

# CMPSC/Math 451, Numerical Computation

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# Numerical solution of nonlinear equations; Introduction

**Problem:** Given  $f(x)$ : continuous, real-valued, possibly non-linear. Find a root  $r$  of  $f(x)$  such that  $f(r) = 0$ .

**Example 1.** Quadratic polynomials:  $f(x) = x^2 + 5x + 6$ .

$$f(x) = (x + 2)(x + 3) = 0, \quad \Rightarrow \quad r_1 = -2, \quad r_2 = -3.$$

Observation: Roots are not unique.

**Example 2.**  $f(x) = x^2 + 4x + 10 = (x + 2)^2 + 6$ . No roots.

Observation: There are no real  $r$  that would satisfy  $f(r) = 0$ .

**Example 3.**  $f(x) = x^2 + \cos x + e^x + \sqrt{x + 1}$ .

Observation: Roots can be difficult/impossible to find analytically.

## Overview of the chapter:

- Bisection
- Fixed point iteration
- Newton's method
- Secant method

# Bisection method

## Basic idea:

Given  $f(x)$ , a continuous function.

If we find some  $a$  and  $b$ , such that  $f(a)$  and  $f(b)$  are of opposite sign, then, there exists a point  $c$ , between  $a$  and  $b$ , such that  $f(c) = 0$ .

This fact follows from the continuity of  $f$ .

We can iterate on this idea!

## Procedure:

- Initialization: Find  $a, b$  such that  $f(a) \cdot f(b) < 0$ .  
This means there is a root  $r \in (a, b)$  s.t.  $f(r) = 0$ .
- Let  $c = \frac{a+b}{2}$ , mid-point.
- If  $f(c) = 0$ , done (lucky!)  
else:
  - if  $f(c) \cdot f(a) < 0$ , pick the interval  $[a, c]$
  - if  $f(c) \cdot f(b) < 0$ , pick the interval  $[c, b]$ ,
- Iterate the procedure until stop criteria satisfied.

## Stop Criteria:

- 1) interval small enough, i.e.,  $(b - a) \leq \epsilon$ ,
- 2)  $|f(c)|$  very small, i.e,  $|f(c)| \leq \epsilon$
- 3) max number of iteration reached. (to avoid dead loop, in case the method does not converge.)
- 4) any combination of the previous ones.

# Convergence analysis:

Consider  $[a_0, b_0]$ ,  $c_0 = \frac{a_0+b_0}{2}$ , let  $r \in (a_0, b_0)$  be a root.

$$\text{The error: } e_0 = |r - c_0| \leq \frac{b_0 - a_0}{2}$$

Denote the further intervals as  $[a_n, b_n]$  for iteration number  $n$ .

$$e_n = |r - c_n| \leq \frac{b_n - a_n}{2} \leq \frac{b_0 - a_0}{2^{n+1}} = \frac{e_0}{2^n}.$$

If the error tolerance is  $\varepsilon$ , we require  $e_n \leq \varepsilon$ , then

$$\frac{b_0 - a_0}{2^{n+1}} \leq \varepsilon \quad \Rightarrow \quad n \geq \frac{\ln(b - a) - \ln(2\varepsilon)}{\ln 2}, \quad (\# \text{ of steps})$$

Remark: very slow convergence.

# Fixed point iterations

We rewrite the equation  $f(x) = 0$  into the form  $x = g(x)$ .

Remark: This can always be achieved, for example:  $x = f(x) + x$ . However, the choice of  $g$  makes a difference in convergence.

## Main idea:

Make a guess of the solution, say  $\bar{x}$ . If the function  $g(x)$  is “nice”, then hopefully,  $g(\bar{x})$  should be closer to the answer than  $\bar{x}$ . If that is the case, then we can iterate.



Iteration algorithm:

- Choose a start point  $x_0$ ,
- Do the iteration  $x_{k+1} = g(x_k)$ ,  $k = 0, 1, 2, \dots$  until meeting stop criteria.

Stop Criteria: Let  $\varepsilon$  be the tolerance

- $|x_k - x_{k-1}| \leq \varepsilon$ ,
- max # of iteration reached,
- any combination.

