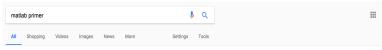
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Math128A algorithms vs. General Purpose algorithms

For any vector $\mathbf{x} \in \mathbf{R}^n$, compute its norm

$$\|\mathbf{x}\|_2 = (x_1^2 + x_2^2 + \dots + x_n^2)^{\frac{1}{2}} = \left(\sum_{k=1}^n x_k^2\right)^{\frac{1}{2}}.$$

- ▶ INPUT: n, x_1, \dots, x_n .
- OUTPUT: Norm.
- **▶ Step 1**: Set **SUM** = 0.
- ▶ **Step 2**: For $k = 1, \dots, n$ set **SUM** = **SUM** + $x_k * x_k$.
- ▶ **Step 3**: Set **Norm** = $\sqrt{\text{SUM}}$.
- ► **Step 4**: Output **Norm**. STOP.

Math128A algorithms vs. General Purpose algorithms

```
|>> n=10;
>> x = (1:n)';
>> sum = 0;
>> for k=1:n
sum = sum + x(k)*x(k);
end
>> nrm = sqrt(sum);
>> disp([nrm, abs(nrm-sqrt(n*(n+1)*(2*n+1)/6))])
  1.9621e+01 0.0000e+00
```

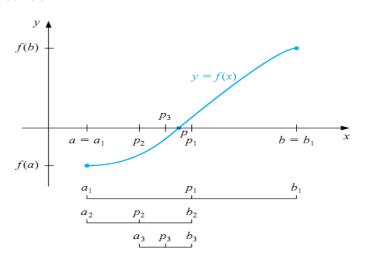
Math128A algorithms vs. General Purpose algorithms

```
>> n=10;
>> x = 1e200*(1:n)';
>> sum = 0;
>> for k=1:n
sum = sum + x(k)*x(k);
end
>> nrm = sqrt(sum);
>> disp([nrm, abs(norm(x)-1e200*sqrt(n*(n+1)*(2*n+1)/6))])
          Inf 0.0000e+00
```

Bisection Method

- ▶ **Given** continuous function f(x) on the interval [a, b] with $f(a) \cdot f(b) < 0$, there must be a root in (a, b).
- ▶ To find a root: set $[a_1, b_1] = [a, b]$.
- set $p_1 = \frac{a_1 + b_1}{2}$ and compute $f(p_1)$.
 - if $f(p_1) = 0$, then quit with root p_1 (MUST BE VERY LUCKY.)
 - if $f(a_1) \cdot f(p_1) < 0$, then set $[a_2, b_2] = [a_1, p_1]$,
 - otherwise $(f(p_1) \cdot f(b_1) < 0)$ set $[a_2, b_2] = [p_1, b_1]$,
- repeat with $p_2 = \frac{a_2 + b_2}{2}$.

Bisection Method



Naive Bisection Method

```
% Bisection Method
%Input: f(x) continuous on [a, b]
       f(a) * f(b) < 0
%Output: p in (a, b) so f(p) = 0.
fa = f(a);
fb = f(b);
repeat
   c = (a+b)/2;
   fc = f(c);
if (fc ==0)
      p = c;
      return:
   end
   if (fc * fa < 0)
      b = c;
   else
      a = c:
   end
end
```

function [x, out] = bisect(FunFcnIn, Intv, params) TOL = params.tol; NO = params.MaxIt: [FunFcn,msq] = fcnchk(FunFcnIn,0); if ~isempty(msg) error('InvalidFUN'.msq); end а = Intv.a: = Intv.b; b fa = sign(FunFcn(a)); fb = sign(FunFcn(b)); if (fa*fb >0) error('Initial Interval may not contain root'.msg): end if a>=b error('a must be smaller than b', msq); end It = 0: out.x = [a:b]:out.f =[FunFcn(a);FunFcn(b)]; while (It <= NO) c = (a+b)/2; out.x = [out.x;c]; out.f =[out.f:FunFcn(c)]: fc = sign(FunFcn(c)); if (fc ==0) x = c;out.fla = 0: out.it = It: return: end if (fc * fa < 0)b = c: else a = c: end if $(abs(b-a) \le TOL)$ x = (a+b)/2: out.flq = 0; out.it = It: return: end It = It + 1: end out.fla =1: out.it = NO: x = (a+b)/2:

