

2 Methods for solving nonlinear eqs $f(x)=0$ numerically

(1) Picard iteration

(1)

$$f(x) = \frac{1}{2}x^2 - x + \frac{1}{4}$$

Solve $f(x) = 0$

$$x^2 - 2x + \frac{1}{2} = 0$$

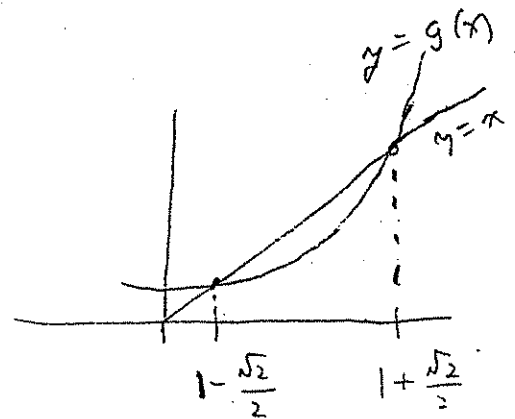
$$x = \frac{2 \pm \sqrt{4-2}}{2} = 1 \pm \frac{\sqrt{2}}{2} \approx .292893219... \text{ or } 1.707106781...$$

Note $f(x) = g(x) - x$

where $g(x) = \frac{1}{2}x^2 + \frac{1}{4}$

To solve $f(x) = 0$

Find x s.t. $x = g(x)$



Picard iteration

initial guess x_0
iteration $x_{m+1} = g(x_m)$

(Def: $\lim_{n \rightarrow \infty} x_n = x^*$ if for all $\epsilon > 0$ there exists an $N > 0$ s.t. $n > N \Rightarrow |x^* - x_n| < \epsilon$.)

If g is continuous then

$$x^* = \lim_{n \rightarrow \infty} x_{m+1} = \lim_{n \rightarrow \infty} g(x_m) = g(x^*)$$

So $x^* = g(x^*)$ and x^* is the solution

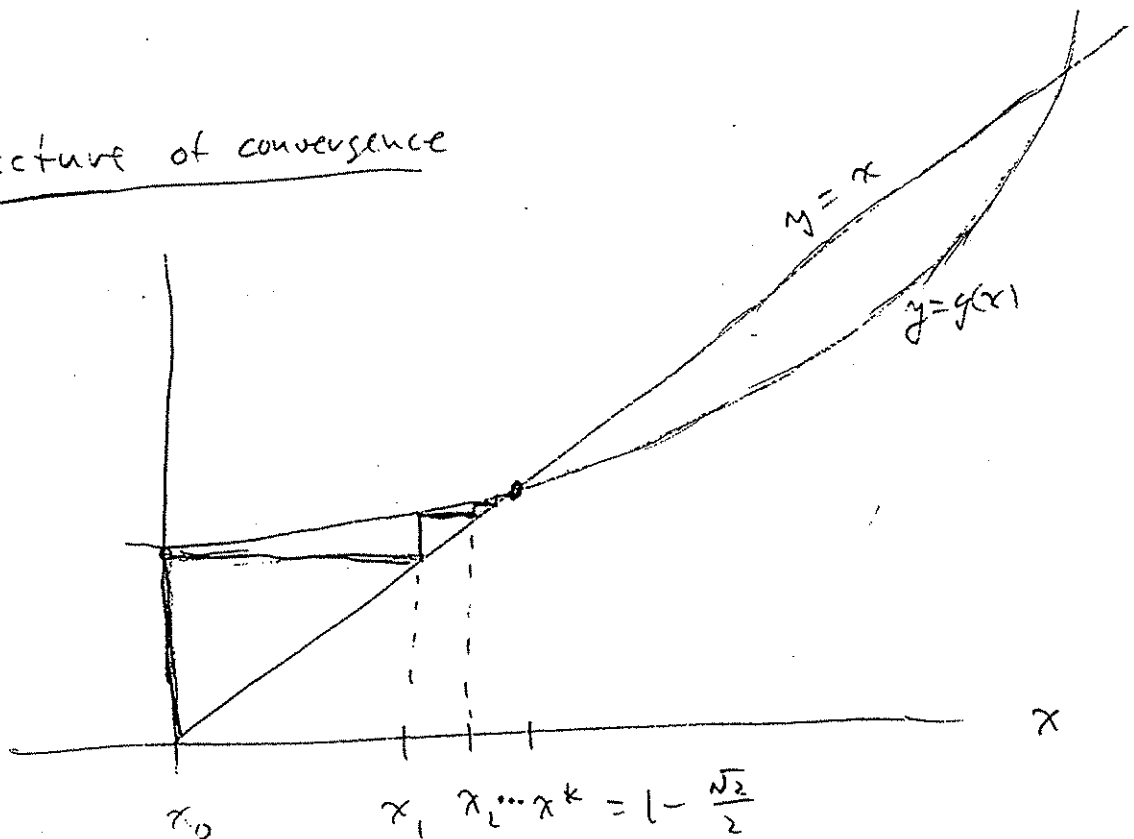
The x_m 's are called the iterates and are said to converge to the solution, x^* .

