



## Course Plan



- lec. 1: **Introduction** ●
- lec. 2: **Mathematics** for 3D Games ●●●●●●●●
- lec. 3: **Scene Graph** ▶▶
- lec. 4: **Game 3D Physics** ▶●●📍●●+●●▶
- lec. 5: **Game Particle Systems** ▶
- lec. 6: **Game 3D Models** ●●▶
- lec. 7: **Game Textures** ▶●●
- lec. 8: **Game Materials** ▶
- lec. 9: **Game 3D Animations** ▶●●●
- lec. 10: **3D Audio** for 3D Games ●
- lec. 11: **Networking** for 3D Games ●
- lec. 12: **Interactive Agents** for 3D Games ●
- lec. 13: **Rendering Techniques** for 3D Games ●

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## Attrition (or friction) forces



- **Isotropic friction force**:  $\vec{f} = -k \vec{v}$ 
  - a force that opposes motion, dissipating energy
  - isotropic = magnitude independent on the direction of motion
  - direction of force: always opposite of current velocity direction
  - Magnitude of force: proportional to the speed =  $||\vec{v}||$
  - note: this force depends on velocity, not position(s).
  - models the resistance of the medium where the motion happens (air, water, thin space...)
  - the denser the medium, the largest  $k$ , stronger the force (water >> air >> thin space)
- **Planar friction forces**:
  - A force that happens when things slide against each other
  - Always parallel to the contact plane (orthogonal to the normal)
  - Part of the collision response (see next topic!)

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## Attrition (or friction) forces:



💡 simulate them with *velocity damping* !

- A simpler way to simulate (isotropic) friction: “velocity damping” (or “viscous damping”)
  - simply reduce all velocity vectors by a fixed proportion
  - for example: scale velocity down by 2% per second (“drag factor” = 0.02 / sec) (that is, scale velocity vectors by a factor 0.98)
  - Why it makes sense: Higher speed = more attrition = more loss of speed. So, attrition = a “fixed tax” (in %) on velocity.
- For planar friction:
  - Split velocity into parallel / orthogonal parts
  - Apply different Drag factors to each parts

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## Velocity Damping: how to (as an example)



- Objective: “reduce speed by 1.5% every second”
- So:
  - After a second:  $\vec{v} \leftarrow (1.0 - 0.015) \vec{v}$
  - After 2 seconds?  $\vec{v} \leftarrow (1.0 - 0.015)^2 \vec{v}$
  - After  $k$  seconds?  $\vec{v} \leftarrow (1.0 - 0.015)^k \vec{v}$
  - After  $dt$  seconds?  $\vec{v} \leftarrow (1.0 - 0.015)^{dt} \vec{v}$   
 e.g.  $1/60 = 0.017$  sec
  - Which can be approximated with  $\vec{v} \leftarrow (1.0 - 0.015 \cdot dt) \vec{v}$
  - The approximation is good when *this* is small

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### Equivalent formulas for friction

(controlled by different constants  $k_0, k_1, k_2$ )

- $\vec{f} = -k_0\vec{v} \Rightarrow \vec{a} = -\frac{k_0}{m}\vec{v} \Rightarrow \vec{v} \leftarrow \vec{v} - \frac{k_0}{m}\vec{v} dt \Rightarrow$   
update:  $\vec{v} \leftarrow \left(1 - \frac{k_0}{m}\right)^{dt} \vec{v}$
- update:  $\vec{v} \leftarrow (1 - k_1)^{dt} \vec{v}$
- update:  $\vec{v} \leftarrow e^{-k_2 \cdot dt} \vec{v}$

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### Equivalent formulas for friction

(controlled by different constants  $k_0, k_1, k_2$ ) - notes

- 1st way:  $k_0$  is the ratio force magnitude / speed.  
Physically based, and auto-adapts with mass,  
but difficult to tune in your game?
- 2nd way:  $k_1$  (i.e.,  $\frac{k_0}{m}$ ) is the velocity loss per second (%).  
Intuitive to set! (here, we will adopt this way).
- 3rd way:  $k_2$  (i.e.,  $-\log(1 - k_1)$ )  
is the exponential decay.  
Also quite used in videogames  
(e.g., in Unity, Godot, Unreal)

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## Velocity Damping: pseudo-code



```
Vec3 position = ...
Vec3 velocity = ...

void initState(){
    position = ...
    velocity = ...
}

void physicsStep( float dt )
{
    Vec3 acceleration = force( positions ) / mass;
    position += velocity * dt;
    velocity += acceleration * dt;
    velocity *= (1.0 - DRAG * dt);
}

void main(){
    initState();
    while (1) do physicsStep( 1.0 / FPS );
}
```

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## Velocity Damping: notes



- Velocity Damping is useful for robustness
  - Prevents the energy to ever increase
- Limitations:
  - it may exaggerate frictions of, e.g., air, especially in absence of contacts
  - it's a crude approximation: attrition forces are not really *linear* with speed
- In practice:
  - low drag: hardly noticeable (in the short run), increases robustness
  - high drag (e.g. 2% per sec): everything feels like to be moving in molasses (ita: *melassa*); everything quickly grinds to a halt
  - super high drag: (e.g. >25% per sec) basically, no inertia anymore. May be useful to converge to (local) minimal energy states: your simulator is now basically a solver for **statics**, not **dynamics**

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## Continuity of pos and vel



- In real Newtonian physics the state (pos and vel) can only change *continuously*
  - No sudden jump!
- In practice, sometimes is useful to artificially break continuity in the simulations
- Discontinuous changes:
  - for positions: “teleports”
  - for velocity: “impulses”
  - In the real world, those variations can well be consequences of forces, but these forces are not modelled as such, in our system

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## Dynamics displacements VS kinematic

a discontinuous  
change of state (position)

$$\dots$$
$$p = p + \vec{v} \cdot dt$$
$$\dots$$

aka **dynamic**  
displacements

Justified  
by physics

$$\dots$$
$$p = p + dp$$
$$\dots$$

aka **Kinematic**  
displacements

Just  
“teleportation”

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## Impulses VS Forces

...

$$\vec{v} = \vec{v} + (\vec{f} / m) \cdot dt$$

...

...

$$\vec{v} = \vec{v} + (\vec{i} / m)$$

...

- **Forces** (continuous)
    - Continuous application
    - every frame
  - **Impulses**
    - Infinitesimal time
    - una tantum

they model very intense but short forces (such as impacts)

a discontinuous change of state (velocity)

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## Impulses VS Forces

- **Force** :
  - it determines an **acceleration**
  - **acc** determines a (continuous!) change of **vel**
  - it's sustained over meaningful periods of time
- **Impulse** :
  - a (**discontinuous!**) change of **vel**
  - in reality: just a force with
    - **very large** magnitude
    - **very short** application times
  - we model it as a force "pre-multiplied" with its application time
  - it's useful to:
    - model phenomena with a time scale  $\ll dt$   
*e.g.* a tennis ball rebounding against a tennis racket
    - control the simulation (direct change of velocity)

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## Impulses VS Forces

● what does *truly* happen when it bounces off the ground?

