

# NAG Library Function Document

## nag\_1d\_cheb\_fit (e02adc)

### 1 Purpose

nag\_1d\_cheb\_fit (e02adc) computes weighted least squares polynomial approximations to an arbitrary set of data points.

### 2 Specification

```
#include <nag.h>
#include <nage02.h>

void nag_1d_cheb_fit (Integer m, Integer kplus1, Integer tda,
    const double x[], const double y[], const double w[], double a[],
    double s[], NagError *fail)
```

### 3 Description

nag\_1d\_cheb\_fit (e02adc) determines least squares polynomial approximations of degrees  $0, 1, \dots, k$  to the set of data points  $(x_r, y_r)$  with weights  $w_r$ , for  $r = 1, 2, \dots, m$ .

The approximation of degree  $i$  has the property that it minimizes  $\sigma_i$  the sum of squares of the weighted residuals  $\epsilon_r$ , where

$$\epsilon_r = w_r(y_r - f_r)$$

and  $f_r$  is the value of the polynomial of degree  $i$  at the  $r$ th data point.

Each polynomial is represented in Chebyshev series form with normalized argument  $\bar{x}$ . This argument lies in the range  $-1$  to  $+1$  and is related to the original variable  $x$  by the linear transformation

$$\bar{x} = \frac{(2x - x_{\max} - x_{\min})}{(x_{\max} - x_{\min})}.$$

Here  $x_{\max}$  and  $x_{\min}$  are respectively the largest and smallest values of  $x_r$ . The polynomial approximation of degree  $i$  is represented as

$$\frac{1}{2}a_{i+1,1}T_0(\bar{x}) + a_{i+1,2}T_1(\bar{x}) + a_{i+1,3}T_2(\bar{x}) + \dots + a_{i+1,i+1}T_i(\bar{x}),$$

where  $T_j(\bar{x})$  is the Chebyshev polynomial of the first kind of degree  $j$  with argument  $\bar{x}$ .

For  $i = 0, 1, \dots, k$ , the function produces the values of  $a_{i+1,j+1}$ , for  $j = 0, 1, \dots, i$ , together with the value of the root mean square residual  $s_i = \sqrt{\sigma_i/(m - i - 1)}$ . In the case  $m = i + 1$  the function sets the value of  $s_i$  to zero.

The method employed is due to Forsythe (1957) and is based upon the generation of a set of polynomials orthogonal with respect to summation over the normalized dataset. The extensions due to Clenshaw (1960) to represent these polynomials as well as the approximating polynomials in their Chebyshev series forms are incorporated. The modifications suggested by Reinsch and Gentleman (Gentleman (1969)) to the method originally employed by Clenshaw for evaluating the orthogonal polynomials from their Chebyshev series representations are used to give greater numerical stability.

For further details of the algorithm and its use see Cox (1974), Cox and Hayes (1973).

Subsequent evaluation of the Chebyshev series representations of the polynomial approximations should be carried out using nag\_1d\_cheb\_eval (e02aec).

## 4 References

Clenshaw C W (1960) Curve fitting with a digital computer *Comput. J.* **2** 170–173

Cox M G (1974) A data-fitting package for the non-specialist user *Software for Numerical Mathematics* (ed D J Evans) Academic Press

Cox M G and Hayes J G (1973) Curve fitting: a guide and suite of algorithms for the non-specialist user *NPL Report NAC26* National Physical Laboratory

Forsythe G E (1957) Generation and use of orthogonal polynomials for data fitting with a digital computer *J. Soc. Indust. Appl. Math.* **5** 74–88

Gentleman W M (1969) An error analysis of Goertzel's (Watt's) method for computing Fourier coefficients *Comput. J.* **12** 160–165

Hayes J G (ed.) (1970) *Numerical Approximation to Functions and Data* Athlone Press, London

## 5 Arguments

- 1: **m** – Integer *Input*  
*On entry:* the number  $m$  of data points.  
*Constraint:*  $\mathbf{m} \geq \mathit{mdist} \geq 2$ , where  $\mathit{mdist}$  is the number of distinct  $x$  values in the data.
  
- 2: **kplus1** – Integer *Input*  
*On entry:*  $k + 1$ , where  $k$  is the maximum degree required.  
*Constraint:*  $0 < \mathbf{kplus1} \leq \mathit{mdist}$ , where  $\mathit{mdist}$  is the number of distinct  $x$  values in the data.
  
