

Symplectic Integration of Hamiltonian Systems

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Outline

1 Background and Problem

- Introduction of Background
- Pendulum Problem

2 Theory and basic definitions

- Lagrange's and Hamilton's Equations
- Symplectic Transformations
- Examples of Symplectic Integrators

3 Numerical experiments

- Lotka-Volterra Problem
- Numerical results and conclusions

- Introduction of Background

They throw geometry out the door, and it comes back through the window. ————— — H.G.Forder, Auckland 1973

Introduction of Background

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General ordinary differential equations (non-stiff and stiff)

solved

Runge–Kutta methods or linear multistep methods

• However

- astronomy
- molecular dynamics
- mechanics
- theoretical physics
- numerical analysis
- applied and pure mathematics

Need develop

- Numerical methods which preserve geometric properties of the flow.

- Introduction of Background

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- Hamiltonian systems form the most important class of ordinary differential equations.
- Some famous problems of Hamiltonian systems in physics:
 1. Kepler Problem
 2. Outer Solar system
 3. Highly Oscillatory Problem
- The flow of Hamiltonian systems is symplectic.

Here,

- We study Symplectic integrators which preserve geometric properties of Hamiltonian systems' flow.

- The Pendulum Problem

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The Pendulum Problem

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- **The Pendulum as a Hamiltonian System**

- Hamiltonian problem

$$\dot{p} = -H_q(p, q) \quad \dot{q} = H_p(p, q)$$

- where
- the *Hamiltonian* $H(p_1, \dots, p_d, q_1, \dots, q_d)$ represents the total energy;
- q_i are the position coordinates;
- p_i the momenta for $i = 1, \dots, d$, with d the number of degrees of freedom;
- H_p and H_q are the vectors of partial derivatives.

- the solution curves of

$$H(p(t), q(t)) = \text{Const}$$

- i.e., the Hamiltonian is an invariant or a *first integral*.

- The Pendulum Problem

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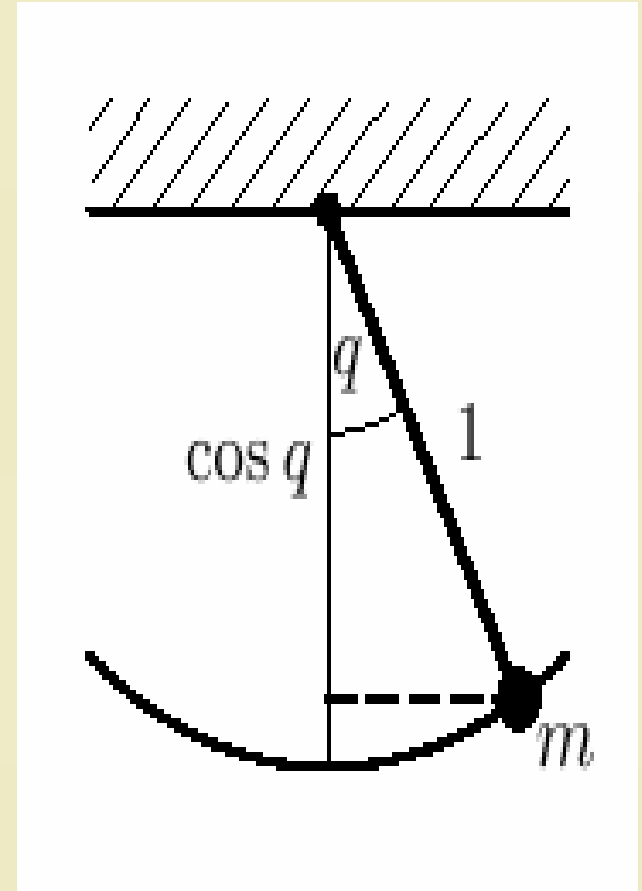
- **Pendulum.**

- The mathematical pendulum (mass $m = 1$, massless rod of length $= 1$, gravitational acceleration $g = 1$) is a system with one degree of freedom having the Hamiltonian

$$H(p, q) = \frac{1}{2}p^2 - \cos q$$

- So that the equations of motion become

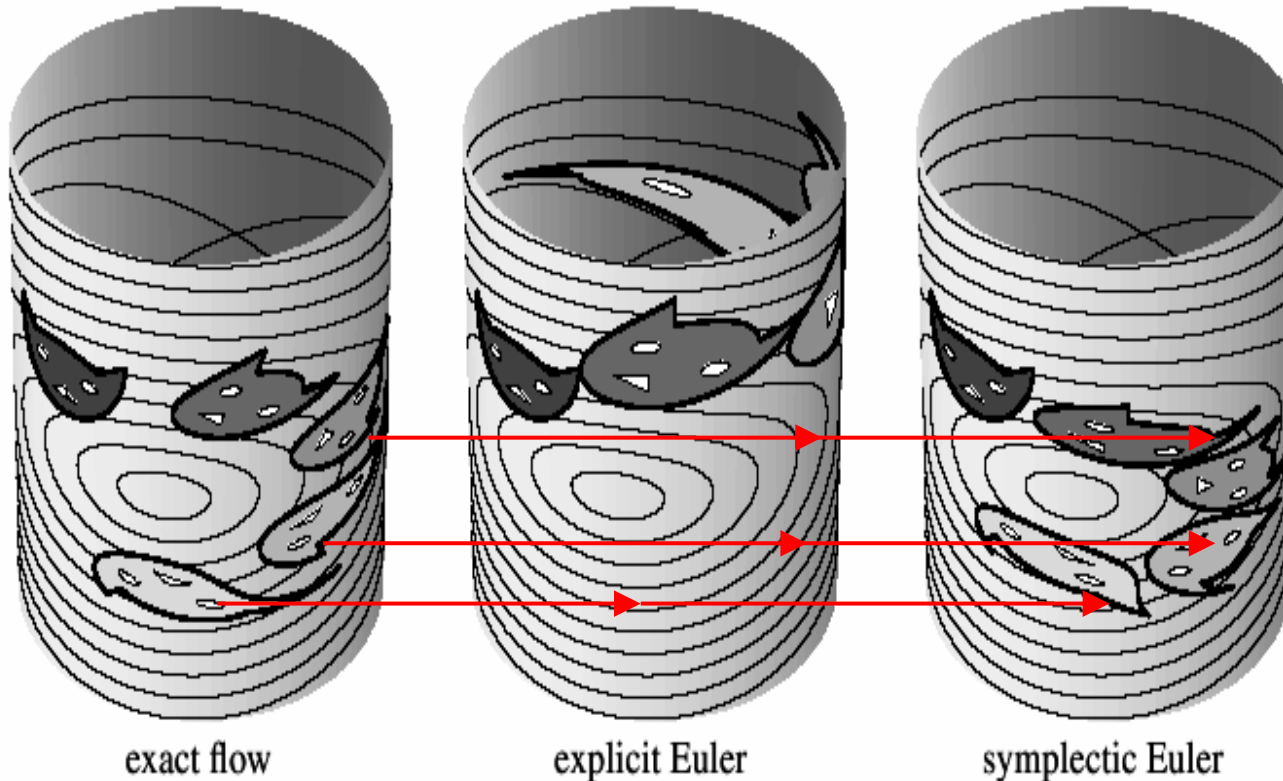
$$\dot{p} = -\sin q \qquad \dot{q} = p$$



