On the equivalence of the forecast value construction in the “pyramidal” extrapolation method and cubic forecast

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Abstract—The paper is devoted to the new method of short time series extrapolation, which was developed in recent years and was called “pyramidal”. This method can be used for predicting the deterministic component of an arbitrary economic, environmental and geophysical parameters. Numerical results show significant advantages of the proposed method in comparison with approaches to extrapolate, based on the use of polynomials, including Newton’s extrapolation. In this paper we study the procedure for constructing the forecast value in accordance with the pyramidal method and show that it is equivalent to the cubic forecast.

Keywords—extrapolation; forecast; divided differences; interpolation, time series recognition problems

I. INTRODUCTION

The most common methods of short time-series forecasting are the extrapolation methods [6]. Among the known methods of extrapolation we can distinguish methods based on interpolation polynomials – the most commonly used Newton-Gregory backward difference formula [3], generalized polynomial interpolation based on the systems of Chebyshev functions – polynomials, exponential, trigonometric functions [2], methods based on the special functions and trend analyze [4]-[9], methods based on spline – cubic, B-spline, Bezier curves [10] and methods, based on the statistical approaches [6],[12]-[13].

There are a number of problems dealing with the study of small time series [8]. For example, in the absence of additional information it is often impossible to conclude that process is determinate or indeterminate, which significantly effects on the model construction. In determinate case is obvious that for any set of points there are many curves that pass through them or something to bring, and it is difficult to argue that one curve (model) is precisely the law that comprehensively describes the phenomenon and will effectively predict its behavior in the future. The hypothesis, which corresponds to the observation point, is effective if it is appropriate to predict with reasonable accuracy the process starting from any step of observations. In practice, the finding an effective hypothesis is impossible. The main idea of this article is to develop a technique that allows you to build the optimal predict point without direct use of any specific classes of extrapolation function. In the paper [1] was proposed a new method of short time series extrapolation, which can be used for predicting of economic, environmental and geophysical parameters. This method is based on the property of the rows of modified finite differences that the best cubic approximation is in the range of convexity. Numerical results show significant advantages of the proposed method in comparison with approaches to extrapolate, based on the use of polynomials, including Newton’s extrapolation. This method is comparatively new and requires deep in-depth research and data series validation. In this paper we study the procedure for constructing the forecast value in accordance with the pyramidal method and show that it is equivalent to the cubic forecast.

II. PIRAMIDAL METHOD OF DATA EXTRAPOLATION

Pyramidal method of data extrapolation was proposed in [1]. The main feature of this method is construction a special divided differences and finding their order, for which a better predicted value in a certain sense can be found. Then the value of the original function at the point located outside the interpolation interval is based on the predictive value for the divided differences using a special computational procedure.

Let $f(x)$ be a continuous function of one independent variable defined over a closed interval $[a,b]$ having continuous derivatives and let values $f_1, f_2, \ldots, f_N$ of this function are
known at the points \( x_1, x_2, \ldots, x_n \) respectively. It is needed to assess the value of this function at the point \( x > x_n \). Consider the finite differences modified as follows:

\[
A_i^j f_i = \frac{A_i^{j+1} f_{i+1} - A_i^{j-1} f_i}{r_i^j - l_i^j},
\]

where \( r_i^j = \frac{x_j - 2l_i}{2}, \quad l_i^j = \frac{x_j - 2r_i}{2} \),

\( j = 1, n-1, i = 1, n-j \).

Define the middle points \( x_j^c = \frac{x_j + x_{j+1}}{2} \). Let \( f_i^c, i = 1, n-1 \) be the functions values in intermediate points \( x_j^c, i = 1, n-1 \). Then we can define the finite differences in the intermediate points as follows:

\[
A_i^j f_i^c = \frac{A_i^{j+1} f_{i+1}^c - A_i^{j-1} f_i^c}{r_i^j - l_i^j},
\]

where

\[
\hat{r}_i^j = \frac{x_j - 2l_i}{2}, \quad \hat{l}_i^j = \frac{x_j - 2r_i}{2}.
\]

\( j = 1, n-2, i = 1, n-j \).