- 3: **tda** – Integer *Input*  
*On entry:* the stride separating matrix column elements in the array **a**.  
*Constraint:*  $\mathbf{tda} \geq \mathbf{kplus1}$ .
  
- 4: **x[m]** – const double *Input*  
*On entry:* the values  $x_r$  of the independent variable, for  $r = 1, 2, \dots, m$ .  
*Constraint:* the values must be supplied in nondecreasing order with  $\mathbf{x}[m - 1] > \mathbf{x}[0]$ .
  
- 5: **y[m]** – const double *Input*  
*On entry:* the values  $y_r$  of the dependent variable, for  $r = 1, 2, \dots, m$ .
  
- 6: **w[m]** – const double *Input*  
*On entry:* the set of weights,  $w_r$ , for  $r = 1, 2, \dots, m$ . For advice on the choice of weights, see the e02 Chapter Introduction.  
*Constraint:*  $\mathbf{w}[r] > 0.0$ , for  $r = 0, 1, \dots, \mathbf{m} - 1$ .
  
- 7: **a[kplus1 × tda]** – double *Output*  
*On exit:* the coefficients of  $T_j(\bar{x})$  in the approximating polynomial of degree  $i$ .  $\mathbf{a}[(i) \times \mathbf{tda} + j]$  contains the coefficient  $a_{i+1,j+1}$ , for  $i = 0, 1, \dots, k$  and  $j = 0, 1, \dots, i$ .
  
- 8: **s[kplus1]** – double *Output*  
*On exit:*  $\mathbf{s}[i]$  contains the root mean square residual  $s_i$ , for  $i = 0, 1, \dots, k$ , as described in Section 3. For the interpretation of the values of the  $s_i$  and their use in selecting an appropriate degree, see the e02 Chapter Introduction.

9: **fail** – NagError \*

*Input/Output*

The NAG error argument (see Section 2.7 in How to Use the NAG Library and its Documentation).

## 6 Error Indicators and Warnings

### NE\_2\_INT\_ARG\_GT

On entry, **kplus1** =  $\langle value \rangle$  while the number of distinct  $x$  values,  $mdist = \langle value \rangle$ . These arguments must satisfy **kplus1**  $\leq$   $mdist$ .

### NE\_2\_INT\_ARG\_LT

On entry, **tda** =  $\langle value \rangle$  while **kplus1** =  $\langle value \rangle$ . The arguments must satisfy **tda**  $\geq$  **kplus1**.

### NE\_ALLOC\_FAIL

Dynamic memory allocation failed.

### NE\_INT\_ARG\_LT

On entry, **kplus1** must not be less than 1: **kplus1** =  $\langle value \rangle$ .

### NE\_NO\_NORMALISATION

On entry, all the  $\mathbf{x}[r]$  in the sequence  $\mathbf{x}[r]$ ,  $r = 0, 1, \dots, \mathbf{m} - 1$  are the same.

### NE\_NOT\_NON\_DECREASING

On entry, the sequence  $\mathbf{x}[r]$ ,  $r = 0, 1, \dots, \mathbf{m} - 1$  is not in nondecreasing order.

### NE\_WEIGHTS\_NOT\_POSITIVE

On entry, the weights are not strictly positive:  $\mathbf{w}[\langle value \rangle] = \langle value \rangle$ .

## 7 Accuracy

No error analysis for the method has been published. Practical experience with the method, however, is generally extremely satisfactory.

## 8 Parallelism and Performance

nag\_1d\_cheb\_fit (e02adc) is not threaded in any implementation.

## 9 Further Comments

The time taken by nag\_1d\_cheb\_fit (e02adc) is approximately proportional to  $m(k+1)(k+11)$ .

The approximating polynomials may exhibit undesirable oscillations (particularly near the ends of the range) if the maximum degree  $k$  exceeds a critical value which depends on the number of data points  $m$  and their relative positions. As a rough guide, for equally spaced data, this critical value is about  $2 \times \sqrt{m}$ . For further details see page 60 of Hayes (1970).

## 10 Example

Determine weighted least squares polynomial approximations of degrees 0, 1, 2 and 3 to a set of 11 prescribed data points. For the approximation of degree 3, tabulate the data and the corresponding values of the approximating polynomial, together with the residual errors, and also the values of the approximating polynomial at points half-way between each pair of adjacent data points.

