

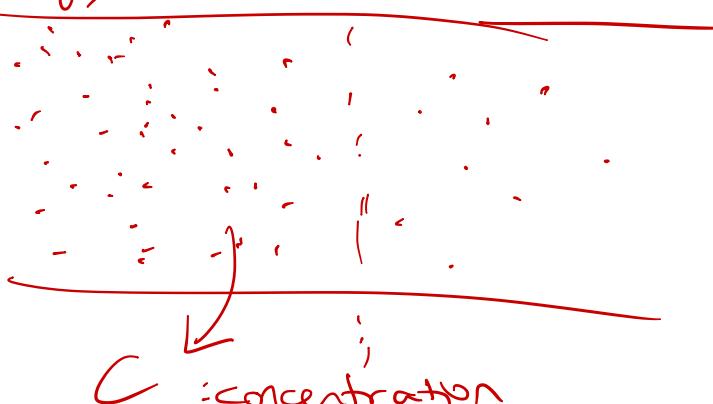
BLG 202 E

Week 13 – 03.05.2016

ODEs

Chapter 16 Differential Equations

Diffusion :



$$\underbrace{J}_{\text{Flux}} = D \frac{\partial C}{\partial x}$$

$$\frac{\partial J}{\partial x} = \frac{\partial C}{\partial t}$$

$$\frac{\partial C}{\partial t} = D \frac{\partial^2 C}{\partial x^2}$$

ODE : Ordinary Diff Eqs.

$$\boxed{\frac{dy}{dt} = y' = f(t, y(t))} \rightarrow y(t_0) = y_0$$

$$\int_{t_i}^{t_{i+1}} \frac{dy}{ds} ds = \int_{t_i}^{t_{i+1}} f(s, y(s)) ds$$

$$y(t_{i+1}) - y(t_i) = \int_{t_i}^{t_{i+1}} f(s, y(s)) ds$$

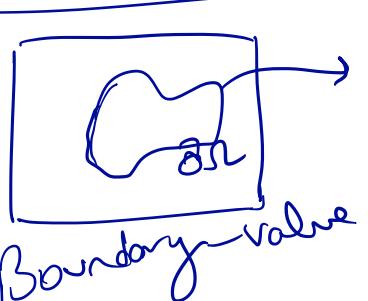
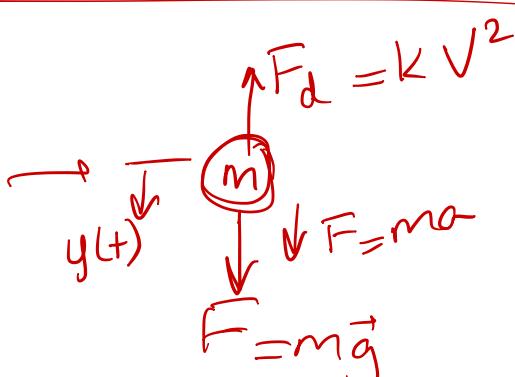
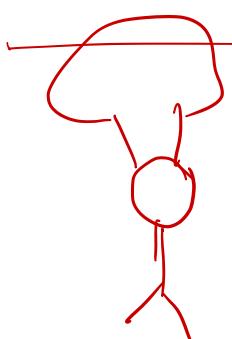
Initial-value ODES

