

MM6 Lagrange Interpolation Method

- kl.8.15-9.00 review of MM5 and some examples
- kl.9.00 – 10.30 exercise (see notes)
- kl.10.40-11.30 MM6 lecture (I)

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Newton's Method (MM4)

- Newton-Raphson method

Suppose to solve $f(x) = 0$

The first-order Taylor expansion of f about a point x_0 is

$$f(x) \approx f(x_0) + (x - x_0)f'(x_0)$$

If the point x_0 is close to the required solution, then we expect that

$$f(x_0) + (x - x_0)f'(x_0) = 0$$

$$x_1 = x_0 - \frac{f(x_0)}{f'(x_0)}$$

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}, \quad n = 0, 1, 2, \dots$$

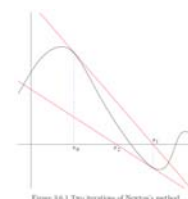


Figure 10.1 The iteration of Newton's method

$$g(x) = x - \frac{f(x)}{f'(x)}$$

$$g'(x) = 1 - \frac{f'(x)}{f'(x)} + \frac{f(x)f''(x)}{(f'(x))^2}$$

$$= 1 - 1 + 0 = 0$$

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Secant Method (MM5)

- Select two starting points (different signs)
- X_{n+1} is the point at which the secant line joining $(x_n, f(x_n))$ and $(x_{n-1}, f(x_{n-1}))$ cuts the x-axis

$$\frac{f(x_2) - f(x_1)}{x_2 - x_1} = \frac{f(x_1) - f(x_0)}{x_1 - x_0}$$

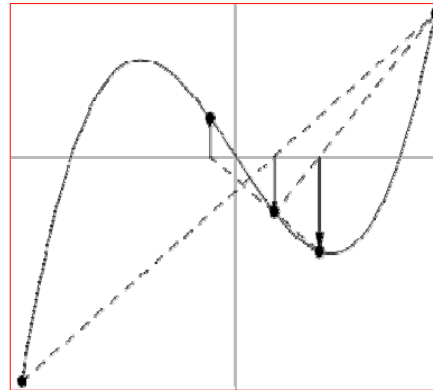
$$f(x_2) = 0$$

$$\Downarrow$$

$$x_2 = x_1 + \frac{x_1 - x_0}{f(x_1) - f(x_0)}$$

$$\Downarrow$$

$$x_{n+1} = x_n + \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})}, \quad n = 1, 2, \dots$$



$$e_{n+1} \approx c e_n^\alpha \quad \text{where} \quad \alpha = \frac{1 + \sqrt{5}}{2} \approx 1.6$$

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Matlab Codes of Secant Method

```
function [y,its]=secant(f,x0,x1,tol)
format long
f0=feval(f,x0);
f1=feval(f,x1);
maxit=100;
its =0;
while (abs(x0-x1)>tol) & (its <maxit)
    x2=x1-(x1-x0)*f1/(f1-f0);
    x0=x1;
    x1=x2;
    f0=feval(f,x0);
    f1=feval(f,x1);
    its=its+1;
end
y=x1;
```

$$x_{n+1} = x_n + \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})}, \quad n = 1, 2, \dots$$

```
function y=example10(x);
format long
y=exp(x)-2*x-1;
```

```
>> [y,its]=secant(@example10,1,2,10e-6)
y = 1.256431208621747
its = 7
```

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2 Equations in 2 Unknowns Using Newton's Method (MM5)

$$\begin{cases} f_1(x, y) = 0 \\ f_2(x, y) = 0 \end{cases}$$

↓

$$\begin{cases} f_1(x+h, y+k) = f_1(x, y) + f_{1x}(x, y)h + f_{1y}(x, y)k \\ f_2(x+h, y+k) = f_2(x, y) + f_{2x}(x, y)h + f_{2y}(x, y)k \end{cases}$$

set

$$f_1(x+h, y+k) = f_2(x+h, y+k) = 0$$

↓

$$\begin{cases} h = \frac{-f_1 f_{2y} + f_2 f_{1y}}{f_{1x} f_{2y} - f_{1y} f_{2x}} \\ k = \frac{f_1 f_{2x} - f_2 f_{1x}}{f_{1x} f_{2y} - f_{1y} f_{2x}} \end{cases} \Rightarrow \begin{cases} x_{n+1} = x_n + h \\ y_{n+1} = y_n + k \end{cases}, \quad n = 1, 2, \dots$$

Jacobian matrix:

$$\begin{bmatrix} \frac{\partial f_1(x, y)}{\partial x} & \frac{\partial f_1(x, y)}{\partial y} \\ \frac{\partial f_2(x, y)}{\partial x} & \frac{\partial f_2(x, y)}{\partial y} \end{bmatrix} = \begin{bmatrix} f_{1x} & f_{1y} \\ f_{2x} & f_{2y} \end{bmatrix}$$

