

Lecture 4

Numerical Interpolation / TDMA

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0. Interpolation

Problem statement

For given data

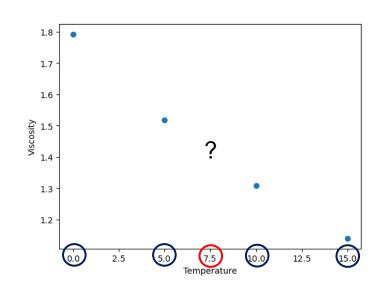
$$(x_i, y_i)$$
 with $i = 1, ..., n$

determine function $f: \mathbb{R} \to \mathbb{R}$ such that

$$f(x_i) = y_i \text{ with } i = 1, ..., n$$

- Given a new x^* , we can interpolate its function value $\hat{y}(x^*)$. $\hat{y}(x)$ is **interpolating function.**
- Example

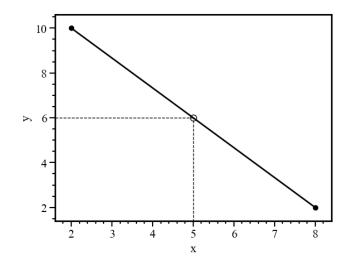
Temperature	0°	5°	10°	15°
Viscosity	1.792	1.519	1.308	1.140



Linear interpolation

- The estimated point is assumed to lie on the line joining the nearest points to the left and right.
- Linear interpolation at x is

$$p(x) = \left(\frac{x - x_1}{x_0 - x_1}\right) y_0 + \left(\frac{x - x_0}{x_1 - x_0}\right) y_1$$
$$= y_0 + \left(\frac{y_1 - y_0}{x_1 - x_0}\right) (x - x_0)$$



$$\begin{array}{c|ccc} x & 2 & 8 \\ \hline y & 10 & 2 \end{array} \longrightarrow \hat{y}(5) = 10 + \left(\frac{2-10}{8-2}\right)(5-2) = 6$$

Lagrange interpolation

- Lagrange polynomial interpolation finds a single polynomial, P(x).
- As an interpolation function, it should have the property $P(x_i) = f(x_i)$ for every point in the dataset.
- It is useful to write them as a linear combination of Lagrange basis polynomials, $l_i(x)$

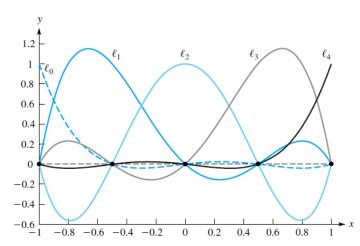
$$\ell_i(x) = \prod_{\substack{j \neq i \\ j=0}}^n \left(\frac{x - x_j}{x_i - x_j} \right) \qquad (0 \le i \le n)$$

$$\ell_i(x) = \left(\frac{x - x_0}{x_i - x_0}\right) \left(\frac{x - x_1}{x_i - x_1}\right) \cdots \left(\frac{x - x_{i-1}}{x_i - x_{i-1}}\right) \left(\frac{x - x_{i+1}}{x_i - x_{i+1}}\right) \cdots \left(\frac{x - x_n}{x_i - x_n}\right)$$

And

$$p_n(x) = \sum_{i=0}^n \ell_i(x) f(x_i)$$

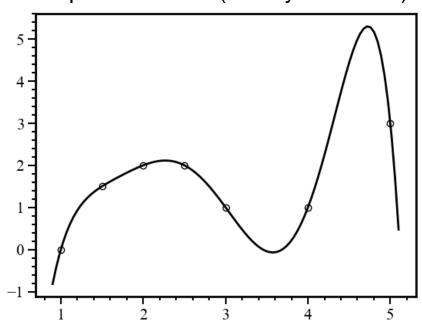
$$p_n(x_j) = \sum_{i=0}^n \ell_i(x_j) f(x_i) = \ell_j(x_j) f(x_j) = f(x_j)$$



