# CS5643 02 Systems of particles

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### Kinematics of a particle

### Described completely by its position $\mathbf{x} \in \mathbb{R}^d$ where d = 2 or 3

### Particle has no extent, no orientation, etc.

but we can model these properties later by using multiple interacting particles...

### This is animation, so the particles are moving!

- We'll use the notation of dots for time derivatives so  $\dot{\mathbf{x}}$  is the velocity of the particle
- Because velocity is important we also give it the name v

X/+) X=| y(t))

$$V = \dot{X} = \begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot$$

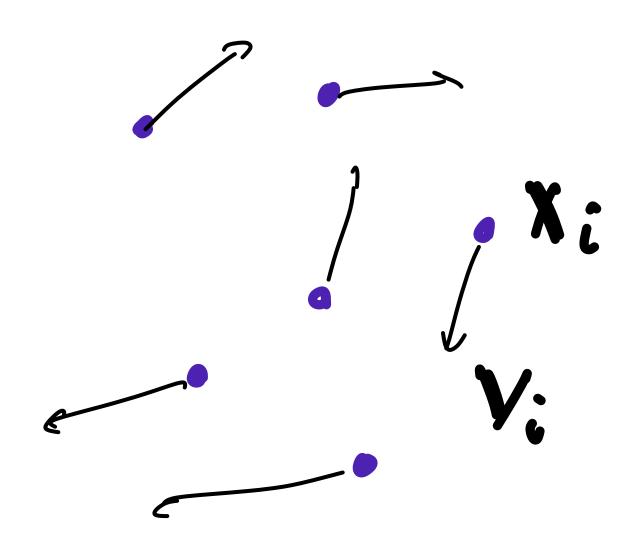
### Systems of particles

#### Usually there is more than one particle

often thousands or millions!

### In this case we let $\mathbf{x}, \mathbf{v} \in \mathbb{R}^{dN}$ represent the positions and velocities of all the particles

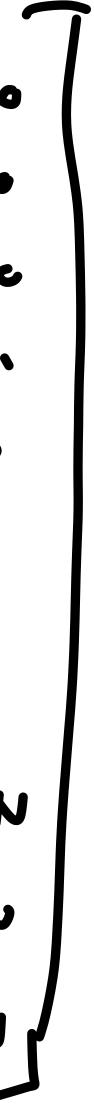
 sometimes N changes as particles are created or deleted (caution: bookkeeping!)





 $V_{\mu}$ 

**y.** 



## Defining particle dynamics

### How to make particles go where we want?

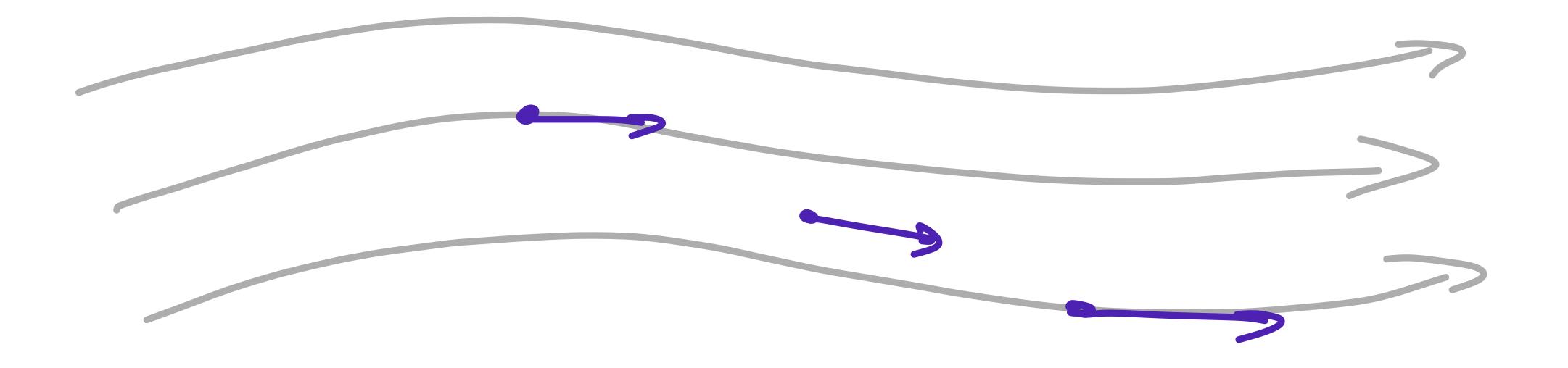
### **0: Script their position directly**

