
§4.1 Roots of Equations

- Given $f(x)$, determine values of x for which $f(x)=0$; these values of x are called roots or zeros of $f(x)$.
- Roots may be real or complex numbers but in this course we focus on computing real roots.
- How many real roots may $f(x)$ have?

Roots of Equations

- A function $f(x)$ can have any number of real roots or none at all.
- Examples:
$$f(x) = \sin(x) - x = 0$$
$$f(x) = \tan(x) - x = 0$$
- Root-finding methods are iterative in nature, i.e., they require a starting point or guess.
- Common approach: bracket the root estimate.

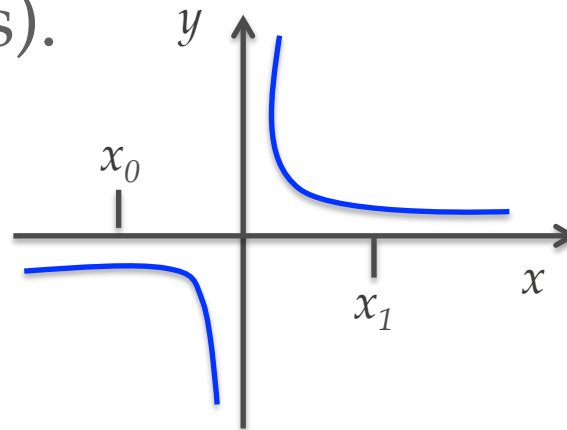
§4.2 Incremental Search Method

- Idea: if $f(x_1)$ and $f(x_2)$ have opposite signs, then there is at least one root in (x_1, x_2) .
- Continue to evaluate $f(x)$ at increments of x (say Δx) for which there is a sign change in $f(x)$.
- Caveats to this approach?

Incremental Search Method

- Possible to miss close roots if search increment Δx is larger than the root separation.
- Will not detect a double root.
- Singularities of $f(x)$ could mistakenly be taken as roots (e.g., asymptotes).

$f(x_0)$ is negative and
 $f(x_1)$ is positive but
 $x=0$ is not a root.



Incremental Search Method

- Review `rootsearch.py` on p. 147; searches for a zero of the user-supplied function $f(x)$ in the interval (a,b) .
- Returns bounds (x_1, x_2) of root or “None”; $dx = \Delta x$

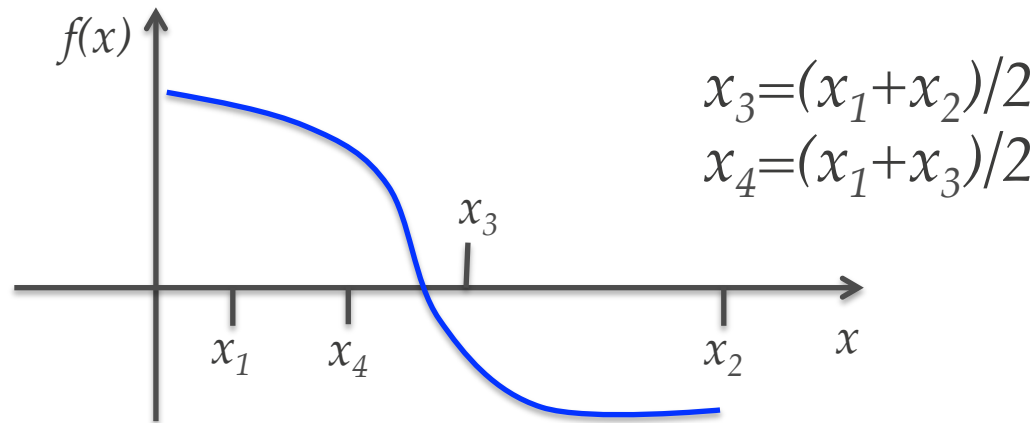
§4.3 Bisection

- Interval halving method that is very reliable (but not very fast).
- Idea: If there is a root in (x_1, x_2) , then $f(x_1) \times f(x_2) < 0$. Compute $f(x_3)$, where $x_3 = (x_1 + x_2) / 2$ and test if $f(x_1) \times f(x_3) < 0$?
- Repeat until the current interval (bounding the root) has a very small length, i.e., $|x_2 - x_1| \leq \varepsilon$.

Bisection

- Number of iterations to converge given an original interval (a,b) is $n = \frac{\ln(|\Delta x|/\epsilon)}{\ln 2}$, $\Delta x = b - a$.

See **bisect.py**
on p.149 and
Example 4.3 code
on pp. 150-151.



