

Lecture 23 Outline

Friday, November 15, 2019 11:53 AM

IVP + BVP

Root Finding

Bisection ; Regula Falsi ; Newton-Raphson ;
Secant Method ; Fixed Point

Many PDE problems are IVP + BVP

Example: Time-evolving heat problem (diffusion)

$$\frac{\partial u}{\partial t} = \alpha \nabla^2 u = 0, \quad \alpha > 0$$

Describes how the temperature $u(x, t)$ evolves over time due to conduction

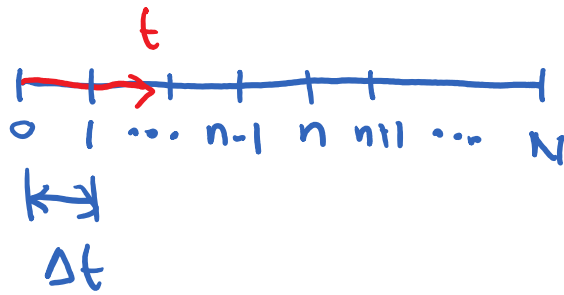
Need to combine IVP and BVP concepts

1D grid over space



1D time points





$$u(t_n, x_i) = u_i^n \quad \text{time step; not power}$$

$$\frac{\partial u}{\partial t} - \alpha^2 \frac{\partial^2 u}{\partial x^2} = 0$$

$$\frac{u_i^{n+1} - u_i^n}{\Delta t} - \alpha^2 \frac{u_{i+1}^n - 2u_i^n + u_{i-1}^n}{(\Delta x)^2} = 0$$

Forward
Euler
in time

Central difference
in space at t_n

$$\frac{u_i^{n+1} - u_i^n}{\Delta t} - \alpha^2 \frac{u_{i+1}^{n+1} - 2u_i^{n+1} + u_{i-1}^{n+1}}{(\Delta x)^2} = 0$$

Backward
Euler
in time

Central difference
in space at t_{n+1}

Many other schemes are possible

Example:

$$\text{Let } \frac{du}{dt} = F\left(x, t, u, \frac{\partial u}{\partial x}, \frac{\partial^2 u}{\partial x^2}, \dots\right)$$

$$\text{Denote } F_i^n = F\left(x_i, t_n, u_i^n, \frac{\partial u_i^n}{\partial x}, \frac{\partial^2 u_i^n}{\partial x^2}, \dots\right)$$

$$\begin{array}{l} \text{Forward} \\ \text{Euler} \end{array} \quad \frac{u_i^{n+1} - u_i^n}{\Delta t} = F_i^n \quad \mathcal{O}(\Delta t)$$

$$\begin{array}{l} \text{Backward} \\ \text{Euler} \end{array} \quad \frac{u_i^{n+1} - u_i^n}{\Delta t} = F_i^{n+1} \quad \mathcal{O}(\Delta t)$$

$$\begin{array}{l} \text{Crank-} \\ \text{Nicholson} \end{array} \quad \frac{u_i^{n+1} - u_i^n}{\Delta t} = \frac{1}{2} \left(F_i^n + F_i^{n+1} \right) \quad \mathcal{O}(\Delta t^2)$$