0 msec      1 msec      2 msec      3 msec      4 msec

- very strong forces (but not infinite)
- applied for a very short time (but not instantaneous)
- see *collision response* later for details about the impulse based approximations

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## Impulses VS Forces

● what does *truly* happen when it bounces off the ground?

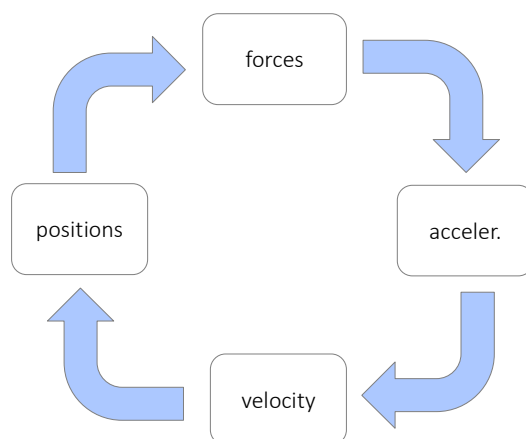
no impact force      huge force      no impact force

$dt$

- This can only be modelled as an *impulse*, not a force
- See also *collision response*, later

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## Next: better integration methods for (Newtonian) dynamics



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## Other numerical integrators ("numerical ways to compute integrals")

- Some commonly used alternatives (among MANY!):
  - "Forward" Euler method (the one seen so far)
  - Symplectic Euler method
  - Leapfrog method
  - Verlet method
- These are just variants of each other – let's see them!
  - From the code point of view, no big change
  - They can differ in accuracy / behavior
  - They can have different "orders of accuracy"
  - Note: a more accurate method is also more efficient (larger  $dt$  are possible, so fewer steps are necessary)

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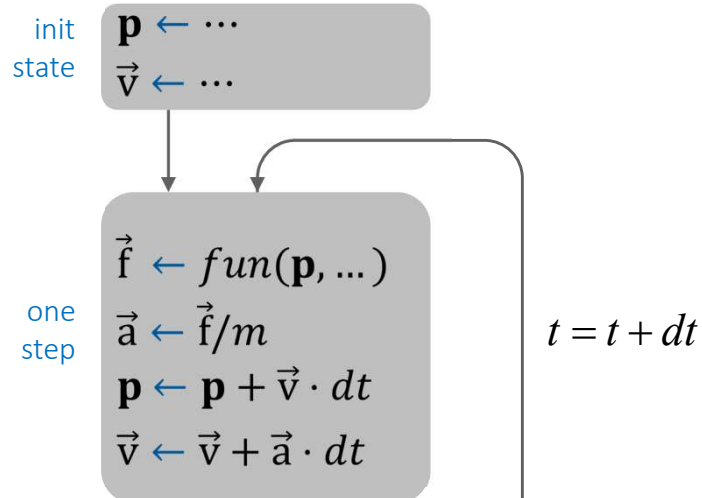
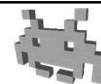
## Forward Euler Method: limitations



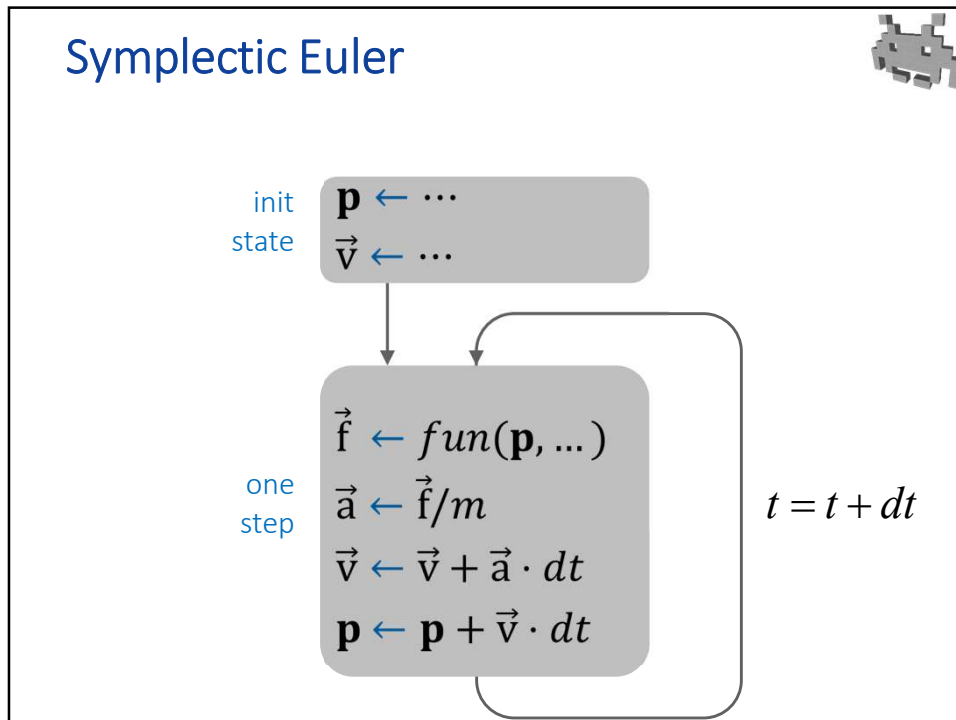
- efficiency / accuracy: not too good
  - error accumulated over time = linear in  $dt$
  - it's only a "first order" method
  - Doubles the steps = halve the  $dt$ , only halves the errors (can be better, but no guarantees)
- scarce stability for large  $dt$
- minor problem: no reversibility, *even in theory*
  - real Newtonian Physics is reversible: flip all velocities and forces  $\Rightarrow$  go backward in time.
  - In our simulation (with Euler): this doesn't work exactly
  - Ability to go reverse a simulation would be useful in games! E.g. replays in a soccer game ?
  - Pro tip: basically, reverse time direction never done like this To go backward in time accurately, store states

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## Forward Euler



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### Forward Euler *pseudo code*

```

Vec3 position = ...
Vec3 velocity = ...

void initState() {
    position = ...
    velocity = ...
}

void physicsStep( float dt )
{
    Vec3 acceleration = compute_force( position ) / mass;
    position += velocity * dt;
    velocity += acceleration * dt;
}

void main() {
    initState();
    while (1) do physicsStep( 1.0 / FPS );
}
    
```

Equivalent to...

$$\vec{\mathbf{f}}_i \leftarrow \text{function}(p_i, \dots)$$

$$\vec{\mathbf{a}}_i \leftarrow \vec{\mathbf{f}}/m$$

$$\vec{\mathbf{v}}_{i+1} \leftarrow \vec{\mathbf{v}}_i + \vec{\mathbf{a}}_i \cdot dt$$

$$p_{i+1} \leftarrow p_i + \vec{\mathbf{v}}_i \cdot dt$$

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## Symplectic Euler *pseudo code* (aka semi-implicit Euler)

```

Vec3 position = ...
Vec3 velocity = ...

void initState(){
    position = ...
    velocity = ...
}

void physicsStep( float dt )
{
    Vec3 acceleration = compute_force( position ) / mass;
    velocity += acceleration * dt;
    position += velocity * dt;
}

void main(){
    initState();
    while (1) do physicsStep( 1.0 / FPS );
}
        
```

Equivalent to...

$$\vec{f}_i \leftarrow \text{function}(p_i, \dots)$$

$$\vec{a}_i \leftarrow \vec{f}/m$$

$$\vec{v}_{i+1} \leftarrow \vec{v}_i + \vec{a}_i \cdot dt$$

$$p_{i+1} \leftarrow p_i + \vec{v}_{i+1} \cdot dt$$

↑

just flip the order

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## Forward Euler:

time:	0 dt	1 dt	2 dt	3 dt	4 dt	5 dt	6 dt	7 dt	...
pos:	*	*	*	*	*	*	*	*	...
vel:	*	*	*	*	*	*	*	*	...
acc:	*	*	*	*	*	*	*	*	...