§4.5 Newton-Raphson Method

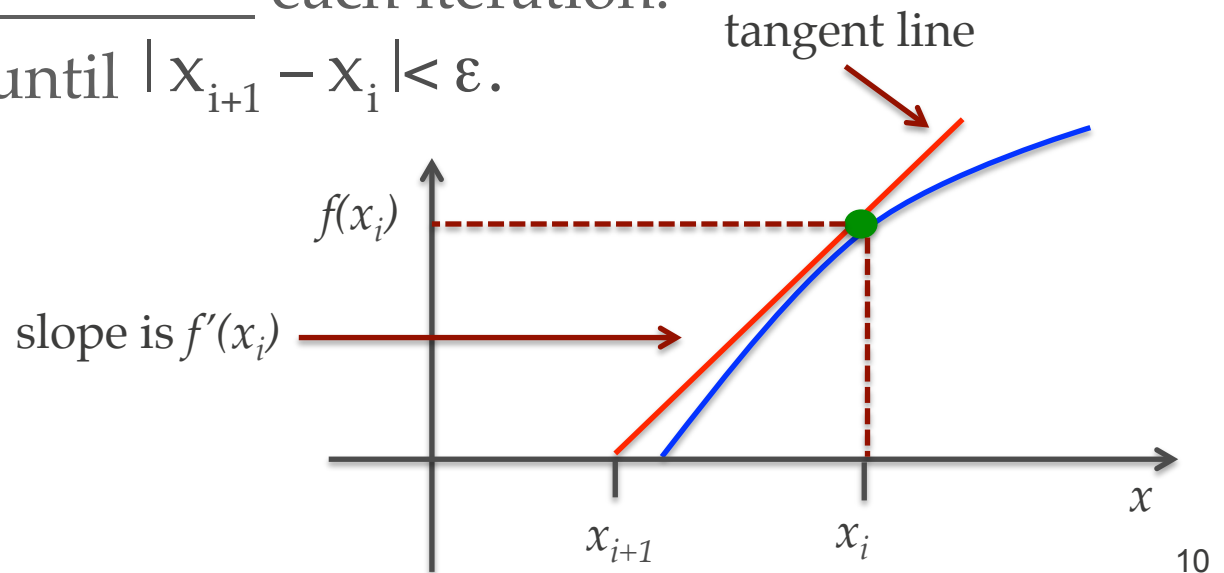
- Best method for root-finding: simple and fast.
- Caveat: requires the derivative $f'(x)$ as well as the function $f(x)$.
- Derivation is from Taylor Series expansion of $f(x)$ about x : $f(x_{i+1}) = f(x_i) + f'(x_i)(x_{i+1} - x_i) + O(x_{i+1} - x_i)^2$.
- If x_{i+1} is a root of $f(x)$ then the equation above becomes: $0 = f(x_i) + f'(x_i)(x_{i+1} - x_i) + O(x_{i+1} - x_i)^2$.

Newton-Raphson Method

- Assuming x_i is close to x_{i+1} , we can simplify the equation further: $0 = f(x_i) + f'(x_i)(x_{i+1} - x_i)$, or $0 = f(x_i) + f'(x_i)x_{i+1} - f'(x_i)x_i$.
- Newton-Raphson Formula: $x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$.
- If x is the exact root that x_i approximates, then the error in x_i is $E_i = x - x_i$, and we can represent the error in x_{i+1} as ...

Newton-Raphson Method

- Error in x_{i+1} is defined by: $E_{i+1} = \frac{f''(x_i)}{2f'(x_i)} E_i^2$.
Converges quadratically, i.e.,
the number of significant digits in the current
approximation _____ each iteration.
Typically iterate until $|x_{i+1} - x_i| < \epsilon$.



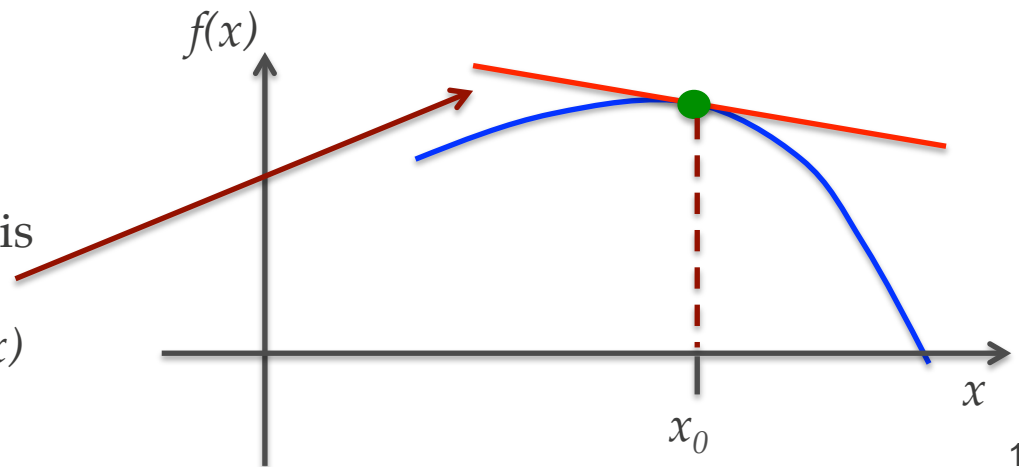
Newton-Raphson Method

- Algorithm:

1. Determine initial guess x for root of $f(x)=0$;
2. Compute $\Delta x = f(x) / f'(x)$;
3. Set $x = x + \Delta x$, repeat Steps 2 and 3 until $|\Delta x| < \epsilon$.

