

Ch. 4 Stiff Equations

Key Points:

- Examples of stiff ODEs
- A central problem $y' = \lambda y$ for understanding numerical stability
 - Euler, Backward Euler, Trapezoid, ...
 - Runge-Kutta Methods
 - Multi-step Methods

... see also Heath, p. 401-404

K.E. Atkinson - "An Introduction to Numerical Analysis" (2nd Edition) ch 6

EXAMPLE

(aka Competitive Lotka-Volterra Equations)

Consider the 3-species Lotka-Volterra system

$$\begin{cases} \frac{dx}{dt} = ax - bxy \\ \frac{dy}{dt} = -cy + dx - eyz \\ \frac{dz}{dt} = -fz + gyz \end{cases} \quad a, b, c, d, e, f, g > 0$$

Ref: "A Lotka-Volterra Three-Species Food Chain"

(x, y, z) : (mouse, snake, owl)
 (vegetation, hare, lynx)
 (warren, robin, falcon)

E. Chauvet
 D. E. Paullet
 Z. Walls

I.C. $x(0) = x_0, y(0) = y_0, z(0) = z_0$

Math Mag.
 75 (4) Oct. 2002
 243-255

2 equilibrium solutions

$\frac{dx}{dt} = \frac{dy}{dt} = \frac{dz}{dt} = 0$. (assume $\frac{a}{b} \neq \frac{f}{g}$)

$(x, y, z) = (0, 0, 0)$
 $(\frac{f}{d}, \frac{a}{b}, 0)$

$x=0$
 $y = a/b$
 $c = dx + ez$
 $z=0$
 $x = \frac{f}{d}$

Jacobian:

$$J = \begin{bmatrix} a & -bx & 0 \\ dy & -c + dx - ez & -ey \\ 0 & gz & -f + gy \end{bmatrix}$$

Before trying to analyze this system let's just try to solve it numerically using Euler's Method...

See `euler_system.m` + `feuler_system.m`

Here I've set

$a = b = d = e = 1$

$c = 2 \quad f = 100 \quad g = 0.1$

$x_0 = 0.5$
$y_0 = 1$
$z_0 = 2$

↑
fast decay of y (snakes)

↑
very fast decay of z (owls)

mice grow slowly if left alone but... then the nonlinear terms start to matter...

Numerical Experiment #1

• set $t_F = 5$... increase n_k for 'sufficient resolution'

$n_k = 1000$ seems quite good $h = \frac{5}{1000} = 0.005$

observe: $z \rightarrow 0$ quickly

x, y : some dynamics

• let's run things longer ~~to~~ to observe the x, y dynamics

by increasing t_F but keeping n_k fixed (— not necessarily a good idea since this causes $h = \frac{t_F}{n_k}$ to increase)

• set $t_F = 10$... seems OK ... observe
 repeating motion in x, y ... $z \approx 0$ (no oscills)

• set $t_F = 20$... z behaving strangely ...
 (large t -oscillations)

Note: $\frac{dz}{dt} = -fz + gyz = -(f - gy)z$

$f = 100$ $y \in [0, 5]$
 $g = 0.1$

so $f - gy > 0$

why isn't z decaying
 rapidly to zero?

• set $t_F = 40$... code crashes...

∴ more reasonable-looking results restored by increasing n_k

($n_k = 4000, \dots$)

~~even~~ even here you should be asking if we
 are ~~restoring this solution~~
 correctly solving this problem.

Numerical Experiment #2

same a, b, c, d, e, f, g as in N.E. #1

x_0, y_0, z_0

- set $\begin{cases} t_F = 0.5 \\ nk = 100 \end{cases}$... looks OK relative to $\begin{cases} t_F = 0.5 \\ nk = 100 \end{cases}$

these look OK...

$h = 0.005$ x grows, y decays, z decays fast

- set $\begin{cases} t_F = 1.0 \\ nk = 100 \end{cases}$ - seems OK more of the same x, y, z trends

$h = 0.01$

- set $\begin{cases} t_F = 1.5 \\ nk = 100 \end{cases}$ \leftarrow $h = 0.015$ some initial oscillations in z but they die out. x, y behave as before.

- set $\begin{cases} t_F = 2.0 \\ nk = 100 \end{cases}$ \leftarrow $h = 0.02$ z has major oscillations between -2 and 2 ... not good

- set $\begin{cases} t_F = 2.01 \\ nk = 100 \end{cases}$ \leftarrow $h > 0.02$ - z has growing oscillations

Again observe $\frac{dz}{dt} = \underbrace{(f-g-y)}_z z$

≈ 100 since $f=100, g=0.1$ and $0 < y < 1$

Why is z not decaying?

