

Final exam: Mon, Dec 11

5-7:45PM in normal room

Monday last day of  
class Monday:  
Review

11-29-17, Wed



## Difference equations: (Chapter 1)

We are interested in linear operators  $T: V \rightarrow V$ .

Example: Shift operator  $E$

$$Ex = (x_2, x_3, \dots), \quad x = (x_1, x_2, \dots).$$

$L$  is a polynomial in  $E$ .  $L = P(E)$ .

Theorem: If  $P$  is a polynomial and  $\lambda$  is a root of  $P$  then one solution of the difference equation

$$P(E)x = 0$$

is  $(\lambda, \lambda^2, \dots)$ . If all the roots of  $P$  are simple and non-zero then all the solutions of the difference equation are linear combinations of these solutions.

Example:

$$(E^2 - 3E + 2E^0)x = 0$$



$$P(\lambda) = \lambda^2 - 3\lambda + 2\lambda = 0.$$

Theorem 3: (on stable difference equations)

For a polynomial  $P$  satisfying  $p(0) \neq 0$ , these properties are equivalent.

① The difference equation

$$P(E)x = 0$$

is stable.

② All roots of  $P$  satisfy  $|\lambda| \leq 1$  and all multiple roots satisfy  $|\lambda| < 1$ .

Example:

$$\begin{cases} x_{n+1} = \frac{13}{3} x_n - \frac{4}{3} x_{n-1}, & n \geq 1 \\ x_0 = 1 \\ x_1 = \frac{1}{3} \end{cases}$$

Postulate:  $\text{soln} = \left(\frac{1}{3}\right)^n$ .

Proof: Use induction on  $n$ .

Because  $x_0 = 1 = \left(\frac{1}{3}\right)^0$  and  $x_1 = \frac{1}{3} = \left(\frac{1}{3}\right)^1$ .

Induction step:

$$x_{m+1} = \frac{13}{3} x_m - \frac{4}{3} x_{m-1}$$

$$\begin{aligned} x_{m+1} &= \frac{13}{3} \left(\frac{1}{3}\right)^m - \frac{4}{3} \left(\frac{1}{3}\right)^{m-1} = \\ &= \left(\frac{1}{3}\right)^{m-1} \left[ \frac{13}{9} - \frac{4}{3} \right] = \left(\frac{1}{3}\right)^{m-1} \left[ \frac{1}{9} \right] = \\ &= \left(\frac{1}{3}\right)^{m+1}. \quad \square \end{aligned}$$

On a 32-bit computer, we get

$$x_0 = 1.000000$$

$$x_1 = 0.333333 \quad 7 \text{ correct digits}$$

$$x_2 = 0.111112 \quad 6 \text{ correct digits}$$

$$x_3 = 0.0370373 \quad 5 \text{ correct digits}$$

$$x_4 = 0.0123456 \quad 4 \text{ correct digits (true: } 0.012345679)$$

⋮

$$x_7 = 0.0005131 \quad 1 \text{ correct digit (true } 4.57 \cdot 10^{-4})$$

$$x_8 = 0.0003757 \quad 0 \text{ correct digit (true } 1.52415 \cdot 10^{-4})$$

Algorithm is unstable.

Any error present in  $x_n$  is multiplied by  $\frac{13}{3}$  to get  $x_{n+1}$ . So an error in  $x_1$  propagates into  $x_{15}$  with a factor of  $(\frac{13}{3})^{14}$ .

Error of  $x_1 \approx 10^{-8}$ . So

$$(\frac{13}{3})^{14} \approx 823207590 \approx 1 \cdot 10^9.$$

This equation

$$x_{n+1} = \frac{13}{3}x_n - \frac{4}{3}x_{n-1}$$

is a difference equation. So

$$x_{n+1} - \frac{13}{3}x_n + \frac{4}{3}x_{n-1} = 0.$$

$$u_{k+1} = A u_k$$

$$\begin{bmatrix} x_{k+1} \\ x_k \end{bmatrix} = \begin{bmatrix} \frac{13}{3} & -\frac{4}{3} \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_k \\ x_{k-1} \end{bmatrix}$$

$$(\frac{13}{3} - \lambda)(-\lambda) + \frac{4}{3} = 0$$

$$\lambda^2 - \frac{13}{3}\lambda + \frac{4}{3} = 0.$$

$$(\lambda - \frac{1}{3})(\lambda - 4) = 0.$$

So  $\lambda = \frac{1}{3}, 4$ . So solution is a linear combination of powers of these 2 solutions. So general solution is

$$x_n = A(\frac{1}{3})^n + B(4)^n$$

where  $A, B$  depend on initial conditions  $x_0, x_1$ .

True solution is

$$x_n = (\frac{1}{3})^n \Rightarrow A = 1, B = 0.$$

But the difference equation solution gotten from this recurrence relation is contaminated by the term  $B(4)^n$ .

Eventually, in fact, the  $(4)^n$  term dominates the true solution  $(\frac{1}{3})^n$ .

Consider now

$$x_0 = 1, \text{ and } x_1 = 4.$$

Then the true solution is  $x_n = 4^n$ .

Proof:

Base case:  $x_0 = 1 = 4^0$ , and  $x_1 = 4 = 4^1$ .