Consider the values of the form:

\[
\hat{A}_i^j f_i^c = 2 \left( \frac{A_i^{j-2} f_{n-i}^c - A_i^{j-2} f_{n-i+1}^c}{r_i - c_i} \right) - \frac{A_i^{j-2} f_{n-i}^c - A_i^{j-2} f_{n-i+1}^c}{r_i - l_i},
\]

where

\[
r_i = \frac{x_i - 2l_i}{2}, \quad r_i = \frac{x_i - 2r_i}{2}.
\]

\( j = 1, n-1, i = 1, n-j \).

\[
c_i = \left\{ \begin{array}{ll}
\frac{x_i}{n-i}, & i = 2k, \\
\frac{x_i}{n-i+1}, & i = 2k+1.
\end{array} \right.
\]

\[
l_i = \left\{ \begin{array}{ll}
\frac{x_i}{n-i}, & i = 2k, \\
\frac{x_i}{n-i+1}, & i = 2k+1.
\end{array} \right.
\]

According to [1] if for any index \( i \) the condition

\[
\frac{A_i^{j-2} f_{n-i}^c - A_i^{j-2} f_{n-i+1}^c}{c_i - l_i} = \frac{c_i - l_i}{2}
\]

are satisfied and the condition of linearity of slope change velocity for the broken line whith nodes at the points

\( (l_i, A_i^{j-2} f_{n-i}^c) \), \( (l_i, A_i^{j-2} f_{n-i+1}^c) \), \( (c_i, A_i^{j-2} f_{n-i+2}^c) \), \( (r_i, A_i^{j-2} f_{n-i+2}^c) \) it is satisfied equality:

\[
\hat{A}_i^j f_i^c = A_i^j f_i^c.
\]

Than a procedure of finding of the function unknown value at the point \( x_n^c \) can be built as follows:

\[
A_i^{j-1} f_{n-i}^c = A_i^{j-1} f_{n-i}^c + A_i^j f_{n-i}^c (\hat{r}_i^j - \hat{l}_i^j)
\]

\( j = 1,2 \).

Consequently, the following algorithm of extrapolation was proposed:
1. Construct the table of finite difference according to (1).
2. Index \( i^* \) is finding according to the formula

\[
i^* = \text{argmin} \left\{ \frac{A_i^{j-2} f_{n-i}^c - A_i^{j-2} f_{n-i+1}^c}{c_i - l_i} \right\}
\]

\( j = 1,2 \).

3. The predictive function value formula:

\[
A_i^{j-1} f_{n-j}^c = A_i^{j-1} f_{n-j}^c + A_i^j f_{n-j}^c (\hat{r}_i^j - \hat{l}_i^j)
\]

Without loss of generality we can consider uniform grid, \( x_k - x_{k-1} = 1 \). In this case finite differences (3) can be easy to calculate. The illustration is given on the Table 1. In this transposed table, the values \( \Delta^k f_1, \Delta^k f_{l+1}, \Delta^k f_{l+2}, \Delta^k f_{l+3} \) are known. \( \Delta^k f_{l+4} \) is unknown. Other values recorded in selected cells are also unknown. According to (3) we can find \( 4\Delta^k f_{l+3} - 8\Delta^k f_{l+2} + 4\Delta^k f_{l+1} \) and it is easy to find another
unknown values according to procedure (6), for example
\[ 4\Delta^k f_{i+3} - 8\Delta^k f_{i+2} + 4\Delta^k f_{i+1} + (\Delta^k f_{i+2} - \Delta^k f_{i}) \]
and so on.

