

EXTRAPOLATING TIME SERIES BY  
DISCOUNTED LEAST SQUARES

by

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### Abstract

An approximating function is fitted to a time series, such as daily observation. The fitting is carried out over all past time by weighted least squares with an exponential weight factor. The approximating function is restricted to be a solution of a certain linear differential equation of the  $m$ th order having constant coefficients. The solution which minimizes the least square expression can be continued into the future. In particular tomorrow's extrapolated value is defined by this continuation. To obtain an explicit solution of the problem a formula is constructed which gives the extrapolated value as a linear combination of the last  $m$  observed values and the last  $m$  extrapolated values. The coefficients of this extrapolation formula prove to be simply related to the coefficients of the differential equation. Another extrapolation formula is of vectorial nature. The components of a vector are  $m$  independent functionals of the past observations. Then tomorrow's vector is given as a linear function of today's vector and today's scalar observation.

## EXTRAPOLATING TIME SERIES BY DISCOUNTED LEAST SQUARES\*

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This paper is concerned with extrapolation of an infinite sequence  $y_1, y_2, \dots$ , of real (or complex) numbers. This is accomplished by fitting the sequence  $(y_n)$  by a function  $p(n)$  taken from a space of functions termed exponentials. The criteria for the fit is given by discounted least squares. This means that  $p(n)$  is that exponential which minimizes the 'error' expression

$$E = \sum_1^{\infty} e^{-\epsilon n} |y_n - p(n)|^2.$$

Here  $\epsilon$  is a positive constant termed the discount factor. Then the extrapolated value of the sequences at the point  $x$  is defined to be  $p(x)$ .

In the previous paper an exponential was defined as an exponential polynomial of the form

$$p(x) = \sum_1^m d_j \beta_j^x$$

where the  $\beta_j$  are fixed complex numbers assumed distinct and nonzero. The coefficients  $d_j$  are arbitrary complex numbers so an  $m$ -dimensional vector space results. In this paper the definition of exponential is extended so as to permit polynomial terms of the form  $p_j^* x^k \beta_j^*$ ,  $\dots$ . It is seen, therefore, that the space of exponentials can be defined to be the solution set of a certain linear differential equation,

$$\sum_{j=1}^m \frac{d^j p}{dx^j} + \dots + \epsilon p = 0$$

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where the coefficients  $e_i$  are complex constants. This space of exponents will be denoted as  $XP$ . An important special case is defined by the equation  $d^m p/dx^m = 0$ . Then  $XP$  is the space of polynomials of degree less than  $m$ .

There are three reasons why the extrapolation procedure just described can treat a large class of problems in applied mathematics. The first reason is that the discounted least squares criteria is suited to problems of mechanics and economics for which the progressive discount of the past seems natural. Another reason is that the space  $XP$  is invariant under an arbitrary translation of the  $x$  axis. This invariance property makes exponents attractive functions for approximating time series. The third reason for the utility of this extrapolation method is that there is an underlying algebraic structure which is both interesting and significant.

The bases  $f_i$  are arbitrary. The selection of the  $f_i$  and the discount factor  $\rho$  should take into account first the genesis of the data. Also, account must be taken of the genesis of data error and the smoothing property of the extrapolation. These questions are not treated in this paper.

A central problem of this paper is the one step extrapolation of the sequence  $\{y_n\}$  to obtain an extrapolated value at  $n = 0$ . The extrapolated value is denoted as  $y_0^*$  and is defined as  $y_0^* = p(0)$ . When the minimization is carried out, it results that  $y_0^*$  is given by a linear functional

$$y_0^* = \sum_{n=1}^{\infty} Q_n y_n.$$

Here the coefficients  $Q_n$  do not depend on the sequence

$y_1, y_2, \dots$ . It is then natural to write

$$y_k^* = \sum_{n=1}^{\infty} Q_n y_{n+k}$$

and define  $y_k^*$  as the predicted value of  $y_k$  based only on the 'previous values'  $y_{k+1}, y_{k+2}, \dots$

Theorem 3b to follow gives a simple generating function to evaluate the coefficients  $Q_n$ . Thereby the problem of one-step extrapolation is essentially solved. Nevertheless, the above formula is not satisfactory from a computational point of view. This is so because it is an infinite series and so an infinite memory is needed. This situation is remedied by Theorem 2 which provides the following short memory formula,

$$*f = \sum_{n=1}^m T_n y_{n+k} + \sum_{n=1}^m I_n y_{n+k}^*$$

Here the  $T_n$  and the  $I_n$  are constants which do not depend on the sequence  $y_n$ . Thus, this identity gives the extrapolation  $y_k^*$  as a linear combination of the  $m$  previous values and the  $m$  previous extrapolations. Consequently, the short memory formula is readily adaptable for computer evaluation.