Nonlinear equations of one variable

Recall that $g(x)$ is a linear function

$$\text{if } g(x) + g(y) = g(x+y)$$

$$\text{Example: } g(x) = ax, \quad a \in \mathbb{R}$$

$$g(x) + g(y) = ax + ay = a(x+y) = g(x+y)$$

\therefore linear

$$\text{Example: } g(x) = ax^2, \quad a \in \mathbb{R}$$

$$g(x) + g(y) = ax^2 + ay^2 \neq a(x+y)^2 = g(x+y)$$

\therefore non-linear

Consider next several methods for solving $g(x) = a$

Rewrite this as $f(x) = g(x) - a$

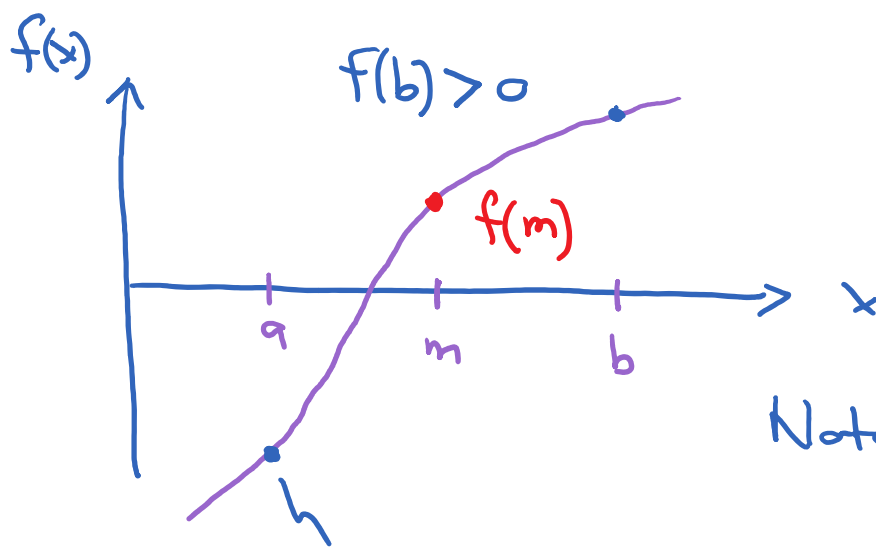
Then find the roots of $f(x) = 0$

- Methods :
- ① Bisection
 - ② Regula falsi
 - ③ Newton-Raphson
 - ④ Secant
 - ⑤ Fixed point

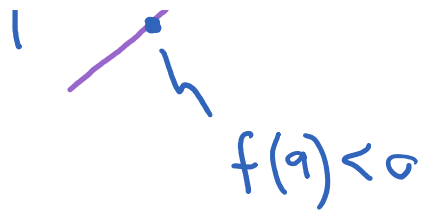
① Bisection Method: Requires that a zero exists in the range $[a, b]$

How can one check this requirement?

Simply require that $f(a)f(b) < 0$



Note: $f(a)f(b) < 0$



$$f(a) \cdot f(b) < 0$$

$$\text{Let } m = \frac{a+b}{2}$$

If $f(a)f(m) < 0$, redo
with range $[a, m]$;

otherwise redo with
range $[m, b]$

Keep going with these iterations until

$$|f(m)| < \epsilon$$

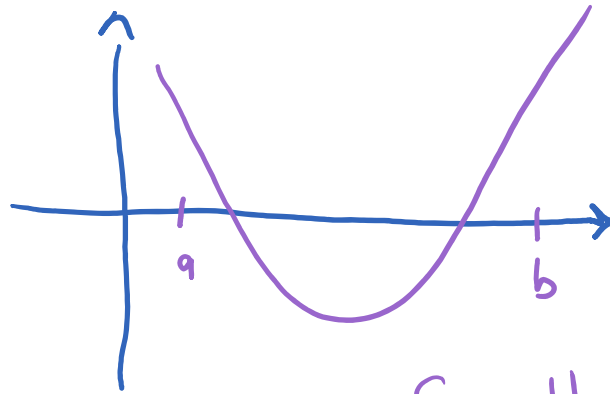
↙ error tolerance

Convergence is linear, because the bound
on the root decreases by $1/2$ for each iteration

Issue: One must have $f(a)f(b) < 0$ for
this scheme to work

→ If $f'(x) = 0$ anywhere in $[a, b]$,
then the root may be missed

then the root may be missed



Can these roots be found?

but if $f(a)f(b) < 0$ is true, then this approach will always find the root (perhaps slowly)

If there are multiple roots within $[a, b]$, then we have no idea which one will be chosen

② Regula falsi (rule of false position)