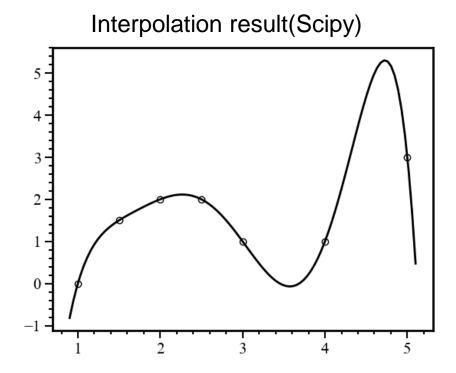
Lagrange interpolation

```
import numpy as np
        import matplotlib.pyplot as plt
In [2]:
        def Lagrange(x,y,xval):
            yval = 0
            deg = Ien(x) - 1
            for i in range(deg+1):
                LagBase = 1.
                for k in range(deg+1):
                    if(k != i):
                        LagBase \star= (xval-x[k])/(x[i]-x[k])
                vval += v[i]*LagBase
            return yval
In [3]: x = np.array([1, 1.5, 2, 2.5, 3, 4, 5])
        y = np.array([0, 1.5, 2, 2, 1, 1, 3])
In [4]:
        xa = np.linspace(0.9, 5.1, 100)
        ya = Lagrange(x, y, xa)
```

Interpolation result(line-by-line code)



Lagrange interpolation



Bivariate functions

The methods we have discussed for interpolating functions of one variable by polynomials extend to *some* cases of functions of two or more variables. An important case occurs when a function $(x, y) \mapsto f(x, y)$ is to be approximated on a rectangle. This leads to what is known as **tensor-product interpolation**. Suppose the rectangle is the Cartesian product of two intervals: $[a, b] \times [\alpha, \beta]$. That is, the variables x and y run over the intervals [a, b], and $[\alpha, \beta]$, respectively. Select n nodes x_i in [a, b], and define the *Lagrangian polynomials*

$$\ell_i(x) = \prod_{\substack{j \neq i \\ j=1}}^n \frac{x - x_j}{x_i - x_j} \qquad (1 \le i \le n)$$

Similarly, we select m nodes y_i in $[\alpha, \beta]$ and define

$$\overline{\ell}_i(y) = \prod_{\substack{j \neq i \\ j=1}}^m \frac{y - y_j}{y_i - y_j} \qquad (1 \le i \le m)$$

Then the function

$$P(x, y) = \sum_{i=1}^{n} \sum_{j=1}^{m} f(x_i, y_j) \ell_i(x) \overline{\ell_j}(y)$$



Linear algebra

An $n \times n$ system of linear equations can be written in matrix form

$$Ax = b$$

where the coefficient matrix A has the form

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \cdots & a_{nn} \end{bmatrix}$$

Diagonal 만 저장해서 저장공간 낭비를 줄임

$$\begin{pmatrix} d_0 & a_0 & 0 & \dots & \dots & 0 \\ b_1 & d_1 & a_1 & 0 & \dots & \dots & 0 \\ 0 & b_2 & d_2 & a_2 & 0 & \dots & 0 \\ 0 & 0 & * & * & * & 0 & 0 \\ 0 & 0 & 0 & * & * & * & 0 \\ 0 & \dots & \dots & 0 & b_{N-1} & d_{N-1} & a_{N-1} \\ 0 & \dots & \dots & 0 & b_N & d_N \end{pmatrix}$$



