

# 11. Nonlinear equations with one variable

- definition and examples
- bisection method
- Newton's method
- secant method

## Definition and examples

$x$  is a *zero* (or *root*) of a function  $f$  if  $f(x) = 0$

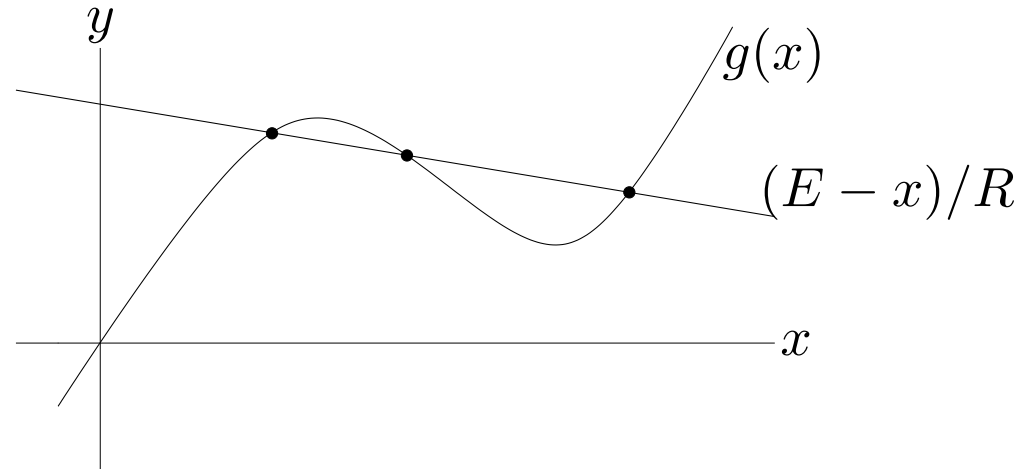
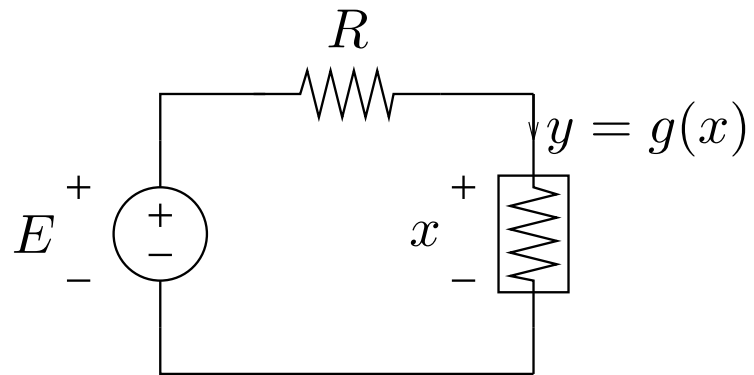
### examples

- $f(x) = e^x$  has no zeros
- $f(x) = e^x - e^{-x}$  has one zero
- $f(x) = e^x - e^{-x} - 3x$  has three zeros
- $f(x) = \cos x$  has infinitely many zeros

*cf.*, one linear equation in one variable  $ax = b$

- a unique solution if  $a \neq 0$
- no solution if  $a = 0, b \neq 0$
- any  $x \in \mathbf{R}$  is a solution if  $a = b = 0$

## Example: nonlinear static circuit



operating point satisfies

$$g(x) - \frac{E - x}{R} = 0$$

- one nonlinear equation in one variable  $x$
- three solutions

# Characteristics of algorithms for nonlinear equations

## how $f$ is described

- user provides subroutine to compute  $f(x)$  (and possibly  $f'(x)$ ) at  $x$
- called 'black box' or 'oracle' model for describing  $f$
- evaluating  $f$  and  $f'$  can be expensive (*e.g.*, require a circuit simulation)

## limitations of algorithms

- there exist no algorithms that are guaranteed to find all solutions
- most algorithms find at most one solution
- need prior information from the user: *e.g.*, an interval that contains a zero, or a point near a solution

## methods for solving nonlinear equations are iterative

- generate a sequence of points  $x^{(k)}$ ,  $k = 0, 1, 2, \dots$  that converge to a solution;  $x^{(k)}$  is called the  $k$ th *iterate*;  $x^{(0)}$  is the *starting point*
- computing  $x^{(k+1)}$  from  $x^{(k)}$  is called one *iteration* of the algorithm
- each iteration typically requires one evaluation of  $f$  (or  $f$  and  $f'$ ) at  $x^{(k)}$
- algorithms need a stopping criterion, *e.g.*, terminate if

$$|f(x^{(k)})| \leq \text{specified tolerance}$$

- speed of the algorithm depends on:
  - the cost of evaluating  $f(x)$  (and possibly,  $f'(x)$ )
  - the number of iterations

# Analyzing speed of convergence

suppose  $x^{(k)} \rightarrow x^*$  with  $f(x^*) = 0$ ; how fast does  $x^{(k)}$  go to  $x^*$ ?