- control  $\mathbf{x}(t)$  directly
- need a function that takes a time and returns a position
- this implicitly sets v because it's the derivative of x:  $v(t) = \dot{x}(t)$
- often used to control a particle that interacts with others
  - e.g. hold up one corner of a sheet
  - e.g. let the user control one particle to push others away e.g. attach one end of a rope to some game element

## Defining particle dynamics

1: Set particle velocity as a function of position

- control  $\mathbf{v}(t) = \mathbf{v}(t, \mathbf{x}(t))$
- need initial position, then particle path is the solution to an initial value problem
  - general form y'(t) = f(t, y(t)) a first-order ordinary differential equation (ODE)
- need a function that takes a position and returns a velocity (don't need to store particle velocity)
- can do the simulation with various methods for solving ODEs







### First-order dynamics: advection

particles are advected by the velocity field of a (fake) fluid flow

- no notion of particle mass or inertia here (e.g. dust or smoke particles, or fluffy snowflakes) for elaborate examples flow fields can come from fluid simulations
- more commonly flow fields are generated procedurally

#### generating plausible flow fields

- randomly generated vector fields tend to have sources and sinks
- particles will soon accumulate at the sinks not a nice swirling motion!
- key is generating divergence free velocity fields:  $\nabla \cdot \mathbf{v} = 0$
- handy fact: if your velocity field is the curl of something, it has zero divergence
- so cook up any potential  $\phi(\mathbf{x})$  you like, then define  $\mathbf{v} = \nabla \times \phi$
- demo in a bit!



