

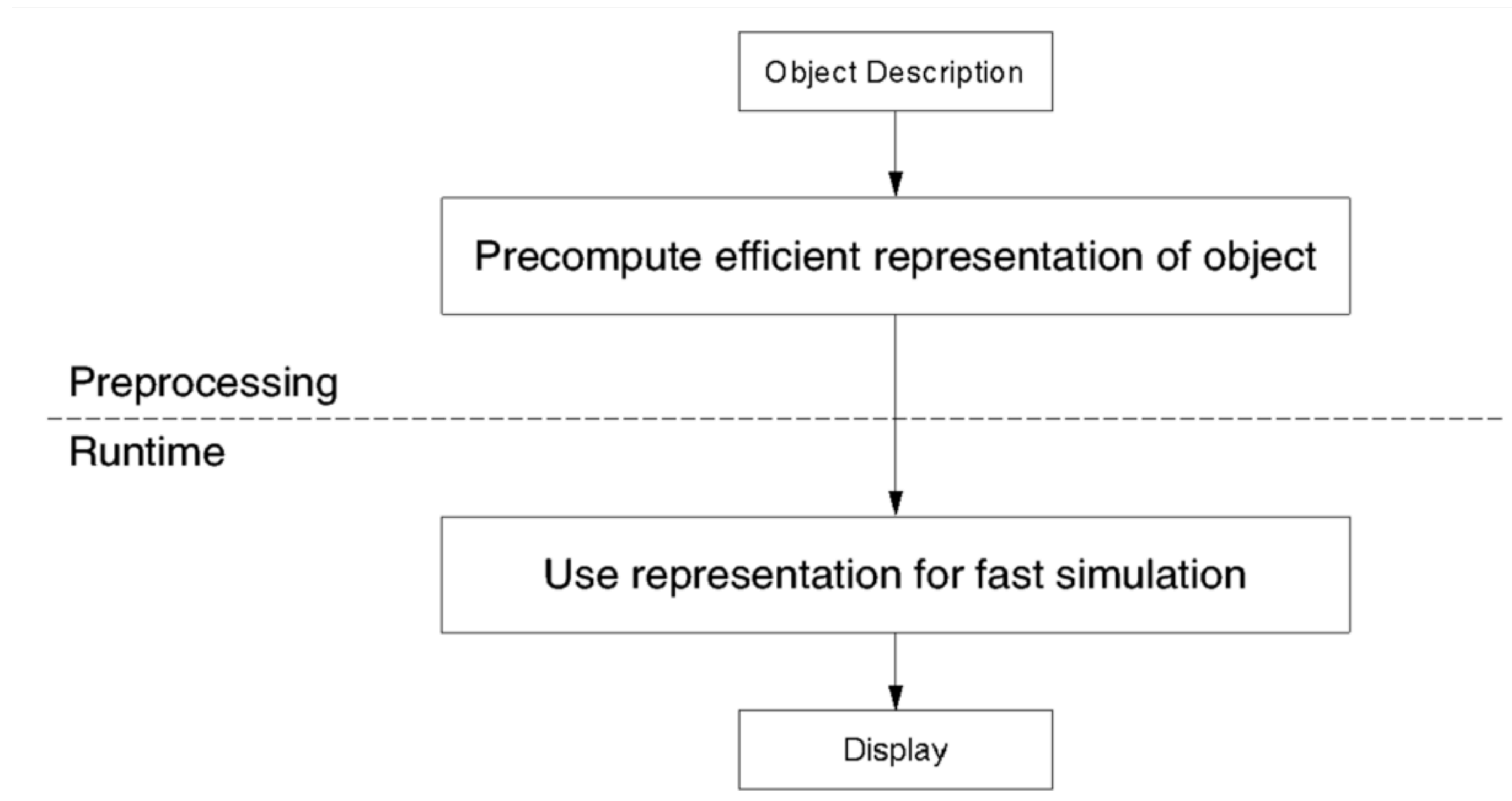
Lecture 21a:

Modal Analysis

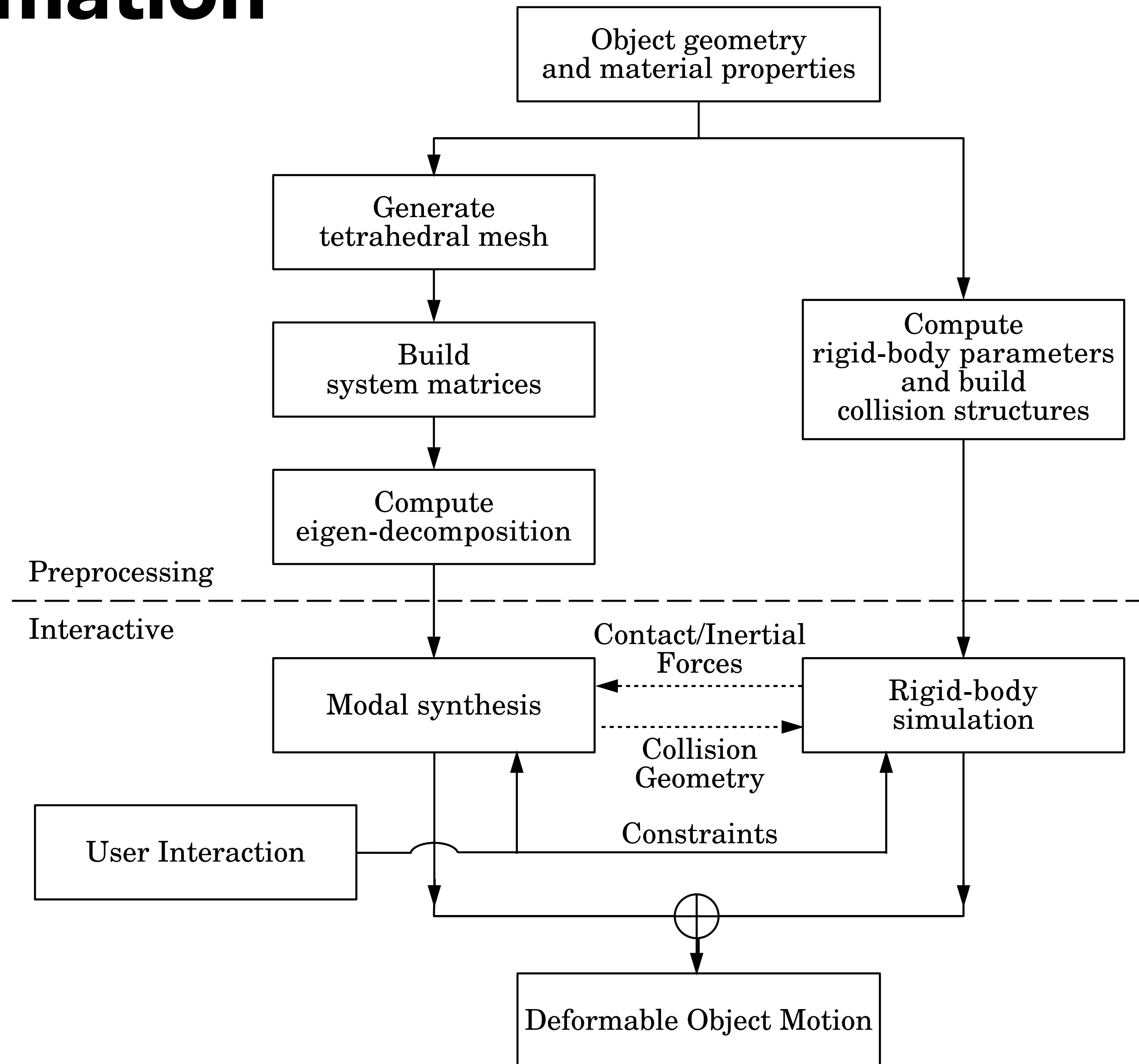
Computer Graphics and Imaging
UC Berkeley CS184/284A

Note: The math on gray slides will not be on the exams.

