Journal of Chemical and Pharmaceutical Research, 2014, 6(2):261-268



Research Article

ISSN: 0975-7384 CODEN(USA): JCPRC5

Decathlon each interaction regression factors analysis based on GRA and FAM

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ABSTRACT

Decathlon has early started in our country, but it is always the backwards event in our country's athletics which still has greater disparity with the world standard. Although in recently years our country excellent athlete Qi Hai-Feng has achieved a historic breakthrough in the event, China overall level in the event appears preliminary declining tendency. This paper adopts grey relational analysis and factor analysis, researches Chinese and foreign country excellent decathlon athletes total performance and each single performance internal relations as well as athletes performance internal structure, and makes contrast analysis, finds out main gap between our country athletes and world level athletes performance, provides a certain theoretical basis for China men decathlon training plan designing, athletes scientific material selections as well as athletics events development. Applies multiple regression analysis method establishing world excellent decathlon athletes' performance prediction model and test its prediction precise.

Key words: Grey Relational Analysis, Factor Analysis Method, Multiple Regression Analysis, Men decathlon

INTRODUCTION

Decathlon is one of the oldest events in athletics with high requests of human functions, is a comprehensive sports event that combines techniques, physical ability, intellectuals into one, whose competitive levels reflects a country athletics levels to some extent. Modern men decathlon by far has nearly 100 years history, human closely focus on its changes and development since it appeared. From world range, decathlon researches experienced events meticulous designing and combination, athletes' selection and training, grading method development and revision, different periods' cases summarizing and analysis and other stages [1-3]. From 1990s to the beginning of this century, Chinese and foreign countries scholars exploration on decathlon has arrived at peak period makes it rapidly developing and its records continuously adapting [4-6]. With Olympic Games and other huge competitive levels still are the key factors that restraint our country sports competitive levels improvement, especially that competitive level of decathlon in athletics generates great gap with world athletics competitive levels, and according to data analysis though recently times it is still a long target for our country decathlon catching up with world advanced level, only we correctly find our country decathlon backwards internal causes together with improvements and learning from each other, it still can fully achieve Asian leading position in a short time [7].

Grey system theory was first proposed by our country scholar professor Deng Ju-Long in 1982. Through 20 more years' perfection and development, the theory has been widely applied in agriculture, economics, medicine, military, sports, industry control and other dozens of fields. Due to grey theory adaptation range is "small sample", "poor data information", "undefined" grey hazy set, many problems in sports system are just in the field. Therefore, it tends to be able to provide a reasonable answer for some questions that are hard to explain by traditional methods as probability statistics (such as SPSS). According to grey system theory, it can regard decathlon as a grey system project, through relevancy analysis, reveal each single item performance in decathlon training position as well as its

functions on total performance, get acknowledge of mutual relations among them. Factor analysis concept derives from the beginning of 20th century Karl Pearson and Charles Spearmen statistical analysis with regard to intellectual test. At present, factor analysis has already successfully applied in psychology, medicine, weather, address, economics and other fields, and therefore foster theory continuously enrichment and perfection. Factor analysis is premised on minimum information loss; integrate multiple original variables into few comprehensive indicators, called factors. Usually, factor analysis generated factors can finally obtain name explanatory through each way. Factor name explanation to each single item, it is impossible to achieve fully balance, system and things are composed of multiple factors. Imbalance and dynamic changes among factors also decide system and things prediction complex [3, 5].

Based on above, this paper according to Chinese and foreign 28 excellent decathlon athletes competition performance, adopts SPSS software respectively carry out factor analysis and establish each single item performance and total performance multiple linear regression prediction model ,so as to better explore decathlon internal elements structure and item laws features. Through recently years' our country decathlon athletes 'exhibition and foreign excellent athletes contrast analysis, and combining China current training, competition system, analyze China decathlon development current situations and backwards main causes from macro fields, and propose reasonable suggestions and counter measurements on China decathlon promotion.

CHINESE AND FOREIGN MEN DECATHLON TOTAL PERFORMANCE AND EACH SINGLE ITEM PERFORMANCE GREY CORRELATION ANALYSIS

Through 11th national games information, select China and world previous 14 excellent men decathlon athletes as research objects, their total scores as well as each single item scores are as following Table 1 and Table 2 shows, basically represents China and foreign men decathlon highest level.