1 2 3

## Symplectic Euler:

time:	0 dt	1 dt	2 dt	3 dt	4 dt	5 dt	6 dt	7 dt	...
pos:	*	*	*	*	*	*	*	*	...
vel:	*	*	*	*	*	*	*	*	...
acc:	*	*	*	*	*	*	*	*	...

1 2 3

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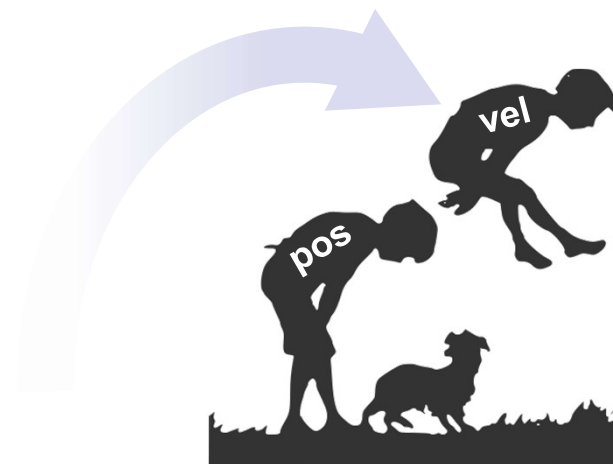
## Forward Euler VS Symplectic Euler (warning: over-simplifications)



- From the code point of view, they are very similar
- The semantics changes:
  - in Symplectic Euler  
the position altered using *next frame* velocity
  - (it's "wrong", in a sense – but tends to work better)
- Similar properties, but better in practice
  - Same order of convergence (still just 1 ☹️)
  - On average,  
Symplectic tends to be more stable and accurate

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## Leapfrog Integration Method



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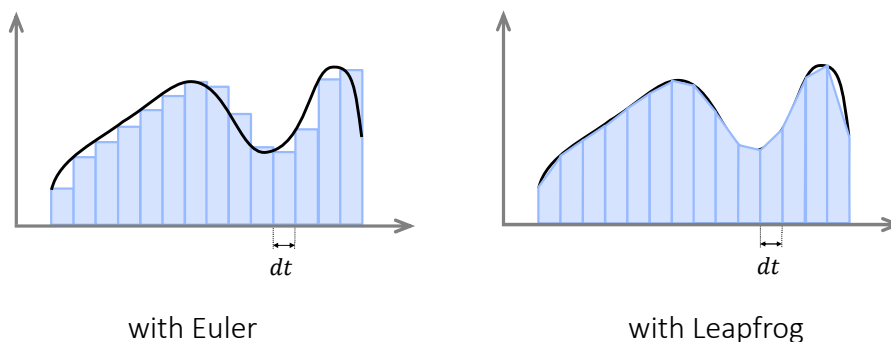
## Leapfrog Integration Method



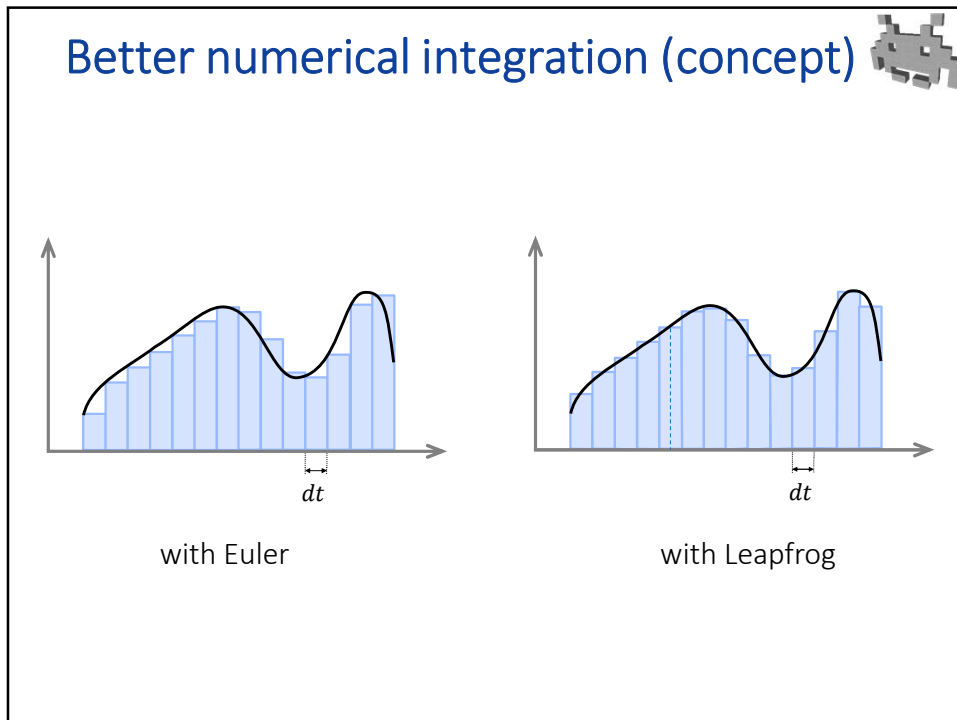
- Basic idea:  
store positions at time  $k \cdot dt$   
(that is, at  $t = 0, 1dt, 2dt, 3dt...$ )  
but store velocities at time  $k \cdot dt + \frac{1}{2} dt$   
(that is, at  $t = 0.5 dt, 1.5 dt, 2.5dt, 3.5dt...$ )
- (roughly) equivalent to use a summatory of the areas of trapezoids,  
(having base  $dt$  and average height  $v(t + \frac{1}{2} dt)$ )  
not rectangles, to compute the integral

