2072U Computational Science I Winter 2022

| Week | Topic |
|-------|---|
| 1 | Introduction |
| 1–2 | Solving nonlinear equations in one variable |
| 3–4 | Solving systems of (non)linear equations |
| 5–6 | Computational complexity |
| 6–8 | Interpolation and least squares |
| 8–10 | Integration & differentiation |
| 10-12 | Additional Topics |
| | |

- 1. First method: bisection
- 2. Bisection
- 3. Newton iteration
- 4. Comparing the two
- 5. Secant method
- 6. Recursion

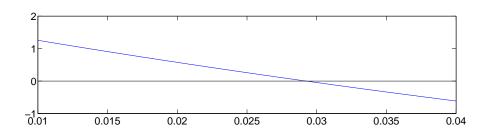
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The problem

Suppose we have a continuous function f on some domain [a, b]. Find $x^* \in [a, b]$ such that

$$f(x^*)=0.$$

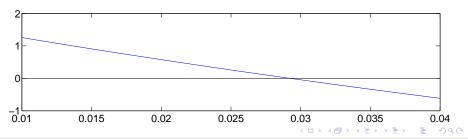




Intermediate Value Theorem

Suppose we have a continuous function f on some domain [a, b]. Then if k is some number between f(a) and f(b) then there exists at least one number c in the interval [a, b] such that f(c) = k. That is,

$$f(a) < k < f(b)$$
 or $f(a) > k > f(b)$,
 $\rightarrow \exists c \in [a, b]$ s.t. $f(c) = k$

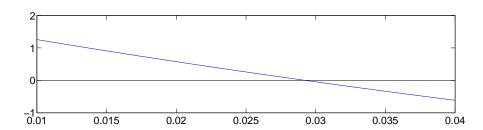




Intermediate Value Theorem for k = 0

Suppose we have a continuous function f on some domain [a,b]. Then if f(a)f(b)<0 then there exists at least one number c in the interval [a,b] such that f(c)=0. That is,

$$f(a)f(b) < 0 \rightarrow \exists c \in [a,b] \text{ s.t. } f(c) = 0$$



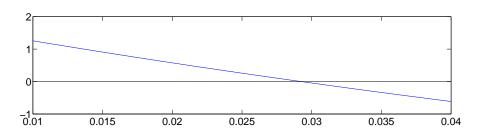
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Bisection

Suppose we have a continuous function f on some domain [a,b]. Find a, b such that f(a)f(b)<0, then by the Intermediate Value Theorem, there exists at least one solution $x^*\in[a,b]$ to the equation

$$f(x^*)=0.$$





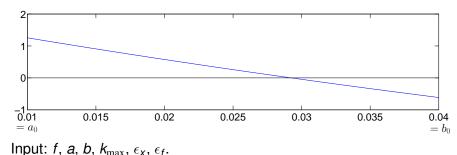
Bisection in *pseudo-code*:

Input: f, a, b, k_{max} , ϵ_X , ϵ_f .

- 1. Set $a_0 = a$, $b_0 = b$.
- **2**. Do for $k = 1, ..., k_{max}$
 - ► Let $c_k = (a_{k-1} + b_{k-1})/2$ and $f_k = f(c_k)$.
 - If $f_k f(a_{k-1}) > 0$ then let $a_k = c_k$ and $b_k = b_{k-1}$ else let $a_k = a_{k-1}$ and $b_k = c_k$.
 - If $|b_k a_k| < \epsilon_x$ or $|f(c_k)| < \epsilon_f$ break.
- 3. Return $x^* = c_k$ and $|b_k a_k|$.

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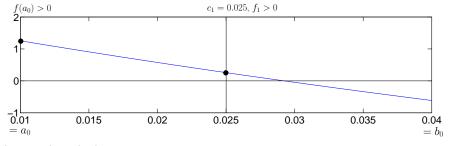
1. Set $a_0 = a$, $b_0 = b$.

