

# Lecture 7

Finding the Roots of  $f(x) = 0$ : Fixed Point, Bisection, Newton

L. Olson

Department of Computer Science  
University of Illinois at Urbana-Champaign

Slides based on NMM slides from Recktenwald

February 6, 2006

# Root-Finding Algorithms

Recall, once we find an initial guess (`brackPlot.m`) we can use a root-finding algorithm:

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

# Fixed Point Iteration

To solve

$$f(x) = 0$$

we rewrite as

$$x_{\text{new}} = g(x_{\text{old}})$$

## Listing 1: Fixed Point Iteration

```
1 initialize:  $x_0 = \dots$ 
2 for  $k = 1, 2, \dots$ 
3      $x_k = g(x_{k-1})$ 
4     if converged, stop
5 end
```

# Fixed Point Iteration Example (1)

To solve

$$x - x^{1/3} - 2 = 0$$

rewrite as

$$x_{\text{new}} = g_1(x_{\text{old}}) = x_{\text{old}}^{1/3} + 2$$

or

$$x_{\text{new}} = g_2(x_{\text{old}}) = (x_{\text{old}} - 2)^3$$

or

$$x_{\text{new}} = g_3(x_{\text{old}}) = \frac{6 + 2x_{\text{old}}^{1/3}}{3 - x_{\text{old}}^{2/3}}$$

Are these  $g(x)$  functions equally effective?

## Fixed Point Iteration Example (2)

$$g_1(x) = x^{1/3} + 2$$

$$g_2(x) = (x - 2)^3$$

$$g_3(x) = \frac{6 + 2x^{1/3}}{3 - x^{2/3}}$$

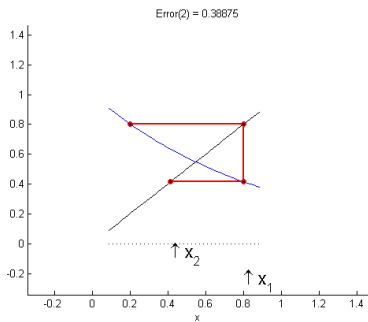
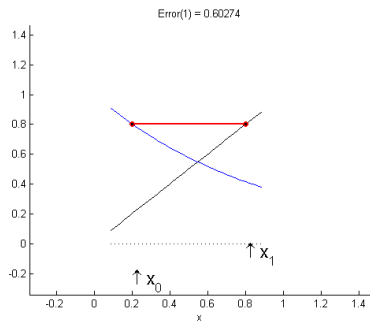
$k$	$g_1(x_{k-1})$	$g_2(x_{k-1})$	$g_3(x_{k-1})$
0	3	3	3
1	3.4422495703	1	3.5266442931
2	3.5098974493	-1	3.5213801474
3	3.5197243050	-27	3.5213797068
4	3.5211412691	-24389	3.5213797068
5	3.5213453678	$-1.451 \times 10^{13}$	3.5213797068
6	3.5213747615	$-3.055 \times 10^{39}$	3.5213797068
7	3.5213789946	$-2.852 \times 10^{118}$	3.5213797068
8	3.5213796042	$\infty$	3.5213797068
9	3.5213796920	$\infty$	3.5213797068

### Summary

$g_1(x)$  converges,  $g_2(x)$  diverges,  $g_3(x)$  converges very quickly

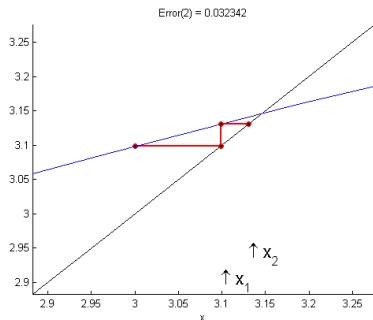
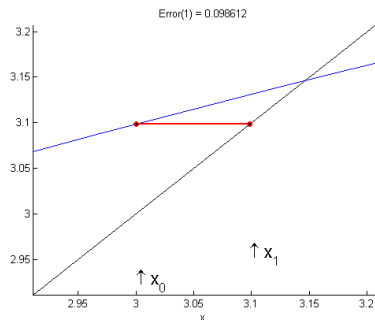
# Fixed Point