$y(t, x_0) = a$
Boundary value

~~We cannot numerically solve it.~~
 ~~y is implicitly defined~~

$$m\vec{g} + m\vec{a}(t) - \frac{k}{m} V^2(t) = 0$$

$$\frac{d^2 y(t)}{dt^2} = -g + \frac{k}{m} \left(\frac{dy(t)}{dt} \right)^2$$



$$v(x, y)|_{\partial\Omega} = v_0(x, y)$$

$$\frac{\partial v}{\partial t} = k |\nabla v|$$

PDE
Curvature

Initial Value ODE problem

Find $y(t)$ that satisfies

$$y'(t) = \frac{dy(t)}{dt} = f(t, y(t))$$

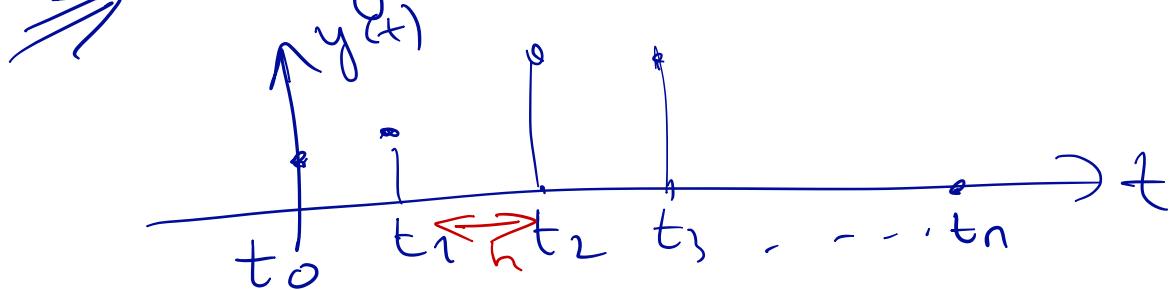
with initial condition

$$y(t_0) = y_0$$

makes the soln. unique.

- $y(t)$ is defined implicitly thru the differential eqn.
 - Numerical soln to this problem generates a sequence of values $t_0, t_1, \dots, t_n, \dots$ and a corresponding seq of values $y_0, y_1, \dots, y_n, \dots$ so that

$$y_n \approx y(t_n), \quad n = 0, 1, 2, \dots$$



$h = t_{i+1} - t_i$: fixed step size.

- ODEs : system of differential equations:

$$\frac{d\mathbf{y}}{dt} = f(t, \mathbf{y}(t))$$

$$\mathbf{y}(t_0) = \mathbf{y}_0$$

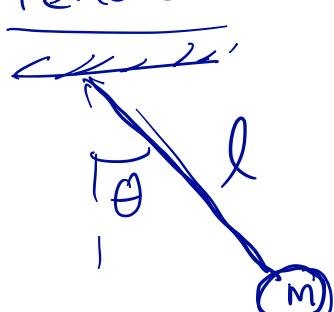
$$\mathbf{y}(t) = \begin{pmatrix} y_1(t) \\ y_2(t) \\ \vdots \\ y_m(t) \end{pmatrix}$$

t : independent variable.

$y(t)$: trajectory (solution) of the ODE.
 for Initial value ODE.

Note on theory: The initial value has a unique soln. provided f is "regular enough" (Lipschitz cont) (We'll not cover here.).

Ex: Pendulum : Angular acceleration : (Newton's laws forces)



$$\frac{d^2\theta}{dt^2} = -\frac{g}{l} \sin\theta$$

constant of gravity

2nd order ODE :

$$= f(t, \underbrace{y(t)}_{\theta(t)})$$

Write as first-order

ODE system :

$$y_1(t) = \theta(t) \Rightarrow y_1'(t) = y_2(t)$$

$$y_2(t) = \theta'(t) \Rightarrow y_2'(t) = -\frac{g}{l} \sin\theta = f(t, \theta(t)) \\ = f(\theta(t))$$

ODE system for the pendulum

$$\underline{f}(t, \underline{y}) = \begin{pmatrix} y_2 \\ -\frac{g}{l} \sin(y_1) \end{pmatrix}, \quad \underline{y} = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$$

$$\left\{ \begin{array}{l} \frac{dy}{dt} = \underline{f}(t, \underline{y}) \\ y_1(t_0) = \theta(t_0) \\ y_2(t_0) = \theta'(t_0) \end{array} \right. \quad \text{vector notation}$$

} initial values are given.

Ex.: Mathematical ecology: populations of prey, and predators.

Prey-predator model: $U(t)$: prey (rabbit) population
 $V(t)$: predator (fox) population

$$\begin{cases} U' = \alpha U - \beta VU = (\alpha - \beta V)U \end{cases}$$

$$\boxed{\alpha, \beta, \gamma, \delta > 0}$$

$$\begin{cases} V' = -\gamma V + \delta UV = (\delta U - \gamma)V \end{cases}$$

The soln. to this nonlinear system of equations

$$\begin{pmatrix} \underline{y}' = \begin{pmatrix} U' \\ V' \end{pmatrix} = f \\ \underline{y} = \begin{pmatrix} U \\ V \end{pmatrix}, f(\underline{y}) = \begin{pmatrix} (\alpha - \beta V)U \\ (\delta U - \gamma)V \end{pmatrix} \end{pmatrix}$$

can be obtained numerically. The solns are periodic. (See figure).