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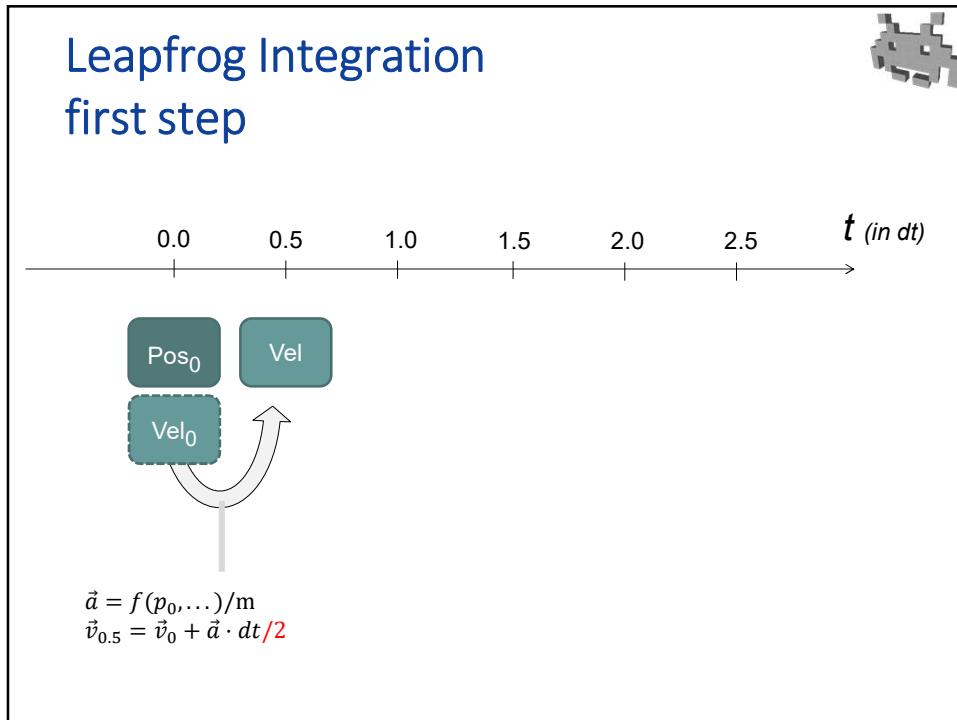
## Better numerical integration (concept)



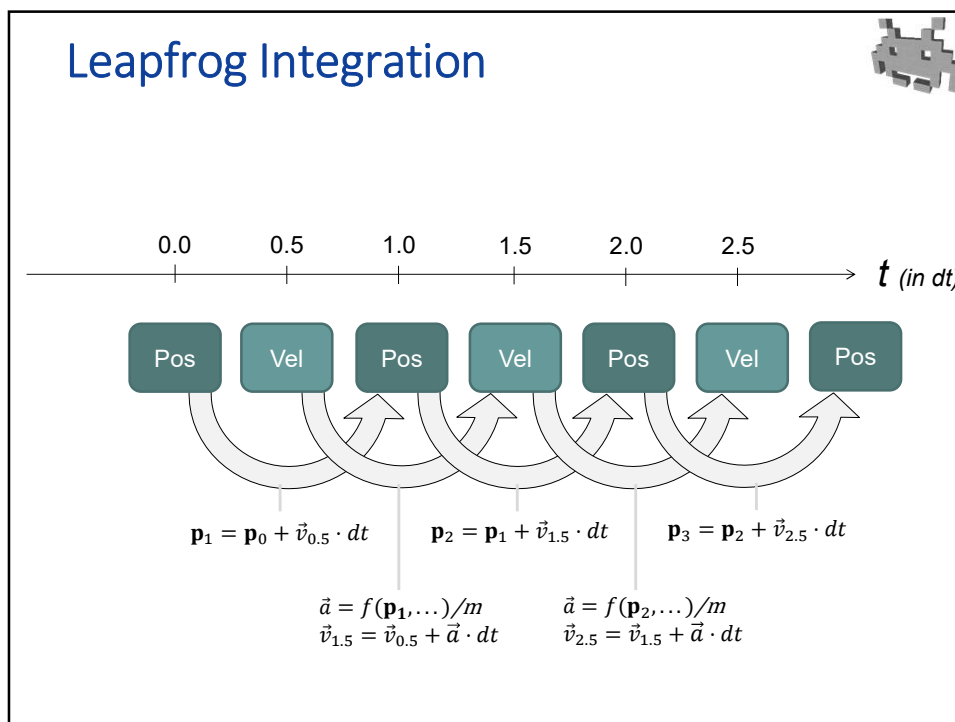
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### Leapfrog method: pros and cons

- Same cost as Euler – and basically same code
  - Velocity stored in status = velocity “half a  $dt$  ago” (and after updating it: “half a frame in the future”)
  - Only real difference: the initialization of velocities
- Better theoretical accuracy, for the same  $dt$ 
  - better asymptotic behavior: it’s a “second order” system instead of first!
  - cumulated error: proportional to  $dt^2$  instead of  $dt$
  - error per frame: proportional to  $dt^3$  instead of  $dt^2$
- Bonus: **fully reversible!**
  - in theory only. Beware of numerical errors.
- But: **requires fixed  $dt$**  during all the simulation
  - Otherwise, updates of vel required in all particles

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## Verlet integration method

- Idea: remove velocity from state  
 Instead, store previous position
- Velocity is now implicit
- It's defined by:
  - current pos  $\mathbf{p}_{now}$
  - last pos  $\mathbf{p}_{old}$   
 which we need to record

$\mathbf{p}_{old} \xrightarrow{\vec{v} \cdot dt} \mathbf{p}_{now}$

$$\mathbf{p}_{now} = \mathbf{p}_{old} + \vec{v} \cdot dt \quad \leftarrow \text{Euler \& variants}$$

$$\Rightarrow$$

$$\vec{v} = (\mathbf{p}_{now} - \mathbf{p}_{old})/dt \quad \leftarrow \text{Verlet}$$

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## Verlet integration method: (modifying Euler integration...)

init state  $\mathbf{p}_{now} = \dots$   
 $\mathbf{p}_{old} = \dots$

one step

$$\vec{f} = \text{funct}(\mathbf{p}_{now}, \dots)$$

$$\vec{a} = \vec{f}/m$$

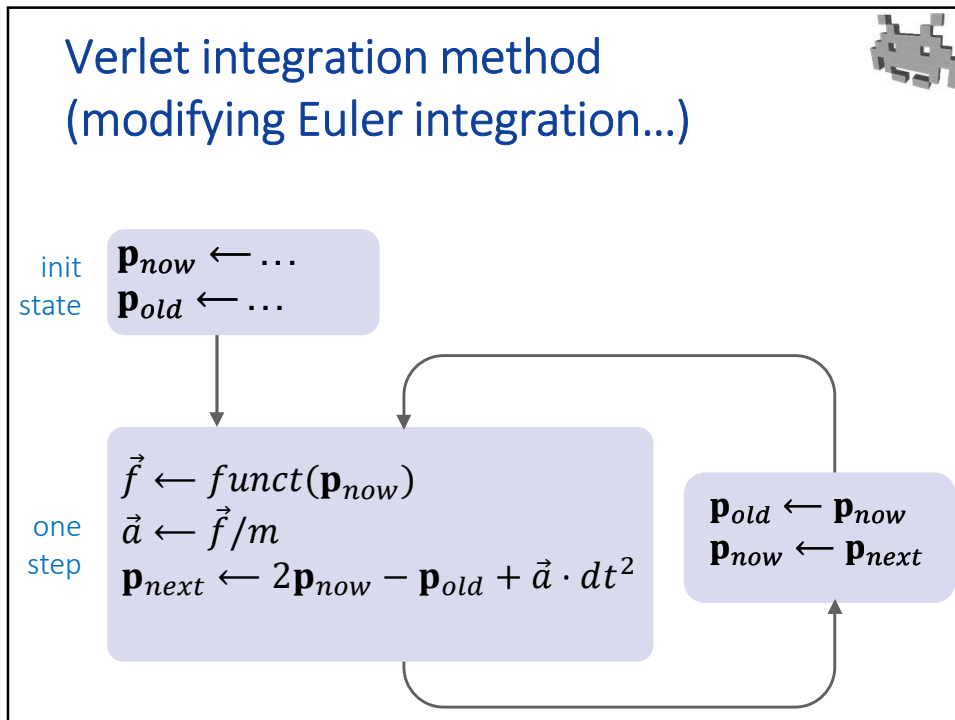
$$\vec{v} = (\mathbf{p}_{now} - \mathbf{p}_{old})/dt$$