Picard iteration from initial guess $x_0 = 0$

(2)

n	x_n	$g(x_n) = \frac{1}{2}x_n^2 + \frac{1}{4}$
0	0	.25
1	.25	.28125
2	.28125	.289550781
3	.289550781	.291919827
4	.291919827	.292608593
5	.292608593	.292809894
6	.292809894	.292868817
⋮	⋮	⋮
	x^*	= .292893219

picture of convergence



Convergence analysis for Picard iteration

(3)

Suppose $f(x) = g(x) - x$ where $g'(x)$ exists
and $f(x^*) = 0$, i.e. $x^* = g(x^*)$.

The error at the n^{th} iterate is

$$e_n = x^* - x_n$$

$$= g(x^*) - g(x_{n-1})$$

$$= g'(c_n)(x^* - x_{n-1})$$

$$= g'(c_n) e_{n-1}$$

⋮

$$\approx [g'(x^*)]^n e_0$$

← for some c_n
between x^* and
 x_{n-1} by the
Mean Value Theorem

(linear convergence,
not so fast)

since $g'(c_n) \approx g'(x^*)$ if x_n is near x^* .

That is, if we have a close enough
initial guess the error will improve
by a factor of about $g'(x^*)$ at each
step. So if $g'(x^*) \approx \frac{1}{10}$ we would
gain a decimal place each step

Note for g in our example

$$g'(x^*) = x^* = .292 \dots$$

(2) Newton's method for $f(x) = \frac{1}{2}x^2 - x + \frac{1}{4}$

(4)

$$f'(x) = x - 1$$

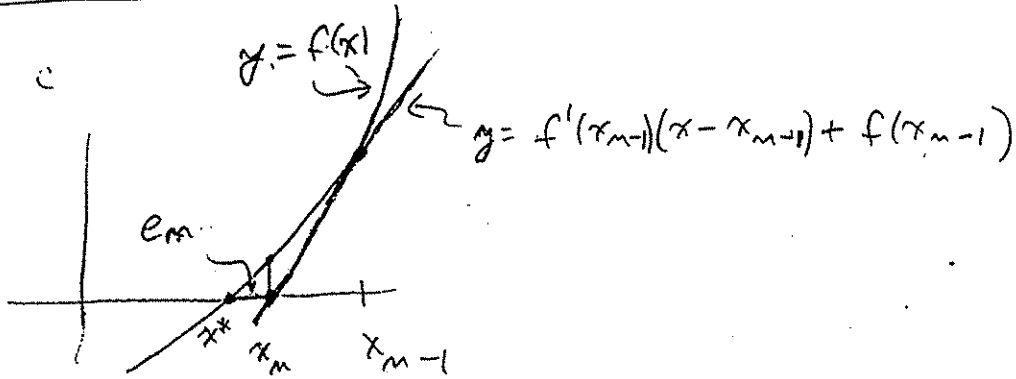
$$\begin{aligned}x_{m+1} &= x_m - \frac{f(x_m)}{f'(x_m)} = x_m - \frac{\frac{1}{2}x_m^2 - x_m + \frac{1}{4}}{x_m - 1} \\ &= \frac{x_m^2 - x_m - \frac{1}{2}x_m^2 + x_m - \frac{1}{4}}{x_m - 1} \\ &= \frac{\frac{1}{2}x_m^2 - \frac{1}{4}}{x_m - 1} = \frac{x_m^2 - \frac{1}{2}}{2(x_m - 1)}\end{aligned}$$

n	x_n	$x_{n+1} = \frac{x_n^2 - \frac{1}{2}}{2(x_n - 1)}$
0	0	$\frac{1}{4} = .25$
1	.25	$\frac{7}{24} = .291666667$
2	.291666667	$.292892157$
3	.292892157	$.292893219$

