

Lecture 11

Rootfinding

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Some Motivation

- Not all problems are as easy to solve as linear systems of equations
- Even the 5th-order polynomial has no closed form solution
- Many problems fall into this category:
 - Where should wireless access points be placed? How many?
 - Where do two curved surfaces intersect?
 - What is the subsurface geology in the Los Angeles basin?
- How do we solve these problems? Iterate, getting closer (we hope!) each time.



What Next?

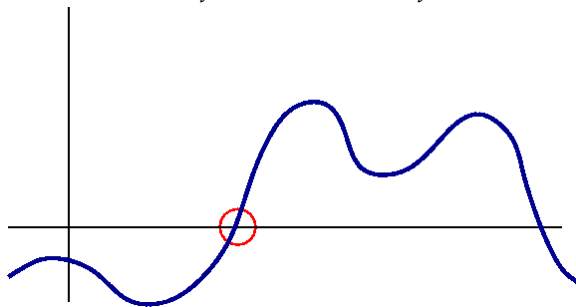
We need to study some iterations.

- iteratively finding a root to an equation
- iteratively finding the solution to an algebraic system
- iteratively finding solutions to Ordinary Differential Equations (ODEs)
- ...



Root Finding

Given a function $f(x)$, find x so that $f(x) = 0$



Rootfinding

Goals:

- Find roots to equations
 - Compare usability of different methods
 - Compare convergence properties of different methods
- 1 bracketing methods
 - 2 Bisection Method
 - 3 Newton's Method
 - 4 Secant Method
 - 5 (opt) fixed point iterations
 - 6 (opt) special Case: Roots of Polynomials

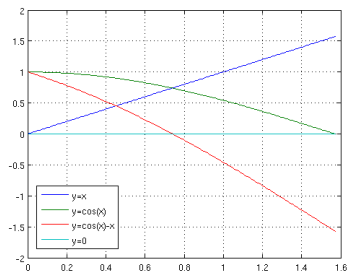


Roots of $f(x)$

- Any single valued function can be written as $f(x) = 0$

Example

- Find x so that $\cos(x) = x$
- That is, find where $f(x) = \cos(x) - x = 0$



Analyze your Application

- Is the function complicated to evaluate?
 - lots of expressions?
 - singularities?
 - simplify? polynomial?
- How accurate does our root need to be?
- How fast/robust should our method be?

!

From this, you can pick the right method...



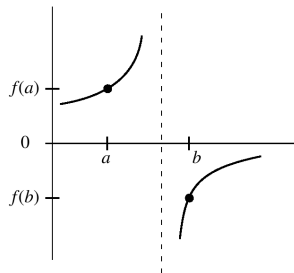
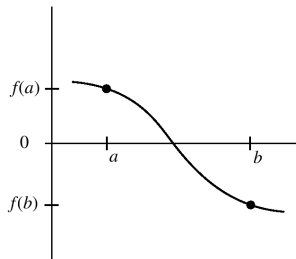
Basic Root Finding Strategy

- 1 Plot the function
 - ▶ Get an initial guess
 - ▶ Identify problematic parts
- 2 Start with the initial guess and iterate



Bracket Basics

- A root x is *bracketed* on $[a, b]$ if $f(a)$ and $f(b)$ have opposite sign.
- Changing signs does not guarantee bracketed, however: singularity



- Bracketing helps get an initial guess

Bracket Algorithm

Listing 1: Bracket Algorithm

```
1 given:  $f(x)$ ,  $x_{min}$ ,  $x_{max}$ ,  $n$ 
2
3  $dx = (x_{max} - x_{min})/n$ 
4  $x_{left} = x_{min}$ 
5  $i = 0$ 
6
7 while  $i < n$ 
8    $i = i + 1$ 
9    $x_{right} = x_{left} + dx$ 
10  if  $f(x)$  changes sign in  $[x_{left}, x_{right}]$ 
11    save  $[x_{left}, x_{right}]$  as an interval with a root
12  end
13   $x_{left} = x_{right}$ 
```



Testing Sign

$$f(a) \times f(b) < 0$$

Should we use?

```
fa = myfunc(a);  
fb = myfunc(b);
```

```
if(fa*fb<0)  
    (save)  
end
```



Better Sign Test

!

Nope. Underflow...

sign()

Use Python's sign

```
fa = myfunc(a);  
fb = myfunc(b);  
import numpy as np  
  
if(np.sign(fa) != np.sign(fb))  
    (save)  
end
```



Moving forward...