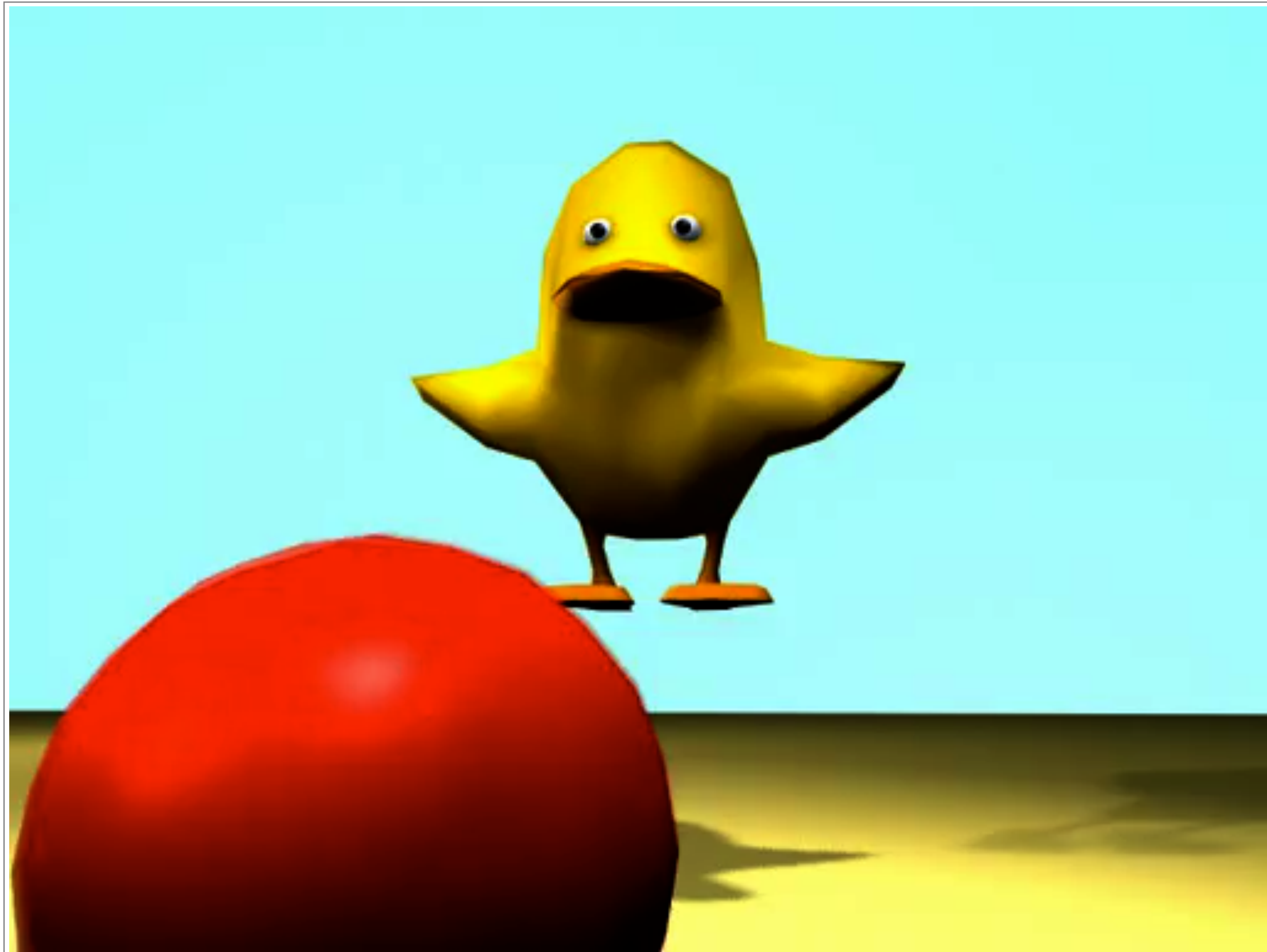
Modal Analysis



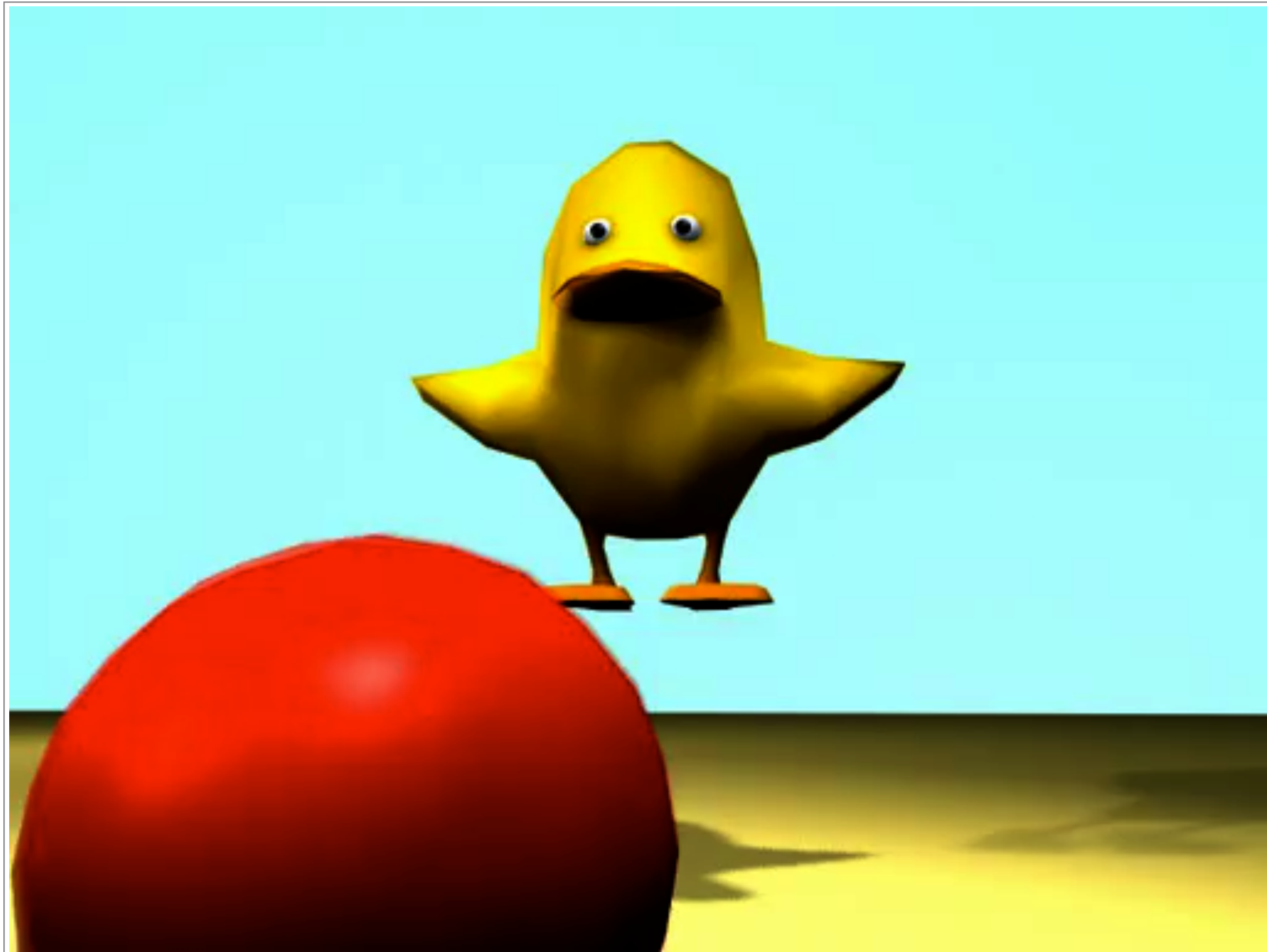
Deformation

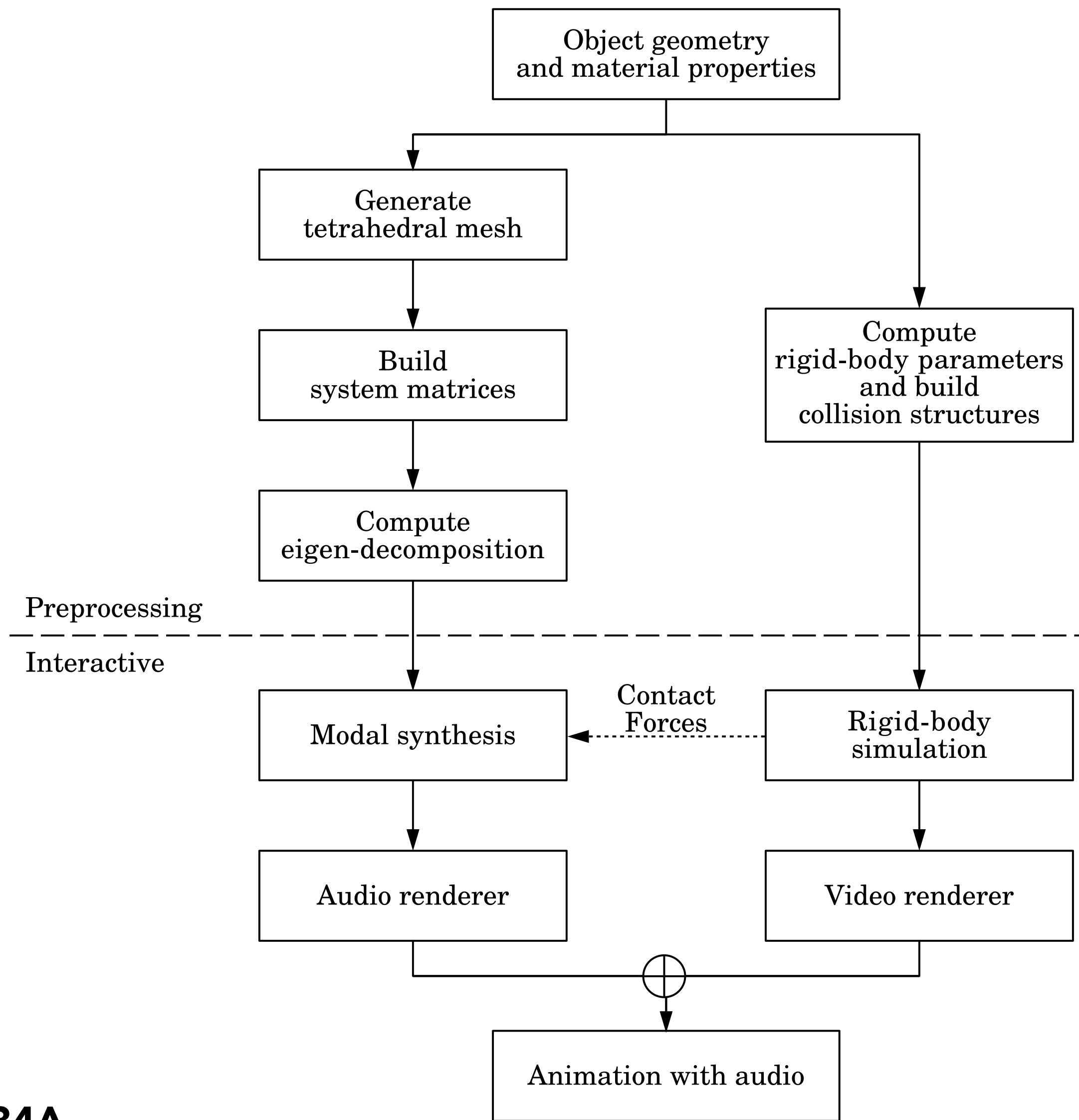


Example

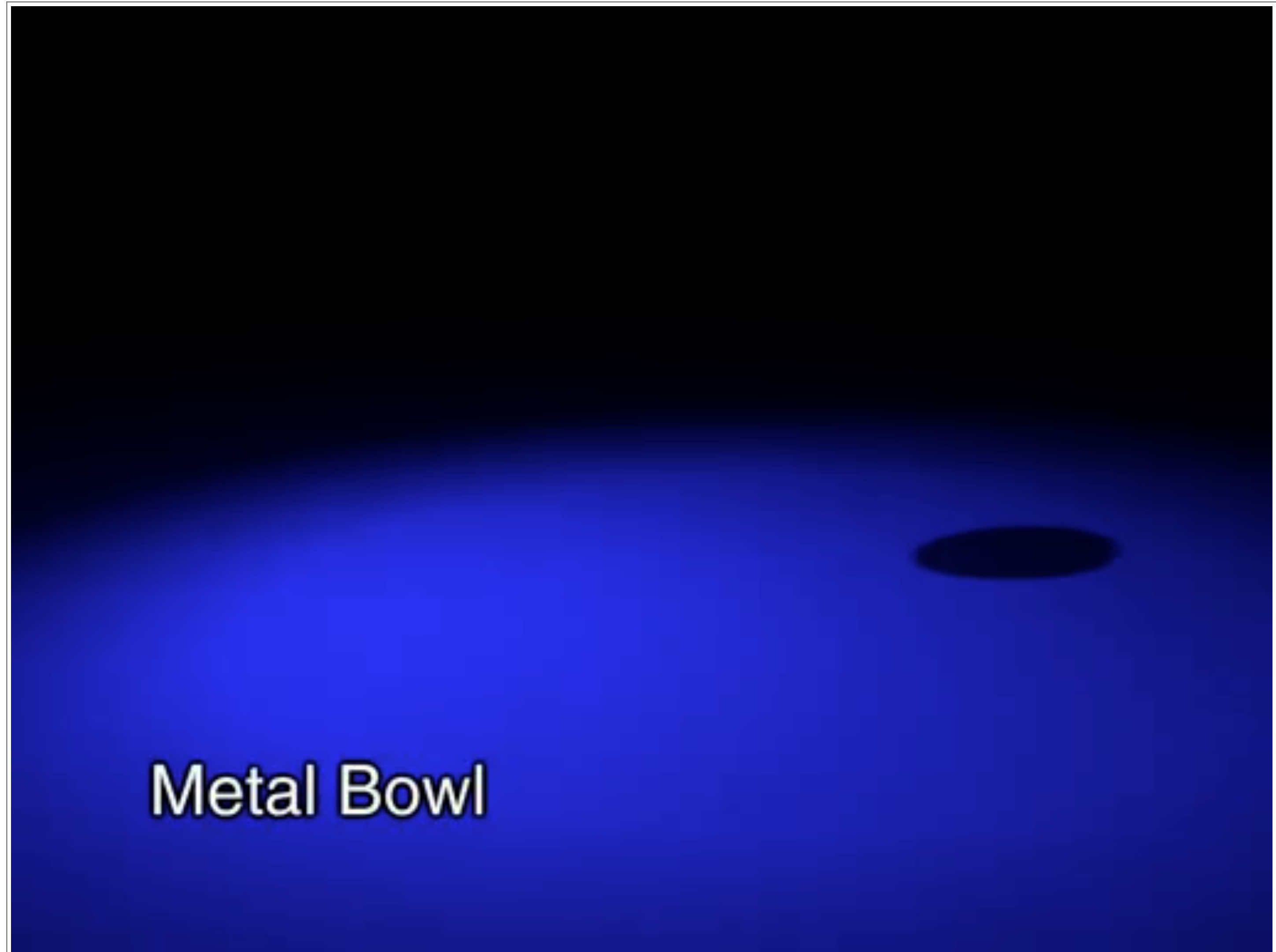


Example

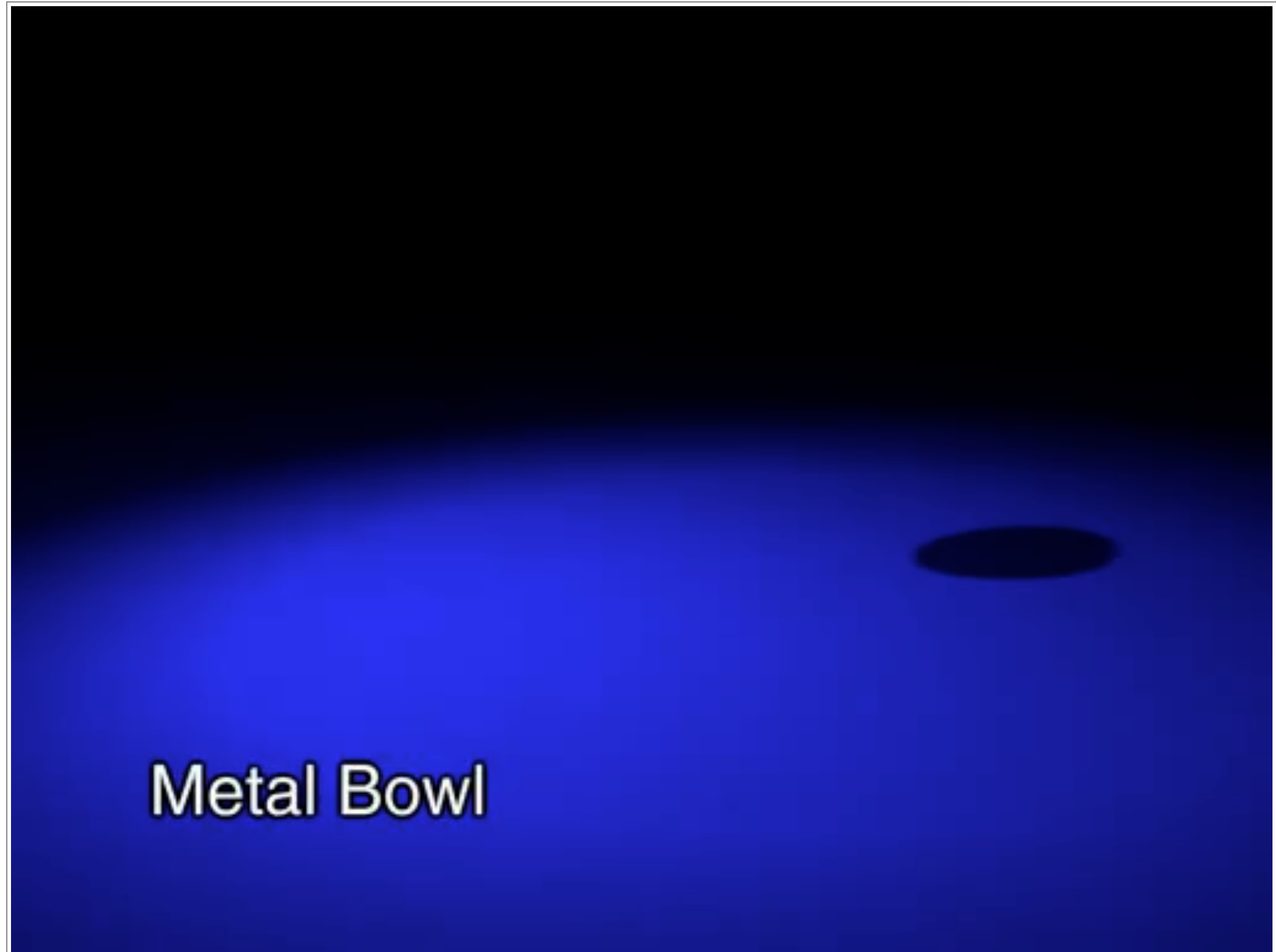