Theorem 2.1 Suppose that $f \in C[a,b]$ and $f(a) \cdot f(b) < 0$. The Bisection method generates a sequence $\{p_n\}_{n=1}^{\infty}$ approximating a zero p of f with

$$|p_n-p| \le \frac{b-a}{2^n}$$
, when $n \ge 1$.

Proof of Thm 2.1

Assume that $f(p_n) \neq 0$ for all n.

By construction

$$a_1 \leq a_2 \leq \cdots \leq a_n \leq \cdots \leq \cdots \leq b_n \leq \cdots \leq b_2 \leq b_1$$
.

Thus sequences $\{a_n\}$ and $\{b_n\}$ monotonically converge to limits $a_{\infty} \leq b_{\infty}$, respectively.

- ▶ Since $f(a_n) \cdot f(b_n) < 0$ for all n, it follows that $f(a_\infty) \cdot f(b_\infty) \le 0$, and thus a root $p \in [a_\infty, b_\infty] \subset [a_n, b_n]$ exists.
- ▶ Since $p_n = \frac{a_n + b_n}{2}$, it follows that $|p_n p| \leq \frac{b_n a_n}{2}$.
- By construction

$$b_n-a_n=\frac{b_{n-1}-a_{n-1}}{2}=\frac{b_{n-2}-a_{n-2}}{2^2}=\cdots=\frac{b_1-a_1}{2^{n-1}}=\frac{b-a}{2^{n-1}}.$$

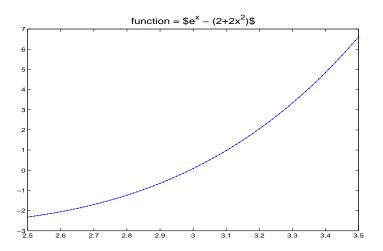
Put together,

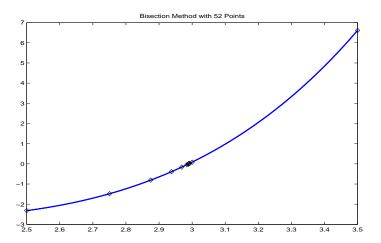
$$|p_n-p|\leq \frac{b-a}{2^n}.$$

▶ In fact, $a_{\infty} = b_{\infty} = p$.



Example Function with Root





Fixed Point Iteration

The number p is a **fixed point** for a given function g if g(p) = p.

Given a root-finding problem f(p) = 0, we can define functions g with a fixed point at
p in a number of ways, for example, as

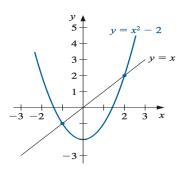
$$g(x) = x - f(x)$$
 or as $g(x) = x + 3f(x)$.

Conversely, if the function g has a fixed point at p, then the function defined by

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

has a zero at p.

Fixed Point Example



Fixed Point Theorem (I)

Theorem 2.3

- (i) If $g \in C[a, b]$ and $g(x) \in [a, b]$ for all $x \in [a, b]$, then g has at least one fixed point in [a, b].
- (ii) If, in addition, g'(x) exists on (a, b) and a positive constant k < 1 exists with

```
|g'(x)| \le k, for all x \in (a, b),
```

then there is exactly one fixed point in [a, b]. (See Figure 2.4.)

Proof of Thm 2.3

- ▶ If g(a) = a or g(b) = b, then g has a fixed point at an endpoint.
- ▶ Otherwise, g(a) > a and g(b) < b. The function h(x) = g(x) x is continuous on [a, b], with

$$h(a) = g(a) - a > 0$$
 and $h(b) = g(b) - b < 0$.

