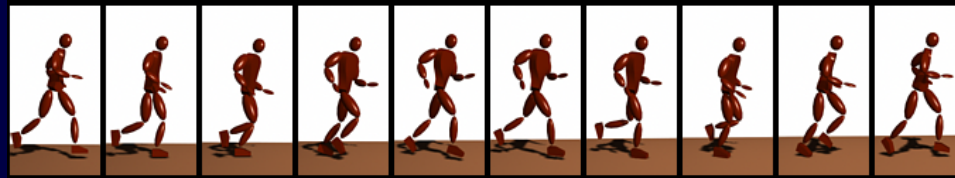
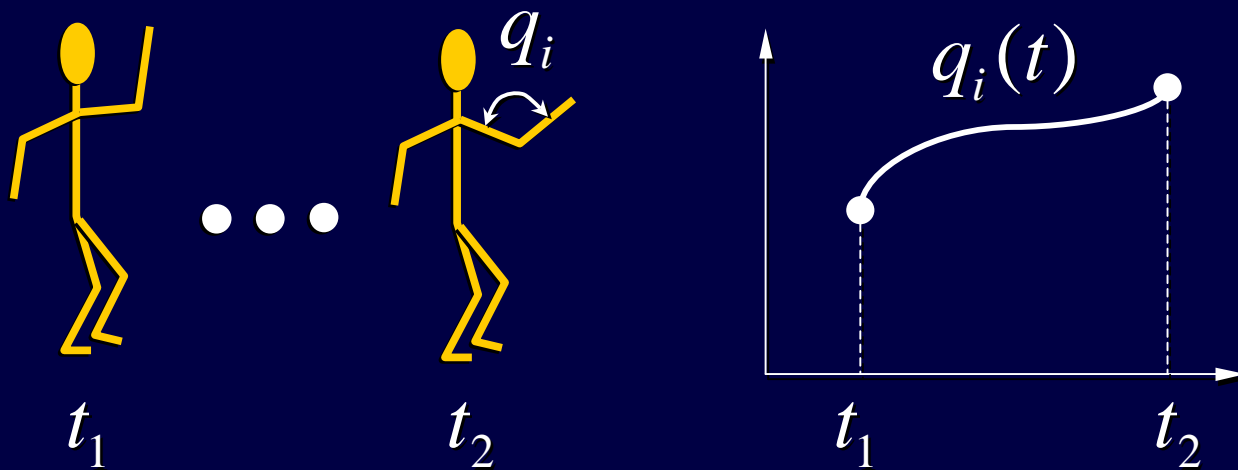


# Physically Based Motion Transformation



# Digital Character Animation

- **Character** – animated object with a number of degrees of freedom (DOFs)
- **Motion** – set of functions  $\mathbf{q}(t)$  depicting how each DOF changes through time



# The Animation Problem

*Automatic generation of expressive/realistic motion that achieves a given set of tasks*

- An open problem
- Realism vs. control tradeoff

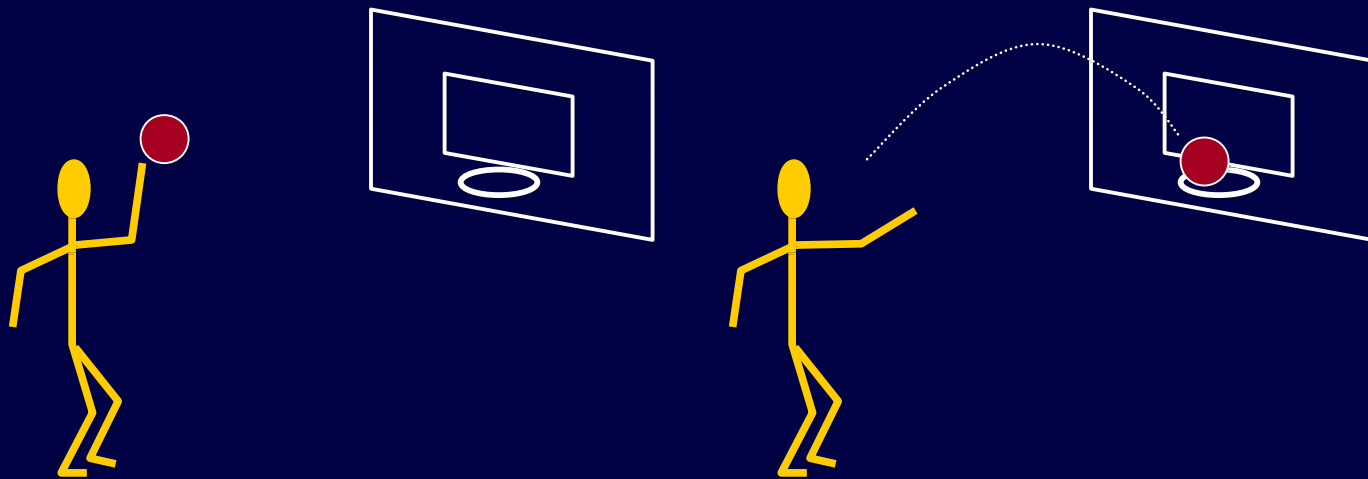


# Physically-based Methods

- Forward simulation [Baraff]
  - Highly realistic
  - Simulated character very hard to control
- Controllers [Raibert, Hodgins, Ngo, van de Pane]
  - Fast motion generation once controllers are computed
  - No set rules on controller generation

