

Numerical Analysis

Mathematics of Scientific Computing

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目录

内容提要

下一节内容

bisection method

- Intermediate-Value Theorem: If f is a continuous function on the interval $[a, b]$ and if $f(a)f(b) < 0$, then f must have a zero in (a, b) .
- The process of bisection method:
 - If $f(a)f(b) < 0$, then $c = (a + b)/2$
 - If $f(a)f(c) < 0$, then f has a zero in $[a, c]$, then $b \leftarrow c$
 - If $f(b)f(c) < 0$, then f has a zero in $[c, b]$, then $a \leftarrow c$
 - If $f(a)f(c) = 0$, then c is the zero of f
 - repeat this process
- The bisection method finds one zero but not all the zeros in the interval $[a, b]$.

Bisection Algorithm

- **Bisection pseudocode need some additional explanation:**

First, $c \leftarrow a + (b - a)/2$ rather than $c \leftarrow (a + b)/2$

second, $\text{sign}(f(c)) \neq \text{sign}(f(a))$ rather than $f(a)f(c) < 0$

- **three stopping criteria**

M give the maximum number of steps that present the computation going into an infinite loop.

When the value of $f(c)$ is small enough, the calculation can be stopped.

When $b - a$ is small enough, the calculation can be stopped.

The bisection algorithm can be written as follows

- input $a, b, M, \delta, \varepsilon$
 $u \leftarrow f(a)$
 $v \leftarrow f(b)$
 $e \leftarrow (b - a)$
 output a, b, u, v
 If $\text{sign}(u) = \text{sign}(v)$ then stop
 for $k = 1$ to M do
 $e \leftarrow e/2$
 $c \leftarrow a + e$
 $w \leftarrow f(c)$
 output k, c, w, e
 If $|e| < \delta$ or $|w| < \varepsilon$ then stop

- If $\text{sign}(w) \neq \text{sign}(u)$ then

$b \leftarrow c$

$v \leftarrow w$

else

$a \leftarrow c$

$u \leftarrow w$

end if

end do

EXAMPLE 1

- Use the bisection method to find the root of the equation $e^x = \sin x$ closest to 0.

```
1 function f=equation(x)
2     f=exp(1)^x-sin(x);
```

```
1 clc
2 a=-4;b=-3;n=16;p=10^(-5); q=10^(-5);
3 [c,w]=Bisection(a,b,n,p,q)
```

EXAMPLE 1

```
1 function [c,w]=Bisection(a,b,n,p,q)
2 u=equation(a);
3 v=equation(b);
4 if sign(u)==sign(v)
5     error('function has same sign at both end points')
6 end
7 for i=1:n
8     e=b-a;
9     c=a+0.5*e
10    w=equation(c)
11    if abs(w)<p || abs(e)<q
12        break
13    end
14    if sign(w)==sign(u)
15        a=c;u=w;
16    else
17        b=c; v=w;
```

Error Analysis

- Let us denote the successive intervals that arise in the process by $[a_0, b_0], [a_1, b_1]$, and so on.

$$a_0 \leq a_1 \leq a_2 \leq a_3 \leq \dots \leq b_0$$

$$b_0 \geq b_1 \geq b_2 \geq b_3 \geq \dots \geq a_0$$

$$b_{n+1} - a_{n+1} = 1/2(b_n - a_n) \quad (n \geq 0) \quad (1)$$

If we apply Equation(1) repeatedly, we find that

$$b_n - a_n = 2^{-n}(b_0 - a_0)$$

$$\lim_{n \rightarrow \infty} b_n - \lim_{n \rightarrow \infty} a_n = \lim_{n \rightarrow \infty} 2^{-n}(b_0 - a_0) = 0$$

$$r = \lim_{n \rightarrow \infty} a_n = \lim_{n \rightarrow \infty} b_n$$

- then, by taking a limit in the inequality $f(a_n)f(b_n) \leq 0$, we obtain $[f(r)]^2 \leq 0$, whence $f(r) = 0$
The best estimate of the root at this stage is not a_n or b_n but the midpoint of the interval: $c_n = (a_n + b_n)/2$
The error is the bounded as follows:

$$|r - c_n| \leq 1/2(b_n - a_n) = 2^{-(n+1)}(b_0 - a_0)$$

Theorem on bisection method

THEOREM 1 (Theorem on bisection method)

