

# Error analysis

## Definition (Order of convergence)

Suppose  $p_n \rightarrow p$ . If  $\exists \lambda, \alpha > 0$  s.t.

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

then  $\{p_n\}$  is said to converge to  $p$  of **order**  $\alpha$ , with asymptotic error constant  $\lambda$ .

# Error analysis

## Definition (Convergence order of numerical methods)

An iterative method  $p_n = g(p_{n-1})$  is of **order**  $\alpha$  if the generated  $\{p_n\}$  converges to the solution  $p$  of  $p = g(p)$  at order  $\alpha$ .

In particular:

- ▶  $\alpha = 1$ : **linearly convergent**
- ▶  $\alpha = 2$ : **quadratically convergent**

# Example

## Example (Speed comparison: linear vs quadratic)

Suppose  $p_n$  (and  $q_n$  respectively) converges to 0 linearly (quadratically) with constant 0.5, enumerate the upper bound of  $|p_n|$  and  $|q_n|$ .

**Solution.** By definition of convergence order, we know

$$\lim_{n \rightarrow \infty} \frac{|p_{n+1}|}{|p_n|} = 0.5 \quad \text{and} \quad \lim_{n \rightarrow \infty} \frac{|q_{n+1}|}{|q_n|^2} = 0.5$$

Suppose that  $p_0$  and  $q_0$  are close enough to 0 s.t.

$|p_{n+1}|/|p_n| \approx 0.5$  and  $|q_{n+1}|/|q_n|^2 \approx 0.5$  for all  $n$ , then

$$|p_n| \approx 0.5|p_{n-1}| \approx 0.5^2|p_{n-2}| \approx \dots \approx 0.5^n|p_0|$$

$$|q_n| \approx 0.5|q_{n-1}|^2 \approx 0.5 \cdot 0.5^2|q_{n-2}|^4 \approx \dots \approx 0.5^{2^n-1}|q_0|^{2^n}$$

# Example

## Example (Speed comparison: linear vs quadratic)

Suppose  $p_0, q_0 \approx 0.5$ . Then

$n$	Linear $0.5^n$	Quadratic $0.5^{2^n-1}$
1	$5.0000 \times 10^{-1}$	$5.0000 \times 10^{-1}$
2	$2.5000 \times 10^{-1}$	$1.2500 \times 10^{-1}$
3	$1.2500 \times 10^{-1}$	$7.8125 \times 10^{-3}$
4	$6.2500 \times 10^{-2}$	$3.0518 \times 10^{-5}$
5	$3.1250 \times 10^{-2}$	$4.6566 \times 10^{-10}$
6	$1.5625 \times 10^{-2}$	$1.0842 \times 10^{-19}$
7	$7.8125 \times 10^{-3}$	$5.8775 \times 10^{-39}$

# Convergence rate of fixed point iteration algorithm

## Theorem (FPI alg has linear convergence rate)

Suppose  $g \in C[a, b]$  s.t.  $g(x) \in [a, b]$ ,  $\forall x \in [a, b]$ . If  $\exists k \in (0, 1)$  s.t.  $|g'(x)| \leq k$ ,  $\forall x \in (a, b)$ , then  $\{p_n\}$  generated by FPI algorithm converges to the unique FP of  $g(x)$  on  $[a, b]$  **linearly**.

## Proof.

We already know  $p_n \rightarrow p$  where  $p$  is the unique fixed point of  $g$  by FPI theorem. Also  $p_{n+1} - p = g(p_n) - g(p) = g'(\xi(p_n))(p_n - p)$  where  $\xi(p_n)$  is between  $p_n$  and  $p$ . So

$$\lim_{n \rightarrow \infty} \frac{|p_{n+1} - p|}{|p_n - p|} = \lim_{n \rightarrow \infty} |g'(\xi(p_n))| = |g'(\lim_{n \rightarrow \infty} \xi(p_n))| = |g'(p)| \leq k < 1$$

So  $p_n \rightarrow p$  linearly with constant  $k$ . □

# Improve convergence order of FPI to quadratic

## Theorem (Additional condition for quadratic rate)

*If  $g \in C^2[a, b]$  and  $g'(p) = 0$  for a FP  $p \in (a, b)$ , then  $\exists M > 0$  s.t.  $|g''(x)| \leq M, \forall x \in [a, b]$  and  $\exists \delta > 0$  s.t. sequence  $\{p_n\}$  by FPI started in  $[p - \delta, p + \delta]$  satisfies*

$$|p_{n+1} - p| \leq \frac{M}{2} |p_n - p|^2, \quad \forall n$$

# Improve convergence order of FPI

Proof.