According to grey system theory, it can regard decathlon as a grey system, make grey correlation analysis of selected research objects performance. Through correlation degree analysis, it can know each single event performance effects on total performance as well as its position in decathlon training.

| Rank | Athlete | 1500m | Javelin | Pole vault | 110mHurdle | 100m | 400m | Shot | High jump | Long jump | Discus | Total performance |
|------|----------------|-------|---------|------------|------------|------|------|------|-----------|-----------|--------|-------------------|
| 1 | Qi Hai-Feng | 643 | 752 | 761 | 872 | 893 | 849 | 699 | 804 | 906 | 769 | 7946 |
| 2 | Yu Bin | 594 | 772 | 674 | 888 | 895 | 813 | 744 | 723 | 942 | 754 | 7798 |
| 3 | Zhu Heng-Jun | 601 | 705 | 761 | 944 | 929 | 874 | 711 | 697 | 855 | 636 | 7712 |
| 4 | Liu Hai-Bo | 640 | 669 | 618 | 832 | 785 | 779 | 673 | 916 | 809 | 713 | 7433 |
| 5 | Hao Ming | 605 | 546 | 820 | 844 | 800 | 747 | 720 | 832 | 772 | 692 | 7376 |
| 6 | Wang Jian-Bo | 522 | 743 | 731 | 836 | 755 | 709 | 724 | 697 | 835 | 800 | 7351 |
| 7 | Zhao De-Ning | 601 | 680 | 703 | 843 | 835 | 816 | 626 | 645 | 840 | 679 | 7266 |
| 8 | Lu Huan-Yong | 629 | 658 | 646 | 794 | 800 | 691 | 695 | 887 | 838 | 623 | 7260 |
| 9 | Guo Wei -Zhao | 616 | 671 | 703 | 850 | 837 | 823 | 659 | 723 | 795 | 556 | 7233 |
| 10 | Li Xian-Gui | 591 | 585 | 674 | 871 | 806 | 794 | 660 | 723 | 847 | 560 | 7111 |
| 11 | Lin Qing-Quan | 520 | 799 | 618 | 726 | 811 | 753 | 696 | 697 | 734 | 664 | 7017 |
| 12 | Yang Wen-Liang | 640 | 702 | 790 | 728 | 768 | 726 | 620 | 645 | 700 | 695 | 7014 |
| 13 | Tang Jun | 561 | 551 | 618 | 934 | 832 | 739 | 539 | 777 | 730 | 485 | 6766 |
| 14 | Zhou Bin | 590 | 300 | 703 | 807 | 783 | 755 | 484 | 671 | 779 | 456 | 6327 |

Table 1: Chinese 14 excellent men decathlon athletes each single event performance and total performance

Table 2: World 14 excellent men decathlon athletes each single event performance and total performance

| Rank | Athlete | Country | 1500m | Javelin | Pole vault | 110m hurdle | 100m | 400m | Shot | High jump | Long jump | Discus | Total performance |
|------|--------------------|--------------|-------|---------|------------|-------------|------|------|------|-----------|-----------|--------|-------------------|
| 1 | Ashton Eaton | America | 721 | 767 | 972 | 1032 | 1011 | 963 | 769 | 850 | 1068 | 716 | 8869 |
| 2 | Terre Hardy | America | 674 | 838 | 849 | 1035 | 994 | 904 | 807 | 794 | 942 | 834 | 8671 |
| 3 | Lionel Suarez | Cuba | 744 | 996 | 819 | 917 | 801 | 859 | 759 | 906 | 940 | 782 | 8523 |
| 4 | Hans Van Alphen | Belgium | 795 | 763 | 849 | 863 | 850 | 853 | 819 | 850 | 970 | 835 | 8447 |
| 5 | Warner | Canada | 746 | 780 | 819 | 926 | 980 | 899 | 712 | 850 | 945 | 785 | 8442 |
| 6 | Rico Freimuth | Germany | 695 | 698 | 880 | 989 | 940 | 906 | 782 | 714 | 864 | 852 | 8320 |
| 7 | Oleksiy Kasyanov | Ukraine | 721 | 661 | 790 | 963 | 961 | 888 | 756 | 794 | 947 | 802 | 8283 |
| 8 | SergeySviridov | Russia | 702 | 865 | 790 | 799 | 910 | 866 | 754 | 794 | 922 | 817 | 8219 |
| 9 | Kerzen | South Africa | 768 | 810 | 760 | 955 | 841 | 882 | 715 | 850 | 854 | 738 | 8173 |
| 10 | Pascal Behrenbruch | Germany | 696 | 810 | 819 | 932 | 847 | 813 | 831 | 767 | 850 | 761 | 8126 |
| 11 | Elko | Holland | 737 | 720 | 1004 | 920 | 894 | 868 | 739 | 740 | 903 | 509 | 8034 |
| 12 | Newdic | New Zealand | 692 | 735 | 819 | 847 | 838 | 804 | 795 | 767 | 900 | 791 | 7988 |
| 13 | Barroilhet | Chile | 629 | 697 | 1035 | 959 | 821 | 766 | 758 | 850 | 767 | 690 | 7972 |
| 14 | Garcia | Cuba | 689 | 736 | 790 | 944 | 906 | 873 | 758 | 794 | 755 | 711 | 7956 |