$$g(x) = 3^{-x}$$



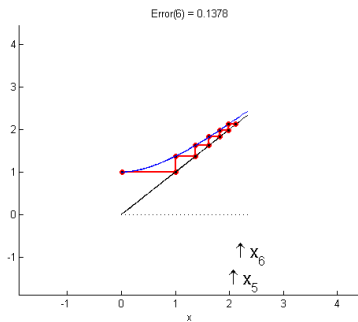
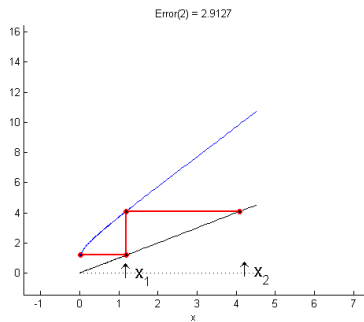
# Fixed Point

$$g(x) = \log(x) + 2$$



# Fixed Point

$$g(x) = e^{x^{0.4}} + x \quad g(x) = e^{-x} + x$$



# Fixed Point

`fixedpointplot.m`

# Fixed Point Theorem

## Theorem

*Fixed Point Let  $g \in C[a, b]$  be such that  $g(x) \in [a, b]$ , for all  $x$  in  $[a, b]$ . Suppose, in addition that  $g'(x)$  exists on  $(a, b)$  and that a constant  $k \in (0, 1)$  exists with*

$$|g'(x)| \leq k, \quad \text{for all } x \in (a, b).$$

*Then, for any guess  $x_0 \in [a, b]$ , the sequence defined by*

$$x_n = g(x_{n-1})$$

*converges to the unique fixed point  $x$  in  $[a, b]$ .*

## Fixed Point Theorem: in words

Assuming  $g(x)$  is smooth and bounded (in  $[a, b]$ ) and that it's not too steep ( $|g'(x)| \leq 1$ ), then we converge to a fixed point.

- in fact,  $k$  is associated with the rate of convergence:

$$|x_n - x| \leq k^n \max\{x_0 - a, b - x_0\}$$

and

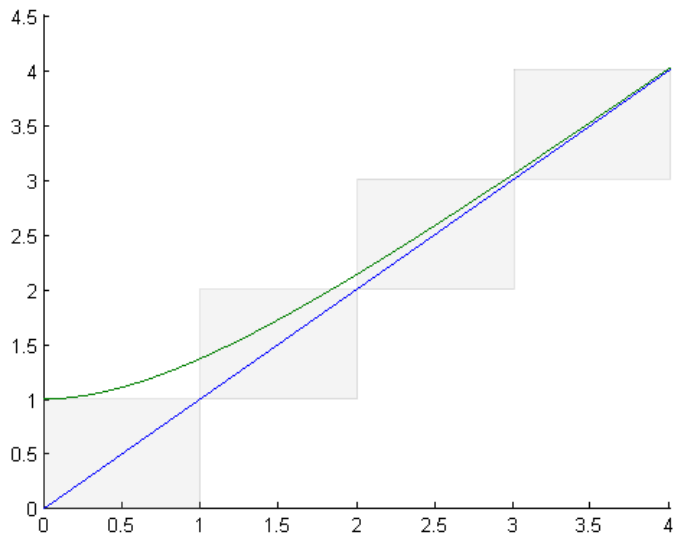
$$|x_n - x| \leq \frac{k^n}{1 - k} |x_1 - x_0|$$

- so convergence is fast if  $k \approx 0$
- and convergence is slow if  $k \approx 1$

## Fixed Point Theorem: cases

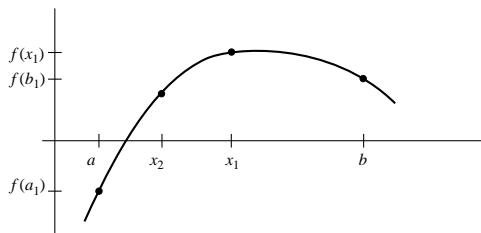
- $g(x) = 3^{-x}$  gives  $g'(x) = -3^{-x} \log(3)$  which can be bounded arbitrarily small
- $g(x) = \log(x) + 2$  gives  $g'(x) = \frac{1}{x}$  which can be bounded arbitrarily small.
- $g(x) = e^{x^{0.4}} + x$  gives  $g'(x) = e^{x^{0.4}} \frac{0.4}{x^{0.6}} + 1$  which is always greater than one
- $g(x) = e^{-x} + x$  gives  $g'(x) = -e^{-x} + 1$  which is always less than one, but  $g(x)$  is not in  $[a, b]$