**error** after  $k$  iterations:

- **absolute error:**  $|x^{(k)} - x^*|$
- **relative error:**  $|x^{(k)} - x^*|/|x^*|$  (defined if  $x^* \neq 0$ )
- **number of correct digits:**

$$\left\lfloor -\log_{10} \left( \frac{|x^{(k)} - x^*|}{|x^*|} \right) \right\rfloor$$

(defined if  $x^* \neq 0$  and  $|x^{(k)} - x^*|/|x^*| \leq 1$ )

**rates of convergence** of a sequence  $x^{(k)}$  with limit  $x^*$

- linear convergence: there exists a  $c \in (0, 1)$  such that

$$|x^{(k+1)} - x^*| \leq c |x^{(k)} - x^*| \quad \text{for sufficiently large } k$$

- R-linear convergence: there exists  $c \in (0, 1)$ ,  $M > 0$  such that

$$|x^{(k)} - x^*| \leq M c^k \quad \text{for sufficiently large } k$$

- quadratic convergence: there exists a  $c > 0$  s.t.

$$|x^{(k+1)} - x^*| \leq c |x^{(k)} - x^*|^2 \quad \text{for sufficiently large } k$$

- superlinear convergence: there exists a sequence  $c_k$  with  $c_k \rightarrow 0$  s.t.

$$|x^{(k+1)} - x^*| \leq c_k |x^{(k)} - x^*| \quad \text{for sufficiently large } k$$

**interpretation** (if  $x^\star \neq 0$ ): let

$$r^{(k)} = -\log_{10}\left(\frac{|x^{(k)} - x^\star|}{|x^\star|}\right)$$

(i.e.,  $r^{(k)} \approx$  the number of correct digits at iteration  $k$ )

- linear convergence: we gain roughly  $-\log_{10} c$  correct digits per step

$$r^{(k+1)} \geq r^{(k)} - \log_{10} c$$

- quadratic convergence: for  $k$  sufficiently large, number of correct digits roughly doubles in one step

$$r^{(k+1)} \geq -\log(c|x^\star|) + 2r^{(k)}$$

- superlinear convergence: number of correct digits gained per step increases with  $k$

$$r^{(k+1)} - r^{(k)} \rightarrow \infty$$



**examples** (with  $x^* = 1$ )

- $x^{(k)} = 1 + 0.5^k$  converges linearly (with  $c = 1/2$ ):

$$\frac{|x^{(k+1)} - 1|}{|x^{(k)} - 1|} = \frac{2^k}{2^{k+1}} = \frac{1}{2}$$

- $x^{(k)} = 1 + 0.5^{2^k}$  converges quadratically (with  $c = 1$ )

$$\frac{|x^{(k+1)} - 1|}{|x^{(k)} - 1|^2} = \frac{(2^{2^k})^2}{2^{2^{k+1}}} = 1$$

- $x^{(k)} = 1 + (1/(k+1))^k$  converges superlinearly

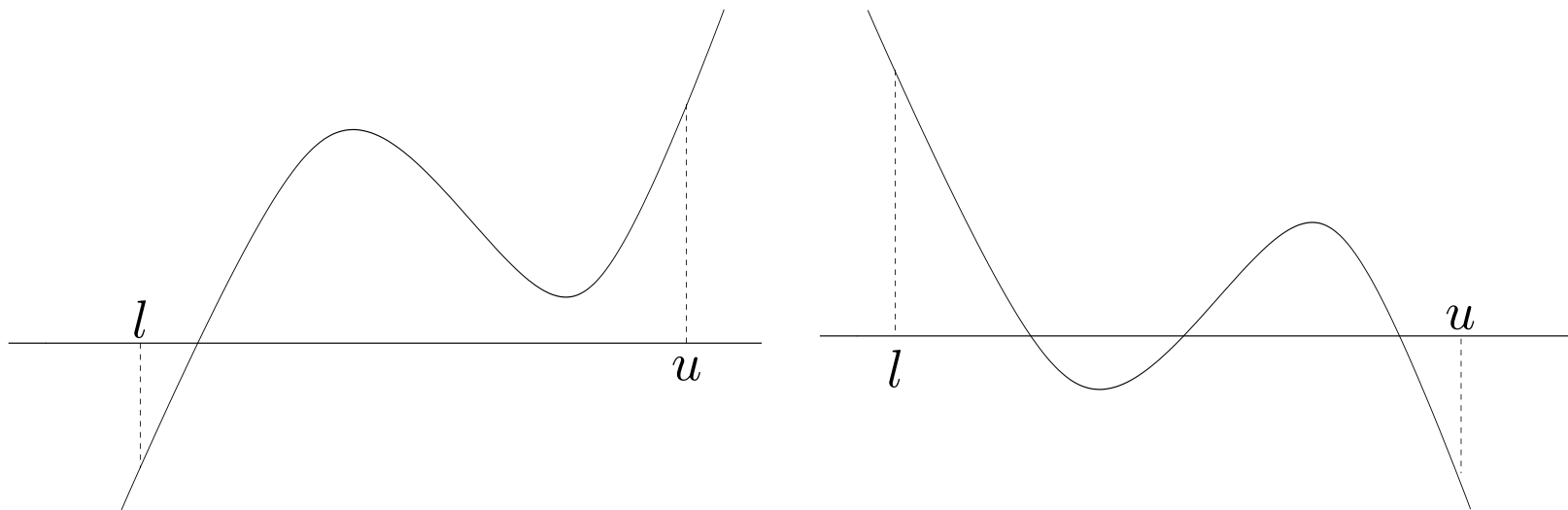
$$\frac{|x^{(k+1)} - 1|}{|x^{(k)} - 1|} = \frac{(k+1)^k}{(k+2)^{k+1}} \rightarrow 0$$