**Example 1.** Find an approximate solution to  $f(x) = x - \cos x = 0$ , with 4 digits accuracy.

Choose  $g(x) = \cos x$ , we have  $x = \cos x$ .

Choose  $x_0 = 1$ , and do the iteration  $x_{k+1} = \cos(x_k)$ :

$$x_1 = \cos x_0 = 0.5403$$

$$x_2 = \cos x_1 = 0.8576$$

$$x_3 = \cos x_2 = 0.6543$$

$$\vdots$$

$$x_{23} = \cos x_{22} = 0.7390$$

$$x_{24} = \cos x_{23} = 0.7391$$

$$x_{25} = \cos x_{24} = 0.7391 \quad \text{stop here}$$

Our approximation to the root is 0.7391.

**Example 2.** Consider  $f(x) = e^{-2x}(x - 1) = 0$ . (root:  $r = 1$ ).

Rewrite as

$$x = g(x) = e^{-2x}(x - 1) + x$$

Choose an initial guess  $x_0 = 0.99$ , very close to the real root. Iterations:

$$x_1 = g(x_0) = 0.9886$$

$$x_2 = g(x_1) = 0.9870$$

$$x_3 = g(x_2) = 0.9852$$

$$\vdots$$

$$x_{27} = 0.1655$$

$$x_{28} = -0.4338$$

$$x_{29} = -3.8477 \quad \text{Diverges. It does not work.}$$

What went wrong?

# Fixed point iteration; Convergence analysis.

Our iteration is  $x_{k+1} = g(x_k)$ . Let  $r$  be the exact root, s.t.,  $r = g(r)$ . Define the error:  $e_k = |x_k - r|$ .

$$\begin{aligned} e_{k+1} &= |x_{k+1} - r| = |g(x_k) - r| = |g(x_k) - g(r)| \\ &= |g'(\xi)| |(x_k - r)| \quad (\xi \in (x_k, r), \text{ since } g \text{ is continuous}) \\ &= |g'(\xi)| e_k \end{aligned}$$

$$\Rightarrow e_{k+1} = |g'(\xi)| e_k.$$

Observation:

- If  $|g'(\xi)| < 1$ , then  $e_{k+1} < e_k$ , error decreases, the iteration converges. (linear convergence)
- If  $|g'(\xi)| > 1$ , then  $e_{k+1} > e_k$ , error increases, the iteration diverges.

## Convergence condition:

There exists an interval around  $r$ , say  $[r - a, r + a]$  for some  $a > 0$ , such that  $|g'(x)| < 1$  for almost all  $x \in [r - a, r + a]$ , and the initial guess  $x_0$  lies in this interval.

In Example 1,  $g(x) = \cos x$ ,  $g'(x) = \sin x$ ,  $r = 0.7391$ ,

$$|g'(r)| = |\sin(0.7391)| < 1. \quad \text{OK, convergence.}$$

In Example 2, we have

$$\begin{aligned} g(x) &= e^{-2x}(x-1) + x, \\ g'(x) &= -2e^{-2x}(x-1) + x^{-2x} + 1 \end{aligned}$$

With  $r = 1$ , we have

$$|g'(r)| = e^{-2} + 1 > 1, \quad \text{Divergence.}$$

# Pseudo code

```
r=fixedpoint('g', x,tol,nmax)
  r=g(x); % first iteration
  nit=1;
  while (abs(r-g(r))>tol and nit < nmax) do
    r=g(r);
    nit=nit+1;
  end
end
```

# A practical error estimate

Assume  $|g'(x)| \leq m < 1$  in  $[r - a, r + a]$ . We have  $e_{k+1} \leq me_k$ .