$$\vec{v} = \vec{v} + \vec{a} \cdot dt$$

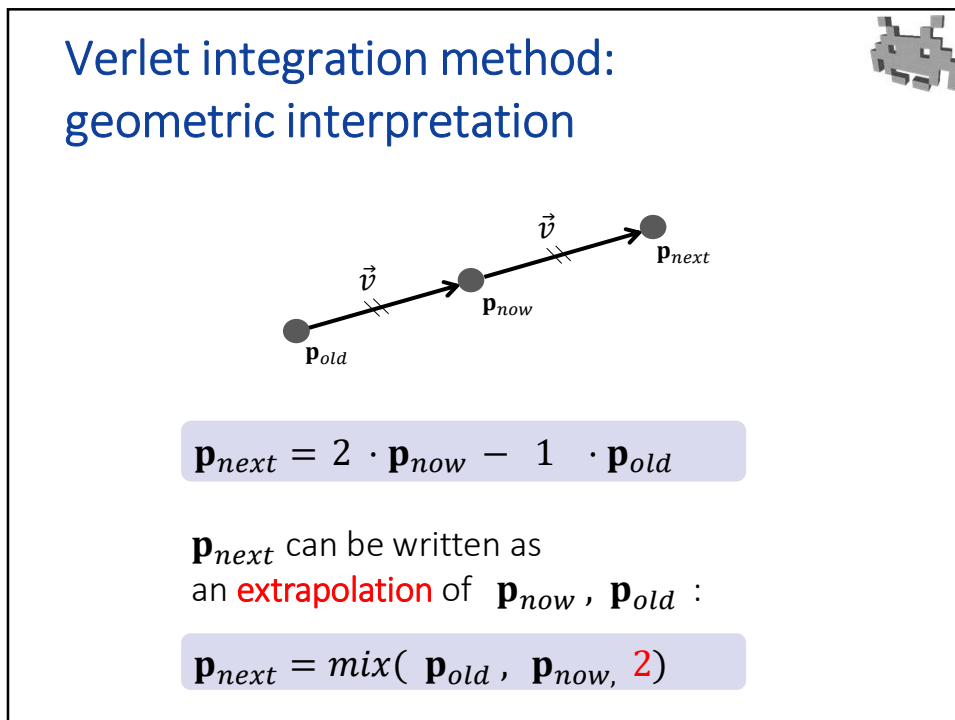
$$\mathbf{p}_{next} = \mathbf{p}_{now} + \vec{v} \cdot dt$$

expanding this...

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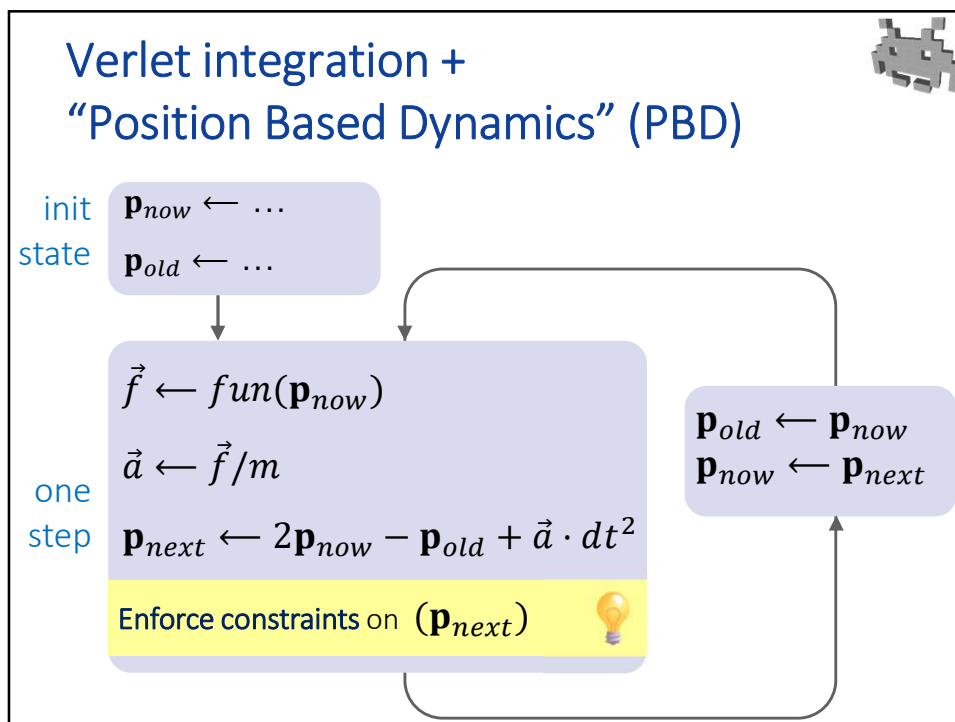
## Verlet: characteristics



- Velocity is kept implicit
  - but that doesn't save RAM:  
we need to store previous position instead
  - (a point instead of a vector: same memory)
- Good efficiency / accuracy ratio
  - accumulated error: order of  $dt^2$  (second order method)
- Extra bonus: reversibility
  - it's possible to go backward\* in  $t$  and reach the initial state from any state
  - \* (just swap  $p\_now$  and  $p\_old$ )
  - only in theory... careful with implementation details

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## Verlet integration + "Position Based Dynamics" (PBD)



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## Position Based Dynamics (PDB)



- A **positional constraint** is an equality/inequality involving the *positions* of particles.
  - Useful, for example, to model consistency conditions
  - Like “*solid objects don’t compenetrates each other*”, or “*steel bars won’t become shorter or longer than they are*”
  - We will see many examples
- 💡 We **enforce** (impose) positional constraint directly by displacing the *positions* of particles
  - Thanks to Verlet: this displacement automatically causes some appropriate update of the velocity!
  - it’s not necessarily correct, but it’s plausible and robust

a formula  
with ‘=’ ‘>’ ‘<’ etc.

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## Example of a positional constraint



«I want all particles to stay above ground  
(that is, their y must never be negative) »

Enforce this constraint: trivial!

```
for (each particle i)
{
    if (p[i].y < 0) p[i].y = 0;
}
```



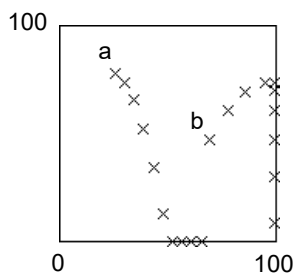
Imposing constraints like this one is a first part of **collision response**.  
For re-bounces, **impulses** must still be added (see collisions).

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## Example of a positional constraint (here, in 2D physics)



«I want particles to stay  
inside a 2D box [0 – 100] x [0 – 100] »



Enforce this constraint: simple clamp!

```
for (each particle i)
{
  p[i].x = clamp( p[i].x, 0, 100 );
  p[i].y = clamp( p[i].y, 0, 100 );
}
```



Imposing constraints like this one is a first part of **collision response**.  
For re-bounces, **impulses** must still be added (see collisions).