$k$	$1 + 0.5^k$	$1 + 0.5^{2^k}$	$1 + (1/(k + 1)^k)$
0	2.0000000000000000	1.5000000000000000	2.0000000000000000
1	1.5000000000000000	1.2500000000000000	1.5000000000000000
2	1.2500000000000000	1.0625000000000000	1.1111111111111111
3	1.1250000000000000	1.0039062500000000	1.0156250000000000
4	1.0625000000000000	1.00001525878906	1.0016000000000000
5	1.0312500000000000	1.00000000023283	1.00012860082305
6	1.0156250000000000	1.0000000000000000	1.00000849985975
7	1.0078125000000000	1.0000000000000000	1.00000047683716
8	1.0039062500000000	1.0000000000000000	1.00000002323057
9	1.00195313125000	1.0000000000000000	1.00000000100000
10	1.00097656250000	1.0000000000000000	1.00000000003855

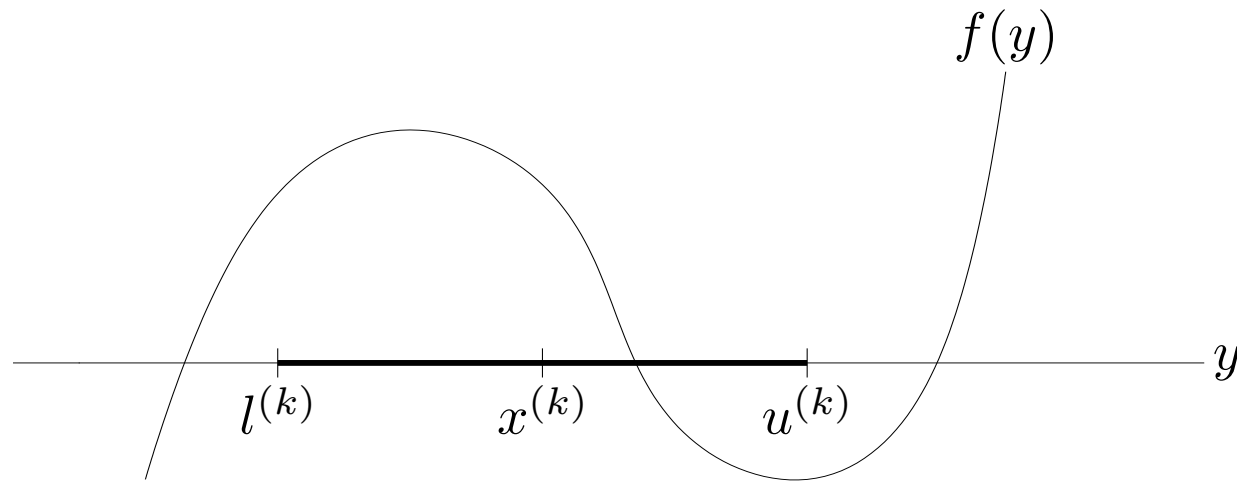
- sequence 1: we gain roughly  $-\log_{10}(c) = 0.3$  correct digits per step
- sequence 2: number of correct digits roughly doubles at each step
- sequence 3: number of correct digits gained per step increases slowly (from 0.5 initially to 2 near the end)

# Bisection method

$f : \mathbf{R} \rightarrow \mathbf{R}$ , continuous



if  $f(l)f(u) < 0$ , then the interval  $[l, u]$  contains at least one zero



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**given**  $l, u$  with  $l < u$  and  $f(l)f(u) < 0$ ; a required tolerance  $\epsilon > 0$

**repeat**

1.  $x := (l + u)/2$ .
2. Compute  $f(x)$ .
3. **if**  $f(x) = 0$ , **return**  $x$ .
4. **if**  $f(x)f(l) < 0$ ,  $u := x$ , **else**,  $l := x$ .