Through Table 1. Table 2 observation, it is known that athletes each single event performance and total performance have the same dimension and so no need to take dimensionless treatment on data.

Select each athletes total performance $x_0(k)$ as reference series:

$$x_0 = \{x_0(k) | k = 1, 2, ..., n\} = \{x_0(1), x_0(2), ..., x_0(n)\}$$
 (From which k represents the k athlete)

Select each athlete each single event performance $x_i(k)$ as contrast series:

$$x_i = \{x_i(k) | k = 1, 2, \dots, n\} = \{x_i(1), x_i(2), \dots, x_i(n)\}$$

(From which i represents the i single event)

It can be concluded that the k athlete contrasts series x_i correlation coefficient to reference series x_0 :

$$\xi_{i}(k) = \frac{\min_{s} \min_{t} |x_{0}(t) - x_{s}(t)| + \rho \max_{s} \max_{t} |x_{0}(t) - x_{s}(t)|}{|x_{0}(k) - x_{i}(k)| + \rho \max_{s} \max_{t} |x_{0}(t) - x_{s}(t)|}$$
(1)

Among them, $\rho \in [0,1]$ is resolution coefficient? It is called in formula(1) $\min_{s} \min_{t} |x_0(t) - x_s(t)|$ is two-level maximum difference. Generally speaking the bigger

minimum difference, ρ is, the bigger resolution percentage would be; the smaller ρ is, the smaller resolution percentage would be.

Formula (1)defined correlation coefficient is an indicator that describes one athlete contrast series and reference series correlation degree, due to each athlete has a correlation number, information turns to be too scattering, not helpful for comparation, therefore it can adopt average correlation degree to carry out comparation.

$$r_{i} = \frac{\sum_{k=1}^{n} \xi_{i}(k)}{n} \qquad (r_{i} \text{ is series } x_{i} \text{ to reference series } x_{0} \text{ correlation degree})$$
(2)

According to Table 1 and Table 2 data, apply MATLAB software program calculation get China and world excellent athletes decathlon performance and each single event correlation degree as following Table 3 and Table 4 shows.

Table 3: Chinese excellent athletes' decathlon performance and each single event correlation degree

| Item | 110m hurdle | 100m | Long Jump | 400m | High Jump |
|--------------------|-------------|--------|-----------|--------|-----------|
| Correlation degree | 0.9123 | 0.9106 | 0.9096 | 0.9064 | 0.9037 |
| Sequence | 1 | 2 | 3 | 4 | 5 |
| Item | Pole vault | Shot | Javelin | Discus | 1500m |
| Correlation degree | 0.8998 | 0.896 | 0.8951 | 0.8948 | 0.8907 |
| Sequence | 6 | 7 | 8 | 9 | 10 |

Table 4: World excellent athletes' decathlon performance and each single event correlation degree

| Item | 110m hurdle | Long Jump | 100m | 400m | Pole vault |
|--------------------|-------------|-----------|--------|--------|------------|
| Correlation degree | 0.964 | 0.9612 | 0.9611 | 0.9584 | 0.9577 |
| Sequence | 1 | 2 | 3 | 4 | 5 |
| Item | High Jump | Javelin | Shot | Discus | 1500m |
| Correlation degree | 0.9536 | 0.9509 | 0.9503 | 0.9495 | 0.9459 |
| Sequence | 6 | 7 | 8 | 9 | 10 |

From Table3 and Table 4, it is clear that foreign athletes each single event influence on total performance as well as correlation sequence as 110m hurdle > 100m > long jump > 400m > high jump > pole vault > shot > javelin > discus > 1500m; While world sequence is: 110m hurdle > long jump > 100m > 400m > pole vault > high jump > javelin