If $[a_0, b_0], [a_1, b_1], \dots, [a_n, b_n], \dots$ denote the intervals in the bisection method, then the limits $\lim_{n \rightarrow \infty} a_n$ and $\lim_{n \rightarrow \infty} b_n$ exist, are equal, and represent a zero of f . If $r = \lim_{n \rightarrow \infty} c_n$ and $c_n = 1/2(a_n + b_n)$, then

$$|r - c_n| \leq 2^{-(n+1)}(b_0 - a_0)$$

EXAMPLE 2

- Suppose that the bisection method is started with the interval $[50, 63]$. How many steps should be taken to compute a root with relative accuracy of one part in 10^{-12} ?
relative accuracy means that

$$|r - c_n|/|r| \leq 10^{-12}$$

We know that $r \geq 50$, so

$$|r - c_n|/50 \leq 10^{-12}$$

$$|r - c_n| \leq 2^{-(n+1)}(63 - 50)$$

$$2^{-(n+1)} * (13/15) \leq 10^{-12}$$

We conclude that $n \geq 37$

下一节内容

Newton's Method

- We have a function f whose zeros are to be determined numerically. Let r be a zero of f and let x be an approximation to r . If f' exists and is continuous, then by Taylor's theorem,

$$0 = f(r) = f(x + h) = f(x) + hf'(x) + O(h^2)$$

where $h = r - x$, ignore the $O(h^2)$, then $h = -f(x)/f'(x)$

$$r = h + x = x - f(x)/f'(x)$$

Newton's method begins with an estimate x_0 of r and then defines inductively

$$x_{n+1} = x_n - f(x_n)/f'(x_n) (n \geq 0)$$

Newton's Algorithm

- input $x_0, M, \delta, \varepsilon$
 $v \leftarrow f(x_0)$
 output $0, x_0, v$
 If $|v| < \varepsilon$ then stop
 for $k = 1$ to M do
 $x_1 \leftarrow x_0 - v/f'(x_0)$
 $v \leftarrow f(x_1)$
 output k, x_1, v
 If $|x_1 - x_0| < \delta$ or $|v| < \varepsilon$ then stop
 $x_0 \leftarrow x_1$
 end do

EXAMPLE 3

- Use Newton's method, to find the negative zero of the function

$$f(x) = e^x - 1.5 - \tan^{-1}x$$

```
1 function f=newf(x)
2 f=exp(1)^(x)-1.5-atan(x);
```

```
1 function df=newdf(x)
2 df=exp(1)^x-(1+x^2)^(-1);
```

```
1 clc
2 x0=-7;M=10;p=10^(-8);q=10^(-8);
3 [x1,v]=newton(x0,M,p,q)
```

EXAMPLE 3

```
1 function [x1,v]=newton(x0,M,p,q)
2 v=newf(x0)
3 if abs(v)<p
4     return
5 end
6 for k=1:M
7     x1=x0-v/newdf(x0)
8     v=newf(x1)
9     if abs(x1-x0)<p || abs(v)<q
10         return
11     end
12     x0=x1;
13 end
```

Graphical Interpretation

- Newton's method involves linearizing the function. Replace f by the first two in his Taylor series.

$$f(x) = f(c) + f'(c)(x - c) + 1/2f''(c)(x - c)^2 + \dots$$

then the linearization (at c) produces the linear function:

$$l(x) = f(c) + f'(c)(x - c)$$

l is a good approximation to f in the vicinity of c , and we have $l(c) = f(c)$ and $l'(c) = f'(c)$

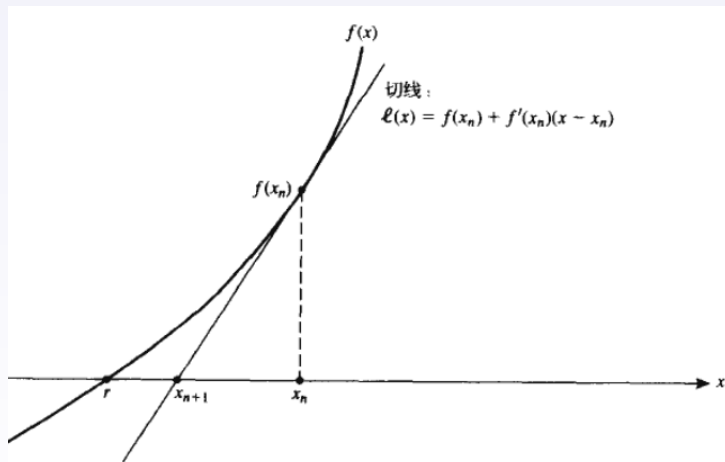


图 3-4 牛顿法的图形解释

Error Analysis

- Now we shall analyze the errors in Newton's method.