Also of importance are extrapolation formulas for other linear functionals of the sequence  $y_1, y_2, \dots$ . For example  $\int_0^1 p(x) dx$ ,  $p(1/2)$ ,  $p'(0)$ ,  $p''(-2)$  are linear functionals. We term such functions extrapolators. Let  $w(x)$  be a vector whose  $m$  components are linearly independent extrapolators. Then Theorem 7 gives the following extrapolation formula,

$$w(x) = Aw(x+1) + by_{x+1}$$

where  $A$  is constant matrix and  $b$  is a constant vector. This is termed a very short memory formula because the extrapolation

is based only on one previous value of  $w$  and one previous value of  $y$ .

The extrapolation of time series based on a discounted least squares criteria has previously been treated by Duffin and Schmidt [1], Duffin and Whidden [2], and Morrison [3].

(Various other authors have proposed similar extrapolation formulae but their work is not based on discounted least squares.) The present paper gives a more general treatment of problems posed by references [1], [2], and [3]. In particular the theorems of this paper are aimed at evaluating and interrelating the constants,  $p_i$ ,  $\theta$ ,  $Q_n$ ,  $T_n$ ,  $S_n^* A$ , and  $\vec{b}$  by algebraic formulae.

To begin the proof, let  $G(z)$  be a polynomial of degree  $m$  in the variable  $z$  and let  $G(0) = 1$ . Then

$$G(z) = \sum_{j=0}^m g_j z^j.$$

Here the  $g_j$  are arbitrary complex constants except that  $g_0 = 1$  and  $g_m \neq 0$ . Of central concern are functions  $p(x)$  of the real variable  $x$  satisfying the difference equation

$$0 = g_0 p(x) + g_1 p(x+1) + \dots + g_m p(x+m).$$

Let  $X$  denote the translation operation defined by the relation  $Xf(x) = f(x+1)$ . Then the difference equation can be written in operation form as

$$0 = G(X)p(x) = \sum_{j=1}^m (1 - \lambda_j^{-1} X) p(x)$$

where the  $\lambda_j$  are the roots of  $G(z) = 0$ . First suppose that  $x$  is restricted to the integers. If none of the roots are repeated it is seen that there are  $m$  linearly independent

solutions having the form  $p(x) = \lambda^x$ . The general solution is a linear combination of such solutions. If the root  $\lambda_j$  is repeated  $k$  times there are  $k$  linearly independent solutions:

$\lambda_j^x, x\lambda_j^{x-1}, \dots, x^{k-1}\lambda_j^{x-k}$ . The general solution is again a linear combination of such solutions and defines an  $m$ -dimensional vector space. This is the space of exponentials and it is denoted as  $XP$ .

For some problems of extrapolation it is required to continue exponentials to non-integral values. One way to do this is to let  $\lambda = e^{B}$ , and then define  $\lambda^x = e^{Bx}$  where  $L$  is an integer. A "natural" choice for  $L$  is

$$-\pi \leq \text{imag } B + 2\pi L \leq \pi.$$

This choice minimizes oscillation. The ambiguity here stems from the fact that multiplying a solution of the equation by an arbitrary function of period 1 gives another solution.

Since an exponential satisfies the difference equation it is seen that prescribing the value of the exponential at the integers  $1, 2, \dots, m$  determines the value of the exponential for all other positive integers. Thus the space  $XP$  is  $m$ -dimensional even if  $x$  is restricted to the integers  $1, 2, \dots, m$ .

The approximation scheme employed in this paper results from embedding the finite dimensional space  $XP$  in an infinite Hilbert space of "discounted squares". This Hilbert space is denoted as  $DS$ . The elements of  $DS$  are infinite sequences of complex numbers  $y_1, y_2, \dots$ , such that

$$\sum_{n=1}^{\infty} \alpha^n |y_n|^2 < CD.$$

Here  $\alpha$  is a positive constant termed the discount factor. The discount factor is required to satisfy the inequality

$$0 < |\epsilon_j| < 1; j = 1, 2, \dots, m.$$

It is readily shown from this inequality that exponentials are in DS. Thus the space XP is a finite dimensional subspace of DS. Let v and y denote two elements of DS then the Hermitian bilinear form is defined as

$$[v, y] = \sum_{n=1}^{\infty} \epsilon^n \bar{v}_n y_n$$

where  $\bar{v}_n$  denotes the complex conjugate of  $v_n$ . The norm of a element y of DS is defined as

$$\|y\| = [y, y]^{1/2}.$$

It is instructive in what follows to regard the sequence  $\{y_n\}$  as a time series of observations'. Thus  $y_x$  can be regarded as the value of the observation at time -x. To extrapolate the sequence  $\{y_n\}$  for values of n not a positive integer we first approximate y by an element p of the subspace XP. By this is meant that p is chosen to minimize the expression

$$E = \sum_{n=1}^{\infty} \epsilon^n |y_n - p(n)|^2.$$

As in the references [1], [2], and [3] this is termed approximation by ~~discounted least squares~~.