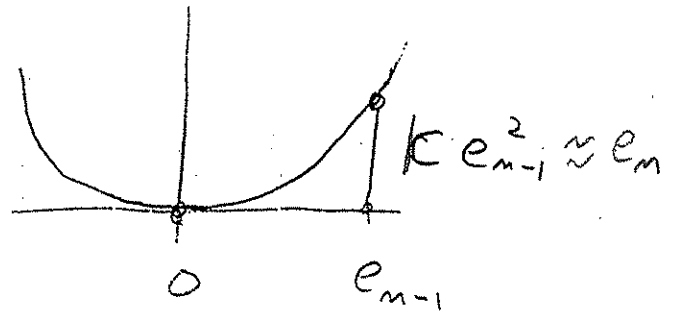
Note how much faster Newton's method converges.

Convergence analysis for Newton's method

(5)



This is like a parabola (more in Calculus II)



error at nth iteration =

$$\begin{aligned}
 e_m &= x^* - x_m \\
 &= x^* - x_{m-1} + f'(x_{m-1})^{-1} f(x_{m-1}) \\
 &= f'(x_{m-1})^{-1} [f'(x_{m-1})(x^* - x_{m-1}) + f(x_{m-1})] \\
 &= f'(x_{m-1})^{-1} [C \cdot (x^* - x_{m-1})^2] \quad (\text{Calc II}) \\
 &= \underbrace{f'(x_{m-1})^{-1} C}_{K} \cdot e_{m-1}^2 \quad \left(C = -\frac{1}{2} f''(\eta) \right. \\
 &= K e_{m-1}^2 \quad \left. \begin{array}{l} \text{for some } \eta \text{ between} \\ x^* \text{ and } x_{m-1} \end{array} \right)
 \end{aligned}$$

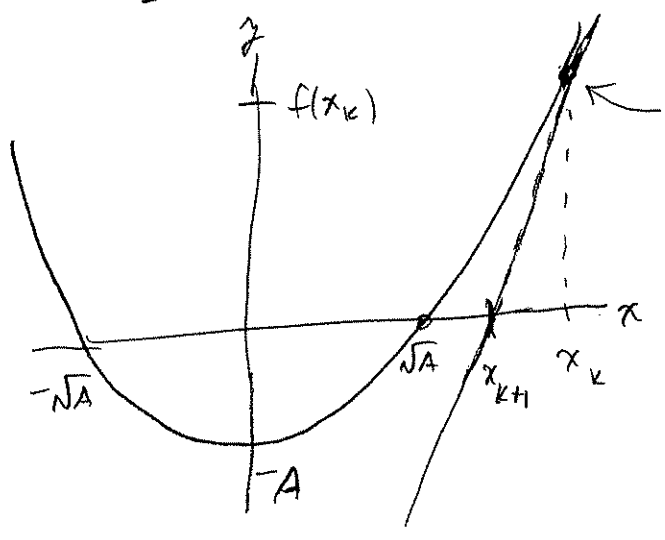
Suppose $K \approx 1$ then if

$$\left. \begin{array}{l}
 e_1 \approx 10^{-1} \\
 e_2 \approx 10^{-2} \\
 e_3 \approx 10^{-4} \\
 e_4 \approx 10^{-8}
 \end{array} \right\} \begin{array}{l} \text{quadratic} \\ \text{convergence} \\ \text{very fast} \end{array}$$

i.e. $e_m = O(e_{m-1}^2)$

example Newton's method for finding \sqrt{A} ($A > 0$).

Solve $f(x) = 0$ where $f(x) = x^2 - A$



tangent line at $(x_k, f(x_k))$

$$y = f(x_k) + f'(x_k)(x - x_k)$$

intersection of x -axis at x_{k+1}

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

⇓

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

$$= x_k - \frac{x_k^2 - A}{2x_k}$$

$$= \frac{1}{2} \left(x_k + \frac{A}{x_k} \right)$$

(Note: $f'(x)$ must exist.)