Example



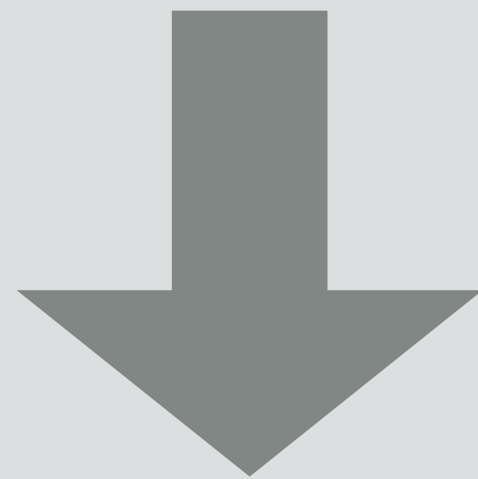
Example



Modal Decomposition

Linearize non-linear system

$$\mathcal{K}(d) + \mathcal{C}(d, \dot{d}) + \mathcal{M}(\ddot{d}) = f$$

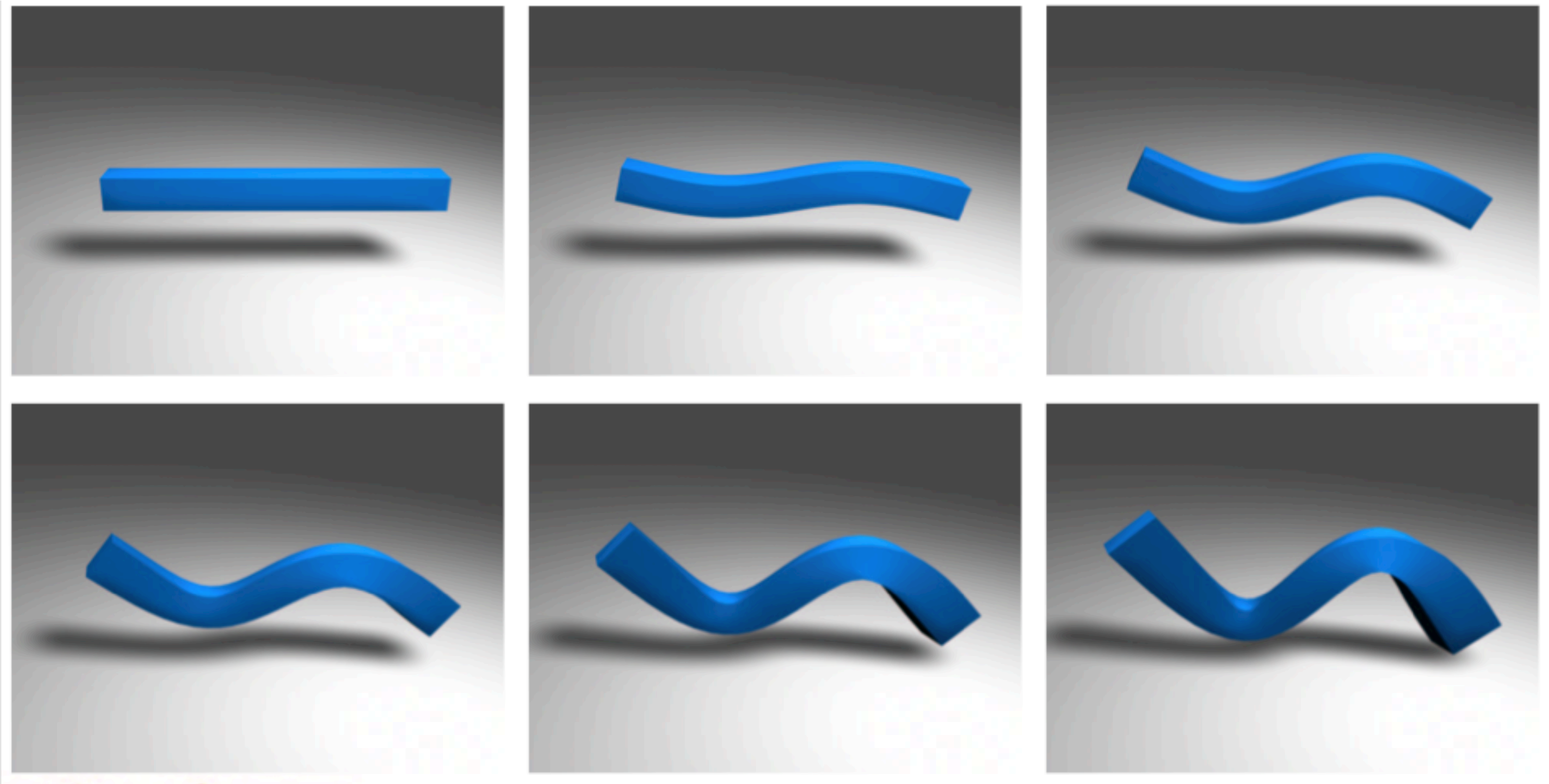


$$Kd + C\dot{d} + M\ddot{d} = f$$