# Fixed Point Theorem: $g(x) = \exp(-x) + x$



# Bisection

Given a bracketed root, halve the interval while continuing to bracket the root



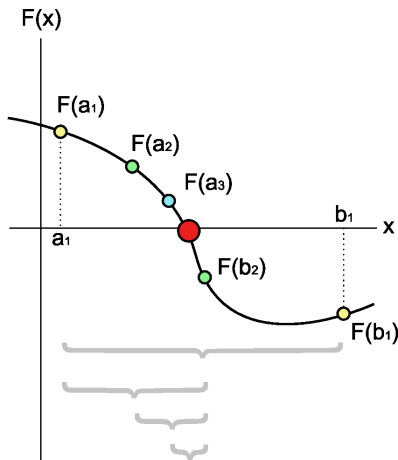
## Bisection (2)

For the bracket interval  $[a, b]$  the midpoint is

$$x_m = \frac{1}{2}(a + b)$$

idea:

- 1 split bracket in half
- 2 select the bracket that has the root
- 3 goto step 1



# Bisection Algorithm

Listing 2: Bisection

```
1 initialize:  $a = \dots, b = \dots$ 
2 for  $k = 1, 2, \dots$ 
3    $x_m = a + (b - a)/2$ 
4   if  $\text{sign}(f(x_m)) = \text{sign}(f(x_a))$ 
5      $a = x_m$ 
6   else
7      $b = x_m$ 
8   end
9   if converged, stop
10 end
```

# Bisection Example

Solve with bisection:

$$x - x^{1/3} - 2 = 0$$

$k$	$a$	$b$	$x_{mid}$	$f(x_{mid})$
0	3	4		
1	3	4	3.5	-0.01829449
2	3.5	4	3.75	0.19638375
3	3.5	3.75	3.625	0.08884159
4	3.5	3.625	3.5625	0.03522131
5	3.5	3.5625	3.53125	0.00845016
6	3.5	3.53125	3.515625	-0.00492550
7	3.51625	3.53125	3.5234375	0.00176150
8	3.51625	3.5234375	3.51953125	-0.00158221
9	3.51953125	3.5234375	3.52148438	0.00008959
10	3.51953125	3.52148438	3.52050781	-0.00074632

## Analysis of Bisection (1)

Let  $\delta_n$  be the size of the bracketing interval at the  $n^{\text{th}}$  stage of bisection. Then

$$\delta_0 = b - a = \text{initial bracketing interval}$$

$$\delta_1 = \frac{1}{2}\delta_0$$

$$\delta_2 = \frac{1}{2}\delta_1 = \frac{1}{4}\delta_0$$

$\vdots$

$$\delta_n = \left(\frac{1}{2}\right)^n \delta_0$$

$$\implies \frac{\delta_n}{\delta_0} = \left(\frac{1}{2}\right)^n = 2^{-n}$$

$$\text{or} \quad n = \log_2 \left( \frac{\delta_0}{\delta_n} \right)$$

## Analysis of Bisection (2)

$$\frac{\delta_n}{\delta_0} = \left(\frac{1}{2}\right)^n = 2^{-n} \quad \text{or} \quad n = \log_2 \left(\frac{\delta_n}{\delta_0}\right)$$

$n$	$\frac{\delta_n}{\delta_0}$	function evaluations
5	$3.1 \times 10^{-2}$	7
10	$9.8 \times 10^{-4}$	12
20	$9.5 \times 10^{-7}$	22
30	$9.3 \times 10^{-10}$	32
40	$9.1 \times 10^{-13}$	42
50	$8.9 \times 10^{-16}$	52

# Convergence Criteria

An automatic root-finding procedure needs to monitor progress toward the root and stop when current guess is close enough to the desired root.

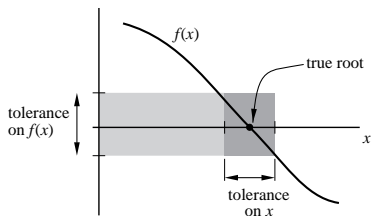
- Convergence checking will avoid searching to unnecessary accuracy.
- Check how closeness of successive approximations

$$|x_k - x_{k-1}| < \delta_x$$

- Check how close  $f(x)$  is to zero at the current guess.

$$|f(x_k)| < \delta_f$$

# Convergence Criteria on $x$



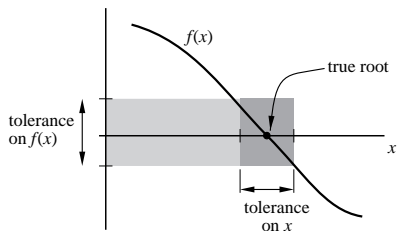
$x_k$  = current guess at the root

$x_{k-1}$  = previous guess at the root

**Absolute tolerance:**  $|x_k - x_{k-1}| < \delta_x$

**Relative tolerance:**  $\frac{|x_k - x_{k-1}|}{|x_k|} < \hat{\delta}_x$

# Convergence Criteria on $f(x)$



**Absolute** tolerance:  $|f(x_k)| < \delta_f$

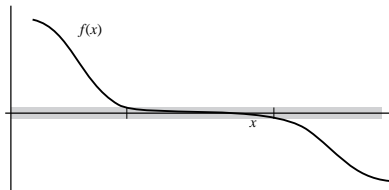
**Relative** tolerance:

$$|f(x_k)| < \hat{\delta}_f \max\{|f(a_0)|, |f(b_0)|\}$$

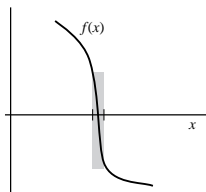
where  $a_0$  and  $b_0$  are the original brackets

