

Nonlinear Equations

- ♣ Bisection Method
- ♣ Newton Method
- ♣ Secant Method
- ♣ Fixed Point and Functional Iteration
- ♠ A System of Nonlinear Equations
- ◇ Other Topics with Applications

1. $x^3 + 3x - 1 = 0$

2. $e^x - \sin(x) = 0, \quad e^x - 1.5 - \tan^{-1}(x) = 0$

3. $3 + \frac{1}{2} \sin(x) - x = 0, \quad 4 + \frac{1}{3} \sin(2x) - x = 0$

4. $x^3 - \sinh(x) + 4x^2 + 6x + 9 = 0$

MATLAB Codes for Plot Nonlinear Functions

```

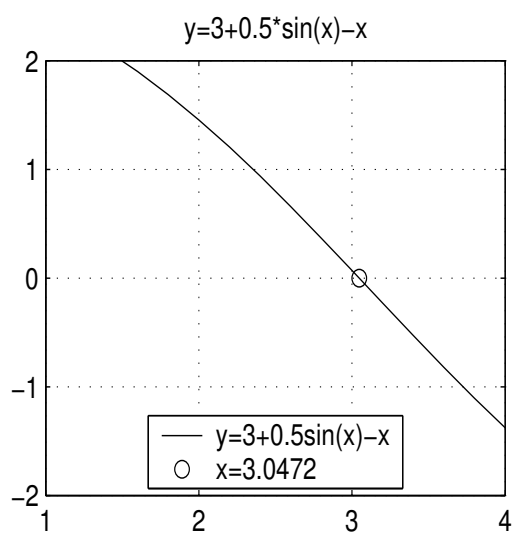
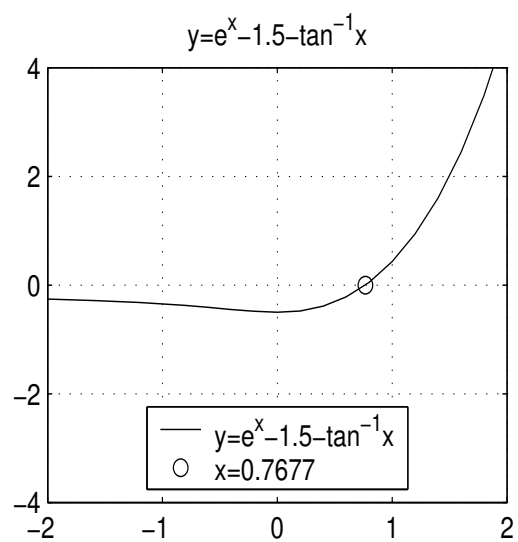
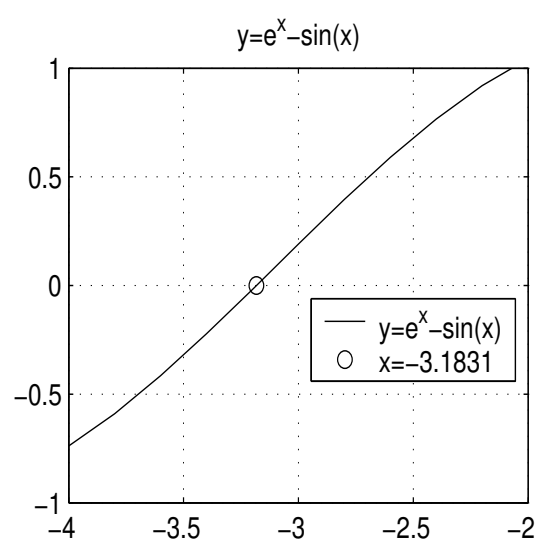
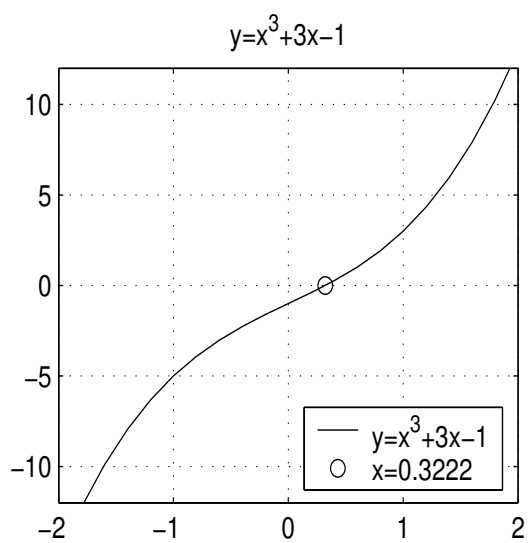
% Plot of four nonlinear functions
% Function funcA
subplot(2,2,1)
X=-2:0.2:2;
Y=X.^3+3*X-1;
V=[-2 2,-12 12];
plot(X,Y,'b-',[0.3222],[0],'ro'); axis(V); grid
% set(gca,'Xtick',[]), set(gca,'YTick',[0]); grid on