**until**  $u - l \leq \epsilon$

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one function evaluation per iteration

## convergence rate

- $u^{(k)} - l^{(k)}$  measures our uncertainty in localizing a zero  $x^*$ :

$$|x^{(k)} - x^*| \leq u^{(k)} - l^{(k)}$$

- uncertainty is halved at each iteration:

$$u^{(k)} - l^{(k)} = \left(\frac{1}{2}\right)^k (u^{(0)} - l^{(0)})$$

$$|x^{(k)} - x^*| \leq \left(\frac{1}{2}\right)^k (u^{(0)} - l^{(0)})$$

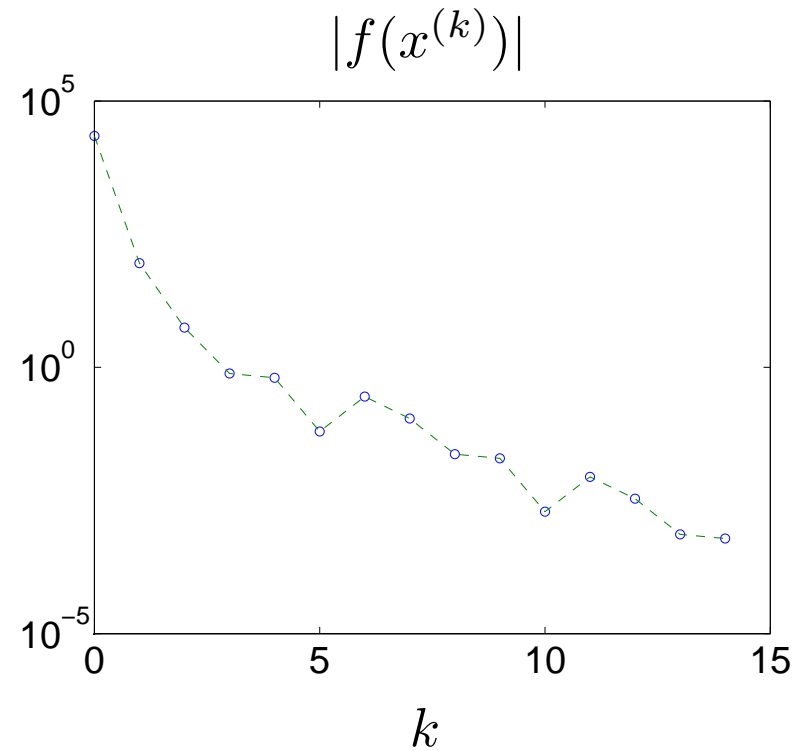
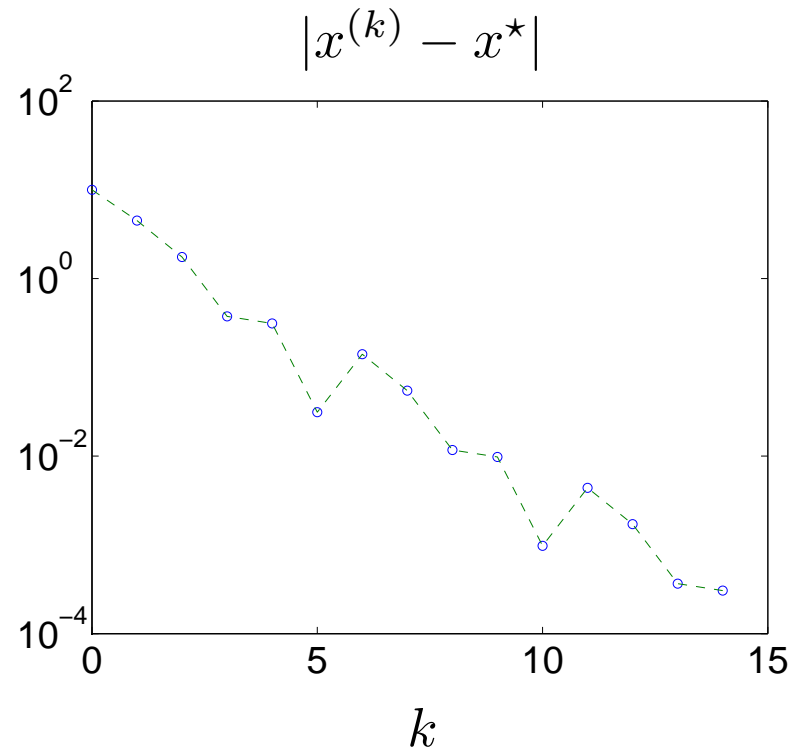
*i.e.*, R-linear convergence with  $c = 1/2$ ,  $M = u^{(0)} - l^{(0)}$

- number of iterations required for  $u^{(k)} - l^{(k)} \leq \epsilon$ :

$$\log_2 \frac{u^{(0)} - l^{(0)}}{\epsilon}$$

**example:**  $f(x) = e^x - e^{-x}$

- unique zero  $x^* = 0$
- start bisection method with  $l = -1$ ,  $u = 21$



# Newton's method

$f : \mathbf{R} \rightarrow \mathbf{R}$ , differentiable

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**given** initial  $x$ , required tolerance  $\epsilon > 0$