# Convergence Criteria on $f(x)$

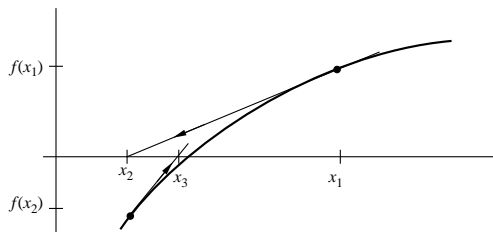
If  $f'(x)$  is small near the root, it is easy to satisfy tolerance on  $f(x)$  for a large range of  $\Delta x$ . The tolerance on  $\Delta x$  is more conservative



If  $f'(x)$  is large near the root, it is possible to satisfy the tolerance on  $\Delta x$  when  $|f(x)|$  is still large. The tolerance on  $f(x)$  is more conservative



# Newton's Method (1)



For a current guess  $x_k$ , use  $f(x_k)$  and the slope  $f'(x_k)$  to predict where  $f(x)$  crosses the  $x$  axis.

## Newton's Method (2)

Expand  $f(x)$  in Taylor Series around  $x_k$

$$f(x_k + \Delta x) = f(x_k) + \Delta x \left. \frac{df}{dx} \right|_{x_k} + \frac{(\Delta x)^2}{2} \left. \frac{d^2f}{dx^2} \right|_{x_k} + \dots$$

Substitute  $\Delta x = x_{k+1} - x_k$   
and neglect 2<sup>nd</sup> order terms to get

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

where

$$f'(x_k) = \left. \frac{df}{dx} \right|_{x_k}$$

## Newton's Method (3)

Goal is to find  $x$  such that  $f(x) = 0$ .

Set  $f(x_{k+1}) = 0$  and solve for  $x_{k+1}$

$$0 = f(x_k) + (x_{k+1} - x_k)f'(x_k)$$

or, solving for  $x_{k+1}$

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

# Newton's Method Algorithm

```
1 initialize:  $x_1 = \dots$   
2 for  $k = 2, 3, \dots$   
3    $x_k = x_{k-1} - f(x_{k-1})/f'(x_{k-1})$   
4   if converged, stop  
5 end
```

# Newton's Method Example (1)

Solve:

$$x - x^{1/3} - 2 = 0$$

First derivative is

$$f'(x) = 1 - \frac{1}{3}x^{-2/3}$$

The iteration formula is

$$x_{k+1} = x_k - \frac{x_k - x_k^{1/3} - 2}{1 - \frac{1}{3}x_k^{-2/3}}$$

## Newton's Method Example (2)

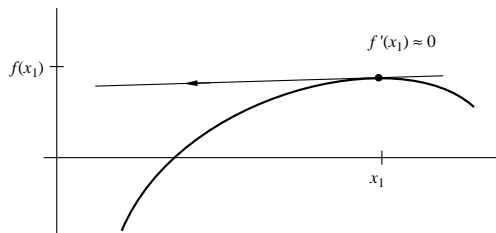
$$x_{k+1} = x_k - \frac{x_k - x_k^{1/3} - 2}{1 - \frac{1}{3}x_k^{-2/3}}$$

$k$	$x_k$	$f'(x_k)$	$f(x)$
0	3	0.83975005	-0.44224957
1	3.52664429	0.85612976	0.00450679
2	3.52138015	0.85598641	$3.771 \times 10^{-7}$
3	3.52137971	0.85598640	$2.664 \times 10^{-15}$
4	3.52137971	0.85598640	0.0

### Conclusion

- Newton's method converges *much* more quickly than bisection
- Newton's method requires an analytical formula for  $f'(x)$
- The algorithm is simple as long as  $f'(x)$  is available.
- Iterations are not guaranteed to stay inside an ordinal bracket.

# Divergence of Newton's Method



Since

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

the new guess,  $x_{k+1}$ , will be far from the old guess whenever  $f'(x_k) \approx 0$