Take e.g. $A = 2$, since $f(0) = -2$ and $f(2) = 2$ and f is continuous, by the Intermediate Value Theorem f must have a root in $[0, 2]$

we just need an initial guess, e.g. $x_0 = 2$, to get started

simple
MATLAB
example



This version is for educational classroom use only. ❄

To get started, type one of these: helpwin, helpdesk, or demo.
For product information, type tour or visit www.mathworks.com.

```
> format long
> x=2;
> x=(x + 2/x)/2
```

← initial guess

```
x =
  1.500000000000000
```

← 1 digit correct
error = $O(10^{-1})$

```
> x=(x + 2/x)/2
```

```
x =
  1.416666666666667
```

← 3 digits correct
error = $O(10^{-3})$

```
> x=(x + 2/x)/2
```

```
x =
  1.41421568627451
```

← 6 digits correct
error = $O(10^{-6})$

```
> x=(x + 2/x)/2
```

```
x =
  1.41421356237469
```

← 12 digits correct
error = $O(10^{-12})$

```
> x=(x + 2/x)/2
```

```
x =
  1.41421356237309
```

← almost all available
digits correct

```
> x=(x + 2/x)/2
```

```
x =
  1.41421356237309
```

```
> sqrt(2)
```

```
ans =
  1.41421356237310
```

Simple MATLAB
example: Newton's
method for finding
 $f(x) = x^2 - 2 = 0$
($x = \sqrt{2}$)

What can go wrong with Newton's method?

(11)

i) $f'(x_*) = 0$ or $f'(x_0) \approx 0$ for $f(x_*) = 0$.

If $f'(x_*) = 0$ and f is sufficiently differentiable then

$$f(x) = \underbrace{f(x_*)}_0 + \underbrace{f'(x_*)}_0 (x - x_*) + \frac{f''(x_*)}{2!} (x - x_*)^2 + O(x - x_*)^3$$

i.e. $f(x) = \frac{f''(x_*)}{2} (x - x_*)^2 + O(x - x_*)^3$ for x near x_* .

and $f'(x) = f''(x_*) (x - x_*) + O(x - x_*)^2$ " " "

Assuming $f''(x_*) \neq 0$ Newton's method for

$x = x_k$ near x_* gives

$$\begin{aligned} x_{k+1} &= x_k - \frac{f(x_k)}{f'(x_k)} \\ &= x_k - \frac{\frac{f''(x_*)}{2} (x_k - x_*)^2 + O(x_k - x_*)^3}{f''(x_*) (x_k - x_*) + O(x_k - x_*)^2} \\ &= x_k - \frac{1}{2} (x_k - x_*) + O(x_k - x_*)^2 \end{aligned}$$

linear convergence with convergence ratio $\frac{1}{2}$

\therefore error at $k+1$ st step

$$\begin{aligned} = e_{k+1} = x_{k+1} - x_* &= \frac{1}{2} (x_k - x_*) + O(x_k - x_*)^2 \\ &\approx \frac{1}{2} e_k = \frac{1}{2} \text{ error at } k\text{th step} \end{aligned}$$

e.g. $f(x) = x^2$, $f(x_*) = 0$ for $x_* = 0$

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

$$\text{so } x_{k+1} - x_* = \frac{1}{2} (x_k - x_*)$$


```

» format long
» x=.1;
» x=(x + .000001/x)/2

x =
    0.050005000000000

» x=(x + .000001/x)/2

x =
    0.02501249900010

» x=(x + .000001/x)/2

x =
    0.01252623950585

» x=(x + .000001/x)/2

x =
    0.00630303596239

» x=(x + .000001/x)/2

x =
    0.00323084483305

» x=(x + .000001/x)/2

x =
    0.00177018069983

» x=(x + .000001/x)/2

x =
    0.00116754738950

» x=(x + .000001/x)/2

x =
    0.00101202183654

» x=(x + .000001/x)/2

x =
    0.00100007140387

» x=(x + .000001/x)/2

x =
    0.00100000000255

» x=(x + .000001/x)/2

x =
    0.00100000000000

```

Newton's method for

$$f(x) = x^2 - 10^{-6} = 0$$

$$\left(\begin{array}{l} f(x_*) = 0 \text{ for } x_* = 10^{-3} = .001 \\ \text{and } f'(x_*) = 2x_* = .002 \approx 0 \end{array} \right)$$

