

# Structure-Preserving Integration for LQR and Filtering

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## Symplectic Integration

HAIRER, WANNER, LUBICH 2004 | MARSDEN, WEST 2001  
OBER-BLÖBAUM, JUNGE, MARSDEN 2011

- Hamilton's equations for  $H : T^*Q \rightarrow \mathbb{R}$  with external forces

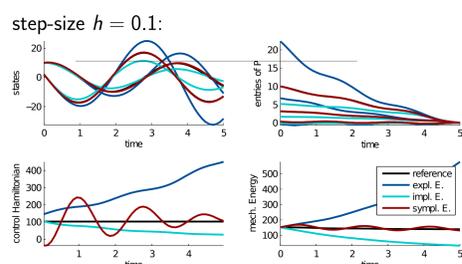
$$\dot{q} = \frac{\partial H}{\partial p}, \quad \dot{p} = -\frac{\partial H}{\partial q} + f(q, p, u)$$

- flow of a (free) Hamiltonian system is *symplectic*
- AIM**: preserve symplecticity in numerical simulations
- symplectic integration scheme* defines a symplectic one-step map  $\Phi_k : (q_k, p_k) \rightarrow (q_{k+1}, p_{k+1})$
- symplectic partitioned RK schemes, e.g. *Symplectic Euler*

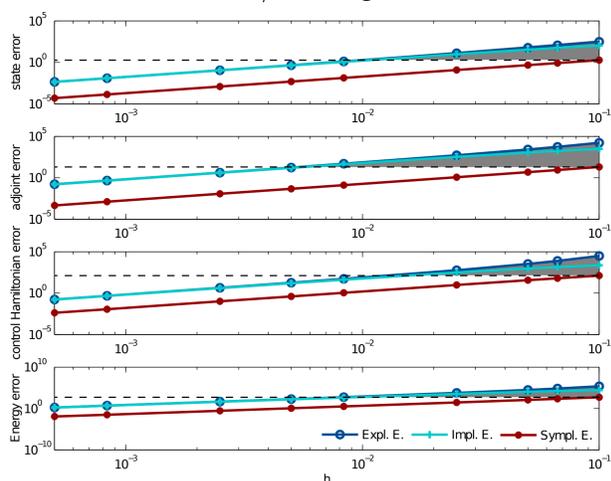
$$\begin{pmatrix} q_{k+1} \\ p_{k+1} \end{pmatrix} = \begin{pmatrix} q_k \\ p_k \end{pmatrix} + h \cdot \begin{pmatrix} \frac{\partial H}{\partial p}(q_{k+1}, p_k) \\ -\frac{\partial H}{\partial q}(q_{k+1}, p_k) + f(q_{k+1}, p_k, u_k) \end{pmatrix}$$

## (Stochastic) Harmonic Oscillator

- comparison of 1st-order methods
- LQR problem:  
 $x^0 = (10, 2)$ ,  $T = 5.0$ ,  
 $A = \begin{pmatrix} 0 & 1 \\ -3 & 0 \end{pmatrix}$ ,  $B = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$ ,  
 $Q = \mathbb{I}$ ,  $R = 100$ ,  $P_T = 0$

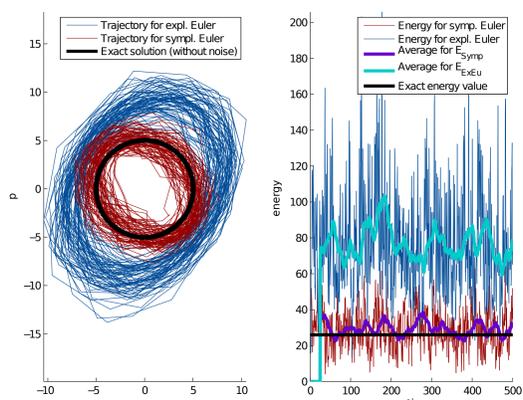


- symplectic discretization allows step-sizes one magnitude larger  $\Rightarrow$  important in real-time robotic applications with low bandwidth actuation/ sensing