Modal Decomposition

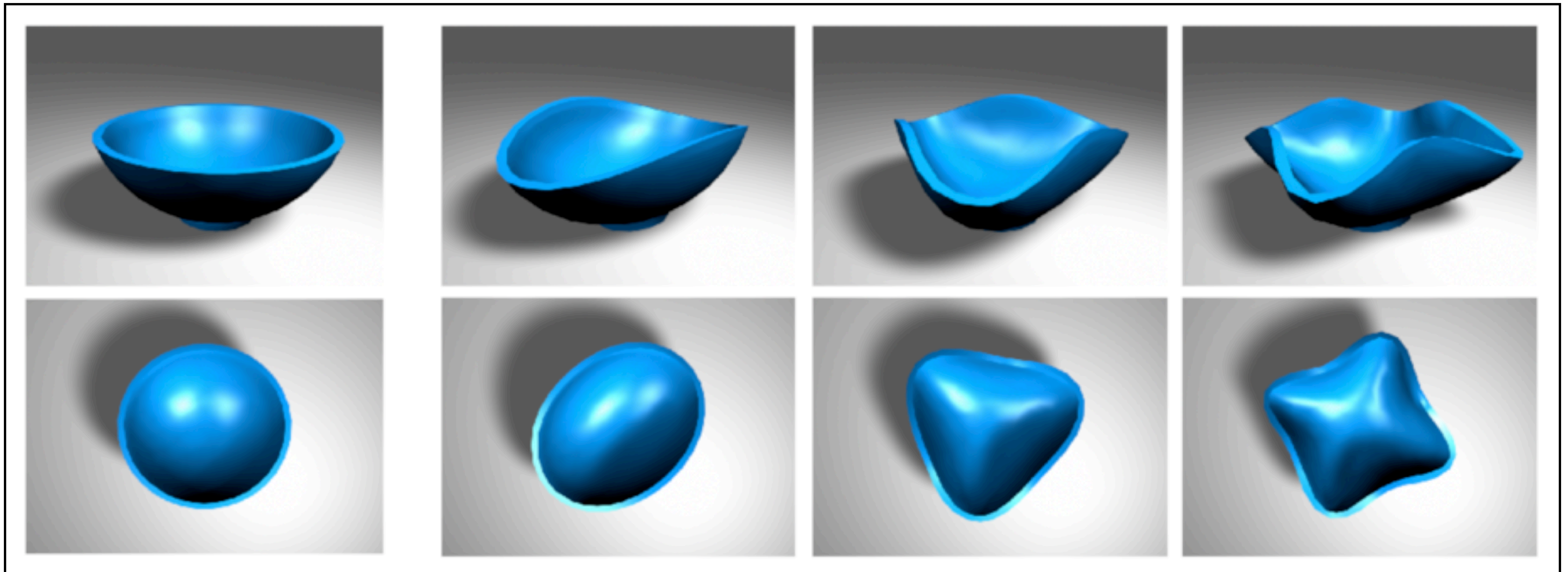
Consequences of linearization

- No local rotations



Modal Decomposition

$$\Lambda \mathbf{z} + (\alpha_1 \Lambda + \alpha_2 \mathbf{I}) \dot{\mathbf{z}} + \ddot{\mathbf{z}} = \mathbf{g}$$



