# Nonlinear Rootfinding (Introduction)

# Introduction

### **Problem Statement**

### **Rootfinding Problem**

Given a continuous scalar function of a scalar variable, find a real number r such that f(r)=0.

- r is a **root** of the function f.
- The formulation f(x)=0 is general enough; e.g., to solve g(x)=h(x), set f=g-h and find a root of f.

### **Iterative Methods**

- Unlike the earlier linear problems, the root cannot be produced in a finite number of operations.
- Rather, a sequence of approximations that formally converge to the root is pursued.

#### **Iteration Strategy for Rootfinding.** To find the root of f:

- **1** Start with an initial iterate, say  $x_0$ .
- **2** Generate a sequence of iterates  $x_1, x_2, \ldots$  using an iteration algorithm of the form

$$x_{k+1} = g(x_k), \quad k = 0, 1, \dots$$

**3** Continue the iteration process until you find an  $x_i$  such that  $f(x_i)=0$ . (In practice, continue until some member of the sequence seems to be "good enough".)

### MATLAB's FZERO

fzero is MATLAB's general purpose rootfinding tool.

#### Syntax:

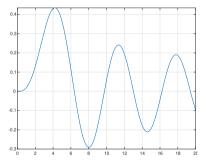
```
x_zero = fzero( <function>, <initial iterate> )
x_zero = fzero( <function>, <initial interval> )
[x_zero, fx_zero] = ....
```

### Example

#### The roots of $J_m$ , a Bessel function of the first kind, is found by

- Plot the function.
- Find approximate locations of roots.

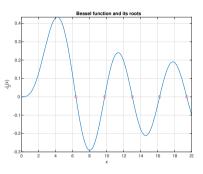
```
J3 = @(x) besselj(3,x);
fplot(J3,[0 20])
grid on
guess = [6,10,13,16,19];
```



# Example (cont')

• Then use fzero to locate the roots:

```
omega = zeros(size(guess));
for j = 1:length(guess)
  omega(j) = fzero(J3,guess(j));
end
hold on
plot(omega,J3(omega),'ro')
```



# Conditioning

 Sensitivity of the rootfinding problem can be measured in terms of the condition number:

(absolute condition number) = 
$$\frac{|abs. error in output|}{|abs. error in input|}$$
,

where, in the context of finding roots of f,

• input: f (function)

• output: r (root)

- Denote the changes by:
  - error/change in input:  $\epsilon a$ , where  $\epsilon > 0$  is small

$$(f \mapsto f + \epsilon g)$$
  
 $(r \mapsto r + \Delta r)$ 

• error/change in output:  $\Delta r$ 

$$(r \mapsto r + \Delta r)$$

# Conditioning (cont')

The perturbed equation

$$f(r + \Delta r) + \epsilon g(r + \Delta r) = 0$$

is linearized to (Taylor expansion)

$$f(r) + f'(r)\Delta r + g(r)\epsilon + g'(r)\epsilon\Delta r \approx 0,$$

ignoring  $O((\Delta r)^2)$  terms<sup>1</sup>.

• Since f(r) = 0, we solve for  $\Delta r$  to get

$$\Delta r \approx -\epsilon \frac{g(r)}{f'(r) + \epsilon g'(r)} \approx -\epsilon \frac{g(r)}{f'(r)},$$

for small  $\epsilon$  compared with f'(r).

<sup>&</sup>lt;sup>1</sup>That is, terms involving  $(\Delta r)^2$  and higher powers of  $\Delta r$ 

# Conditioning (cont')

Therefore, the absolute condition number of the rootfinding problem is

$$\kappa_{f \mapsto r} = \frac{1}{\left| f'(r) \right|},$$

which implies that the problem is highly sensitive whenever  $f'(r) \approx 0$ .

• In other words, if |f'| is small at the root, a computed *root estimate* may involve large errors.

### Residual and Backward Error

- Without knowing the exact root, we cannot compute the error.
- But the **residual** of a root estimate  $\tilde{r}$  can be computed:

(residual) = 
$$f(\tilde{r})$$
.

- Small residual might be associated with a small error.
- The residual  $f(\tilde{r})$  is the *backward error* of the estimate.

# **Multiple Roots**

### Definition 1 (Multiplicity of Roots)

Assume that r is a root of the differentiable function f. Then if

$$0 = f(r) = f'(r) = \dots = f^{(m-1)}(r)$$
 but  $f^{(m)}(r) \neq 0$ ,

we say that f has a root of **multiplicity** m at r.

- We say that f has a **multiple root** at r if the multiplicity is greater than 1.
- A root is called **simple** if its multiplicity is 1.
- If r is a multiple root, the condition number is infinite.
- Even if r is a simple root, we expect difficulty in numerical computation if  $f'(r) \approx 0$ .