

# Computer Modelling Techniques

# Numerical Methods Lecture 3: Roots of equations

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A differential equation such as:  $\frac{d}{dx}\left(\lambda \frac{dT}{dx}\right) + S(x) = 0$ 

is linear in the temperature if  $\lambda \neq \lambda(T)$ . In this case, the linear ODE can be discretised with any method and this leads to a linear algebraic equation,  $a_W T_W + a_P T_P + a_E T_E = b$ , where the coefficients *a* are independent of the temperature (L1). Assembling the *n* eqs for *n* control volumes yields a system of linear eqs, that can be solved with any direct/iterative method (L2).

However, the same equation may become nonlinear if λ = λ(T); in this case, the discretisation equation is nonlinear because the coefficients *a* depend on the temperature, and assembling the eqs leads to a system of nonlinear equations.
 This can still be solved (i) with iterations or (ii) with the methods that we see today.

### Today's menu

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- Newton-Raphson method to find roots of one equation
- Newton-Raphson method for a system of two nonlinear equations
- Newton-Raphson method for a system of *n* nonlinear equations
- Worked examples Newton-Raphson method in Matlab

Expected outcome: know the principles of the N-R method, advantages and

pitfalls; be able to implement it using matlab.

# **Solution of nonlinear equations**

In general, equations in mathematics can be recast in the form f(x) = 0.

Example:  $x^4 = 5 \implies x^4 - 5 = 0$ , or:  $e^{-x} = x \implies e^{-x} - x = 0$ 

Therefore, solving the equation means finding the root(s) of the equation.

Several methods are available:

- Bisection (bracketing method)
- Newton-Raphson (open method)
- Secant (open method)
- Brent method (bracketing+open)
- ...and so on

To be clear:

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- **Bracketing methods**: the search interval is "bracketed" between two values where the function changes sign, and this interval is narrowed down iteratively.
- **Open methods**: use only one starting guess, no bracketing is done.



The N-R method works by using an initial guess, and then successively improves it by using iterations based on the slope (gradient) of the curve.

From the previous guess  $x_i$ , we want to find the next guess  $x_{i+1}$ . First, we find the tangent to f(x) in  $x_i$ . Then we extend the tangent line till it crosses the x-axis, and set  $x_{i+1}$  as the abscissa of the zero crossing. How do we "translate" this into an iteration equation?

First-order Taylor series expansion nearby  $x_i$ :

$$f_T(x) = f(x_i) + f'(x_i)(x - x_i)$$

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This crosses the x-axis at some point where  $f_T(x) = 0$ , which identifies  $x_{i+1}$ :

$$0 = f_T(x_{i+1}) = f(x_i) + f'(x_i)(x_{i+1} - x_i)$$



Iterative procedure for new guess



## **Newton-Raphson method**

Example: consider the equation

$$f(x) = e^{-x} - x = 0$$

The derivative of f(x) is:

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 $f'(x) = -e^{-x} - 1$ 

Therefore the equation to use to find a new guess is:

$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)} = x_i - \frac{e^{-x_i} - x_i}{-e^{-x_i} - 1}$$

If we take the initial guess  $x_i = 0$ , the next guess will be:

$$x_{i+1} = 0 - \frac{1-0}{-1-1} = 0.5$$

If we continue, we will eventually converge to the solution, which is  $x \approx 0.56714$ 



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 $f(x) = e^{-x} - x = 0$ 

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Route to convergence and impact of a different initial guess:



# Newton-Raphson method - Pitfalls



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• The initial guess has to be "sufficiently" close to the root

 Convergence depends on the nature of the function, in particular its derivative (see figure):

a) nearby an inflection point (f'' = 0), iterations diverge

b) nearby a max/min, iterations oscillate or diverge (d)

c) nearby max/min, the guess jumps to another root

- No bracketing is done, divergence may occur
- Convergence is not guaranteed
- Needs knowledge of the first derivative

#### **Remedies:**

- Always set a max number of iterations
- Check that the solution is converging,  $|f(x)| \rightarrow 0$
- Alert if the guess shoots out
- Secant method derivative calculated with two guesses
- Brent method first bisection, then open methods; try:

fzero(@(x) exp(-x)-x,0,optimset('DISP','ITER'))

Figure from: Numerical Methods for Engineers, S. C. Chapra, R. P. Canale, McGraw-Hill 2015

#### Nottingham IKI CHINA I MALAYSIA Newton-Raphson method for two nonlinear eqs.

Suppose we have to solve the system of equations: u(x, y) = 0, v(x, y) = 0.