- ▶ This implies that there exists $p \in (a, b)$, h(p) = 0.
- g(p) p = 0, or p = g(p).

If $|g'(x)| \le k < 1$ for all x in (a, b), and p and q are two distinct fixed points in [a, b]. Then a number ξ exists

$$\frac{g(p)-g(q)}{p-q}=g'(\xi)<1.$$

So

$$1 = \frac{p-q}{p-q} = \frac{g(p)-g(q)}{p-q} = g'(\xi) < 1.$$

Fixed Point Iteration

Given initial approximation p_0 , define Fixed Point Iteration

$$p_n = g(p_{n-1}), \quad n = 1, 2, \cdots,$$

If iteration converges to p, then

$$p = \lim_{n \to \infty} p_n = \lim_{n \to \infty} g(p_{n-1}) = g(p).$$

Fixed Point Theorem (II)

Theorem 2.4 (Fixed-Point Theorem)

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' exists on (a, b) and that a constant 0 < k < 1 exists with

$$|g'(x)| \le k$$
, for all $x \in (a, b)$.

Then for any number p_0 in [a, b], the sequence defined by

$$p_n = g(p_{n-1}), \quad n \ge 1,$$

converges to the unique fixed point p in [a, b].



Fixed Point Example $x - x^2 = 0$: no convergence

$$g(x) = x^2 \in [1, \infty]$$
 for $x \in [1, \infty]$, $|g'(x)|$ unbounded in $[1, \infty]$.

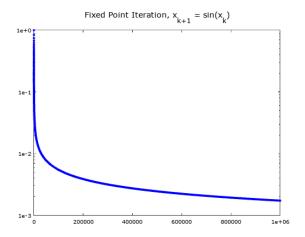
```
>> n = 200;
>> x = zeros(n,1);
>> x(1) = 2;
>> for k=2:n
x(k) = x(k-1)*x(k-1);
end
>> disp([x(n)])
Inf
```

Fixed Point Example $x - \sin(x) = 0$: slow convergence

```
g(x) = \sin(x) \in [-1, 1] for x \in [-1, 1],
 |g'(x)| < 1 \in [-1, 1].
>> n = 1000000;
>> x = zeros(n,1);
>> x(1) = 1;
>> for k=2:n
x(k) = \sin(x(k-1));
end
>> semilogy(abs(x), 'b.-')
>> title('Fixed Point Iteration, x_{k+1} = sin(x_k)', 'FontSize', 14)
```

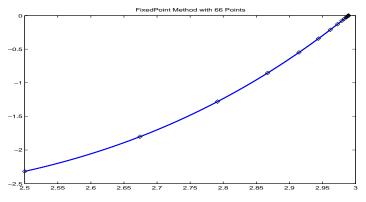
Fixed Point Example $x - \sin(x) = 0$: slow convergence

$$g(x) = \sin(x) \in [-1, 1]$$
 for $x \in [-1, 1]$,
 $|g'(x)| \le 1 \in [-1, 1]$.



Fixed Point Example $x - \log(2 + 2x^2) = 0$: normal convergence

$$g(x) = \log(2 + 2x^2) \in [2, 3]$$
 for $x \in [2, 3]$, $|g'(x)| \le \frac{4}{5} \in [2, 3]$.



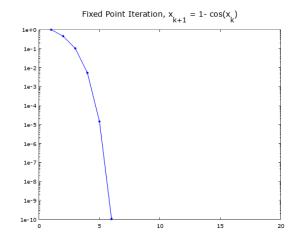
Fixed Point Example $x - (1 - \cos(x)) = 0$: fast convergence

```
|g'(x)| \approx \text{with.}
 >>
 >> n=20;
 >> x = zeros(n,1);
 >> x(1) = 1;
 >> for k=2:n
 x(k) = 1 - \cos(x(k-1));
 end
 >> semilogy(abs(x), 'b.-')
 warning: axis: omitting non-positive data in log plot
```

 $g(x) = 1 - \cos(x) \in [-1, 1]$ for $x \in [-1, 1]$,

Fixed Point Example $x - (1 - \cos(x)) = 0$: fast convergence

$$g(x) = 1 - \cos(x) \in [-1, 1]$$
 for $x \in [-1, 1]$, $|g'(x)| \le 1 \in [-1, 1]$.



