Application of GM(1,1) Model Based on Residual Error Correction in Athletic Performance Prediction

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Abstract. In allusion to such problems in the athletic performance prediction as "poor information", "small sample" and "dynamics" difficult to solve through traditional statistical methods, the application of the grey model will more effectively and accurately solve the above problems. However, the application of the grey model in the athletic performance prediction also encounters many problems, for example: the traditional GM(1,1) model cannot obtain corresponding accuracy in many prediction problems. The application of GM(1,1) grey model based on residual error correction in the athletic performance prediction is researched in this article. Specifically, the best performances of the world men's 200m races during $2003\sim2013$ are taken as the samples; then, GM(1,1) grey model based on residual error correction is adopted for sample modeling, wherein the prediction accuracy level of the model is the first level; then, the model obtained thereby is adopted for predicting the best performances of the world men's 200m race in 2014, and the predicted performance is 20.59s.

Introduction

In the past more than 30 years since Professor Deng Julong created grey mathematics in 1982, the application of grey mathematics and relevant theoretical research have had rapid and significant development, wherein the modeling fields involve decision-making, control, planning, prediction, etc., and the application fields thereof involve medical treatment, economy, education, sports, etc. Nevertheless, the application of grey model has more rapid development in the athletic field. According to the statistical information of relevant theses, the athletic theses with the grey model as the research method are increased every year at a speed of $9.81\% \sim 10.10\%$ in recent five years^[1]. The rapid development of the application of the grey model in the athletic performance prediction has not only significantly improved the scientificity of the research on the athletic performance prediction, but also greatly promoted the continuous development of the grey mathematics theory. However, the grey model is sometimes abused in athletic field, and many problems not applicable to the grey model or not applicable to the traditional GM(1,1) grey prediction model are introduced into GM(1,1) grey model for modeling. Like this, the development of the grey mathematics theory and the application of the same in the athletic performance prediction will be undoubtedly influenced. At present, the application research on the grey model has become one of the most rapidly developed subjects in the athletic field. In this article, the performances of the world men's 200m races are taken as the example to research the application of the grey model in the athletic performance prediction. Finally, GM(1,1) grey prediction model based on residual error correction is established through relevant research and analysis, thus not only providing reference for predicting the performances of men's 200m races in China, but also promoting the application and development of the grey model in the athletic field.

GM(1,1) Model of the Best Performances of World Men's 500m Race

Grey prediction is one of the main research directions in the grey theory, and the method application thereof is second only to the grey correlation analysis in the grey mathematics theory. GM(1,1) model has the features of simple method, wide application scope, high accuracy and

strong practical applicability, and the application thereof has occupied 75.61% of all applications of the grey prediction model, and relevant theses are increased every year at a speed of $19.69\% \sim 21.08\%$. The variation tendency of the best performances of the world men's 200m races is analyzed in this article through relevant research, and meanwhile GM(1,1) model is applied to predict the development and change tendency thereof.

Establishment of Residual Error Sequence GM (1,1) Model

The accuracy of the prediction model is an important issue in the application of the grey model. In practical application, the accuracy of the grey prediction models including the athletic performance prediction model cannot meet corresponding requirement due to the data change and development characteristics, the model application, the knowledge quantity of the problem solver and other limitations. How to solve the prediction accuracy problem of the grey model and effectively improve and process the grey model becomes a major problem in the development of the grey model. It is known from the calculation result in the previous chapter that for the grey model established according to the best performances of the world men's 200m races, the mean square error ratio test index thereof is 0.5289, and the accuracy level is the third level, namely: the prediction accuracy of the prediction model is relatively low. In allusion to the problem about how to improve the prediction accuracy of the mare mainly focused on processing the sample data and selecting the whitenization differential equation of the grey model. In order to improve the prediction accuracy of the best performances of the world men's 200m races, the residual error sequence grey prediction is implemented for this model to correct GM(1,1) grey prediction model.