Newton-Raphson in 2 dimensions: first-order Taylor series expansion,

$$u_T(x_{i+1}, y_{i+1}) = u(x_i, y_i) + (x_{i+1} - x_i)\frac{\partial u}{\partial x}(x_i, y_i) + (y_{i+1} - y_i)\frac{\partial u}{\partial y}(x_i, y_i)$$

$$v_T(x_{i+1}, y_{i+1}) = v(x_i, y_i) + (x_{i+1} - x_i)\frac{\partial v}{\partial x}(x_i, y_i) + (y_{i+1} - y_i)\frac{\partial v}{\partial y}(x_i, y_i)$$

Jacobian

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y

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$$u_{T,i+1} = 0 \implies \frac{\partial u}{\partial x}\Big|_{i} x_{i+1} + \frac{\partial u}{\partial y}\Big|_{i} y_{i+1} = -u_{i} + x_{i} \frac{\partial u}{\partial x}\Big|_{i} + y_{i} \frac{\partial u}{\partial y}\Big|_{i} \qquad J_{i} = \begin{bmatrix} \frac{\partial u}{\partial x}\Big|_{i} & \frac{\partial u}{\partial y}\Big|_{i} \\ \frac{\partial v}{\partial x}\Big|_{i} & \frac{\partial v}{\partial y}\Big|_{i} \end{bmatrix}$$
$$v_{T,i+1} = 0 \implies \frac{\partial v}{\partial x}\Big|_{i} x_{i+1} + \frac{\partial v}{\partial y}\Big|_{i} y_{i+1} = -v_{i} + x_{i} \frac{\partial v}{\partial x}\Big|_{i} + y_{i} \frac{\partial v}{\partial y}\Big|_{i} \qquad F_{i} = \begin{bmatrix} u_{i} \\ v_{i} \end{bmatrix}$$

$$\longrightarrow J_i \cdot x_{i+1} = -F_i + J_i \cdot x_i$$

Iterative procedure for new guess

unknown

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In 2 dimensions: 
$$x_{i+1} = x_i - \frac{u_i \frac{\partial v_i}{\partial y} - v_i \frac{\partial u_i}{\partial y}}{\det(J_i)}$$
  $y_{i+1} = y_i - \frac{v_i \frac{\partial u_i}{\partial x} - u_i \frac{\partial v_i}{\partial x}}{\det(J_i)}$ 

#### Example:

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$$u(x, y) = x^{2} + xy - 10 = 0 \qquad v(x, y) = y + 3xy^{2} - 57 = 0$$

Initial guesses:  $x_0 = 1.5$ ,  $y_0 = 3.5$ 

We start off with evaluating the elements of the Jacobian:

$$\frac{\partial u}{\partial x}\Big|_{0} = 2x_{0} + y_{0} = 6.5, \quad \frac{\partial u}{\partial y}\Big|_{0} = x_{0} = 1.5$$

$$det(\boldsymbol{J}_{0}) = \frac{\partial u}{\partial x}\Big|_{0} \cdot \frac{\partial v}{\partial y}\Big|_{0} - \frac{\partial u}{\partial y}\Big|_{0} \cdot \frac{\partial v}{\partial x}\Big|_{0} = 156.125$$

$$= 156.125$$

Values of the functions at the initial guesses:

$$u_0 = x_0^2 + x_0 y_0 - 10 = -2.5$$
,  $v_0 = y_0 + 3x_0 y_0^2 - 57 = 1.625$ 

$$x_{1} = x_{0} - \frac{u_{0} \frac{\partial v}{\partial y}\Big|_{0} - v_{0} \frac{\partial u}{\partial y}\Big|_{0}}{det(\boldsymbol{J}_{0})} = 2.03603, \quad y_{1} = y_{0} - \frac{v_{0} \frac{\partial u}{\partial x}\Big|_{0} - u_{0} \frac{\partial v}{\partial x}\Big|_{0}}{det(\boldsymbol{J}_{0})} = 2.84388$$

The computation is converging towards the exact solution x = 2, y = 3.