- 2. Do for $k = 1, \ldots, k_{max}$
 - ▶ Let $c_k = (a_{k-1} + b_{k-1})/2$ and $f_k = f(c_k)$.
 - If $f_k f(a_{k-1}) > 0$ then let $a_k = c_k$ and $b_k = b_{k-1}$ else let $a_k = a_{k-1}$ and $b_k = c_k$.
 - If $|b_k a_k| < \epsilon_x$ or $|f(c_k)| < \epsilon_f$ break.
- 3. Return $x^* = c_k$ and $|b_k a_k|$.

For $\epsilon_x = \epsilon_f = 0.001$, $x^* \approx 0.0296875$ with error 9.375×10^{-4} .

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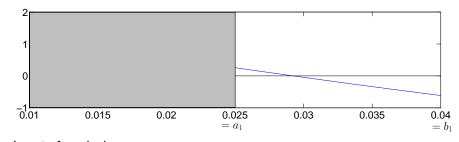


Input: f, a, b, k_{max} , ϵ_X , ϵ_f . 1. Set $a_0 = a$, $b_0 = b$.

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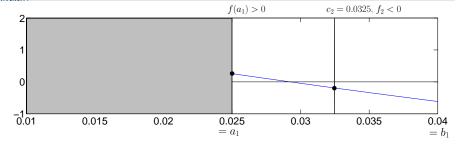
Input: f, a, b, k_{max} , ϵ_X , ϵ_f . 1. Set $a_0 = a$, $b_0 = b$.

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 - ▶ Let $c_k = (a_{k-1} + b_{k-1})/2$ and $f_k = f(c_k)$.
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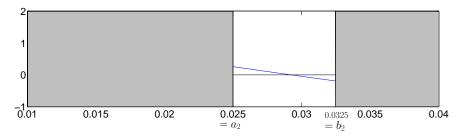


Input: f, a, b, k_{max} , ϵ_X , ϵ_f . 1. Set $a_0 = a$, $b_0 = b$.

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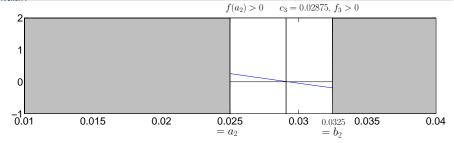


Input: f, a, b, k_{max} , ϵ_{x} , ϵ_{f} . 1. Set $a_{0} = a$, $b_{0} = b$.

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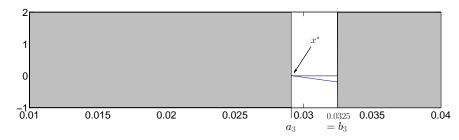
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- 2 De fer /c 1
- 2. Do for $k = 1, ..., k_{max}$
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 - ▶ If $|b_k a_k| < \epsilon_x$ or $|f(c_k)| < \epsilon_f$ break.
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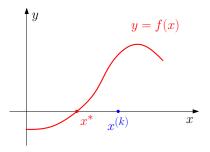
Input: f, a, b, k_{max} , ϵ_{x} , ϵ_{f} . 1. Set $a_{0} = a$, $b_{0} = b$.

- 2. Do for $k = 1, \ldots, k_{\text{max}}$
 - Let $c_k = (a_{k-1} + b_{k-1})/2$ and $f_k = f(c_k)$.
 - If $f_k f(a_{k-1}) > 0$ then let $a_k = c_k$ and $b_k = b_{k-1}$
 - else let $a_k = a_{k-1}$ and $b_k = c_k$. If $|b_k - a_k| < \epsilon_x$ or $|f(c_k)| < \epsilon_f$ break.
- 3. Return $x^* = c_k$ and $|b_k a_k|$.

For $\epsilon_X = \epsilon_f = 0.001$, $X^* \approx 0.0296875$ with error 9.375×10^{-4} .