Establishment of Residual Error Sequence GM(1,1) Prediction Model

Residual error refers to the difference between the predicted value of the grey prediction model and the actual value, and the residual error sequence grey prediction model refers to the improved prediction model obtained through firstly taking the residual error sequence as the original sequence of the grey prediction model for GM(1,1) modeling and then adopting the residual error prediction model obtained thereby to correct originally established GM(1,1) grey prediction model for improving the prediction accuracy of the prediction model. According to the above calculation, the residual errors of the best performances of the world men's 200m races are as shown in Table 1: Table 1 Residual Errors of the Prediction Model of the Best Performances of

Mell's 20011 Races During 2005~2015						
Year	2003	2004	2005	2006	2007	2008
Performance	0	-0.1542	-0.0477	0.1688	-0.1647	0.1618
Year	2009	2010	2011	2012	2013	
Performance	0.1883	-0.0153	0.0612	-0.0623	-0.1358	

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The residual error sequence generated according to the residual errors of the best performances of the world men's 200m races during 2007~2013 is as follows:

 $\varepsilon_0 = (\varepsilon_0(1), \varepsilon_0(2), \cdots, \varepsilon_0(7))$

According to Table 5, some residual errors are negative values and cannot be directly used for establishing GM(1,1) grey model. Meanwhile, for adopting these residual errors to establish the grey prediction model, the sequence class ratio $\lambda(k)$ thereof shall be within the interval of $(e^{-\frac{2}{n+1}}, e^{\frac{2}{n+2}})$, or else, these residual errors shall not be used for establishing the grey prediction model. n = 7 is put into $(e^{-\frac{2}{n+1}}, e^{\frac{2}{n+2}})$ to obtain the interval (0.7788, 1.2488).

In order to establish GM(1,1) grey prediction model, the residual error sequence shall be processed to obtain the new sequence able to meet the grey model establishment conditions. In this article, the translation transformation is implemented for the residual error sequence ε_0 , namely: take a suitable positive value *c* to obtain $\varepsilon_1^{(0)}(k) = \varepsilon_0(k) + c$. Here, c = 15 is taken to obtain the

converted sequence as shown in Table 2:

		a
Table 2 New Re	sidual Error	Sequence

Year	2007	2008	2009	2010
Performance	-0.1647	0.1618	0.1883	-0.0153
Year	2011	2012	2013	
Performance	0.0612	-0.0623	-0.1358	

According to the class ratio test, the class ratio scope of the new residual error sequence is [0.9785, 1.0136], and the establishment condition of GM(1,1) grey prediction model can be met. Afterwards, we can establish GM(1,1) grey prediction model according to the modeling steps.

Result and Verification of Residual Error Sequence GM(1,1) Model

GM(1,1) model is established for the new residual error sequence. The accumulative sequence thereof is set as $\varepsilon_1^{(1)}$ to obtain the solution of the whitenization differential equation thereof as follows:

∧ (1)

 ε_1 (k+1) = 3718.11-3703.27exp(-.00410962k)

The model error values are obtained according to relevant verification and the contrastive analysis of the predicted value and the original value, as shown in the following table: Table 3 Error Values of New Pasidual Error Sequence CM(1, 1) Model

		fror values of N	ew Residual Errol	r Sequence GM(1	,1) Model
Year	Original Value	Model Value	Residual Error	Relative Error	Class Ratio Difference
2007	14.8353	14.8353	0	0.0035	
2008	15.1618	15.1878	-0.0260	0.0002	0.0255
2009	15.1883	15.1255	0.0628	0.0025	0.0058
2010	14.9847	15.0635	-0.0788	0.0079	-0.0094
2011	15.0612	15.0017	0.0595	0.0052	0.0092
2012	14.9377	14.9402	-0.0025	0.0088	-0.0041
2013	14.8642	14.8789	-0.0147	0.0013	-0.0008

By virtue of the error values and the residual error values of the model in Table 3, residual error GM(1,1) model is calculated to obtain various verification indexes as shown in Table 4:

Table 4 Accuracy of Residual Error GM(1,1) Model

First Level 0.00	0.9978	0.3272	

According to Table 8, the relative error of the model is 0.0023, the accuracy level is the first level, the absolute relevancy is 0.9978 and the mean square error ratio is 0.3272. Residual error GM(1,1) grey system has high accuracy for grey prediction model and can be used for residual error prediction.