Simplest Method to solve numerically an ODE

Forward Euler Method:

$$\begin{cases} y' = f(t, y) \\ y(t_0) = y_0 \end{cases}$$

Grid (mesh) points t_0, t_1, \dots, t_N
 $h = t_{i+1} - t_i$.
initial value
fixed step size.

Find Approximate soln. $y(t_i) \approx y_i$ ←

* Apply forward-difference:

$$y'(t_i) = \frac{y(t_{i+1}) - y(t_i)}{h} - \frac{h}{2} y''(\xi)$$

$$f(t_i, y(t_i))$$

$$\Rightarrow y(t_{i+1}) = y(t_i) + h f(t_i, y(t_i)) + \frac{h^2}{2} y''(\xi) \quad \rightarrow$$

The initial value problem is then solved by :

$$\begin{aligned} \underline{y}_{i+1} &= \underline{y}_i + h f(t_i, \underline{y}_i) \\ t_{i+1} &= t_i + h \\ \underline{y}_0 &= \underline{y}(t_0) \end{aligned}$$

Forward Euler method

Ex: $\begin{cases} y' = y \\ y(0) = 1 \end{cases}$ $\cdot y(t) = e^t$ exact soln.

Forget about this, go to numerical soln

⇒ Apply forward Euler:

$$y_0 = 1$$

$$y_{i+1} = y_i + h \cdot y_i$$

$$y_{i+1} = (1+h) y_i$$

$$y_0 = 1$$

$$i = 0, 1, 2, \dots$$

Backward Euler Method:

Instead of forward, use the backward difference:

$$y'(t_{i+1}) = \frac{\underline{y}(t_{i+1}) - \underline{y}(t_i)}{h} + \frac{h}{2} y''(\xi)$$

Recall
 $\underline{y}(t+h) = \dots$
 $\underline{y}(t-h) = \dots$

Derivation:

$$\underline{y}(t_{i-1}) = \underline{y}(t_i) - h \underline{y}'(t_i) + \frac{h^2}{2} y''(\xi)$$

$$\underline{y}'(t_i) \leftarrow \frac{\underline{y}(t_i) - \underline{y}(t_{i-1})}{h} + \frac{h}{2} y''(\xi)$$

$$\underline{y}'(t_{i+1}) = \frac{\underline{y}(t_{i+1}) - \underline{y}(t_i)}{h} + \frac{h}{2} y''(\xi) \quad \checkmark$$



So, set : $y_{i+1} = y_i + h f(t_{i+1}, \underline{y}_{i+1})$

 $t_{i+1} = t_i + h$

Note that y_{i+1} (unknown) appears implicitly!

For stepping procedure, solve a system of eqns
(nonlinear)

$$y_{i+1} - y_i - h f(t_{i+1}, y_{i+1}) = 0$$

\therefore More costly than forward Euler method.

* Forward Euler is explicit } methods.
Backward Euler is implicit . }

* Absolute Stability:

During computations, step size choice becomes important -

- Consider a simple test equation:

$$\begin{cases} y' = \lambda y \\ y(0) = y_0 \end{cases}$$

soln: $y(t) = y_0 e^{\lambda t}$.
 $\lambda > 0$: models growth,
 $\lambda < 0$: models decay
(death)

Apply forward Euler method:

$$y_{i+1} = y_i + h f(t_i, y_i) = y_i + h \lambda y_i$$

$$y_{i+1} = (1 + h\lambda) y_i$$

Require that approx soln does not grow: $|y_{i+1}| \leq |y_i|$

$$|y_{i+1}| \leq |y_i| \quad y_{i+1} = (1+\lambda h) y_i$$

$$\Rightarrow |1+\lambda h| \leq 1 \quad \text{for stability.}$$

$$-1 \leq 1+\lambda h \leq 1$$

$$y' = \lambda y$$

$$y(t) = y_0 e^{\lambda t}$$

$\lambda > 0$: models growth, numerically also grows.