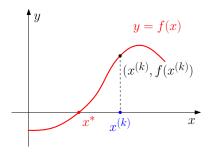


Start with $x^{(k)}$ (intended to approximate x^* such that $f(x^*) = 0$)



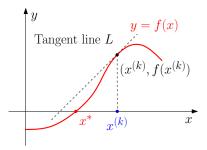


- Start with $x^{(k)}$ (intended to approximate x^* such that $f(x^*) = 0$)
- ightharpoonup Evaluate f at $x^{(k)}$



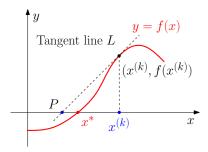


- Start with $x^{(k)}$ (intended to approximate x^* such that $f(x^*) = 0$)
- ightharpoonup Evaluate f at $x^{(k)}$
- ► Extend tangent line *L* from $(x^{(k)}, f(x^{(k)}))$ (using $f'(x^{(k)})$)





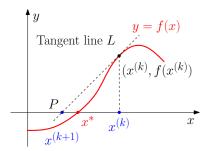
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- Extend tangent line L from $(x^{(k)}, f(x^{(k)}))$ (using $f'(x^{(k)})$)
- ► Follow *L* to *P* (where it cuts *x*-axis)







- Start with $x^{(k)}$ (intended to approximate x^* such that $f(x^*) = 0$)
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- Extend tangent line L from $(x^{(k)}, f(x^{(k)}))$ (using $f'(x^{(k)})$)
- ► Follow *L* to *P* (where it cuts *x*-axis)
- \triangleright $x^{(k+1)}$ defined as x-coordinate of point P







► Slope of *L* is $f'(x^{(k)})$ & *x*-intercept is $x^{(k+1)}$, so

$$f'(x^{(k)}) = \frac{f(x^{(k)}) - 0}{x^{(k)} - x^{(k+1)}}.$$

▶ Solving for $x^{(k+1)}$ gives the formula for Newton's method.

$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f'(x^{(k)})} \qquad (k \ge 0)$$

► Hopefully, $x^{(k+1)}$ closer to true zero x^* than $x^{(k)}$.



Newton's method

Given an iterate $x^{(k)}$ approximating a zero of f, the next iterate is

$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f'(x^{(k)})} \qquad (k \ge 0)$$

- Iterative procedure to locate zeros of f.
- ▶ Requires initial iterate $x^{(0)}$ to start.
- Near true zero x^* of f, iteration converges quickly.



Algorithm for Newton's method

Input:
$$f, f', x^{(0)}$$
 for $k = 0, 1, 2, \ldots$ until convergence $r^{(k)} \leftarrow f\left(x^{(k)}\right)$ (evaluate nonlinear residual) $\delta x^{(k)} \leftarrow -\left[f'\left(x^{(k)}\right)\right]^{-1} r^{(k)}$ (compute Newton step) $x^{(k+1)} \leftarrow x^{(k)} + \delta x^{(k)}$ (compute next iterate) Test for convergence (break if necessary) end for Output: $x^{(k)}$

- ► Terminology: $r^{(k)} := f(x^{(k)}) =$ residual
- ► Terminology: $\delta x^{(k)} := -\left[f'\left(x^{(k)}\right)\right]^{-1} r^{(k)} = \text{Newton step}$



Example:

Carry out Newton's method starting from $x^{(0)} = 0.75$ to find $x^{(3)}$ that approximates the real solution of $x = \cos x$.

▶ Define $f(x) = x - \cos x$, so $f'(x) = 1 + \sin x$





Example:

- Then, Newton's method is $x^{(k+1)} = x^{(k)} \frac{f(x^{(k)})}{f'(x^{(k)})}$, or, in steps,

$$r^{(k)} = f(x^{(k)}) = x^{(k)} - \cos(x^{(k)}),$$

$$\delta x^{(k)} = \frac{-r^{(k)}}{f'(x^{(k)})} = \frac{-r^{(k)}}{1 + \sin(x^{(k)})},$$

$$x^{(k+1)} = x^{(k)} + \delta x^{(k)}$$



Example:

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$$x^{(k+1)} = x^{(k)} + \delta x^{(k)}$$

