

CSC418/2504 Computer Graphics

A digital artwork depicting a scene from Star Wars. In the foreground, a Stormtrooper stands on a beach, holding a blaster. In the background, a droid is visible on the left, and a TIE fighter flies in the sky. The scene is set during a sunset or sunrise, with a calm sea and a cloudy sky.

Rob Kaz

Some Slides/Images adapted from Marschner and Shirley

Physics-Based Animation



Announcements

Assignment 4 graded (Avg 85.6% -- Very Good!)

If you want to see your midterm come to office hours

Handle midterm remark requests through MarkUs

Assignment 8 out soon, due March 20th

What are we doing about COVID-19

1. All assignments and lecture notes will, by tomorrow, be online. Continue to use github issues/email for online help
2. This is the last lecture on testable material
3. No online office hours for now (if there is strong demand I will figure out how to set this up).
4. If you miss the final test for a valid reason (with documentation) your grade will be redistributed 50-50 to assignments and the previous test

Physics-Based Animation

Newton's Laws of Motion

The Mass-Spring System

Implicit Integration via Optimization

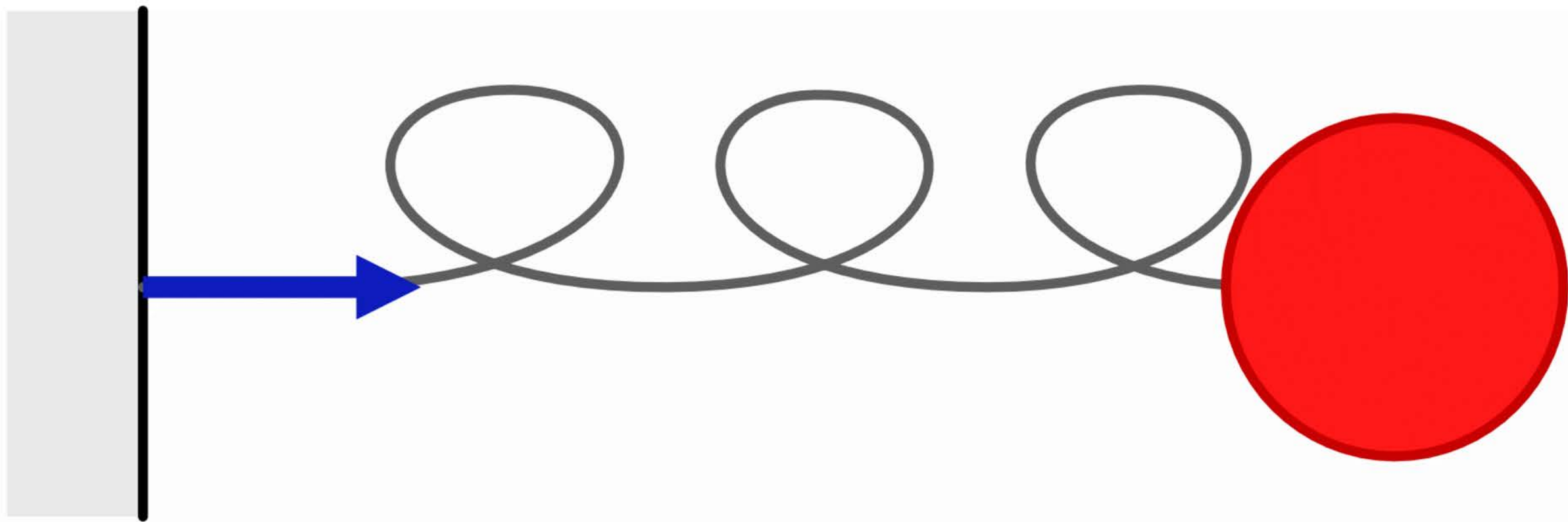
A Local-Global Solver for Fast-Mass Springs

Newton's Laws

1. Every object will remain at rest or in uniform motion in a straight line unless compelled to change its state by the action of an external force
2. The force acting on an object is equal to the time rate-of-change of the momentum
3. For every action there is an equal and opposite reaction

Newton's Second Law

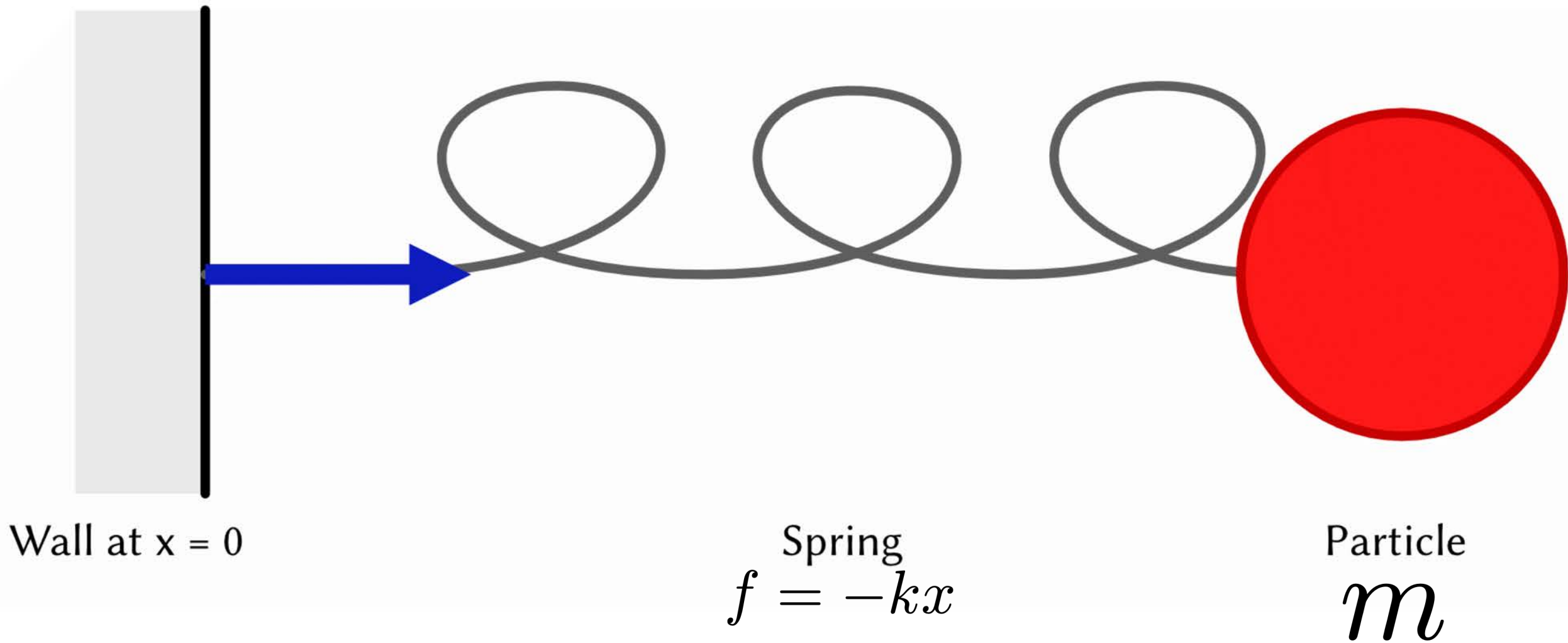
$$\underset{\text{Mass}}{m} \overset{\text{Acceleration}}{a} = \underset{\text{force}}{f}$$