| Table 1. Illustration of the calculation (3) in the case of uniform grid |
|------------------|------------------|
| \( \Delta^k f_{i+4} \) | \( 4\Delta^k f_{i+3} - 8\Delta^k f_{i+2} + 4\Delta^k f_{i+1} + (\Delta^k f_{i+2} - \Delta^k f_{i}) \) |
| \( \Delta^k f_{i+3} \) | \( 4\Delta^k f_{i+3} - 8\Delta^k f_{i+2} + (\Delta^k f_{i+2} - \Delta^k f_{i}) \) |
| \( \Delta^k f_{i+2} \) | \( \Delta^k f_{i+3} - \Delta^k f_{i+1} + (\Delta^k f_{i+2} - \Delta^k f_{i}) \) |
| \( \Delta^k f_{i+1} \) | \( \Delta^k f_{i+3} - \Delta^k f_{i+1} + (\Delta^k f_{i+2} - \Delta^k f_{i}) \) |
| \( \Delta^k f_{i} \) | \( \Delta^k f_{i+1} - \Delta^k f_{i-1} \) |
| \( \Delta^k f_{i+4} = 4\Delta^k f_{i+3} - 8\Delta^k f_{i+2} + 4\Delta^k f_{i+1} + (\Delta^k f_{i+2} - \Delta^k f_{i}) \) |

Spatial generalization of the “pyramidal” method was proposed in [14]. To construct the “predictive” value of some surface at the selected point, it is proposed to consider paths passing through lattice nodes, where the values of the corresponding surface are known and a special parameter (measure) of the predictability of the function is determined. Then, a predictive value is the result of one-dimensional “pyramidal” approach for the function values through the path for which the degree of predictability is maximal.

![Fig.1 Illustration of the spatial generalization of the “pyramidal” method](image)

III. EQUIVALENCE OF THE RELATIONS (4)-(5) AND CUBIC PREDICTION

The main idea of the pyramidal method justification in [1] is based on the consumption that the best for the prediction row of the finite differences “pyramid” can be found if this row is linear. This is equivalent to the cubic behavior in the row \( i^2 \). But the condition for optimizing row selection requires a more detailed study.

Let \( \Delta^k f_i = z_1 \), \( \Delta^k f_{i+1} = z_2 \), \( \Delta^k f_{i+2} = z_3 \), \( \Delta^k f_{i+3} = z_4 \).

We can consider this values in the points \( x, x + d, x + 2d, x + 3d \) respectively. Thus, we'll get simple system of linear equation:

\[
\begin{align*}
&c_1 x^3 + d_1 x^2 + e_1 x + f_1 = z_1 \\
&c_1(x + d)^3 + d_1(x + d)^2 + e_1(x + d) + f_1 = z_2 \\
&c_1(x + 2d)^3 + d_1(x + 2d)^2 + e_1(x + 2d) + f_1 = z_3 \\
&c_1(x + 3d)^3 + d_1(x + 3d)^2 + e_1(x + 3d) + f_1 = z_4
\end{align*}
\]

Matrix of the coefficients of this systems is Vandermonde matrix:

\[
A = \begin{bmatrix}
 x^3 & x^2 & x & 1 \\
(x + d)^3 & (x + d)^2 & x + d & 1 \\
(x + 2d)^3 & (x + 2d)^2 & x + 2d & 1 \\
(x + 3d)^3 & (x + 3d)^2 & x + 3d & 1
\end{bmatrix}
\]