Gauss elimination (forward sweep)

$$d'_{i} = d_{i} - \frac{b_{i}}{d'_{i-1}} a_{i-1}$$

$$b'_{i} = b_{i} - \frac{b_{i}}{d'_{i-1}} d'_{i-1} = 0$$

$$c'_{i} = c_{i} - \frac{b_{i}}{d'_{i-1}} c'_{i-1}$$

$$i = 1, 2, \dots, N$$

$$\begin{pmatrix} d_0 & a_0 & 0 & \dots & \dots & 0 \\ 0 & d'_1 & a_1 & 0 & \dots & \dots & 0 \\ 0 & 0 & d'_2 & a_2 & 0 & \dots & 0 \\ 0 & 0 & 0 & * & * & 0 & 0 \\ 0 & 0 & 0 & 0 & * & * & 0 \\ 0 & \dots & \dots & 0 & 0 & d'_{N-1} & a_{N-1} \\ 0 & \dots & \dots & 0 & 0 & 0 & d'_{N} \end{pmatrix} \begin{pmatrix} u_0 \\ u_1 \\ u_2 \\ \vdots \\ u_{N-1} \\ u_N \end{pmatrix} = \begin{pmatrix} u_0 \\ u_1 \\ u_2 \\ \vdots \\ u_{N-1} \\ u_N \end{pmatrix}$$

2. TDMA

Back substitution (forward sweep)

Find unknowns starting with u_N

$$u_N = c'_N/d'_N$$
.

then going backward $i = N - 1, N - 2, \dots, 0$

$$u_i = (c_i' - a_i u_{i+1})/d_i'$$

Forward + backward sweeps require $\sim N$ arithmetic operations Standard Gauss elimination, which does not take into account special structure of matrix, requires $\sim N^3$ arithmetic operations

2. TDMA

Example

$$\begin{bmatrix} 2 & 3 & 0 & 0 \\ 6 & 3 & 9 & 0 \\ 0 & 2 & 5 & 2 \\ 0 & 0 & 4 & 3 \end{bmatrix} \begin{bmatrix} x_0 \\ x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 21 \\ 69 \\ 34 \\ 22 \end{bmatrix}$$



$$\begin{bmatrix} 2 & 3 & 0 & 0 \\ 6 & 3 & 9 & 0 \\ 0 & 2 & 5 & 2 \\ 0 & 0 & 4 & 3 \end{bmatrix} \begin{bmatrix} 3 \\ 5 \\ 4 \\ 2 \end{bmatrix} = \begin{bmatrix} 21 \\ 69 \\ 34 \\ 22 \end{bmatrix}$$

```
In [11]: import numpy as np
         def TDMA(a,b,c,d):
             n = len(c)
             dp = np.zeros(n)
             cp = np.zeros(n)
             x = np.zeros(n)
             # forward sweep
             dp[0] = d[0]
             cp[0] = c[0]
             for i in range(1,n):
                 dp[i] = d[i] - b[i]*a[i-1]/dp[i-1]
                 cp[i] = c[i] - b[i] * cp[i-1] / dp[i-1]
             # backward substitution
             x[n-1] = cp[n-1]/dp[n-1]
             for i in range(n-2,-1,-1):
                 x[i] = (cp[i] - a[i]*x[i+1])/dp[i]
             return x
```

```
In [24]: upp = np.array([3,9,2,0])
    dig = np.array([2,3,5,3])
    low = np.array([0,6,2,4])

    rhs = np.array([21,69,34,22])

X = TDMA(upp,low,rhs,dig)
    print(X)
```

[3. 5. 4. 2.]

Cubic Spline

Let $g_i(x)$ be the cubic in the interval $x_i \le x \le x_{i+1}$ and let g(x) denote the collection of all the cubics for the entire range of x. Since g is piecewise cubic its second derivative, g'', is piecewise linear. For the interval $x_i \le x \le x_{i+1}$, we can write the equation for the corresponding straight line as

$$g_i''(x) = g''(x_i) \frac{x - x_{i+1}}{x_i - x_{i+1}} + g''(x_{i+1}) \frac{x - x_i}{x_{i+1} - x_i}.$$
 (1.3)

Integrating (1.3) twice we obtain

$$g_i(x) = \frac{g''(x_i)}{x_i - x_{i+1}} \frac{(x - x_{i+1})^3}{6} + \frac{g''(x_{i+1})}{x_{i+1} - x_i} \frac{(x - x_i)^3}{6} + C_1 x + C_2.$$