- set $\begin{cases} t_F = 2.05 \\ nk = 100 \end{cases}$ \leftarrow $h > 0.02$ - now x, y 'blow up' z - oscillates badly...

This is a fairly 'rich' mathematical system that we can easily have trouble with if we are not careful.

What's going on?

The presence of multiple 'time scales' such as the widely different decay rates ~~are~~ a , c and f ~~make it~~ not to mention the other oscillatory dynamics makes this a 'stiff' system (see also Iserles, p.56)

Consider the system of linear differential equations

$$\begin{cases} \vec{y}' = A \vec{y} \\ \vec{y}(0) = \vec{y}_0 \end{cases} \quad A = \text{constant } m \times m \text{ matrix}$$

Assume A is diagonalizable (has a full set of linearly-independent eigenvectors) so $A = V D V^{-1}$ is its spectral factorization (D = diagonal matrix with eigenvalues λ_j of A along the diagonal and V = matrix whose columns are the linearly-independent eigenvectors).

This problem has exact solution

$$\vec{y}_{ex} = V e^{Dt} V^{-1} \vec{y}_0 \quad \text{where } \vec{x}_j \text{ are multiples related to } V, \vec{y}_0.$$

$$e^{Dt} = \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & e^{\lambda_2 t} & \\ 0 & & \ddots \\ & & & e^{\lambda_n t} \end{bmatrix} \quad \text{or } \vec{y}_{ex} = e^{\lambda_1 t} \vec{x}_1 + \dots + e^{\lambda_n t} \vec{x}_n$$

What happens if we apply Euler's method to this problem?

$$\vec{y}_{n+1} = \vec{y}_n + h A \vec{y}_n = (I + h A) \vec{y}_n$$

So

$$\vec{y}_{n1} = (I + h A) \vec{y}_0$$

$$\vec{y}_{n2} = (I + h A) \vec{y}_1 = (I + h A)^2 \vec{y}_0$$

$$\boxed{\vec{y}_n = (I + h A)^n \vec{y}_0} \quad n = 0, 1, 2, \dots$$

Note: $A = VDV^{-1}$

$$\begin{aligned}
& (I + hA)(I + hA) \\
&= I + 2hA + h^2A^2 \\
&= I + 2hVDV^{-1} + h^2(VDV^{-1})(VDV^{-1}) \\
&= I + 2hVDV^{-1} + h^2(VD^2V^{-1}) \\
&= VV^{-1} + 2hVDV^{-1} + h^2(VD^2V^{-1}) \\
&= V(I + 2hD + h^2D^2)V^{-1} \\
&= V(I + hD)^2V^{-1}
\end{aligned}$$

etc. so

$$(I + hA)^n = V(I + hD)^n V^{-1}$$

and then

$$\vec{y}_n = V(I + hD)^n V^{-1} \vec{y}_0$$

but also note

$$I + hD = \begin{bmatrix} 1+h\lambda_1 & & & \\ & 1+h\lambda_2 & & \\ & & \ddots & \\ & & & 1+h\lambda_m \end{bmatrix}$$

so

$$(I + hD)^n = \begin{bmatrix} (1+h\lambda_1)^n & & & \\ & (1+h\lambda_2)^n & & \\ & & \ddots & \\ & & & (1+h\lambda_m)^n \end{bmatrix}$$

so $\vec{y}_n = V \begin{bmatrix} (1+h\lambda_1)^n \\ \vdots \\ (1+h\lambda_m)^n \end{bmatrix} V^{-1} \vec{y}_0$

$$\vec{y}_n = (1+h\lambda_1)^n \vec{x}_1 + (1+h\lambda_2)^n \vec{x}_2 + \dots + (1+h\lambda_m)^n \vec{x}_m$$

again, recall exact soln...

$$\vec{y}_{ex} = e^{\lambda_1 t} \vec{x}_1 + e^{\lambda_2 t} \vec{x}_2 + \dots + e^{\lambda_m t} \vec{x}_m$$

EXAMPLE (Zseries, p. 53-54)

$$A = \begin{bmatrix} -100 & 0 \\ 0 & -\frac{1}{10} \end{bmatrix} \Rightarrow A = D \quad \begin{matrix} \lambda_1 = -100 \\ \lambda_2 = -\frac{1}{10} \end{matrix}$$

$$\vec{y}_n = (1-100h)^n \vec{x}_1 + (1-\frac{1}{10}h)^n \vec{x}_2$$

$$\vec{y}_{ex} = e^{-100t} \vec{x}_1 + e^{-\frac{1}{10}t} \vec{x}_2$$

Comments

- the exact solution has $\vec{y}_{ex} \rightarrow 0$ as $t \rightarrow \infty$ for any initial conditions.
- Does Euler's solution have this property? (i.e. $\vec{y}_n \rightarrow 0$ as $n \rightarrow \infty$ for fixed h ?)