This gives

$$e_1 \leq me_0, \quad e_2 \leq me_1 \leq m^2 e_0, \quad \dots \quad e_k \leq m^k e_0.$$

This result is useless unless we find a way to estimating  $e_0$ .

$$e_0 = |r - x_0| = |r - x_1 + x_1 - x_0| \leq e_1 + |x_1 - x_0| \leq me_0 + |x_1 - x_0|$$

then

$$e_0 \leq \frac{1}{1 - m} |x_1 - x_0|, \quad (\text{can be computed})$$

Put together

$$e_k \leq \frac{m^k}{1 - m} |x_1 - x_0|.$$



$$e_k \leq \frac{m^k}{1-m} |x_1 - x_0|.$$

If the error tolerance is  $\varepsilon$ , then

$$\frac{m^k}{1-m} |x_1 - x_0| \leq \varepsilon,$$

$$m^k \leq \frac{\varepsilon(1-m)}{|x_1 - x_0|}$$

$$k \geq \frac{\ln(\varepsilon(1-m)) - \ln |x_1 - x_0|}{\ln m}$$

which give the minimum number of iterations needed to achieve an error  $\leq \varepsilon$ .

**Example 3.** We want to solve  $\cos x - x = 0$  with the fixed point iteration

$$x = g(x) = \cos x, \quad x_0 = 1,$$

with error tolerance  $\varepsilon = 10^{-5}$ . Find the min# iterations.

We know  $r \in [0, 1]$ .

We see that the iteration happens between  $x = 0$  and  $x = 1$ .

For  $x \in [0, 1]$ , we have

$$|g'(x)| = |\sin x| \leq \sin 1 = 0.8415, \quad m \doteq 0.8415$$

And  $x_1 = \cos x_0 = 0.5403$ . Using the formula

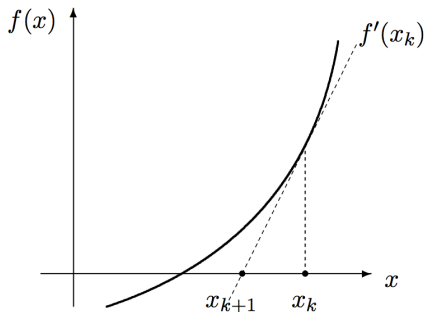
$$k \geq \frac{\ln(\varepsilon(1 - m)) - \ln |x_1 - x_0|}{\ln m} \approx 73 \quad \text{\#iterations needed}$$

In actual simulation  $k = 25$  is enough.

# Newton's method

**Goal:** Given  $f(x)$ , find a root  $r$  s.t.  $f(r) = 0$ .

**Main idea:** Given  $x_k$ , the next approximation  $x_{k+1}$  is determined by approximating  $f(x)$  as a linear function at  $x_k$ .



We have

$$\frac{f(x_k)}{x_k - x_{k+1}} = f'(x_k)$$

which gives

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}$$

# Newton's method is a fixed point iteration.

Newton's method is a fixed point iteration

$$x_{k+1} = g(x_k), \quad g(x) = x - \frac{f(x)}{f'(x)}.$$

If  $r$  is a fixed point, assume  $f'(r) \neq 0$ , then

$$r = g(r), \quad r = r - \frac{f(r)}{f'(r)}, \quad \frac{f(r)}{f'(r)} = 0, \quad f(r) = 0.$$

then  $r$  is a root for  $f$ .

# Newton's method is the “best” fixed point iteration.

Observation: A fixed point iteration  $x = g(x)$  is optimal if  $g'(r) = 0$  where  $r = g(r)$ .

For Newton, we have

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

Then

$$g'(r) = \frac{f''(r)f(r)}{(f'(r))^2} = 0$$

which is the “best” possible scenario!

# Newton's method; Convergence analysis.

Let  $r$  be the root so  $f(r) = 0$  and  $r = g(r)$ . Recall  $g'(r) = 0$ .

$$\text{Define Error: } e_{k+1} = |x_{k+1} - r| = |g(x_k) - g(r)|$$

Taylor expansion for  $g(x_k)$  at  $r$ :

$$\begin{aligned} g(x_k) &= g(r) + (x_k - r)g'(r) + \frac{1}{2}(x_k - r)^2 g''(\xi), \quad \xi \in (x_k, r) \\ &= g(r) + \frac{1}{2}(x_k - r)^2 g''(\xi) \end{aligned}$$

$$e_{k+1} = \frac{1}{2}(x_k - r)^2 |g''(\xi)| = \frac{1}{2}e_k^2 |g''(\xi)|$$

Write now  $M = \frac{1}{2} \max_x |g''(\xi)|$ , we have

$$e_{k+1} \leq M (e_k)^2$$

This is called *quadratic convergence*.