**repeat**

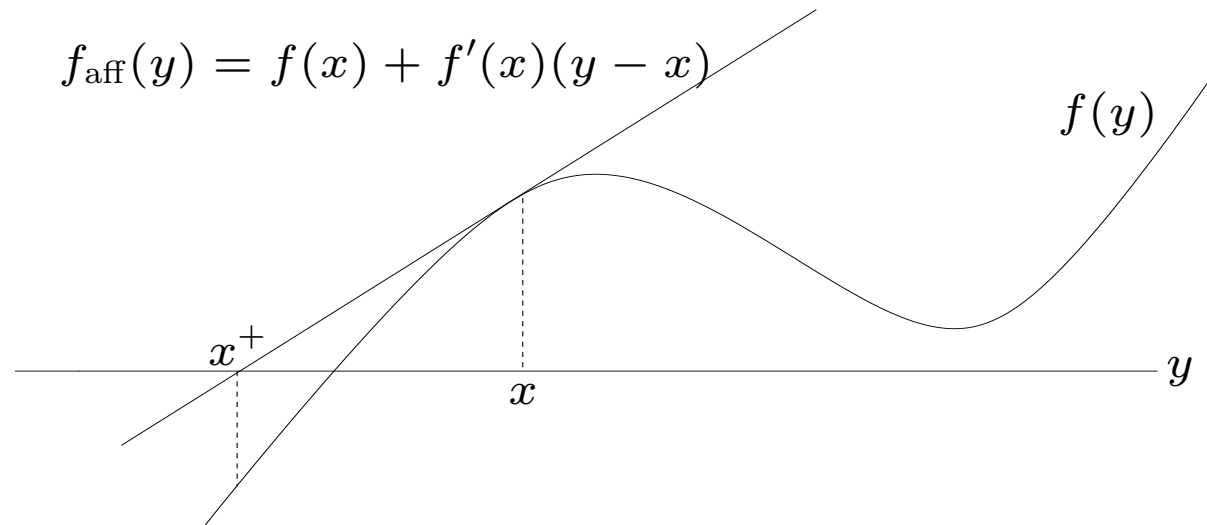
1. Compute  $f(x)$  and  $f'(x)$ .
2. **if**  $|f(x)| \leq \epsilon$ , **return**  $x$ .
3.  $x := x - f(x)/f'(x)$ .

**until** maximum number of iterations is exceeded.

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- each iteration requires one evaluation of  $f$  and  $f'$
- there exist other (more sophisticated) stopping criteria
- we assume  $f'(x^{(k)}) \neq 0$ , all  $k$

**interpretation** (with notation  $x = x^{(k)}$ ,  $x^+ = x^{(k+1)}$ )



- make affine approximation of  $f$  around  $x$  using Taylor series expansion:

$$f_{\text{aff}}(y) = f(x) + f'(x)(y - x)$$

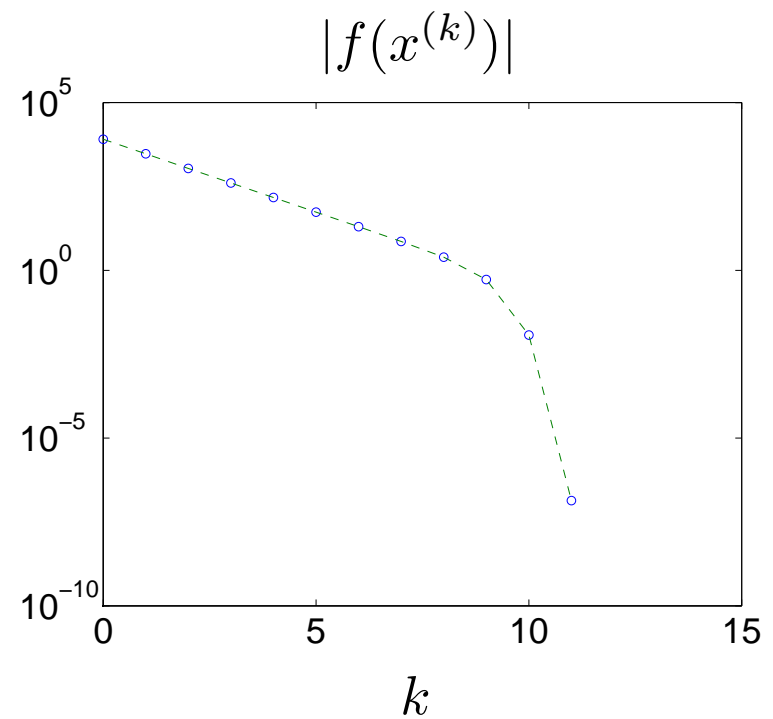
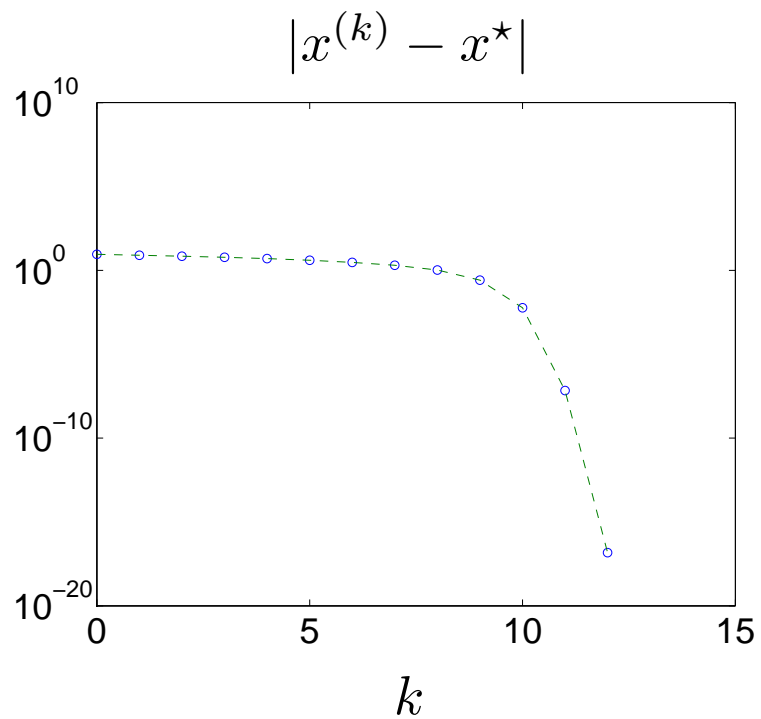
- solve the linearized equation  $f_{\text{aff}}(y) = 0$  and take the solution  $y$  as  $x^+$ :