- The Pendulum Problem

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- Area Preservation.



From Book 'Geometric Numerical Integration'

- Lagrange's and Hamilton's Equations

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Lagrange's and Hamilton's Equations

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- derive the Hamilton's Equations from a mechanical problem.
- By theory from mechanics :

$$T = T(q, \dot{q})$$

- which represents the *kinetic energy*, where $q = (q_1, \dots, q_d)^T$ as generalized coordinates.

$$U = U(q)$$

- representing the *potential energy*. Then, after denoting by

$$L = T - U$$

- the corresponding *Lagrangian*, the coordinates $q_1(t), \dots, q_d(t)$ obey the differential equations

$$\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}} \right) = \frac{\partial L}{\partial q}$$

- which constitute the *Lagrange equations* of the system.

- Lagrange's and Hamilton's Equations

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- Hamilton (1834) simplified the structure of Lagrange's equations by
- introducing Poisson's variables, the conjugate *momenta*

$$p_k = \frac{\partial L}{\partial \dot{q}_k}(q, \dot{q}) \quad \text{for } k=1, \dots, d,$$

- considering the *Hamiltonian*

$$H := p^T \dot{q} - L(q, \dot{q})$$

- $H = H(p, q)$ obtained by expressing \dot{q} as a function of p and q
- for every q , a continuously differentiable bijection $\dot{q} \leftrightarrow p$. This map is called the *Legendre transform*.

- Lagrange's and Hamilton's Equations

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- Theorem 1.** *Lagrange's equations are equivalent to Hamilton's equations*

$$\dot{p}_k = -\frac{\partial H}{\partial q_k}(p, q) \quad \dot{q}_k = \frac{\partial H}{\partial p_k}(p, q) \quad K=1, \dots, d$$

- Proof.* The definitions conjugate momenta and Hamiltonian for the momenta p and for the Hamiltonian H imply that

$$\frac{\partial H}{\partial p} = \dot{q}^T + p^T \frac{\partial \dot{q}}{\partial p} - \frac{\partial L}{\partial \dot{q}} \frac{\partial \dot{q}}{\partial p} = \dot{q}^T$$

$$\frac{\partial H}{\partial q} = p^T \frac{\partial \dot{q}}{\partial q} - \frac{\partial L}{\partial q} - \frac{\partial L}{\partial \dot{q}} \frac{\partial \dot{q}}{\partial q} = -\frac{\partial L}{\partial q} = -\dot{p}$$

- Symplectic Transformations

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Symplectic Transformations

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- Properties of Hamiltonian systems

1. the Hamiltonian $H(p, q)$ is a *first integral* of the system

$$\dot{p}_k = -\frac{\partial H}{\partial q_k}(p, q) \qquad \dot{q}_k = \frac{\partial H}{\partial p_k}(p, q)$$

2. Its flow has a property called *symplecticity*.

- Symplectic Transformations

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- two-dimensional parallelograms lying in R^{2d}

$$\xi = \begin{pmatrix} \xi^p \\ \xi^q \end{pmatrix} \quad \eta = \begin{pmatrix} \eta^p \\ \eta^q \end{pmatrix}$$

- In the (p, q) space ($\xi_p, \xi_q, \eta_p, \eta_q$ are in R^d) as

$$P = \{t\xi + s\eta \mid 0 \leq t \leq 1, 0 \leq s \leq 1\}$$

- In the case $d = 1$, the *oriented area*

$$\text{or. area } (P) = \det \begin{pmatrix} \xi^p & \eta^p \\ \xi^q & \eta^q \end{pmatrix} = \xi^p \eta^q - \xi^q \eta^p$$

- In higher dimensions, we replace this by the *sum of the oriented areas of the projections of P onto the coordinate planes* (p_i, q_i) , i.e., by

$$\omega(\xi, \eta) := \sum_{i=1}^d \det \begin{pmatrix} \xi_i^p & \eta_i^p \\ \xi_i^q & \eta_i^q \end{pmatrix} = \sum_{i=1}^d (\xi_i^p \eta_i^q - \xi_i^q \eta_i^p)$$

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- In matrix notation, this bilinear map acting on vectors of R^{2d} has the form

$$\omega(\xi, \eta) = \xi^T J \eta \quad \text{with} \quad J = \begin{pmatrix} 0 & I \\ -I & 0 \end{pmatrix}$$

where I is the identity matrix of dimension d .