The following lemma is aimed at determining p given a sequence  $\{y_n\}$ .

~~Lemma 1. Let p be the exponential best approximating y in the Hilbert space. Then~~

$$[r, y] = [r, p]$$

~~for every exponential r.~~

Proof. This is merely a reformulation of the basic theorem concerning Hilbert space which states that if p is the best approximation



to  $y$  for  $p$  constrained to a subspace then  $p$  is the orthogonal projection of  $y$  into the subspace.

By choosing a basis for the subspace  $XP$  the orthogonality relation stated by this lemma leads to a system of  $m$  linear equations which could be used to determine  $p$  as a unique linear combination of the basis elements.

Then  $p$  is determined uniquely at the positive integer points. The convention introduced above permits an exponential, given at the positive integer points, to be determined for all real  $x$ . Thus it is possible to define the extrapolation of the sequence  $y^{\wedge}y^{\wedge}****$  at the point  $x$  to be  $p(x)$ .

Lemma 2. Let  $r$  be a given exponential and let  $k$  be a given number. Then there is a unique sequence of numbers

$c_1, c_2, \dots, c_m$  such that the formula

$$[r, v] = c_1 v(k+1) + c_2 v(k+2) + \dots + c_m v(k+m)$$

holds for every exponential  $v$ .

Conversely given a number  $k$  and a sequence of numbers

$c_1, c_2, \dots, c_m$  there is a unique exponential  $r$  for which the above formula holds»

Proof. As is well known the linear functionals  $f(v)$  in an  $m$ -dimensional space such as  $XP$  have the form  $f(v) = [r, v]$ .

Moreover, these linear functionals themselves form an  $m$ -dimensional space. Clearly the expression on the right side of the formula

of the lemma is a linear functional for any choice of constants

$c_1, c_2, \dots, c_m$ . Suppose that for some choice  $c_1 v(k+1) + \dots + c_m v(k+m)$

vanishes for all  $v$  in  $XP$ . However, an exponential can be prescribed arbitrarily at a sequence of  $m$  points obtained by

successive unit translations to the right. Hence all the constants

$c_i^*$  must be zero. This shows that  $v(k+1), v(k+2), \dots, v(k+m)$  are  $m$  independent functionals. There can be no more than  $m$  independent functionals and consequently the left and right side of the formula represent the same space of linear functionals. The proof of the lemma follows from this observation.

Theorem 1. Let  $r$  be a given exponential and let  $k$  be a given number, then there is a unique sequence of numbers  $c_1, c_2, \dots, c_m$  such that the formula

$$[r, y] = c_1 p(k+1) + c_2 p(k+2) + \dots + c_m p(k+m)$$

holds for every  $y$  of the Hilbert space  $DS$  provided that  $p$  is the orthogonal projection of  $y$  into the exponential subspace  $XP$ .

Conversely given a sequence of numbers  $c_1, c_2, \dots, c_m$  there is a unique exponential  $r$  satisfying this formula.

Proof. By Lemma 1 we have  $[r, y] = [r, p]$ . Then apply Lemma 2 with  $v = p$  and the proof is seen to be complete.

A general problem of extrapolation is to extrapolate the sequence  $y^1, y^2, \dots$  to obtain a value for  $y$  at a point  $x$  not a positive integer. This extrapolated value is denoted by the symbol  $\text{ext}(y)$  and is defined as  $\text{ext}(y) = p(x)$ . The

following corollary of Theorem 1 gives a formula for computing  $\text{ext}(y)$ .

Corollary 1. Let  $p(x)$  be the approximating exponential to the sequence  $y^1, y^2, \dots$ . Then there is a kernel function  $q(x, n)$  such that

$$p(x) = \sum_{n=1}^{\infty} q(x, n) y_n = [q, y].$$

For  $n$  fixed  $q(x, n)$  is an exponential in  $x$ . For  $x$  fixed

$\bar{q}(x^n)$  is an exponential in n. The kernel function may depend on 0 but not on y.

Proof. In Theorem 1 take  $k = x - 1$ ,  $c_1 = 1$ ,  $c_2 = 0, \dots, c_m = 0$ . This proves the formula of Corollary 1 with  $\bar{q} = r$ . To show that the kernel function is an exponential in  $x$  take  $y_n = 0$  except for  $n = n_0$ . This is seen to complete the proof.