$$x^+ = x - f(x)/f'(x)$$



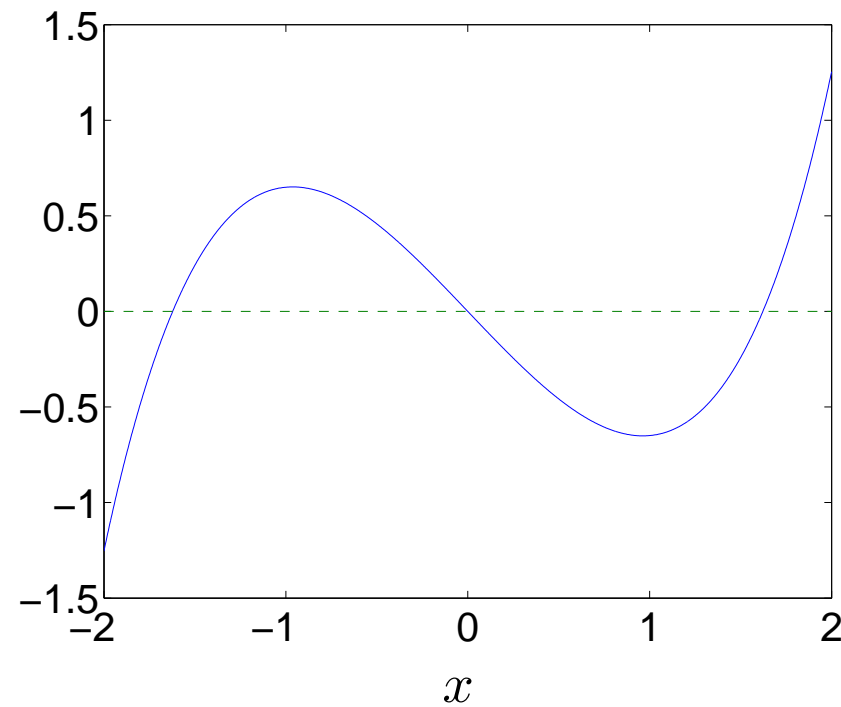
# Examples

- $f(x) = e^x - e^{-x}$ , start at  $x^{(0)} = 10$



asymptotic convergence is much faster than bisection method

- $f(x) = e^x - e^{-x} - 3x$



- start at  $x^{(0)} = -1$ : converges to  $x = -1.62$
- start at  $x^{(0)} = -0.8$ : converges to  $x = 1.62$
- start at  $x^{(0)} = -0.7$ : converges to  $x = 0$

converges to a different solution depending on the starting point

- $f(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$  (unique root at  $x = 0$ )

– start at  $x^{(0)} = 0.9$ :