Step 1: Linearization

$$\mathcal{K}(d) + \mathcal{C}(\dot{d}) + \mathcal{M}(\ddot{d}) = f$$

$$Kd + C\dot{d} + M\ddot{d} = f$$

Assume: $C = \alpha_1 K + \alpha_2 M$

$$K(d + \alpha_1 \dot{d}) + M(\alpha_2 \dot{d} + \ddot{d}) = f$$

Step 2: Normalize Mass

Normalize for mass by change of coordinates:

- Cholesky decompositions: $\mathbf{M} = \mathbf{L}\mathbf{L}^\top$
- Change of variables: $\mathbf{y} = \mathbf{L}^\top \mathbf{d}$ $\mathbf{d} = \mathbf{L}^{-\top} \mathbf{y}$

$$\mathbf{K}(\mathbf{d} + \alpha_1 \dot{\mathbf{d}}) + \mathbf{M}(\alpha_2 \dot{\mathbf{d}} + \ddot{\mathbf{d}}) = \mathbf{f}$$





$$\mathbf{L}^{-1} \mathbf{K} \mathbf{L}^{-\top} (\mathbf{y} + \alpha_1 \dot{\mathbf{y}}) + (\alpha_2 \dot{\mathbf{y}} + \ddot{\mathbf{y}}) = \mathbf{L}^{-1} \mathbf{f}$$

Step 3: Diagonalize

Diagonalize with second change of coordinates:

- Eigen decompositions: $\mathbf{L}^{-1}\mathbf{K}\mathbf{L}^{-\top} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^{\top}$
- Change of variables: $\mathbf{z} = \mathbf{V}^{\top}\mathbf{y}$ $\mathbf{y} = \mathbf{V}\mathbf{z}$

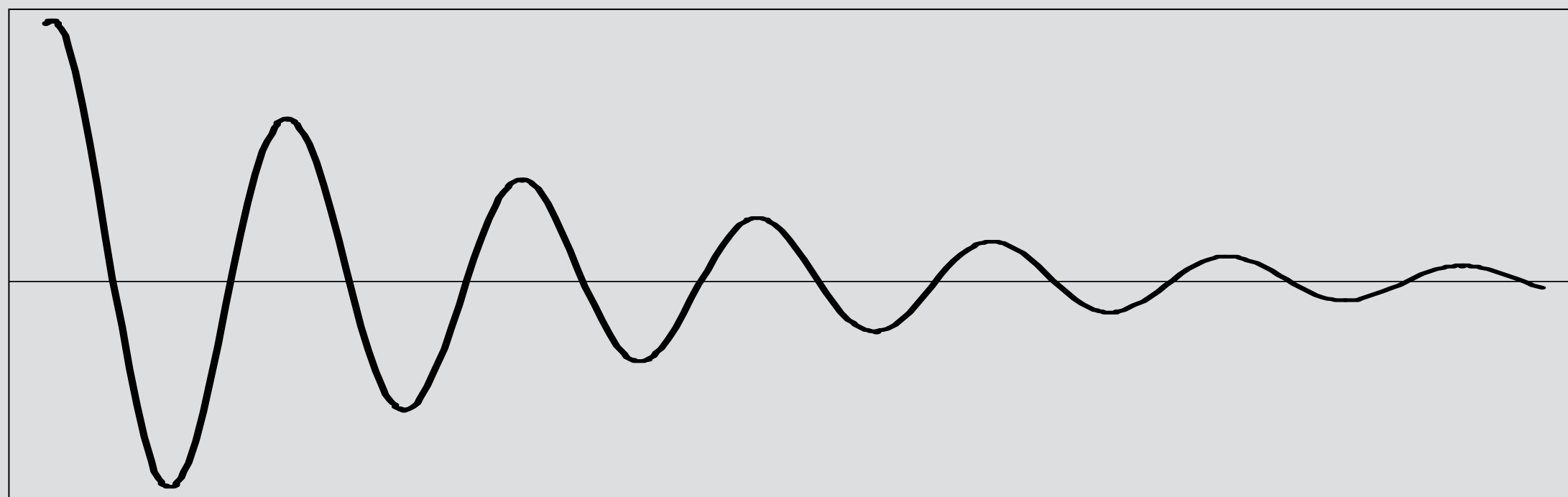
$$\mathbf{L}^{-1}\mathbf{K}\mathbf{L}^{-\top}(\mathbf{y} + \alpha_1\dot{\mathbf{y}}) + (\alpha_2\dot{\mathbf{y}} + \ddot{\mathbf{y}}) = \mathbf{L}^{-1}\mathbf{f}$$

$$\mathbf{\Lambda}(\mathbf{z} + \alpha_1\dot{\mathbf{z}}) + (\alpha_2\dot{\mathbf{z}} + \ddot{\mathbf{z}}) = \mathbf{V}^{\top}\mathbf{L}^{-1}\mathbf{f}$$

$$\mathbf{\Lambda}\mathbf{z} + (\alpha_1\mathbf{\Lambda} + \alpha_2\mathbf{I})\dot{\mathbf{z}} + \ddot{\mathbf{z}} = \mathbf{g}$$

Result: Individual Modes

$$\lambda_i z_i + (\alpha_1 \lambda_i + \alpha_2) \dot{z}_i + \ddot{z}_i = g_i$$

$$z_i = c_1 e^{t\omega_i^+} + c_2 e^{t\omega_i^-}$$

$$\omega_i^\pm = \frac{-(\alpha_1 \lambda_i + \alpha_2) \pm \sqrt{(\alpha_1 \lambda_i + \alpha_2)^2 - 4\lambda_i}}{2}$$