A significant special case of the formula of Corollary 1 is  $x = 0$ . This extrapolates the sequence  $y_{-1}, y_0, y_1, \dots$  one unit to the left to obtain  $ext(y_0)$ . This one step extrapolation is sufficiently important to warrant a special notation and we write  $y_0^* = p(0)$ . The general extrapolation formula becomes

$$y_0^* = \sum_{n=1}^{\infty} e^{-n} q(n) y_n.$$

Here  $q(n) = q(0, n)$  and  $q(n)$  is also termed a kernel.

If the sequence  $y_0, y_1, \dots$  is in the Hilbert space  $DS$  then the sequences  $y_{x-1}, y_{x+2}, \dots$  is also in  $DS$  for any positive or negative integer  $x$ . This follows from the relation

$$\sum_{n=1}^{\infty} e^{-n} |y_{x+n}|^2 = e^{-k} \sum_{n=1}^{\infty} e^{-n} |y_n|^2 < \infty.$$

As a natural extension of the previous notation let  $y_x^*$  denote the extrapolation of the sequence  $y_{x-1}, y_{x+2}, \dots$  to the point  $x$ . The extrapolation formula given by discounted least squares is

$$y_x^* = \sum_{n=1}^{\infty} e^{-n} q_x(n) y_{n+x}$$

This formula involves all values of the sequence  $y_{x-1}, y_{x+2}, \dots$  and so is an infinite series. For this reason a formula of this nature may be termed a long memory formula. It is now to be shown that  $y_x^*$  is also given by a finite recursion relation and hence by a short memory formula.

Theorem 2. Let  $y_k^*$  be defined by the formula

$$y_k^* = \sum_{n=1}^{CD} \theta^n q(n) y_{n+k}.$$

Then  $y_k^*$  the one step extrapolation, is given by the recursion formula

$$y_k^* = - \sum_{n=1}^m g_n y_{n+k} - \sum_{n=1}^m f_n \theta^n \delta_{n+k}.$$

Here, for convenience of notation,  $f_n = \bar{g}_{n-m} / \bar{g}_m$ . Also  $\delta_n = y_n^* - y_n$  and is termed the discrepancy.

Proof. Let  $q^1 = 0$  for  $n < 0$  and  $q^1 = q$  for  $n > 0$  so

$$y_x^* = Z \cdot \bigoplus_{-CD}^n q^1(n) W_{n+x},$$

$$(\theta^x y_x^*) = \sum_{-CD}^{00} \theta^n q^1(n-x) y_n.$$

Since  $\bar{q}(x)$  is a polynomial  $G(X)\bar{q}(x) = 0$ . Thus  $\sum_{j=0}^m \bar{g}_j q(x+j) = 0$

or  $Z_0 g^{m-j} q(x+m-j) = 0$ . A polynomial  $F(z)$  related to the polynomial  $G(z)$  plays an important role in what follows. It is defined as

$$F(z) = \sum_{j=0}^m f_j z^j = \sum_{j=0}^m \bar{g}_{-j} z^j / \bar{g}_m.$$

It is seen that  $F(X)q(n-x) = 0$  for all  $n$ . Consequently

$F(X)q^1(n-x) = 0$  for  $n > x+m$  or  $n \leq x$  so

$$F(X) \theta^x y_x^* = \sum_{n=-\infty}^{CD} \sum_{j=0}^m [H f_j \theta^{q^1(n-x-j)}] \theta^n y_n$$

$$= \sum_{n=x+1}^{x+m} s_n y_n$$

where the  $s_n$  are certain absolute constants. To evaluate these constants,, first set  $x = 0$  in the last relation and obtain

$$(*) \quad \sum_{j=0}^m f_j \theta^j y_j^* = \sum_{n=1}^m s_n y_n.$$

Next let  $y_k = r(k)$  an exponential. Since the extrapolation of exponentials is error-free,  $y_k^* = r(k)$  also. Note that  $f_0 = 1$  so after substituting  $y_k = r(k)$  the relation (\*) can be written as

$$r(0) = \sum_{j=1}^m (s_j - f_j \cdot 9^j) r(j).$$

But since  $G(X)r(x) = 0$  it is also true that

$$r(0) = - \sum_{j=1}^m g_j r(j).$$

Subtracting these two equations for  $r(0)$  gives

$$0 = \sum_{j=1}^m (s_j - f_j G^j + g_j) r(j).$$

But an exponential can be defined arbitrarily on the integers  $1, 2, \dots, m$ . Thus  $s_j = f_j \cdot 9^j - g_j$ , and relation (\*) becomes

$$y_0^* = \sum_{j=1}^m (f_j \cdot e^D - g_j) y_j - \sum_{j=1}^m f_j \theta^j y_j^*.$$

This is seen to be equivalent to the short memory formula stated in Theorem 2.