So we can use wellknown Cramer’s rule. Let consider matrix:

\[
A_i = \begin{bmatrix}
 z_1 & x^2 & x & 1 \\
 z_2 & (x + d)^2 & x + d & 1 \\
 z_3 & (x + 2d)^2 & x + 2d & 1 \\
 z_4 & (x + 3d)^2 & x + 3d & 1
\end{bmatrix}
\]
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\[ A_2 = \begin{pmatrix} x^3 & z_1 & x & 1 \\ (x+d)^3 & z_2 & x + d & 1 \\ (x+2d)^3 & z_3 & x + 2d & 1 \\ (x+3d)^3 & z_4 & x + 3d & 1 \end{pmatrix}, \]

\[ A_3 = \begin{pmatrix} x^3 & x^2 & z_1 & 1 \\ (x+\Delta)^3 & (x+\Delta)^2 & z_2 & 1 \\ (x+2\Delta)^3 & (x+2\Delta)^2 & z_3 & 1 \\ (x+3\Delta)^3 & (x+3\Delta)^2 & z_4 & 1 \end{pmatrix}, \]

\[ A_4 = \begin{pmatrix} x^3 & x^2 & x & z_1 \\ (x+\Delta)^3 & (x+\Delta)^2 & x + \Delta & z_2 \\ (x+2\Delta)^3 & (x+2\Delta)^2 & x + 2\Delta & z_3 \\ (x+3\Delta)^3 & (x+3\Delta)^2 & x + 3\Delta & z_4 \end{pmatrix}. \]

Obviously, finding the appropriate determinants is a cumbersome task. We can use symbolic operation in Mathcad environment. Thus det \( A = 12d^6 \),

\[ c_1 = \det A_1 / \det A = \frac{-6d^3x^2z_2+6d^4xz_2+6d^4x^2z_4}{(12d^6)}, \]

\[ d_1 = \det A_2 / \det A = \frac{30d^4x^2z_2-12d^4x^2z_1-24d^4x^2z_3+6d^4x^4z_4-6d^3x^3z_1+18d^3x^3z_2-18d^3x^3z_3+6d^3x^3z_4)}{1(12d^6)}, \]

\[ e_1 = \det A_3 / \det A = \frac{36d^5x^2z_2-22d^5x^2z_1-18d^5x^3z_2+4d^5x^5z_4-6d^3x^3z_1+18d^3x^3z_2-24d^3x^3z_3+6d^3x^3z_4+160d^4x^4z_2-48d^4x^4z_3+12d^4x^4z_4}{12d^6}, \]

\[ f_1 = \det A_4 / \det A = \frac{(12d^6x^2z_1+2d^4x^2z_3z_1+12d^4x^2z^2-6d^3x^3z_2-30d^4x^4z_2+6d^3x^3z_3+24d^4x^4z_3-2d^3x^3z_4-6d^4x^4z_2+22d^3x^3z_4-36d^5x^5z_2+34d^5x^5z_3-2d^3x^3z^4+12d^5x^5z_4)}{12d^6}. \]

After substitution all expression into \( c_1(x+4d)^3 \) and \( d_1(x+4d)^2 + e_1(x+4d) + f_1 \), we get:

\[ c_1(x+4d)^3 + d_1(x+4d)^2 + e_1(x+4d) + f_1 = 4xz - z_1 - 6xz + 4xz. \]

We’ve get expression (3). Thus, the task of determining the forecast value in the corresponding row of the pyramidal method is equivalent to the cubic forecast based on the last 4 values of the data series.

IV CONCLUSIONS

Thus, in this paper we show that the procedure of the predicted value constructing in the pyramidal method is equivalent to the cubic forecast. This result makes it possible to assert that the pyramidal method allows one to construct an optimal cubic forecast in a certain sense in the corresponding row of the pyramidal method of finite differences. In paper [1] authors also discuss the cubic forecast, but the corresponding statements are not strict enough. Ultimately, pyramidal method allows to build the forecast based on the optimal degree polynomial. The degree of the polynomial is chosen in accordance with minimizing the error.

The second interesting result that follows from our article is the method of constructing a cubic forecast of an arbitrary function. Indeed, to construct a cubic forecast by arbitrary 4 points, it is needed to build a fragment of the finite difference table (see Table 1) and carry out simple calculations in accordance with the procedure described in Section 2. In this case, there is no need to solve a system of linear algebraic equations to find the corresponding coefficients.

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