- Kalman filtering for stochastic harmonic oscillator

$$\begin{aligned} \dot{x} &= Ax + Gw, \\ z &= Cx + v, \text{ with } \\ &\text{uncorrelated noise} \\ x(0) &\sim \mathcal{N}(x^0, P_0), \\ w &\sim \mathcal{N}(0, Q), v \sim \mathcal{N}(0, R) \\ A &= \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}, G = \mathbb{I}, Q = \mathbb{I}, \\ R &= \text{diag}(1, 100), C = \mathbb{I}, \\ x^0 &= (5, 1), T = 500, h = 0.5 \end{aligned}$$



References: K. Flaßkamp, T. Murphey: Variational Integrators in Linear Optimal Filtering, The 2015 American Control Conference, July 1–3, 2015, Chicago, IL  
K. Flaßkamp, T. Murphey: Structure-Preserving Integration for Linear Optimal Control and Filtering of Mechanical Systems, submitted, 2015.

## Linear Quadratic Optimal Control

$$\min_u J(u) = \frac{1}{2} \int_0^T x^T Q x + u^T R u dt + \frac{1}{2} x(T)^T P_T x(T)$$

w.r.t.  $\dot{x}(t) = Ax(t) + Bu(t)$ ,  $x(0) = x_0$ .

- state-adjoint optimality system

$$\begin{aligned} \dot{x} &= \frac{\partial \mathcal{H}}{\partial p} = Ax + Bu, \quad x(0) = x^0, \quad u = -R^{-1} B^T \rho, \\ \dot{\rho} &= -\frac{\partial \mathcal{H}}{\partial x} = -Qx - A^T \rho, \quad \rho(T) = P_T x(T), \end{aligned}$$

for the LQR problem's Hamiltonian

$$\mathcal{H} = \frac{1}{2} (x^T Q x + u^T R u) + \rho^T (Ax + Bu)$$

- Riccati eq.  $\dot{P} = -PA - A^T P + PBR^{-1} B^T P - Q$ ,  $P(T) = P_T$

## Symplectic Discrete-Time LQR

- AIM**: preserve symplecticity of mechanics and state-adjoint system in discretization
- aside*: standard discrete-time LQR corresponds to explicit Euler discretization of the dynamics and only preserves the symplectic state-adjoint system's flow
- transform LQR problem into Mayer form and apply e.g. symplectic Euler scheme to obtain optimization problem
- The optimal solution satisfies a symplectic discretization of the Hamiltonian state-adjoint system

$$\begin{pmatrix} q_{k+1} \\ p_{k+1} \\ \lambda_{k+1} \\ \mu_{k+1} \end{pmatrix} = \begin{pmatrix} q_k \\ p_k \\ \lambda_k \\ \mu_k \end{pmatrix} + h \begin{pmatrix} A & -BR^{-1}B^T \\ -Q & -A^T \end{pmatrix} \begin{pmatrix} q_k \\ p_k \\ \lambda_k \\ \mu_k \end{pmatrix},$$

with  $(q_0, p_0) = (q^0, p^0)$  and  $(\lambda_N, \mu_N)^T = P_T (q_N, p_N)^T$ .

- The discrete-time Riccati matrices are obtained iteratively from the scheme's one step map  $\Phi_k$ , starting at  $P_N = P_T$ , by

$$P_k = (P_{k+1} \Phi_k^{12} - \Phi_k^{22})^{-1} (\Phi_k^{21} - P_{k+1} \Phi_k^{11}).$$

- symp. state-adjoint scheme as in OBER-BLÖBAUM ET AL. 2011 | HAGER 2000
- structure-preserving feedback controller can be used in projection-based nonlinear trajectory optimization
- applicable to linear (Kalman) filtering via duality principle



**Kathrin Flaßkamp**

- Education
  - Postdoc at Northwestern University, USA
  - Ph.D. and Diploma in math from U Paderborn
- Research interests
  - Optimal Control
  - Variational Integrators for Robotics
  - Hybrid Systems