Corollary 2. The kernel function  $q(x)$  satisfies the constant coefficient difference equation  $G(X) \tilde{q}(x) = 0$ , an equation of the  $m$ th order. But  $q(x)$  does not satisfy any such equation of lower order.

Proof. Suppose  $q(x)$  satisfied the equation  $G^{fT}(X) \tilde{q}(x) = 0$  of order  $m^{fT} < m$ . Then the proof of Theorem 2 could be carried out with  $G^{fT}$  replacing  $G$ . This would lead to a relation of the form

$$r(0) = \sum_{j=1}^{m^{fT}} (f_j \cdot 9^j - g_j) r(j)$$

for every exponential  $r(j)$ . But an exponential can be defined

arbitrarily at  $m$  successive points. Thus take  $r(0) = 1$  but  $r(j) = 0$  for  $j = 1, \dots, m$ . This contradicts the assumed relation and the proof is complete.

So far the kernel function  $q(n)$  of the long formula has only been defined implicitly. On the other hand the coefficients of the short formula are given explicitly in Theorem 2. However since the short memory formula and the long memory formula are essentially equivalent it is possible to use the short formula to give an explicit procedure for evaluating  $q(n)$ . Different ways of doing this are given in Theorem 3a and Theorem 3b to follow.

Theorem 3a. The kernel function  $q(n)$  satisfies the recursion formulae:

$$q(n) = f_n - \sum_{j=1}^{n-1} \theta^j g_j q(n-j), \quad 1 \leq n \leq m$$

$$q(n) = -\sum_{j=1}^m \theta^j g_j q(n-j), \quad n > m.$$

Thus

$$q(1) = f_1 - \theta g_1$$

$$q(2) = f_2 - \theta^2 g_2 - f_1^2 + \theta f_1 g_1$$

$$q(3) = f_3 - \theta^3 g_3 - 2f_1 f_2 + f_1^2 - \theta f_1^2 + \theta^2 g_2^2 + \theta f_2 g_1$$

Proof. Let  $y_j = 0$  for all  $j$  except that  $y_n = 1$ . Then it follows directly from the long formula that  $y_n^* = G^n q(n)$ . It is also seen that:  $y_j^* = \theta^{n-j} q(n-j)$  if  $j < n - 1$ ,  $y_j^* = 0$  if  $j \geq n$ . The short formula may be written in the form

$$y_n^* = \sum_{j=1}^m (\theta^j f_j - g_j) y_j - \sum_{j=1}^m \theta^j f_j y_j^*$$

Substituting the above special values of  $y_j$  and  $y_j^*$  in this formula leads directly to the formulae of Theorem 3a.

Theorem 3b. The kernel function  $q(n)$  satisfies the generating identity

$$\frac{F(Qz) - G(z)}{F(\theta z)} = \sum_{n=1}^{\infty} \theta^n q(n) z^n$$

provided  $|z|$  is small.

Proof. If  $|z|$  is small the sequence  $1, z, z^2, \dots$  is in the Hilbert space DS. Thus, substituting  $y_n = z^n$  in the long formula gives

$$y_n^* = \sum_{j=1}^{\infty} q(j) z^j = z y_n^*$$

Then substituting  $y_n = z^n$  and  $y_n^* = z^n$  in the short formula gives

$$y_n^* = \sum_{j=1}^m (e^{f_j} - g_j z^{j-1}) z^j = \sum_{j=1}^m (e^{f_j} - g_j z^{j-1}) z^j$$

But  $f_0 = g_0 = 1$  so

$$y_n^* = \sum_{j=0}^m f_j z^j = \sum_{j=0}^m f_j z^j - \sum_{j=0}^m g_j z^j$$

and the proof is complete.

We now turn from one-step extrapolation to multi-step extrapolation. Thus the two-step extrapolation of the sequence  $y_n, y_{n+1}, y_{n+2}, \dots$  is given by  $\text{ext}(y_{n+1}) = p(-1)$  etc.

Theorem 4a. Multistep extrapolators of the sequence  $y_n, y_{n+1}, y_{n+2}, \dots$  are given by the long memory formulae:

$$p(0) = \sum_{n=1}^{\infty} q_n y_n$$

$$p(-1) = \sum_{n=1}^{\infty} \theta^{n+1} (q_{n+1} + q_1 q_n) y_n$$

$$p(-2) = \sum_{n=1}^{\infty} \theta^{n+2} (q_{n+2} + q_1 q_{n+1} + q_2 q_n + q_1^2 q_n) y_n$$

$$p(-3) = \sum_{n=1}^{\infty} e^{n+3} [q_{n+3} + q_1 q_{n+2} + (q_1^2 + q_2) q_{n+1} + (q_1^3 + 2q_1 q_2 + q_3) q_n] y_n.$$

Here  $q_n$  is a condensed notation for the kernel  $q(n)$ .