↳ Not necessarily...

• $(1-100h)^n \rightarrow 0$ as $n \rightarrow \infty$ for fixed h if

$$|1-100h| < 1$$

$$-1 < 1-100h < 1$$

$$-2 < -100h < 0$$

always true
for $h > 0$

here we need

$$h < \frac{2}{100}$$

← step size restriction!

• we ~~also~~ also need $(1-\frac{1}{10}h)^n \rightarrow 0$ as

$$|1-\frac{1}{10}h| < 1$$

$$-1 < 1-\frac{1}{10}h < 1$$

$$-2 < -\frac{1}{10}h < 0$$

↳ here we need

$$h < \frac{2}{(1/10)} = 20$$

← step size restriction (but probably not a very restrictive one).

~~The~~ Isodes (Fig. 4.1) shows what happens if you violate the first one of these (i.e. they use $h = \frac{1}{10}$ and $\frac{2}{100} < \frac{1}{10} < 20$).

Everything looks ok for a while in this example until the $(1-100h)^n$ terms start getting large!

The above examples and discussion highlight the

idea of a 'stiff' differential equation:

$$\vec{y}' = \vec{f}(t, \vec{y}) \quad \vec{y}(t_0) = \vec{y}_0$$

Some characteristics of stiff differential equations

- some numerical methods for solving these equations require a significant reduction in step size - maybe far beyond that expected necessary for accuracy - to avoid instability.
- the equations contain modes ~~or~~ or describe physical phenomena that are associated with ~~different time scales~~ a broad range of time scales. (e.g. we saw exponential decay rates of 100 and $\frac{1}{10}$ in the Iseries example). Fast dynamics and slow dynamics both present.
- the equations have associated with them a large 'stiffness ratio' = ratio of largest to smallest eigenvalue (in modulus) of the Jacobian $\frac{\partial f_i}{\partial y_j}$ matrix.

* stiffness is in general a relative concept

- depends on method used (Euler, Backward Euler, ...)
 - " " accuracy required
 - " " length of interval (i.e. t_{final})
- i.e. the consequences of stiffness are not always apparent

(see discussion in Heath, p. 401)

* stiff systems of ODEs often arise from a semidiscretization of a PDE (e.g. using the method of lines ...)

Many of the key concepts associated with stiff equations and the stability of numerical methods can be understood by examining a very simple linear, scalar ODE

*
$$\begin{cases} y' = \lambda y & \lambda \in \mathbb{C} \text{ (Complex } \lambda) \\ y(0) = y_0 \end{cases}$$

where we are focused here on the case with $\text{Re}(\lambda) < 0$ so that $y \rightarrow 0$ as $t \rightarrow \infty$ in the exact solution.

Note:

- we saw the $\lambda_1, \lambda_2, \dots, \lambda_n$ come out of the analysis in our linear system $\vec{y}' = A\vec{y}$ (so think in terms of the worst case scenario... largest $|\lambda_i|$).
- In a nonlinear system we still have interest in the eigenvalues of the Jacobian $\frac{\partial \vec{f}}{\partial \vec{y}}$ where again an understanding of equation * is central.

Linear stability + A-stability Domain

Consider
$$\begin{cases} y' = \lambda y & \lambda \in \mathbb{C} \\ y(0) = 1 & (y_0 = 1 \text{ for simplicity}) \end{cases}$$

$\text{Re}(\lambda) < 0$, so $y \rightarrow 0$ as $t \rightarrow \infty$ is a feature of the exact sol.