The example program supplied is written in a general form that will enable polynomial approximations of degrees  $0, 1, \dots, k$  to be obtained to  $m$  data points, with arbitrary positive weights, and the approximation of degree  $k$  to be tabulated. `nag_ld_cheb_eval` (e02aec) is used to evaluate the approximating polynomial. The program is self-starting in that any number of datasets can be supplied.

### 10.1 Program Text

```

/* nag_ld_cheb_fit (e02adc) Example Program.
 *
 * NAGPRODCODE Version.
 *
 * Copyright 2016 Numerical Algorithms Group.
 *
 * Mark 26, 2016.
 */

#include <nag.h>
#include <stdio.h>
#include <nag_stdlib.h>
#include <nage02.h>

int main(void)
{
#define A(I, J) a[(I) *tda + J]

    Integer exit_status = 0, i, iwght, j, k, m, r, tda;
    NagError fail;
    double *a = 0, *ak = 0, dl, fit, *s = 0, *w = 0, *x = 0, x1, xarg, xcapr,
           xm, *y = 0;

    INIT_FAIL(fail);

    printf("nag_ld_cheb_fit (e02adc) Example Program Results \n");

    /* Skip heading in data file */
#ifdef _WIN32
    scanf_s("%*[\n]");
#else
    scanf("%*[\n]");
#endif

#ifdef _WIN32
    while ((scanf_s("%" NAG_IFMT "", &m)) != EOF)
#else
    while ((scanf("%" NAG_IFMT "", &m)) != EOF)
#endif
    {
        if (m >= 2) {
            if (!(x = NAG_ALLOC(m, double)) ||
                !(y = NAG_ALLOC(m, double)) || !(w = NAG_ALLOC(m, double)))
            {
                printf("Allocation failure\n");
                exit_status = -1;
                goto END;
            }
        }
        else {
            printf("Invalid m.\n");
            exit_status = 1;
            return exit_status;
        }
    }

#ifdef _WIN32
    scanf_s("%" NAG_IFMT "", &k);
#else
    scanf("%" NAG_IFMT "", &k);
#endif
    if (k >= 1) {

```

```

    if (!(a = NAG_ALLOC((k + 1) * (k + 1), double)) ||
        !(s = NAG_ALLOC(k + 1, double)) || !(ak = NAG_ALLOC(k + 1, double)))
    {
        printf("Allocation failure\n");
        exit_status = -1;
        goto END;
    }
    tda = k + 1;
}
else {
    printf("Invalid k.\n");
    exit_status = 1;
    return exit_status;
}

#ifdef _WIN32
    scanf_s("%" NAG_IFMT "", &iwght);
#else
    scanf("%" NAG_IFMT "", &iwght);
#endif
    for (r = 0; r < m; ++r) {
        if (iwght != 1) {
#ifdef _WIN32
            scanf_s("%lf", &x[r]);
#else
            scanf("%lf", &x[r]);
#endif
#ifdef _WIN32
            scanf_s("%lf", &y[r]);
#else
            scanf("%lf", &y[r]);
#endif
#ifdef _WIN32
            scanf_s("%lf", &w[r]);
#else
            scanf("%lf", &w[r]);
#endif
        }
        else {
#ifdef _WIN32
            scanf_s("%lf", &x[r]);
#else
            scanf("%lf", &x[r]);
#endif
#ifdef _WIN32
            scanf_s("%lf", &y[r]);
#else
            scanf("%lf", &y[r]);
#endif
        }
        w[r] = 1.0;
    }
}
/* nag_ld_cheb_fit (e02adc).
 * Computes the coefficients of a Chebyshev series
 * polynomial for arbitrary data
 */
nag_ld_cheb_fit(m, k + 1, tda, x, y, w, a, s, &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_ld_cheb_fit (e02adc).\n%s\n", fail.message);
    exit_status = 1;
    goto END;
}

for (i = 0; i <= k; ++i) {
    printf("\n");
    printf(" %s%4" NAG_IFMT "%s%12.2e\n", "Degree", i,
        " R.M.S. residual =", s[i]);
    printf("\n  J Chebyshev coeff A(J) \n");
    for (j = 0; j < i + 1; ++j)
        printf(" %3" NAG_IFMT "%15.4f\n", j + 1, A(i, j));
}

