

# Ch 2.3, part 2: details/variants on Newton's Method

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Newton's method has been around awhile, so there are many variants / extensions etc.

Outline:

- History/example
- Convergence revisited (multiple roots)
- Modified Newton
- Deflation
- Practical Newton's method
- Secant Method
- Pros/cons of Newton

## Ex. Babylonian Algo

2000+ years ago, Archimedes claimed

How did he find it? Not sure, but

maybe via the Babylonian Algorithm, aka Heron's method.

Algo: Find  $x = \sqrt{a}$

$$\text{Iterate } x_{k+1} = \frac{1}{2} \left( x_k + \frac{a}{x_k} \right)$$

It turns out this is Newton's method

applied to  $f(x) = x^2 - a$  (i.e.  $f'(x) = 2x$ )

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

$$= x_k - \frac{(x_k^2 - a)}{2x_k}$$

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

1.73202

$$\frac{265}{153}$$

$$< \sqrt{3} < \frac{1351}{780} . \text{Not bad!}$$

1.73205



Square w/ area  $a$



rectangle with same area  $a$   
if one side is  $x$ , other  
side must be  $a/x$ .

If  $x > \sqrt{a}$  then  $a/x < \sqrt{a}$

and vice-versa

(one side too short  $\Rightarrow$  other side too long)

So we have under and over approximations...  
so average these

## Convergence, revisited

Recall our result from last time:

Thm: (combining Thm 2.6 and Thm 2.9)

| Let  $f \in C^2([a,b])$  have a root  $p \in (a,b)$  with multiplicity 1 (i.e.  $f'(p) \neq 0$ )

| then if initialized sufficiently close to  $p$ , Newton's method will converge to  $p$ .

| Furthermore, if additionally  $|g''(x)|$  is bounded on some open interval around  $p$ ,

| then the convergence rate is quadratic.  $\rightarrow g(x) := x - \frac{f(x)}{f'(x)}$

OK, but what if  $p$  isn't a simple root?

aka multiplicity 1

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Def Recall, a root  $p$  of  $f$  is multiplicity  $m$  if the 1<sup>st</sup>  $m-1$  derivatives of  $f$  are 0 at  $p$

$$\text{i.e., } f(p) = f'(p) = \dots = f^{(m-1)}(p) = 0 \text{ and } f^{(m)}(p) \neq 0$$

Equivlently, if  $f(x) = (x-p)^m g(x)$  and  $g(p) \neq 0$

Ex:  $f(x) = x(x-1)(x-2)$  has a simple root at  $x=0$  (and at  $x=1, x=2$ )  $m=1$

$f(x) = x^2(x-1)$  has a double root at  $x=0$   $m=2$

$f(x) = x^3$  has a triple root at  $x=0$   $m=3$

Our theorem doesn't apply. Often we still get convergence but just not at a quadratic rate

Ex  $f(x) = x^2$ ,  $x=0$  is a double root.  $f'(x) = 2x$

$$\text{Newton's method is } x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)} = x_k - \frac{x_k^2}{2x_k} = \frac{1}{2}x_k$$

$$\text{i.e., } x_{k+1} = \frac{1}{2}x_k,$$

so, e.g.,  $x_0 = 1$ , then we iterate  $1, \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \frac{1}{16}, \dots$

error at step  $k$  is  $\frac{1}{2^k}$ ,  $e_k = \frac{1}{2^k}$ . Is this quadratic?

$$\text{Check: } \lim_{k \rightarrow \infty} \frac{e_{k+1}}{e_k^2} = \lim_{k \rightarrow \infty} \left( \frac{\frac{1}{2^{k+1}}}{\left(\frac{1}{2^k}\right)^2} \right) = \frac{\frac{1}{2^{k+1}}}{\frac{1}{2^{2k}}} = \frac{2^{2k}}{2^{k+1}} = 2^{k-1} = \infty$$

No, not quadratic conv.

In fact, this is linear convergence since it fits the form  $\rho^k$  ( $\rho = \frac{1}{2}$ ) which we're already discussed.

One fix to this multiplicity issue is...

