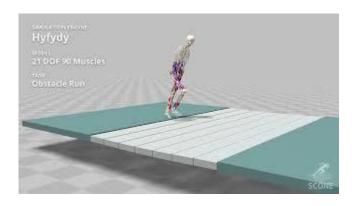
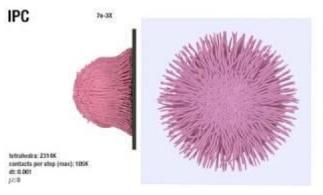
How does simulation work?

Thousands of things we can do in a simulation





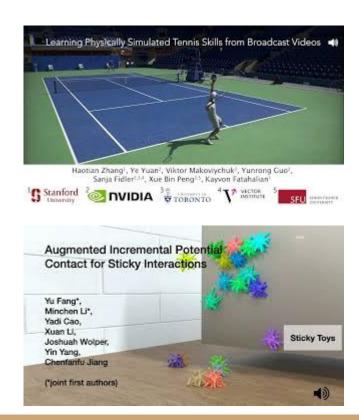


Table of Contents

- State parametrization how to describe a system?
- 2. **Time Evolution** how to describe how the system is changing?
- 3. **Numerical Integration** why and how to approximate time evolution?

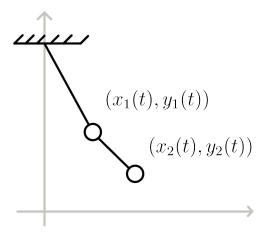
State Parameterisation

How to describe state of a system?

State Parameterisation - General Idea

<u>General idea:</u> choose a set of parameters (coordinates) \rightarrow values of these parameters determine how the state of the system changes in time

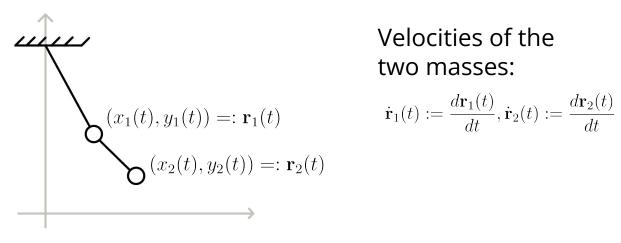
Example - Double Pendulum and Cartesian Coordinates:



Derivatives: Velocity, Acceleration, ...

We care not only about the current values, but also how they change: **velocity**, **acceleration** (or higher order derivatives)

Example - Double Pendulum and Cartesian Coordinates:



Velocities of the two masses:

$$\dot{\mathbf{r}}_1(t) := \frac{d\mathbf{r}_1(t)}{dt}, \dot{\mathbf{r}}_2(t) := \frac{d\mathbf{r}_2(t)}{dt}$$

Generalized Coordinates - Holonomic Constraints

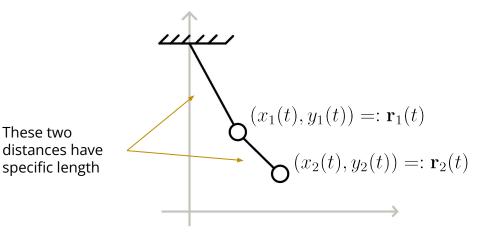
Usually the system has to satisfy **constraints**, i.e. the state of the system is somehow restricted

Holonomic constraints \rightarrow do not depend on the derivatives:

These two

$$f(\mathbf{r}_k(t), t) = 0$$

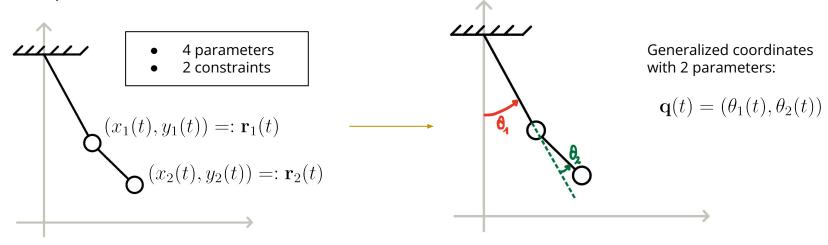
Example:



May in general change in time! E.g. changing the length of a pendulum

Generalized Coordinates - Degrees of Freedom

- Holonomic constraints → we do not need all parameters
- Degrees of Freedom (DoF) = # parameters # holonomic constraints
- Generalized coordinates → choose parameters s.t. #parameters = #DoF

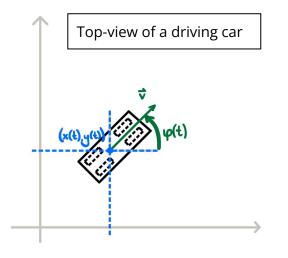


Time Evolution

How to describe how the system is changing?