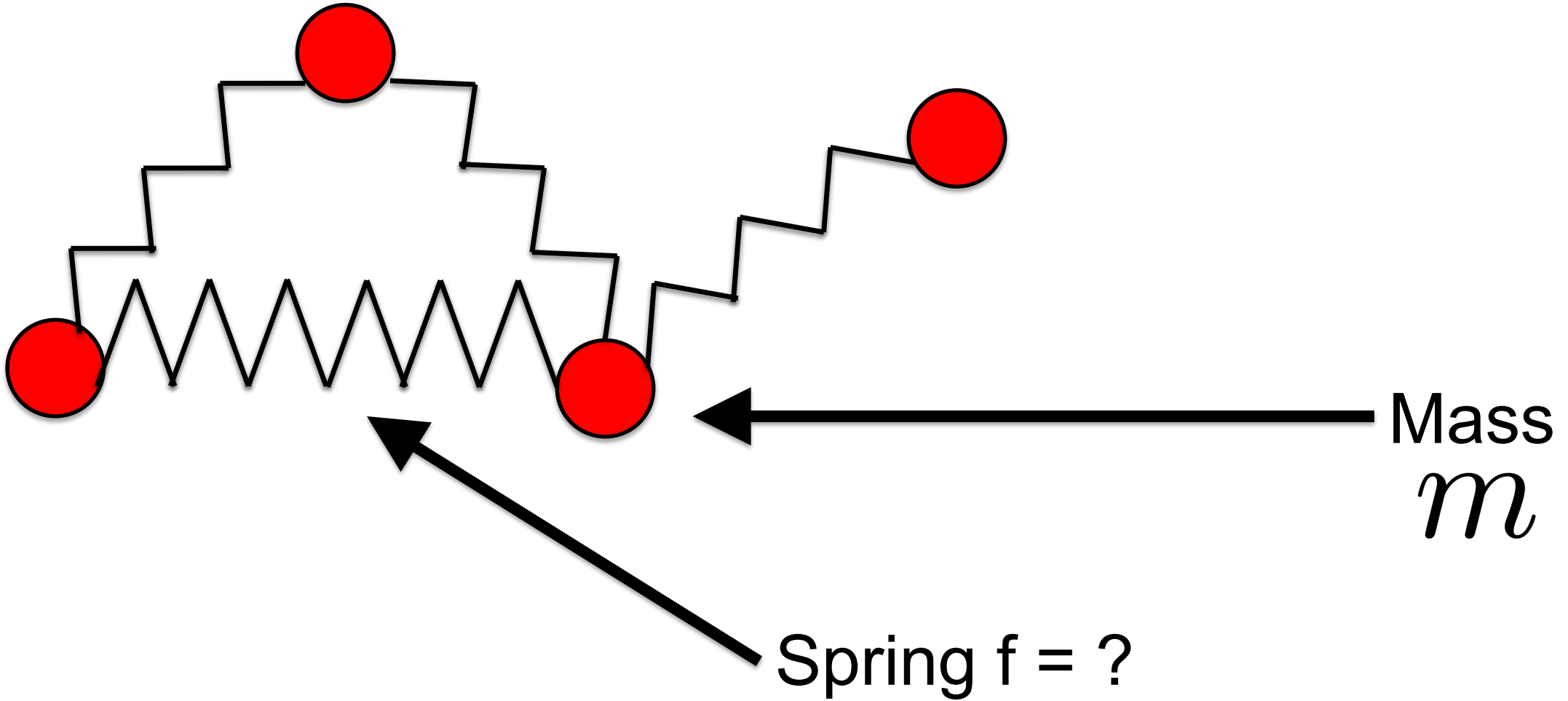
Wall at $x = 0$

Spring

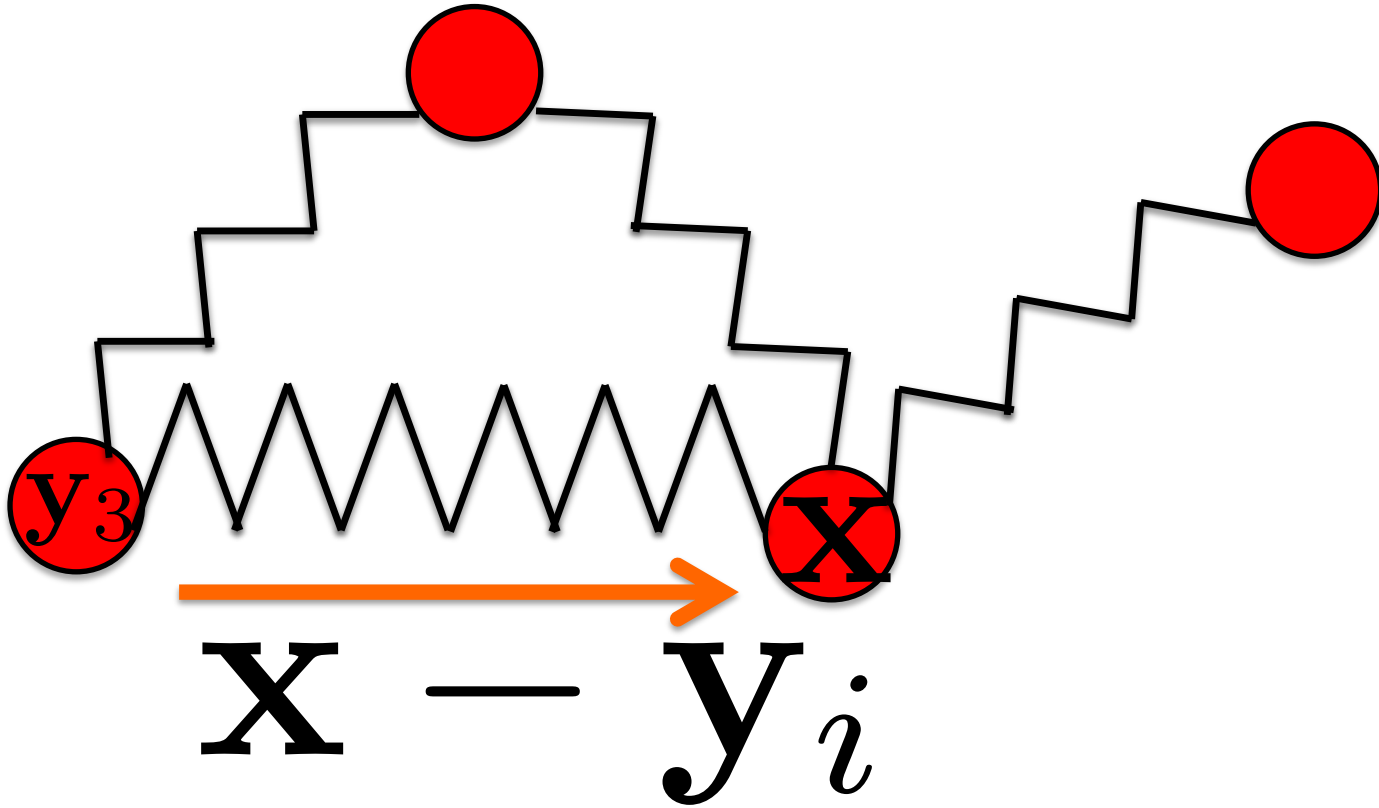
Particle



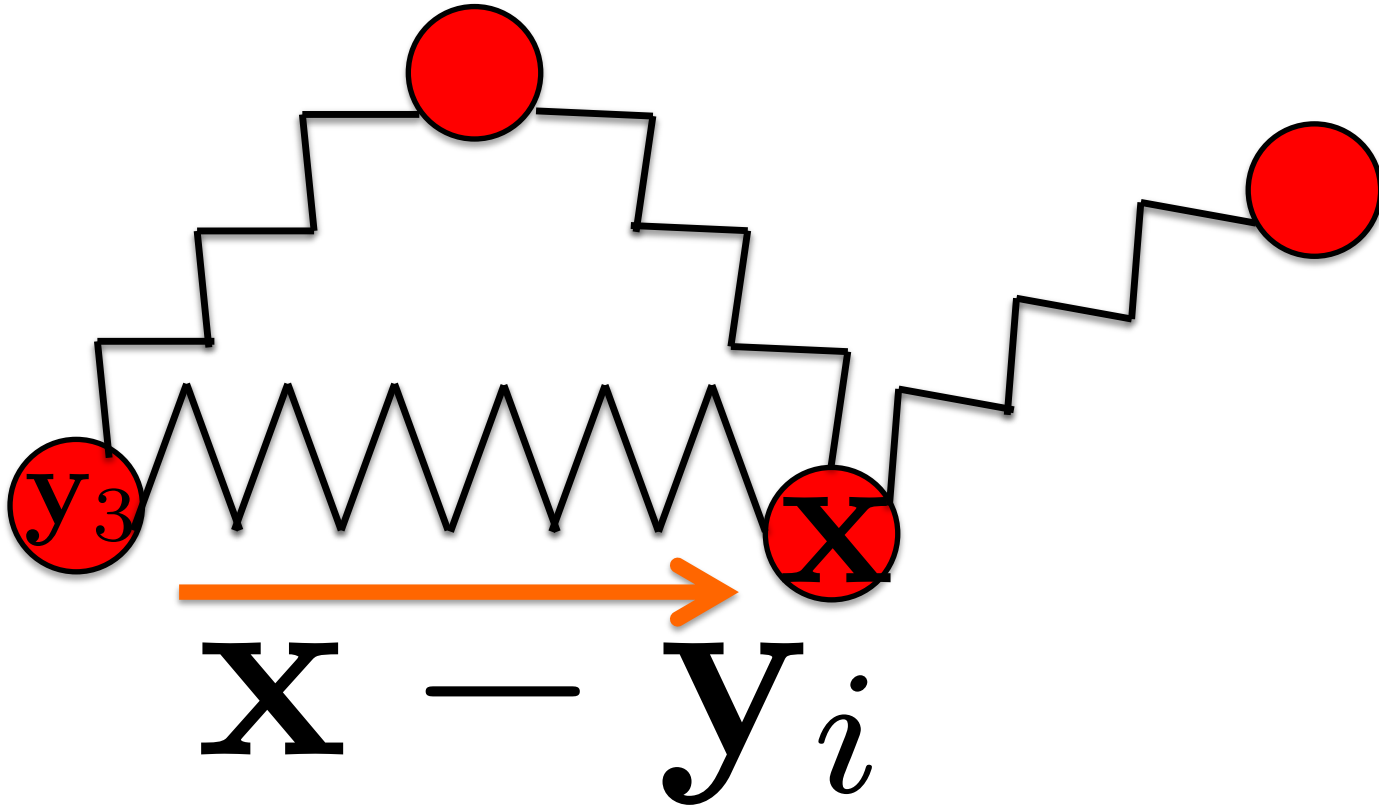
The Mass-Spring System



The Mass-Spring System

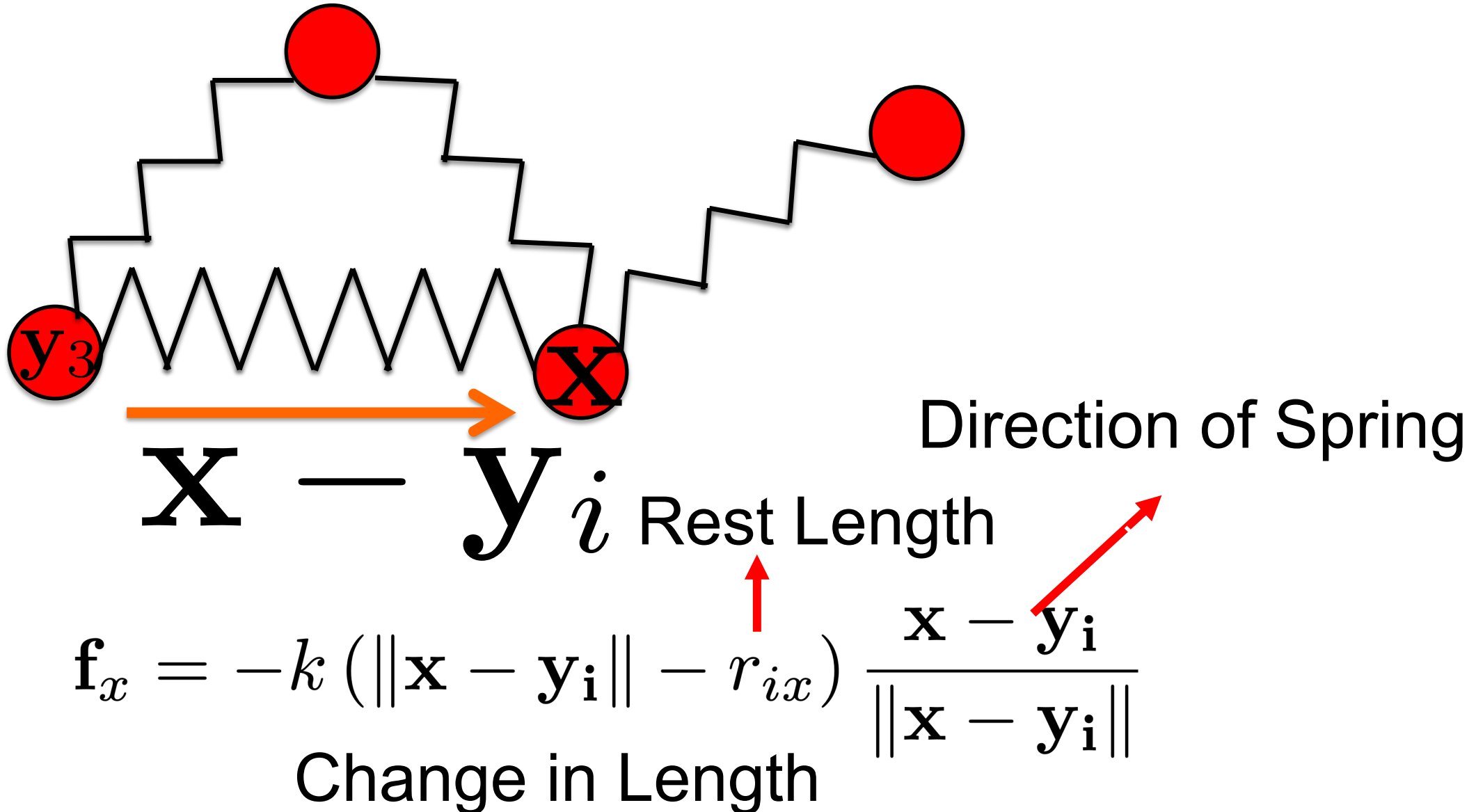


The Mass-Spring System

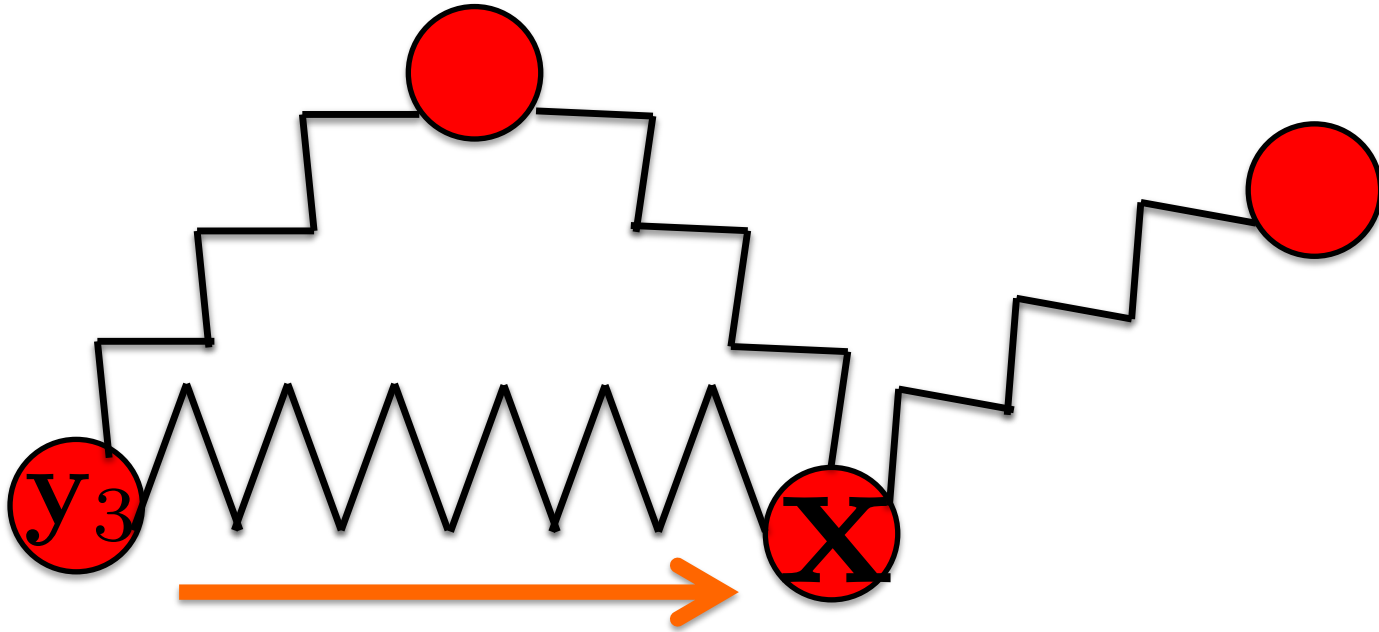


$$\mathbf{f}_x = -k (\|\mathbf{x} - \mathbf{y}_i\| - r_{ix}) \frac{\mathbf{x} - \mathbf{y}_i}{\|\mathbf{x} - \mathbf{y}_i\|}$$