Proof. Let  $Y_n = y_n$  for  $n > 0$  and let  $Y_n = p(n)$  for  $n \leq 0$ .

Let  $E_0 = \sum_{n=0}^{\infty} |Y_n - P(n)|^2$  where  $P(n)$  is an exponential. Thus

$$E_0 = |p(0) - P(0)|^2 + \sum_{n=1}^{\infty} e^n |y_n - P(n)|^2.$$

Clearly  $E_0$  is minimized when  $P(n) = p(n)$ . Hence applying the one-step extrapolation formula to the sequence  $Y_0, Y_1, Y_2, \dots$  gives

$$p(-1) = e^{q_1} p(0) + \sum_{n=2}^{\infty} e^{n q_n} y_{n-1}.$$

But

$$p(0) = \sum_{n=1}^{\infty} e^{n q_n} y_n$$

so

$$p(-1) = \sum_{n=1}^{\infty} e^{n+1} q_{n+1} y_n + \sum_{n=1}^{\infty} e^{n+1} q_{n+1} y_n.$$

This proves the formula for the two-step extrapolator.

Next consider  $E_{-1} = \sum_{n=-1}^{\infty} |Y_n - P(n)|^2$  and following similar reasoning to that given above shows that

$$p(-2) = \sum_{n=1}^{\infty} e^{n q_n} y_{n-2}$$

$$p(-2) = e^{q_1} p(-1) + e^{2q_2} p(0) + \sum_{n=3}^{\infty} e^{n q_n} y_{n-2}$$

Substituting the series expression just derived for  $p(-1)$  is seen to prove the formula given for the three-step extrapolation. Further formulae are derived analogously and the proof is complete.



Theorem 4b. Multistep extrapolators of the sequence  $Y_1, Y_2, \dots$   
are given by the short memory formulae;

$$\begin{aligned}
 -p(0) &= \begin{bmatrix} g_1 y_1 + I e^{n_1} \delta_1 \\ \vdots \\ g_m y_m + I e^{n_m} \delta_m \end{bmatrix} \\
 -p(-1) &= \begin{bmatrix} g_1 y_1 - g_1 g_n y_n + \sum_{l=1}^m e^{n_l} (\theta f_{n+1} - g_1 f_n) \delta_n \\ \vdots \\ g_m y_m - g_1 g_n y_n + \sum_{l=1}^m e^{n_l} (\theta f_{n+1} - g_1 f_n) \delta_n \end{bmatrix} \\
 -p(-2) &= \begin{bmatrix} g_{n+2} - g_1 g_{n+1} - g_2 g_n + g_1^2 g_n y_n \\ \vdots \\ g_m y_m - g_1 g_n y_n + \sum_{l=1}^m e^{n_l} (\theta^2 f_{n+2} - \theta g_1 f_{n+1} - g_2 f_n + g_1^2 f_n) \delta_n \end{bmatrix}
 \end{aligned}$$

where  $g_n = 0$  and  $f_n = 0$   $\forall n > m$ .

Proof: Apply the short memory formula to the sequence  $Y_0, Y_1, \dots$  which was introduced in the proof of Theorem 4a. But  $Y_{-1}^* = p(-1)$

so

$$* \quad \tilde{L}^{-m} \wedge_{n, n-1} \sum_{l=1}^m e^{n_l} \Delta_{n-1}$$

Also  $Y_0 = p(0)$ ,  $\wedge_0 = 0$  and so

$$\begin{aligned}
 -p(-D) &= \begin{bmatrix} g^{\wedge 0} \\ \vdots \\ g^{\wedge m} \end{bmatrix} + \begin{bmatrix} g_1 y_1 \\ \vdots \\ g_m y_m \end{bmatrix} + \begin{bmatrix} e^{n_1} \delta_1 \\ \vdots \\ e^{n_m} \delta_m \end{bmatrix} \\
 -p(-1) &= \begin{bmatrix} g_1 y_1 \\ \vdots \\ g_m y_m \end{bmatrix} + \begin{bmatrix} e^{n_1} \delta_1 \\ \vdots \\ e^{n_m} \delta_m \end{bmatrix} + \sum_{l=1}^{m-1} g_{n+1} y_n + \\
 &\quad \sum_{l=1}^{m-1} g_{n+1}^2 y_n
 \end{aligned}$$

This seen to prove the stated formulae for  $p(-1)$ . Further formulae are derived analogously.