▶ Use formulas above 3 times starting from k = 0, $x^{(0)} = 0.75$ to obtain sequence $x^{(1)}, x^{(2)}, x^{(3)}$ (see following table)



$$f(x) = x - \cos x \qquad f'(x) = 1 + \sin x$$

$$\begin{array}{c|ccc} k & x^{(k)} & r^{(k)} & \delta x^{(k)} \\ \hline 0 & 0.75 & & & & \\ \end{array}$$



$$x^{(0)} = 0.75,$$

 $r^{(0)} = f(x^{(0)}) = x^{(0)} - \cos(x^{(0)}) = 0.75 - \cos(0.75) = 1.83111 \times 10^{-2},$





$$x^{(0)} = 0.75,$$

$$r^{(0)} = f(x^{(0)}) = x^{(0)} - \cos(x^{(0)}) = 0.75 - \cos(0.75) = 1.83111 \times 10^{-2},$$

$$\delta x^{(0)} = -\frac{r^{(0)}}{f'(x^{(0)})} = -\frac{1.83111 \times 10^{-2}}{1 + \sin(0.75)} = -1.08889 \times 10^{-2},$$





Carry out Newton's method starting from $x^{(0)} = 0.75$ to find $x^{(3)}$ that approximates the real solution of $x = \cos x$.

$$x^{(0)} = 0.75,$$

$$r^{(0)} = f(x^{(0)}) = x^{(0)} - \cos(x^{(0)}) = 0.75 - \cos(0.75) = 1.83111 \times 10^{-2},$$

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$$x^{(1)} = x^{(0)} + \delta x^{(0)} = 0.75 + (-1.08889 \times 10^{-2}) = 0.739111138752579$$

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$$x^{(1)} = 0.739111138752579,$$

 $r^{(1)} = f(x^{(1)}) = 0.739111 - \cos(0.739111) = 4.35234 \times 10^{-5},$





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 $r^{(1)} = f(x^{(1)}) = 0.739111 - \cos(0.739111) = 4.35234 \times 10^{-5},$
 $\delta x^{(1)} = -\frac{r^{(1)}}{f'(x^{(1)})} = -\frac{4.35234 \times 10^{-5}}{1 + \sin(0.739111)} = -2.60055 \times 10^{-5},$





Carry out Newton's method starting from $x^{(0)} = 0.75$ to find $x^{(3)}$ that approximates the real solution of $x = \cos x$.

$$x^{(1)} = 0.739111138752579,$$

$$r^{(1)} = f(x^{(1)}) = 0.739111 - \cos(0.739111) = 4.35234 \times 10^{-5},$$

$$\delta x^{(1)} = -\frac{r^{(1)}}{f'(x^{(1)})} = -\frac{4.35234 \times 10^{-5}}{1 + \sin(0.739111)} = -2.60055 \times 10^{-5},$$

$$x^{(2)} = x^{(1)} + \delta x^{(1)} = 0.739111 + (-2.60055 \times 10^{-5}) = 0.73908513336448$$



$$f(x) = x - \cos x \qquad f'(x) = 1 + \sin x$$

$$k \qquad x^{(k)} \qquad r^{(k)} \qquad \delta x^{(k)}$$

$$0 \qquad 0.75 \qquad 1.83111 \times 10^{-2} \qquad -1.08889 \times 10^{-2}$$

$$1 \qquad 0.739111138752579 \qquad 4.35234 \times 10^{-5} \qquad -2.60055 \times 10^{-5}$$

$$2 \qquad 0.739085133364485 \qquad 2.49910 \times 10^{-10}$$

$$x^{(2)} = 0.739085133364485,$$

 $r^{(2)} = f(x^{(2)}) = 0.739085 - \cos(0.739085) = 2.49910 \times 10^{-10},$





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Carry out Newton's method starting from $x^{(0)} = 0.75$ to find $x^{(3)}$ that approximates the real solution of $x = \cos x$.

$$x^{(2)} = 0.739085133364485,$$

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$$\delta x^{(2)} = -\frac{r^{(2)}}{f'(x^{(2)})} = -\frac{2.49910 \times 10^{-10}}{1 + \sin(0.739085)} = -1.49324 \times 10^{-10},$$

$$x^{(3)} = x^{(2)} + \delta x^{(2)} = 0.739085 + (-1.49324 \times 10^{-10}) = 0.73908513321519$$



Example 2:

Carry out Newton's method starting from $x^{(0)} = 0.5$ to find $x^{(3)}$ that approximates a zero of the equation $xe^x = 2$.