legend('y=x^3+3x-1','x=0.3222',0);
title('y=x^3+3x-1')

% Function funcB
subplot(2,2,2)
X=-4:0.2:-2;
Y=exp(X)-sin(X);
V=[-4 -2, -1 1];
plot(X,Y,'b-',[-3.1831],[0],'ro'); axis(V); grid
% set(gca,'Xtick',[]), set(gca,'YTick',[0]); grid on
legend('y=e^x-sin(x)','x=-3.1831',0);
title('y=e^x-sin(x)')

% Function funcC
subplot(2,2,3)
X=-2:0.2:2;
Y=exp(X)-1.5-atan(X);
V=[-2 2,-4 4];
plot(X,Y,'b-',[0.7677],[0],'ro'); axis(V); grid
% set(gca,'Xtick',[]), set(gca,'YTick',[0]); grid on
legend('y=e^x-1.5-tan^{-1}x','x=0.7677',0);
title('y=e^x-1.5-tan^{-1}x')

% Function funcD
subplot(2,2,4)
X=1:0.2:4;
Y=3+0.5*sin(X)-X;
V=[1 4,-2 2];
plot(X,Y,'b-',[3.0472],[0],'ro'); axis(V); grid
% set(gca,'Xtick',[]), set(gca,'YTick',[0]); grid on
legend('y=3+0.5sin(x)-x','x=3.0472',0);
title('y=3+0.5sin(x)-x')