The Mass-Spring System

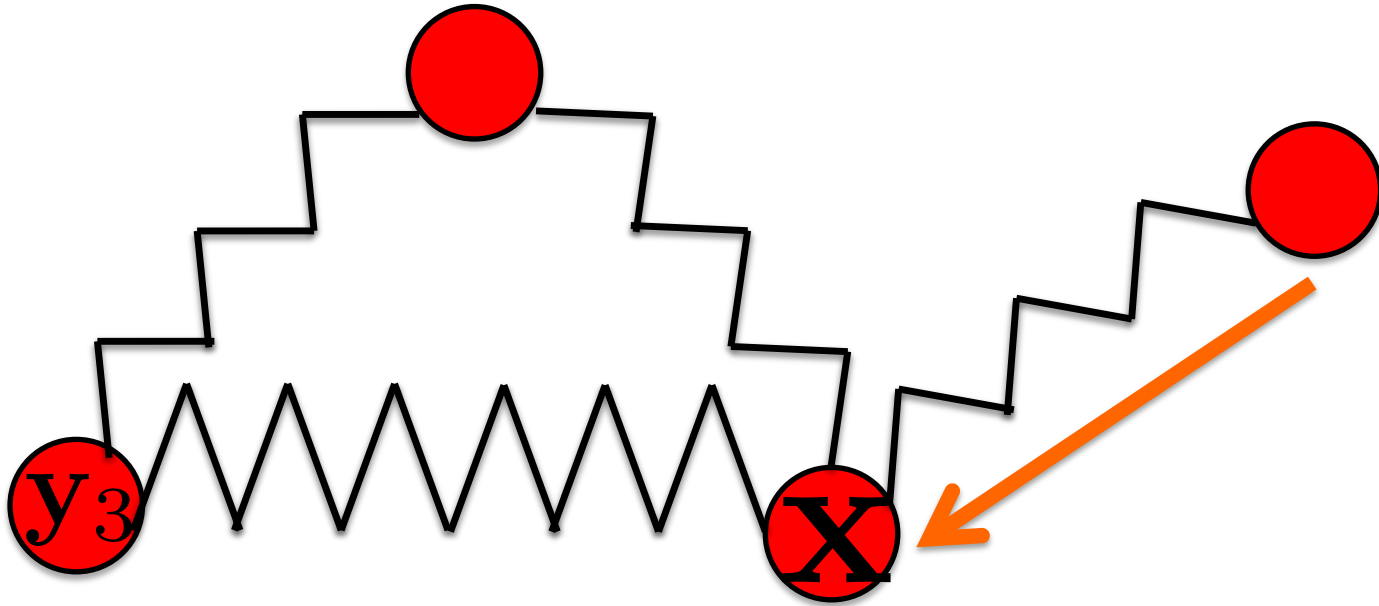


Newton's Second Law for Each Particle



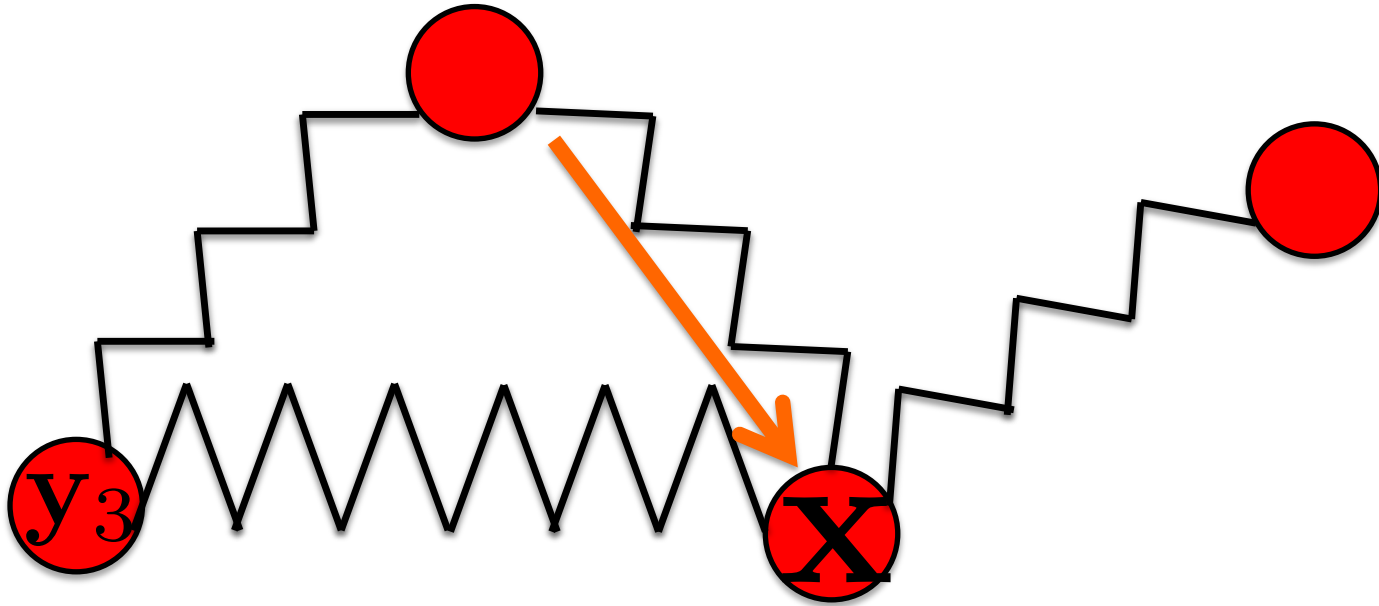
$$m_x \mathbf{a}_x = \sum_i \mathbf{f}_x (\mathbf{y}_i)$$

Newton's Second Law for Each Particle



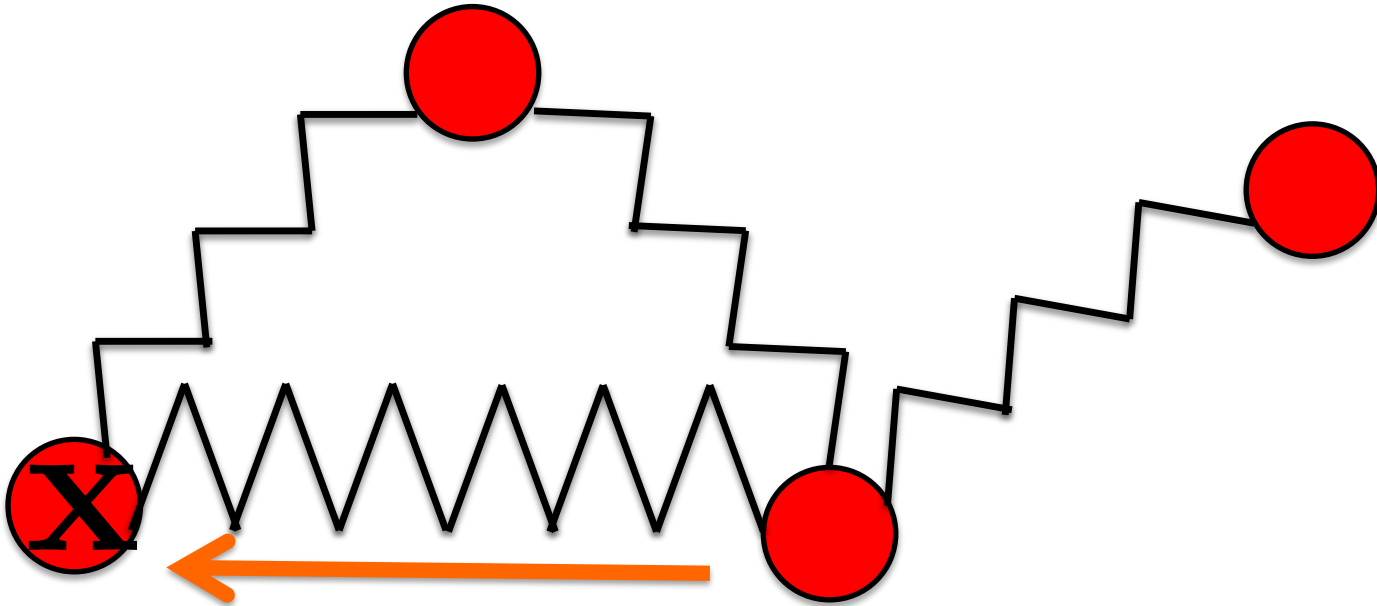
$$m_x \mathbf{a}_x = \sum_i \mathbf{f}_x (\mathbf{y}_i)$$

Newton's Second Law for Each Particle



$$m_x \mathbf{a}_x = \sum_i \mathbf{f}_x (\mathbf{y}_i)$$

Newton's Second Law for Each Particle



$$m_x \mathbf{a}_x = \sum_i \mathbf{f}_x(\mathbf{y}_i)$$

One Equation for each particle
We will solve them all together





Cloth

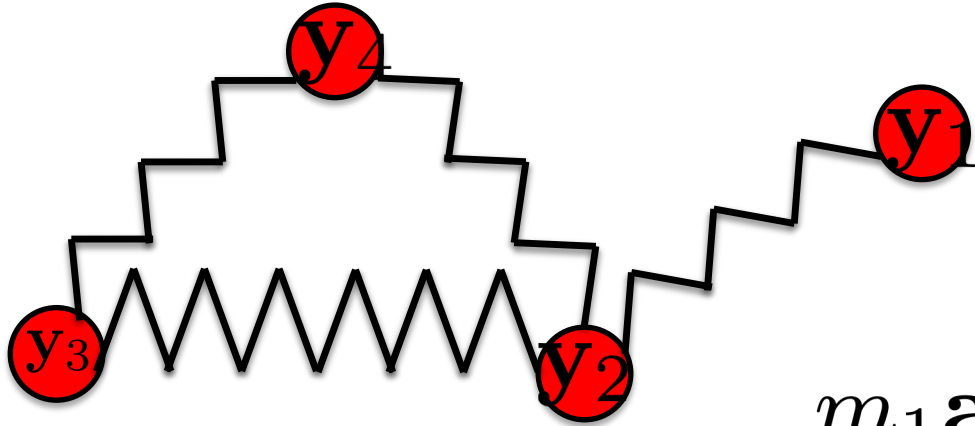
SIMIT GPU

15,630 Triangles

7,988 Verts

14 FPS

Newton's Second Law: System of Equations



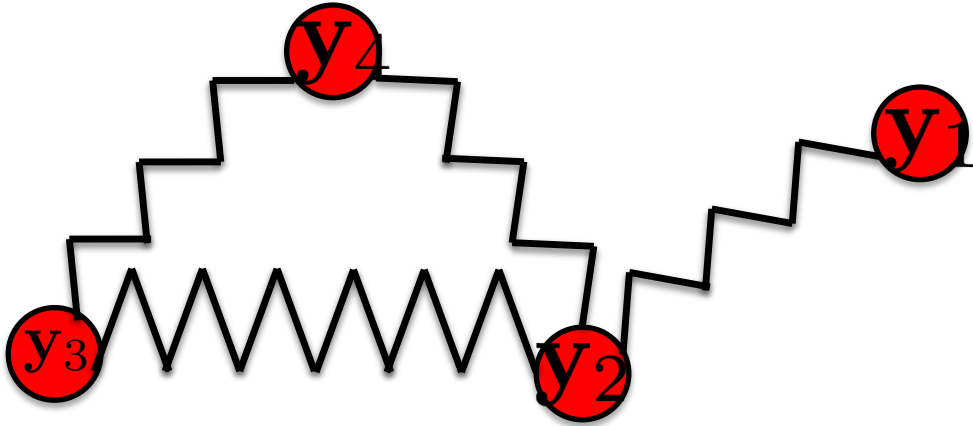
$$m_1 \mathbf{a}_1 = \sum_i \mathbf{f}_1 (\mathbf{y}_i)$$

$$m_2 \mathbf{a}_2 = \sum_i \mathbf{f}_2 (\mathbf{y}_i)$$

$$m_3 \mathbf{a}_3 = \sum_i \mathbf{f}_3 (\mathbf{y}_i)$$

$$m_4 \mathbf{a}_4 = \sum_i \mathbf{f}_4 (\mathbf{y}_i)$$

Newton's Second Law: System of Equations



$$\begin{pmatrix} m_1 \cdot I & 0 & 0 & 0 \\ 0 & m_2 \cdot I & 0 & 0 \\ 0 & 0 & m_3 \cdot I & 0 \\ 0 & 0 & 0 & m_4 \cdot I \end{pmatrix} \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \mathbf{a}_3 \\ \mathbf{a}_4 \end{pmatrix} = \begin{pmatrix} \mathbf{f}_1 \\ \mathbf{f}_2 \\ \mathbf{f}_3 \\ \mathbf{f}_4 \end{pmatrix}$$

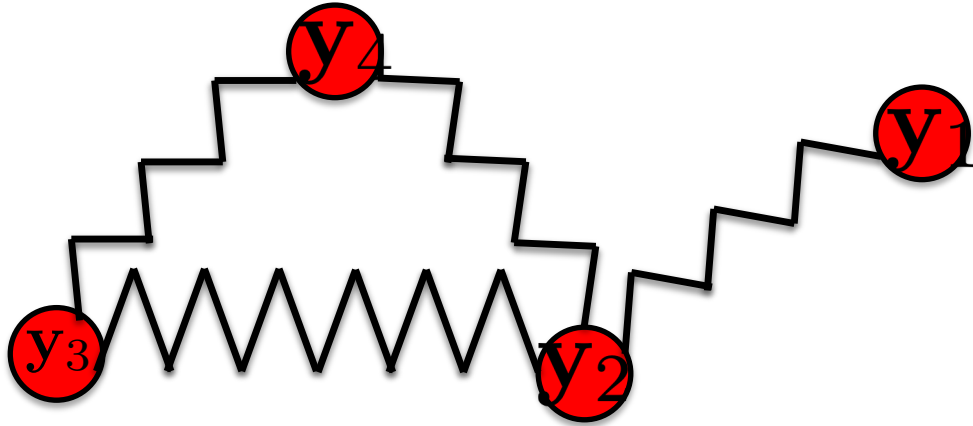
Mass Matrix $\mathbf{a}(t)$ $\mathbf{f}(t)$

Time Integration



Time Integration Converts Accelerations to Positions

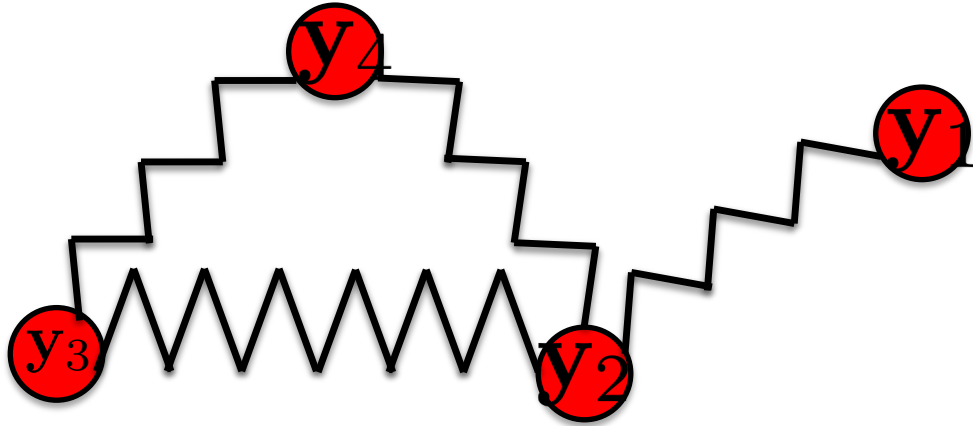
Time Integration



$$\begin{pmatrix} m_1 \cdot I & 0 & 0 & 0 \\ 0 & m_2 \cdot I & 0 & 0 \\ 0 & 0 & m_3 \cdot I & 0 \\ 0 & 0 & 0 & m_4 \cdot I \end{pmatrix} \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \mathbf{a}_3 \\ \mathbf{a}_4 \end{pmatrix} = \begin{pmatrix} \mathbf{f}_1 \\ \mathbf{f}_2 \\ \mathbf{f}_3 \\ \mathbf{f}_4 \end{pmatrix}$$