Theorem: If  $e_{k+1} \leq M e_k^2$ , then  $\lim_{k \rightarrow +\infty} e_k = 0$  if  $e_0$  is sufficiently small.  
(This means,  $M$  can be big, but it would not effect the convergence!)

Proof for the convergence: We have

$$e_1 \leq (M e_0) e_0$$

If  $e_0$  is small enough, such that  $(M e_0) < 1$ , then  $e_1 < e_0$ .

This means  $(M e_1) < M e_0 < 1$ , and so

$$e_2 \leq (M e_1) e_1 < e_1, \quad \Rightarrow \quad M e_2 < M e_1 < 1$$

By an induction argument, we conclude that  $e_{k+1} < e_k$  for all  $k$ , i.e., error is strictly decreasing after each iteration.  $\Rightarrow$  convergence.

# Newton's iterations; Examples

**Example** . Find a numerical method to compute  $\sqrt{a}$  using only  $+$ ,  $-$ ,  $*$ ,  $/$  arithmetic operations. Test it for  $a = 3$ .

**Answer.** It's easy to see that  $\sqrt{a}$  is a root for  $f(x) = x^2 - a$ .  
Newton's method gives

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)} = x_k - \frac{x_k^2 - a}{2x_k} = \frac{x_k}{2} + \frac{a}{2x_k}$$

Test it on  $a = 3$ : Choose  $x_0 = 1.7$ .

	error
$x_0 = 1.7$	$7.2 \times 10^{-2}$
$x_1 = 1.7324$	$3.0 \times 10^{-4}$
$x_2 = 1.7321$	$2.6 \times 10^{-8}$
$x_3 = 1.7321$	$4.4 \times 10^{-16}$

Note the extremely fast convergence.



## Sample Code:

```
r=newton('f','df',x0,nmax,tol)
x=x0;
n=0;
dx=f(x)/df(x);
while ((dx>tol) and (f(x)>tol)) or (n<nmax) do
    n=n+1;
    x=x-dx;
    dx=f(x)/df(x);
end
r=x-dx;
```

# Secant method

If  $f(x)$  is complicated,  $f'(x)$  might not be available.

Solution for this situation: using approximation for  $f'$ , i.e.,

$$f'(x_k) \approx \frac{f(x_k) - f(x_{k-1})}{x_k - x_{k-1}}$$

This is *secant method*:

$$x_{k+1} = x_k - \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})} f(x_k)$$

Initial guess needed:  $x_0, x_1$ .

**Advantages** include

- No computation of  $f'$ ;
- One  $f(x)$  computation each step;
- Also rapid convergence.

A bit on convergence: One can show that

$$e_{k+1} \leq C e_k^\alpha, \quad \alpha = \frac{1}{2}(1 + \sqrt{5}) \approx 1.62$$

This is called *super linear convergence*. ( $1 < \alpha < 2$ )

It converges for all function  $f$  if  $x_0$  and  $x_1$  are close to the root  $r$ .

**Example** Use secant method for computing  $\sqrt{a}$ .

**Answer.** The iteration now becomes

$$x_{k+1} = x_k - \frac{(x_k^2 - a)(x_k - x_{k-1})}{(x_k^2 - a) - (x_{k-1}^2 - a)} = x_k - \frac{x_k^2 - a}{x_k + x_{k-1}}$$

Test with  $a = 3$ , with initial data  $x_0 = 1.65$ ,  $x_1 = 1.7$ .

	error
$x_1 = 1.7$	$7.2 \times 10^{-2}$
$x_2 = 1.7328$	$7.9 \times 10^{-4}$
$x_3 = 1.7320$	$7.3 \times 10^{-6}$
$x_4 = 1.7321$	$1.7 \times 10^{-9}$
$x_5 = 1.7321$	$3.6 \times 10^{-15}$

It is a little but slower than Newton's method, but not much.

## The most practical algorithm: hybrid methods

- Sample the function to find  $a$  and  $b$  such that  $f(a) \cdot f(b) < 0$ .
- Use bisection's method, maybe 5-6 iterations, to get some good initial guess  $x_0$ ;
- Use either Newton or secant method, with this  $x_0$ , for 3-4 iterations.