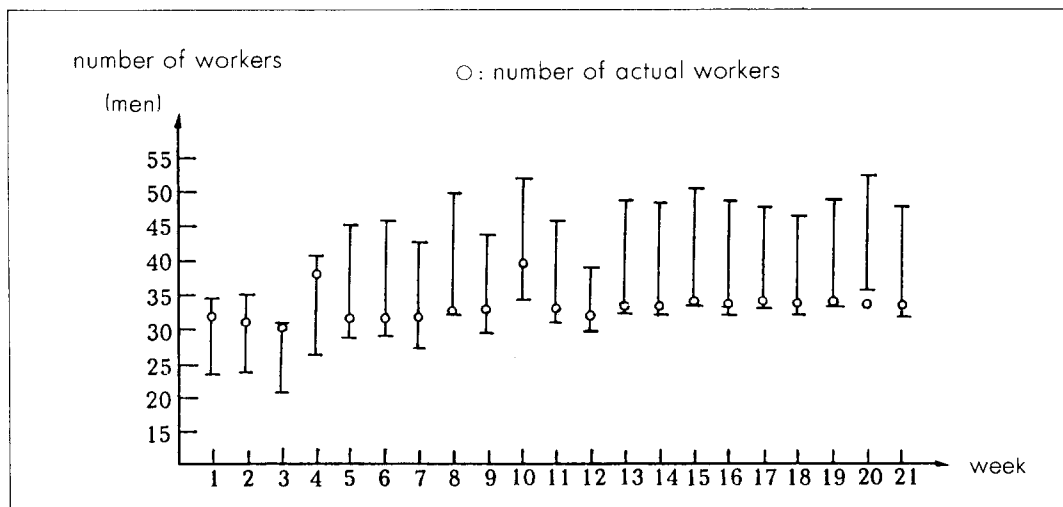


Fig. 2 The number of estimated workers in a processing work team



workers in this figure shows the average, which is the number of workers of a daily processing team. This value is obtained through dividing the number of total actual workers by the number of working days in a week. From estimation of the number of workers, which has been done here, we can make an evaluation of the number of present workers such as the following.

The number of workers is given by the interval shown in Fig. 2. It denotes the number of possible workers. If the number of workers is placed in the left side of the interval, the work is done by a comparatively few number of workers. Then we can think of a small allowance of work. If the number of workers is shown to right side of the interval, we can think that there is a certain allowance of work, because the work is done by a comparatively large number of workers. The number of necessary workers for the amount of future work can be estimated from Eq.(14). For example, when values such as  $x_{11}=1.0$ ,  $x_{12}=3.0$ ,  $x_{13}=2.0$ ,  $x_{14}=1.5$ ,  $x_{15}=1.8$ ,  $x_{16}=2.2$ ,  $T_1=50$  are given, and the working days are six, we can estimate the number of workers in one day as the interval[24.6, 36.9].

#### 4. Conclusions

In this paper, we estimated the actual working time by using an interval linear

regression analysis with constraint conditions. Also, the number of workers in a future plan is estimated by using the estimated actual working time. The interval linear regression analysis with constraint conditions is the method in which the information of the expert is accepted. The estimated value is denoted by an interval. In actual situations, we must consider not only the information which is obtained about parameters, but also parameters which are obtained by the result of regression analysis. We can obtain parameters for actual application, by means of adding constraint conditions into parameters.

#### References

- [1] Niebel, B.W., Time Standard, *Handbook of Industrial Engineering*, Ch. 4.4, Wiley & Sons, 4.4.1-4.4.37, 1982.
- [2] Barnes, R.M., *Motion and Time Study Design and Measurement of work*, 7th ed., John Wiley & Sons, 1980.
- [3] Senjyu, S., *Work Study*, Japanese Standard Association, Tokyo, 1980(in Japanese).
- [4] Zadeh, L.A., "Fuzzy Sets as a Basis for a Theory of Possibility", *Fuzzy Sets and Systems*, Vol. 1, No. 1, pp.3-28, 1978.
- [5] Tanaka, H., Uejima, S. and Asai, K., "A Linear Regression Model with Fuzzy Functions", *J. of the Operations Research Society of Japan*, Vol. 25, No. 2, pp.162-174,

1982.

[6] Tanaka, H., Uejima, S. and Asai, K., "Linear Regression Analysis with Fuzzy Model", *IEEE Trans.*, SMC-12, pp.903-907, 1982.

[7] Tanaka, H. and Asai, K., "Fuzzy Solution in Fuzzy Linear Programming Problems", *IEEE Trans.*, SMC-14, pp.325-328, 1984.

[8] Tanaka, H., "Possibilistic Model and its Applications", *Systems and Control*, Vol. 28, No. 7, pp.447-451, 1984(in Japanese).

[9] Tanaka, H., Simomura, T. and Watada, J., "Identification of Learning Curve Based on Fuzzy Regression Model", *The Japanese Journal of Ergonomics*, Vol. 22, No. 5, pp.253-258, 1986(in Japanese).

[10] Tanaka, H., Watada, J., Hayashi, I. and Asai, K., "The Formulation of the Fuzzy GMDH", *Systems and Control*, Vol. 30, No. 9, pp.581-587, 1986(in Japanese).

[11] Tanaka, H., Watada, J. and Hayashi, I., "On Three Formulations of Fuzzy Linear Regression Analysis", *Trans. of the Society of Instrument and Control Engineers*, Vol. 22, No. 10, pp.45-51, 1986(in Japanese).

[12] Tanaka, H., "Fuzzy Data Analysis by Possibilistic Linear Models", *Fuzzy Sets and Systems*, Vol. 24, No. 3, pp.363-375, 1987.

[13] Seo, Y.J., Hwang, S.G., Hayashi, I. and Tanaka, H., "Institution of Standard

Time by Possibilistic Linear Regression with Restriction", *APORS' 88*, Seoul, Korea on August, 24-26, pp.348-351, 1988.

[14] Tanaka, H., Hayashi, I. and Watada, J., "Interval Regression Analysis", *3rd Fuzzy System Symposium*, pp.9-12, 1987(in Japanese).

[15] Dubois, D. and Prade, H., *Fuzzy Sets and Systems, Theory and Application*, Academic Press, 1980.

[16] Hwang, S.G., Hayashi, I. and Tanaka, H., "Analysis of Human Judgement by Interval", *IEA 10th International Congress Proceedings*, Sydney Australia on August 1-5, pp.484-486, 1988.

[17] Chisman, J.A., "Using Linear Programming to Determine Time Standards", *The Journal of Industrial Engineering*, Vol. 17, No. 4, pp.189-191, 1966.

[18] Shafer, G., *A Mathematical Theory of Evidence*, Princeton Univ. Press, 1976.

[19] Alefeld, G. and Herzberger, J., *Introduction to Interval Computations*(translated by J. Rokne), Academic Press, New York, 1983.

[20] Moore, R.E., *Methods and Applications of Interval Analysis*, SIAM, Philadelphia, 1979.

[21] Konou, H., *Measurement of Amount of Office Work and Actual Affairs of the Fixed Personnel Estimation*, Japan Management Press, Tokyo, 1976(in Japanese).