> shot > discus > 1500m. Compared with foreign athletes sequence, ten single event 110 hurdle, 400m, discus and 1500m sequence are the same, successively rank in 1, 4, 9, 10, domestic top three are also long jump, 100m and 110m hurdle, but sequence are different, these three items all require horizontal speed cover absolute dominates; Foreign pole vault, javelin are respectively rank in the front of high jump and shot, while it is on the contrary at home. It is clear that domestic athletes has longer gap from foreign excellent athletes in special techniques high requested pole vault and javelin such two events, which indicates that our country athletes still have shortcomings on technical motions completion with high speed, high rhythm that should be taken seriously by our country all-round coaches.

CHINESE AND FOREIGN MEN DECATHLON PERFORMANCE FACTOR ANALYSIS

Men decathlon is a special competitive sport event that composed of running, jumping, throwing three large kinds of athletic events, its score rules as well as its comprehensive features decide that participating athletes must be versatile, any one event imbalance development would influence on decathlon total performance. However, to athletes, it is impossible to achieve each event fully balance. Due to each event has connection and each event develop status decides athletes total performance, the model according to statistics principal and method, apply SPSS 19 statistics software making statistics and handling with 28 Chinese and foreign men decathlon athletes' sport performance as well as each single event performance, research each large kinds of event levels influences on total performance.

Correlation coefficient matrix test

The model adopts KMO test making factor analysis adaptation test, KMO is Kaiser-Meyer-Olkin sampling adequacy measure. The higher KMO measurement value is (close to 1.0), indicates that more variables common factors would be, research data fit for factor analysis. Normally explain the indicator value with following criterion: KMO value arrives at above 0.9 means excellent, 0.8~0.9 means good, 0.7~0.8 means normal, 0.6~0.7 means bad, 0.5~0.6 means worse. If KMO measurement values are lower than 0.5, indicates samples are too small and needs to expand samples.

Input Table 1 Table 2 data into SPSS software, operate and get as following Table 5:

Table 5: KMO and Bartlett test

| Sampling enough measure Ka | aiser-Meyer-Olkin measurement | .765 |
|----------------------------|-------------------------------|---------|
| | approximate Chi-square | 173.277 |
| Bartlett sphericity test | df | 45 |
| | Sig. | .000 |

From Table 5, it can know KMO test result is 0.765 larger than 0.5, indicates data weaker than correlation, which is fit for factor analysis. Bartlett value equal to 173.277, P<0.001, indicates related matrix is not a unit matrix which can implement factor analysis.

Factor analysis

At first extract principal component, convert each variable value into standard value. Principal component analysis is constructing a multiple dimensions space with variety of variables, then project line in space to explain maximum variance, achieved line is common factor which can represent each variable natures to the maximum degree, while

values in the line construct a variable is the first common factor, or is called the first factor (F_1). But there is still remaining variance in space; therefore it needs to project the second line to explain variance. At this time, it needs to follow the second criterion that project the second line and first line form direct relations (that is uncorrelated), which means represent different aspects. Values in the second line construct a variable is called the second factor

 (F_2) . According to the principal it can determine the third, the fourth or more factors. Principally, factors numbers and original variables numbers are the same, but after extracting main factors, if rest variance is very small, it can give up other factors so as to achieve the purpose simplify data.

The model defines 10 events into variables, successively are 1500m, javelin, pole vault, 110m hurdle, 100m, 400m, shot, high jump, long jump, discus. Take athletes total performance as samples observation value; apply SPSS implementing factor analysis get as following Table 6:

| | | | Explanatory total va | riance | | | | |
|---------|-------|--------------|----------------------|--------|----------------------------|----------------|--|--|
| Element | | Initial feat | ture value | | Extract square sum loading | | | |
| | Total | Variance % | Accumulation % | Total | Variance % | Accumulation % | | |
| 1 | 5.333 | 53.332 | 53.332 | 5.333 | 53.332 | 53.332 | | |
| 2 | 1.446 | 14.457 | 67.789 | 1.446 | 14.457 | 67.789 | | |
| 3 | .927 | 9.274 | 77.063 | .927 | 9.274 | 77.063 | | |
| 4 | .730 | 7.297 | 84.359 | .730 | 7.297 | 84.359 | | |
| 5 | .502 | 5.025 | 89.384 | \ | / | | | |
| 6 | .363 | 3.627 | 93.011 | \ | \ | | | |
| 7 | .301 | 3.012 | 96.023 | \ | \ | | | |
| 8 | .183 | 1.826 | 97.850 | \ | | | | |
| 9 | .151 | 1.509 | 99.359 | \ | \ | | | |
| 10 | .064 | .641 | 100.000 | \ | / | \ | | |