Bracketing is fine. But we need to find the actual root:

- Bisection
- Newton's Method
- Secant Method
- Fixed Point Iteration

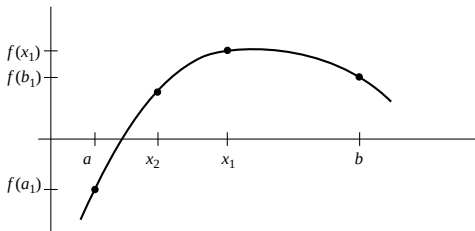
Process:

- 1 use bracket.py to get a visual and brackets
- 2 search brackets with these methods



Bisection

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



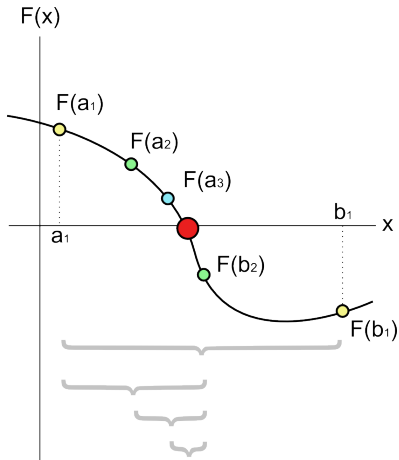
Bisection (2)

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

Let δ_n be the size of the bracketing interval at the n^{th} stage of bisection. Then

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

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

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

$$\vdots$$

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

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

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



Analysis of Bisection

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

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

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

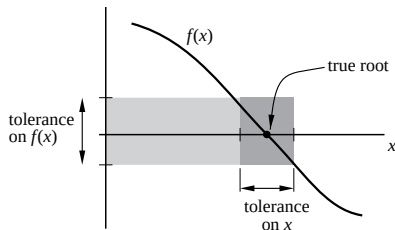
- Check how close $f(x)$ is to zero at the current guess.

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

- Which one you use depends on the problem being solved



Convergence Criteria on x



x_k = current guess at the root

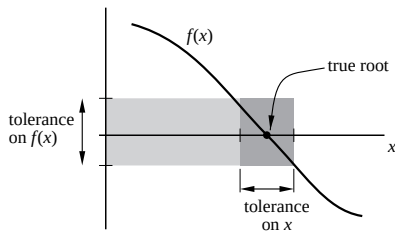
x_{k-1} = previous guess at the root

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

Relative tolerance: $\left| \frac{x_k - x_{k-1}}{b - a} \right| < \hat{\delta}_x$



Convergence Criteria on $f(x)$



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

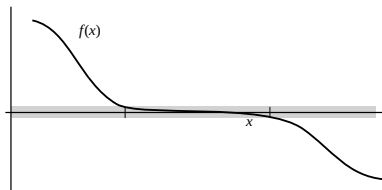
Relative tolerance:

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

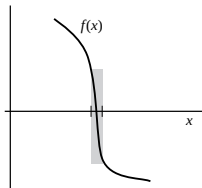
where a_0 and b_0 are the original brackets

Convergence Criteria Compared

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



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



Relationship Between Criteria

- How are the criteria on x and $f(x)$ related? Consider the ratio of the two criteria

$$\frac{f(x_b) - f(x_a)}{x_b - x_a}$$

- The limit of this as x_a and x_b converge to the exact answer x^* is just $f'(x^*)$.
- We can thus expect (this is not yet a proof) that

$$|f(x_b) - f(x_a)| \approx |f'(x^*)| |x_b - x_a|$$

as x_a and x_b approach the solution x^* .



Convergence rate of a root finding iteration

- Let $e_n = x^* - x_n$ be the error.
- In general, a sequence is said to converge with rate r if

$$\lim_{k \rightarrow \infty} \frac{|e_{n+1}|}{|e_n|^r} = C$$

Special Cases:

- If $r = 1$ and $C < 1$, then the rate is *linear*
- If $r = 2$ and $C > 0$, then the rate is *quadratic*
- If $r = 3$ and $C > 0$, then the rate is *cubic*

Example

Convergence Rate

- 1 $10^{-2}, 10^{-3}, 10^{-4}, 10^{-5} \dots$
- 2 $10^{-2}, 10^{-4}, 10^{-6}, 10^{-8} \dots$
- 3 $10^{-2}, 10^{-4}, 10^{-8}, 10^{-16} \dots$
- 4 $10^{-2}, 10^{-6}, 10^{-18}, \dots$



Example

Convergence Rate

- 1 $10^{-2}, 10^{-3}, 10^{-4}, 10^{-5} \dots$ (linear with $C = 10^{-1}$)
- 2 $10^{-2}, 10^{-4}, 10^{-6}, 10^{-8} \dots$ (linear with $C = 10^{-2}$)
- 3 $10^{-2}, 10^{-4}, 10^{-8}, 10^{-16} \dots$ (quadratic)
- 4 $10^{-2}, 10^{-6}, 10^{-18}, \dots$ (cubic)