For recovering the residual error sequence, the predicted values of the residual errors during 2003~2006 are set as 0 to obtain the predicted residual error values as shown in Table 5:

Table 5 Predicted Values of Residual Errors of the Best Performances of

Men's 200m Races during 2003~2013						
Year	2003	2004	2005	2006	2007	2008
Performance	0	0	0	0	-0.1647	0.1878
Year	2009	2010	2011	2012	2013	
Performance	0.1255	0.0635	0.0017	-0.0598	-0.1211	

Establishment of GM(1,1) Grey Prediction Model Based on Residual Error Correction

According to relevant calculation in the previous chapter, the solution of the whitenization differential equation of the new residual error sequence is as follows:

^ (1)

 ε_1 (k) = 3718.11-3703.27exp(-.00410962k)

The solution of the above whitenization differential equation is obtained through adopting the sequence of the residual errors during $2007 \sim 2013$ to establish GM(1,1) grey prediction model. In order to combine the solution with the result of GM(1,1) grey prediction model established according to the sequence of the best performances of the world men's 200m races during 2003~2013, the above formula shall be processed.

$$\varepsilon_2^{(1)}(k) = \begin{cases} 0 & k \le 4 \\ \varepsilon_1^{(1)}(k-4) & k > 4 \end{cases}$$

If $y^{(0)}(k)$ ($k = 1, 2, \dots, 13$) is assumed as the predicted value of GM(1,1) grey prediction model based on residual error correction, then $y^{(1)}(k)$ is the accumulative sequence of the predicted values of the residual error grey prediction model. Afterwards, the following formula can be obtained.

$$y^{(1)}(k) = x^{(1)}(k) + \varepsilon_2^{(1)}(k-4)$$

The prediction sequence $y^{(0)}(k)$ is recovered through the accumulative sequence $y^{(1)}(k)$ in order to obtain the predicted values and the corresponding error values as shown in Table 6:

Year	Original Value	Model Value	Residual Error	Relative Error
2003	20.13	21.3000	0	0
2004	20.58	20.7342	-0.1542	0.0075
2005	20.69	20.7377	-0.0477	0.0023
2006	20.91	20.7412	0.1688	0.0081
2007	20.58	20.5800	0	0
2008	20.91	20.9360	-0.0260	0.0012
2009	20.94	20.8772	0.0628	0.0030
2010	20.74	20.8188	-0.0788	0.0038
2011	20.82	20.7605	0.0595	0.0029
2012	20.70	20.7025	-0.0025	0.0001
2013	20.63	20.6447	-0.0147	0.0007

Table 6 Error	Values of	Grey Pre	diction Seq	uence GM	(1,1) Model
					() /

By virtue of the error values and the residual error values of the model in Table 9, GM(1,1) grey prediction model based on residual error correction is calculated to obtain various verification indexes as shown in Table 7:

Table 7 Accuracy of GM(1,1) Prediction Model Based on Residual Error Correction

Accuracy Level	Relative Error	Absolute Relevancy	Mean Square Error Ratio
First Level	0.0027	0.9980	0.3328

According to Table 10, the relative error of the model is 0.0027, the accuracy level is the first level, the absolute relevancy is 0.9980 and the mean square error ratio is 0.3328. In other words, the prediction accuracy level of GM(1,1) grey prediction model based on residual error correction is the first level and this model can be used for accurately predicting the best performances of the world men's 200m race.

Then, the above model is used to predict the best performances of the world men's 200m race in 2014.

^ (1)

 $x \quad (k+1) = 122821e^{-0.000168803k} - 122801$

According to the above formula, when k = 10 and k = 11 are true, we can obtain $\hat{x}^{(1)}(11) = 227.5006$ and $\hat{x}^{(1)}(12) = 248.2699$.

Namely, the predicted value obtained from GM(1,1) grey prediction model for 2004

 $is x^{(0)}(12) = 20.7639$.

Similarly, the predicted value of the residual error for 2014 is $\hat{\mathcal{E}}_{2}(8) = -0.1821$. Therefore, the best performance of the world men's 200m race in 2014 is calculated as 20.59 according to GM(1,1) model based on residual error correction.

Conclusion

In this article, the grey prediction model is established according to the best performances of the world men's 200m races in order to not only predict the performance of the world men's 200m race in 2014, but also discuss the application method of the grey model in the athletic performance prediction, thus promoting the application of the grey model in the athletic sports. GM(1,1) grey model is applicable to the modeling process with the sample data distribution presented in index movement. Through establishing relevant model for the statistical data for analysis, the grey prediction model is extended. Meanwhile, GM(1,1) grey prediction model based on residual error correction is applied to predict the best performance of the world men's 200m race in 2014, and this model is also compared with GM(1,1) grey model for contrastive analysis. According to the comparison result, we can find that GM(1,1) grey prediction model based on residual error correction has the features of low complexity and high prediction accuracy, thus applicable to the athletic performance prediction.

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