Table 6: Each event variance and accumulative contribution ratio

From Table 6, it is clear that according to professional knowledge judging, select 4 principal components. These four principal components accumulative contribution ratio arrive at 84.359% larger than 80%. It can be known that these four principal component loaded sport single events are athletes' main scoring events in decathlon.

Then factor rotation, usually extract initial factors, then cannot make effect explanation on factors. At this time, it usually needs to make factors rotation, through coordinate transformation can make factor solution significance more easily explaining. The purpose of rotating shaft is changing subjects each factor load quantity size, adjust each factor loading quantity size according to subject and factor structural relations closely degrees when rotating shaft. After rotating shaft, variables in each factor loading quantity is growing bigger (close to 1) or smaller (close to 0), while each factor loading quantity size almost the same before non-shaft that let name common factors and explain variables more easily. After revolving shaft, every common factor feature values would change, but each variable communality would not change.

After extracting factors and rotating, principal component accumulative contribution ratio as following Table 7:

Table 7: After factors rotation each principal component accumulative contribution ratio

| | Explanatory total variance | | | | | | | | | |
|---------|-----------------------------|------------|----------------|--|--|--|--|--|--|--|
| Element | Rotation square sum loading | | | | | | | | | |
| Element | Total | Variance % | Accumulation % | | | | | | | |
| 1 | 3.083 | 30.826 | 30.826 | | | | | | | |
| 2 | 2.582 | 25.820 | 56.646 | | | | | | | |
| 3 | 1.391 | 13.913 | 70.558 | | | | | | | |
| 4 | 1.380 | 13.801 | 84.359 | | | | | | | |
| 5 | \ | \ | | | | | | | | |
| 6 | \ | \ | | | | | | | | |
| 7 | \ | \ | | | | | | | | |
| 8 | \ | \ | | | | | | | | |
| 9 | \ | \ | | | | | | | | |
| 10 | \ | \ | \ | | | | | | | |

From Table 7, it is clear that after extracting factors and rotating each principal component accumulative contribution ratio hasn't changes still is 84.359%, but feature values tend to concentration and values all above 1.

Through great variance orthogonal rotation, after 5 times rotation get as following Table 8:

Table 8: After rotation factor loading

| | Rotation components matrix | | | | | | | | |
|-------------|----------------------------|-----------|------|------|--|--|--|--|--|
| 1 | | Component | | | | | | | |
| \ | 1 | 2 | 3 | 4 | | | | | |
| 100m | .917 | .192 | 028 | .139 | | | | | |
| 400m | .882 | .268 | .106 | .217 | | | | | |
| 110m hurdle | .742 | .009 | .187 | .406 | | | | | |
| Long jump | .706 | .417 | .279 | .026 | | | | | |
| Discus | .186 | .880 | .107 | .064 | | | | | |
| Javelin | .169 | .856 | .165 | .051 | | | | | |
| Shot | .250 | .784 | .179 | .403 | | | | | |
| High jump | .080 | .186 | .947 | .058 | | | | | |
| 1500m | .446 | .327 | .539 | .393 | | | | | |
| Pole vault | .292 | .184 | .092 | .906 | | | | | |

From Table 8 can know the first principal component larger loading variable is 100m, 400m, 110m hurdle, long jump; The second principal component larger loading variable is discus, javelin, shot. The third principal component larger loading variable is pole vault.