- Linear: Adds one digit of accuracy at each step
- Quadratic: Doubles the number of digits at each step



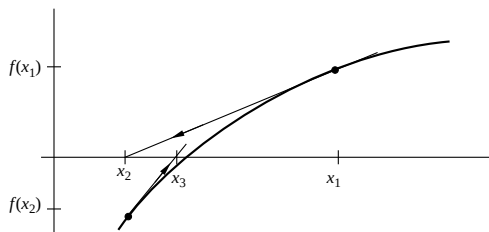
Performing Division

- Ever wondered how a computer process performs division?
- “Long” division requires lookup, subtraction, shifts
- Generates one digit and a time. Can we do better?

To answer this, we need to look at faster methods than bisection



Newton's Method



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

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 2nd 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

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

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

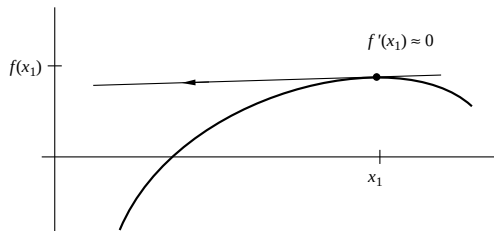
$$x_{k+1} = x_k - \frac{x_k - x_k^{1/3} - 2}{1 - \frac{1}{3}x_k^{-2/3}}$$

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	3.771×10^{-7}
3	3.52137971	0.85598640	2.664×10^{-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

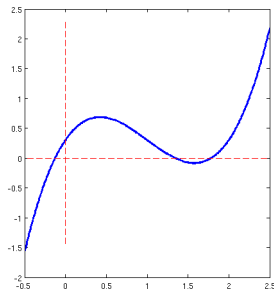
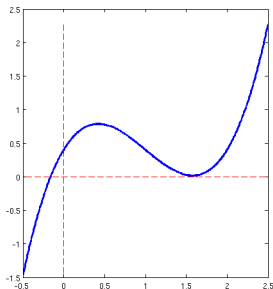
$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}$$

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



Divergence of Newton's Method

Can you guess?



<http://www.math.umn.edu/~garrett/qy/Newton.html>



Newton's Method: Convergence

Recall

Convergence of a method is said to be of order r if there is a constant $C > 0$ such that

$$\lim_{k \rightarrow \infty} \frac{|e_{k+1}|}{|e_k|^r} = C$$

Newton's method is of order 2 (quadratic) when $f'(x_*) \neq 0$. For ξ between x_k

and x_*

$$f(x_*) = f(x_k) + (x_* - x_k)f'(x_k) + \frac{1}{2}(x_* - x_k)^2 f''(\xi) = 0$$

So

$$\frac{f(x_k)}{f'(x_k)} + x_* - x_k + \frac{(x_* - x_k)^2 f''(\xi)}{2 f'(x_k)} = 0$$

Then

$$x_* - x_{k+1} + \frac{1}{2}(x_* - x_k)^2 \frac{f''(\xi)}{f'(x_k)} = 0$$

Thus

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

Reciprocal Approximation

- Consider the task of computing $1/q$ for some q without using division.
- We can write this as: find the root x of $f(x) = 1/(xq) - 1 = 0$.
- What is Newton's Method for this?
- $f'(x) = -1/(x^2q)$. Thus

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

or

$$x_{n+1} = x_n - \frac{1/(x_n q) - 1}{-1/(x_n^2 q)}$$



Reciprocal Approximation

- Consider the task of computing $1/q$ for some q without using division.
- We can write this as: find the root x of $f(x) = 1/(xq) - 1 = 0$.
- What is Newton's Method for this?
- $f'(x) = -1/(x^2q)$. Thus

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

or

$$x_{n+1} = x_n - \frac{1/(x_n q) - 1}{-1/(x_n^2 q)} \frac{x_n^2 q}{x_n^2 q}$$

$$x_{n+1} = x_n + x_n - x_n^2 q = 2x_n - x_n^2 q$$



Example: Compute $1/3$

- Find the bracket:

- $1/2 > 1/3 > 1/4$

① $x_0 = 1/4$

② $x_1 = 2x_0 - x_0^2q = 1/2 - 3/16 = 5/16$

③ $x_2 = 2 \times 5/2^4 - 3 \times 25/2^8 = (160 - 75)/2^8 = 85/2^8$

④ $x_3 = 2 \times 85/2^8 - 3 \times 85^2/2^{16} = 21845/2^{16}$

In 3 steps, computed 16 bits in $1/3$