- **Definition 2.1.** A linear mapping $A : R^{2d} \rightarrow R^{2d}$ is called *symplectic* if

$$A^T J A = J$$

Or, equivalently, if $\omega(A\xi, A\eta) = \omega(\xi, \eta)$ for all $\xi, \eta \in R^{2d}$

- Symplectic Transformations

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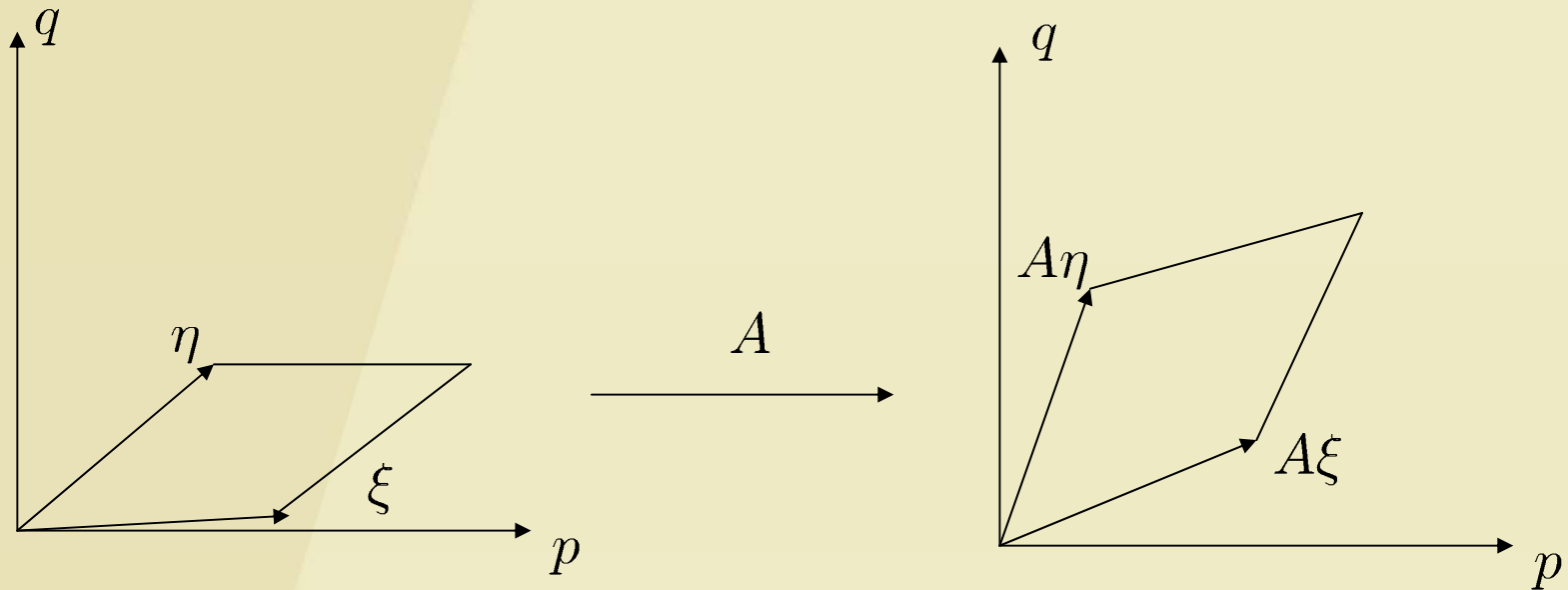


Fig. 2.1. Symplecticity (area preservation) of a linear mapping

- Symplectic Transformations

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- For nonlinear mappings

- **Definition 2.2.** A differentiable map $g : U \rightarrow R^{2d}$ (where $U \subset R^{2d}$ is an open set) is called *symplectic* if the Jacobian matrix $g(p, q)$ is everywhere symplectic, i.e., if

$$g'(p, q)^T J g'(p, q) = J \quad \text{or} \quad \omega(g'(p, q)\xi, g'(p, q)\eta) = \omega(\xi, \eta)$$

- all symplectic mappings (also nonlinear ones) are *area preserving*.

- Symplectic Transformations

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- We use the notation $y = (p, q)$, and we write the Hamiltonian system in the form

$$\dot{y} = J^{-1} \nabla H(y)$$

Where J is the matrix we see before and $\nabla H(y) = H'(y)^T$.

- The flow $\varphi_t : U \rightarrow \mathbb{R}^{2d}$ of a Hamiltonian system is the mapping that advances the solution by time t , i.e., $\varphi_t(p_0, q_0) = (p(t, p_0, q_0), q(t, p_0, q_0))$
- where $(p(t, p_0, q_0), q(t, p_0, q_0))$ is the solution of the system corresponding to initial values $p(0) = p_0, q(0) = q_0$.

- Symplectic Transformations

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- **Theorem 2.3 (Poincar'e 1899).** *Let $H(p, q)$ be a twice continuously differentiable function on $U \subset \mathbb{R}^{2d}$. Then, for each fixed t , the flow φ_t is a symplectic transformation wherever it is defined.*

- Symplectic Transformations

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- *locally Hamiltonian:*

$$\dot{y} = f(y)$$

if for every $y_0 \in U$ there exists a neighbourhood where $f(y) = J^{-1} \nabla H(y)$ for some function H .

- **Theorem 2.4. (characteristic property for Hamiltonian systems)** Let $f: U \rightarrow \mathbb{R}^{2d}$ be continuously differentiable. Then, $\dot{y} = f(y)$ is locally Hamiltonian if and only if its flow $\varphi_t(y)$ is symplectic for all $y \in U$ and for all sufficiently small t .

- **Lemma 2.5 (Integrability Lemma).** Let $D \subset \mathbb{R}^n$ be open and $f: D \rightarrow \mathbb{R}^n$ be continuously differentiable, and assume that the Jacobian $f'(y)$ is symmetric for all $y \in D$. Then, for every $y_0 \in D$ there exists a neighbourhood and a function $H(y)$ such that

$$f(y) = \nabla H(y)$$

on this neighbourhood

- Examples of Symplectic Integrators

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Examples of Symplectic Integrators

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- In the following we show the symplecticity of various numerical methods when they are applied to the Hamiltonian system in the variables $y = (p, q)$,

$$\dot{p} = -H_q(p, q)$$

$$\dot{q} = H_p(p, q)$$

Or equivalently $\dot{y} = J^{-1} \nabla H(y)$

- where H_p and H_q denote the column vectors of partial derivatives of the Hamiltonian $H(p, q)$ with respect to p and q , respectively.