```

```

for (j = 0; j < k + 1; ++j)
    ak[j] = A(k, j);
x1 = x[0];
xm = x[m - 1];
printf("\n %s%4" NAG_IFMT "\n\n",
    "Polynomial approximation and residuals for degree", k);
printf("  R  Abscissa  Weight  Ordinate  Polynomial  Residual \n");
for (r = 1; r <= m; ++r) {
    xcapr = (x[r - 1] - x1 - (xm - x[r - 1])) / (xm - x1);
    /* nag_ld_cheb_eval (e02aec).
     * Evaluates the coefficients of a Chebyshev series
     * polynomial
     */
    nag_ld_cheb_eval(k + 1, ak, xcapr, &fit, &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_ld_cheb_eval (e02aec).\n%s\n", fail.message);
        exit_status = 1;
        goto END;
    }

    dl = fit - y[r - 1];
    printf(" %3" NAG_IFMT "%11.4f%11.4f%11.4f%11.4f%11.2e\n", r, x[r - 1],
        w[r - 1], y[r - 1], fit, dl);
    if (r < m) {
        xarg = (x[r - 1] + x[r]) * 0.5;
        xcapr = (xarg - x1 - (xm - xarg)) / (xm - x1);
        /* nag_ld_cheb_eval (e02aec), see above. */
        nag_ld_cheb_eval(k + 1, ak, xcapr, &fit, &fail);
        if (fail.code != NE_NOERROR) {
            printf("Error from nag_ld_cheb_eval (e02aec).\n%s\n", fail.message);
            exit_status = 1;
            goto END;
        }
        printf("      %11.4f                %11.4f\n", xarg, fit);
    }
}
END:
NAG_FREE(a);
NAG_FREE(x);
NAG_FREE(y);
NAG_FREE(w);
NAG_FREE(s);
NAG_FREE(ak);
}
return exit_status;
}

```

## 10.2 Program Data

nag\_ld\_cheb\_fit (e02adc) Example Program Data

```

11
3  2
1.00  10.40  1.00
2.10  7.90  1.00
3.10  4.70  1.00
3.90  2.50  1.00
4.90  1.20  1.00
5.80  2.20  0.80
6.50  5.10  0.80
7.10  9.20  0.70
7.80  16.10 0.50
8.40  24.50 0.30
9.00  35.30 0.20

```

**10.3 Program Results**

nag\_ld\_cheb\_fit (e02adc) Example Program Results

Degree 0 R.M.S. residual = 4.07e+00

J	Chebyshev coeff A(J)
1	12.1740

Degree 1 R.M.S. residual = 4.28e+00

J	Chebyshev coeff A(J)
1	12.2954
2	0.2740

Degree 2 R.M.S. residual = 1.69e+00

J	Chebyshev coeff A(J)
1	20.7345
2	6.2016
3	8.1876

Degree 3 R.M.S. residual = 6.82e-02

J	Chebyshev coeff A(J)
1	24.1429
2	9.4065
3	10.8400
4	3.0589

Polynomial approximation and residuals for degree 3

R	Abcissa	Weight	Ordinate	Polynomial	Residual
1	1.0000	1.0000	10.4000	10.4461	4.61e-02
	1.5500			9.3106	
2	2.1000	1.0000	7.9000	7.7977	-1.02e-01
	2.6000			6.2555	
3	3.1000	1.0000	4.7000	4.7025	2.52e-03
	3.5000			3.5488	
4	3.9000	1.0000	2.5000	2.5533	5.33e-02
	4.4000			1.6435	
5	4.9000	1.0000	1.2000	1.2390	3.90e-02
	5.3500			1.4257	
6	5.8000	0.8000	2.2000	2.2425	4.25e-02
	6.1500			3.3803	
7	6.5000	0.8000	5.1000	5.0116	-8.84e-02
	6.8000			6.8400	
8	7.1000	0.7000	9.2000	9.0982	-1.02e-01
	7.4500			12.3171	
9	7.8000	0.5000	16.1000	16.2123	1.12e-01
	8.1000			20.1266	
10	8.4000	0.3000	24.5000	24.6048	1.05e-01
	8.7000			29.6779	
11	9.0000	0.2000	35.3000	35.3769	7.69e-02

---