



# Nonlinear Optimization of GM(1,1) Model Based on Multi-parameter Background Value

Tangsen Zhan, Hongyan Xu

## ► To cite this version:

Tangsen Zhan, Hongyan Xu. Nonlinear Optimization of GM(1,1) Model Based on Multi-parameter Background Value. 5th Computer and Computing Technologies in Agriculture (CCTA), Oct 2011, Beijing, China. pp.15-19, 10.1007/978-3-642-27275-2\_2 . hal-01361113

**HAL Id: hal-01361113**

**<https://inria.hal.science/hal-01361113>**

Submitted on 6 Sep 2016

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.



Distributed under a Creative Commons Attribution 4.0 International License

# Nonlinear Optimization of GM(1,1)Model based on Multi-Parameter Background Value

Tangsen-ZHAN Hongyan-XU

( School of Information Engineering,Jingdezhen Ceramic Institute, Jingdezhen, Jiangxi, China,  
e-mail:ztangsen@yahoo.com.cn)

**Abstract.** By studying the existing algorithms for background value in GM(1,1), a nonlinear optimization model of GM(1,1) based on multi-parameter background value is given. The paper uses the invertible matrix of the parameter to optimize and estimate the parameters  $\hat{a}$ ; in addition, the parameter estimate  $\hat{a}$  obtained from the multi-parameter background value has higher prediction accuracy, thus overcoming the restriction on the prediction based on the fixed average background value in other literatures. the simulated values obtained by the optimized model (NOGM(1,1)) are more precise, and the maximum error is reduced by 15%. The nonlinear optimization model of GM(1,1) based on multi-parameter background value provides algorithms for further study of GM(1,1) model.

**Keywords.** background value; parameter; nonlinear optimization; ratio

## 1 INTRODUCTION

The grey model GM(1,1) is a forecast model in grey system. It is important to the application of grey system theory. Successful examples of using GM(1,1) for prediction can be found in many fields [1-2]. Much research has been done on this model, and a number of methods have been proposed to improve it [3-9]. Fundamentally, the effect of the change of the background value on the fitted value has been analyzed. However, when solving GM(1,1), the area is partitioned and the background value is obtained through approximate iteration, or the solution can be obtained through a parameter expression derived from a special background value model.

As for the background value, the  $(1 - \delta : \delta)$  ratio established between the background value and the two endpoint prototype values restricts the expression and estimation of grey differential equation, which causes the estimated parameters to produce a greater prediction error. The model solution is probably imprecise. This

paper discusses the GM(1,1) model of the background value with two or more parameters obtained through the extension of the background value, the nonlinear optimization of the analog value with parameters called the nonlinear optimization model of GM(1,1) based on the background value with multi-parameters (NOGM(1,1)).

## 2 THE NONLINEAR OPTIMIZATION MODEL OF GM(1,1) BASED ON THE MULTI-PARAMETER BACKGROUND VALUE

### 2.1 The grey model GM(1,1) of background value with two parameters

Denote the original sequence by  $X^{(0)} = (X_{(1)}^{(0)}, X_{(2)}^{(0)}, \dots, X_{(N)}^{(0)})$ , calculate an AGO

$X_{(k)}^{(1)} = \sum_{i=1}^k X_{(i)}^{(0)}$ , get the new sequence  $X^{(1)} = (X_{(1)}^{(1)}, X_{(2)}^{(1)}, \dots, X_{(N)}^{(1)})$ , albinism

differential form of GM(1,1) is  $\frac{dX^{(1)}}{dt} + aX^{(1)} = b$  (1)

Difference form of GM(1,1) is  $X_{(k)}^{(0)} + a * z_{(k)}^{(1)} = b$  (k=2, 3, 4, ..., N) (2)

There is  $z_{(k+1)}^{(1)} = r_k * X_{(k)}^{(1)} + s_k * X_{(k+1)}^{(1)}$ , solving (2) and getting parameter

$$\hat{a} = [a, b]^T = (B^T B)^{-1} B^T Y_N \quad (3)$$

Where  $Y_N = (X_{(2)}^{(0)}, X_{(3)}^{(0)}, \dots, X_{(N)}^{(0)})$ ,

$B = \begin{bmatrix} -z_{(2)}^{(1)} & -z_{(3)}^{(1)} & \dots & -z_{(N)}^{(1)} \\ 1 & 1 & \dots & 1 \end{bmatrix}^T$ , the model of the background value

$z_{(k+1)}^{(1)}$  with parameters  $r_k$  and  $s_k$  is obtained, and it is called the grey model

GM(1,1) of the background value with two parameters.

