

# Uncovering the Lagrangian of a system from discrete observations

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# Background

## The Problem

Given discrete measurements of a Lagrangian system, can we recover the Lagrangian?

## The Discrete Euler-Lagrange Equations

Given a system with Lagrangian  $L(x, v)$ , we can discretize the action with time step  $\tau$  by summing the **discrete Lagrangian**

$$L_d(x, y) = \tau \cdot L\left(\frac{x + y}{2}, \frac{y - x}{\tau}\right).$$

The principle of stationary action yields the **discrete Euler-Lagrange equations**

$$D_2 L_d(x, y) + D_1 L_d(y, z) = 0,$$

relating any three consecutive points  $x$ ,  $y$ , and  $z$  on a discrete trajectory.

# Recovering the Discrete Lagrangian

## The Problem

Given a pair of points  $(x_0, y_0)$ , we would like to use data points on trajectories that pass nearby to estimate the Taylor expansion of the discrete Lagrangian  $L_d$  at  $(x_0, y_0)$ .

## A Caveat

Two Lagrangians might be **equivalent** in the sense that they yield the same equations of motion.

- For example, if  $L(x, y)$  is a discrete Lagrangian, then the Lagrangian

$$L'(x, y) = \alpha L(x, y) + \beta(y^2 - x^2) + \gamma(y - x) + \delta$$

produces the same discrete Euler-Lagrange equations, for any choice of  $\alpha$ ,  $\beta$ ,  $\gamma$ , and  $\delta$ .

- Trajectory data can't distinguish between equivalent Lagrangians, nor would it be useful to do so.

# A Second-Order Approximation

## Taylor Expansion of the Discrete Lagrangian

Given a pair of points  $(x_0, y_0)$ , we approximate  $L_d$  with its second-degree Taylor polynomial at  $(x_0, y_0)$ . We rewrite it in the form

$$L_d \approx a(x - p)^2 + 2b(x - p)(y - p) + c(y - p)^2 + d_p(x - p) + e_p(y - p) + f_p,$$

where  $p = (x_0 + y_0)/2$ . The expression in higher dimensions is analogous.

## Discrete Euler-Lagrange Equations

For three consecutive points  $x$ ,  $y$ , and  $z$  on a trajectory, we can apply the discrete Euler-Lagrange equations  $D_2L_d(x, y) + D_1L_d(y, z) = 0$  to find

$$0 \approx 2(a + c)(y - p) + 2b(x - p + z - p) + (d_p + e_p).$$

We will use nearby triplets  $(x, y, z)$  from our trajectory measurements to estimate  $a + c$ ,  $b$ , and  $d_p + e_p$ .

# New Parameters for the Lagrangian

## Scaling the Parameters

- In order to estimate the parameters using many data points, we need to assign weights to the parameters appropriately.
- The parameter  $d_p + e_p$  has different units from  $a + c$  and  $b$ .
- Taylor approximations of the discrete Euler-Lagrange equations suggest that an appropriate rescaling of the parameters at  $(x_0, y_0)$  is

$$A := (a + c) \|y_0 - x_0\|, \quad B := 2b \|y_0 - x_0\|, \quad D := d_p + e_p.$$

## Discrete Euler-Lagrange Equations

For three consecutive points  $(x, y, z)$  on a trajectory, we have

$$0 \approx A(2y - x_0 - y_0) + B(x + z - x_0 - y_0) + D \|y_0 - x_0\|.$$

# Estimating the Parameters

## Discrete Euler-Lagrange Equations

For three consecutive points  $(x, y, z)$  on a trajectory, we have

$$0 \approx A(2y - x_0 - y_0) + B(x + z - x_0 - y_0) + D \|y_0 - x_0\|.$$

## Estimating Lagrangian Parameters with Several Data Points

Given data of consecutive triplets  $(x_i, y_i, z_i)$ , we estimate  $A$ ,  $B$ , and  $D$  up to scaling at a point  $(x_0, y_0)$  as follows.

- Construct a matrix  $M$  whose rows are

$$w_i \cdot (2y_i - (x_0 + y_0) \quad x_i + z_i - (x_0 + y_0) \quad \|y_0 - x_0\|).$$

- For the correct values of the parameters, we will have  $M \begin{pmatrix} A \\ B \\ D \end{pmatrix} \approx 0$ .
- Estimate  $A$ ,  $B$ , and  $D$  by the eigenvector corresponding to the least eigenvalue of  $M^T M$ .