The undetermined constants C_1 and C_2

$$g_i(x_i) = f(x_i) \equiv y_i$$
 $g_i(x_{i+1}) = f(x_{i+1}) \equiv y_{i+1}$

Cubic Spline

$$g_{i}(x) = \frac{g''(x_{i})}{6} \left[\frac{(x_{i+1} - x)^{3}}{\Delta_{i}} - \Delta_{i}(x_{i+1} - x) \right]$$

$$+ \frac{g''(x_{i+1})}{6} \left[\frac{(x - x_{i})^{3}}{\Delta_{i}} - \Delta_{i}(x - x_{i}) \right]$$

$$+ f(x_{i}) \frac{x_{i+1} - x}{\Delta_{i}} + f(x_{i+1}) \frac{x - x_{i}}{\Delta_{i}}, \qquad \Delta_{i} = x_{i+1} - x_{i}$$

 $g''(x_i)$ and $g''(x_{i+1})$ unknowns.

the continuity of the first derivatives: $g'_i(x_i) = g'_{i-1}(x_i)$

$$\frac{\Delta_{i-1}}{6}g''(x_{i-1}) + \frac{\Delta_{i-1} + \Delta_i}{3}g''(x_i) + \frac{\Delta_i}{6}g''(x_{i+1})
= \frac{f(x_{i+1}) - f(x_i)}{\Delta_i} - \frac{f(x_i) - f(x_{i-1})}{\Delta_{i-1}} \qquad i = 1, 2, 3, \dots, N-1.$$

N-1 equations for the N+1 unknowns

Cubic Spline

N-1 equations for the N+1 unknowns \rightarrow required 2 more constraints

$$\begin{bmatrix} \frac{\Delta_0 + \Delta_1}{3} & \frac{\Delta_1}{6} & 0 & \cdots & 0 & 0 & 0 \\ \frac{\Delta_1}{6} & \frac{\Delta_1 + \Delta_2}{3} & \frac{\Delta_2}{6} & \cdots & 0 & 0 & 0 \\ \vdots & & \ddots & & & \vdots \\ 0 & 0 & 0 & \cdots & \frac{\Delta_{n-3}}{6} & \frac{\Delta_{n-3} + \Delta_{n-2}}{3} & \frac{\Delta_{n-2}}{6} \\ 0 & 0 & 0 & \cdots & 0 & \frac{\Delta_{n-2}}{6} & \frac{\Delta_{n-2} + \Delta_{n-1}}{3} \end{bmatrix} \begin{bmatrix} g''(x_1) \\ g''(x_2) \\ \vdots \\ g''(x_{n-2}) \\ g''(x_{n-1}) \end{bmatrix} = \begin{bmatrix} \frac{f(x_2) - f(x_1)}{\Delta_1} - \frac{f(x_1) - f(x_0)}{\Delta_0} - \frac{\Delta_0}{6} g''(x_0) \\ \frac{f(x_2) - f(x_1)}{\Delta_1} - \frac{f(x_1) - f(x_0)}{\Delta_0} \\ \vdots \\ \frac{f(x_n) - f(x_{n-2})}{\Delta_{n-2}} - \frac{f(x_{n-2}) - f(x_{n-3})}{\Delta_{n-3}} \\ \frac{f(x_n) - f(x_{n-1})}{\Delta_{n-1}} - \frac{f(x_{n-1}) - f(x_{n-2})}{\Delta_{n-2}} - \frac{\Delta_{n-1}}{6} g''(x_n) \end{bmatrix}$$