$g \in C^2$ ,  $g(p) = p$ ,  $g'(p) = 0$  together imply  $\exists \delta > 0$  and  $k \in (0, 1)$  s.t.  $|g'(x)| \leq k < 1$  for all  $x \in [p - \delta, p + \delta]$  and  $g : [p - \delta, p + \delta] \rightarrow [p - \delta, p + \delta]$ . Also

$$g(p_n) = g(p) + g'(p)(p_n - p) + \frac{1}{2}g''(\xi(p_n))(p_n - p)^2$$

where  $\xi(p_n)$  is between  $p_n$  and  $p$ .

Since  $p_{n+1} = g(p_n)$ ,  $g(p) = p$ , and  $g'(p) = 0$ , we have  $p_{n+1} = p + \frac{1}{2}g''(\xi(p_n))(p_n - p)^2$ . So

$$\frac{|p_{n+1} - p|}{|p_n - p|^2} = \frac{1}{2}|g''(\xi(p_n))| \leq \frac{M}{2}$$

□

# Improve convergence order of FPI

Suppose we have a fixed point method with  $g(x) = x - \phi(x)f(x)$ .  
How to choose  $\phi$  such that FPI converges quadratically?

We need  $g$  s.t.  $g'(p) = 0$  at a FP  $p$  (root of  $f$ ):

$$g'(p) = 1 - \phi'(p)f(p) - \phi(p)f'(p) = 0$$

Since  $f(p) = 0$  we have  $\phi(p) = \frac{1}{f'(p)}$ . Choose  $\phi(x) = \frac{1}{f'(x)}$  s.t.

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

This is exactly Newton's method!

So Newton's method converges quadratically.



# Convergence of Newton's method when $f'(p) = 0$

We mentioned condition  $f'(p) \neq 0$  at the root  $p$  of  $f$  in the convergence proof of Newton's method above.

What if  $f'(p) = 0$ ? When will this happen and how to address it?

# Multiple roots

$f'(p) = 0$  at root  $p$  means  $p$  is not a “simple root”.

## Definition (Root multiplicity)

A solution  $p$  of  $f(x)$  is a **root (zero) of multiplicity  $m$**  if  $f(x) = (x - p)^m q(x)$  for some  $q$  s.t.  $\lim_{x \rightarrow p} q(x) \neq 0$ .

## Definition (Simple root)

$p$  is a **simple root (zero)** of  $f$  if its multiplicity  $m = 1$ .

# Multiple roots

## Theorem (S.N.C. for simple root)

$f \in C^1[a, b]$  has a simple root  $p \in (a, b)$  iff  $f(p) = 0$  and  $f'(p) \neq 0$ .

### Proof.

“ $\implies$ ”:  $f(x) = (x - p)q(x)$  where  $\lim_{x \rightarrow p} q(x) \neq 0$ . Then  $f'(x) = q(x) + (x - p)q'(x)$ . So  $f \in C^1$  implies

$$f'(p) = \lim_{x \rightarrow p} f'(x) = \lim_{x \rightarrow p} (q(x) + (x - p)q'(x)) \neq 0$$

“ $\impliedby$ ”:  $f(x) = f(p) + f'(\xi(x))(x - p)$  where  $\xi(x)$  between  $x$  and  $p$ . Define  $q(x) = f'(\xi(x))$  then

$$\lim_{x \rightarrow p} q(x) = \lim_{x \rightarrow p} f'(\xi(x)) = f'(\lim_{x \rightarrow p} \xi(x)) = f'(p) \neq 0$$

So  $f$  has a simple root at  $p$ . □

# Multiple roots

## Theorem (S.N.C. for multiple root)

$f \in C^m[a, b]$  has a zero  $p$  of multiplicity  $m$  iff

$$f(p) = f'(p) = \cdots = f^{(m-1)}(p) = 0 \quad \text{and} \quad f^{(m)}(p) \neq 0$$

## Proof.

Hint: Follow the proof above and use

$$(uv)^{(n)} = \sum_{k=0}^n \binom{n}{k} u^{(k)} v^{(n-k)}$$



# Example

## Example (Multiple root)

Let  $f(x) = e^x - x - 1$ , show that  $f(x)$  has a zero of multiplicity 2 at  $x = 0$ .