- Examples of Symplectic Integrators

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- Definition 3.1.** A numerical one-step method is called *symplectic* if the one-step map

$$y_1 = \phi_h(y_0)$$

is symplectic whenever the method is applied to a smooth Hamiltonian system.

- Theorem 3.1 (de Vogelaere 1956).** *The so-called symplectic Euler methods*

$$\begin{array}{l} p_{n+1} = p_n - hH_q(p_{n+1}, q_n) \\ q_{n+1} = q_n + hH_p(p_{n+1}, q_n) \end{array} \quad \text{or} \quad \begin{array}{l} p_{n+1} = p_n - hH_q(p_n, q_{n+1}) \\ q_{n+1} = q_n + hH_p(p_n, q_{n+1}) \end{array}$$

are symplectic methods of order 1.

- Examples of Symplectic Integrators

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- Proof.*

- We consider only the method to the left. Differentiation with respect to (p_n, q_n) yields

$$\begin{pmatrix} 1 + hH_{qp}^T & 0 \\ -hH_{pp} & I \end{pmatrix} \begin{pmatrix} \frac{\partial(p_{n+1}, q_{n+1})}{\partial(p_n, q_n)} \end{pmatrix} = \begin{pmatrix} I & -hH_{qq} \\ 0 & I + hH_{qp} \end{pmatrix}$$

where the matrices H_{pq}, H_{pp}, \dots of partial derivatives are all evaluated at

(p_{n+1}, q_n) . This relation allows us to compute $\frac{\partial(p_{n+1}, q_{n+1})}{\partial(p_n, q_n)}$ and to check in

a straightforward way the symplecticity condition

$$\begin{pmatrix} \frac{\partial(p_{n+1}, q_{n+1})}{\partial(p_n, q_n)} \end{pmatrix}^T J \begin{pmatrix} \frac{\partial(p_{n+1}, q_{n+1})}{\partial(p_n, q_n)} \end{pmatrix} = J$$

- Examples of Symplectic Integrators

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- Theorem 3.2.** *The implicit midpoint rule*

$$y_{n+1} = y_n + hJ^{-1}\nabla H((y_{n+1} + y_n)/2)$$

is a symplectic method of order 2.

- Proof.*
- Differentiation yields

$$\left(I - \frac{h}{2}J^{-1}\nabla^2 H\right)\left(\frac{\partial y_{n+1}}{\partial y_n}\right) = \left(I + \frac{h}{2}J^{-1}\nabla^2 H\right)$$

- Again it is straightforward to verify that $\left(\frac{\partial y_{n+1}}{\partial y_n}\right)^T J \left(\frac{\partial y_{n+1}}{\partial y_n}\right) = J$. Due to its symmetry, the midpoint rule is known to be of order 2

- Numerical experiments

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Numerical experiments

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- **The Lotka–Volterra Mode**

- We start with an equation from mathematical biology which models the growth of animal species.

- Assumptions:

1. $u(t)$ is to represent the number of individuals of a certain species at time t
2. Its evolution is $du/dt = u \cdot a$, where a is the reproduction rate.
3. $u(t)$ denotes the number of predators and $v(t)$ denotes the number of prey.

The we have *Lotka–Volterra model* :

$$\dot{u} = u(v - 2)$$

$$\dot{v} = v(1 - u)$$

- where the dots on u and v stand for differentiation with respect to time. (We have chosen the constants 2 and 1 arbitrarily.)

- Numerical experiments

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$$\dot{y} = f(y)$$

- Every y represents a point in the *phase space*, $y = (u, v)$ is in the phase plane .
- The vector-valued function $f(y)$ represents a *vector field*

- Flow of the System.**

- A fundamental concept is the *flow* over time t . denoted by φ_t , is thus defined by

$$\varphi_t(y_0) = y(t) \quad \text{if} \quad y(0) = y_0$$

- Numerical experiments

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- Invariants.**

- If we divide the two equations by each other, we obtain a single equation between the variables u and v . After separation of variables we get

$$0 = \frac{1-u}{u} \dot{u} - \frac{v-2}{v} \dot{v} = \frac{d}{dt} I(u, v)$$

Where

$$I(u, v) = \ln u - u + 2 \ln v - v$$

- so that $I(u(t), v(t)) = \text{Const}$ for all t .
- We call the function I an *invariant* of the system.
- Every solution of model thus lies on a level curve of I .
- Some of these curves are drawn in the pictures of Fig. 4.1.

- Numerical experiments

- three stages:
 - U: prey
 - V: predator
- the prey population increases.
 - the predator population increases by feeding on the prey.
 - the predator population diminishes due to lack of food