$\lambda < 0$:

$$-1 \leq 1+\lambda h \Rightarrow h\lambda \geq -2$$

$$h \leftarrow \frac{-2}{\lambda}$$

$$h \leq \frac{2}{|\lambda|}$$

$\lambda < 0$, Absolute stability requirement

$$h \leq \frac{2}{|\lambda|}$$

Bound on the Step size h

\Rightarrow for $\lambda < 0$, we must require h to satisfy this bound for stable solns.

$$\text{Ex: } y' = -\frac{1000(y - \cos(t))}{1000} - \sin(t)$$

$$\begin{aligned} y' &= -1000 \frac{(y - \cos(t))}{1000} - \sin(t) \\ y' &= -1000 \frac{y}{1000} + 1000 \frac{\cos(t)}{1000} - \frac{\sin(t)}{1000} \\ y(0) &= 1 \end{aligned}$$

$$y(t) = \cos(t)$$

If we want to apply forward Euler method.

We have to choose $h \leq \frac{2}{1000} = \underline{\underline{0.002}}$

\Rightarrow otherwise, you end up with a non-correct soln.

$\Rightarrow \underline{\underline{\text{HW!}}}$

* Stability: Test eqn : $y' = \lambda y$

Apply the Backward Euler method :

$$y_{i+1} = y_i + h\lambda y_{i+1}$$

$$y_{i+1} = \frac{1}{1-h\lambda} y_i \Rightarrow \text{we require } |y_{i+1}| \leq |y_i| \text{ for absolute stability}$$

$$\Rightarrow \frac{1}{|1-h\lambda|} \leq 1$$

$$-(1-h\lambda) \leq 1$$

$$h\lambda - 1 \leq 1$$

$$h\lambda \leq +2$$

$$\lambda < 0$$

$$h \geq \frac{2}{\lambda}$$

$$-1 \leq \frac{1}{1-h\lambda} \leq 1$$

$$1-h\lambda \geq 1$$

$$h\lambda \leq 0$$

$$\lambda < 0$$

$$h > 0$$

$$h > 0$$

$$\lambda < 0$$

No bound / (absolute stability criterion)

w/ Forward Euler method

Ex (*) : $h > 0.002$ with Forward Euler method

$$h = 0.01\pi \rightarrow \text{the result blows up.}$$

w/ Backward Euler $\frac{h=0.01\pi}{h=0.1\pi}$ gives error $= 3.2e^{-9}$
 \Downarrow " $= 1.7e^{-5}$

Forward & Backward Euler are $O(h)$.

Runge-Kutta Methods: higher order methods

The Euler methods we've seen are only 1st order accurate. Want higher accuracy.

* ODE: $y' = f(t, y)$, integrate the ODE from t_i to t_{i+1} (in 1 step):

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

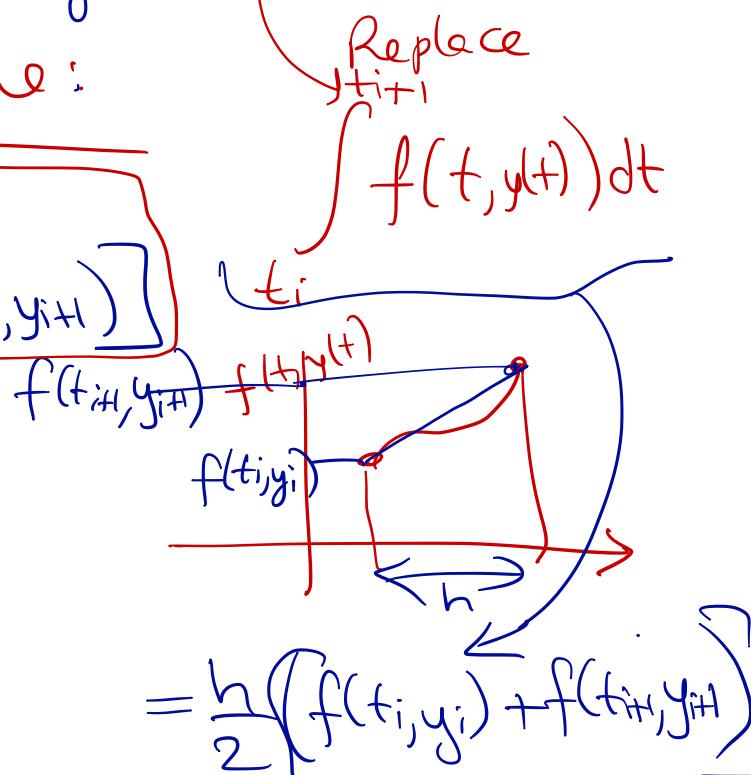
discretize this integral by a basic quadrature rule.