Case when N-R diverges
(does not converge):

Tangent line is
not a good
approx. to $f(x)$

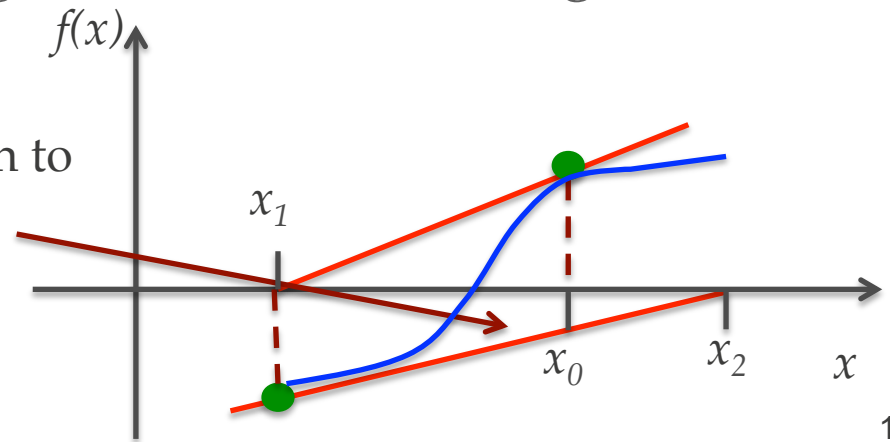


Newton-Raphson Method

- See `newtonRaphson.py` on pp. 158-159 in textbook. Assumes an initial interval (a,b) for the root; if it *wanders* outside of (a,b) , the bisection method is used to get closer to root (hybrid method).

Another case when N-R diverges (does not converge):

If x_0 is not close enough to the root, N-R may not converge (globally).



Newton-Raphson Method

- See also Example 4.7 (`example4_7.py`) on pp. 159-160; find the smallest zero of

$$f(x)=x^4-6.4x^3+6.45x^2+20.538x-31.752$$

(should be 2.1)

§4.6 Systems of Equations

- Consider n simultaneous nonlinear equations:

$$\vec{f}(\vec{x}) = \vec{0} \quad \left\{ \begin{array}{l} f_1(x_1, x_2, \dots, x_n) = 0 \\ f_2(x_1, x_2, \dots, x_n) = 0 \\ \vdots \\ f_n(x_1, x_2, \dots, x_n) = 0 \end{array} \right.$$

Need to extend N-R to solve for $\vec{x}^T = (x_1, x_2, \dots, x_n)$.

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Systems of Equations

- Taylor Series expansion of $f_i(\bar{x})$ about \bar{x} is

$$f_i(\bar{x} + \Delta\bar{x}) = f_i(\bar{x}) + \sum_{j=1}^n \frac{\partial f_i}{\partial x_j} \Delta\bar{x}_j + O(\Delta\bar{x}^2).$$

Dropping the last term yields:

$$f_i(\bar{x} + \Delta\bar{x}) = f_i(\bar{x}) + J(\bar{x})\Delta\bar{x},$$

where $J(\bar{x})$, $J_{ij} = \frac{\partial f_i}{\partial x_j}$, is the n -by- n Jacobian matrix.

Systems of Equations

- Assume that \bar{x} is the current approx. to the solution of $\vec{f}(\bar{x}) = \vec{0}$, and let $\bar{x} + \Delta\bar{x}$ be an improved approximation; how do we determine $\Delta\bar{x}$?

$$\text{Set } \vec{f}(\bar{x} + \Delta\bar{x}) = \vec{0} \rightarrow J(\bar{x})\Delta\bar{x} = -\vec{f}(\bar{x}).$$

Systems of Equations

- N-R Method for solving a nonlinear system of equations:

1. Estimate $\bar{\mathbf{x}}$
2. Evaluate $\vec{f}(\bar{\mathbf{x}})$
3. Compute $J(\bar{\mathbf{x}})$
4. Solve $J(\bar{\mathbf{x}})\Delta\bar{\mathbf{x}} = -\vec{f}(\bar{\mathbf{x}})$ for $\Delta\bar{\mathbf{x}}$
5. Set $\bar{\mathbf{x}} = \bar{\mathbf{x}} + \Delta\bar{\mathbf{x}}$ and repeat Steps 2—5.

Stopping criterion: $\|\Delta\bar{\mathbf{x}}\|_2 < \varepsilon$.

Systems of Equations

- What happens if $\frac{\partial f_i}{\partial x_j}$ is difficult to impractical to compute?

Can approximate it using finite differences:

$$\frac{\partial f_i}{\partial x_j} \approx \frac{f_i(\bar{x} + \bar{e}_j h) - f_i(\bar{x})}{h}, \bar{e}_j = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ 0 \end{bmatrix} \leftarrow \text{jth component}$$

Systems of Equations

- See **newtonRaphson2.py** on p. 163 for code to implement the N-R method for nonlinear systems of equations; for an example of its use see the **Example4_9.py** code (pp. 165-166).

Two interesting applications of the N-R method are found in Problems 16 (p. 167, structural mechanics) and 26 (pp. 170-171, circle drawing).