```



Bisection Algorithm

Let f be a continuous function on $[a, b]$ and $f(a)f(b) < 0$. This algorithm is to find a $c \in [r, s] \subseteq [a, b]$ such that $|f(c)| < \varepsilon$ and $|s - r| < \delta$, where δ are user-specified small numbers.

```

input  $a, b, M, \delta, \varepsilon$ 
 $u \leftarrow f(a)$ 
 $v \leftarrow f(b)$ 
 $e \leftarrow b - a$ 
 $k \leftarrow 0$ 
print  $k, a, b, u, v$ 
if  $\text{sgn}(u) = \text{sgn}(v)$  return
for  $k=1, 2 \dots, M$ 

     $e \leftarrow e/2$ 
     $e \leftarrow a + e$ 
     $w \leftarrow f(c)$ 
print  $k, a, b, c, f(c)$ 
    if  $|e| < \delta$  or  $|w| < \varepsilon$  return
    if  $\text{sgn}(w) \neq \text{sgn}(u)$  then
         $b \leftarrow c$ 
         $v \leftarrow w$ 
    else
         $a \leftarrow c$ 
         $u \leftarrow w$ 
    endif
endfor

```

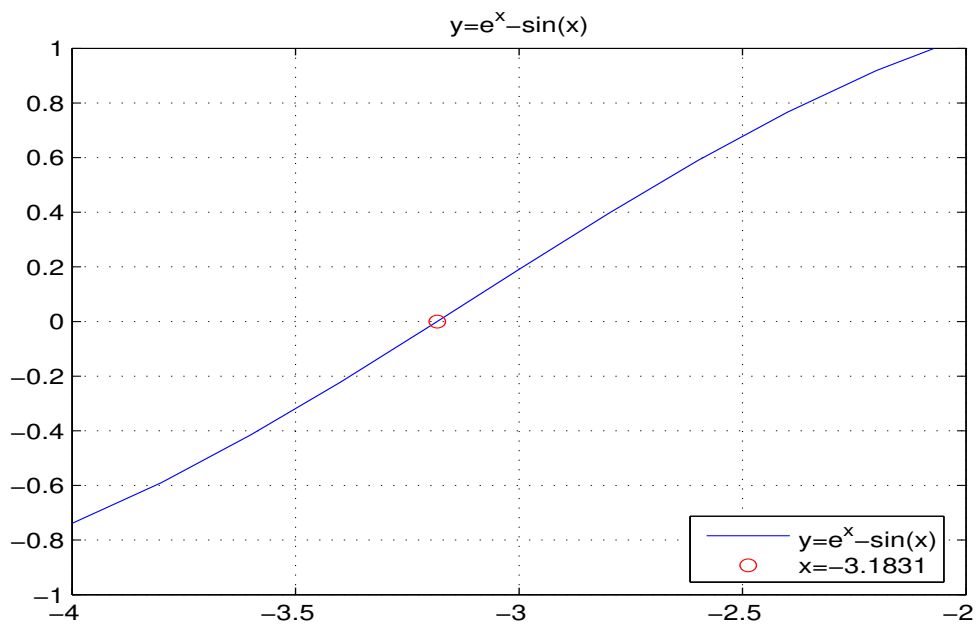


Figure 1: The function of $e^x - \sin(x)$

```

/* Brute-force method for finding Zero of f(x) */

#include <stdio.h>
#include <math.h>
main()
{
    int i, max_num=51;
    double x, y,dx=0.02;

    x=-4.0; y=exp(x)-sin(x);
    for(i=0; i<max_num; i++)
    {
        printf("%11.6f, %11.6f ",x,y);
        x+=dx; y=exp(x)-sin(x);
    }
}

```

Newton's Method

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

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)} \quad \text{for } n \geq 0$$

Algorithm

```

input  $x, M$ 
 $k \leftarrow 0$ 
 $y \leftarrow f(x)$ 
print  $k, x, y$ 
for  $k = 1, 2, \dots, M$ 
     $x \leftarrow x - \frac{y}{f'(x)}$ 
     $y \leftarrow f(x)$ 
    print  $k, x, y$ 
endfor

```

Example

$$f(x) = e^x - 1.5 - \tan^{-1} x \implies f'(x) = e^x - \frac{1}{1+x^2}$$

0	-7.0000000, -0.070189
1	-10.677096, -0.022567
2	-13.279167, -0.004366
3	-14.053656, -0.000239
4	-14.101110, -0.000001
5	-14.101270, -0.000000
6	-14.101270, -0.000000

```
/* Newton's method for finding the zero of  $f(x)=\exp(x)-1.5-\operatorname{atan}(x)$  */

#include <stdio.h>
#include <sys/time.h>
#include <math.h>

main( )
{
    int k, max_num=10;
    double x, y, z;

    x=-7.0; y=exp(x)-1.5-atan(x);

    for(k=0; k<max_num; k++)
    {
        printf("t%2d %11.6f %11.6f\n",k,x,y);
        z=exp(x)-1.0/(1.0+x*x);
        x-=(y/z);
        y=exp(x)-1.5-atan(x);
    }
}
```

Secant Method

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

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

Algorithm

```

input  $x, y, M,$ 

 $k \leftarrow 0$ 

 $zx \leftarrow f(x)$ 

 $zy \leftarrow f(y)$ 

print  $k, x, zx$ 

for  $k=1, 2 \dots, M$ 

    print  $k, y, zy,$ 

     $t \leftarrow y - zy * (y - x) / (zy - zx)$ 

     $x \leftarrow y$ 

     $zx \leftarrow zy$ 

     $y \leftarrow t$ 

     $zy \leftarrow f(y)$ 

endfor