# Assigning Weights to the Data Points

## The matrix of coefficients

The  $i$ th row of  $M$  is

$$w_i \cdot (2y_i - (x_0 + y_0) \quad x_i + z_i - (x_0 + y_0) \quad \|y_0 - x_0\|).$$

## Distance

We define the distance between  $(x_0, y_0)$  and  $(x, y)$  to be

$$\delta((x_0, y_0), (x, y))^2 = \left\| \frac{x+y}{2} - \frac{x_0+y_0}{2} \right\|^2 + \tau_s^2 \left\| \frac{y-x}{\tau} - \frac{y_0-x_0}{\tau} \right\|^2,$$

where  $\tau_s$  is a parameter and  $\tau$  is the timestep.

## Weights

$$w_i = \exp \left( -\frac{1}{2\sigma^2} (\delta((x_0, y_0), (x_i, y_i))^2 + \delta((x_0, y_0), (y_i, z_i))^2) \right),$$

where  $\sigma$  is another parameter.

# The Simple Pendulum

## The Lagrangian

$$L_d(x, y) = \tau \left( \frac{1}{2} \left( \frac{y - x}{\tau} \right)^2 - \left( 1 - \cos \left( \frac{x + y}{2} \right) \right) \right).$$

## True Values of Lagrangian Parameters

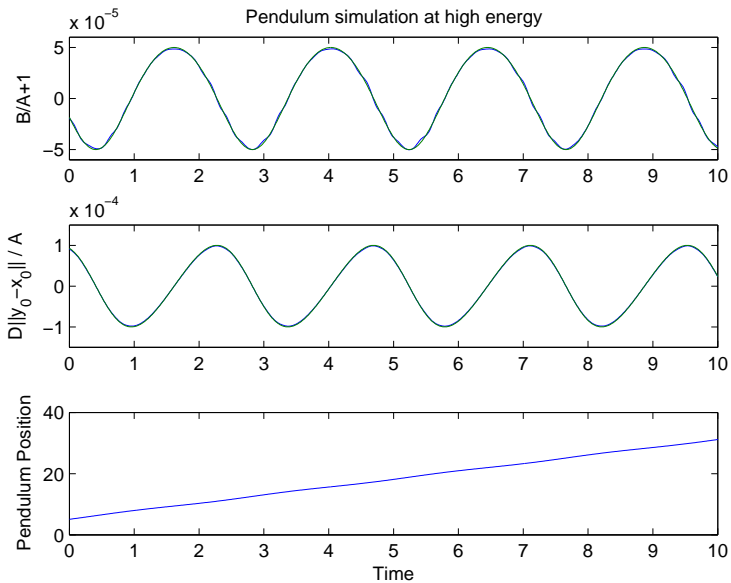
Using a Taylor approximation to the Lagrangian, we find that

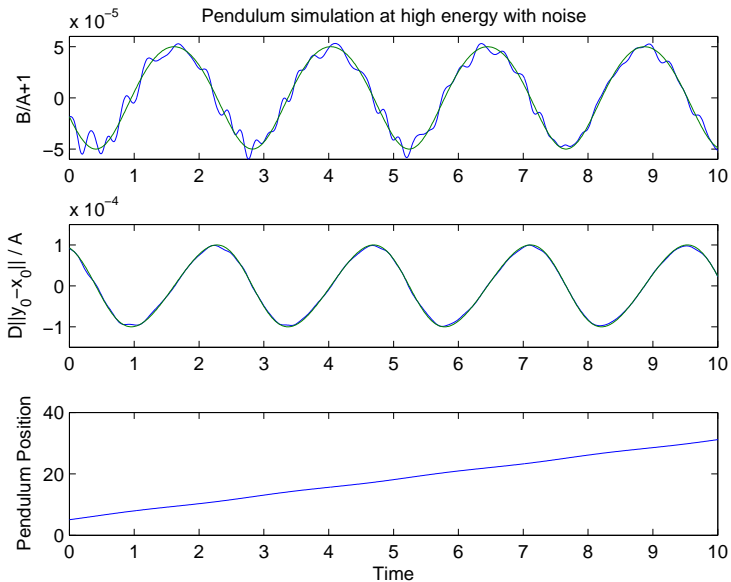
$$\frac{B}{A} = -\frac{4 + \tau^2 \cos \left( \frac{x_0 + y_0}{2} \right)}{4 - \tau^2 \cos \left( \frac{x_0 + y_0}{2} \right)}, \quad \frac{D}{A} = -\frac{4\tau^2}{\|y_0 - x_0\|} \cdot \frac{\sin \left( \frac{x_0 + y_0}{2} \right)}{4 - \tau^2 \cos \left( \frac{x_0 + y_0}{2} \right)}.$$