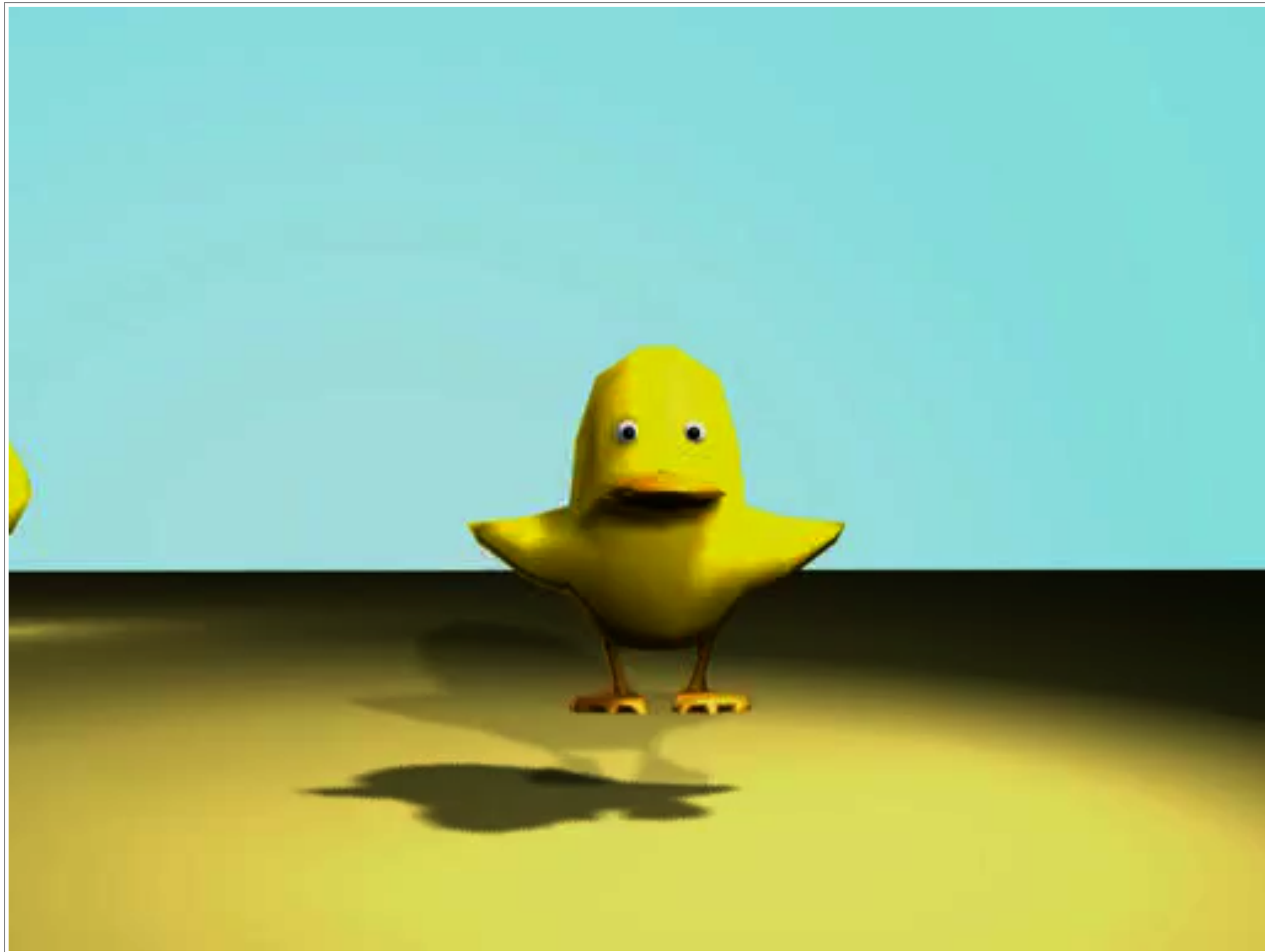
Fast Simulation

Only a pair of complex multiplies per time step

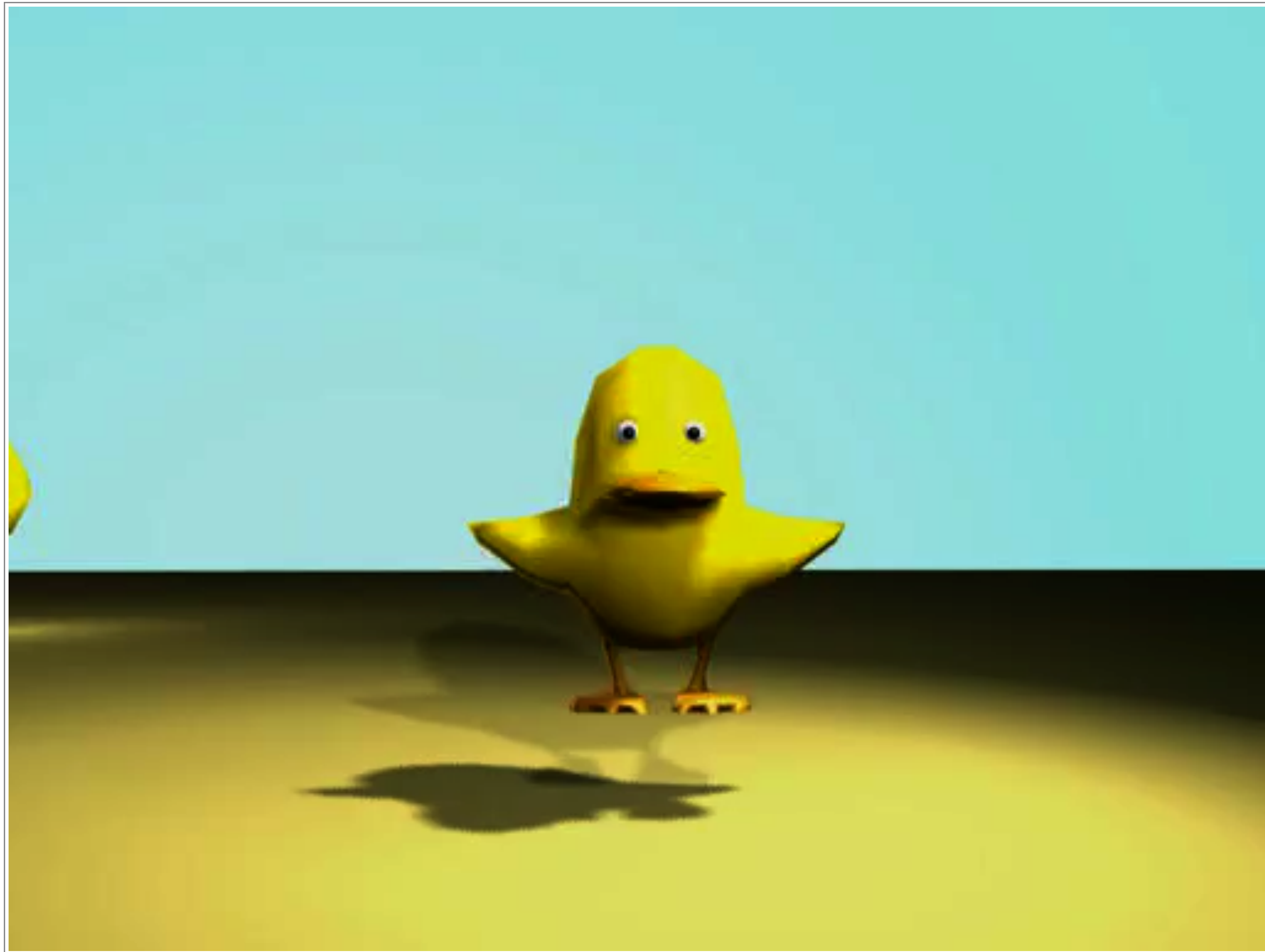
$$e^{\omega(t+\Delta t)} = e^{\omega(t)} e^{\omega(\Delta t)}$$

- No stability limit on step size
- Jump to arbitrary point in time
- Only keep useful modes