```


Example

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

where

$$\sinh(x) = \frac{e^x - e^{-x}}{2}$$

0	7.000000, 41.683877
1	8.000000, -665.478826
2	7.058945, 20.798251
3	7.087465, 10.037673
4	7.114068, -0.401497
5	7.113045, 0.007307
6	7.113063, -0.000005
7	7.113063, 0.000000
8	7.113063, -0.000000

```

/* Secant method for finding the zero of */
/* f(x)=x sup 3 -sinh(x)+ 4x sup 2 + 6x + 9 */

#include <stdio.h>
#include <sys/time.h>
#include <math.h>

main()
{
    int k, max_num=10;
    /*double t, x, y, zx, zy */
    float t, x, y , zx, zy;

    x=7.0; y=8.0;
    zx=x*x*x-0.5*(exp(x)-exp(-x))+4.0*x*x+6.0*x+9.0;
    zy=y*y*y-0.5*(exp(y)-exp(-y))+4.0*y*y+6.0*y+9.0;

    printf("\t %2d %11.6f, %11.6f\n",k,x,zx);

    for (k=1; k<max_num; k++)
    {
        printf("\t %2d %11.6f, %11.6f\n",k,y,zy);
        t=y-zy*(y-x)/(zy-zx);
        x=y;
        zx=zy;
        y=t;
        zy=y*y*y-0.5*(exp(y)-exp(-y))+4.0*y*y+6.0*y+9.0;
    }
}

```

Fixed Point and Functional Iteration

$$x_{n+1} = F(x_n) = x_n - \frac{f(x_n)}{f'(x_n)} \quad \forall n \geq 0 \quad (\text{Newton})$$

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

Definition: A mapping $F : X \rightarrow Y$ is said to be contractive if \exists a number $\lambda < 1$ such that

$$|F(x) - F(y)| < \lambda|x - y| \quad \forall x, y \in X$$

Theorem: Let F be a contractive mapping of a compact set \mathbf{S} of \mathbf{R}^n into \mathbf{S} . Then F has a unique fixed point. Moreover, this fixed point is the limit of every sequence obtained from $\mathbf{x}_{n+1} = F(\mathbf{x}_n)$ with any starting point $\mathbf{x}_0 \in \mathbf{S}$.

Hint of Proof

show every sequence in the contractive mapping theorem satisfies Cauchy criterion for convergence. For any $\varepsilon > 0$, \exists an N such that if $n \geq m \geq N$, we have

$$\|\mathbf{x}_n - \mathbf{x}_m\| \leq \|\mathbf{x}_n - \mathbf{x}_{n-1}\| + \|\mathbf{x}_{n-1} - \mathbf{x}_{n-2}\| + \dots + \|\mathbf{x}_{m+1} - \mathbf{x}_m\|$$

$$\leq \lambda^{n-1}\|\mathbf{x}_1 - \mathbf{x}_0\| + \lambda^{n-2}\|\mathbf{x}_1 - \mathbf{x}_0\| + \dots + \lambda^m\|\mathbf{x}_1 - \mathbf{x}_0\|$$

$$\leq \frac{\lambda^N}{1 - \lambda}\|\mathbf{x}_1 - \mathbf{x}_0\| \leq \varepsilon$$

Example 1. $F(x) = 3 + \frac{1}{2}\sin x$

Example 2. $F(x) = 4 + \frac{1}{3}\sin(2x)$

Error Analysis and Accelerating Convergence

A sequence $\{p_n\}$ of approximations to a number p converges

linearly if $|p_{n+1} - p| \leq M|p_n - p|$ for large n and $0 < M < 1$

quadratically if $|p_{n+1} - p| \leq M|p_n - p|^2$ for large n and $0 < M < 1$

- *Aitken's Δ^2 method and Convergence*

If $\{p_n\}_0^\infty$ is a sequence that converges linearly to p and $(p_{n+1} - p)(p_n - p) > 0$, define

$$q_n = p_n - \frac{(p_{n+1} - p_n)^2}{p_{n+2} - 2p_{n+1} + p_n} \text{ or } q_n = p_n - \frac{(\Delta p_n)^2}{\Delta^2 p_n},$$

then $\lim_{n \rightarrow \infty} \frac{q_n - p}{p_n - p} = 0$ that is, $\{q_n\}_0^\infty$ converges to p more rapidly.