Fixed Point Theorem (II)

Theorem 2.4 (Fixed-Point Theorem)

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' exists on (a, b) and that a constant 0 < k < 1 exists with

$$|g'(x)| \le k$$
, for all $x \in (a, b)$.

Then for any number p_0 in [a, b], the sequence defined by

$$p_n = g(p_{n-1}), \quad n \ge 1,$$

converges to the unique fixed point p in [a, b].



Proof of Thm 2.4

▶ A unique fixed point $p \in [a, b]$ exists.

$$|p_n-p|=|g(p_{n-1})-g(p)|=|g'(\xi_n)(p_{n-1}-p)|\leq k|p_{n-1}-p|$$

Þ

$$|p_n - p| \le k|p_{n-1} - p| \le k^2|p_{n-2} - p| \le \cdots \le k^n|p_0 - p|.$$

Since

$$\lim_{n\to\infty} k^n = 0,$$

 $\{p_n\}_{n=0}^{\infty}$ converges to p.

Newton's Method for solving f(p) = 0

- ▶ Suppose that $f \in C^2[a, b]$.
- ▶ Let $p_0 \in [a, b]$ be an approximation to p with

$$f'(p_0) \neq 0$$
, and $|p - p_0|$ "small".

▶ Taylor expand f(x) at x = p:

$$0 = f(p) = f(p_0) + (p - p_0)f'(p_0) + \frac{(p - p_0)^2}{2}f''(\xi(p)).$$

"Solve" for p:

$$p = p_0 - \frac{f(p_0)}{f'(p_0)} - \frac{(p - p_0)^2}{2f'(p_0)} f''(\xi(p))$$

$$\approx p_0 - \frac{f(p_0)}{f'(p_0)} \stackrel{def}{=} p_1.$$

Newton's Method:
$$p_{k+1} = p_k - \frac{f(p_k)}{f'(p_k)}, \quad k = 0, 1, \cdots$$

Newton's Method for solving f(p) = 0

$$p = p_0 - \frac{f(p_0)}{f'(p_0)} - \frac{(p - p_0)^2}{2f'(p_0)}$$

$$\approx p_0 - \frac{f(p_0)}{f'(p_0)} f''(\xi(p)) \stackrel{def}{=} p_1.$$

- ▶ If p_0 "close to" p, we can expect fast convergence.
- Best hope in practice: p₀ "not too far from" p. Newton's method may or may not converge.
- If Newton's method converges, it converges quickly.

Geometry of Newton's Method

▶ Taylor expand f(x) at x = p:

$$0 = f(p) = f(p_0) + (p - p_0)f'(p_0) + \frac{(p - p_0)^2}{2}f''(\xi(p)).$$

▶ Replace f(x) by a straight line:

$$f(p_0) + (p - p_0)f'(p_0) \approx 0.$$

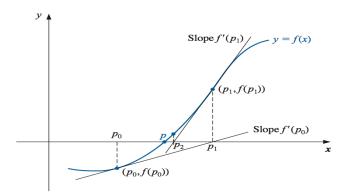
•

$$p\approx p_0-\frac{f(p_0)}{f'(p_0)}$$

is the horizontal intercept of straight line

$$y = f(p_0) + (x - p_0)f'(p_0)$$

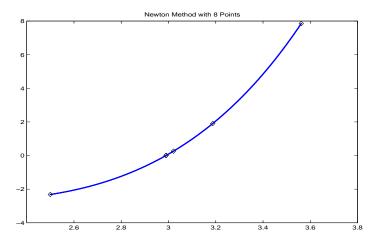
Newton Method



```
function [fun, dfun, x, out] = NewtonMethod(Fun, dFun, x0, params)
[FunFcn.msa] = fcnchk(Fun.0):
if ~isempty(msq)
   error('InvalidFUN', msq);
end
[dFunFcn,msq] = fcnchk(dFun,0);
if ~isempty(msg)
   error('InvalidFUN', msq);
end
out.flq = 1;
x(1) = x0:
N = params.MaxIt;
tol = params.tol:
out.x = []:
out.f =[];
for k = 1:N
   fun(k) = FunFcn(x(k)):
   dfun(k) = dFunFcn(x(k)):
   out.x = [out.x;x(k)];
   out.f = [out.f;fun(k)];
   if (abs(fun(k)) < tol)
       out.flq = 0;
       out.it = k:
       return:
   end
   if (dfun(k) == 0)
       out.it = k:
      return;
   end
   x(k+1) = x(k) - fun(k)/dfun(k):
end
```

Theorem 2.6 Let $f \in C^2[a,b]$. If $p \in (a,b)$ is such that f(p) = 0 and $f'(p) \neq 0$, then there exists a $\delta > 0$ such that Newton's method generates a sequence $\{p_n\}_{n=1}^{\infty}$ converging to p for any initial approximation $p_0 \in [p-\delta,p+\delta]$.

Newton Method for $f(x) = e^x - (2 + 2x^2)$



Computing square root with Newton's Method

▶ Given a > 0, $p \stackrel{def}{=} \sqrt{a}$ is positive root of equation

$$f(x) \stackrel{\text{def}}{=} x^2 - a = 0.$$

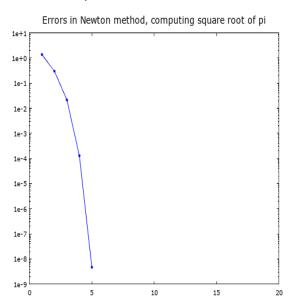
Newton's Method

$$p_{k+1} = p_k - \frac{p_k^2 - a}{2p_k} = \frac{1}{2} \left(p_k + \frac{a}{p_k} \right), k = 0, 1, 2, \cdots,$$

▶ Newton's Method is well defined for any $p_0 > 0$.



Newton Method for square root



Theorem 2.6 Let $f \in C^2[a,b]$. If $p \in (a,b)$ is such that f(p) = 0 and $f'(p) \neq 0$, then there exists a $\delta > 0$ such that Newton's method generates a sequence $\{p_n\}_{n=1}^{\infty}$ converging to p for any initial approximation $p_0 \in [p-\delta,p+\delta]$.

Proof of Theorem 2.6

- Newton's method is fixed point iteration $p_n = g(p_{n-1}), g(x) = x \frac{f(x)}{f'(x)}$.
- Since $f'(p) \neq 0$, there exists an interval $[p \delta_1, p + \delta_1] \subset [a, b]$ on which $f'(x) \neq 0$. Thus, g(x) is defined on $[p \delta_1, p + \delta_1]$.

$$g'(x) = 1 - \frac{f'(x) f'(x) - f(x) f''(x)}{(f'(x))^2} = \frac{f(x) f''(x)}{(f'(x))^2} \in C[p - \delta_1, p + \delta_1].$$

▶ Since g'(p) = 0, there exists $0 < \delta < \delta_1$ so that

$$|g'(x)| \le \kappa \quad (=\frac{1}{2}), \quad \text{for all} \quad x \in [p-\delta, p+\delta].$$

• If $x \in [p - \delta, p + \delta]$, then

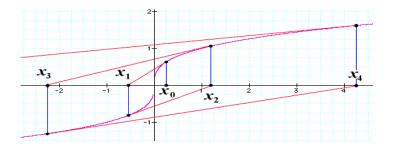
$$|g(x)-p| = |g(x)-g(p)| = |g'(\xi)(x-p)| \le \kappa |x-p| \le |x-p|.$$

Therefore $g(x) \in [p - \delta, p + \delta]$.

 \triangleright { p_n } converges to p by Fixed Point Theorem.