► Define $g(x) = x \exp(x) - 2$, so $g'(x) = (x + 1) \exp(x)$



- ▶ Define $g(x) = x \exp(x) 2$, so $g'(x) = (x + 1) \exp(x)$
- ► Then, Newton's method is $x^{(k+1)} = x^{(k)} \frac{g(x^{(k)})}{g'(x^{(k)})}$, or, in steps,

$$r^{(k)} = g(x^{(k)}) = x^{(k)} \exp(x^{(k)}) - 2,$$

$$\delta x^{(k)} = \frac{-r^{(k)}}{g'(x^{(k)})} = \frac{-r^{(k)}}{(x^{(k)} + 1) \exp(x^{(k)})},$$

$$x^{(k+1)} = x^{(k)} + \delta x^{(k)}$$



Carry out Newton's method starting from $x^{(0)} = 0.5$ to find $x^{(3)}$ that approximates a zero of the equation $xe^x = 2$.

- ▶ Define $g(x) = x \exp(x) 2$, so $g'(x) = (x + 1) \exp(x)$
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$$r^{(k)} = g(x^{(k)}) = x^{(k)} \exp(x^{(k)}) - 2,$$

$$\delta x^{(k)} = \frac{-r^{(k)}}{g'(x^{(k)})} = \frac{-r^{(k)}}{(x^{(k)} + 1) \exp(x^{(k)})},$$

$$x^{(k+1)} = x^{(k)} + \delta x^{(k)}$$

▶ Use formulas above 3 times starting from k = 0, $x^{(0)} = 0.5$ to obtain sequence $x^{(1)}, x^{(2)}, x^{(3)}$ (see following table)





$$g(x) = x \exp(x) - 2$$
 $g'(x) = (x + 1) \exp(x)$
 $x^{(k)}$ $x^{(k)}$



$$g(x) = x \exp(x) - 2$$
 $g'(x) = (x + 1) \exp(x)$
 $x^{(k)}$ $r^{(k)}$ $\delta x^{(k)}$
 0 0.5 -1.17564

$$x^{(0)} = 0.5,$$

 $r^{(0)} = g(x^{(0)}) = x^{(0)} \exp(x^{(0)}) - 2 = 0.5 \exp(0.5) - 2 = -1.17564,$





$$g(x) = x \exp(x) - 2$$
 $g'(x) = (x + 1) \exp(x)$
 $x^{(k)}$ x

$$x^{(0)} = 0.5,$$

 $r^{(0)} = g(x^{(0)}) = x^{(0)} \exp(x^{(0)}) - 2 = 0.5 \exp(0.5) - 2 = -1.17564,$
 $\delta x^{(0)} = -\frac{r^{(0)}}{g'(x^{(0)})} = -\frac{(-1.17564)}{(0.5 + 1) \exp(0.5)} = 0.475374,$





Carry out Newton's method starting from $x^{(0)} = 0.5$ to find $x^{(3)}$ that approximates a zero of the equation $xe^x = 2$.

$$g(x) = x \exp(x) - 2$$
 $g'(x) = (x+1) \exp(x)$
 $k \mid x^{(k)} \mid r^{(k)} \mid \delta x^{(k)}$
 $0 \mid 0.5 \mid -1.17564 \mid 0.475374$
 $1 \mid 0.975374212950178$

$$x^{(0)} = 0.5,$$

$$r^{(0)} = g(x^{(0)}) = x^{(0)} \exp(x^{(0)}) - 2 = 0.5 \exp(0.5) - 2 = -1.17564,$$

$$\delta x^{(0)} = -\frac{r^{(0)}}{g'(x^{(0)})} = -\frac{(-1.17564)}{(0.5+1) \exp(0.5)} = 0.475374,$$