## 2.2 The nonlinear optimization model of GM(1,1) for the multi-parameter background value

In order to achieve parameter estimation based on the multi-parameter background value, inverse matrix must be directly expressed by easy other matrix. The following is the process to get the parameter estimate through invertible matrix.

Theorem 1: If  $F = \sum_{k=2}^N (z_{(k)}^{(1)})^2$ , and  $C = \sum_{k=2}^N z_{(k)}^{(1)}$ ,

$$(B^T B)^{-1} = \frac{1}{(N-1)F - C^2} \begin{bmatrix} N-1 & C \\ C & F \end{bmatrix}$$

Proof: Because  $B^T B = \begin{bmatrix} \sum_{k=2}^N (z_{(k)}^{(1)})^2 & -\sum_{k=2}^N z_{(k)}^{(1)} \\ -\sum_{k=2}^N z_{(k)}^{(1)} & N-1 \end{bmatrix} = \begin{bmatrix} F & -C \\ -C & N-1 \end{bmatrix}$ ,

$$(B^T B)^{-1} = \frac{(B^T B)^*}{|B^T B|} = \frac{1}{(N-1)F - C^2} \begin{bmatrix} N-1 & C \\ C & F \end{bmatrix}.$$

Theorem 2: If  $D = \sum_{k=2}^N X_{(k)}^{(0)}$  and  $E = \sum_{k=2}^N X_{(k)}^{(0)} \cdot z_{(k)}^{(1)}$ ,  $B^T Y_N = \begin{bmatrix} -E \\ D \end{bmatrix}$

Theorem 2 is easily established.

Derived from Theorem 1 and Theorem 2 is equation (3), the parameter

estimate  $\hat{a} = [a, b]^T = (B^T B)^{-1} B^T Y_N$ , where

$$a = \frac{C * D - (N-1)E}{(N-1) * F - C^2}, b = \frac{D * F - C * E}{(N-1)F - C^2} \quad (4)$$

Substituting  $a$  and  $b$  in the discrete responsive type

$$\hat{X}_{(k+1)}^{(1)} = (X_{(1)}^{(0)} - \frac{b}{a}) * e^{-a * k} + \frac{b}{a} \quad (5)$$

Because matrix B and estimation parameter  $\hat{a}$  include parameters  $r_k$  and  $s_k$ ,

parameters  $a$  and  $b$  cannot be directly calculated by (3), but will be calculated by the nonlinear optimization model as follows:

Step1: Get  $a$  and  $b$  with parameters  $r_k$  and  $s_k$  and the expression of the fitted value.

Get the fitted value  $\hat{X}_{(k+1)}^{(0)}$  of the original data  $X_{(k+1)}^{(0)}$  from form (5),

and  $\hat{X}_{(k+1)}^{(0)} = \hat{X}_{(k+1)}^{(1)} - \hat{X}_{(k)}^{(1)} (k =$

$0, 1, \dots, N-1), \hat{X}_{(1)}^{(0)} = X_{(1)}^{(0)}$

Step2: Get and solve the expression of the minimum object function:

$$\min f(r_i, s_i) = \sum_{k=2}^N (\hat{X}_{(k)}^{(0)} - X_{(k)}^{(0)})^2 \quad (6)$$

Use MATLAB to program and calculate  $r_k$  and  $s_k$ .

Step3: Substitute  $r$  and  $s$  in (3) to calculate  $a$  and  $b$ , substitute  $r_k$  and  $s_k$  in (4) to

calculate the fitted value  $\hat{X}_{(k+1)}^{(1)}$ , and get the error values.