Newton Method Divergence Example: $f(x) = x^{1/3}$



Secant Method: Poor man's Newton Method

Motivation

- Newton method style of fast convergence
- Avoid need for derivative calculations.

Approach

- Newton method: $p_{n+1} = p_n \frac{f(p_n)}{f'(p_n)}$.
- ▶ Replace $f'(p_n)$ by its cheap approximation

$$f'(p_n) = \lim_{x \to \infty} \frac{f(p_n) - f(x)}{p_n - x} \approx \frac{f(p_n) - f(p_{n-1})}{p_n - p_{n-1}}.$$

Secant method

$$p_{n+1} = p_n - \frac{f(p_n)(p_n - p_{n-1})}{f(p_n) - f(p_{n-1})}, n = 1, 2, \cdots.$$



Secant Method: Geometry

▶ "Approximate" f(x) by a straight line

$$f(x) \approx \frac{(x-p_0)f(p_1)-(x-p_1)f(p_0)}{p_1-p_0}.$$

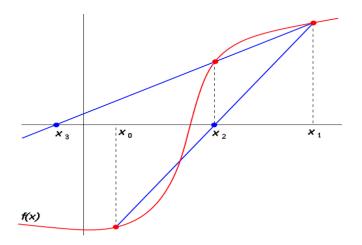
Both f(x) and straight line go through points $(p_0, f(p_0)), (p_1, f(p_1))$.

Let approximate root p_2 be the x-intercept of the straight line

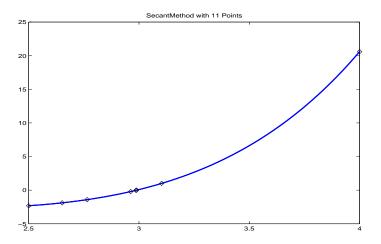
$$p_2 = \frac{p_0 f(p_1) - p_1 f(p_0)}{f(p_1) - f(p_0)} = p_1 - \frac{f(p_1)(p_1 - p_0)}{f(p_1) - f(p_0)}.$$



Secant Method: Illustration



Fixed point for $g(x) = log(2 + 2x^2)$



Performance: number of iterations vs. error in the solution

Function to be considered

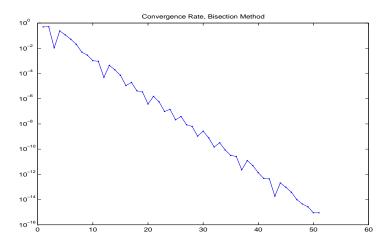
$$g(x) = log(2 + 2x^2), \quad f(x) = x - g(x) = x - log(2 + 2x^2).$$

▶ Root p of f (i.e., f(p) = 0)

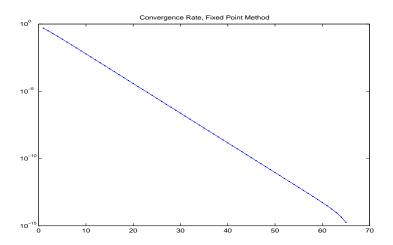
$$p = 2.98930778246493e + 00.$$

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

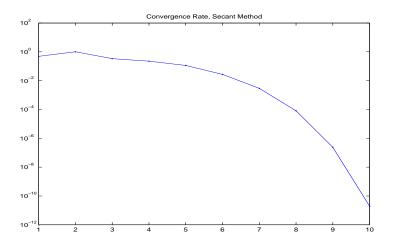
Bisection Method Order of Convergence



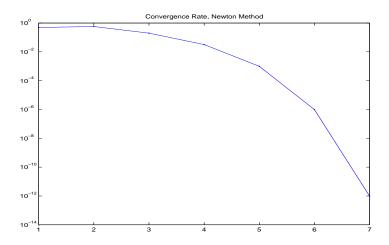
Fixed Point Iteration Order of Convergence



Secant Method Order of Convergence



Newton Method Order of Convergence



Order of convergence

Suppose $\{p_n\}_{n=0}^{\infty}$ is a sequence that converges to p, with $p_n \neq p$ for all n. If positive constants λ and α exist with

$$\lim_{n\to\infty}\frac{|p_{n+1}-p|}{|p_n-p|^{\alpha}}=\lambda,$$

then $\{p_n\}_{n=0}^{\infty}$ converges to p of order α , with asymptotic error constant λ .