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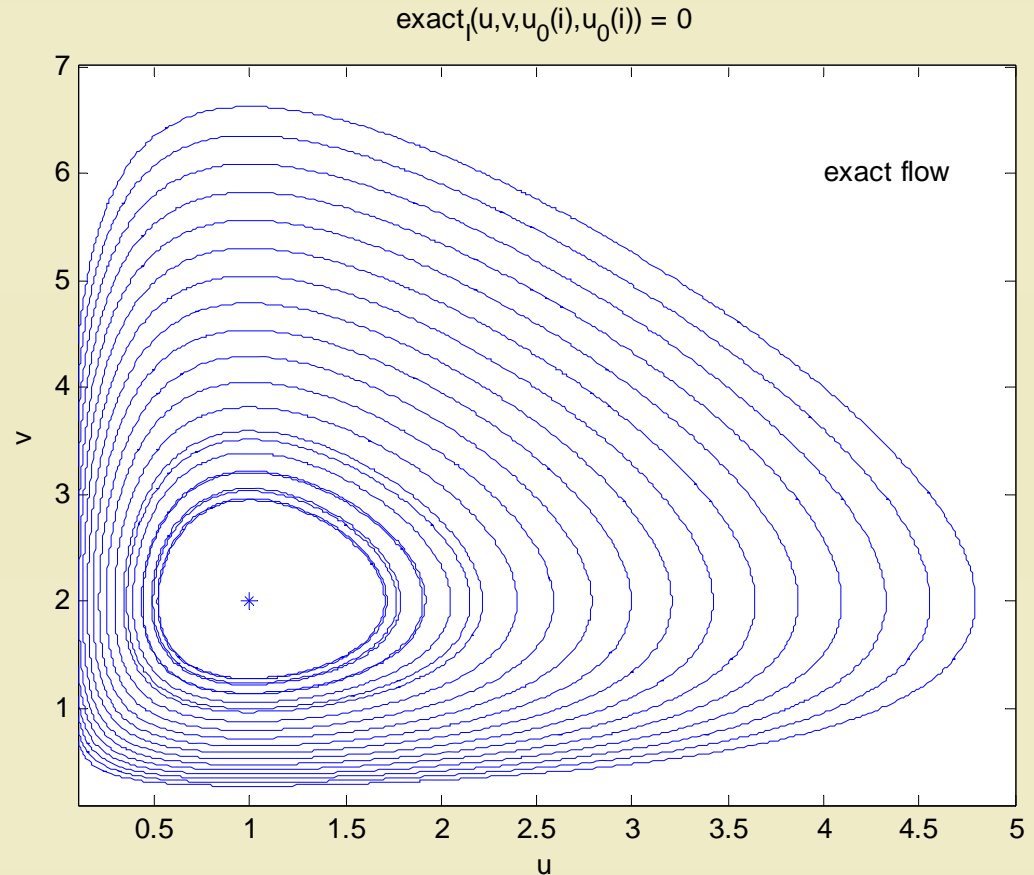


Fig. 4.1. exact flow

- Numerical results and conclusions

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- **Explicit Euler Method.**

- The simplest of all numerical methods for the system $\dot{y} = f(y)$ is the method formulated by Euler (1768),

$$y_{n+1} = y_n + hf(y_n)$$

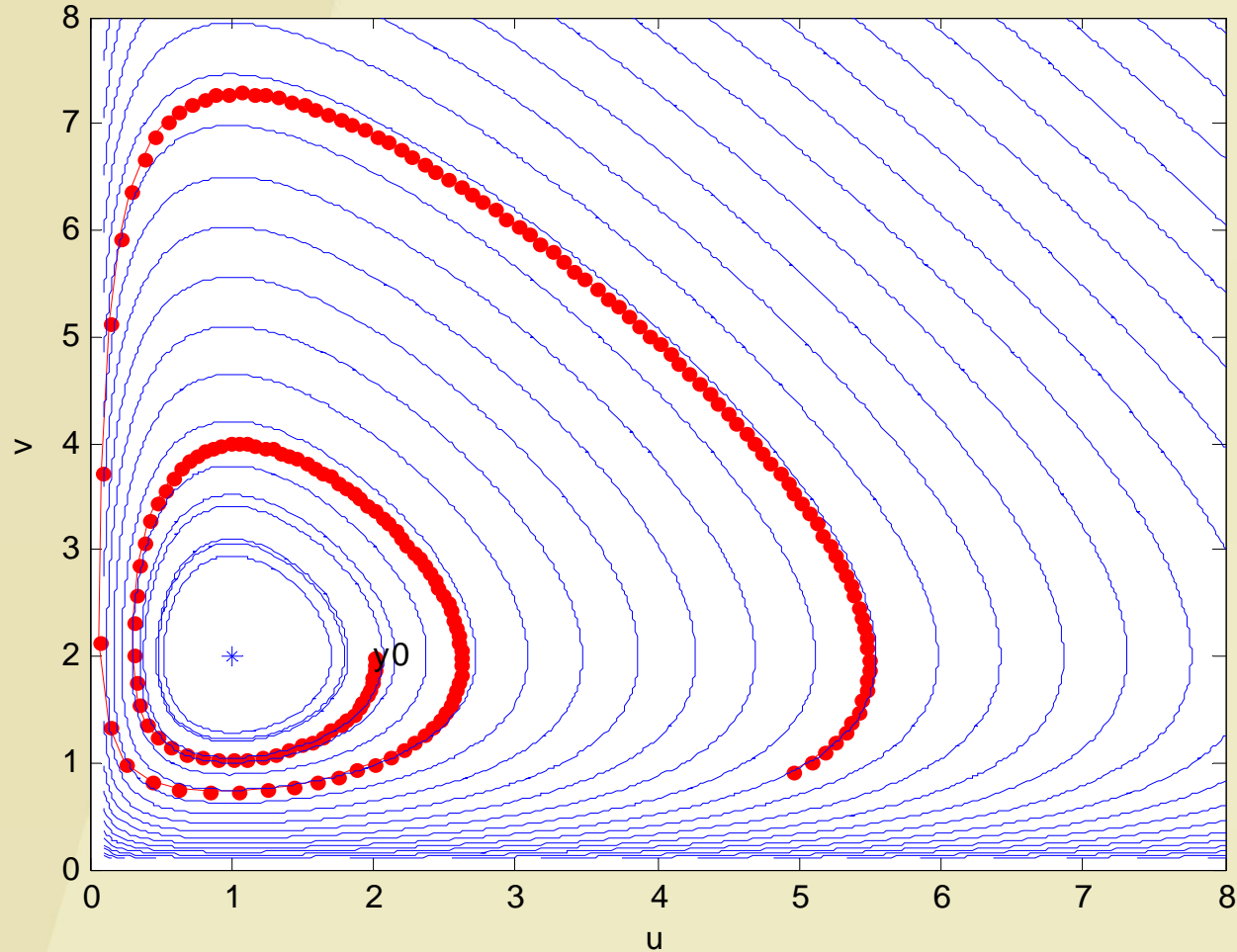
- constant step size h
- given initial value $y(0) = y_0$.
- *Discrete or numerical flow*