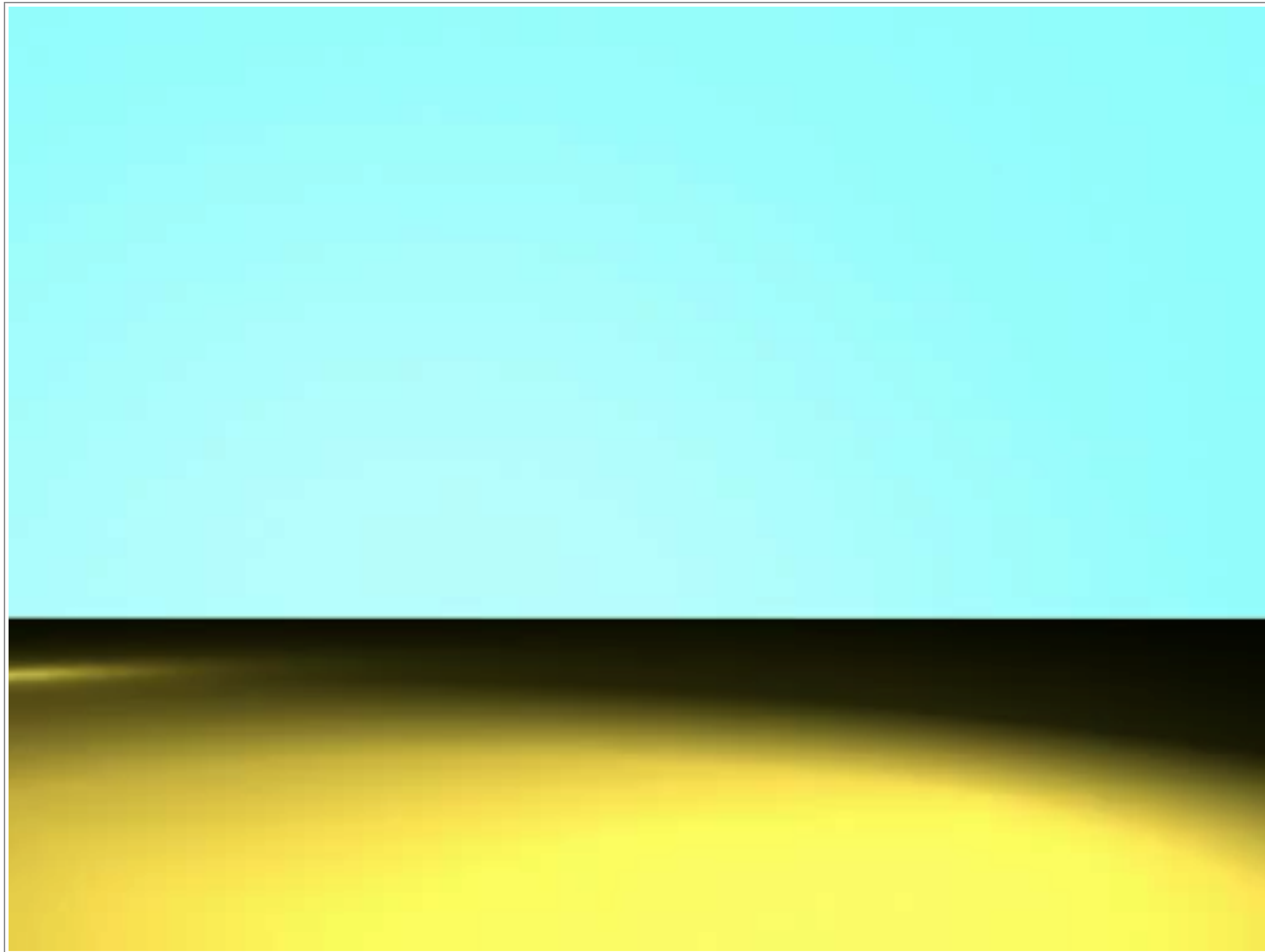
Examples



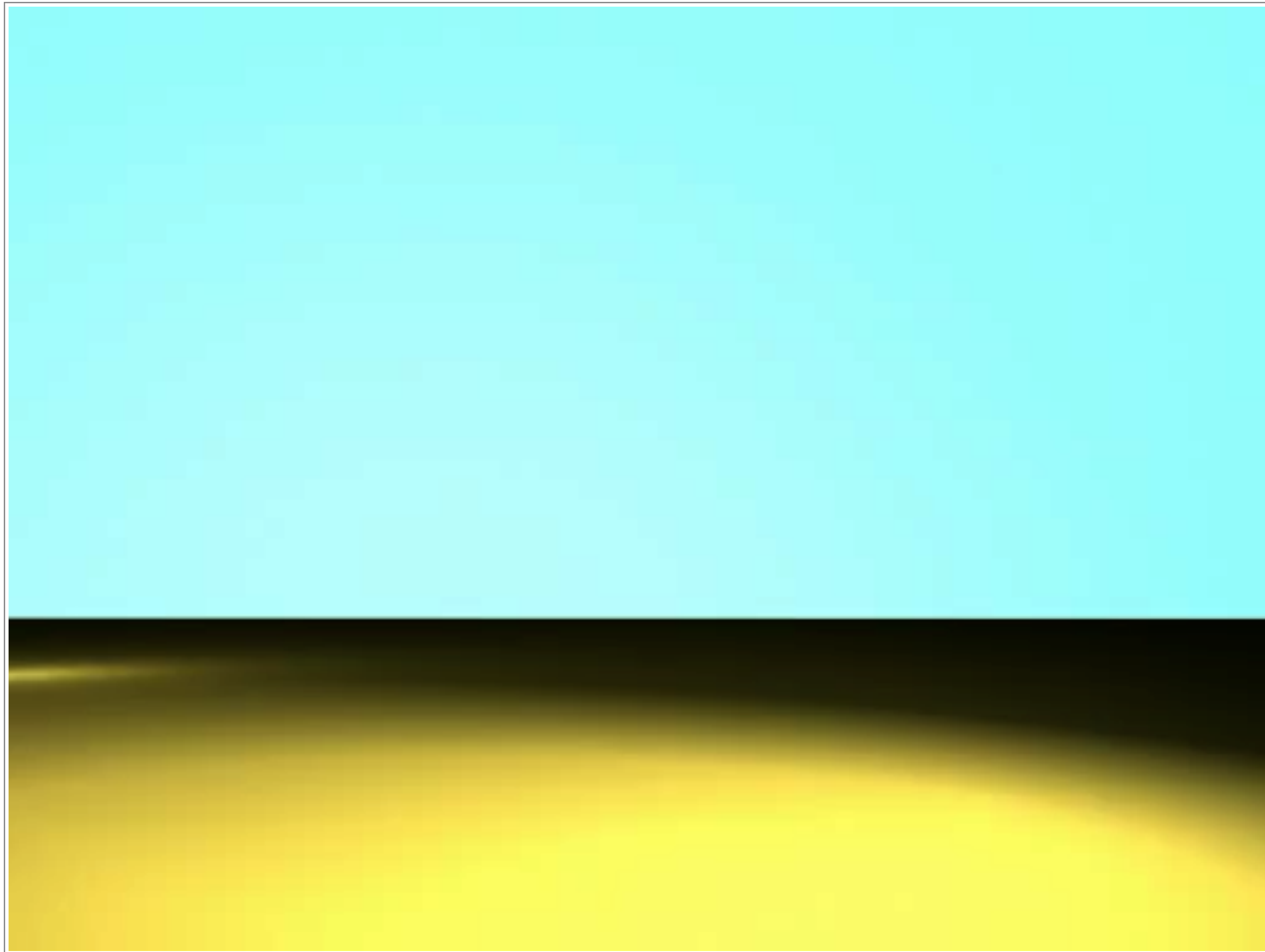
Examples



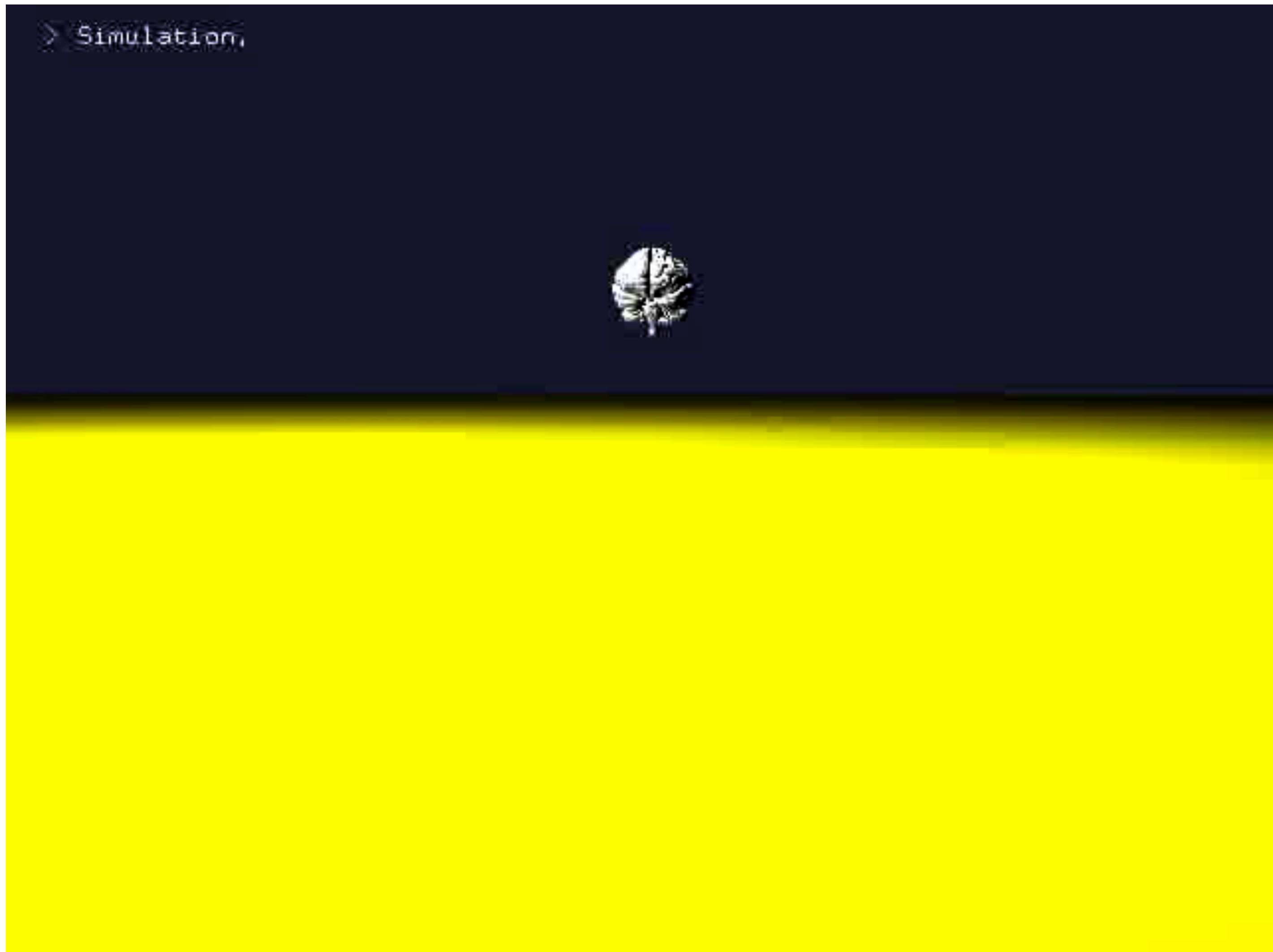
Examples



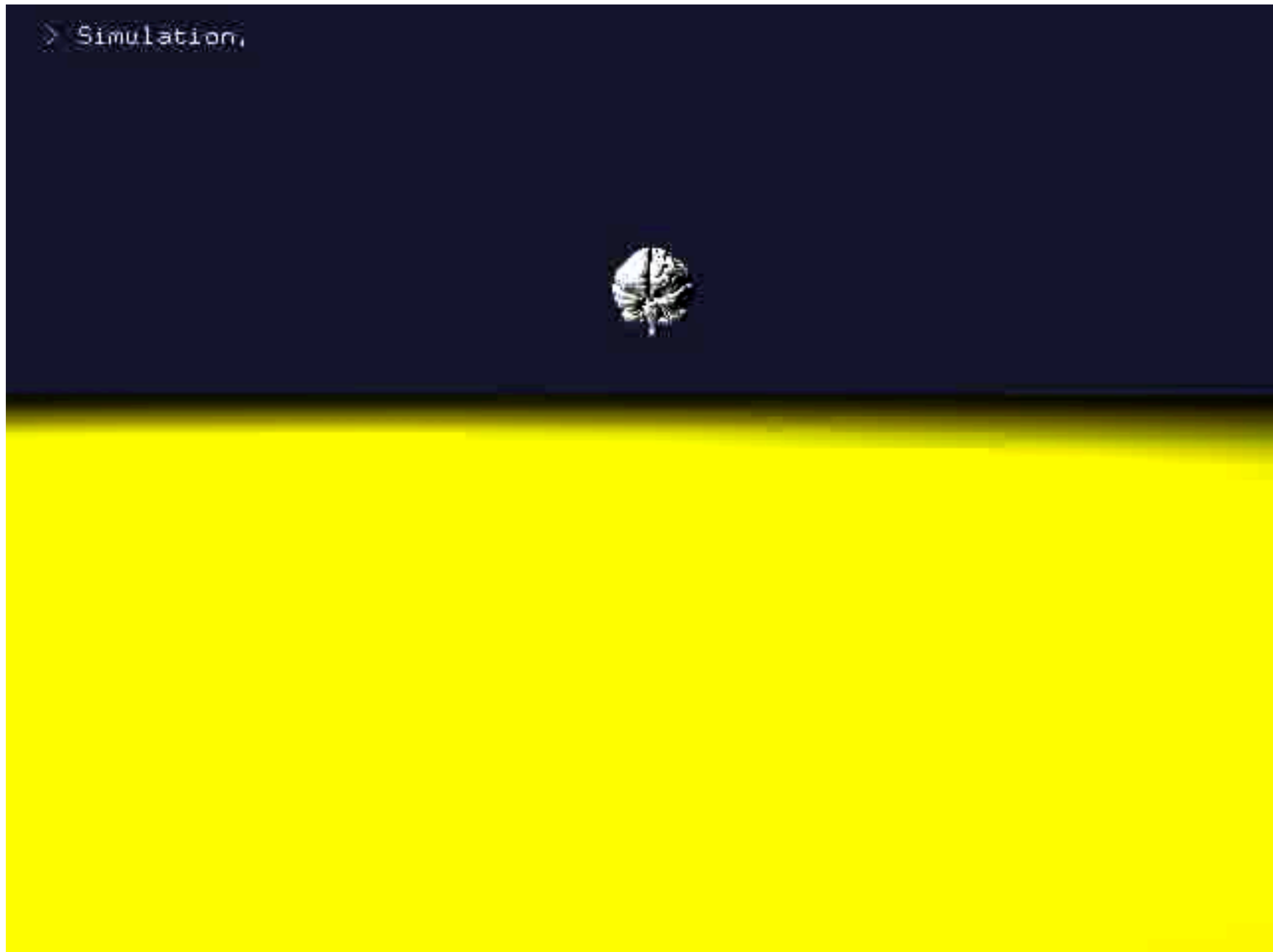
Examples



Examples



Examples



Examples

Synthesizing Sounds from Rigid–Body Simulation

James F. O'Brien
Chen Shen
Christine M. Gatchalian

University of California, Berkeley

ACM SIGGRAPH Symposium on Computer Animation 2002

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High-order Differentiable Autoencoder for Nonlinear Model Reduction

SIGGRAPH 2021, Shen, *et. al*

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Backward vs Forward Euler Integration

$$\mathbf{x}^{t+1} = \mathbf{x}^t + \Delta t \dot{\mathbf{x}}^t$$

$$\dot{\mathbf{x}}^{t+1} = \dot{\mathbf{x}}^t + \Delta t \ddot{\mathbf{x}}^t = \dot{\mathbf{x}}^t + \Delta t \mathbf{M}^{-1} \mathbf{f}(\mathbf{x}^t, \dot{\mathbf{x}}^t)$$

Forward Euler

$$\Delta \mathbf{x}_F = \Delta t \dot{\mathbf{x}}^t$$

$$\Delta \dot{\mathbf{x}}_F = \Delta t \mathbf{M}^{-1} \mathbf{f}^t$$

Backward (semi-implicit) Euler

$$\Delta \mathbf{x}_B = \Delta t \dot{\mathbf{x}}^{t+1}$$

$$\Delta \dot{\mathbf{x}}_B = \Delta t \mathbf{M}^{-1} \mathbf{f}^{t+1} \approx \Delta t \mathbf{M}^{-1} (\mathbf{f}^t + \mathbf{K} \Delta \mathbf{x})$$

Backward vs Forward Euler Integration

$$\Delta \mathbf{x}_B = \Delta t \dot{\mathbf{x}}^{t+1}$$
$$\Delta \dot{\mathbf{x}}_B = \Delta t \mathbf{M}^{-1} (\mathbf{f}^t + \mathbf{K} \Delta \mathbf{x}_B)$$

$$\Delta \mathbf{x}_F = \Delta t \dot{\mathbf{x}}^t$$
$$\Delta \dot{\mathbf{x}}_F = \Delta t \mathbf{M}^{-1} \mathbf{f}^t$$

$$\Delta \mathbf{x}_B = \Delta t \dot{\mathbf{x}}^t + \Delta t \Delta \dot{\mathbf{x}}_B$$
$$\Delta \dot{\mathbf{x}}_B = \Delta t \mathbf{M}^{-1} \mathbf{f}^t + \Delta t \mathbf{M}^{-1} \mathbf{K} \Delta \mathbf{x}_B$$

Rewrite with deltas