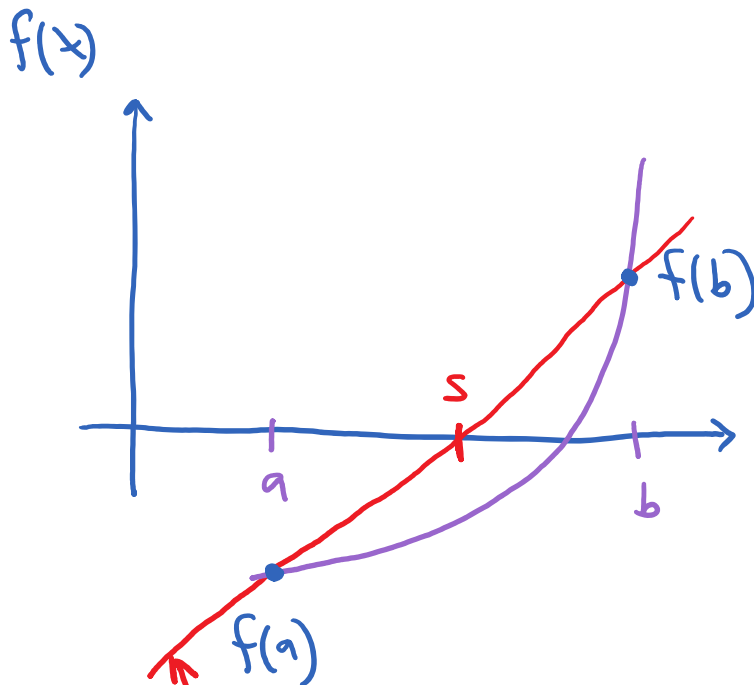
Modification of bisection method

Let $f(a)f(b) < 0$ for range $[a, b]$

Instead of checking $\left[a, \frac{a+b}{2}\right]$,

check whether $f(a)f(s) < 0$, where

$$s = b - \left(\frac{b-a}{f(b)-f(a)} \right) f(b)$$



$$y(x) = f(b) - \left(\frac{f(b)-f(a)}{b-a} \right) (b-x)$$

$$\text{Then, } y(s) = f(b) - \left(\frac{f(b)-f(a)}{b-a} \right) (b-s) = 0$$

$$(b-s) = \frac{b-a}{f(b)-f(a)} f(b)$$

$$s = b - \left(\frac{b-a}{f(b)-f(a)} \right) f(b) //$$

If $f(a)f(s) < 0$, then use $[a, s]$ for next iteration; otherwise, use $[s, b]$

Still has only linear convergence

If $f(a)f(b) < 0 \rightarrow$ Guaranteed to converge!

③ Newton-Raphson Method

Let x_1 be a point near the root x^*

Then, from Taylor series,

$$f(x) = f(x_1) + f'(x_1)(x - x_1) + \frac{1}{2}f''(x_1)(x - x_1)^2 + \dots$$

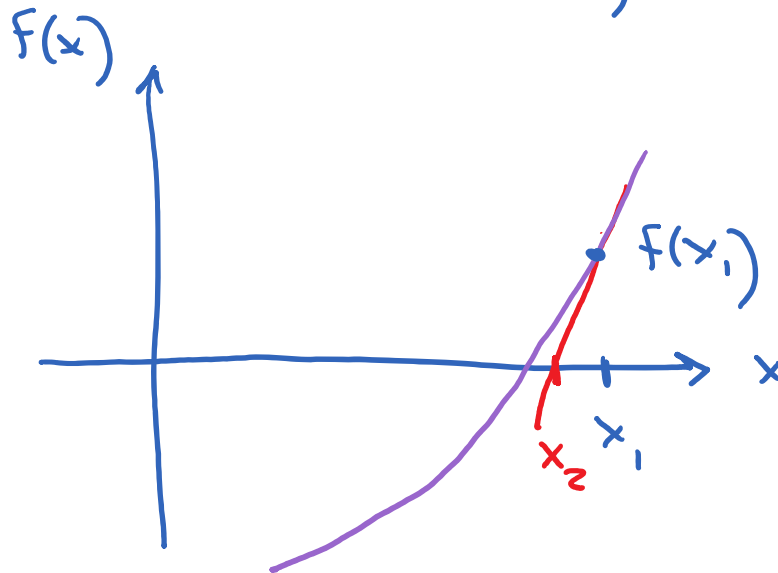
Approximate as

$$f(x) = f(x_1) + f'(x_1)(x - x_1) + O((x - x_1)^2)$$

and solve for approximation to root

$$f(x_2) = f(x_1) + f'(x_1)(x_2 - x_1) = 0$$

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



Each iteration is then

$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$$

Consider convergence characteristics

Let x^* be the true root and x_i, x_{i+1}

be estimates at iterations $i, i+1$, such that

$$|x^* - x_i| = \delta \ll 1$$

Define errors $e_i = x^* - x_i$, $e_{i+1} = x^* - x_{i+1}$

However, we know

$$f(x^*) = 0 = f(x_i) + f'(x_i)(x^* - x_i) + \frac{f''(\xi)}{2}(x^* - x_i)^2$$

for some $\xi \in (x^*, x_i)$ such that

$$\frac{f''(\xi)}{2}(x^* - x_i)^2 = \frac{f''(x_i)}{2}(x^* - x_i)^2 + \text{H.O.T.}$$

In Newton-Raphson

$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)} \Rightarrow f(x_i) = f'(x_i)(x_i - x_{i+1})$$

Then,

$$f(x^*) = 0 = f(x_i) + f'(x_i)(x^* - x_i) + \frac{f''(\xi)}{2}(x^* - x_i)^2$$

$$0 = f'(x_i)(x_i - x_{i+1}) + f'(x_i)(x^* - x_i) + \frac{f''(\xi)}{2}(x^* - x_i)^2$$