Example 1. $\{p_n = \cos(1/n)\}_{n=1}^\infty$.

n	$p_n = \cos(\frac{1}{n})$	q_n
1	0.54030	0.96178
2	0.87758	0.98213
3	0.94496	0.98979
4	0.96891	0.99342
5	0.98007	0.99541
6	0.98614	
7	0.98981	

Table 1: **The Speed of Convergence for $\cos(\frac{1}{n})$**

Proof from P.260, Kincaid

Let $h_n = p_n - p$, then

$$\begin{aligned}
 q_n &= p_n - \frac{(p_{n+1}-p_n)^2}{p_{n+2}-2p_{n+1}+p_n} \\
 &= \frac{(p_n p_{n+1} - 2p_n p_{n+1} + p_n^2) - (p_{n+1}^2 - 2p_n p_{n+1} + p_n^2)}{p_{n+2} - 2p_{n+1} + p_n} \\
 &= \frac{(p+h_n)(p+h_{n+1}) - (p+h_{n+1})^2}{(p+h_{n+2}) - 2(p+h_{n+1}) + (p+h_n)} \\
 &= \frac{p(h_n+h_{n+2}-2h_{n+1}) + h_n h_{n+1} - h_{n+1}^2}{h_{n+2} - 2h_{n+1} + h_n} \\
 &= p + \frac{h_n h_{n+1} - h_{n+1}^2}{h_{n+2} - 2h_{n+1} + h_n}
 \end{aligned}$$

Since $|p_{n+1}-p| \leq M|p_n-p|$ with $0 < M < 1$ for large n , then $\exists \delta_n \geq 0$ with $\lim_{n \rightarrow \infty} \delta_n = 0$ such that

$$h_{n+1} = (M + \delta_n)h_n, \quad h_{n+2} = (M + \delta_{n+1})(M + \delta_n)h_n.$$

Then

$$\begin{aligned}
 q_n - p &= \frac{h_n^2 [(M+\delta_{n+1})(M+\delta_n) - (M+\delta_n)^2]}{[(M+\delta_{n+1})(M+\delta_n) - 2(M+\delta_n) + 1]h_n} \\
 &= h_n \cdot \frac{(M+\delta_{n+1})(M+\delta_n) - (M+\delta_n)^2}{(M+\delta_{n+1})(M+\delta_n) - 2(M+\delta_n) + 1} \\
 &= (p_n - p) \cdot \frac{(M+\delta_{n+1})(M+\delta_n) - (M+\delta_n)^2}{(M+\delta_{n+1})(M+\delta_n) - 2(M+\delta_n) + 1}
 \end{aligned}$$

Thus

$$\frac{q_n - p}{p_n - p} = \frac{M(\delta_{n+1} - \delta_n) + \delta_{n+1}\delta_n - \delta_n^2}{(M-1)^2 + M(\delta_{n+1} + \delta_n) + \delta_{n+1}\delta_n - 2\delta_n} \rightarrow 0 \text{ as } n \rightarrow \infty$$

Müller's Method

It attempts to overcome two problems, in particular, occurred in Newton's method,

- (a) $f(x_n) \approx 0$ and $f'(x_n) \approx 0$ for some steps.
- (b) unable to find a complex root if starting with a real initial number.

Given 3 initial approximations p_0, p_1, p_2 for a solution of $f(x) = 0$. Consider a quadratic polynomial passing through $[p_0, f(p_0)], [p_1, f(p_1)], [p_2, f(p_2)]$

$$f(x) = a(x - p_2)^2 + b(x - p_2) + c$$

where a, b, c are to be determined, the next approximation p_3 is defined to be one of the roots of $f(x) = 0$ computed by

$$p_3 = p_2 - \frac{2c}{b + \operatorname{sgn}(b)\sqrt{b^2 - 4ac}} \quad (\text{the } p_3 \text{ closest to } p_2)$$

where $c = f(p_2)$

$$b = \frac{(p_0 - p_2)^2[f(p_1) - f(p_2)] - (p_1 - p_2)^2[f(p_0) - f(p_2)]}{(p_0 - p_1)(p_0 - p_2)(p_1 - p_2)}$$

$$a = \frac{(p_1 - p_2)[f(p_0) - f(p_2)] - (p_0 - p_2)[f(p_1) - f(p_2)]}{(p_0 - p_1)(p_0 - p_2)(p_1 - p_2)}$$

Notice that a, b, c are obtained by solving

$$(p_0 - p_2)^2 a + (p_0 - p_2)b + c = f(p_0)$$

$$(p_1 - p_2)^2 a + (p_1 - p_2)b + c = f(p_1)$$

$$(p_2 - p_2)^2 a + (p_2 - p_2)b + c = f(p_2)$$

- $f(x) = 16x^4 - 40x^3 + 5x^2 + 20x + 6 = 0$ has roots 1.241677, 1.970446, $-0.356062 \pm 0.162758j$.