$$x^{(1)} = x^{(0)} + \delta x^{(0)} = 0.5 + (0.475374) = 0.975374212950178$$

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$$g(x) = x \exp(x) - 2 \qquad g'(x) = (x+1) \exp(x)$$

$$\begin{array}{c|c|c|c} k & x^{(k)} & r^{(k)} & \delta x^{(k)} \\ \hline 0 & 0.5 & -1.17564 & 0.475374 \\ 1 & 0.975374212950178 & 0.586848 & 0.586848 \end{array}$$

$$x^{(1)} = 0.975374212950178,$$

 $r^{(1)} = g(x^{(1)}) = 0.975374 \exp(0.975374) - 2 = 0.586848,$





$$g(x) = x \exp(x) - 2$$
 $g'(x) = (x + 1) \exp(x)$
 $x^{(k)}$ x

$$x^{(1)} = 0.975374212950178,$$

 $r^{(1)} = g(x^{(1)}) = 0.975374 \exp(0.975374) - 2 = 0.586848,$
 $\delta x^{(1)} = -\frac{r^{(1)}}{g'(x^{(1)})} = -\frac{0.586848}{(0.975374 + 1)\exp(0.975374)} = -0.112015,$



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Example 2:

0.863359106097814

Carry out Newton's method starting from $x^{(0)} = 0.5$ to find $x^{(3)}$ that approximates a zero of the equation $xe^x = 2$.

$$x^{(1)} = 0.975374212950178,$$

$$r^{(1)} = g(x^{(1)}) = 0.975374 \exp(0.975374) - 2 = 0.586848,$$

$$\delta x^{(1)} = -\frac{r^{(1)}}{g'(x^{(1)})} = -\frac{0.586848}{(0.975374 + 1)\exp(0.975374)} = -0.112015,$$

$$x^{(2)} = x^{(1)} + \delta x^{(1)} = 0.975374212950178 + -0.112015 = 0.863359106097.$$

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$$g(x) = x \exp(x) - 2$$
 $g'(x) = (x+1) \exp(x)$
 $\begin{array}{c|cccc} k & x^{(k)} & r^{(k)} & \delta x^{(k)} \\ \hline 0 & 0.5 & -1.17564 & 0.475374 \\ \hline \end{array}$

| ^ | A · · | , · · | 0 % |
|---|-------------------|--------------------------|-----------|
| 0 | 0.5 | -1.17564 | 0.475374 |
| 1 | 0.975374212950178 | 0.586848 | -0.112015 |
| 2 | 0.863359106097814 | 4.71213×10^{-2} | |
| | | | |
| | | | |

$$x^{(2)} = 0.863359106097814,$$

$$r^{(2)} = g(x^{(2)}) = 0.863359 \exp(0.863359) - 2 = 4.71213 \times 10^{-2}$$





$$\begin{split} x^{(2)} &= 0.863359106097814, \\ r^{(2)} &= g(x^{(2)}) = 0.863359 \exp(0.863359) - 2 = 4.71213 \times 10^{-2}, \\ \delta x^{(2)} &= -\frac{r^{(2)}}{g'(x^{(2)})} = -\frac{4.71213 \times 10^{-2}}{(0.863359 + 1) \exp(0.863359)} = -1.06652 \times 10^{-2}, \end{split}$$



Carry out Newton's method starting from $x^{(0)} = 0.5$ to find $x^{(3)}$ that approximates a zero of the equation $xe^x = 2$.

$$x^{(2)} = 0.863359106097814,$$

$$r^{(2)} = g(x^{(2)}) = 0.863359 \exp(0.863359) - 2 = 4.71213 \times 10^{-2},$$

$$\delta x^{(2)} = -\frac{r^{(2)}}{g'(x^{(2)})} = -\frac{4.71213 \times 10^{-2}}{(0.863359 + 1) \exp(0.863359)} = -1.06652 \times 10^{-2},$$

$$x^{(3)} = x^{(2)} + \delta x^{(2)} = 0.863359106097814 + (-1.06652 \times 10^{-2}) = 0.85269$$

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Now we have to answer the three essential questions:

- 1. Under what conditions does the algorithm converge?
- 2. How accurate will the result be?
- 3. How fast does it converge?