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## Verlet + Position Based Dynamics. Advantages



- **flexibility**: different constraints can be used to model many different phenomena
  - Useful constraints are straightforward to define
  - They are easy to impose (they involve only few particles)
  - They can be used to model many possible phenomena
  - See following slides for examples
- **robustness** : plausibility is ensured by *explicitly* enforcing the conditions we want to see
  - For example: a ball won't ever be seen outside the box containing it – and it will also recover from mistakes
- No forces / impulses are needed to enforce any such consistency conditions

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## Verlet: *caveats*

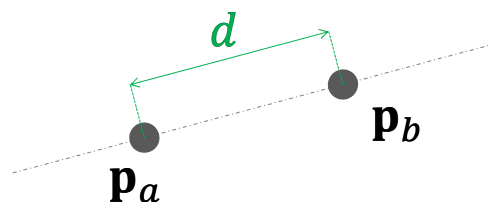
(see next slides for solutions)

- ⚠ it assumes a constant  $dt$  (time-step duration)
  - if  $dt$  varies: corrections are needed! (how?)
- ⚠ Q: how to act on **velocity** (which is now implicit)?
  - for example, how to apply **impulses** ?
  - A: change  $\mathbf{p}_{old}$  instead (how?)
- ⚠ Q: how to act of **positions** w/o impacting velocity?
  - for example, to apply **teleports** / **kinematic motions** ?
  - A: change both  $\mathbf{p}_{new}$  and  $\mathbf{p}_{old}$  (how?)
- ⚠ Q: how to apply **velocity damps**?

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## Example of positional constraint: equidistance constraint

«Particles  $\mathbf{a}$  and  $\mathbf{b}$  must stay at a fixed distance  $d$  »



I require that...  $\|\mathbf{p}_a - \mathbf{p}_b\| = d$

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### Enforce equidistance constraints (assuming equal masses for now)

if  $\|\mathbf{p}_a - \mathbf{p}_b\| > d$

if  $\|\mathbf{p}_a - \mathbf{p}_b\| < d$

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### Enforce equidistance constraints: pseudo code

```
Vector3 pa, pb; // curr positions of a,b
float d;        // distance (to enforce)

Vector3 v = pa - pb;
float currDist = v.length;

v /= currDist; // normalization of v

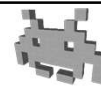
float delta = currDist - d ;

pa += ( 0.5 * delta ) * v;
pb -= ( 0.5 * delta ) * v;
```

↑ we move each particle *half the way*  
(it makes sense, but see later for why exactly)

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## Compare: equidistance constraints vs. springs



- Similar
  - they both mean:  
these 2 particles “want to be” at *this* distance (not more, not less)
- but different
  - equidistance constraint:
    - applied during **constraint enforcement**
    - directly affects positions
    - models a **rigid** rod between the two particles
      - of a given length
    - sometimes called a “HARD” constraint
  - spring:
    - applied during **force evaluation** step
    - affects forces, therefore accelerations
    - models a **deformable** spring between the two particles
      - of a given length
    - sometimes called a “SOFT” constraint
- A physic engine can use both at the same time!

some constant scalar parameter  $D$



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## How to enforce a positional constraint? (see next lecture for the complete answer)



When enforcing constraints...

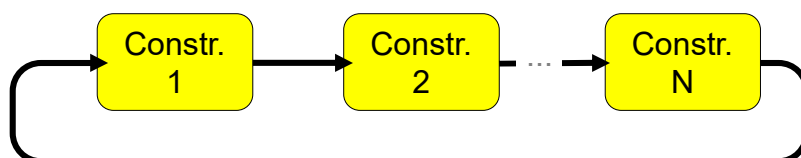
- If a constraint is valid, no problem.
  - If it's not, change particle positions so that it does.
    - Problems: there's usually many possible ways to do
- ⚠ Which one to pick?

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## Enforcing a set of constraints



- There are many constraints to impose: when you solve one → maybe you break another!
- Simultaneous enforcement: computationally expensive
- Practical & easy solution: just enforce them in cascade (similar in concept to Gauss-Seidel solvers):



Repeat until convergence (= max error below threshold)  
...but at most for  $N_{MAX}$  times! (remember: our simulation is *soft* real-time)

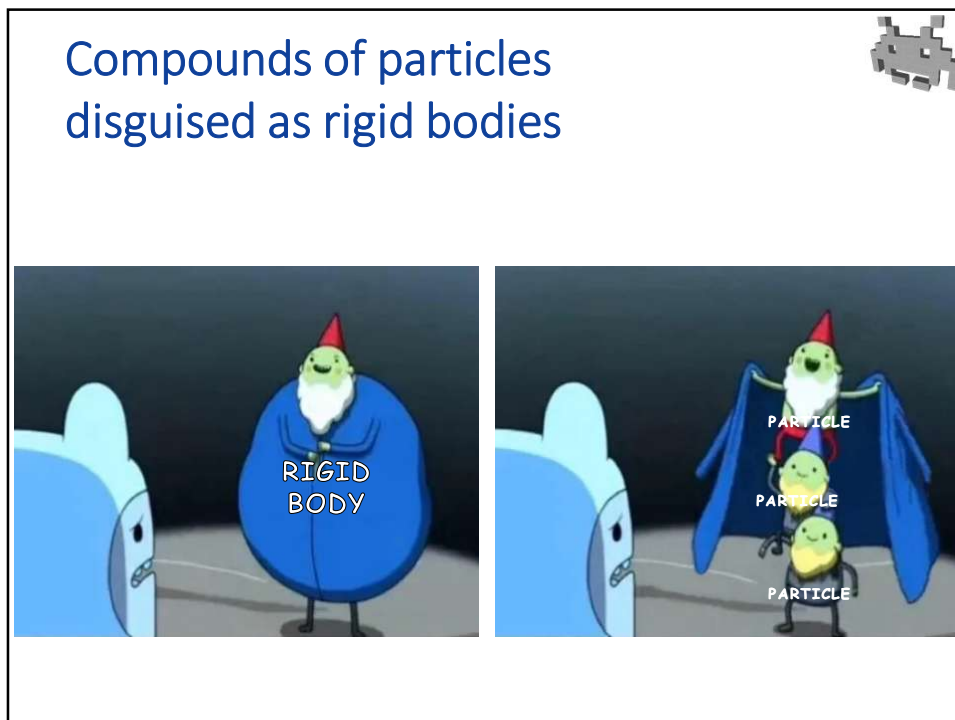
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## Enforcing a set of constraints one after the other (in cascade)



- The whole loop for imposing the constraints happen in the constraint enforcement phase on one physics step
- Notes about convergence:
  - needed iterations (typically) few: e.g. 1 ~ 10 (efficient!).
  - if convergence not reached within a given number of steps: never mind, next frames will fix it (so it's robust)
  - (it is never reached, if constraints are contradictory)
  - Optimization (to reduce the number of needed iterations): solve the most unsatisfied constraints first
- ⚠ Problem: it's a **sequential** approach! ☹
  - **parallelized** versions (similar to Jacobi solvers) are possible
  - they have a worse convergence in practice (they require more iterations), but each iteration is faster