System Evolution - Constraints

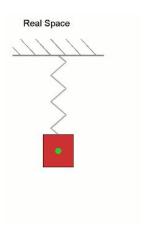
A changing system may have to satisfy more sophisticated constraints \rightarrow non-holonomic constraints: $g(\mathbf{q}(t), \dot{\mathbf{q}}(t), t) = 0$



$$\mathbf{q}(t) = (x(t), y(t), \phi(t)) \implies \dot{\mathbf{q}}(t) = ||\mathbf{v}||(\hat{\mathbf{x}}\cos\phi(t) + \hat{\mathbf{y}}\sin\phi(t))|$$

Differential Equations to Describe Time Evolution

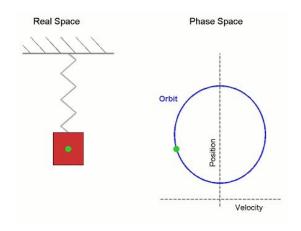
• In general higher order derivatives may describe not only constraints but the evolution of the state in time \rightarrow nth order Ordinary Differential Equations (ODEs)



$$\mathbf{F} = m\mathbf{a} \implies \ddot{\mathbf{x}}(t) = -\frac{k}{m}x(t)$$

Differential Equations to Describe Time Evolution

 In practice we can always translate nth order DE^{qs} to a 1st order ODE by changing the state description



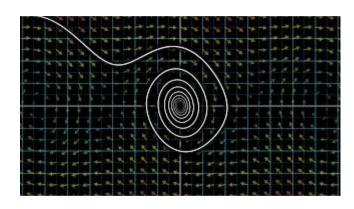
$$\mathbf{q} = egin{bmatrix} x \ v \end{bmatrix}$$
 $\qquad \qquad \qquad \downarrow$ $\qquad \qquad \downarrow$ $\qquad \qquad \dot{\mathbf{q}} = egin{bmatrix} \dot{x} \ \dot{v} \end{bmatrix} = egin{bmatrix} \dot{x} \ \dot{x} \end{bmatrix} = egin{bmatrix} v \ -rac{k}{m}x \end{bmatrix} = egin{bmatrix} 0 & 1 \ -rac{k}{m} & 0 \end{bmatrix} egin{bmatrix} x \ v \end{bmatrix} = A\mathbf{q}$

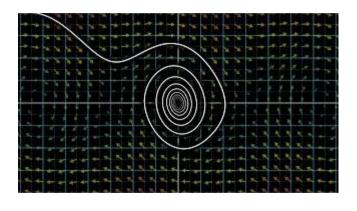
Differential Equations to Describe Physics

So in general motion can be described as 1st order ODE:

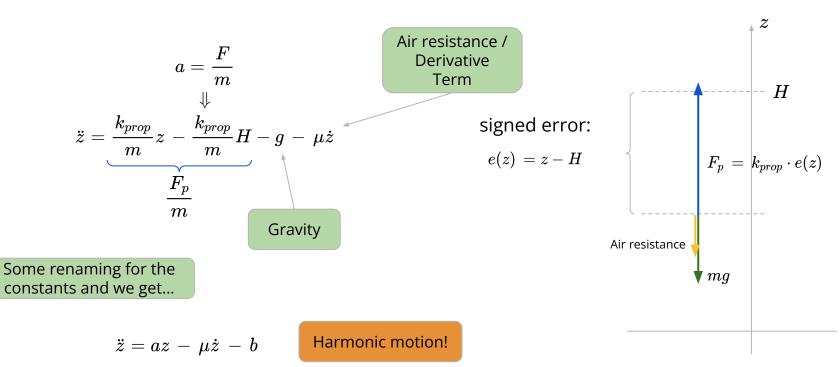
$$\dot{\mathbf{q}} = f(\mathbf{q}(t), t)$$

• Think of a vector field *f* (possibly changing in time) determining how the state changes:

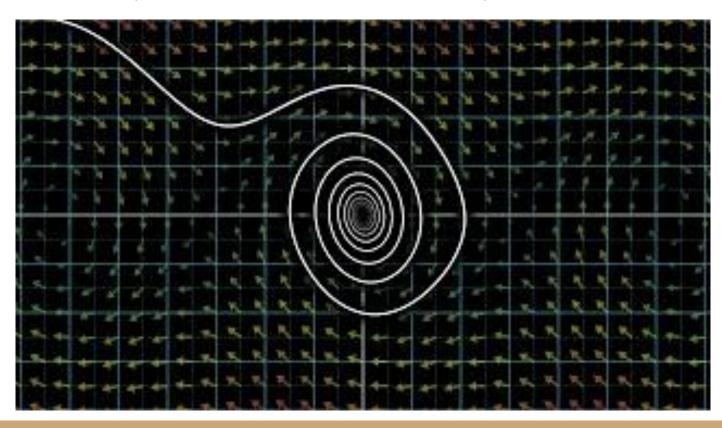




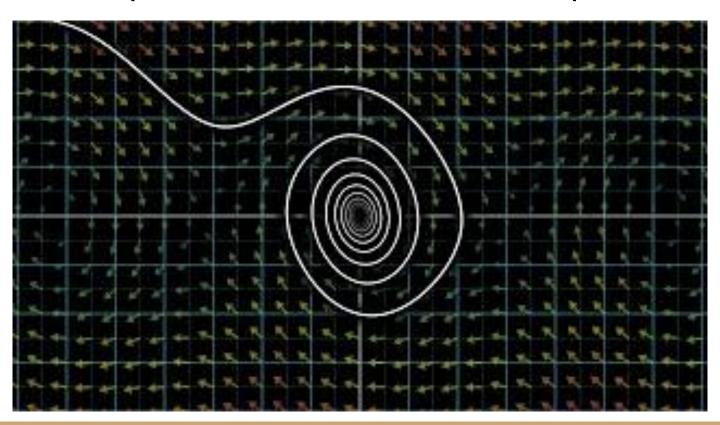
Why bother? Second Example - Drone Control



Third Example - Pendulum - Physics



Third Example - Pendulum - Phase Space



Numerical Integration

How to <u>approximate</u> how the system is changing?

Numerical Integration

- ullet Assume vector field determining system evolution is known, i.e. RHS of: $\dot{f q}=f({f q},t)$
- Approximate derivative
- Determine the next state using obtained approximation

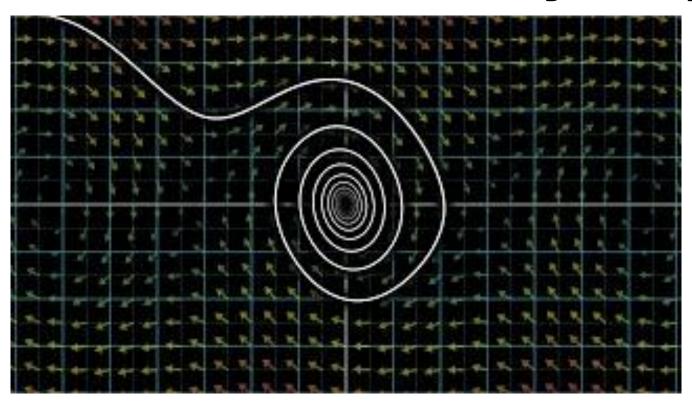
Example - Euler Explicit Integration:

Note that the vector field can be a superposition, e.g. predefined physics + control!

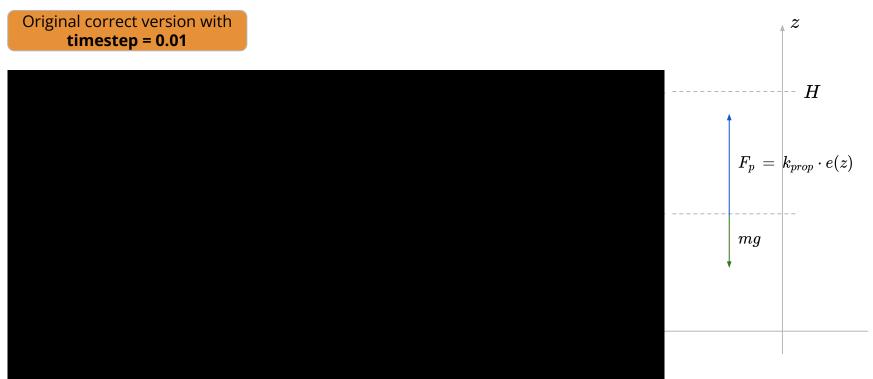
$$f(\mathbf{q}(t),t) \,=\, \mathbf{\dot{q}}(t) pprox \frac{\mathbf{q}(t+\Delta t) - \mathbf{q}(t)}{\Delta t} \implies \mathbf{q}(t+\Delta t) pprox \mathbf{q}(t) \,+\, \Delta t \,\cdot f(\mathbf{q}(t),t)$$

In practice much better approximation schemes are used!