$$\Delta \mathbf{x}_B = \Delta t \dot{\mathbf{x}}^t + \Delta t^2 \mathbf{M}^{-1} \mathbf{f}^t + \Delta t^2 \mathbf{M}^{-1} \mathbf{K} \Delta \mathbf{x}_B$$

Combine

$$(\mathbf{I} - \Delta t^2 \mathbf{M}^{-1} \mathbf{K}) \Delta \mathbf{x}_B = \Delta t \dot{\mathbf{x}}^t + \Delta t^2 \mathbf{M}^{-1} \mathbf{f}^t$$

$$\Delta \mathbf{x}_B = (\mathbf{I} - \Delta t^2 \mathbf{M}^{-1} \mathbf{K})^{-1} (\Delta \mathbf{x}_F + \Delta t^2 \mathbf{M}^{-1} \mathbf{f}^t)$$

Solve

Backward vs Forward Euler Integration

$$\Delta \mathbf{x}_B = \Delta t \dot{\mathbf{x}}^{t+1}$$
$$\Delta \dot{\mathbf{x}}_B = \Delta t \mathbf{M}^{-1} (\mathbf{f}^t + \mathbf{K} \Delta \mathbf{x}_B)$$

$$\Delta \mathbf{x}_F = \Delta t \dot{\mathbf{x}}^t$$
$$\Delta \dot{\mathbf{x}}_F = \Delta t \mathbf{M}^{-1} \mathbf{f}^t$$

$$\Delta \mathbf{x}_B = (\mathbf{I} - \Delta t^2 \mathbf{M}^{-1} \mathbf{K})^{-1} (\Delta \mathbf{x}_F + \Delta t^2 \mathbf{M}^{-1} \mathbf{f}^t)$$

Solve

$$\Delta \mathbf{x}_B = \mathbf{V} (\mathbf{I} - \Delta t^2 \mathbf{\Lambda})^{-1} \mathbf{V}^T (\Delta \mathbf{x}_F + \Delta t^2 \mathbf{f}^t)$$

$$\Delta \mathbf{x}_B = \text{filter}(\Delta \mathbf{x}_F + \Delta t^2 \mathbf{f}^t)$$

Note: \mathbf{K} , \mathbf{V} , $\mathbf{\Lambda}$, and `filter` change over time

Backward vs Forward Euler Integration

$$\Delta \mathbf{x}_F = \Delta t \dot{\mathbf{x}}^t$$

$$\Delta \dot{\mathbf{x}}_F = \Delta t \mathbf{M}^{-1} \mathbf{f}^t$$

$$\Delta \mathbf{x}_B = \mathbf{V}(\mathbf{I} - \Delta t^2 \mathbf{\Lambda})^{-1} \mathbf{V}^T (\Delta \mathbf{x}_F + \Delta t^2 \mathbf{f}^t)$$

$$\Delta \mathbf{x}_B = \text{filter}(\Delta \mathbf{x}_F + \Delta t^2 \mathbf{f}^t)$$

Note: \mathbf{K} , \mathbf{V} , $\mathbf{\Lambda}$, and **filter** change over time

- **PCA vs Eigen System**

- **Dynamic orthogonality**
- **Energy leakage**
- **Limited sample states for PCA**

Acknowledgments

This slide set contain contributions from:

- Kayvon Fatahalian
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