From factor analysis result, it is clear that in Chinese and foreign athletes' performance structure, 100m, 400m, 110m hurdle these three single event have the greatest functions, they can call speed, explosive force factor, discus, javelin and shot these 3 events functions are the secondary, they can call strength factors; High jump functions are the next that call nimble factor; Minimum functions event is 1500m, it can call speed endurance factor. These four factors structure can explain Chinese and foreign excellent athletes still focus on absolute speed training in training; Rank two factors discus, javelin and shot these three single events not only requests athletes has sufficient guarantee in absolute strength, and it have higher correlation in individual speed quality. Though it is not significant in score contribution ratio aspect that compare throw kinds event with other classification events, under current versatile sport balance development, eliminate weak events guiding thoughts, the kind factor functions go without saving; Rank three factors only high jump such one event, due to high jump is horizontal speed and vertical speed common function inter transformation generate dynamics, meanwhile it also asks special for athletes 'body shape, therefore classify separately into one kind; In speed endurance factors, similarly it have requests to athletes body shape but not the same as the first ,second factors, but these four kinds factors always correlated to speed quality. It can be seen that men decathlon core with speed is conform to current development tendency. For men decathlon athletes, it is necessary to comprehensive develop each physical quality ,especially speed ,strength and techniques, in future training it need to focus on strengthen techniques aspects training on the basis of speed, strength consolidation.

MULTIPLE REGRESSION PREDICTION MODELS

Give all-round events 10 single event performance as independent variable, total performance as dependent variable. Adopt full model method carrying out linear regression, apply SPSS software in handling Table 1 can get as following Table 9, Table 10 and Table 11:

Table 9: Model Summary b

| Model | R | R Square | Adjusted R Square | Std. Error of the Estimate | | | | |
|--------------|--|----------|-------------------|----------------------------|--|--|--|--|
| 1 | .995a | .991 | .986 | 78.31389 | | | | |
| a. Predictor | a. Predictors: (Constant), discus, 110m hurdle, high jump, pole vault, long jump, javelin, 1500m, 100m, shot,400 | | | | | | | |
| | b. Dependent Variable: total performance | | | | | | | |

Table 9 is model fitting goodness test, from table can know that both square coefficient and after adjust square coefficient arrive at more than 0.99, indicates model has higher fitting goodness.

Table 10: ANOVA b

| | Model | Sum of Squares | df | Mean Square | F | Sig. |
|--|------------|--------------------|----------|------------------|---------|------------|
| | Regression | 1.135E7 | 10 | 1134536.674 | 184.987 | .000a |
| 1 | Residual | 104262.120 | 17 | 6133.066 | \ | \ |
| | Total | 1.145E7 | 27 | / | \ | \ |
| a. Predictors: (Constant), discus, 110m hurdle, high jump, pole vault, long jump, javelin, 1500m, 100m, shot, 40 | | | | | | shot, 400m |
| | | b. Dependent Varia | ıble: ta | otal performance | | |

Table 10 is model variance test, from Table can know F = 184.987, significant correlation probability is 0.000^a , which is in significant difference.

| | Madal | Unstandardiz | ed Coefficients | Standardized Coefficients | | | | | |
|---|--|--------------|-----------------|---------------------------|--------|------|--|--|--|
| | Wodel | В | Std. Error | Beta | t | Sig. | | | |
| | (Constant) | -298.001 | 232.235 | | -1.283 | .217 | | | |
| | 1500m | 1.396 | .433 | .155 | 3.225 | .005 | | | |
| | Javelin | 1.134 | .200 | .216 | 5.677 | .000 | | | |
| | Pole vault | 1.074 | .211 | .180 | 5.091 | .000 | | | |
| | 110m hurdle | .837 | .339 | .102 | 2.472 | .024 | | | |
| 1 | 100m | 1.962 | .482 | .214 | 4.066 | .001 | | | |
| | 400m | .355 | .676 | .037 | .525 | .606 | | | |
| | Shot | .363 | .407 | .044 | .890 | .386 | | | |
| | High jump | 1.158 | .278 | .137 | 4.160 | .001 | | | |
| | Long jump | .948 | .282 | .125 | 3.366 | .004 | | | |
| | Discus | 1.170 | .230 | .196 | 5.088 | .000 | | | |
| | a. Dependent Variable: total performance | | | | | | | | |

Table 11: Coefficients

Table 11 is regression calculation process each equation coefficient table, from which can know t test shows significance level.

According to Table 11can get relative each event all-round non-standard regression model.