Induction step:

$$\begin{aligned} x_{n+1} &= \frac{13}{3} (4)^n - \frac{4}{3} (4)^{n-1} = \\ &= 4^n \left[ \frac{13}{3} - \frac{1}{3} \right] = 4^n \left[ \frac{12}{3} \right] = 4^{n+1}. \quad \square \end{aligned}$$

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And the numerical results will be stable.

$$x_1 = 4.0000$$

$$x_{10} = 1.048576 \cdot 10^6 \text{ (true } 1048576)$$

Here there is still an error coming from  $(\frac{1}{3})^n$  but it is relatively small compared to true solution.

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Stability = "continuous dependence on initial data".

Back to multistep methods:

Midpoint method

$$y'(t) = f(t, y)$$

$$\int_{t_{n-1}}^{t_{n+1}} y'(t) dt = \int_{t_{n-1}}^{t_{n+1}} f(t, y) dt$$

$$y(t_{n+1}) - y(t_{n-1}) = (t_{n+1} - t_{n-1}) f(t_n, y_n)$$

Midpoint method

$$y(t_{n+1}) = y(t_{n-1}) + 2h f(t_n, y_n).$$

This is a multistep method because it uses more than one prior value of solution  $y$ .

Euler only used previous value. Using prior info leads to more accurate numerical solution.

This is an **explicit** method because we know all the values of  $y$  that we need to get new values  $y(t_{n+1})$ . The order of convergence globally is  $\mathcal{O}(h^2)$ . Local truncation error is  $\mathcal{O}(h^3)$ .

Lets look at model problem:

$$\begin{cases} y'(x) = \lambda x \\ y(0) = 1 \end{cases}$$

By separation of variables,

$$\begin{array}{l|l} \frac{dy}{y} = \lambda dx & y(0) = 1 \\ \ln y = \lambda x + C & C e^{\lambda(0)} = 1 \\ y = C e^{\lambda x} & C = 1. \\ & \boxed{y(x) = e^{\lambda x}} \end{array}$$

Now apply midpoint method for this problem. We get

$$y(x_{n+1}) = y(x_{n-1}) + 2h f(t_n, x_n)$$

$$y_{n+1} = y_{n-1} + 2h \lambda y_n.$$

This is a linear difference equation.

$$u_{k+1} = A u_k$$

$$\begin{bmatrix} y_{n+1} \\ y_n \end{bmatrix} = \begin{bmatrix} 2h\lambda & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} y_n \\ y_{n-1} \end{bmatrix}$$

$$\det \begin{bmatrix} 2h\lambda - \nu & 1 \\ 1 & -\nu \end{bmatrix} = 0$$

$$(2h\lambda - \nu)(-\nu) - 1 = 0.$$

$$\nu^2 - 2h\lambda\nu - 1 = 0$$

$$\text{roots: } \nu = \frac{2h\lambda \pm \sqrt{(2h\lambda)^2 + 4}}{2} = h\lambda \pm \sqrt{(h\lambda)^2 + 1}.$$

General solution to difference equation is

$$\textcircled{1} y_n = A \underbrace{(h\lambda + \sqrt{(h\lambda)^2 + 1})}_{V_0}^n + B \underbrace{(h\lambda - \sqrt{(h\lambda)^2 + 1})}_{V_1}^n, \quad n \geq 0.$$

Write instead

$$y_n = AV_0^n + BV_1^n.$$

Case 1:  $0 < \lambda < \infty$  (for all  $h$ ).

$$V_0 > |V_1| > 0$$

$$V_0 > 1.$$

So  $V_0^n$  dominates the solution.

Case 2:  $-\infty < \lambda < 0$

$$V_1 < -1, \quad h > 0$$

$$0 < V_0 < 1.$$

So in this case,  $V_1^n$  will eventually dominate the solution.

In fact the numerical method (midpoint method) is unstable for these values of  $\lambda$ . One of the eigenvalues has magnitude  $> 1$ .

Example: Consider the problem

$$y'(x) = \lambda y(x),$$

Let  $\lambda = -1$ . So,

$$\begin{cases} y'(x) = -y(x) \\ y(0) = 1 \end{cases}.$$

| $x_n$ | $y_n = \text{midpoint method solution}$ | error   |
|-------|---|---------|
| 0.25  | 0.7500                                  |         |
| 0.50  | 0.6250                                  | 0.0288  |
| 0.75  | 0.4375                                  | -0.0185 |
| 1.00  | 0.4063                                  | 0.0349  |
| 1.25  | 0.2344                                  | -0.0384 |
| 1.50  | 0.2891                                  | 0.0521  |
| ...   |   | -0.0659 |
| 2.25  | -0.0322                                 | 0.4376  |

Note that at  $x_n = 2.25$ , we get a negative value for solution to DE. But true solution is  $y = e^x$  which is never negative.

The trapezoid method is another **multistep method**.

$$y'(t) = f(t, y(t))$$

$$y(t_{n+1}) = y(t_n) + \int_{t_n}^{t_{n+1}} f(t, y(t)) dt$$

Apply trapezoid rule to the integral to get

$$y(t_{n+1}) = y(t_n) + \frac{h}{2} [f(t_n, y_n) + f(t_{n+1}, y_{n+1})].$$



This is a single-step implicit method.

The unknown solution  $y_{n+1} = y(t_{n+1})$   
appears on both sides of the method.

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