- (i) If $\alpha = 1$ (and $\lambda < 1$), the sequence is **linearly convergent**.
- (ii) If $\alpha = 2$, the sequence is quadratically convergent.

Recall rate of convergence: the Big O

Suppose $\{\beta_n\}_{n=1}^{\infty}$ is a sequence known to converge to zero, and $\{\alpha_n\}_{n=1}^{\infty}$ converges to a number α . If a positive constant K exists with

$$|\alpha_n - \alpha| \le K|\beta_n|$$
, for large n ,

then we say that $\{\alpha_n\}_{n=1}^{\infty}$ converges to α with **rate**, **or order**, **of convergence** $O(\beta_n)$. (This expression is read "big oh of β_n ".) It is indicated by writing $\alpha_n = \alpha + O(\beta_n)$.

the Big O() = rate of convergence



Recall rate of convergence: the Big O

Suppose $\{\beta_n\}_{n=1}^{\infty}$ is a sequence known to converge to zero, and $\{\alpha_n\}_{n=1}^{\infty}$ converges to a number α . If a positive constant K exists with

$$|\alpha_n - \alpha| \le K|\beta_n|$$
, for large n ,

then we say that $\{\alpha_n\}_{n=1}^{\infty}$ converges to α with rate of convergence $O(\beta_n)$. (This expression is read "big oh of β_n ".) It is indicated by writing $\alpha_n = \alpha + O(\beta_n)$.

▶ Suppose that $\{p_n\}_{n=1}^{\infty}$ is linearly convergent to 0,

$$\lim_{n\to\infty}\frac{|p_{n+1}|}{|p_n|}=0.5,\quad \text{or roughly}\quad \frac{|p_{n+1}|}{|p_n|}\approx 0.5,$$

hence $p_n \approx (0.5)^n |p_0|$. Suppose that $\{\tilde{p}_n\}_{n=1}^{\infty}$ is quadratically convergent to 0,

$$\lim_{n\to\infty}\frac{|\tilde{p}_{n+1}|}{|\tilde{p}_n|^2}=0.5,\quad\text{or roughly}\quad\frac{|\tilde{p}_{n+1}|}{|\tilde{p}_n|^2}\approx0.5.$$

But now

$$|\tilde{p}_n| \approx 0.5 |\tilde{p}_{n-1}|^2 \approx (0.5)[0.5|\tilde{p}_{n-2}|^2]^2 = (0.5)^3 |\tilde{p}_{n-2}|^4$$

$$\approx (0.5)^3 [(0.5)|\tilde{p}_{n-3}|^2]^4 = (0.5)^7 |\tilde{p}_{n-3}|^8$$

$$\approx \dots \approx (0.5)^{2^n - 1} |\tilde{p}_0|^{2^n}.$$



n	Linear Convergence Sequence $\{p_n\}_{n=0}^{\infty}$ $(0.5)^n$	Quadratic Convergence Sequence $\{\tilde{p}_n\}_{n=0}^{\infty}$ $(0.5)^{2^n-1}$
1	5.0000×10^{-1}	5.0000×10^{-1}
2	2.5000×10^{-1}	1.2500×10^{-1}
3	1.2500×10^{-1}	7.8125×10^{-3}
4	6.2500×10^{-2}	3.0518×10^{-5}
5	3.1250×10^{-2}	4.6566×10^{-10}
6	1.5625×10^{-2}	1.0842×10^{-19}
7	7.8125×10^{-3}	5.8775×10^{-39}

Linear convergence: one more accurate bit per iteration Quadratic convergence: double # of correct bits per iteration.