Mass Matrix
 $\mathbf{a}(t)$
 $\mathbf{f}(t)$

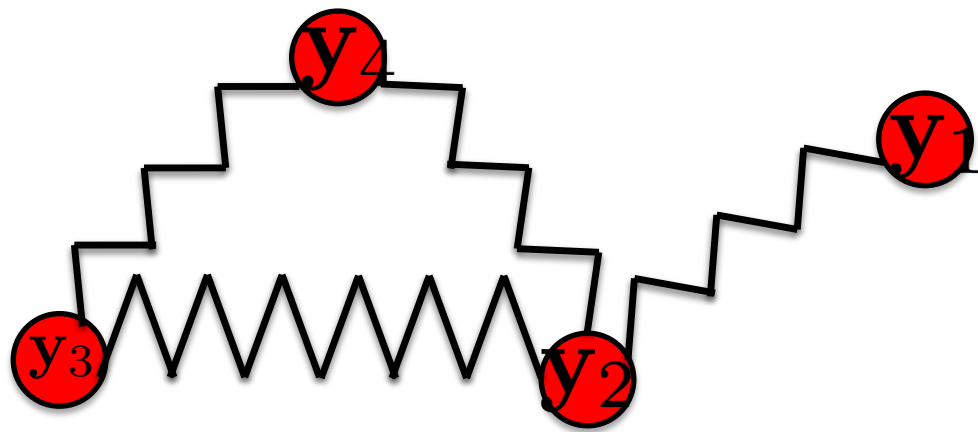
Time Integration



$$M \mathbf{a}(t) = \mathbf{f}(\mathbf{y}(t))$$

Mass Matrix

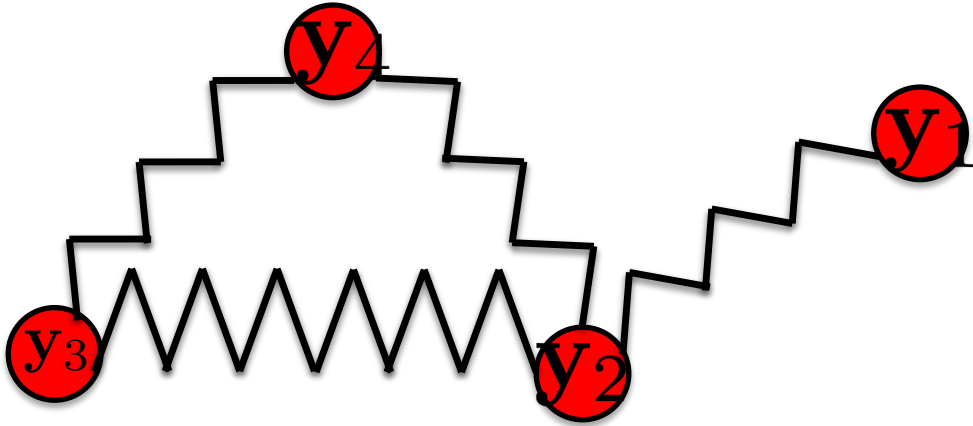
Time Integration



$$M \frac{d^2 \mathbf{y}(t)}{dt^2} = \mathbf{f}(\mathbf{y}(t))$$

Use Finite Differences: $\frac{d^2 \mathbf{y}(t)}{dt^2} \approx \frac{1}{\Delta t^2} (\mathbf{y}^{t+1} - 2\mathbf{y}^t + \mathbf{y}^{t-1})$

Time Integration

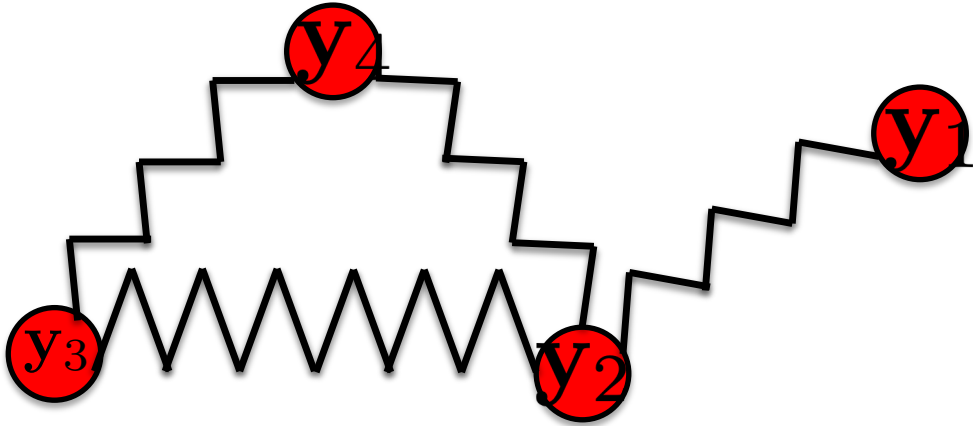


Need to Discretize

$$M \frac{d^2 \mathbf{y}(t)}{dt^2} = \mathbf{f}(\mathbf{y}(t))$$

Use Finite Differences: $\frac{d^2 \mathbf{y}(t)}{dt^2} \approx \frac{1}{\Delta t^2} (\mathbf{y}^{t+1} - 2\mathbf{y}^t + \mathbf{y}^{t-1})$

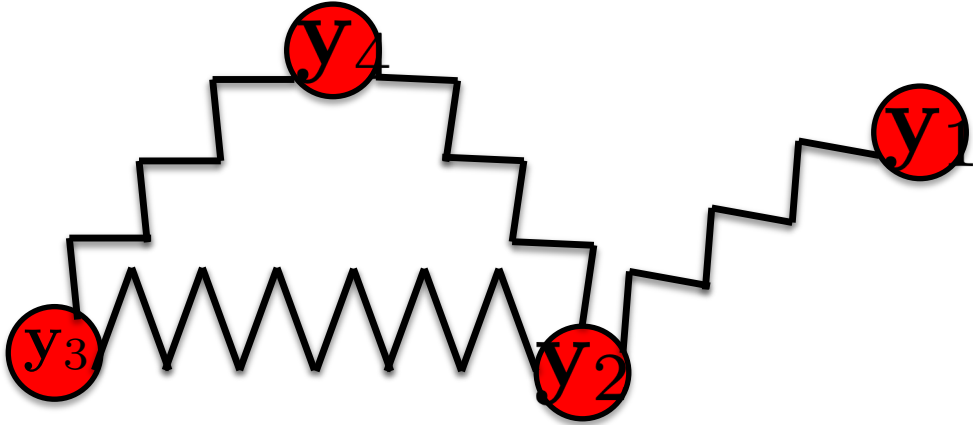
Implicit Time Integration



$$M \frac{d^2 \mathbf{y}}{dt^2} (t) = \mathbf{f} (\mathbf{y}^{t+1})$$

Use Finite Differences: $\frac{d^2 \mathbf{y} (t)}{dt^2} \approx \frac{1}{\Delta t^2} (\mathbf{y}^{t+1} - 2\mathbf{y}^t + \mathbf{y}^{t-1})$

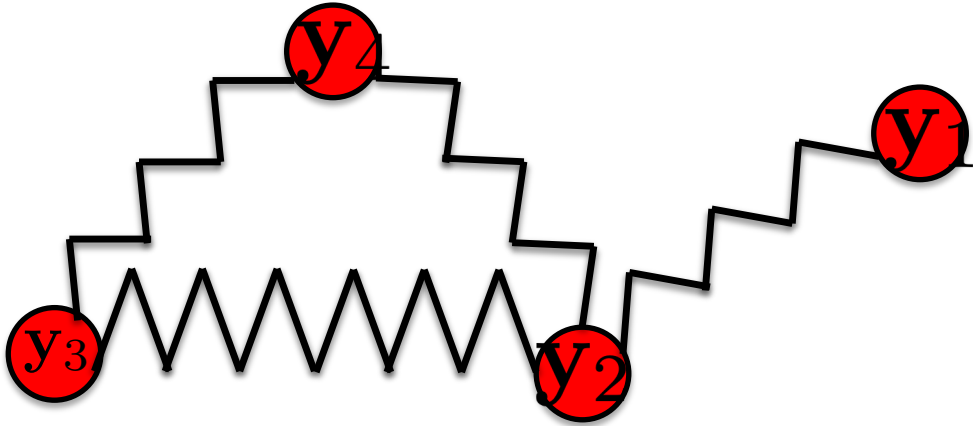
Implicit Time Integration



$$M \mathbf{y}^{t+1} = M (2\mathbf{y}^t - \mathbf{y}^{t-1}) + \Delta t^2 \mathbf{f}(\mathbf{y}^{t+1})$$

Goal: Solve for \mathbf{y}^{t+1}

Implicit Time Integration



$$M\mathbf{y}^{t+1} - M(2\mathbf{y}^t - \mathbf{y}^{t-1}) - \Delta t^2 \mathbf{f}(\mathbf{y}^{t+1}) = \mathbf{0}$$

How to find when some equation = 0?

Goal: Solve for \mathbf{y}^{t+1}

Implicit Integration as Optimization

If we can find a function $E(q)$ such that:

$$\nabla_{\mathbf{q}} E(\mathbf{y}^{t+1}) = 0$$

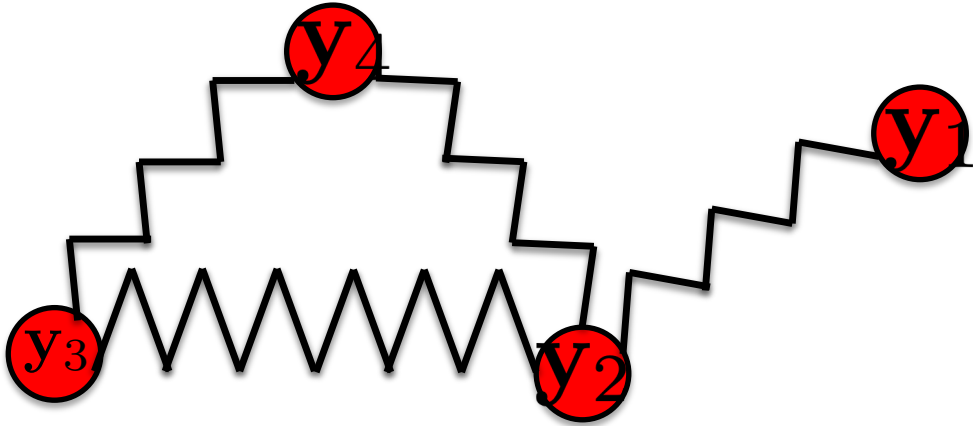
then, rather than solve

$$M\mathbf{y}^{t+1} - M(2\mathbf{y}^t - \mathbf{y}^{t-1}) - \Delta t^2 \mathbf{f}(\mathbf{y}^{t+1}) = \mathbf{0}$$

we can solve

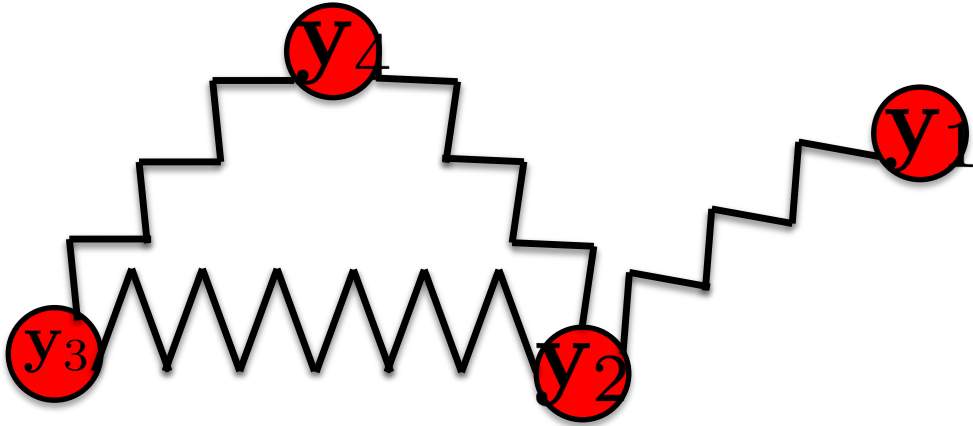
$$\mathbf{y}^{t+1} = \arg \min_{\mathbf{q}} E(\mathbf{q})$$

Implicit Integration as Optimization



$$\underbrace{M\mathbf{y}^{t+1} - M(2\mathbf{y}^t - \mathbf{y}^{t-1})}_{\text{find } \mathbf{E}_1(\mathbf{y}^{t+1})} - \underbrace{\Delta t^2 \mathbf{f}(\mathbf{y}^{t+1})}_{\text{find } \mathbf{E}_2(\mathbf{y}^{t+1})} = \mathbf{0}$$

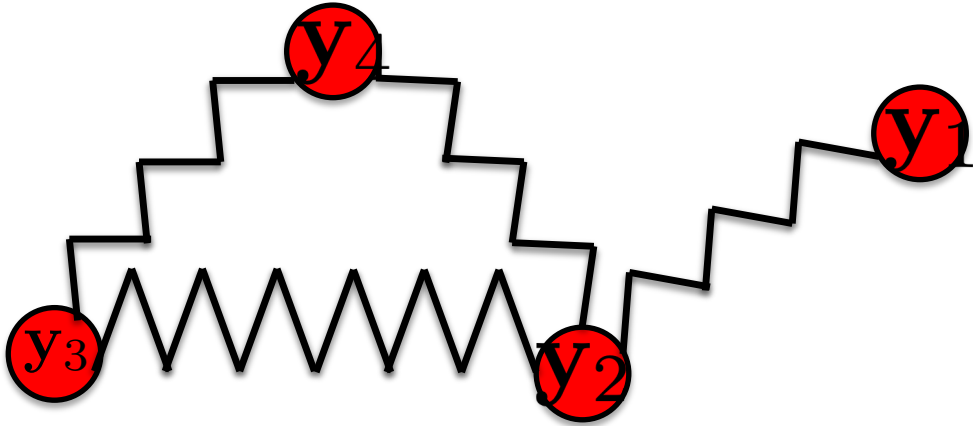
Implicit Integration as Optimization



$$\mathbf{E}_1(\mathbf{y}^{t+1}) = \frac{1}{2} (\mathbf{y}^{t+1})^T M \mathbf{y}^{t+1} - (\mathbf{y}^{t+1})^T M \mathbf{b}$$

$$\mathbf{b} = 2\mathbf{y}^t - \mathbf{y}^{t-1}$$

Implicit Integration as Optimization



$$\underbrace{M\mathbf{y}^{t+1} - M(2\mathbf{y}^t - \mathbf{y}^{t-1})}_{\text{find } \mathbf{E}_1(\mathbf{y}^{t+1})} - \underbrace{\Delta t^2 \mathbf{f}(\mathbf{y}^{t+1})}_{\text{find } \mathbf{E}_2(\mathbf{y}^{t+1})} = \mathbf{0}$$