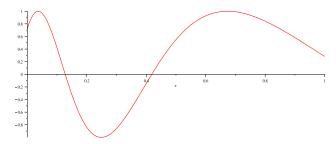
Since bisection and Newton iterations serve the same pupose (find x^* such that $f(x^*) = 0$) we can compare them...



1. Under what conditions does the algorithm converge?

Bisection converges to some x^* such that $f(x^*) = 0$ in [a, b], if f is continuous and f(a)f(b) < 0.

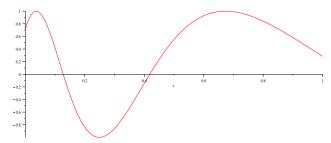
If there are two or more solutions, we don't know to which one it will converge.





1. Under what conditions does the algorithm converge?

Newton iteration converges if x_0 is *sufficiently close* to x^* . Usually, we do not know a priori how close is close enough and we must resort to trial and error. . . .





2. How accurate will the result be?

Both methods can give us x^* up to machine precision.

2. How accurate will the result be?

Both methods can give us x^* up to machine precision.

3. How fast does it converge?

In bisection, the error $|x^* - x^{(k)}|$ decreases by a factor of 1/2 in each iteration.

In Newton iterations, the error is approximately squared in each iteration (provided it is small enough!).

$$\epsilon_0, \frac{\epsilon_0}{2}, \frac{\epsilon_0}{4}, \frac{\epsilon_0}{8}, \dots$$
 vs. $\epsilon_0, \epsilon_0^2, \epsilon_0^4, \epsilon_0^8, \dots$





Newton's method converges very quickly, but requires the computation of f'(x).

Sometimes, we cannot compute it, for instance if f is shorthand for some complicated procedure:



In that case, we have two options:



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In that case, we have two options:

1. bisection, or





Newton's method converges very quickly, but requires the computation of f'(x).

Sometimes, we cannot compute it, for instance if f is shorthand for some complicated procedure:



In that case, we have two options:

- 1. bisection, or
- 2. the secant method.

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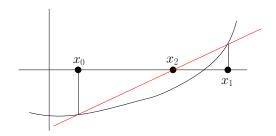


Suppose we have *two* intial points, x_0 and x_1 . Then we can *estimate* the derivative of f at x_1 as

$$f'(x_1) \approx \frac{f(x_1) - f(x_0)}{x_1 - x_0}$$

and substitute this in the Newton iteration:

$$x_2 = x_1 - \frac{f(x_1)}{f'(x_1)} \approx x_1 - \frac{f(x_1)(x_1 - x_0)}{f(x_1) - f(x_0)}$$





Then we iterate to find:

Secant method

Given iterates $x^{(k)}$ and $x^{(k-1)}$ approximating a zero of f, compute

$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)}) \left[x^{(k)} - x^{(k-1)} \right]}{f(x^{(k)}) - f(x^{(k-1)})} \quad (k \ge 1)$$

Secant methods needs *two* initial guesses: $x^{(0)}$ and $x^{(1)}$



Some remarks:

- ▶ This method uses a *finite difference* approximation to f'.
- Asymptotically (meaning if $|x_k x^*|$ is small enough) the secant method converges as fast as Newton's method does.
- The secant method has extensions to problems with more than 1 unknown, but in this case Newton's method tends to be less cumbersome.
- The secant method is a second order recurrence relation. It relates the next approximation to the two previous approximations.
- If we can find an a and b such that $x^* \in [a, b]$, then $x_0 = a$ and $x_1 = b$ is a good starting point.



Recurrence and iteration really mean procedures in which we repeat the same action over and over.

One way to program this is by using for and while loops. We can also make the recurrent nature of the computation explicit by *making the function call itself*. This is called *recursive* programming.

Simple example:

```
def fact(k):
    if k == 1:
        return 1
    else:
        return fact(k-1) * k
```

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