Suppose we have to solve a system with *n* unknows and *n* nonlinear equations:

$$f_1(x_1, x_2, ..., x_n) = 0$$
  

$$f_2(x_1, x_2, ..., x_n) = 0$$
  
:  

$$f_n(x_1, x_2, ..., x_n) = 0$$

Generalisation of the Newton-Raphson method to *n* dimensions:

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#### What to take home from today's lecture

- Advantages/pitfalls of the Newton-Raphson method
- How to set the iterative procedure for the solution of one, two, or a system of nonlinear equations
- How to use Matlab to implement the solution procedure

Implement the Newton-Raphson method in Matlab to solve the equation:

 $f(x) = e^{-x} - x = 0$ 

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starting with initial guess  $x_0 = 0$ , till convergence. For convergence, consider the error evaluated at each iteration as  $err_i = |f(x_i)|$  and use  $tol = 10^{-8}$ .

```
%%%%% Computational Modelling Techniques - Part 1: Numerical Methods
 1
 2
       %%%%% Lecture 4 - Worked example 1: Solve one equation using Newton-Raphson
 3
 4
       clear all; close all; clc; % clears workspace, figures, command window
 5
 6
       %%% function f(x) = e^{(-x)} - x
 7
 8 -
      maxIt=1000; % Max number of iterations
 9 -
       tol=1e-8; % Tolerance on solution
10 -
      x=0; % Initial guess
11
12 -
       i=1;
13 -
       err=abs(exp(-x)-x); % Error is defined based on how much |f(x)| is far from zero
14 -
     while err(i)>tol & i<maxIt
15 -
           x(i+1)=x(i)-(exp(-x(i))-x(i))/(-exp(-x(i))-1);
16 -
           err(i+1)=abs(exp(-x(i+1))-x(i+1));
17 -
           i=i+1;
18 -
       end
19 -
       figure('color','w','units','Centimeters','position',[5 5 7.5 7]);
20 -
      plot(x, 'o-'); grid on; hold on; xlabel('Iterations'); ylabel('x')
21
22 -
       figure('color','w','units','Centimeters','position',[5 5 7.5 7]);
23 -
       semilogy(err, 'o-'); grid on; hold on
       xlabel('Iterations'); ylabel('Error: |f(x)|')
24 -
```

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Implement the Newton-Raphson method in Matlab to solve the system of equations:

$$u(x, y) = x^{2} + xy - 10 = 0,$$
  $v(x, y) = y + 3xy^{2} - 57 = 0$ 

starting with initial guess  $x_0 = 1.5$ ,  $y_0 = 3.5$ , till convergence. For convergence,

consider the error evaluated at each iteration as  $err_i = |u_i| + |v_i|$  and use  $tol = 10^{-8}$ .

1	%%%%% Computational Modelling Techniques - Part 1: Numerical Methods
2	%%%%% Lecture 4 - Worked example 2: Solve two nonlinear equations using
3	%%%%% Newton-Raphson
4	
5 -	clear all; close all; clc; % clears workspace, figures, command window
6	
7	%%% functions $u(x,y)=x^2+xy-10$ and $v(x,y)=y+3xy^2-57$
8	
9 -	maxIt=1000; tol=1e-8; % Max number of iterations and tolerance
10 -	x=1.5; y=3.5; % Initial guesses
11	
12 -	err=sum(abs(x^2+x*y-10)+abs(y+3*x*y^2-57)); % Error is err= u + v
13 -	i=1;
14 -	🖓 while err(i)>tol 💪 i <maxit< td=""></maxit<>
15	
16 -	J(1,1)=2*x(i)+y(i); % du/dx
17 -	J(1,2)=x(i); % du/dy
18 -	$J(2,1)=3*y(1)^2; % dv/dx$
19 -	J(2,2)=1+6*x(i)*y(i); % dv/dy
20 -	F(1,1)=x(i)^2+x(i)*y(i)-10; % u(x_i,y_i)
21 -	F(2,1)=y(i)+3*x(i)*y(i)^2-57; % v(x_i,y_i)
22 -	X(1,1)=x(i); X(2,1)=y(i); % defines vector X_i=[x_i;y_i]
23	
24 -	X=J(-F+J*X); % Backslash operator to solve the linear system
25 -	x(i+1)=X(1); y(i+1)=X(2); % New guess values
26	
27 -	F(1)=x(i+1)^2+x(i+1)*y(i+1)-10; % Needed to compute the new error
28 -	F(2)=y(i+1)+3*x(i+1)*y(i+1)^2-57; % Needed to compute the new error
29 -	err(i+1)=sum(abs(F)); % Error is e= u + v
30	
31 -	i=i+1;
32 -	end
33 -	<pre>figure('color','w','units','Centimeters','position',[5 5 7.5 7]);</pre>
34 -	<pre>plot(x, 'o-'); hold on; plot(y, 'o-');</pre>
35 -	<pre>grid on; xlabel('Iterations'); ylabel('Solutions'); legend('x','y')</pre>
36 -	<pre>figure('color','w'); semilogy(err,'o-'); grid on</pre>
37 -	<pre>xlabel('Iterations'); ylabel('Error:  F(X) ')</pre>