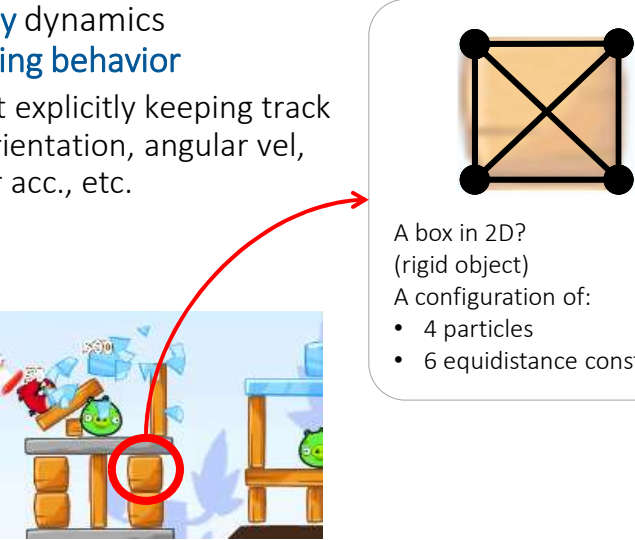
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### We can combine equidistance constraints to obtain rigid objects!

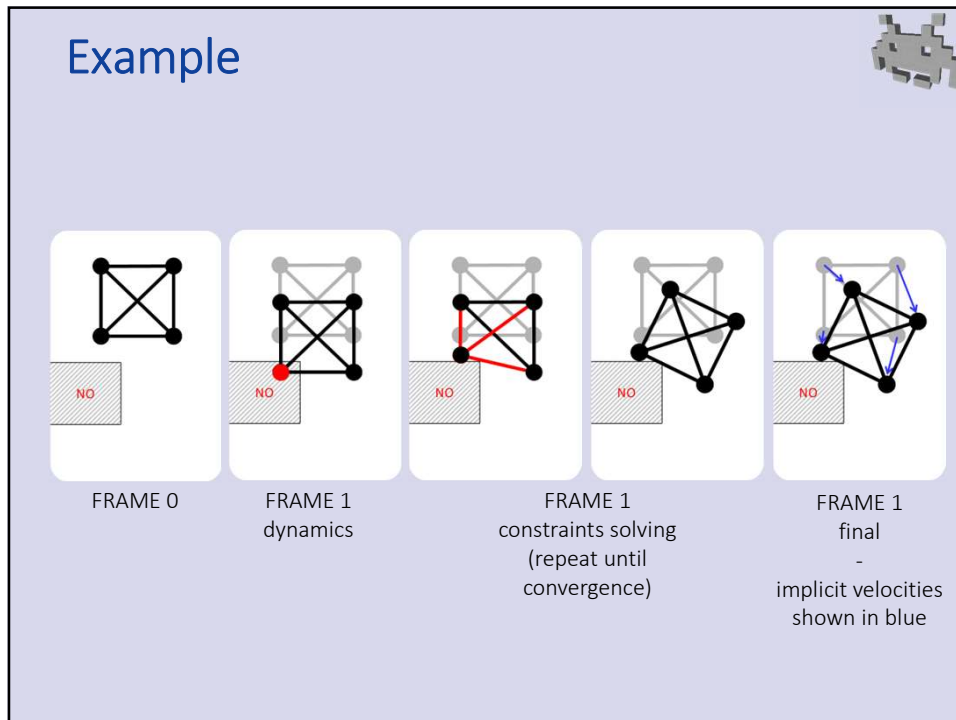
- **Rigid body** dynamics as **emerging behavior**
  - without explicitly keeping track their orientation, angular vel, angular acc., etc.



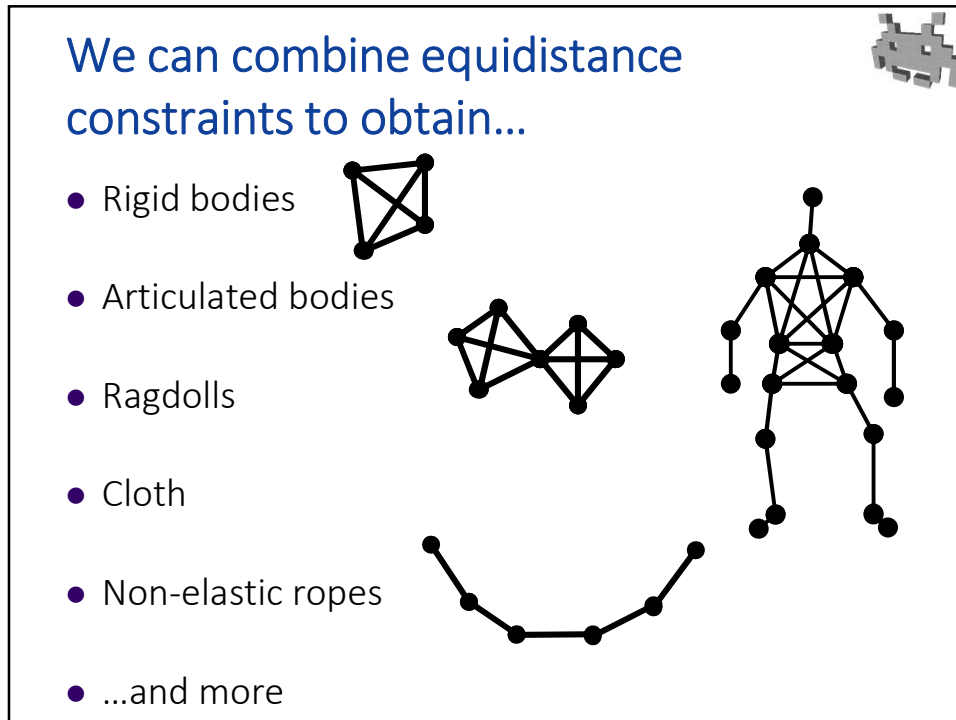
A box in 2D?  
(rigid object)  
A configuration of:

- 4 particles
- 6 equidistance constraints

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## Positional constraints (in general terms, and more formally)



- A predicate defined on the position(s) of a number of particles (usually, a small number: 1 - 4)

$$\mathcal{C}: (\mathbf{p}_a, \mathbf{p}_b, \mathbf{p}_c, \dots) \rightarrow \{ true, false \}$$

- For example, the equidistance constraints is

$$\mathcal{C}(\mathbf{p}_a, \mathbf{p}_b) \Leftrightarrow \|\mathbf{p}_a - \mathbf{p}_b\| = k_{CONST}$$

- They can be an equality (=) or an inequality ( $\leq$  or  $\geq$ )

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## Equality positional constraints: examples



- Equidistance constraint (the one we have seen):  
*«these 2 particles must stay at distance  $k$ »*
  - E.g: because they are linked by a metal rod of length  $k$
- Fixed positions:  
*«this particle must stay in position  $\mathbf{p}_a$  »*
  - the particle is “pinned” in position
  - trivial to impose, but still useful!
- Coplanarity / collinearity:  
*«these  $N$  particles must stay on a line / on a plane»*

