

Last Time:

- Controlled dynamics (continuous time)
- Manipulator dynamics
- Equilibria
- Stability (local)

Today:

- Continuous ODEs \rightarrow Discrete time sim
- More on stability

Motivation:

- For general, we can't solve $\dot{x} = f(x)$ for $x(t)$
- Computationally, need to represent $x(t)$ with discrete values
- Discrete-time models can capture some effects that continuous ODEs can't

Discrete-time Dynamics:

* "Explicit" Form:

$$x_{n+1} = f_n(x_n, u_n)$$

↖ "discrete"

- Simplest discretization:

$$x_{n+1} = x_n + h f(x_n, u_n)$$

↑ time step
fd

≈ "Forward
Euler
Integration"

- Pendulum Simulation

$$l = m = 1, \quad h = 0.1, \quad 0.01$$

(flows up)

Stability of Discrete-time Systems:

- Remember, in continuous time:

$$\operatorname{Re}[\operatorname{eig}\left(\frac{\partial f}{\partial x}\right)] < 0 \Rightarrow \text{stable}$$

- In discrete time, dynamics is an iterated map:

$$x_N = f_a(f_a(f_a(\dots f_a(x_0))))$$

- Linearize + apply chain rule:

$$\frac{\partial x_N}{\partial x_0} = \frac{\partial f_d}{\partial x} \frac{\partial f_a}{\partial x} \frac{\partial f_d}{\partial x} \dots \frac{\partial f_d}{\partial x} \Big|_{x_0} = A_d^N$$

- Assume we set up coordinates so equilibrium is at $x_0 = 0$

$$\text{stable} \Rightarrow \lim_{n \rightarrow \infty} A_d^k x_0 = 0 \quad \forall x_0$$

$$\Rightarrow \lim_{n \rightarrow \infty} A_d^n = 0$$

$$\Rightarrow |\operatorname{eig}(A_d)| < 1$$

(inside unit circle)

- Let's apply this to pendulum + forward Euler

$$x_{n+1} = \underbrace{x_n + h f(x_n)}_{f_d(x_n)}$$

$$A_d = \frac{\partial f}{\partial x_n} = I + h A = I + h \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} \cos(\theta) & 0 \end{bmatrix}$$

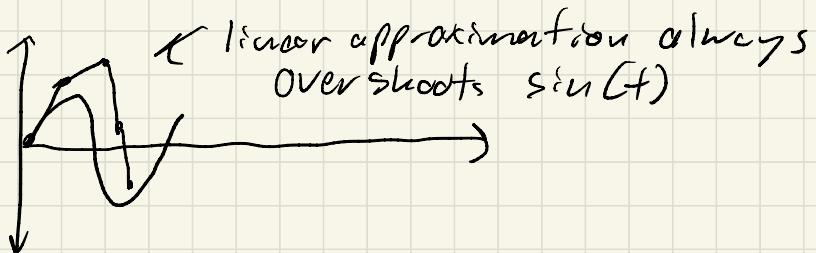
$$\operatorname{eig}(A_d|_{\theta=0}) \approx 1 \pm 0.3i$$

$(h = 0.1)$

- Plot $|\operatorname{eig}(A_d)|$ vs. h

\Rightarrow Only (marginally) stable in [unit $h \rightarrow 0$]

- Intuition:



Take Away Messages:

- Be careful when discretizing ODEs
- Sanity check based on e.g. energy, momentum behavior (conservation and/or dissipation)
- Don't use forward Euler Integration!

A better explicit Integrator

- 4th order Runge Kutta method ("industry standard")
- Intuition:
 - Euler fits a line segment over each time step
 - RK4 fits a cubic polynomial
⇒ much better accuracy!
- Pseudo Code:

$$x_{n+1} = f_{x_n}(x_n)$$

$$k_1 = f(x_n)$$

$$k_2 = f(x_n + \frac{h}{2} k_1)$$

$$k_3 = f(x_n + \frac{h}{2} k_2)$$

$$k_4 = f(x_n + h k_3)$$

$$x_{n+1} = x_n + \frac{h}{6} (k_1 + 2k_2 + 2k_3 + k_4)$$

- Take away:

- Accuracy win \gg additional compute cost
 - Even sophisticated integrators have issues
- \Rightarrow Always sanity check!

* "Implicit" form:

$$f_d(x_{n+1}, x_n, u_n) = 0$$

- Simplest version:

$$x_{n+1} = x_n + h f(x_{n+1}) \approx \text{"Backward Euler"}$$

evaluate f at future time

- How do we simulate?

· Write as:

$$f_d(x_{n+1}, x_n, u_n) = x_n + h f(x_{n+1}) - x_{n+1} = 0$$

- Solve root-finding problem for x_{n+1}

more on this next week

- Pendulum Sim:

- Opposite energy behavior from forward Euler

- Discretization adds damping
- While unphysical, this effect allows simulators to take big steps and is often convenient
 - ⇒ very common in low-f. simulators in graphics/robotics

* Take Aways:

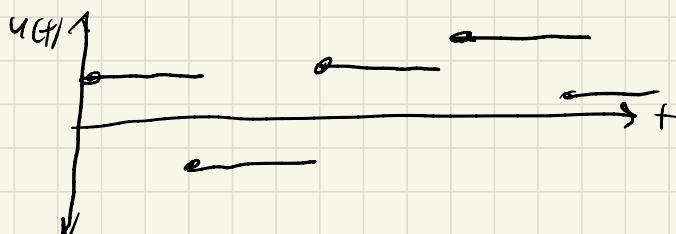
- Implicit methods are often "more stable" than
- For forward simulation, solving explicit equations can be more expensive
- In many "direct" traj opt methods they're not more expensive.

Discretizing Controls:

- So far we've discretized $\dot{x}(t)$
- We also have $u(t)$

* Simplest option:

$$u(t) = u_n, t_n \leq t < t_{n+1} \in \text{"zero-order hold"}$$



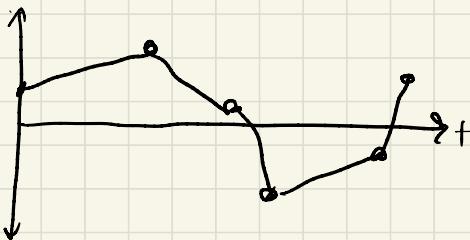
- easy to implement

- May require lots of knot points to accurately capture continuous $u(t)$

* (Possibly) Better Option

$$u(t) = u_n + \left(\frac{u_{n+1} - u_n}{h} \right) (t - t_n)$$

↖ "First-order hold"



- Can approximate continuous $u(t)$ with fewer knot points

- Not much extra work vs. zero-order hold

- Super common (e.g. classic DTRCOL)

* Other Options

- We can keep playing this game with higher-order polynomials

- In many control applications $u(t)$ is not smooth (e.g. bang-bang). Therefore high-order polynomials are not good approximations

⇒ Zero-order and first-order hold
are most common in practice