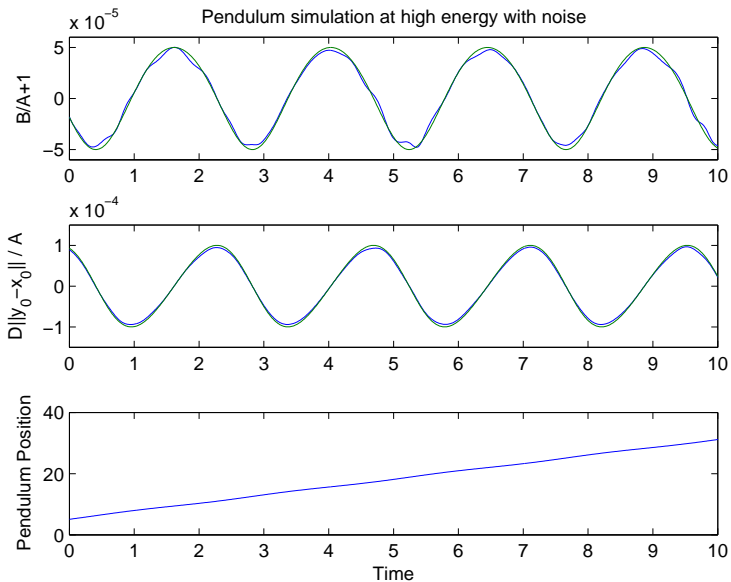
## Parameters Computed From Trajectories

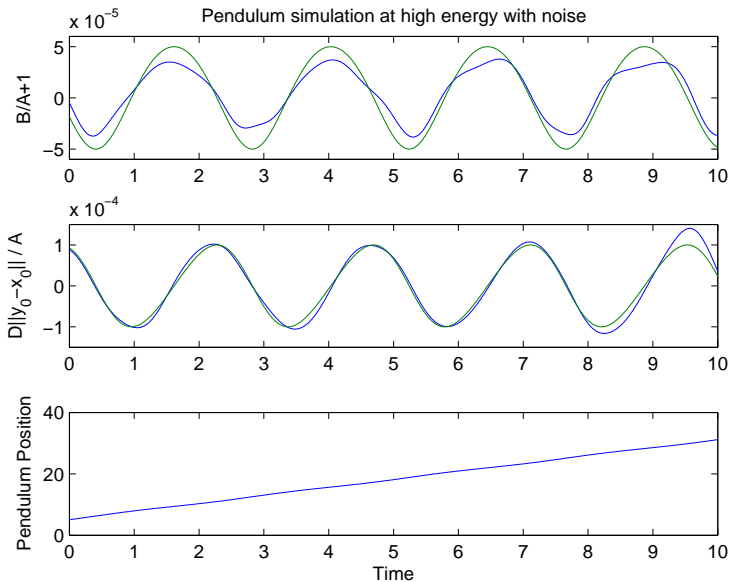
I computed the parameters from the trajectories with Matlab. The graphs of  $\frac{B}{A} + 1$  and  $\frac{D}{A} \|y_0 - x_0\|$  are on the following slides.











## Future Directions

- Recover the Lagrangian from data of several trajectories, and then use it to predict new trajectories.
- Investigate the best choices for  $\tau_S$  and  $\sigma$ .
- Try adding other kinds of noise to the system.
- Try the method with real data.



Evan S. Gawlik, Patrick Mullen, Dmitry Pavlov, Jerrold E. Marsden, and Mathieu Desbrun, *Geometric, variational discretization of continuum theories*, 2010.



Ari Stern and Mathieu Desbrun, *Discrete geometric mechanics for variational time integrators*, *Discrete Differential Geometry: An Applied Introduction*, 2006, pp. 75–80.