$$e_n = x_n - r$$

Let us assume that f'' is continuous and r is a simple zero of f , so that $f(r) = 0 \neq f'(r)$. From the definition of the Newton iteration, We have

$$e_{n+1} = x_{n+1} - r = x_n - \frac{f(x_n)}{f'(x_n)} - r = e_n - \frac{f(x_n)}{f'(x_n)} = \frac{e_n f'(x_n) - f(x_n)}{f'(x_n)}$$

$$0 = f(r) = f(x_n - e_n) = f(x_n) - e_n f'(x_n) + \frac{1}{2} e_n^2 f''(\xi_n)$$

$$e_n f'(x_n) - f(x_n) = \frac{1}{2} f''(\xi_n) e_n^2$$

$$e_{n+1} = \frac{1}{2} \frac{f''(\xi_n)}{f'(x_n)} e_n^2 \approx \frac{1}{2} \frac{f''(r)}{f'(r)} e_n^2 = C e_n^2$$

Theorem on Newton's Method

THEOREM 2 (Theorem on Newton's Method)

Let f' be continuous and let r be a simple zero of f . Then there is a neighborhood of r and a constant C such that if Newton's method is started in that neighborhood, the successive points become steadily closer to r and satisfy

$$|x_{n+1} - r| \leq C(x_n - r)^2$$

- In some situations Newton's iteration can be guaranteed to converge from an arbitrary starting point.

Theorem on Newton's Method for a Convex Function

THEOREM 3 (Theorem on Newton's Method for a Convex Function)

If f belongs to $C^2(R)$, is increasing, is convex, and has a zero, then the zero is unique, and the Newton iteration will converge to it from any starting point.

- proof: f is convex $\rightarrow f''(x) > 0$, f is increasing $\rightarrow f' > 0$ on the R

By Equation, $e_{n+1} = \frac{1}{2} \frac{f''(\xi_n)}{f'(x_n)} e_n^2$, $e_{n+1} > 0$

$e_{n+1} = x_{n+1} - r \rightarrow x_n > r$ for $n \geq 1$

f is increasing, $f(x_n) > f(r) = 0$. by Equation $e_{n+1} = e_n - \frac{f(x_n)}{f'(x_n)}$,

$e_{n+1} < e_n$

Thus, the sequences $[e_n]$ and $[x_n]$ are decreasing and bounded below $(0, r)$. Therefore, the limits $e^* = \lim_{n \rightarrow \infty} e_n$ and $x^* = \lim_{n \rightarrow \infty} x_n$

exist. $e^* = e^* - \frac{f(x^*)}{f'(x^*)}$, Whence $f(x^*) = 0$ and $x^* = r$.

Implicit Functions

- For the equation $G(x, y) = 0$, If x is prescribed, the equation $G(x, y) = 0$ can be solved for y using Newton's method. Form a suitable starting point y_0 , we define y_1, y_2, \dots by, this method can be used to construct a table of the function $y(x)$.

$$y_{k+1} = y_k - \frac{G(x, y_k)}{\frac{\partial G}{\partial y}(x, y_k)}$$

If the table contains an entry (x_n, y_n) , we can start the Newton iteration with (x_{n+1}, y_n) , the result will be the value y_{n+1} .