 $Y = -298.001 + 1.396X_1 + 1.134X_2 + 1.074X_3 + 0.837X_4 + 1.962X_5 + 0.355X_6 + 0.363X_7 + 1.158X_8 + 0.948X_9 + 1.170X_{10} + 1.170X_{10}$

Among them, $X_1 - X_{10}$ are independent variables, respectively corresponding represent 1500m, javelin, pole vault, 110m hurdle, 100m, 400m, shot, high jump, long jump, discus single event performance; Y is dependent variable, represents decathlon total performance.

| Name | Prediction total performance | Actual total performance | Predict accuracy ratio |
|------------------------------|------------------------------|--------------------------|------------------------|
| Qi Hai-Feng | 7902.37434 | 7,941.00 | 0.99513592 |
| Yu Bin | 7833.507706 | 7,791.00 | 0.994543999 |
| Zhu Heng-Jun | 7843.854998 | 7,708.00 | 0.982374806 |
| Liu Hai-Bo | 7898.012259 | 7,427.00 | 0.936581088 |
| Hao Ming | 7849.591999 | 7,370.00 | 0.934926459 |
| Wang Jian-Bo | 7733.213342 | 7,346.00 | 0.947289226 |
| Zhao De-Ning | 7843.356905 | 7,261.00 | 0.919796598 |
| Lu Huan-Yong | 7882.3557 | 7,256.00 | 0.91367755 |
| Guo Wei -Zhao | 7864.222777 | 7,230.00 | 0.912279007 |
| Li Xian-Gui | 7830.45619 | 7,106.00 | 0.898050072 |
| Lin Qing-Quan | 7730.492957 | 7,012.00 | 0.897533805 |
| Yang Wen-Liang | 7898.52487 | 7,009.00 | 0.873088191 |
| Tang Jun | 7788.761997 | 6,761.00 | 0.847986689 |
| Zhou Bin | 7829.138226 | 6,323.00 | 0.761800059 |
| Ashton Eaton | 8011.43 | 8,869.00 | 0.903307024 |
| Terre Hardy | 7945.818 | 8,671.00 | 0.91636697 |
| Lionel Suarez | 8043.538 | 8,523.00 | 0.943744925 |
| Hans Van Alphen | 8114.734 | 8,447.00 | 0.960664615 |
| Warner | 8046.33 | 8,842.00 | 0.910012441 |
| Rico Freimuth | 7975.134 | 8,320.00 | 0.95854976 |
| Oleksiy Kasyanov | 8011.43 | 8,283.00 | 0.96721357 |
| SergeySviridov | 7984.906 | 8,219.00 | 0.971517946 |
| Kerzen | 8077.042 | 8,173.00 | 0.988259146 |
| Pascal Behrenbruch | 7976.53 | 8,126.00 | 0.981605956 |
| Elko | 8033.766 | 8,034.00 | 0.999970874 |
| Newdic | 7970.946 | 7,988.00 | 0.997865048 |
| Barroilhet | 7882.998 | 7,972.00 | 0.988835675 |
| Garcia | 7966.758 | 7,956.00 | 0.998647813 |
| Accuracy ratio average value | \ | \ | 0.939343758 |

Input Chinese and foreign versatile athletes each single event performance into prediction model, get prediction total

performance, compare prediction value with actual value. It can get as following Table12.

From Table 12 can get model prediction accuracy degree average value is 0.9393, it can be concluded that model prediction accuracy degree quiet high. The prediction equation fully can predict Chinese and foreign excellent decathlon athletes total performance.

CONCLUSION

Grey correlation analysis results show that Chinese and foreign athletes each single event influence on total performance as well as correlation sequence as 110m hurdle > 100m > long jump > 400m > high jump > pole vault > shot > javelin > discus > 1500m; While world sequence is: 110m hurdle > long jump > 100m > 400m > pole vault > high jump > javelin > shot > discus > 1500m. It is clear from that Chinese athletes still have greater gap by comparing with foreign excellent athletes on high requested special techniques pole vault and javelin such two events which indicates our country athletes still have shortcomings on technical motions completion with high speed, high rhythm that should be taken seriously by our country all-round coaches.

From factor analysis result, it is clear that in Chinese and foreign excellent athletes' performance structure,100m, 400m, 110m hurdle these three single event have the greatest functions, they can call speed, explosive force factor, discus, javelin and shot these 3 events functions are the secondary, they can call strength factors; High jump functions are the next that call nimble factor; Minimum functions event is 1500m, it can call speed endurance factor. These four factors all are related to speed quality, which reveals that men decathlon feature is core with speed.

Take 10 single event performance as independent variable, take total performance as dependent variable, adopt full model method establish men decathlon performance multiple regression equation, through samples back substitution, equation precise arrives more than 99%, therefore can regard as Chinese and foreign excellent decathlon athletes' sports performance prediction model.

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