Euler

$$y_{n+1} = y_n + h\lambda y_n = (1+h\lambda)y_n$$

$$\Rightarrow y_n = (1+h\lambda)^n \quad n=0,1,2,\dots$$

so for $y_n \rightarrow 0$ as $n \rightarrow \infty$ require

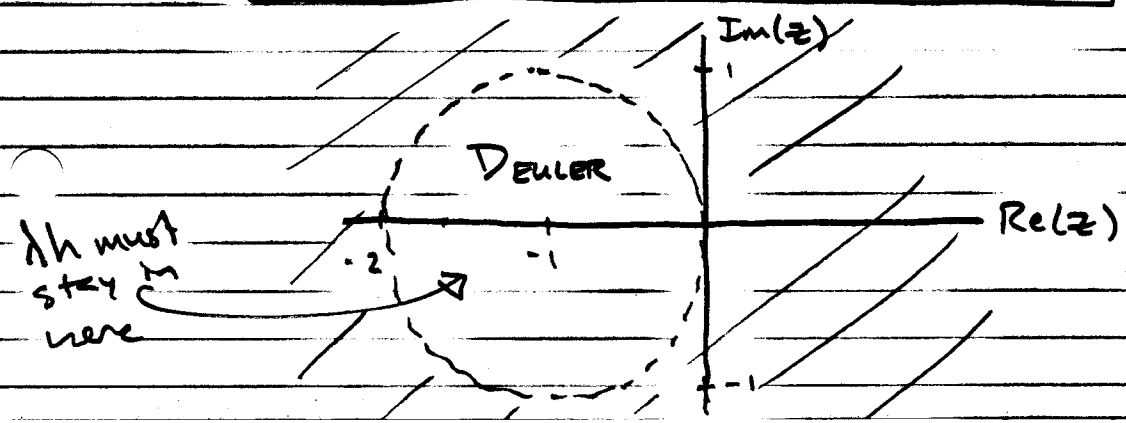
$$|1+h\lambda| < 1 \quad \leftarrow \text{time step \& restriction}$$

Let $z \equiv h\lambda \in \mathbb{C}$.

Define the linear stability domain D to be the set of all $h\lambda$ ($= z$) such that $y_n \rightarrow 0$ as $n \rightarrow \infty$. So

$$D_{\text{EULER}} = \left\{ z \in \mathbb{C} \mid |1+z| < 1 \right\}$$

linear stability domain for Euler



h must stay in here

... so if $|\lambda|$ is big then h may need to be quite small...

Backward Euler

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

$$(1-h\lambda)y_{n+1} = y_n$$

$$y_{n+1} = \frac{1}{(1-h\lambda)} y_n$$

assume $1-h\lambda \neq 0$

$$y_n = \frac{1}{(1-h\lambda)^n} \quad \text{for } n=0,1,2,\dots$$

so for $y_n \rightarrow 0$ as $n \rightarrow \infty$ for fixed h require

$$\frac{1}{|1-h\lambda|} < 1$$

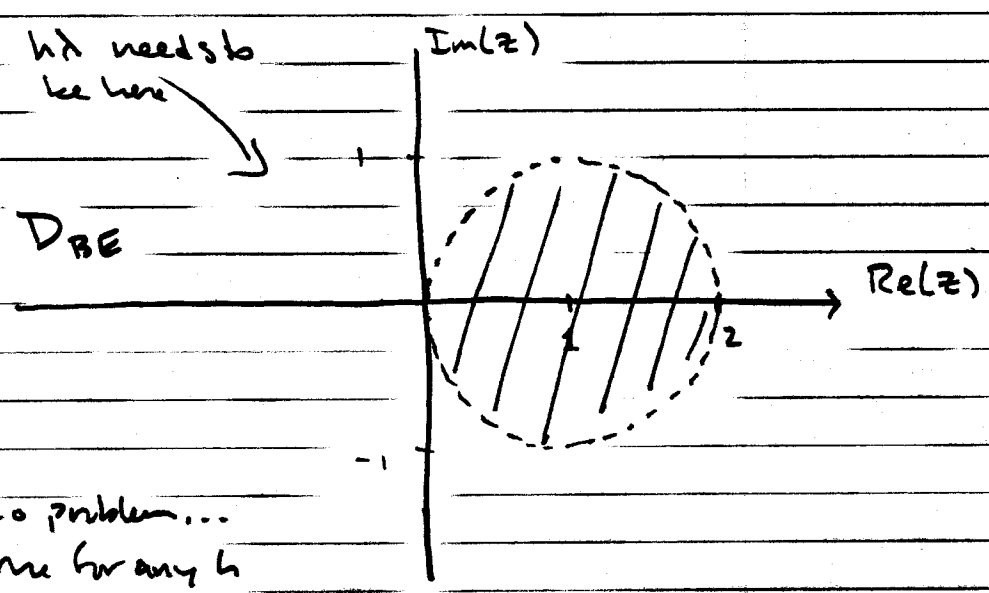
no time step restriction for $\text{Re}(\lambda) < 0$.