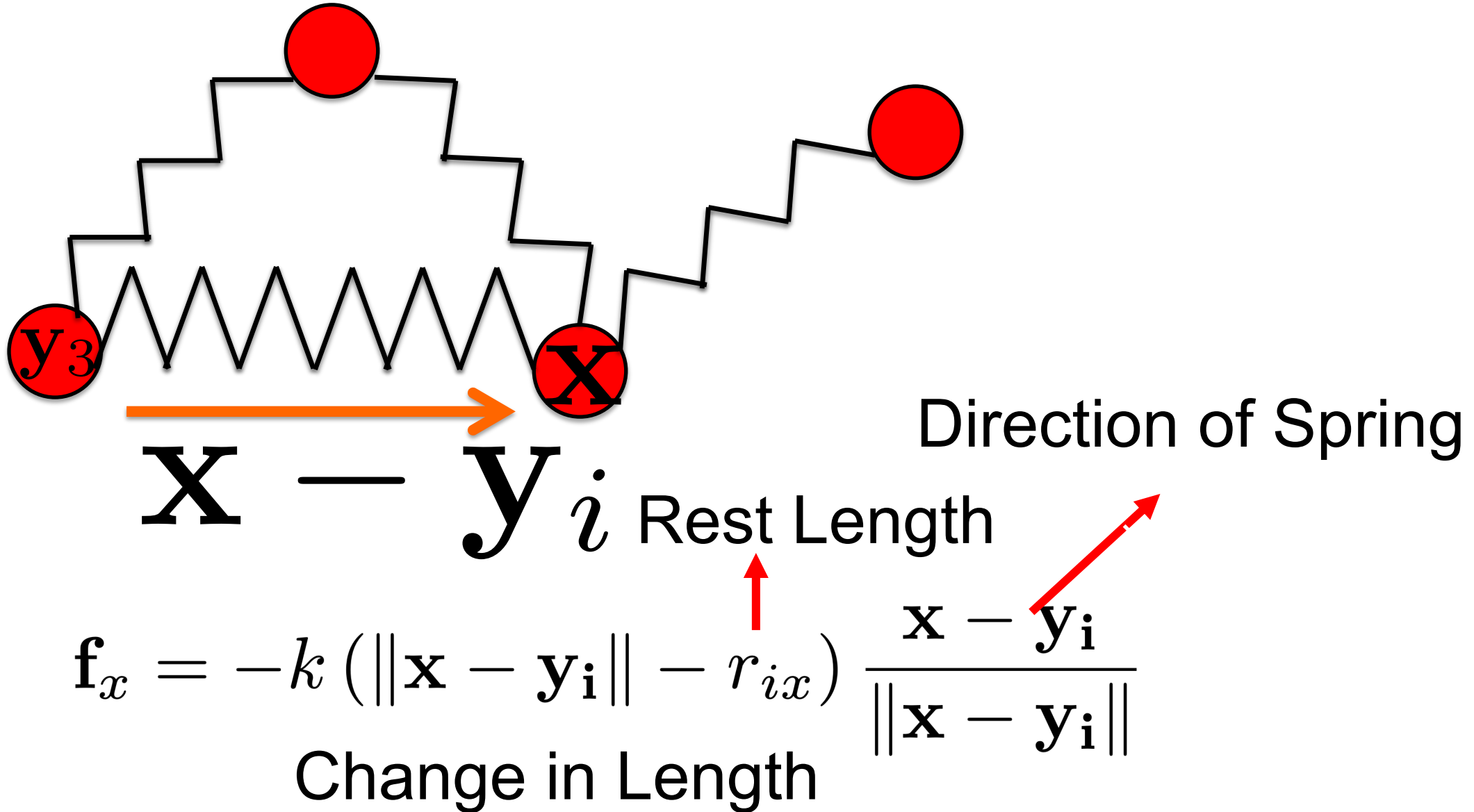
Potential energy

We are going to introduce a special type of energy called potential energy

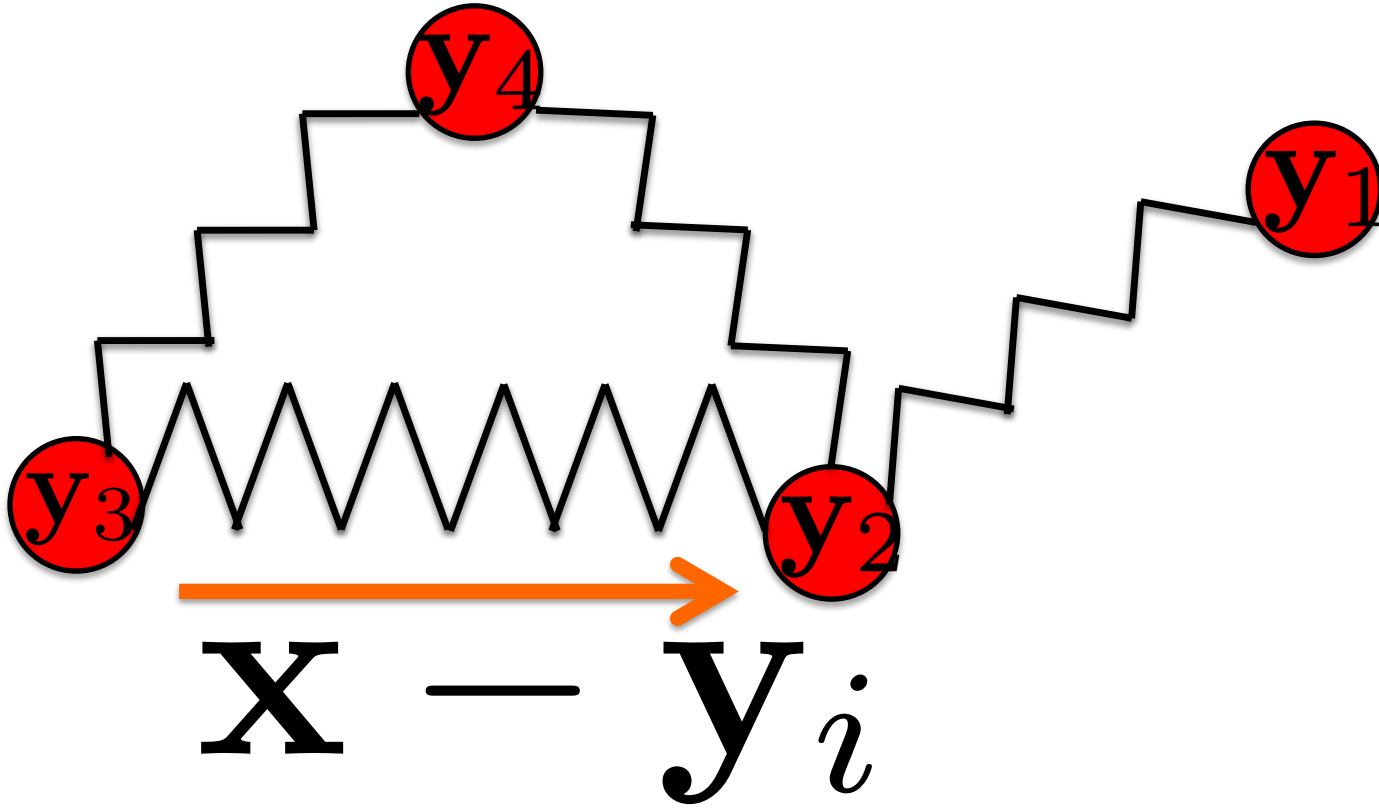
If $E_2(\mathbf{q})$ is a potential energy then

$$\nabla_{\mathbf{q}} E_2 = -\mathbf{f}(\mathbf{q})$$

Potential Energy of a Spring

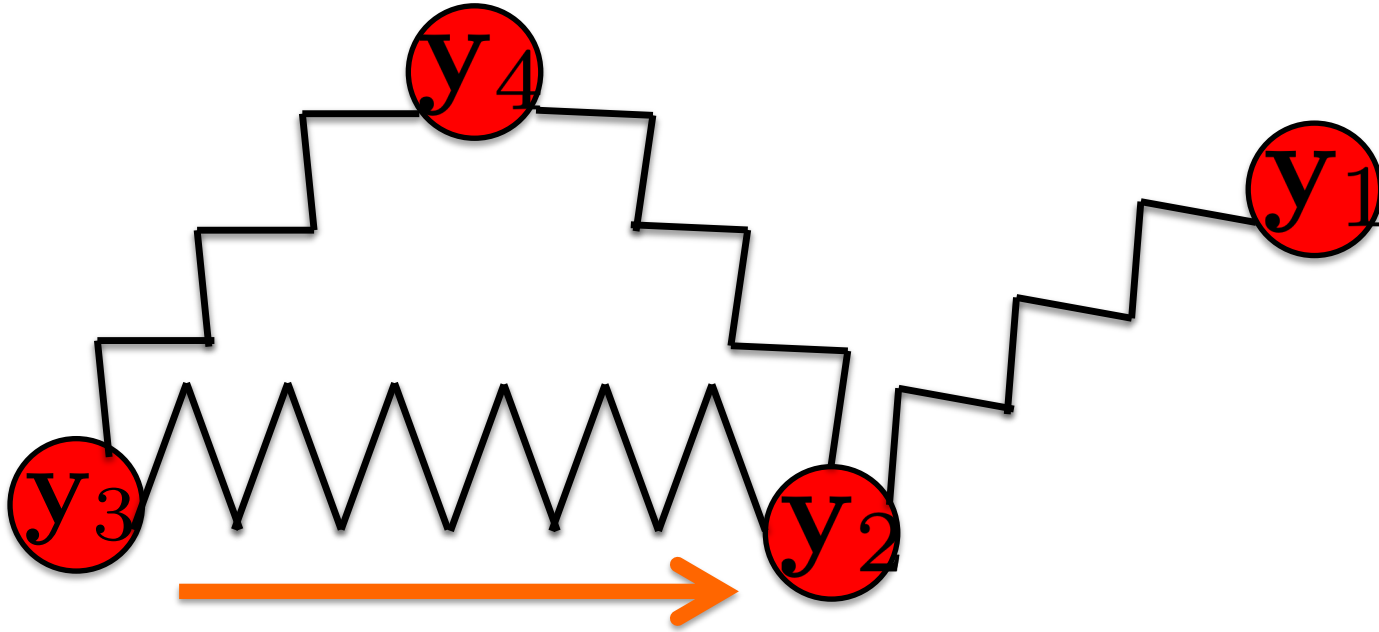


Potential Energy of a Spring



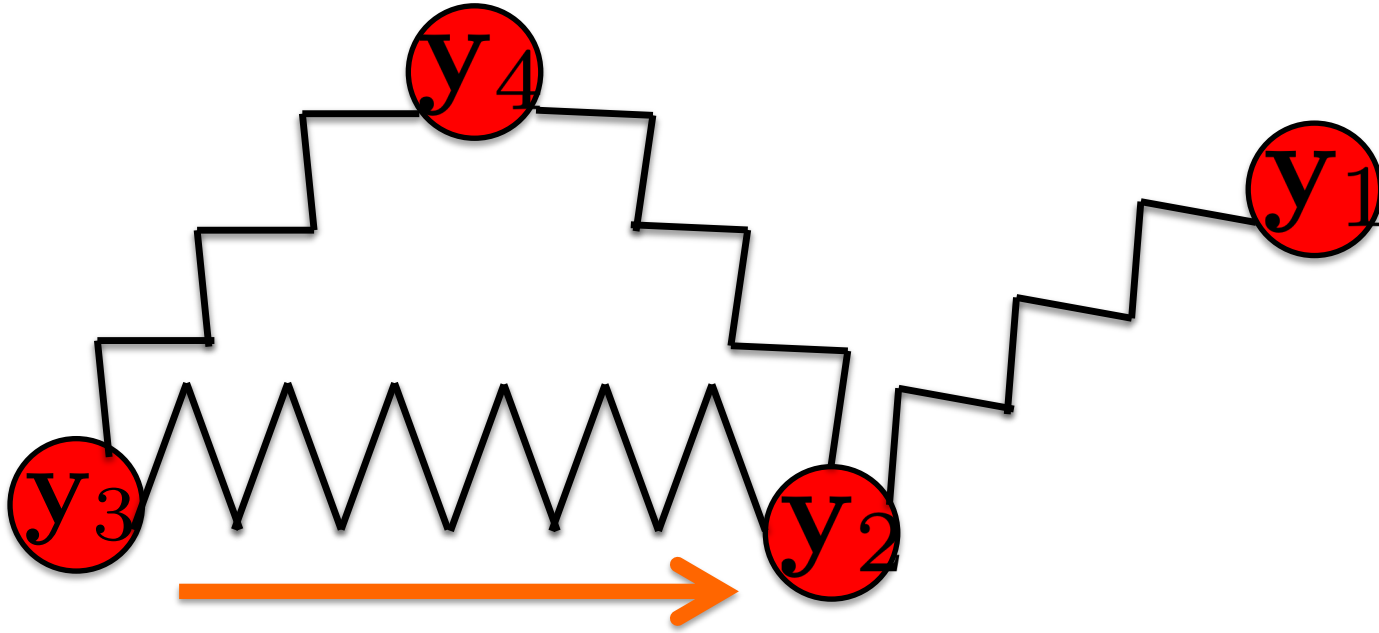
$$\mathbf{f}_{\mathbf{y}_j} = -k (\|\mathbf{y}_j - \mathbf{y}_i\| - r_{ij}) \frac{\mathbf{y}_j - \mathbf{y}_i}{\|\mathbf{y}_j - \mathbf{y}_i\|}$$