$$x^{(1)} = -5.7 \cdot 10^{-1}$$

$$x^{(2)} = 1.3 \cdot 10^{-1}$$

$$x^{(3)} = -1.6 \cdot 10^{-3}$$

$$x^{(4)} = 2.5 \cdot 10^{-9}$$

$$x^{(5)} = -3.0 \cdot 10^{-17}$$

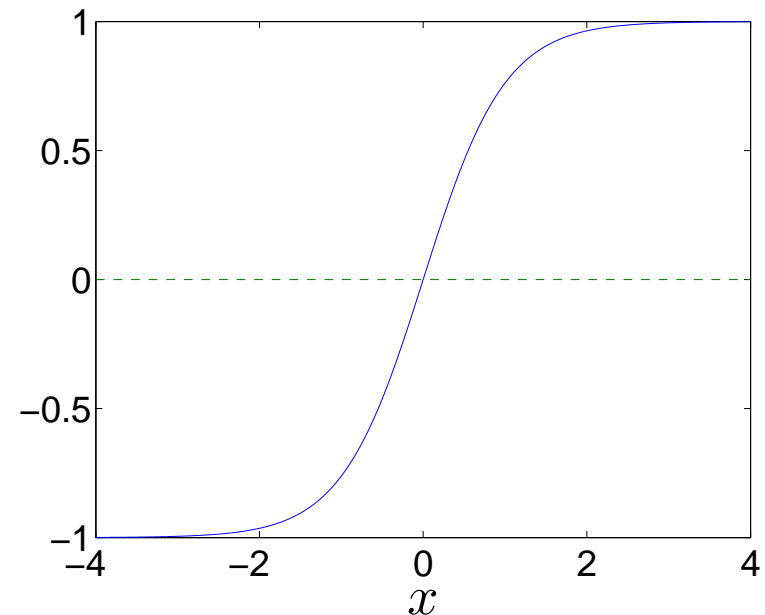
converges very rapidly

– start at  $x^{(0)} = 1.1$ :

$$x^{(1)} = -1.1 \cdot 10^0, \quad x^{(2)} = 1.2 \cdot 10^0, \quad x^{(3)} = -1.7 \cdot 10^0,$$

$$x^{(4)} = 5.7 \cdot 10^0, \quad x^{(5)} = -2.3 \cdot 10^4$$

does not converge



## conclusion

- Newton's method works very well if we start near a solution
- it may not work at all if we start too far from a solution
- if there are multiple solutions, it may converge to a different solution depending on the starting point; it does not necessarily converge to the solution closest to the starting point

# Secant method

$f : \mathbf{R} \rightarrow \mathbf{R}$ , continuous

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**given** two initial points  $x, x_{\text{prev}}$ , required tolerance  $\epsilon > 0$

**repeat**

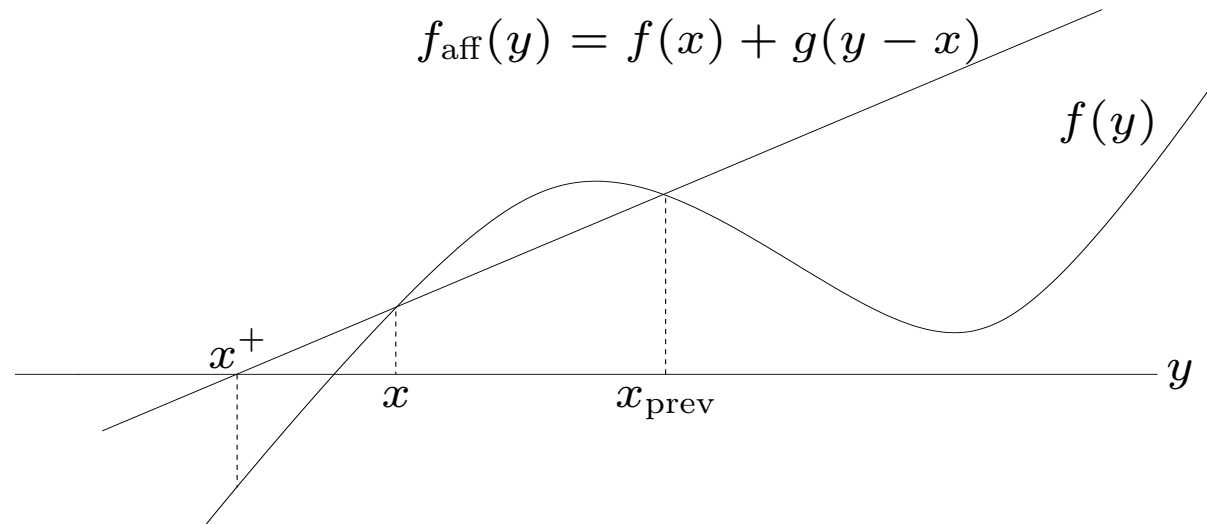
1. Compute  $f(x)$
2. **if**  $|f(x)| \leq \epsilon$ , **return**  $x$ .
3.  $g := (f(x) - f(x_{\text{prev}}))/(x - x_{\text{prev}})$ .
4.  $x_{\text{prev}} := x$ .
5.  $x := x - f(x)/g$ .

**until** maximum number of iterations is exceeded.

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- first iteration requires two evaluations of  $f$  (at  $x$  and  $x_{\text{prev}}$ )
- subsequent iterations require one evaluation (at  $x$ )
- we assume  $g \neq 0$

**interpretation** (with notation:  $x = x^{(k)}$ ,  $x^+ = x^{(k+1)}$ ,  $x_{\text{prev}} = x_{\text{prev}}^{(k)}$ )



- affine approximation  $f_{\text{aff}}$  with  $f_{\text{aff}}(x) = f(x)$ ,  $f_{\text{aff}}(x_{\text{prev}}) = f(x_{\text{prev}})$ :