$$\phi_h : y_n \rightarrow y_{n+1}$$

- Numerical results and conclusions

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explicit Euler



- Numerical results and conclusions

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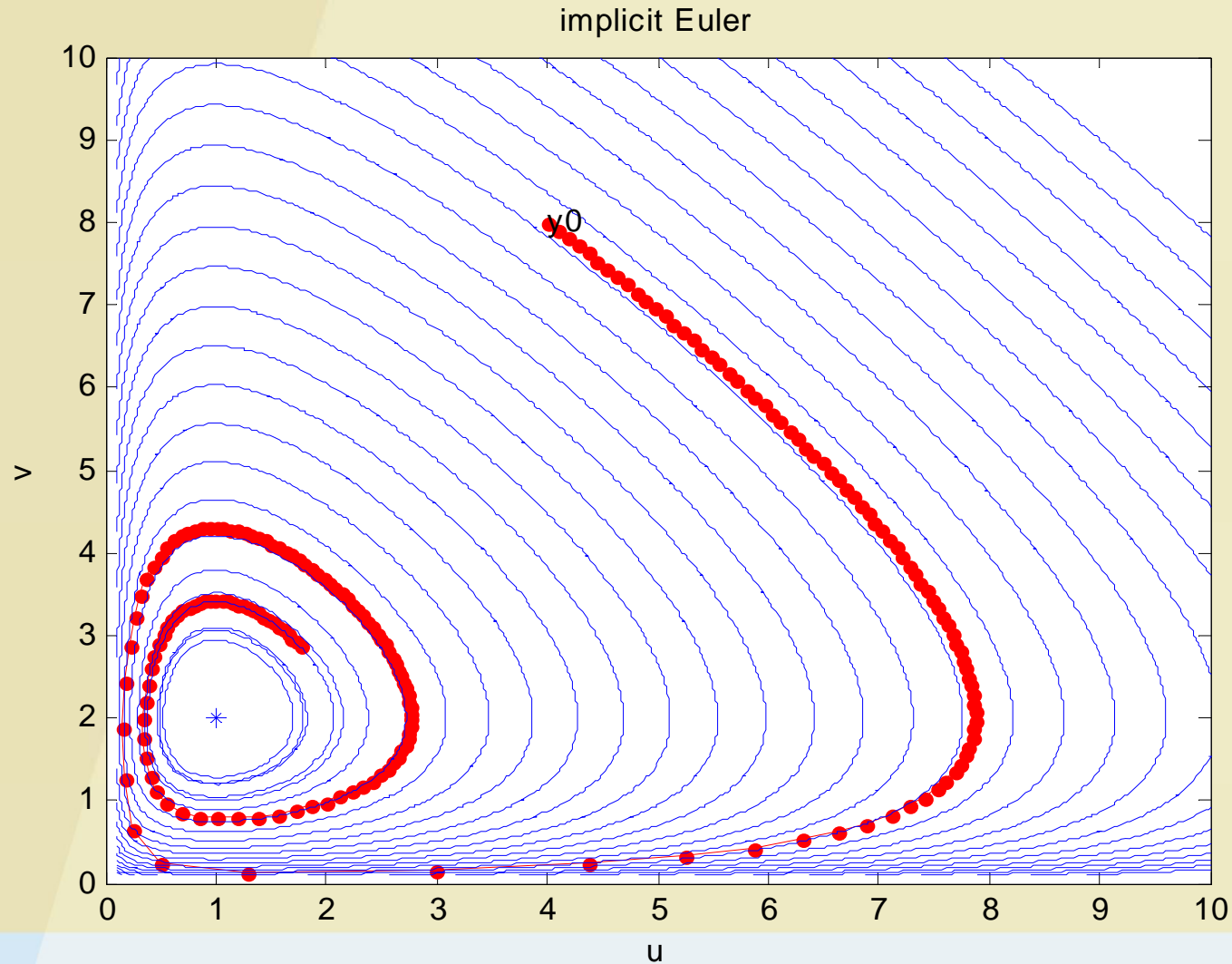
- **Implicit Euler Method.**
- The *implicit Euler method*

$$y_{n+1} = y_n + hf(y_{n+1})$$

- is known for its all-damping stability properties. In contrast to the last one, the approximation y_{n+1} is defined implicitly and the implementation requires the numerical solution of a nonlinear system of equations.

- Numerical results and conclusions

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- Numerical results and conclusions

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- Implicit Midpoint Rule.**

- Taking the mean of y_n and y_{n+1} in the argument of f , we get the *implicit midpoint rule*