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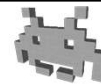
## Equality positional constraints: other examples



- Volume preservation:  
*“The volume delimited by the squishy ballon defined by these particles is a constant  $k_{\text{CONST}}$ ” (e.g., because it’s filled with water)*
- How to impose it (approximation):
  1. Estimate current total volume  $v$
  2. uniform scale the entire object by factor  $\sqrt[3]{f_{\text{CONS}} / v}$  around its barycenter

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## Inequality positional constraints: example



- “please don’t sink below the ground”  
assuming the ground is the plane  $Y = 0$

$$C(\mathbf{p}_a) \Leftrightarrow \mathbf{p}_a \cdot \mathbf{y} \geq 0$$

- Trivial to impose:  
just set the  $\mathbf{y}$  to  $0$ , if it is  $< 0$

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## Inequality positional constraints: example



- “this particle must stay above this fixed (and arbitrary) plane”
  - For example, because the plane is a solid unmovable slab
  - The plane is given by a point on it  $\mathbf{p}_q$  and its normal  $\hat{\mathbf{n}}_q$

$$C(\mathbf{p}_a) \Leftrightarrow (\mathbf{p}_a - \mathbf{p}_q) \cdot \hat{\mathbf{n}}_q \geq 0$$

- To enforce it (when it's not already true) enforce

$$(\mathbf{p}_a - \mathbf{p}_q) \cdot \hat{\mathbf{n}}_q = 0$$

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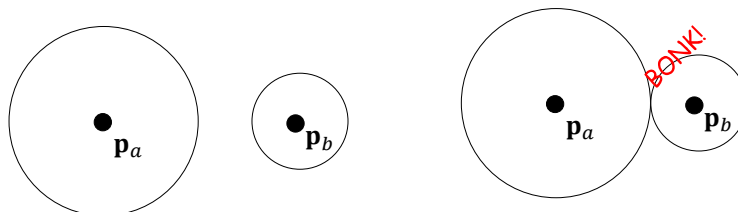
## Inequality positional constraints: example



- These two particles must be *at least*  $k_{CONST}$  apart

$$C(\mathbf{p}_a, \mathbf{p}_b) \Leftrightarrow \|\mathbf{p}_a - \mathbf{p}_b\| \geq k_{CONST}$$

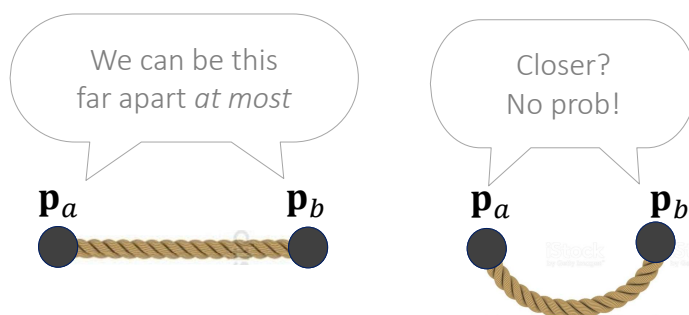
- For example, because they are the centers of two rigid spheres and  $k_{CONST}$  is the sum of their radii
- part of “collision handling” (a topic for later)



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## Inequality positional constraints: example

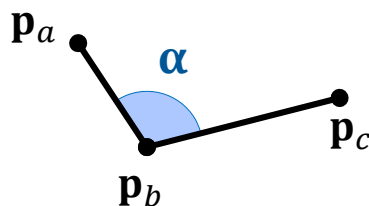
- These two particles must be *at most*  $k_{CONST}$  apart  
$$\mathcal{C}(\mathbf{p}_a, \mathbf{p}_b) \Leftrightarrow \|\mathbf{p}_a - \mathbf{p}_b\| \leq k_{CONST}$$
  - For example, because they are tied by an inextensible rope that has length  $k_{CONST}$  (but can fold)



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## Inequality positional constraints: example

- Angle constraints, e.g.  $\alpha < \alpha_{\max}$   
with  $\alpha$  the angle between  $\mathbf{p}_a, \mathbf{p}_b$  and  $\mathbf{p}_b, \mathbf{p}_c$ 
  - e.g., on joints: «*elbows cannot bend backward*»
  - (a constraint between three particles!)



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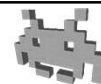
## Enforcing one positional constraint (in general terms)



- **Inequality** constraint:
  1. *Test*: does the inequality already hold?
  2. If so: do nothing
  3. If not: enforce it as an **equality** (=) instead (see below)
- **Equality** constraint:
  - All involved particles must be displaced from that current position, so that it now holds
  - There can be infinite ways to achieve this!  
Question: **Which one to pick?**

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## Enforcing one equality constraint: (assuming for now all particles have same mass)



- Answer:  
**minimize** the sum of *squared* displacements  
(with respect to current position)
- For each kind of constraint, we need to find the minimizer analytically  
("analytically" = in closed form = "with formulas"  
= "solving a simple math problem on paper")
  - That's what we did for the equality constraint

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### Enforcing one equality constraint (assuming for now all particles have same mass)

- We want to enforce a constraint  $\mathcal{C}$  on particles  $a, b, c, \dots$  currently in positions  $\mathbf{p}_a, \mathbf{p}_b, \mathbf{p}_c, \dots$ 

$$\mathcal{C}: (\mathbf{p}_a, \mathbf{p}_b, \mathbf{p}_c, \dots) \rightarrow \{ true, false \}$$
- We must apply the displacements  $\vec{d}_a, \vec{d}_b, \vec{d}_c$  that are the
 
$$\underset{\vec{d}_a, \vec{d}_b, \vec{d}_c, \dots}{\operatorname{argmin}} \left( \|\vec{d}_a\|^2 + \|\vec{d}_b\|^2 + \|\vec{d}_c\|^2 + \dots \right)$$
 such that  $\mathcal{C}(\mathbf{p}_a + \vec{d}_a, \mathbf{p}_b + \vec{d}_b, \mathbf{p}_c + \vec{d}_c, \dots)$ 

among all the choices that satisfy this,  
 we want the one which minimizes this

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### Enforcing one equality constraint (in general, for particles with different mass)

- We want to enforce a constraint  $\mathcal{C}$  on particles  $a, b, c, \dots$  in positions  $\mathbf{p}_a, \mathbf{p}_b, \mathbf{p}_c, \dots$  and with masses  $m_a, m_b, m_c, \dots$ 

$$\mathcal{C}: (\mathbf{p}_a, \mathbf{p}_b, \mathbf{p}_c, \dots) \rightarrow \{ true, false \}$$
- We must apply the displacements  $\vec{d}_a, \vec{d}_b, \vec{d}_c$  which are the:
 
$$\underset{\vec{d}_a, \vec{d}_b, \vec{d}_c, \dots}{\operatorname{argmin}} \left( m_a \|\vec{d}_a\|^2 + m_b \|\vec{d}_b\|^2 + m_c \|\vec{d}_c\|^2 + \dots \right)$$
 such that  $\mathcal{C}(\mathbf{p}_a + \vec{d}_a, \mathbf{p}_b + \vec{d}_b, \mathbf{p}_c + \vec{d}_c, \dots)$ 

among all the choices that satisfy this,  
 we want the one which minimizes this

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