### 3 APPLICATION OF THE NONLINEAR OPTIMIZATION MODEL OF GM(1,1) FOR THE BACKGROUND VALUE WITH PARAMETERS

Use the GDPs of china from 1991 to 2003 to set background values with two parameters  $r$  and  $s$  in NOGM(1,1), and get  $r=0.0267$ ,  $s=0.8945$ , but  $r+s \neq 1$  through (3), (4), and (5). The results show that the model that expresses the background value better is not based on the  $(1-\delta:\delta)$  ratio between the two endpoint values. The values simulated in the NOGM(1,1) are compared with those simulated in NGM(1,1)[7] in the following table 1.

Table 1. Comparison between values simulated in NOGM(1,1) and those in NGM(1,1)

year	Original data	Simulated data in DGM(1,1)	Simulated error in DGM(1,1) %	Simulated data in NOGM(1,1) with two parameters	Simulated error in NOGM(1,1) with two parameters %
				a=-0.1040,b=34468	
1991	21617.8	21617.8	0	21617.8	0
1992	26638.1	40500	52.04	36548	37.21
1993	34634.4	44750	29.22	42754	23.4437
1994	46759.4	49450	5.76	47419	1.41063
1995	58478.1	54640	6.56	52594	10.0621
1996	67884.6	60380	11.05	58332	14.0718
1997	74462.6	66720	10.4	64697	13.1148
1998	78345.2	73720	5.9	71756	8.4107
1999	82067.5	81460	0.73	79586	3.02373
2000	89468.1	90020	0.61	88269	1.34025
2001	97314.8	99470	2.21	97901	0.602375
2002	105172.3	109910	4.51	108580	3.24011
2003	117251.9	121450	3.58	120430	2.71049
Average error			10.1977		9.167

## 4 CONCLUSION

GM(1,1) model has its adaptation range[6]. By optimizing and making background values parametric, the paper uses the invertible matrix of the parameter to optimize and estimate the parameters  $\hat{a}$ , the improved model overcomes the dependence of the background value on the  $(1-\delta:\delta)$  ratio between the two endpoint values in other literatures. It provides a better prediction model for the conversion from a continual form to a discrete form. The examples indicate that the simulated value obtained by the model NOGM(1,1) is more precise, and the maximum error is reduced by 15%. However, because the simulated value is expressed by an exponential function, the simulating error is bigger at some points. Then, how to reduce the overall error will be a concern of the future study.

- Supported by the National Natural Science Foundation of China under Grant No.61066003

## REFERENCES

- [1] Naixiang Shui,Yuchun Qin.On some theoretical problems of GM(1,1) model of grey systems[J].Systems Engineering—Theory & Practice.18(4): 59-63(1998)
- [2] Sifeng Liu,Tianbang Guo,Yaoguo Dang.Grey System Theory and Its Application[M].Beijing: Science Press( 2004).
- [3] Guan-Jun Tan.The structure method and application of background value in grey system GM(1,1) model (I)[J]. Systems Engineering—Theory & Practice, 4(4): 98-103(2004).
- [4] Yinao Wang,Guangzhen Liu. A step by step optimum direct modeling method of GM(1,1)[J]. Systems Engineering—Theory & Practice, 9(6): 99-104(2000).
- [5] Naiming Xie,Sifeng Liu. Discrete GM(1,1) and mechanism of grey forecasting model[J]. Systems Engineering—Theory & Practice, 1(3): 93-99(2005).
- [6] Hui Xu,Youxing Chen, Fei Zhonghua. Discussion of the region suitable for GM(1,1) and its improving model[J]. Mathematics in Practice and Theory, 11(4): 58-63(2008).
- [7] Zhonghua Fei,Hui Xu,Zhongyi Jiang. Theoretical defect of GM(1,1) model and its optimum analysis based on time response function[J]. Mathematics in Practice and Theory, 5(10): 214-219(2009).
- [8] Na Liu,Wanhua Sun, He Yao. Research on improving the fitting and prediction precision of the GM (1,1) Model[J]. Mathematics in Practice and Theory, 4(8): 33-39(2008).
- [9] Shiwei Chen, Zhuguo LI, Shouxi Zhou.Application of non-equal interval GM(1,1) model in oil monitoring of internal combustion engine. Journal of Central South University of Technology. 12(6).705-708(2005).