~~Backward Euler~~

so, assuming $1-h\lambda \neq 0$ require

$$|1-h\lambda| > 1$$

$$\text{let } z = h\lambda \Rightarrow |1-z| > 1$$



(no problem... true for any h with $\text{Re}(\lambda) < 0$)

Trapezoid Method

$$y_{n+1} = y_n + \frac{1}{2}h (\lambda y_n + \lambda y_{n+1})$$

$$(1 - \frac{1}{2}h\lambda) y_{n+1} = (1 + \frac{1}{2}h\lambda) y_n$$

$$y_{n+1} = \frac{(1 + \frac{1}{2}h\lambda)}{(1 - \frac{1}{2}h\lambda)} y_n$$

$1 - \frac{1}{2}h\lambda \neq 0$

$$y_n = \left[\frac{1 + \frac{1}{2}h\lambda}{1 - \frac{1}{2}h\lambda} \right]^n$$

Here we require

$$\left| \frac{1 + \frac{1}{2}h\lambda}{1 - \frac{1}{2}h\lambda} \right| < 1$$

no restriction on h if $\text{Re}(\lambda) < 0$.

i.e.

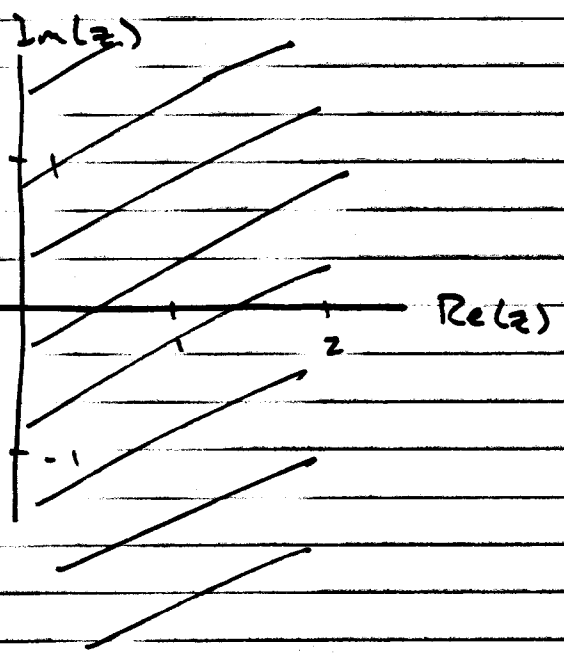
$$\left| \frac{1 + \frac{1}{2}z}{1 - \frac{1}{2}z} \right| < 1$$

always true if $\text{Re}(z) < 0$
always false if $\text{Re}(z) > 0$

$h\lambda$ needs to be here

TRAP

no problem if $\text{Re}(\lambda) < 0$.



For Heun's Method see Homework...

Midpoint Rule (Implicit)

$$y_{n+1} = y_n + h \bar{f}\left(t_n + \frac{1}{2}h, \frac{1}{2}(y_n + y_{n+1})\right)$$