Potential Energy of a Spring



$$E_{ij} = \frac{k}{2} (\|\mathbf{y}_j - \mathbf{y}_i\| - r_{ij})^2$$

Potential Energy for a Mass-Spring System



$$E_2 = \sum_{ij} E_{ij} = \sum_{ij} \frac{k}{2} (\|\mathbf{y}_i - \mathbf{y}_j\| - r_{ij})^2$$

Implicit Integration as Optimization

If we can find a function $E(q)$ such that:

$$\nabla_{\mathbf{q}} E(\mathbf{y}^{t+1}) = 0$$

then, rather than solve

$$M\mathbf{y}^{t+1} - M(2\mathbf{y}^t - \mathbf{y}^{t-1}) - \Delta t^2 \mathbf{f}(\mathbf{y}^{t+1}) = \mathbf{0}$$

we can solve

$$\mathbf{y}^{t+1} = \arg \min_{\mathbf{q}} E_1(\mathbf{q}) + \Delta t E_2(q)$$

Local-Global Solvers for Mass-Spring Systems

WHILE Not done

For Each Spring

 Local Optimization

 Global Optimization

END

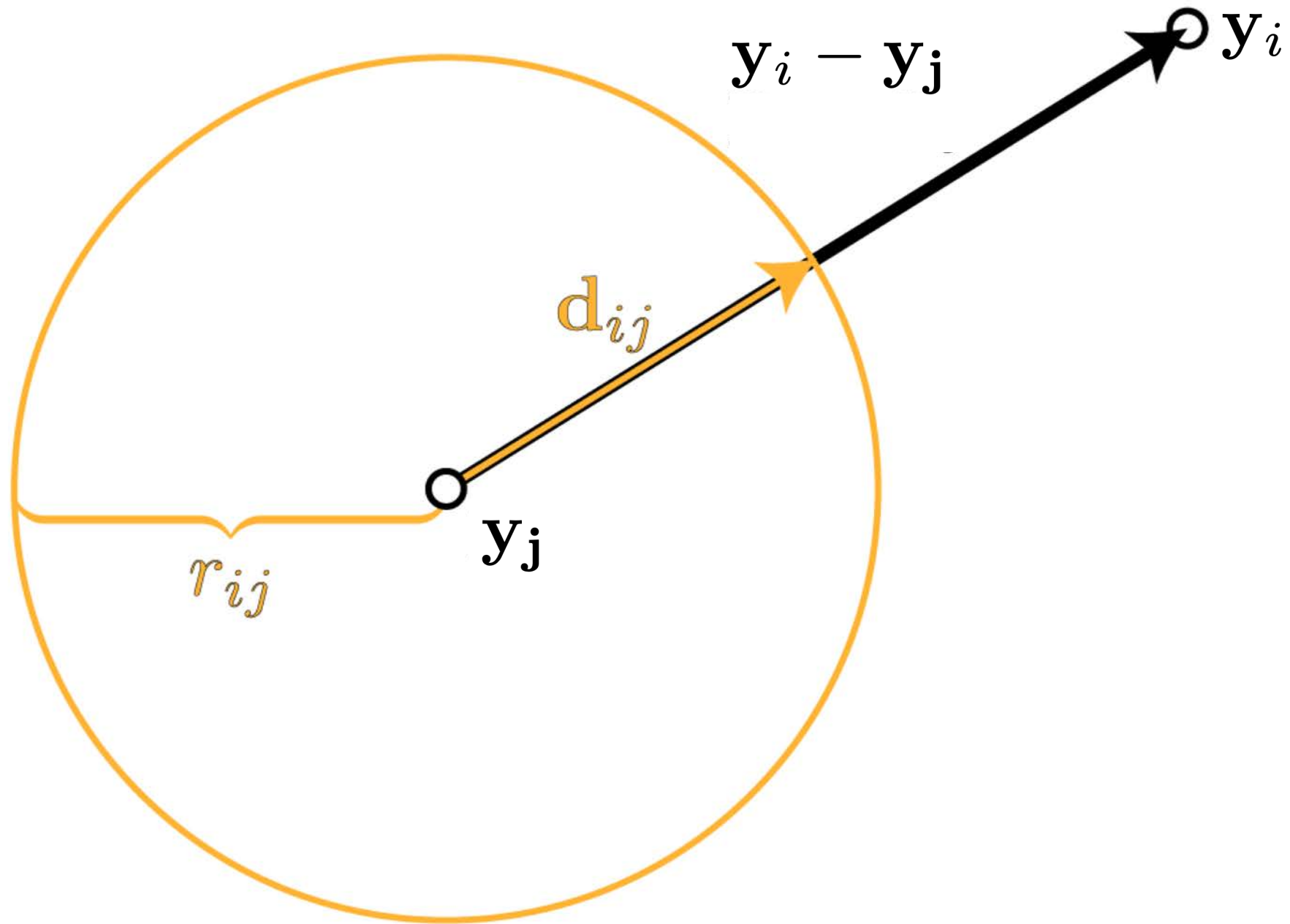
Now we can start defining these steps for mass-springs

Rethinking Potential Energy

$E_{ij} = \frac{k}{2} (\|\mathbf{y}_j - \mathbf{y}_i\| - r_{ij})^2$ is equivalent to

$$E_{ij} = \arg \min_{\mathbf{d}_{ij}, \|\mathbf{d}_{ij}\| = r_{ij}} \frac{k}{2} \|\mathbf{y}_i - \mathbf{y}_j - \mathbf{d}_{ij}\|^2$$

Given $\mathbf{y}_i - \mathbf{y}_j$ we can quickly find \mathbf{d}_{ij}



Why Do This ?

$$E_{ij} = \frac{k}{2} (\|\mathbf{y}_j - \mathbf{y}_i\| - r_{ij})^2 \text{ is equivalent to}$$

$$E_{ij} = \arg \min_{\mathbf{d}_{ij}, |\mathbf{d}_{ij}|=r_{ij}} \frac{k}{2} \|\mathbf{y}_i - \mathbf{y}_j - \mathbf{d}_{ij}\|^2$$

We can expand a bit more ...

$$E_{ij} = \arg \min_{\mathbf{d}_{ij}, |\mathbf{d}_{ij}|=r_{ij}} \frac{k}{2} \left[\|\mathbf{y}_i - \mathbf{y}_j\|^2 - (\mathbf{y}_i - \mathbf{y}_j)^T \mathbf{d}_{ij} + \mathbf{d}_{ij}^T \mathbf{d}_{ij} \right]$$

Aside from the constraints, this is a nice quadratic energy

Local-Global Solvers for Mass-Spring Systems

$$\mathbf{E}_1(\mathbf{y}^{t+1}) = \frac{1}{2} (\mathbf{y}^{t+1})^T M \mathbf{y}^{t+1} - (\mathbf{y}^{t+1})^T M \mathbf{b}$$

$$E_2 = \sum_{ij} \frac{k}{2} \left(\|\mathbf{y}_i - \mathbf{y}_j\|^2 - 2(\mathbf{y}_i - \mathbf{y}_j)^T \mathbf{d}_{ij} + \mathbf{d}_{ij}^T \mathbf{d}_{ij} \right)$$

Both energies are quadratic now. This will let us build a fast algorithm

We will do this using block coordinate descent. First optimize over one set of variables (the \mathbf{d} 's) then the second set (the \mathbf{y} 's) Rinse and repeat!

Local-Global Solvers for Mass-Spring Systems

$$\mathbf{y}^{t+1} = \arg \min_{\mathbf{y}, \mathbf{d}_{ij}, \|\mathbf{d}_{ij}\|=r_{ij}} E_1(\mathbf{y}) + \Delta t \dot{E}_2(\bar{\mathbf{y}}, \dot{\mathbf{d}}_{ij})$$

For **step 1** we will hold \mathbf{y} constant and minimize with respect to \mathbf{d} and its constraints

Note that this recovers the problem

$$\arg \min_{\mathbf{d}_{ij}, \|\mathbf{d}_{ij}\|=r_{ij}} \sum_{ij} \frac{k}{2} \left[\underbrace{\|\mathbf{y}_i - \mathbf{y}_j\|^2 - 2(\mathbf{y}_i - \mathbf{y}_j)^T \mathbf{d}_{ij} + \mathbf{d}_{ij}^T \mathbf{d}_{ij}}_{\text{Each } \mathbf{d} \text{ acts on a spring independently!}} \right]$$

Each \mathbf{d} acts on a spring independently!

The Local Step

This gives us our local step:

$$\arg \min_{\mathbf{d}_{ij}, |\mathbf{d}_{ij}|=r_{ij}} \sum_{ij} \frac{k}{2} \|\mathbf{y}_i - \mathbf{y}_j\|^2 - 2(\mathbf{y}_i - \mathbf{y}_j)^T \mathbf{d}_{ij} + \mathbf{d}_{ij}^T \mathbf{d}_{ij}$$

Can be minimized by visiting each spring and finding \mathbf{d} such that

$$E_{ij} = \arg \min_{\mathbf{d}_{ij}, |\mathbf{d}_{ij}|=r_{ij}} \underbrace{\frac{k}{2} \|\mathbf{y}_i - \mathbf{y}_j - \mathbf{d}_{ij}\|^2}$$

No sum anymore!

Local-Global Solvers for Mass-Spring Systems

$$\mathbf{E}_1(\mathbf{y}^{t+1}) = \frac{1}{2} (\mathbf{y}^{t+1})^T M \mathbf{y}^{t+1} - (\mathbf{y}^{t+1})^T M \mathbf{b}$$

$$E_2 = \sum_{ij} \frac{k}{2} \left(\|\mathbf{y}_i - \mathbf{y}_j\|^2 - 2(\mathbf{y}_i - \mathbf{y}_j)^T \mathbf{d}_{ij} + \mathbf{d}_{ij}^T \mathbf{d}_{ij} \right)$$

Both energies are quadratic now. This will let us build a fast algorithm

We will do this using block coordinate descent. First optimize over one set of variables (the \mathbf{d} 's) then the second set (the \mathbf{y} 's) Rinse and repeat!

The Global Step

Minimizing wrt to \mathbf{y} requires us to find

$$\mathbf{y}^{t+1} \text{ s.t. } \nabla_{\mathbf{y}}(E_1(\mathbf{y}) + \Delta t E_2(\mathbf{y}, \mathbf{d}_{ij})) = \mathbf{0}$$

Recall $\mathbf{E}_1(\mathbf{y}) = \frac{1}{2}\mathbf{y}^T M \mathbf{y} - \mathbf{y}^T M \mathbf{b}$

$$\nabla \mathbf{E}_1 = M \mathbf{y} - M \mathbf{b}$$

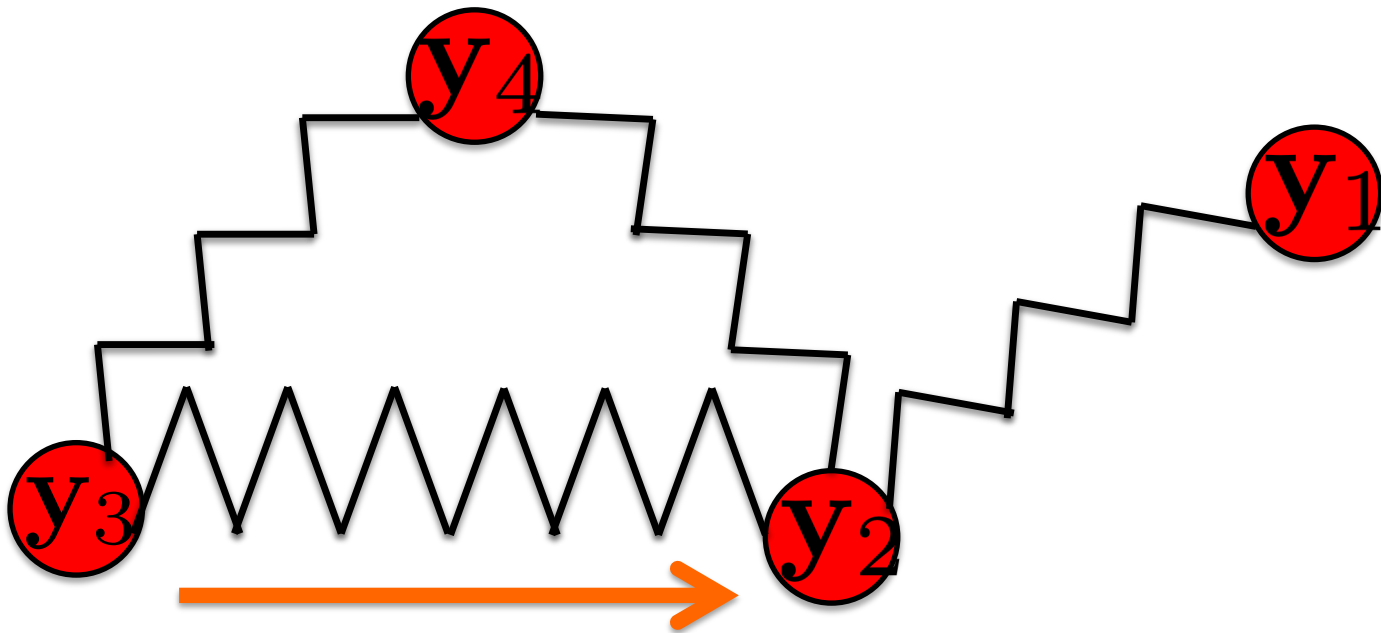
Not so bad ...