Modified Newton's Method

\* there are many ways to modify it, this is just our book's notation

Let  $p$  be a root of  $f$  w/ multiplicity  $m$  ( $m=1$  is ok, but mostly interested in  $m>1$ )

Define  $h(x) = \frac{f(x)}{f'(x)}$ . Then claim  $p$  is a root of  $h$  also

proof:  $m=1$  then  $f'(p) \neq 0$  so immediately  $h(p) = 0$

Furthermore,  $p$  is a simple root of  $h$

$m>1$  then  $f'(p) = 0$ ,  $\frac{f(p)}{f'(p)} = \frac{0}{0}$  ... use L'Hopital

proof:

Can write  $f(x) = (x-p)^m g(x)$

w/  $g(p) \neq 0$

$$\text{so } \lim_{x \rightarrow p} \frac{f(x)}{f'(x)} = \lim_{x \rightarrow p} \frac{f'(x)}{f''(x)} = \dots = \lim_{x \rightarrow p} \underbrace{\frac{f^{(m-1)}(x)}{f^{(m)}(x)}}_{\neq 0 \text{ at } p} = 0$$

then  $h(x) = \frac{(x-p)^m g(x)}{m(x-p)^{m-1} g(x) + (x-p)^m g'(x)}$

$$= (x-p) \frac{g(x)}{m \cdot g(x) + (x-p) g'(x)} \quad \tilde{g}(p) = \frac{g(p)}{m g(p) + (p-p) g'(p)}$$

$$\text{So } h(x) = (x-p) \tilde{g}(x) \quad = \frac{1}{m} \neq 0$$

$\text{with } \tilde{g}(p) \neq 0 \Rightarrow$  is a simple root

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... back to the point.

$P$  is a simple root of  $h(x) = \frac{f(x)}{f'(x)}$

So run Newton on  $h$  instead of  $f$ .

Simplifying  $h'$ , we get

MODIFIED NEWTON

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

ugly quotient rule stuff

Not a perfect fix:

- DRAWBACKS:
  - ① must compute  $f''$
  - ② subtractive cancellation

Note:

If  $m$  is known,  $x_{n+1} = x_n - m \frac{f(x_n)}{f'(x_n)}$  will work and w/  
quadratic convergence

example of a "stepsize"

### Deflation

A related idea to dealing with roots of multiplicity  $> 1$

Suppose  $f$  has simple roots at  $p_1$  and  $p_2$  (if  $p_1 = p_2$  it's a double root)

If  $|p_1 - p_2|$  is very small, starts to act like a double root, especially  
w/<sub>1</sub> roundoff errors. The condition number of the root-finding problem is large

One practical consequence:

Suppose we find  $p_1$ . How to find  $p_2$ ? We have to start sufficiently  
close to it, which is hard since we don't know where it is! We might  
get "sucked into" the  $p_1$  root.

... and a fix: deflation

Define  $h(x) = \frac{f(x)}{x - p_1}$  so  $h$  doesn't have a root at  $p_1$ ,  
but still has a root at  $p_2$

(this is also the name for a broader class of techniques, e.g. in eigenvalue problems)

### Practical Newton's method, i.e. globalization strategies

We won't go into details

(1) combine w/<sub>1</sub> bracketing or another root-finding method

(our book mentions a special version called False Position / Regular Falsi)

(2) safeguarding / linesearch:

don't let  $x_{n+1}$  go too far

(ex: if we must keep  $x_n \geq 0$ )

or  $x_{n+1} = x_n - \alpha \frac{f(x_n)}{f'(x_n)}$ ,  $\alpha \leq 1$  Need  $\alpha = 1$  for quadratic convergence  
but often take  $\alpha < 1$  for the first few iterations

Don't worry about these,  
just use Matlab / Scipy libraries

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### Secant Method

Idea: Newton's method is  $x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}$

Well, rule of thumb: when a step produces an approximate result, you are free to carry it out approximately.

Since Newton's method was derived via Taylor

Series, ignoring higher-order terms, let's try approximating the derivative:

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{(x+h) - (x)}, \text{ i.e., } f'(x_k) \approx \frac{f(x_k) - f(x_{k-1})}{x_k - x_{k-1}} \text{ if } |x_k - x_{k-1}| \text{ small}$$

So

#### SECANT METHOD

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

See demo: if you call scipy's 1D Newton and don't supply "fprime", it doesn't work as well (presumably it tries finite difference)

What if we want to avoid calculating  $f'(x)$ ? \*

\* See lecture on "Automatic Differentiation" ( $\neq$  symbolic diff.) ex. backprop.

(re-use our old computation)

Only needs 1 function evaluation per step, no derivative needed. Nice!

You can extend in 2 ways:

(1) use more previous points, "inverse interpolation". Not very common

(2) In multi-dimensional problems, there is "more freedom", and

we actually have a whole class of QUASI-NEWTON METHODS.