Implicit Trapezoidal rule:

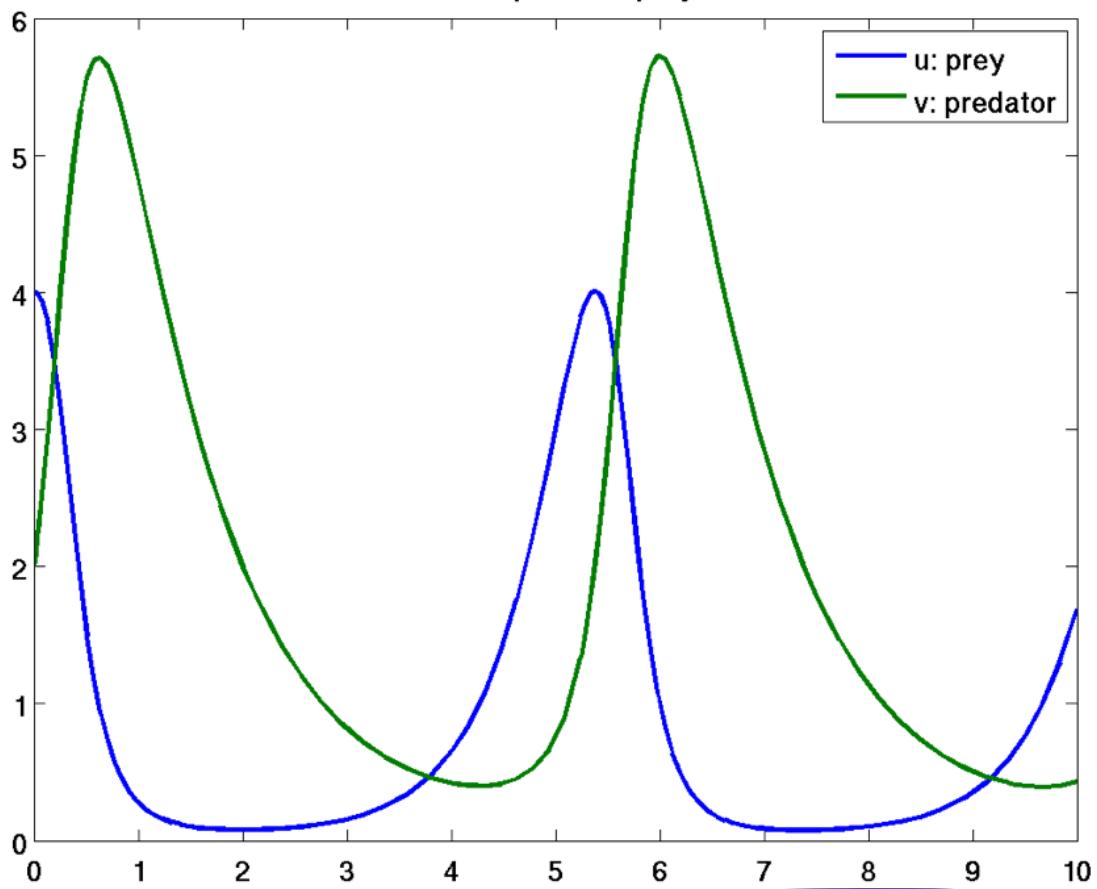
$$y_{i+1} = y_i + \frac{h}{2} [f(t_i, y_i) + f(t_{i+1}, y_{i+1})]$$

Implicit!

$O(h^2)$



Lotka-Volterra predator-prey model



Unknown f :

$$E(f(x)) = \int (f - g)^2 dx + \int \left(\frac{\partial f}{\partial x} \right)^2 dx$$

Cost fn.

$$\arg \min_f E(f)$$

$$\frac{\partial E}{\partial f} = 0 \Rightarrow \boxed{\frac{\partial f}{\partial t} = \dots}$$

Diff equations are also used in optimizing functionals (cost functions).