The Global Step

$$E_2 = \sum_{ij} \frac{k}{2} \left(\|\mathbf{y}_i - \mathbf{y}_j\|^2 - 2(\mathbf{y}_i - \mathbf{y}_j)^T \mathbf{d}_{ij} + \mathbf{d}_{ij}^T \mathbf{d}_{ij} \right)$$

This is a little trickier.

Global Step



$$\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix}$$

$$\Delta \mathbf{y} = \underbrace{\begin{pmatrix} I & -I & 0 & 0 \\ 0 & I & -I & 0 \\ 0 & I & 0 & -I \\ 0 & -I & I & -I \end{pmatrix}}_{\mathbf{G}} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix}$$

Each row is a spring

Global Step

Using this we can rewrite the second energy as

$$E_2 = \frac{k}{2} (\mathbf{y} G^T G \mathbf{y} - 2 \mathbf{y}^T G^T \mathbf{d} + \mathbf{d}^T \mathbf{d}) \quad \mathbf{d} = \begin{pmatrix} \mathbf{d}_{12} \\ \mathbf{d}_{23} \\ \mathbf{d}_{24} \\ \mathbf{d}_{34} \end{pmatrix}$$

So the gradient becomes

$$\nabla E_2 = k G^T G \mathbf{y} - k \mathbf{G}^T \mathbf{d}$$

And the total global step finds \mathbf{y} so that

$$\nabla (E_1 + \Delta t^2 E_2) = (M + \Delta t^2 k G^T G) \mathbf{y} - (M \mathbf{b} - \Delta t^2 k \mathbf{G}^T \mathbf{d}) = 0$$

Global Step

And the total global step finds \mathbf{y} so that

$$\nabla(E_1 + \Delta t^2 E_2) = (M + \Delta t^2 k G^T G) \mathbf{y} - (M \mathbf{b} + \Delta t^2 k \mathbf{G}^T \mathbf{d}) = 0$$

Or

$$(M + \Delta t^2 k G^T G) \mathbf{y} = (M \mathbf{b} + \Delta t^2 k \mathbf{G}^T \mathbf{d})$$

You can solve this linear system using the Cholesky Solver
in Eigen

Local-Global Solvers for Mass-Spring Systems

WHILE Not done

//Local Steps

For Each Spring

$$E_{ij} = \arg \min_{\mathbf{d}_{ij}, |\mathbf{d}_{ij}|=r_{ij}} \frac{k}{2} \|\mathbf{y}_i - \mathbf{y}_j - \mathbf{d}_{ij}\|^2$$

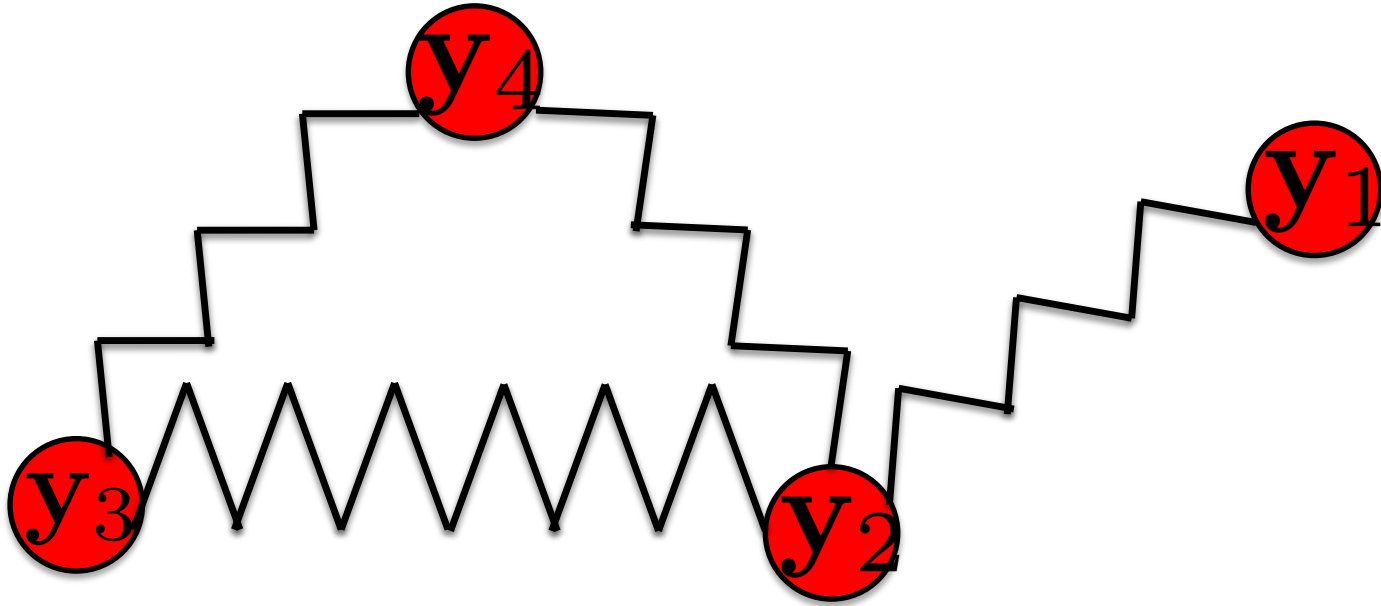
//Global Step

$$\text{Solve } (M + \Delta t^2 k G^T G) \mathbf{y} = (M \mathbf{b} + \Delta t^2 k \mathbf{G}^T \mathbf{d})$$

END



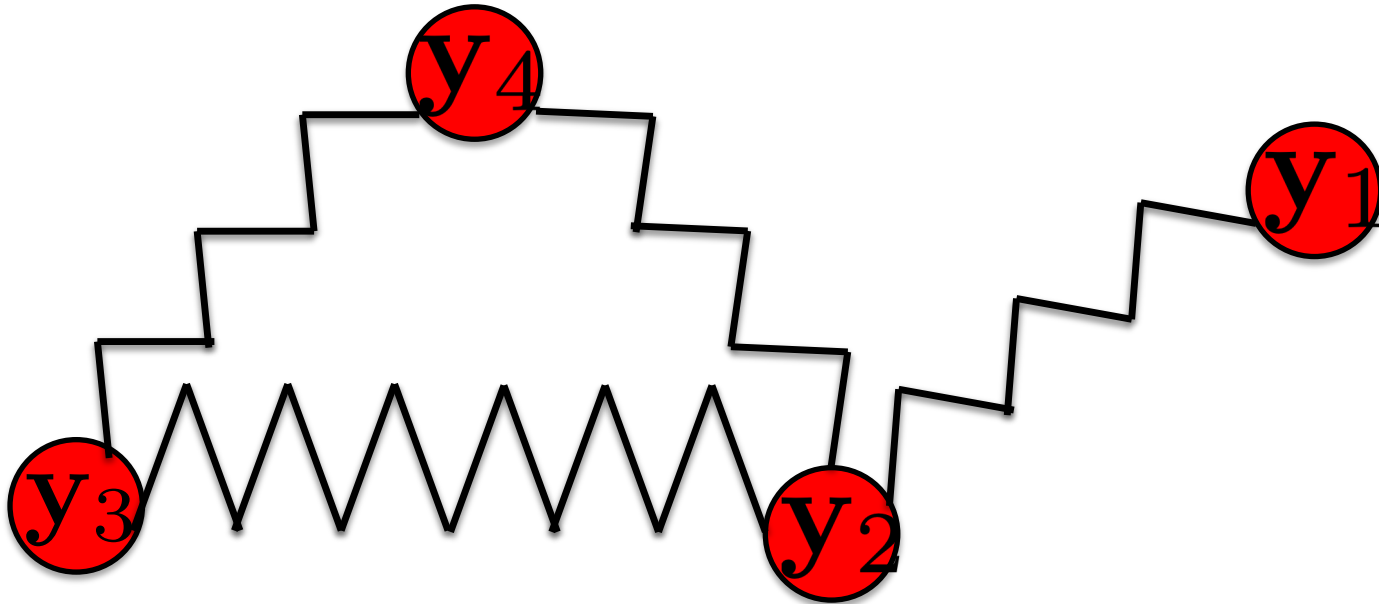
Fixed Points



$$\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix}$$

Let's say we never want y_3 to move

Fixed Points

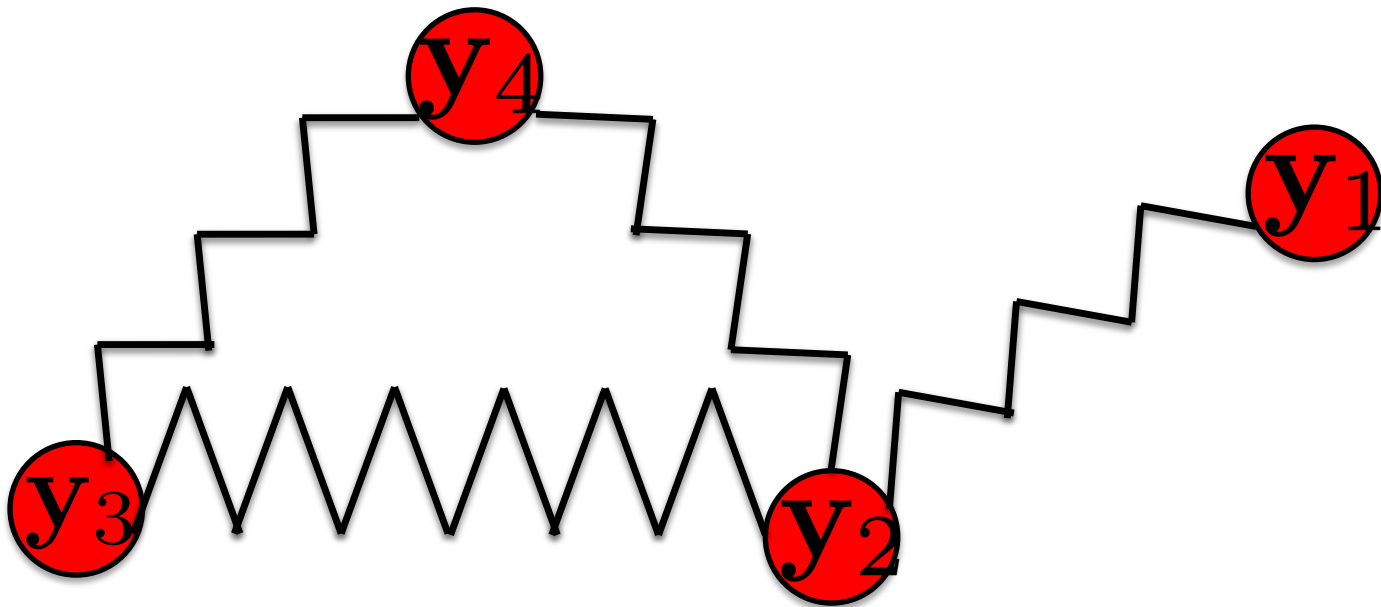


$$\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix}$$

Let's say we never want y_3 to move

i.e $y_3 = \mathbf{c}$ forever and always

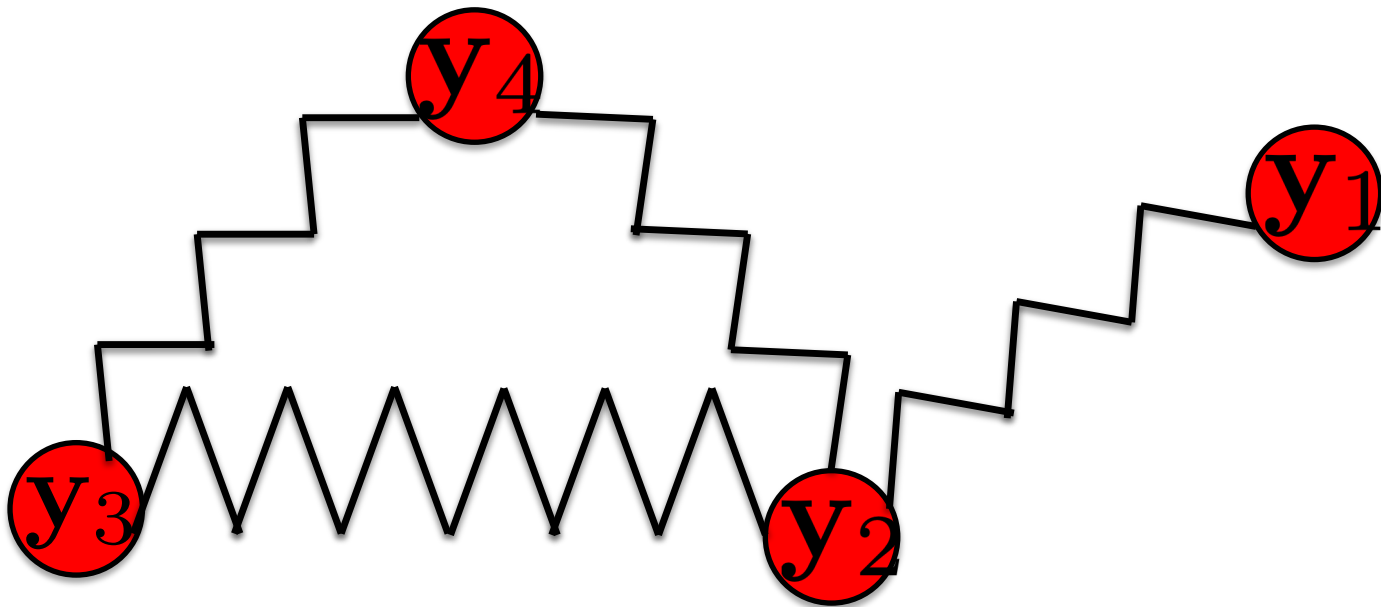
Fixed Points via Projection



$$\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix}$$

$$\mathbf{y} = \underbrace{\begin{pmatrix} I & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & 0 \\ 0 & 0 & I \end{pmatrix}}_{\mathbf{P}} \underbrace{\begin{pmatrix} y_1 \\ y_2 \\ y_4 \end{pmatrix}}_{\tilde{\mathbf{y}}} + \underbrace{\begin{pmatrix} 0 \\ 0 \\ \mathbf{c} \\ 0 \end{pmatrix}}_{\mathbf{c}}$$