**Solution.**  $f(x) = e^x - x - 1$ ,  $f'(x) = e^x - 1$ , and  $f''(x) = e^x$ . So  $f(0) = f'(0) = 0$  and  $f''(0) = 1 \neq 0$ . By Theorem above  $f$  has root (zero) at  $x = 0$  of multiplicity 2.

# Modified Newton's method

Instead of using  $f(x)$  in Newton's method, we can replace  $f$  by

$$\mu(x) := \frac{f(x)}{f'(x)}$$

We need to show:

$p$  is a root (simple or not) of  $f \implies p$  is a simple root of  $\mu$

# Modified Newton's method

Recall that  $f$  has a root  $p$  of multiplicity  $m$  if  $f(x) = (x - p)^m q(x)$  for some  $q$  with  $\lim_{x \rightarrow p} q(x) \neq 0$ .

Now there is

$$\begin{aligned}\mu(x) &= \frac{f(x)}{f'(x)} = \frac{(x - p)^m q(x)}{m(x - p)^{m-1} q(x) + (x - p)^m q'(x)} \\ &= (x - p) \cdot \frac{q(x)}{mq(x) + (x - p)q'(x)}\end{aligned}$$

where  $\frac{q(x)}{mq(x) + (x - p)q'(x)} \rightarrow \frac{1}{m} \neq 0$  as  $x \rightarrow p$ .

By definition,  $\mu(x)$  has simple root at  $p$ , i.e.,  $\mu(p) = 0$  and  $\mu'(p) \neq 0$ .

# Modified Newton's method

Now we use  $\mu(x)$  instead of  $f(x)$  in Newton's method:

$$g(x) = x - \frac{\mu(x)}{\mu'(x)} = x - \frac{(f(x)/f'(x))}{(f(x)/f'(x))'} = \dots = x - \frac{f(x)f'(x)}{(f'(x))^2 - f(x)f''(x)}$$

The **modified Newton's method** is

$$p_n = p_{n-1} - \frac{f(p_{n-1})f'(p_{n-1})}{(f'(p_{n-1}))^2 - f(p_{n-1})f''(p_{n-1})}$$

Drawbacks of the modified Newton's method:

- ▶ Needs  $f''$  in computation.
- ▶ Denominator approximates 0 as  $p_n \rightarrow p$ , so round-off may degrade convergence.



# Accelerating convergence

We showed that FPI generally has linear convergence only. How to improve?

Suppose  $N$  is large, and  $p_n, p_{n+1}, p_{n+2}$  satisfy

$$\frac{p_{n+1} - p}{p_n - p} \approx \frac{p_{n+2} - p}{p_{n+1} - p}$$

$$\iff (p_{n+1} - p)^2 \approx (p_n - p)(p_{n+2} - p) = p_n p_{n+2} - p(p_{n+2} + p_n) + p^2$$

$$\iff p \approx \frac{p_n p_{n+2} - p_{n+1}^2}{p_{n+2} - 2p_{n+1} + p_n} = \dots = p_n - \frac{(p_{n+1} - p_n)^2}{p_{n+2} - 2p_{n+1} + p_n}$$

# Aitken's $\Delta^2$ method

Denote  $\Delta p_n := p_{n+1} - p_n$ , called **forward difference**, and

$$\begin{aligned}\Delta^2 p_n &:= \Delta(\Delta p_n) = \Delta(p_{n+1} - p_n) \\ &= (p_{n+2} - p_{n+1}) - (p_{n+1} - p_n) \\ &= p_{n+2} - 2p_{n+1} + p_n\end{aligned}$$

So the result above can be written as  $p \approx p_n - \frac{(\Delta p_n)^2}{\Delta^2 p_n}$ .