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Matlab Codes for Newton's Method (2 unknown, 2 equations)

```
function [sol,its]=newton2(fcn,Jac,v0,tol)
```

```
x0=v0(1); y0=v0(2);
f=feval(fcn,v0);
f1=f(1);
f2=f(2);
J=feval(Jac,v0);
f1x=J(1,1); f1y=J(1,2); f2x=J(2,1); f2y=J(2,2);
h=(-f1*f2y + f2*f1y)/(f1x*f2y - f1y*f2x);
k=(f1*f2x - f2*f1x)/(f1x*f2y - f1y*f2x);
x1=x0+h; y1=y0+k;
v1=[x1; y1];
its=1;
```

```
while abs(x1-x0)>tol | abs(y1-y0)>tol
    v0=v1; x0=v0(1); y0=v0(2);
    f=feval(fcn,v0);
    f1=f(1); f2=f(2);
    J=feval(Jac,v0);
    f1x=J(1,1); f1y=J(1,2); f2x=J(2,1); f2y=J(2,2);
    h=(-f1*f2y + f2*f1y)/(f1x*f2y - f1y*f2x);
    k=(f1*f2x - f2*f1x)/(f1x*f2y - f1y*f2x);
    v1=[x0+h; y0+k];
    its=its+1;
end
sol=v1; its=its;
```

```
function f=eq2(v)
```

```
% here both f and v are vectors
```

```
x=v(1); y=v(2);
f(1)=4*x^2 + y^2-4;
f(2)=x^2*y^3-1;
```

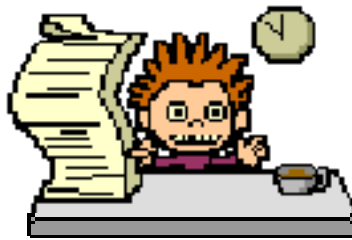
```
function J=Deq2(v)
```

```
% Jacobian matrix for eq2.m
```

```
x=v(1); y=v(2);
J(1,1)=8*x; J(1,2)=2*y;
J(2,1)=2*x*y^3;
J(2,2)=3*x^2*y^2;
```

```
>> [sol,its]=newton2(@eq2,@Deq2,[0.8,1.2],10e-8)
sol =
    0.821684259596287
    1.139889510230440
its = 2
>> [sol,its]=newton2(@eq2,@Deq2,[0.4,1.8],10e-8)
sol =
    0.404149564420627
    1.829386038547648
its = 2
```

Exercises (MM5)



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Question One:

Consider the same equation as we used in MM3 and MM4 Exercise One, i.e.,

$$3x^3 - 5x^2 - 4x + 4 = 0 \quad (1)$$

- Create your m-file to obtain the solution of the above equation located within the interval $[0, 1]$, using secant method with tolerance 10^{-6} ;
- How many iterations would be needed to obtain this solution? Compare the result with that of Newton's method in Exercise MM4.

Question Two:

Consider the following two equations

$$\begin{aligned} 4x^2 + y^2 &= 4 \\ x^2 y^3 &= 1, \end{aligned} \quad (2)$$

- Find the coordinates of the intersections in the second quadrant of curves described by above equations using 2 unknown parameters Newton's Method.

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MM6 Lagrange Interpolation Method

Reading material: section 4.1 and 4.2 in
Textbook

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Introduction to Interpolation

- **Motivation:**
Approximate evaluation of a function which is known only by its values at a set of data points
- **Interpolation:**
the process of fitting a function of a particular nature (typically a **polynomial**, or a **collection of polynomials**) through the data.
- **Classification:**
 - Polynomial interpolation, C
 - Cubic spline approximation,
 - Least square approximation

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Polynomial Interpolation

- Any continuous function can be approximated to any required accuracy by a polynomial (**Weierstrass Theorem**)
- Polynomials are the **only** functions, which we can, theoretically at least, evaluate **exactly**
- Efficient evaluation of a polynomial (**Horner's Rule**)

$$p(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$$

$$p(x) = \{[(a_n x + a_{n-1})x + a_{n-2}]x + \dots + a_1\}x + a_0$$

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Lagrange Interpolation

- **Problem formulation:**
Suppose we are given the values of a function f at $N+1$ distinct points (nodes), we wish to find a polynomial p with minimum degree such as

$$p(x_k) = f(x_k), k=0, 1, 2, \dots, N$$

$$p(x) = a_m x^m + a_{m-1} x^{m-1} + \dots + a_1 x + a_0$$

$$\begin{cases} a_m x_0^m + a_{m-1} x_0^{m-1} + \dots + a_1 x_0 + a_0 = f(x_0) \\ a_m x_1^m + a_{m-1} x_1^{m-1} + \dots + a_1 x_1 + a_0 = f(x_1) \\ \vdots \\ a_m x_N^m + a_{m-1} x_N^{m-1} + \dots + a_1 x_N + a_0 = f(x_N) \end{cases} \Leftrightarrow AX = F$$

Vandermonde determinant :

$$V = \begin{vmatrix} 1 & x_0 & \dots & x_0^N \\ 1 & x_1 & \dots & x_1^N \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_N & \dots & x_N^N \end{vmatrix}$$

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Lagrange Polynomial

If we can find a number of polynomials $l_j(x)$ ($j = 0, 1, \dots, N$) of degree at most N such that

$$l_j(x_k) = \delta_{jk} = \begin{cases} 1 & \text{if } j = k \\ 0 & \text{if } j \neq k \end{cases}$$

then the Lagrange polynomial is defined as

$$p(x) = \sum_{j=0}^N f(x_j) l_j(x)$$

The Lagrange basis polynomials $l_j(x)$ ($j = 0, 1, \dots, N$) are defined as

$$l_j(x) = \frac{(x - x_0) \cdots (x - x_{j-1})(x - x_{j+1}) \cdots (x - x_N)}{(x_j - x_0) \cdots (x_j - x_{j-1})(x_j - x_{j+1}) \cdots (x_j - x_N)}$$