Fixed Points via Projection

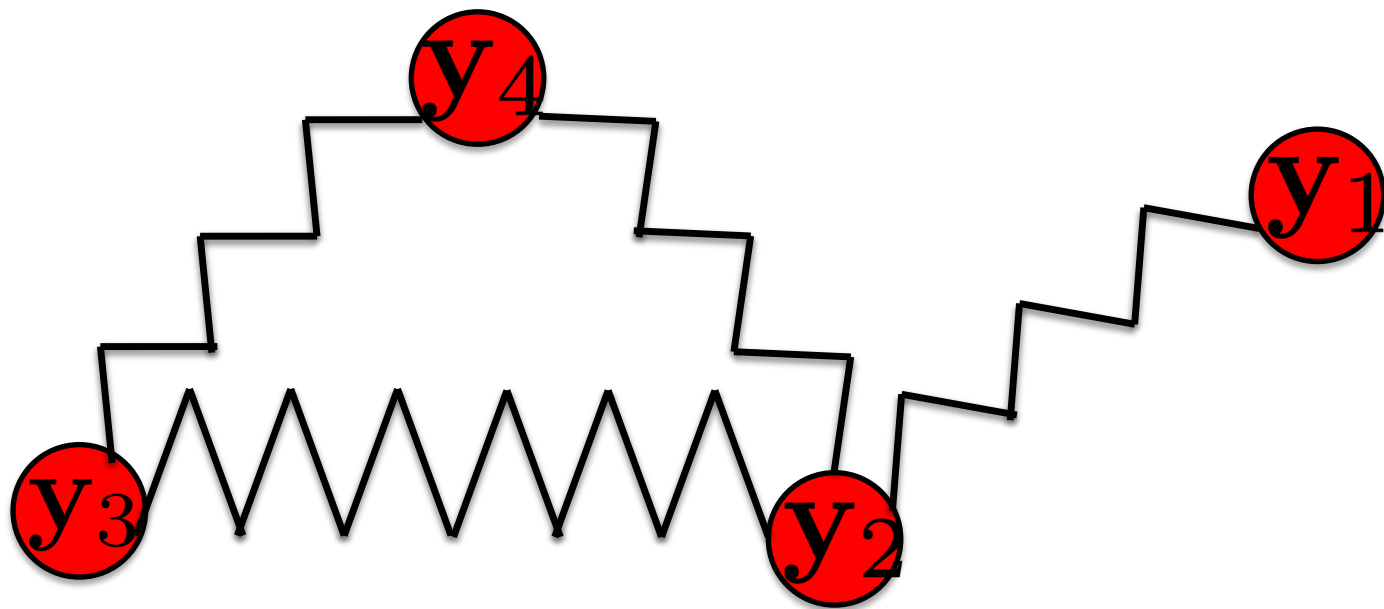


$$(M + \Delta t^2 k G^T G) \mathbf{y} = (M \mathbf{b} - \Delta t^2 k \mathbf{G}^T \mathbf{d})$$

\mathbf{A}

\mathbf{f}

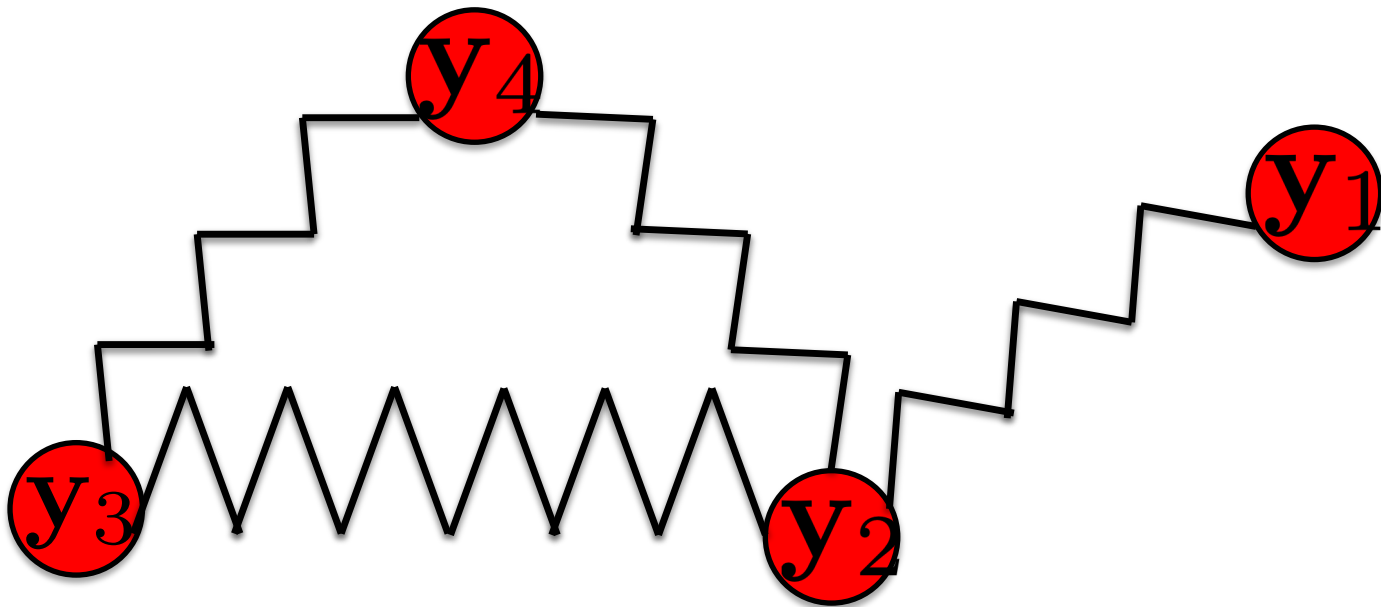
Fixed Points via Projection



$$A\mathbf{y} = \mathbf{f}$$

$$\mathbf{y} = P\tilde{\mathbf{y}} + \mathbf{c}$$

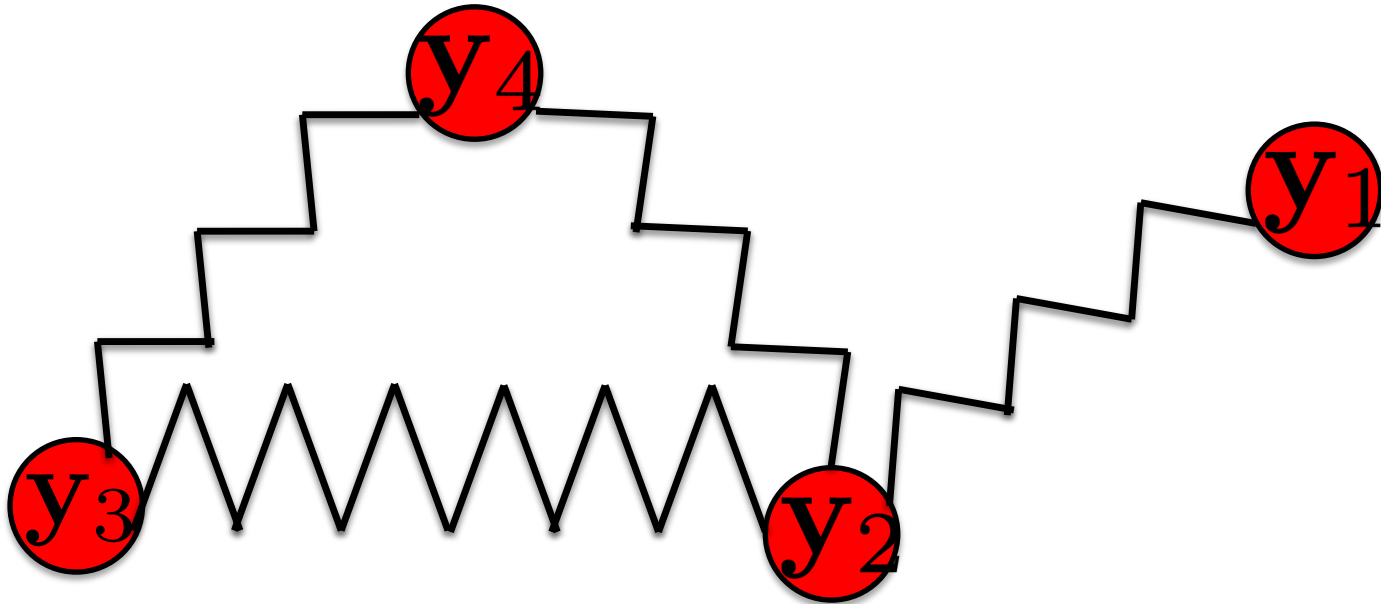
Fixed Points via Projection



$$AP\tilde{\mathbf{y}} = \mathbf{f} - A\mathbf{c}$$

Too many rows now ...

Fixed Points via Projection

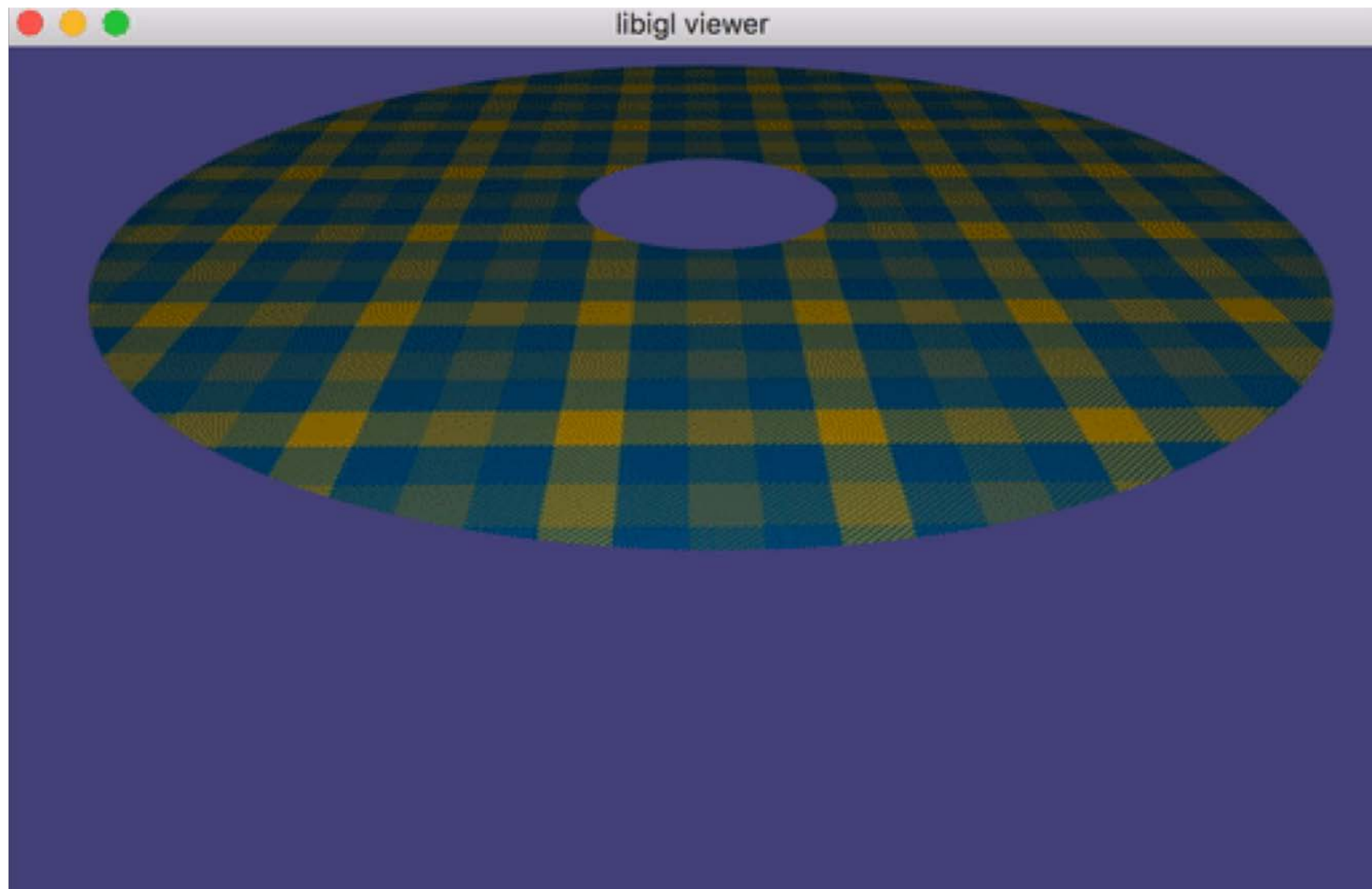


$$P^T A P \tilde{\mathbf{y}} = P^T (\mathbf{f} - A \mathbf{c})$$

Just right ... but don't forget to rebuild \mathbf{y} after solving

Lots more on the Assignment Page

So please read it carefully when doing the assignment



Done for Today

Office hours: Right now! BA5268