Theorem 5. The relation

$$F(\odot z) \sim \wedge e^{t(n)z} z^n$$

gives a one to one linear correspondence between polynomials of the form  $\hat{L}(z) = \sum_{j=0}^m a_j z^{-j}$  and conjugate exponentials  $t(n)$ .

Proof. Given such a polynomial  $\hat{L}(z)$  then  $T(z) = H(z)/F(z)$  is a convergent power series if  $|z|$  is sufficiently small. Thus write

$$T(z) = \sum_{n=0}^{\infty} t(n) z^n$$

where  $t(n) = 0$  for  $n < 0$ . Since  $\hat{L}(z) = F(z)T(z)$  we have

$$\hat{L}(z) = \sum_{n=0}^m \sum_{j=-\infty}^{\infty} f_j e^{j+n} z^{j+n} t(n),$$

Let  $j + n = k$  so

$$\hat{L}(z) = \sum_{k=-\infty}^{\infty} e^k z^k \sum_{j=0}^m f_j t(k-j)$$

Since  $\hat{L}(z)$  is a polynomial of degree not exceeding  $m$  it follows from the above relation that

$$\sum_{j=0}^m f_j t(k-j) = 0 \quad \text{for } k > m, \text{ or}$$

$$\sum_{j=0}^m f_j t(x+m-j) = 0 \quad \text{for } x > 0, \text{ so}$$

$$\sum_{j=0}^m f_{m-j} t(x+j) = 0 \quad \text{for } x > 0.$$

Hence  $G(x)t(x) = 0$  and it follows that  $t(x)$  is an exponential for  $x > 0$ .

We can conclude from this last result that the  $m$ -dimensional vector space  $P$  of polynomials of the form  $\hat{L}(z)$  is mapped linearly into a space  $S$  of conjugate exponentials. Moreover, this is a one to one mapping because power series are unique so  $S$  has

dimension  $m$ . However, the space  $XP$  of all conjugate exponentials has dimension  $m$  so  $S = XP$  and the proof is complete.

Lemma 3. Let  $l(z) = \sum_{i=0}^m a_i z^{-i}$  be a given polynomial. Then  
~~there are polynomials~~

$$H(z) = \sum_{j=1}^m h_j z^j \quad \text{and} \quad K(z) = \sum_{j=1}^m k_j z^j$$

such that

$$*C(z) = -H(z)G(z) + K(z)F(0z).$$

Hence the system of  $2m$  equations

$$a_i A_{\pm} = - \sum_{j=1}^i g_i^{j h_j} + \sum_{j=1}^{i-1} e^{i-j} f_{i-j}^{k_j} \quad * \quad x \leq i \leq 2m$$

can be used to find the coefficients  $h_j$  and  $k_j$ . Here  $A_j = 1$   
for  $j < m$  and  $A_m = 0$  for  $j > m$ .

Proof. The roots of  $G(z)$  are  $\{\beta_j\}$  and the roots of  $F(0z)$  are  $\{\alpha_j\}$ . The discount factor  $\beta$  was chosen so that  
count factor

It follows that the roots of  $G(z)$  are outside a circle of radius  $1/2$  and the roots of  $F(0z)$  are inside this circle. Consequently  $G(z)$  and  $F(0z)$  cannot have a common root so by a basic theorem in the algebra of polynomials

$$*C(z)/z = -H_0(z)G(z) + K_0(z)F(Gz)$$

where  $H_0(z)$  and  $K_0(z)$  are polynomials of degree less than the degree of  $G$  and  $F$ . Of course  $G$  and  $F$  are of degree  $m$  so multiplying through by  $z$  leads to the relation stated in the lemma.

Theorem 6. Given an arbitrary exponential  $t(n)$  let  $w(x)$  be the corresponding extrapolator

$$w(x) = \sum_{n=1}^{\infty} e^{nt} t(n) y_{n+x}.$$

Then a short formula for this extrapolator is

$$w(x) = \sum_{j=1}^m h_j \delta_{j+x} + \sum_{j=1}^m k_j y_{j+x}.$$

~~Here  $h_j$  and  $k_j$  are coefficients defined by the polynomial  $\phi(z)$  which is the image of  $t(n)$  according to Theorem 5 and Lemma 3.~~

Proof. Let  $Q(z) = \sum_{n=1}^{2L} q(n) z^n$  then according to Theorem 3b

$1 - Q(z) = G(z)/F(ez)$ . Let  $T(z) = \sum_{n=1}^{\infty} e^{nt} z^n$ . Then according to Theorem 5 we have  $T(z) = C(z)/F(Gz)$ . Then Lemma 3 with  $X$  replacing  $z$ , gives

$$T(X) = H(X) [Q(X) - 1] + K(X).$$

Here  $X$  is interpreted as the translation operator. Operating on the function  $y_x$  with the above identity we see that  $[Q(X) - 1]y_x =$

•6 and so the short formula follows. This formalism is justified by the absolute convergence of the resulting series when  $y$  is in the Hilbert space and so the proof is complete.