EXAMPLE 4

- Produce a table of x versus y , where y is defined implicitly as a function of x . Use $G(x, y) = 3x^7 + 2y^5 - x^3 + y^3 - 3$ and start at $x = 0$, proceeding in steps of 0.1 to $x = 10$.

```
1 function f=F(x,y)
2 f=3*x^7+2*y^5-x^3+y^3-3;
```

```
1 function fy=Fy(y)
2 fy=10*y^4+3*y^2;
```

```
1 x=0;y=1;h=0.1;M=100;N=4;
2 [i,x,y,G]=newton4(x,y,h,M,N)
```

EXAMPLE 4

```
1 function [i,x,y,G]=newton4(x,y,h,M,N)
2 for i=1:M
3     x = x+h
4     for j=1:N
5         f=F(x,y);
6         fy=Fy(y);
7         y = y-f/fy;
8     end
9     y=y
10    G=f
11 end
```

Systems of Nonlinear Equations

- Newton's method for systems of nonlinear equations follows **linearize** and **solve**, let us illustrate with a pair of equations involving two variables:

$$\begin{cases} f_1(x_1, x_2) = 0 \\ f_2(x_1, x_2) = 0 \end{cases}$$

linear terms in the Taylor expansion

$$\begin{cases} 0 = f_1(x_1 + h_1, x_2 + h_2) \approx f_1(x_1, x_2) + h_1 \frac{\partial f_1}{\partial x_1} + h_2 \frac{\partial f_1}{\partial x_2} \\ 0 = f_2(x_1 + h_1, x_2 + h_2) \approx f_2(x_1, x_2) + h_1 \frac{\partial f_2}{\partial x_1} + h_2 \frac{\partial f_2}{\partial x_2} \end{cases}$$

\mathbf{J} is the **Jacobian matrix** of f_1 and f_2 :

$$\mathbf{J} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix}$$

- the solution is

$$\begin{bmatrix} h_1 \\ h_2 \end{bmatrix} = -J^{-1} \begin{bmatrix} f_1(x_1, x_2) \\ f_2(x_1, x_2) \end{bmatrix}$$

Newton's method for two nonlinear equations in two variables is

$$\begin{bmatrix} x_1^{(k+1)} \\ x_2^{(k+1)} \end{bmatrix} = \begin{bmatrix} x_1^{(k)} \\ x_2^{(k)} \end{bmatrix} + \begin{bmatrix} h_1^{(k)} \\ h_2^{(k)} \end{bmatrix}$$

下一节内容

Secant Method

- **The drawback of Newton's method** is need the derivative of the function at zero.
- How to solve ?
 1. Steffensen's iteration
$$x_{n+1} = x_n - \frac{[f(x_n)]^2}{f(x_n + f(x_n)) - f(x_n)}$$
 2. Secant Method

Secant Method

- The Newton iteration is $x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$, replace $f'(x_n)$ by a **difference quotient**, such as

$$f'(x_n) \approx \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}$$

$$x_{n+1} = x_n - f(x_n) \left[\frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \right]$$

Secant Algorithm

- input $a, b, M, \delta, \varepsilon$ $fa \leftarrow f(a); fb \leftarrow f(b)$
 output $0, a, fa$
 output $1, b, fb$
 for $k = 2$ to M do
 If $|fa| > |fb|$ then
 $a \leftrightarrow b; fa \leftrightarrow fb$
 end if
 $s \leftarrow (b - a)/(fb - fa)$
 $b \leftarrow a$
 $fb \leftarrow fa$
 $a \leftarrow a - fa * s$
 $fa \leftarrow f(a)$
 output k, a, fa
 If $|fa| < \delta$ or $|b - a| < \varepsilon$ then stop
 end do

EXAMPLE 5

Use the secant method to find a zero of the function

$$f(x) = x^3 - \sinh x + 4x^2 + 6x + 9$$

```
1  function f=sca(x)
2  f=x^3-sinh(x)+4*x^2+6*x+9;
```

```
1  clc
2  a=8;b=7;M=10;p=10^(-8);q=10^(-8);
3  [a,fa]=scant(a,b,M,p,q)
```

EXAMPLE 5

```
1 function [a,fa]=scant(a,b,M,p,q)
2 fa=sca(a);fb=sca(b)
3 for k=1:M
4     if abs(fa)>abs(fb)
5         c=b;b=a;a=c
6         fd=fb;fb=fa; fa=fd
7     end
8     s=(b-a)/(fb-fa);
9     b=a;
10    fb=fa;
11    a=a-fa*s
12    fa=sca(a)
13    if abs(fa)<p || abs(b-a)<q
14        break
15    end
16 end
```

Error Analysis

- $|e_{n+1}| \approx A|e_n|^{(1+\sqrt{5})/2}$
- since $(1 + \sqrt{5})/2 \approx 1.62 < 2$, the rapidity of convergence of the secant method is not as good as Newton's method but is better than the bisection method.