$(p_0, p_1, p_2) = (0.5, 1.0, 1.5)$ leads to **1.24168** in 7 iterations

$(p_0, p_1, p_2) = (2.5, 2.0, 2.25)$ leads to **1.97044** in 6 iterations

$(p_0, p_1, p_2) = (0.5, -0.5, 0)$ leads to **-0.356062+0.162758j** in 8 iterations

A System of Nonlinear Equations

A typical problem of n nonlinear equations with n variables is

$$f_1(x_1, x_2, \dots, x_n) = f_1(\mathbf{x}) = \mathbf{0}$$

$$f_2(x_1, x_2, \dots, x_n) = f_2(\mathbf{x}) = \mathbf{0}$$

.....

(1)

$$f_n(x_1, x_2, \dots, x_n) = f_n(\mathbf{x}) = \mathbf{0}$$

or

$$\mathbf{f}(\mathbf{x}) = \mathbf{0}$$

By Taylor's expansion, we have

$$f_i(\mathbf{x} + \delta\mathbf{x}) = f_i(\mathbf{x}) + \sum_{j=1}^n \frac{\partial f_i(\mathbf{x})}{\partial x_j} \delta x_j + O(\|\delta\mathbf{x}\|^2) \quad (2)$$

or

$$f_i(\mathbf{x} + \delta\mathbf{x}) \approx f_i(\mathbf{x}) + \sum_{j=1}^n \frac{\partial f_i(\mathbf{x})}{\partial x_j} \delta x_j \quad (3)$$

Assume $\mathbf{x} + \delta\mathbf{x}$ is the zero of \mathbf{f} , then

$$\sum_{j=1}^n \alpha_{ij}(\delta_j) = \beta_i \quad \text{for } 1 \leq i \leq n \quad (4)$$

where

$$\alpha_{ij} \equiv \frac{\partial f_i(\mathbf{x})}{\partial x_j} \quad \text{and} \quad \beta_i \equiv -f_i(\mathbf{x}) \quad (5)$$

Once the $\delta\mathbf{x}$ is solved by using(4), the next approximated solution can be updated by

$$\mathbf{x}^{(new)} = \mathbf{x}^{(old)} + \delta\mathbf{x} \quad \text{or} \quad x_i^{(new)} = x_i^{(old)} + \delta x_i, \quad 1 \leq i \leq n \quad (6)$$

The process is iterated to convergence if an initial guess \mathbf{x} is close to a zero of \mathbf{f}

Algorithm

Given an initial guess \mathbf{x} for a zero of \mathbf{f} , take $ntrial$ Newton-Raphson iterations to improve the root. Stops if $\|\delta\mathbf{x}\|_1 \leq tol_x$ or $\|\delta\mathbf{f}(\mathbf{x})\|_1 \leq tol_f$

Subroutine Mnewton($ntrial, x, n, tol_x, tol_f, alpha, beta, indx$)

```

real  $x(n), alpha(n, n), beta(n)$ 
integer  $indx(n)$ 
for  $k=1, \dots, ntrial$ 

    call userfun( $x, alpha, beta$ )
    errf=0.
    for  $k=1, \dots, n$ 
        errf=errf +  $abs(beta(i))$ 
    endfor
    if (errf .le.  $tol_f$ ) return
    call LUdecomp( $alpha, n, indx, det$ )
    call LUbacksb( $alpha, n, indx, beta$ )
    errx=0.
    for  $k=1, \dots, n$ 
        errx = errx +  $abs(beta(i))$ 
         $x(i) = x(i) + beta(i)$ 
    endfor
    if (  $errx$  .le.  $tol_x$  ) return
endfor
Return
End