A different proof for the existence of the short formula for  $w(x)$  results from combining Theorem 1 with  $k = -m$  and Theorem 4b.

The short memory formulae given in Theorem 2 and Theorem 6 may be termed  $m$ th order formulae because the right sides are expressed as translation operators of order  $m$ . Thus the extrapolator of Theorem 6 can be written as

$$w(x) = \left( \sum_{j=1}^m h_j X^j \right) \delta_{x+} + \left( \sum_{j=1}^m k_j X^j \right) y_x.$$

Now an  $m$ th order scalar difference equation is equivalent to a first-order vector difference equation. This suggests that the

mth order scalar extrapolation formula can be replaced by a first order vector extrapolation formula. In fact such formulae are to be found in the paper by Morrison [3]. for the special case of polynomial extrapolation. Moreover, Morrison's paper indicates that first order vector extrapolators may be advantageous in numerical work because of economy in memory.

The following is a general theorem on first order vector extrapolation formula.

Theorem 7. Let  $w_1(x), \dots, w_m(x)$  denote a set of  $m$  independent extrapolators. Then there is a set of constants  $a_{1j}$  and  $b_i$  such that

$$w_i(x) = 21^{a_i - M^w}; (x+1) + b_i y_{x+1}; i = 1, \dots, m.$$

This is termed a very short memory formula.

Proof. Then  $w_i(x) = \sum_{j=1}^m a_{ij} r_j(n) y_{n+x+1}$  and the exponentials  $f_1(n), f_2(n), \dots, f_m(n)$  form a basis for the space  $XP$ . Hence it is possible to find constants  $a_{ij}$  such that

$$Gr_i(n+1) = \sum_{j=1}^m a_{ij} r_j(n) y_{n+x+1}$$
 because  $0r_i(n+1)$  is also an exponential. These identities are multiplied by  $0^n y_{n+x+1}$  and summed. This gives

$$\sum_{n=1}^{\infty} 0^n r_i(n) y_{n+x+1} = \sum_{j=1}^m a_{ij} w_j(x+1) + 0r_i(1) y_{x+1}$$

Let  $b_i = 0r_i(1)$  and this is seen to complete the proof of the very short formula.

Corollary 3. The extrapolation of  $y_n$  at  $m$  successive points is given by a very short memory formula.

Proof. The extrapolation of  $y_n$  at  $m$  successive points is  $p(x+1), p(x+2), \dots, p(x+m)$ . By virtue of Theorem 1 these are independent extrapolators and the proof is complete.

Lemma 4. Let  $x_0$  denote an arbitrary fixed point and let  $v(x)$  denote a function of the exponential space  $XP$ . Then

$$v(x_0), \int_{x=x_0}^{\dots} \frac{dx}{dx} \dots \int_{x=x_0}^{\dots} \frac{dx}{dx} \dots$$

~~are a set of  $m$  independent linear functionals on the space  $XP$ .~~

Proof. The above expressions are obviously linear functionals.

If they were not independent then there would be an identity of the form

$$\sum_{i=0}^{m-1} c_i v^{(i)}(x) = 0; x = x_0$$

which holds for  $v$  in  $XP$ . But if  $v(x)$  is in  $XP$  so also is  $v(x+k)$  for every  $k$ . Hence the above identity actually holds for all  $x$ . It is a differential equation of order  $m-1$  and can have at most  $m-1$  independent solutions. This is a contradiction, since there are  $m$  independent functions in  $XP$ ,

~~Corollary 4. The extrapolation of  $y$  together with its derivatives up to order  $m-1$  evaluated of the same point is given by the very short memory formula.~~

Proof. The extrapolators are  $p(x), p^{*1}(x), \dots, p^{m-1}(x)$  evaluated at a certain point  $x = x_0$ . Then the corollary follows from Lemma 4 and Theorem 7.

#### REFERENCES

1. Duffin, R. J. and Schmidt, Th. W., 'An extrapolator and scrutator', J. Math. Anal, and Appl. JL (1960), 215-224.
2. Duffin, R. J. and Whidden, Phillips, 'An exponential extrapolator', J. Math. Anal, and Appl. 3^ (1961), 526-536.

3. Morrison, Norman, <sup>T</sup>Smoothing and extrapolation of time series by means of discrete Laguerre Polynomials<sup>1</sup>, SIAM Jour, (to appear).