**Aitken's  $\Delta^2$  method:**

Given  $\{p_n\}$  generated by FPI, set  $\hat{p}_n = p_n - \frac{(\Delta p_n)^2}{\Delta^2 p_n}$ . Then  $\hat{p}_n \rightarrow p$  faster than  $p_n$ .

# Aitken's $\Delta^2$ method

What does it mean by “faster”?

Theorem (Faster convergence by Aitken's  $\Delta^2$  method)

If  $p_n \rightarrow p$  linearly with  $\lim_{n \rightarrow \infty} \frac{p_{n+1} - p}{p_n - p} < 1$ , then  $\hat{p}_n$  computed by Aitken's  $\Delta^2$  method satisfy

$$\lim_{n \rightarrow \infty} \frac{\hat{p}_n - p}{p_n - p} = 0$$

**Proof.**

Hint: Define  $e_n := p_n - p$ , then  $\Delta e_n = \Delta p_n$ ,  $\Delta^2 e_n = \Delta^2 p_n$ , and  $\frac{e_{n+1}}{e_n} \rightarrow \lambda < 1$ . Then

$$\frac{\hat{p}_n - p}{p_n - p} = \frac{p_n - \frac{(\Delta p_n)^2}{\Delta^2 p_n} - p}{p_n - p} = \frac{e_n - \frac{(\Delta e_n)^2}{\Delta^2 e_n}}{e_n} = \frac{\frac{e_{n+2}}{e_{n+1}} - \frac{e_{n+1}}{e_n}}{\frac{e_{n+2}}{e_{n+1}} - 2 + \frac{e_n}{e_{n+1}}} \rightarrow \frac{\lambda - \lambda}{\lambda - 2 + \frac{1}{\lambda}} = 0$$



# Steffenson's method

Aitken's method computes  $\hat{p}_n$  separately from  $p_n$ . Steffenson's method makes use of  $\hat{p}_n$  to compute future  $p_n$ .

**Steffenson's method:** given  $g$  for FPI, compute

$$\begin{aligned} p_0^{(0)}, & \quad p_0^{(0)} = g(p_0^{(0)}), \quad p_0^{(0)} = g(p_0^{(0)}) \\ p_0^{(1)} = p_0^{(0)} - \frac{(\Delta p_0^{(0)})^2}{\Delta^2 p_0^{(0)}}, & \quad p_1^{(1)} = g(p_0^{(1)}), \quad p_2^{(1)} = g(p_1^{(1)}) \\ p_0^{(2)} = p_0^{(1)} - \frac{(\Delta p_0^{(1)})^2}{\Delta^2 p_0^{(1)}}, & \quad p_1^{(2)} = g(p_0^{(2)}), \quad p_2^{(2)} = g(p_1^{(2)}) \\ & \quad \vdots \end{aligned}$$

# Steffenson's method

## Steffenson's method

- ▶ **Input.** Initial guess  $p_0$ ,  $\epsilon_{\text{tol}}$ ,  $N_{\text{max}}$ . Set  $N = 1$ .
- ▶ While  $N \leq N_{\text{max}}$ , do :
  1. Set  $p_1 = g(p_0)$ ,  $p_2 = g(p_1)$  and  $p = p_0 - \frac{(p_1 - p_0)^2}{p_2 - 2p_1 + p_0}$
  2. If  $|p - p_0| < \epsilon_{\text{tol}}$ , STOP
  3.  $p_0 = p$
  4. Set  $N = N + 1$
- ▶ **Output.** Return  $p$ . If  $N \geq N_{\text{max}}$ , print( "Max iteration reached." ).

# Steffenson's method

## Theorem

*Suppose  $g(x)$  has a fixed point  $p$  and  $g'(p) \neq 1$ . If  $\exists \delta > 0$ , s.t.  $f \in C^3[p - \delta, p + \delta]$ , then Steffenson's method generates a sequence  $\{p_n\}$  converging to  $p$  quadratically for any initial  $p_0 \in [p - \delta, p + \delta]$ .*