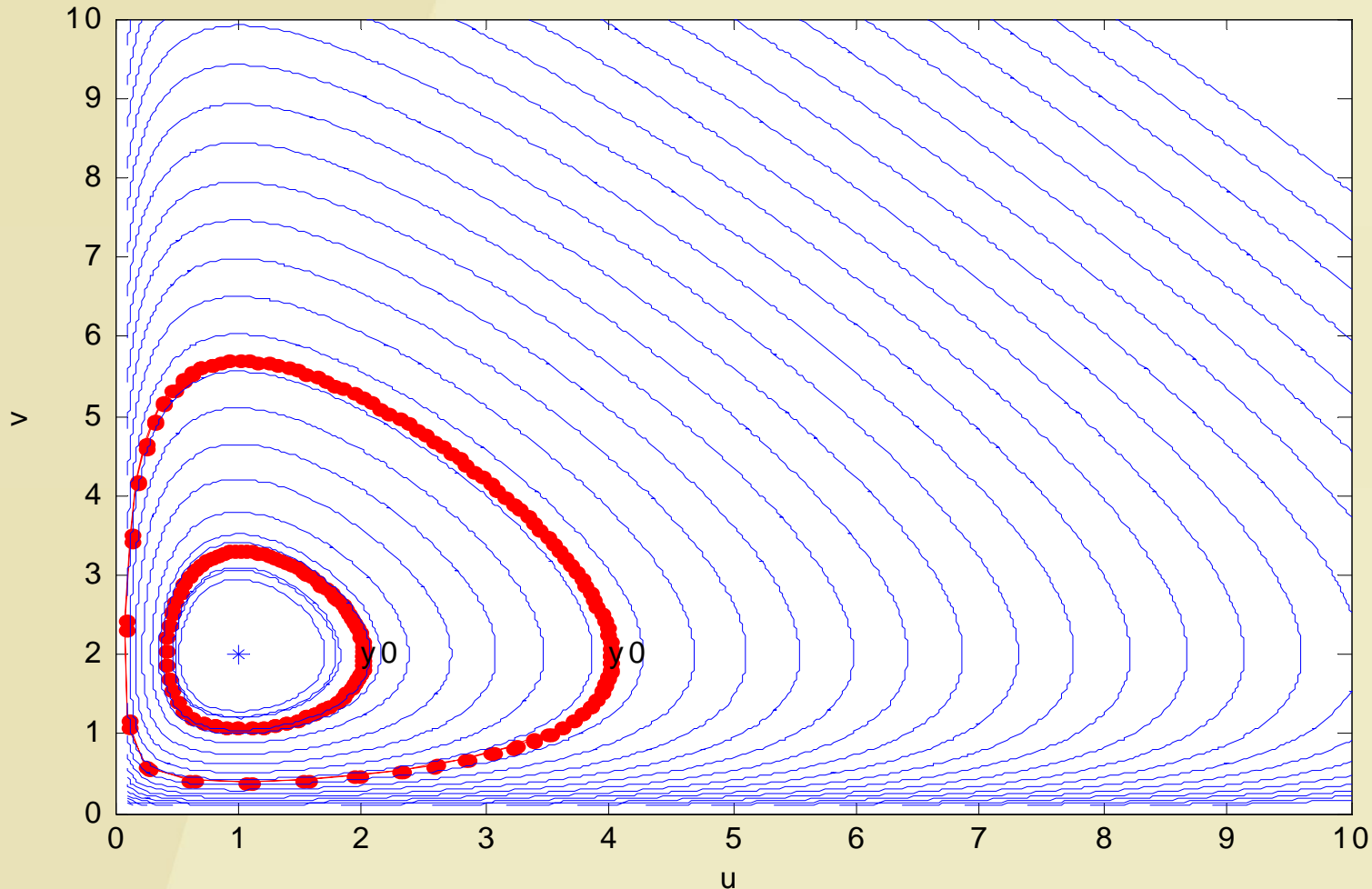
$$y_{n+1} = y_n + hf \left(\frac{y_{n+1} + y_n}{2} \right)$$

- It is a *symmetric* method, which means that the formula is left unaltered after exchanging $y_n \leftrightarrow y_{n+1}$ and $h \leftrightarrow -h$

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implicit Midpoint



- Numerical results and conclusions

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- **Symplectic Euler Methods.**
- For *partitioned* systems