$$0 = f'(x_i)(x^* - x_{i+1}) + \frac{f''(\xi)}{2}(x^* - x_i)^2$$

$$0 = f'(x_i) e_{i+1} + \frac{f''(\xi)}{2} e_i^2$$

$$0 = f'(x_i) e_{i+1} + \frac{f''(\rho)}{2} e_i^2$$

$$\rightarrow e_{i+1} = - \frac{f''(\rho)}{2f'(x_i)} e_i^2$$

$$\rightarrow e_{i+1} \propto e_i^2 \rightarrow \text{2nd order convergence}$$

$$\text{If } e_i \sim 10^{-3}$$

$$e_{i+1} \sim 10^{-6}$$

$$e_{i+2} \sim 10^{-12}$$

Issues: The initial guess at the root, x_0 , must be "close" to x^*

This means that one must test for divergence, as well as convergence

So, convergence occurs if $|f(x_i)| < \varepsilon$

Divergence occurs, if

of iterations is large

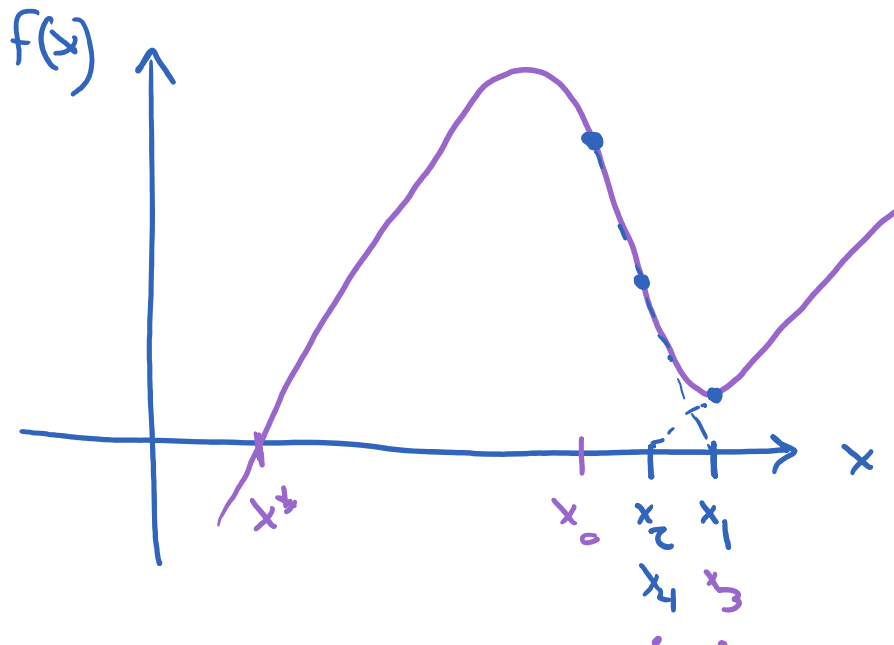
$$\text{or } |f'(x_i)| < \epsilon,$$

$$\text{or } |x_{i+1} - x_i| > \delta$$

Why check $|f'(x_i)| < \epsilon$?

$$\text{Because } x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$$

One may never converge if x_0 is too far from x^* or if one gets oscillations



$$\begin{matrix} x_1 & x_2 \\ \vdots & \vdots \\ x_n & x_{n+1} \end{matrix}$$

Final issue: $f'(x)$ might be difficult or expensive to evaluate

④ Secant Method: Tries to alleviate issue if $f'(x)$ is not known or expensive to evaluate

$$\text{Approximate } f'(x_i) \approx \frac{f(x_i) - f(x_{i-1})}{x_i - x_{i-1}}$$

Then,

$$x_{i+1} = x_i - \left(\frac{x_i - x_{i-1}}{f(x_i) - f(x_{i-1})} \right) f(x_i)$$

Rate of convergence is ≈ 1.62

Some of the same issues as Newton-Raphson

⑤ Fixed Point Method

A fixed point is one where

$$x = h(x), \quad \text{e.g., } y(x) = \sqrt{x}, \quad x = 1$$

Let $f(x)$ have a linear and nonlinear part

$$f(x) = x - g(x) \Rightarrow f(x) = 0 \text{ as}$$

$$x - g(x) = 0 \rightarrow x = g(x)$$

Iteration is then $x_{i+1} = g(x_i)$

Converges if $|x_{i+1} - g(x_{i+1})| < \varepsilon$

$$\text{or } |f(x_{i+1})| < \varepsilon$$

Convergence

Let x^* be the root, such that $x^* = g(x^*)$

$$\text{with } x_{i+1} = g(x_i)$$

$$\text{Then } x^* - x_{i+1} = g(x^*) - g(x_i)$$

$$\text{and } \frac{e_{i+1}}{e_i} = \frac{x^* - x_{i+1}}{x^* - x_i} = \frac{g(x^*) - g(x_i)}{x^* - x_i}$$

The mean value theorem of calculus states that if $g(x)$ is continuous over $[x^*, x_{i+1}]$, then there exists a $\xi \in [x^*, x_i]$ such that

$$g'(\xi) = \frac{g(x^*) - g(x_i)}{x^* - x_i}$$

$$\Rightarrow \frac{e_{i+1}}{e_i} = g'(\xi) \Rightarrow e_{i+1} = g'(\xi)e_i$$

$$\Rightarrow \text{Convergence occurs if } |g'(\xi)| < 1$$