Let's visualise this and see what can go wrong



Example: Wrong Timestep for Drone Control (P)



Example: Wrong Timestep for Drone Control (P)

Timestep increased just to 0.04! HMost recent min: 1.504448389646509 Current altitude: 1.9989670047246892 Most recent max: 1.5055456043845443 Most recent min: 1.504448389646509 Current altitude: 1.9990413752718068 Most recent max: 1.5055456043845443 Most recent min: 1.504448389646509 $F_p = k_{prop} \cdot e(z)$ Most recent min: 1.504448389646509 Current altitude: 1.9991744389899253 Most recent max: 1.5055456043845443 Most recent max: 1.5055456043845443 Most recent min: 1.504448389646509 mqCurrent altitude: 1.9992890326174884 Most recent max: 1.5055456043845443 Current altitude: 1.9993402188327982 Most recent max: 1.5055456043845443 Most recent min: 1.504448389646509 (lab12) → lab12-secret git:(lecture-branch) x

Precision of Numerical Schemes

Remember that we want to approximate the derivative:

$$f(\mathbf{q}(t),t) \,=\, \mathbf{\dot{q}}(t) pprox rac{\mathbf{q}(t+\Delta t) - \mathbf{q}(t)}{\Delta t}$$

Forward Euler

$$\mathbf{q}(t+\Delta t) = \mathbf{q}(t) + \Delta t \cdot \dot{q}(t) + rac{\Delta t^2}{2!} \ddot{q}(t) + rac{\Delta t^3}{3!} \ddot{q}^{\cdot}(t) + \dots \ rac{\mathbf{q}(t+\Delta t) - \mathbf{q}(t)}{\Delta t} = \dot{q}(t) + o(\Delta t)$$

Precision of Numerical Schemes

Remember that we want to approximate the derivative:

$$f(\mathbf{q}(t),t) \,=\, \mathbf{\dot{q}}(t) pprox rac{\mathbf{q}(t+\Delta t) - \mathbf{q}(t)}{\Delta t}$$

Backward Euler

$$egin{aligned} \mathbf{q}(t-\Delta t) &= \mathbf{q}(t) - \Delta t \cdot \dot{q}(t) + rac{\Delta t^2}{2!} \ddot{q}(t) - rac{\Delta t^3}{3!} \ddot{q}^{\cdot}(t) + \dots \ rac{\mathbf{q}(t) - \mathbf{q}(t-\Delta t)}{\Delta t} &= \dot{q}(t) + o(\Delta t) \end{aligned}$$

Precision of Numerical Schemes

Remember that we want to approximate the derivative:

$$f(\mathbf{q}(t),t) \,=\, \mathbf{\dot{q}}(t) pprox rac{\mathbf{q}(t+\Delta t) - \mathbf{q}(t)}{\Delta t}$$

Central Difference

$$\mathbf{q}(t+\Delta t) = \mathbf{q}(t) + \Delta t \cdot \dot{q}(t) + \frac{\Delta t^2}{2!} \ddot{q}(t) + \frac{\Delta t^3}{3!} \ddot{q}(t) + \dots$$
 $\mathbf{q}(t-\Delta t) = \mathbf{q}(t) - \Delta t \cdot \dot{q}(t) + \frac{\Delta t^2}{2!} \ddot{q}(t) - \frac{\Delta t^3}{3!} \ddot{q}(t) + \dots$
 $\mathbf{q}(t+\Delta t) - \mathbf{q}(t-\Delta t) = 2\Delta t \cdot \dot{q}(t) + 2\frac{\Delta t^3}{3!} \ddot{q}(t) + \dots$
 $\frac{\mathbf{q}(t+\Delta t) - \mathbf{q}(t-\Delta t)}{2\Delta t} = \dot{q}(t) + o(\Delta t^2)$

More on these topics

AKA References

References

- Lectures on numerical differentiation and integration:
 https://www.youtube.com/playlist?list=PLMr|AkhleNNTYaOnVI3QpH7igULnAmvPA
- https://research.nvidia.com/labs/toronto-ai/vid2player3d/
- https://hungyuling.com/character-controllers-motion-vaes/
- https://vladlen.info/publications/continuous-character-control-with-low-dimensional-embeddings/
- https://robotic-pretrained-transformer.github.io/
- https://gengshan-y.github.io/ppr/
- https://www.albertpumarola.com/research/D-NeRF/index.html