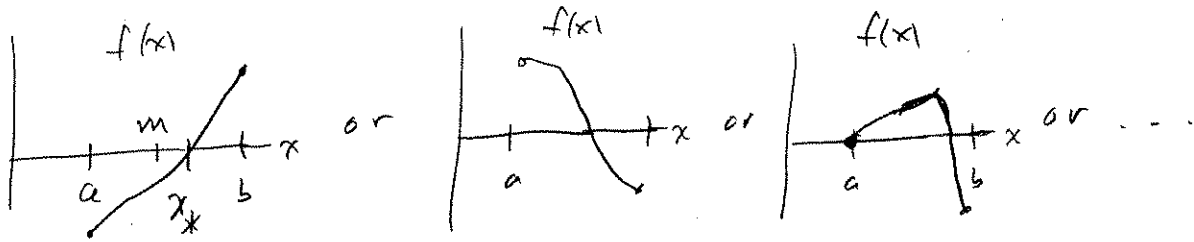
initial convergence linear
with factor $\approx \frac{1}{2}$

See Theorem 8 p. 268
in van Loan's book

final convergence quadratic

Bisection method (see 8.1.2)

Suppose $f(x)$ continuous in $[a, b]$ and $f(a)f(b) \leq 0$,
Then $f(x) = 0$ for some $x \in [a, b]$.



Let $m = \frac{a+b}{2}$. Then either $f(a)f(m) \leq 0$ or $f(m)f(b) \leq 0$.
So there is a root in $[a, m]$ or $[m, b]$

Letting $a_0 = a, b_0 = b,$

$$a_{k+1} = a_k, b_{k+1} = \frac{a_k + b_k}{2} \text{ if } f(a_k)f\left(\frac{a_k + b_k}{2}\right) \leq 0$$

$$\text{and } a_{k+1} = \frac{a_k + b_k}{2}, b_{k+1} = b_k \text{ otherwise}$$

we setting $x_k = \frac{a_k + b_k}{2}$ we have

$$|x_{k+1} - x_k| \leq \frac{1}{2} |x_k - x_k| \leq \frac{1}{2^{k+2}} |a - b|$$

linear convergence with factor $\frac{1}{2}$

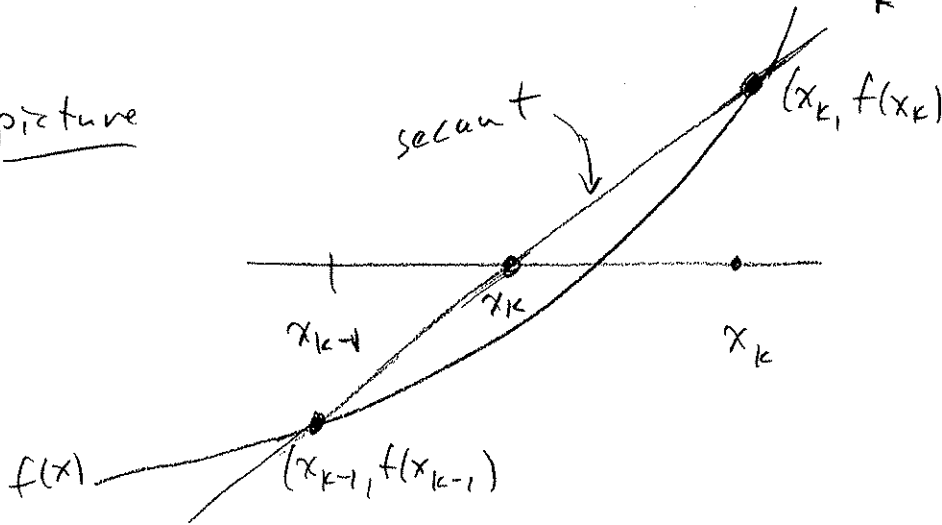
8.1.5 Avoiding Derivatives

(13)

Secant method \approx Newton's method with

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

picture



eq. of secant: $y = f(x_k) + \frac{f(x_k) - f(x_{k-1})}{x_k - x_{k-1}} (x - x_k)$

$y=0$ for $x = \boxed{x_{k+1} = x_k - \frac{f(x_k)(x_k - x_{k-1})}{f(x_k) - f(x_{k-1})}}$

requires 2 starting values x_0, x_1
and 1 evaluation of $f(x_k)$ per step.

CBS convergence is "superlinear"

$$|x_k - x_*| \leq C |x_{k-1} - x_*|^r$$

where $r = \frac{1+\sqrt{5}}{2} \approx 1.6$ $\therefore \underset{\uparrow}{1} \leq r \leq \underset{\uparrow}{2}$
linear conv. quadratic conv.