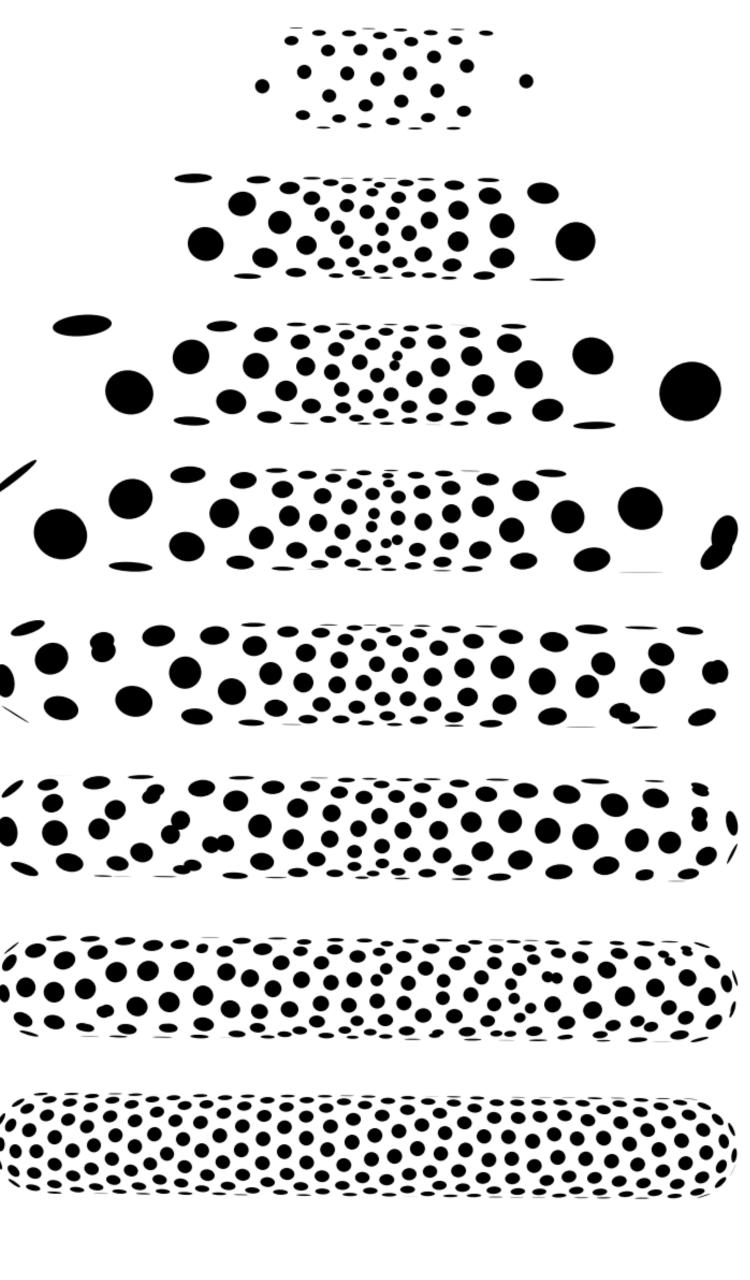
### First-order dynamics: optimization

#### Sometimes the point of particle evolution is the end state

Use particle systems to achieve uniform spacing or other desired arrangements

#### In this case we can use gradient flow on a loss function:

- $\cdot \dot{\mathbf{x}}(t) = -\nabla L(\mathbf{x}(t))$
- that way the value of  $L(\mathbf{x}(t))$  always decreases with time, until **x** reaches a stationary point
- note that L might involve interactions between particles so that  $\dot{\mathbf{x}}_i$  might depend on various  $\mathbf{x}_i$



Witkin and Heckbert, "Using Particles to Sample and Control Implicit Surfaces," SIGGRAPH 94.

## Defining particle dynamics

### 2: Set particle acceleration as a function of position (and maybe velocity)

- control  $\dot{\mathbf{v}}(t) = \mathbf{a}(t, \mathbf{x}(t), \mathbf{v}(t))$
- need initial position and velocity and then future path is determined
- need a function that takes system state and computes the acceleration

### Newtonian mechanics leads to a second order formulation like this

- for each particle,  $\mathbf{f}_i(t, \mathbf{x}_i(t), \mathbf{v}_i(t), \ldots) = m_i \dot{\mathbf{x}}(t)$  (ellipsis because forces could depend on other particles)
- for the whole system  $\mathbf{f}(t, \mathbf{x}(t), \mathbf{v}(t)) = M \ddot{\mathbf{x}}(t)$  (and for a particle system M is diagonal)

### Forces for particles: unary

gravity (constant)

•  $f_i = m_i \mathbf{g}$  (doesn't depend on anything at all...)

drag (depends on own velocity)

•  $f_i = -k\mathbf{v}_i$  (viscous drag, a reasonable model for slow moving objects)

• 
$$f_i = -kv_i \mathbf{v}_i = -kv_i^2 \hat{\mathbf{v}}_i$$
 , where  $v = |\mathbf{v}|$ 

#### anchored spring with zero rest length (depends on own position)

•  $f_i = -k(\mathbf{x} - \mathbf{p})$  (spring anchored at  $\mathbf{p}$ )

gravitational field (depends on own position)

$$f_i = -\frac{GMm_i}{|\mathbf{x}|^2} \hat{\mathbf{x}} = -\frac{GMm_i}{|\mathbf{x}|^3} \mathbf{x} \text{ (for mass )}$$



- (aerodynamic drag, a reasonable model for fast objects)

- M fixed at origin)  $f_i = -GMm_i \frac{\mathbf{x} \mathbf{p}}{|\mathbf{x} \mathbf{p}|^3}$  (M at **p**)



### Forces for particles: binary

spring with zero rest length

$$\cdot f_i = -k(\mathbf{x}_i - \mathbf{x}_j)$$

gravitational force between two bodies (same form for elecrostatic)  $f_i = -\frac{Gm_im_j}{|\mathbf{x}_i - \mathbf{x}_i|^2} \frac{\mathbf{x}_i - \mathbf{x}_j}{|\mathbf{x}_i - \mathbf{x}_j|} = -Gm_im_j$ 



- $f_i = -k(\mathbf{x}_i \mathbf{x}_i) = -f_i$  (forces on *i* and *j* must be opposite, otherwise there is an external force)

$$\frac{\mathbf{x}_i - \mathbf{x}_j}{|\mathbf{x}_i - \mathbf{x}_j|^3}$$



### Fun with demos!

#### ballistic particles

- Newtonian particle system (second order)
- initialize with random velocity
- gravity, drag forces

#### springy particles

- Newtonian particle system still
- initialize with random position and velocity
- spring and gravitational forces

#### particle advection

- first order advection
- random vector fields using Perlin noise and curl noise
- line integral convolution for field visualization