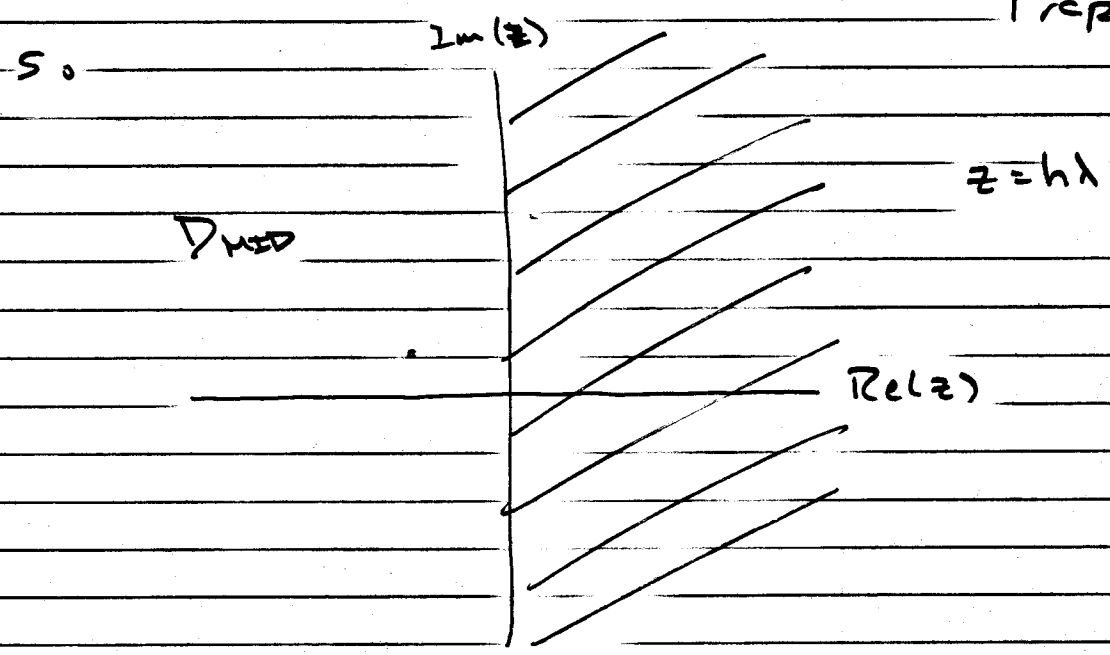
see Isoles
p.12

for $y' = \lambda y$ this reduces to

$$y_{n+1} = y_n + h \left(\lambda \cdot \frac{1}{2}(y_n + y_{n+1}) \right)$$

$$\left(1 - \frac{1}{2}h\lambda\right) y_{n+1} = \left(1 + \frac{1}{2}h\lambda\right) y_n$$

← same result
for this problem
as we saw for
Trapezoid



The ideas we've been using in these examples are captured in the following Lemma

Lemma 4.2 (Iseries, p. 68)

Suppose the application of a numerical method to ~~approx~~ the problem

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

leads to the geometric solution sequence


$$y_n = [r(h\lambda)]^n \quad n=0,1,2,\dots$$

Then, the linear stability domain D is

$$D = \{ z \in \mathbb{C} \mid |r(z)| < 1 \}$$

Def: A-stable (i.e. a numerical method is A-stable when...)

We say that a numerical method is

 = $\mathbb{C}^- = \{z \in \mathbb{C} \mid \text{Re}(z) < 0\} \subseteq D$

where D is the linear stability domain in terms of $z = h\lambda$, where $y' = \lambda y$

So for the previously discussed methods, we have

A-stable

not A-stable

Backward Euler

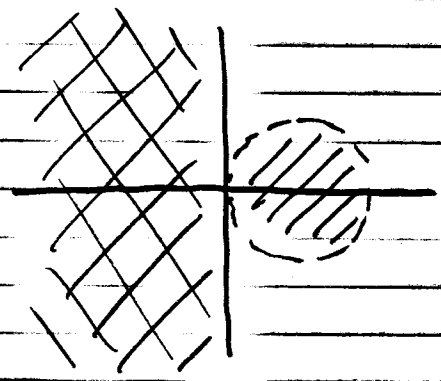
Euler

Trapezoid

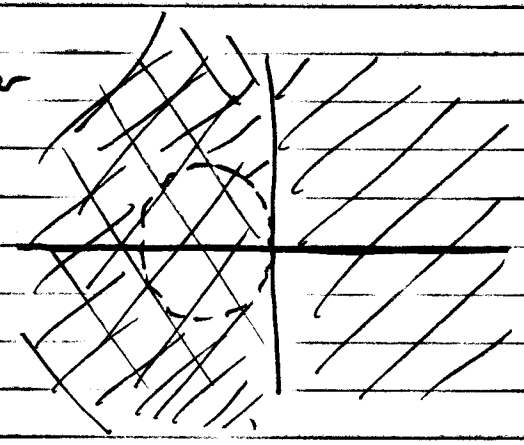
Heun (see HW)

Implicit Midpoint

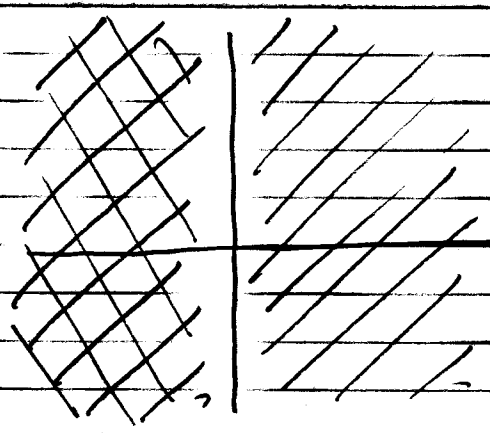
B.E.



Euler



TRAP + IMP. MID



Heun

