



- 1 a) Use divided differences and Newton's interpolation formula to find the interpolating polynomial of lowest possible degree for the points in the table:

x_i	0	2	3
y_i	1	2	4

- b) Add the point $(1, 0)$ to the table in a). What is the new interpolating polynomial?
- 2 In this task, you are to approximate $\sin(x)$ on the interval $[0, \pi]$ using polynomial interpolation.
- a) Choose 4 equidistant nodes on the interval. Find the polynomial and an upper bound for the error $|\sin(x) - p_3(x)|$.
- b) Repeat a), but use Chebyshev nodes instead (see hint). (You do not have to calculate the polynomial, but show the interpolation points.)
- c) In the two cases (equidistant and Chebyshev nodes), find an expression for an upper error bound, expressed by n (the polynomial degree). Plot the error bound as a function of n and compare the two cases.

Hint: The Chebyshev nodes are defined on the interval $[-1, 1]$. To move them over to another interval $[a, b]$, the change of variables

$$x = \frac{a+b}{2} + \frac{b-a}{2}t, \quad t \in [-1, 1]$$

is used. The error formula is adjusted in the same way.

- 3 Write two PYTHON functions, one that calculates the table of divided differences based on a given dataset, and one that calculates the value of the interpolating polynomial in given points, based on this table.

Iterated multiplication may be useful in this task. Let

$$p(x) = f_0 + f_1(x - x_0) + f_2(x - x_0)(x - x_1) + \cdots + f_n(x - x_0)(x - x_1) \cdots (x - x_{n-1}),$$

and define the polynomials b_n, b_{n-1}, \dots, b_0 by

$$\begin{aligned} b_n(x) &= f_n \\ b_{n-1}(x) &= f_{n-1} + (x - x_{n-1})b_n(x) \\ &\vdots \\ b_0(x) &= f_0 + (x - x_0)b_1(x) \end{aligned}$$

Then $b_0(x) = p(x)$. This equality is easily seen by substituting for $b_1(x), b_2(x), \dots, b_n(x)$ in the expression for b_0 and expand the expression. Test the functions on the dataset in Task 1.

- 4 Let the distance h between nodes be such that $x_i = a + ih$, $i = 0, 1, 2, \dots$. Let $f_i = f(x_i)$, $i = 0, 1, 2, \dots$

On such a sequence $\{f_i\}_{i=0}^\infty$, we can define a *forward difference* recursively by

$$\Delta^0 f_0 = f_0, \quad \Delta f_0 = f_1 - f_0, \quad \Delta^k f_0 = \Delta(\Delta^{k-1} f_0) = \Delta^{k-1} f_1 - \Delta^{k-1} f_0, \quad k = 1, 2, \dots$$

We get

$$\Delta^2 f_0 = f_2 - 2f_1 + f_0, \quad \Delta^3 f_0 = f_3 - 3f_2 + 3f_1 - f_0, \quad \text{and so on.}$$

Let $x = x_0 + sh$, where $s \in \mathbb{R}$. The task is about showing that the polynomial interpolating f in the nodes x_i , $i = 0, 1, \dots, n$ can be written

$$p_n(x) = p_n(x_0 + sh) = f_0 + \sum_{k=1}^n \binom{s}{k} \Delta^k f_0 \tag{1}$$

where

$$\binom{s}{k} = \frac{s(s-1) \cdots (s-k+1)}{k!}.$$

- a) Show by induction:

$$f[x_0, x_1, \dots, x_k] = \frac{1}{k!h^k} \Delta^k f_0.$$

- b) Show that

$$\prod_{i=0}^{k-1} (x - x_i) = k!h^k \binom{s}{k}, \quad k \geq 1.$$

- c) Use the results from a) and b) to prove Newton's forward difference formula (1).
d) Apply the formula to the dataset in Task 1b).

Comment: Equivalently, it is possible to show *Newton's backward difference formula*. Backward differences on the sequence $\{f_n\}_{n=0}^{\infty}$ are defined by

$$\nabla^0 f_n = f_n, \quad \nabla f_n = f_n - f_{n-1}, \quad \nabla^k f_n = \nabla^{k-1} f_n - \nabla^{k-1} f_{n-1}, \quad k = 1, 2, \dots$$

Newton's backward difference formula is given by

$$p_n(x) = p_n(x_n + sh) = f_n + \sum_{k=1}^n (-1)^k \binom{-s}{k} \nabla^k f_n.$$

5 Exercise 7.3, chapter 7 from S&M book

6 Exercise 7.6, chapter 7 from S&M book

Note: S&M,

Süli, Endre, and David F. Mayers. *An introduction to numerical analysis*. Cambridge university press, 2003.