Cubic Spline

Free run-out (natural spline):

$$g''(x_0) = g''(x_N) = 0$$
. 미분이 변화가 없다

Parabolic run-out:

$$g''(x_0) = g''(x_1)$$

$$g''(x_{N-1}) = g''(x_N).$$

$$g''(x_0) = \alpha g''(x_1)$$

$$g''(x_{N-1}) = \beta g''(x_N),$$

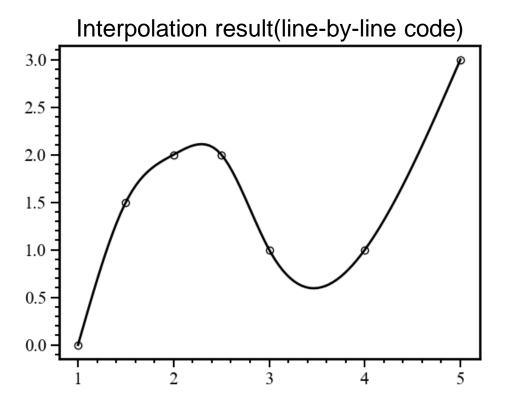
 α and β are constants chosen by the user.

Periodic:

$$g''(x_0) = g''(x_{N-1})$$

 $g''(x_1) = g''(x_N).$

Cubic Spline

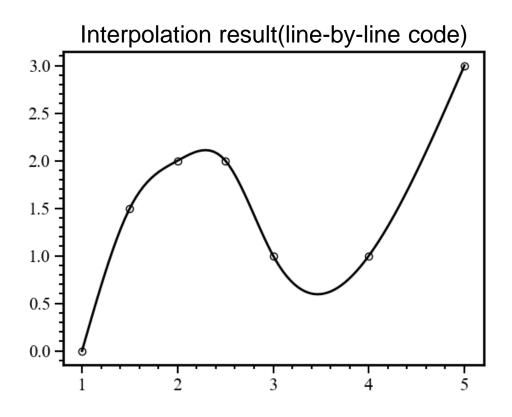


```
In [13]: import numpy as np
        def cubic_spline(x,y,xval):
            # number of intervals
            N = Ien(x) - 1
            # number of points
            ND = N + 1
            # number of intervals minus 1
            # initialize arrays
            h = np,zeros(N) # interval widths
            gdp = np,zeros(Np) # second derivatives of spline
            upp = np.zeros(Nm) # upper diagonal of matrix
            low = np.zeros(Nm) # lower diagonal of matrix
            dig = np,zeros(Nm) # diagonal of matrix
            rhs = np,zeros(Nm) # right-hand side of matrix equation
            gdp[0], gdp[N] = 0.0, 0.0 # free-rum (boundary conditions)
            # calculate interval widths
            h[:] = x[1:] - x[:-1]
            # set up matrix equation to solve for second derivatives of spline
            upp[:-1] = h[1:-1]/6
            dig[ : ] = (h[ :-1]+h[ 1: ])/3
            low[1:] = h[1:-1]/6
            rhs[ : ] = (y[2: ]-y[1:-1])/h[1: ] - (y[1:-1]-y[:-2])/h[:-1]
            rhs[0] -= h[0]*gdp[0]/6
            rhs[-1] = h[-1]*gdp[-1]/6
            # solve matrix equation to obtain second derivatives of spline
            gdp[1:-1] = TDMA(upp, low, rhs, dig)
            # evaluate spline at specified x values
            Ncs = Ien(xval)
            yval= np,zeros(Ncs)
            for i in range(N)
                for j in range(Ncs):
                    if x[i+1]>=xval[j] and x[i]<xval[j]:</pre>
                       yval[j] = gdp[i]/6 * ((x[i+1]-xval[j])**3/h[i] - h[i]*(x[i+1]-xval[j])) #
                               + gdp[i+1]/6 + ((xval[j] - x[i])+3/h[i] - h[i]+(xval[j] - x[i])) #
                               + y[i]*(x[i+1]-xval[j])/h[i] + y[i+1]*(xval[j]-x[i])/h[i]
            # return spline values at specified x values
            return yval
```

n [14]: xa

xa = np.linspace(1,5,100)
ya = cubic_spline(x,y,xa)

Cubic Spline



```
In [28]: from scipy import interpolate as ip

cs = ip.CubicSpline(x,y,bc_type='natural')

xa = np.linspace(1,5,100)
ya = cs(xa)
```

Q&A Thanks for listening