```

Example 1. Find a solution of the following equations by Newton-Raphson method using $\mathbf{x}^{(0)} = [2, 1, 1]^t$ as an initial guess.

$$xy - z^2 = 1$$

$$xyz - x^2 + y^2 = 2$$

$$e^x - e^y - z = 3$$

```
%
% The solution is [1.1659, 1.4493, 1.4625]
%      xy - z^2 = 1
% xyz - x^2 + y^2 = 2
% e^x - e^y - z = 3
% Newton-Raphson's method for Nrun iterations
% X=[x y z]'
n=3; Nrun=50;
X=[2.0 1.0 1.0]';      % initial guess
for k=1:Nrun,
    F=[X(1)*X(2)-X(3)*X(3)-1;
        X(1)*X(2)*X(3)-X(1)*X(1)+X(2)*X(2)-2.0;
        exp(X(1))-exp(X(2))-X(3)-3.0];
    A=[X(2), X(1), -2.0*X(3);
        X(2)*X(3)-2.0*X(1), X(1)*X(3)+2.0*X(2), X(1)*X(2);
        exp(X(1)), -exp(X(2)), -1.0];

    dX=(A+0.005*eye(n))\F;
    X=X-dX;
end
format short
X',F'
```

Example 2. Find a solution of the following equations by Newton-Raphson method using $\mathbf{x}^{(0)} = [4.8, 3.3, 1.5, 1.3, 0.8, 1.2]^t$ as an initial guess, where $\mathbf{f}(\mathbf{x}) = \mathbf{C} = [2, 2, 2, 2, 2, 2]^t$.

$$\begin{aligned}x_1 - x_2 &= C_1 \\x_1x_3 - x_2x_5 &= C_2 \\x_1x_3^2 - x_2x_5^2 &= C_3 \\x_1x_4 - x_2x_6 &= C_4 \\x_1x_4^2 - x_2x_6^2 &= C_5 \\x_1x_3x_4 - x_2x_5x_6 &= C_6\end{aligned}$$

```
% Newton-Raphson's method with Nrun iterations to solve
% A Nonlinear System of Equations from Professor Cheng-Yan Kao at NTU
% [C1,C2,C3,C4,C5,C6]=[2 2 2 2 2 2]
% [X1,X2,X3,X4,X5,X6]=[144.6946, 142.6946, 1.0000, 1.0001, 1.0000, 1.0001];
% n=6;
% fin=fopen('dataC.txt');
% header=fgetL(fin); % read off the header line
% C=fscanf(fin,'%d',n); % read in coefficients C1 ~ C6
% X=[5.3; 3.8; 2.2; 1.2; 1.2; 1.3]; % initial guess
%
X=[4.8; 3.3; 1.5; 1.3; 0.8; 1.2]; % initial guess
C=[2 2 2 2 2 2]';
Nrun=50; format long
for k=1:Nrun,
    F=[X(1)-X(2)-C(1); X(1)*X(3)-X(2)*X(5)-C(2); ...
        X(1)*X(3)*X(3)-X(2)*X(5)*X(5)-C(3); X(1)*X(4)-X(2)*X(6)-C(4); ...
        X(1)*X(4)*X(4)-X(2)*X(6)*X(6)-C(5); X(1)*X(3)*X(4)-X(2)*X(5)*X(6)-C(6)];
    A=[1, -1, eps, eps, eps, eps; ...
        X(3), -X(5), X(1), 0, -X(2), 0; ...
        X(3)*X(3), -X(5)*X(5), 2*X(1)*X(3), 0, -2*X(2)*X(5), 0; ...
        X(4), -X(6), 0, X(1)+eps, 0, -X(2); ...
        X(4)*X(4), -X(6)*X(6), 0, 2*X(1)*X(4), 0, -2*X(2)*X(6); ...
        X(3)*X(4), -X(5)*X(6), X(1)*X(4), X(1)*X(3), -X(2)*X(6), -X(2)*X(5)];
    dX=(A+0.005*eye(n))\F;
    X=X-dX;
end
format short
X',F'
```