(names like "Broyden class", "SR1", "BFGS")

often "state-of-the-art"

Very useful!

Also, some extra computational savings that are irrelevant for scalar problems.

### Convergence Analysis of Secant method [possibly skip in lecture]

Recall for Newton  $\lim \frac{|e_{k+1}|}{|e_k|^2} < \infty$  i.e.  $\alpha = 2$

where  $e_k = p - x_k$  is the error.

For the Secant method, assuming  $e_k$  is small,

we can do a Taylor expansion of our Secant method iteration

(tedious but straightforward) to get  $e_{k+1} \approx -\frac{1}{2} \underbrace{\frac{f''(p)}{f'(p)}}_{\text{Some constant}} e_k e_{k-1}$  (\*)

Let's guess/hope that we have  $\alpha$  convergence and can write

( $\forall k$ )  $e_{k+1} = c \cdot e_k^\alpha$  and solve for  $\alpha$   
 like an "ansatz"

then  $e_{k+1} = c \cdot e_k^\alpha = c \cdot (c e_{k-1}^\alpha)^\alpha = \text{const. } e_{k-1}^{\alpha^2}$

and  $e_k e_{k-1} = c e_k^\alpha e_k = c e_{k-1}^{\alpha+1}$  involves  $c$  and  $-\frac{1}{2} \frac{f''(p)}{f'(p)}$

Plugging into (\*) gives  $e_{k-1}^{\alpha^2} = (\text{some const.}) e_{k-1}^{\alpha+1}$

this should be true for all  $e_{k-1}$  (when  $e_{k-1}$  is near 0)

so need  $\text{const.} = 1$

$$\alpha^2 = \alpha + 1, \quad \alpha > 0$$

solution is the Golden Ratio

$$\alpha = \frac{1 + \sqrt{5}}{2} \approx 1.618$$

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So... Newton's method has rate  $\alpha = 2$

Secant method has rate  $\alpha = 1.62$

... and in fact if you say that Newton's method takes twice as much work as the secant method (since you must evaluate  $f(x_n)$  and  $f'(x_n)$ ) then we can do 2 iterations of the secant method and count the rate as  $(1.62)^2 = 2.62$

(or, keep as 1.62 but call Newton's rate  $\sqrt{2} = 1.41$ )

... meaning the secant method is better than Newton, in this sense.

### Summary of Pros/Cons

#### pros

- Newton's method:
- + No need for bracketing interval  $[a, b]$ .
  - + Doesn't need  $f(a) f(b) < 0$  (which excludes  $f(x) = x^2$ )
  - + Very fast convergence eventually (the gold-standard)
  - + Simple
  - + extends to dimensions  $\geq 1$

#### CONS

- $x_0$  must be close to root, and hard to know how close  
If not close enough, may diverge or converge to wrong root
- Slower convergence for multiple roots  $m > 1$   
Some fixes but not perfect
- practical implementations need more information,  
more complicated
- must supply  $f'( )$

#### Secant method

Same pros/cons except no longer need to provide  $f'( )$ , slightly slower convergence rate

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Addendum skip in lecture

Theorems 2.6 and 2.8 in the book are not stated in the most useful way, especially 2.8. Notation: let  $p$  denote a fixed pt.

Book's Thm 2.8:  $g \in C[a,b]$ ,  $g(x) \in [a,b] \forall x \in [a,b]$ ,  $g'$  continuous on  $(a,b)$ , and  $|g'(x)| \leq k \forall x \in (a,b)$  for some  $k < 1$ .

Then unless  $p_0$  is a fixed point, the sequence  $(p_n)$ , defined by  $p_{n+1} = g(p_n)$ , converges only linearly if  $g'(p) \neq 0$ .

(They state it this way since

Thm 2.9 gives quadratic convergence if  $g'(p) = 0$ )

1. First, recall

(1) root-finding,  $f(p) = 0$ ,  
here  $f'(p) = 0$  is BAD

(2) fixed-pt. iter, e.g. Newton

$$g(x) = x - \frac{f(x)}{f'(x)}, \quad g(p) = p$$

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

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

$$\underline{g'(p) = 0 \text{ is GOOD}}$$

We can strengthen Thm 2.8  
by looking at its proof:

Better Thm 2.8 (same assumptions ...)

... then if  $g'(p) \neq 0$ ,  
 $(p_n)$  converges to  $p$  linearly  
at rate  $|g'(p)|$ .