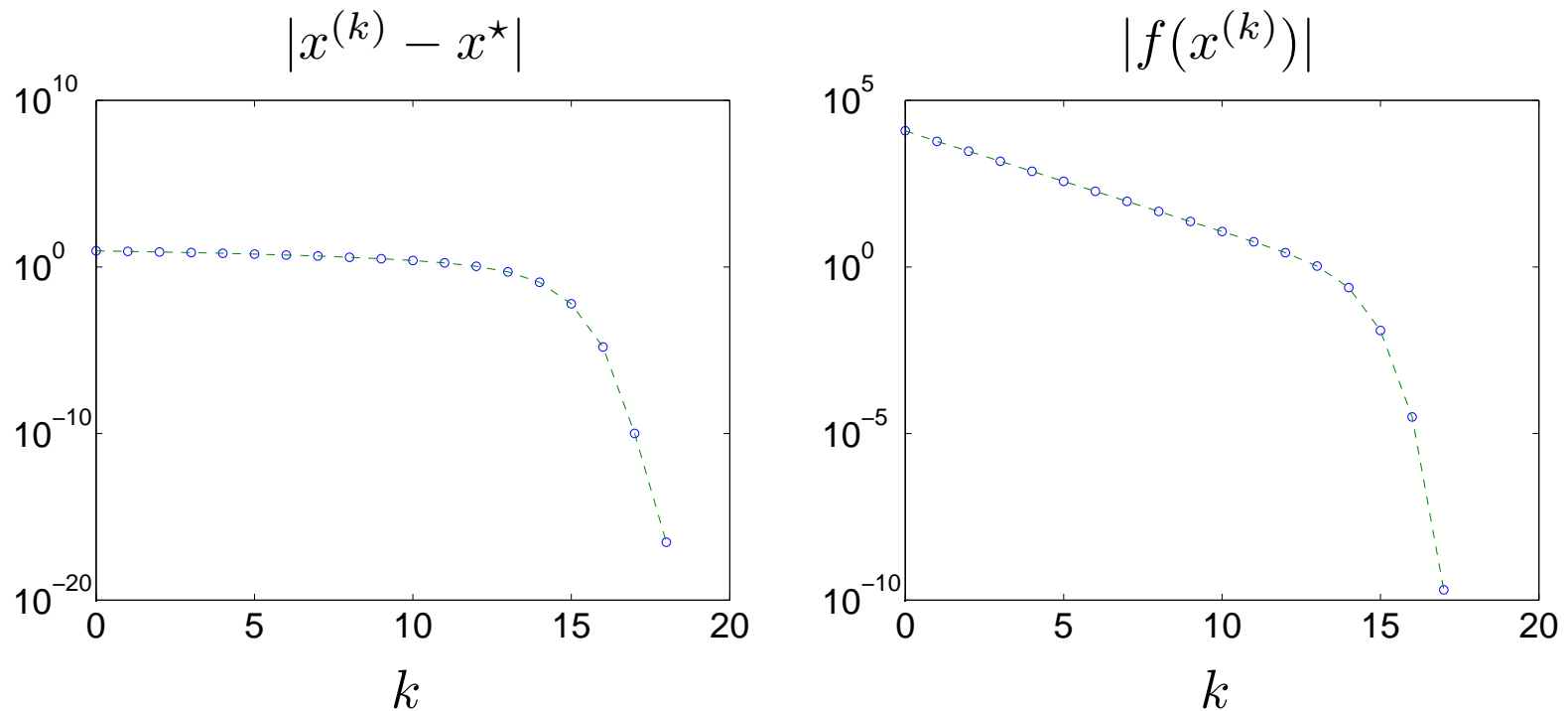
$$f_{\text{aff}}(y) = f(x) + g(y - x) \quad \text{with} \quad g = \frac{f(x) - f(x_{\text{prev}})}{x - x_{\text{prev}}}$$

- solve linear equation  $f_{\text{aff}}(y) = 0$  and take the solution as new iterate  $x^+$ :

$$x^+ = x - f(x)/g$$

# Examples

- $f(x) = e^x - e^{-x}$ , start at  $x^{(0)} = 10$ ,  $x_{\text{prev}}^{(0)} = 11$



fast asymptotic convergence, but slower than Newton method

- other examples: secant method works well if we start near a solution; may not converge otherwise

# Convergence of Newton and secant methods

**Newton method:** if  $f'(x^*) \neq 0$  and  $x^{(0)}$  is sufficiently close to  $x^*$ , then Newton's method converges and there exists a  $c > 0$  such that

$$|x^{(k+1)} - x^*| \leq c |x^{(k)} - x^*|^2$$

*i.e.*, quadratic convergence

**secant method:** if  $f'(x^*) \neq 0$  and  $x^{(0)}$  is sufficiently close to  $x^*$ , then the secant method converges and there exists a  $c > 0$  such that

$$|x^{(k+1)} - x^*| \leq c |x^{(k)} - x^*|^r$$

where  $r = (1 + \sqrt{5})/2 \approx 1.6$

*i.e.*, superlinear convergence



# Summary

## bisection method

- does not require derivatives
- user must provide initial interval  $[l, u]$  with  $f(l)f(u) < 0$
- R-linear convergence

## Newton's method

- requires derivatives
- user must provide starting point near a solution
- quadratic convergence

## secant method

- does not require derivatives
- user must provide two starting points near a solution
- superlinear convergence