$$\dot{u} = a(u, v)$$

$$\dot{v} = b(u, v)$$

- we consider also *partitioned* Euler methods

$$u_{n+1} = u_n + ha(u_n, v_{n+1})$$

$$u_{n+1} = u_n + ha(u_{n+1}, v_n)$$

or

$$u_{n+1} = u_n + hb(u_n, v_{n+1})$$

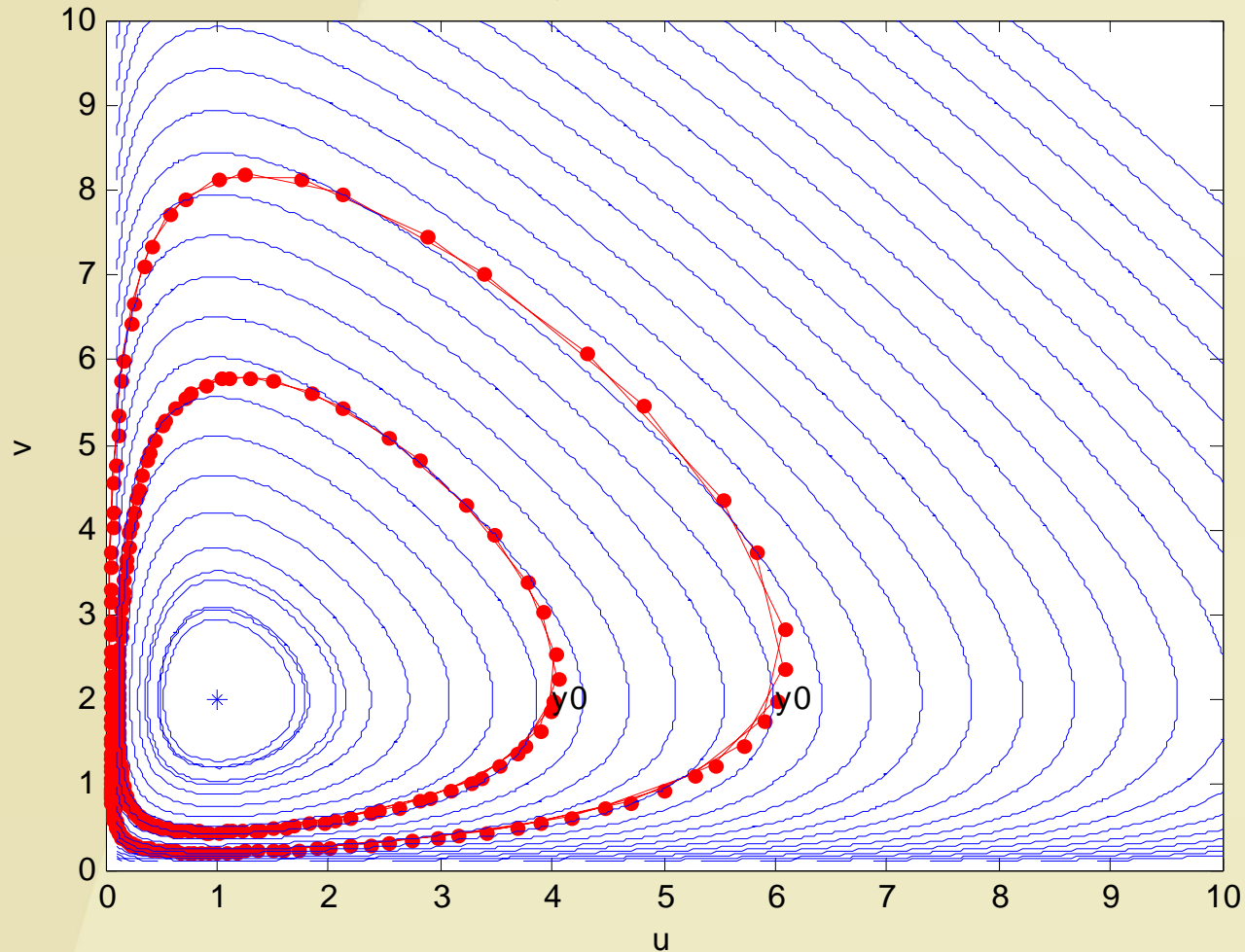
$$u_{n+1} = u_n + hb(u_{n+1}, v_n)$$

- which treat one variable by the implicit and the other variable by the explicit Euler method.

- Numerical results and conclusions

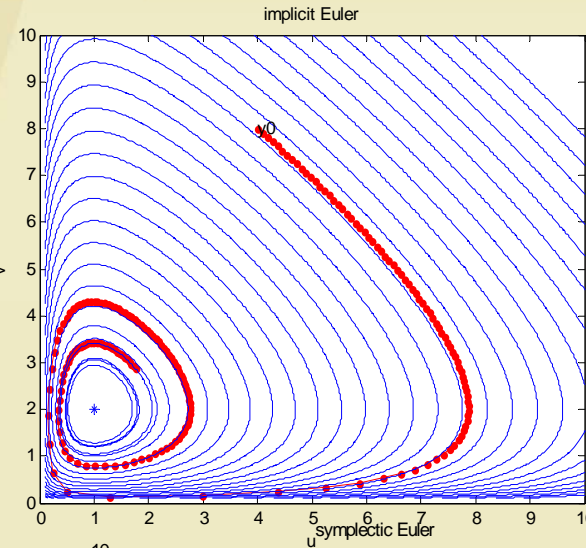
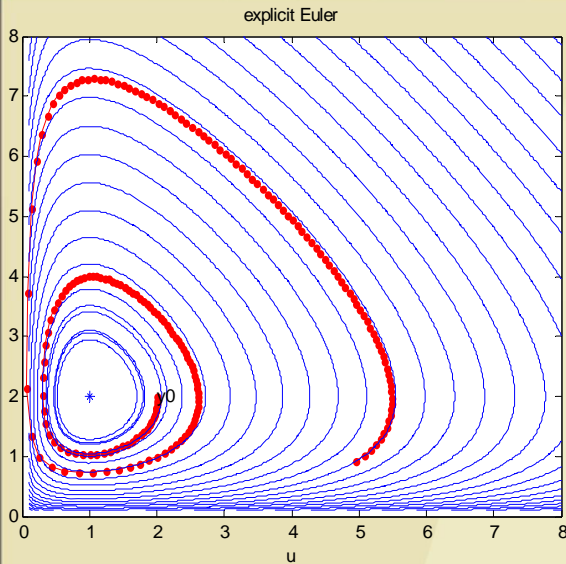
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symplectic Euler

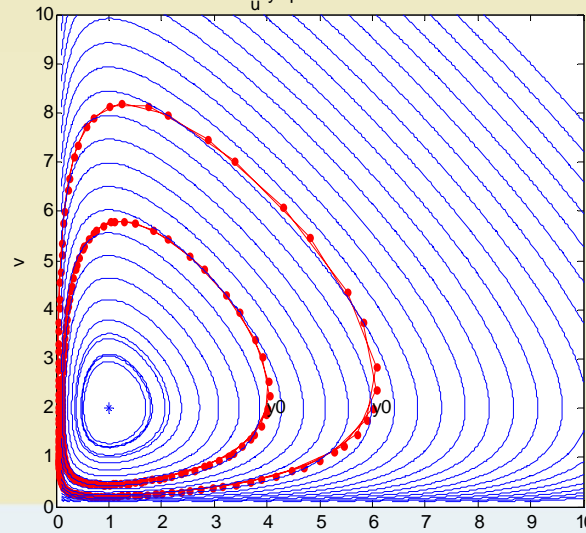


Numerical results and conclusions

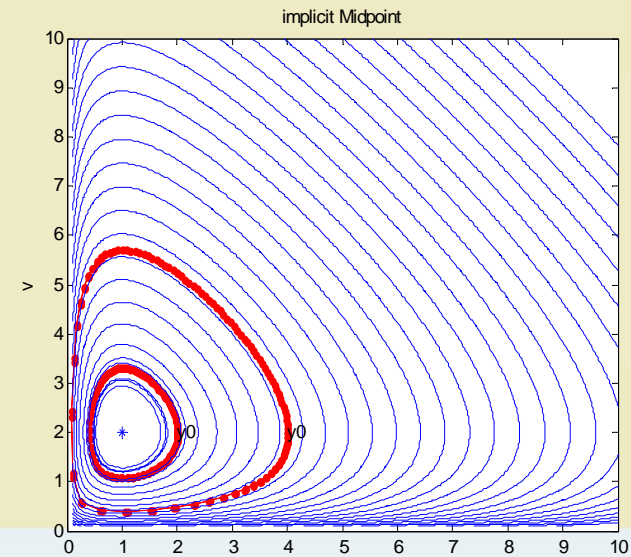
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- (4,2) and (2,2) for implicit midpoint rule
- (4,2) and (6,2) for symplectic Euler method



- step sizes $h = 0.12$
- explicit Euler method has initial values (2, 2)
- (4, 8) for the implicit Euler method



- Numerical results and conclusions

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- **Numerical Results for the Lotka–Volterra Problem**

- the explicit and implicit Euler methods show wrong qualitative behaviour.
- Implicit Midpoint Rule gives a numerical solution that lies apparently on a closed curve as does the exact solution.
- The symplectic Euler method (implicit in u and explicit in v), however, gives a numerical solution that lies apparently on a closed curve as does the exact solution.

Thank you