# Spacetime Constraints

- Provide both realism and control
- Downside
  - Methods do not scale up
  - Sensitivity to the initial position



# Captured Motion

- Sampled DOFs through time gathered from the real world
- Rich and realistic
- Hard to edit



# Motion Warping

- Set poses which the warped motion should interpolate
- Set time constraints and solve for the minimum curve deviation Interpolate the space constructed by a few sample motion capture sequences



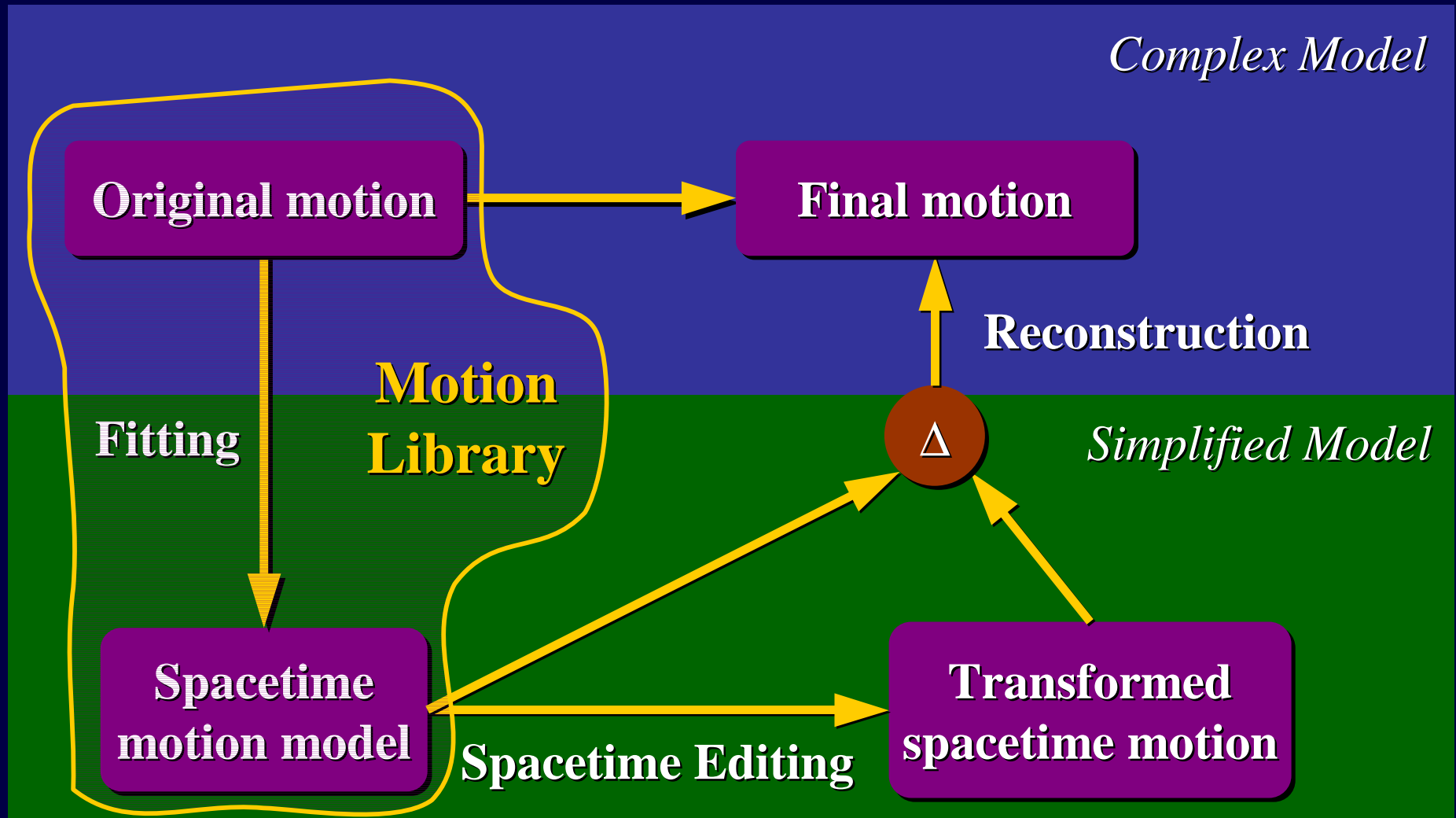
# High Level Control

- Get a limp walk by making one leg stiff
- Reduce gravity to get a “moon walk”
- Change the position and timing of foot placements
- Make a “quiet” run by reducing the floor impact forces

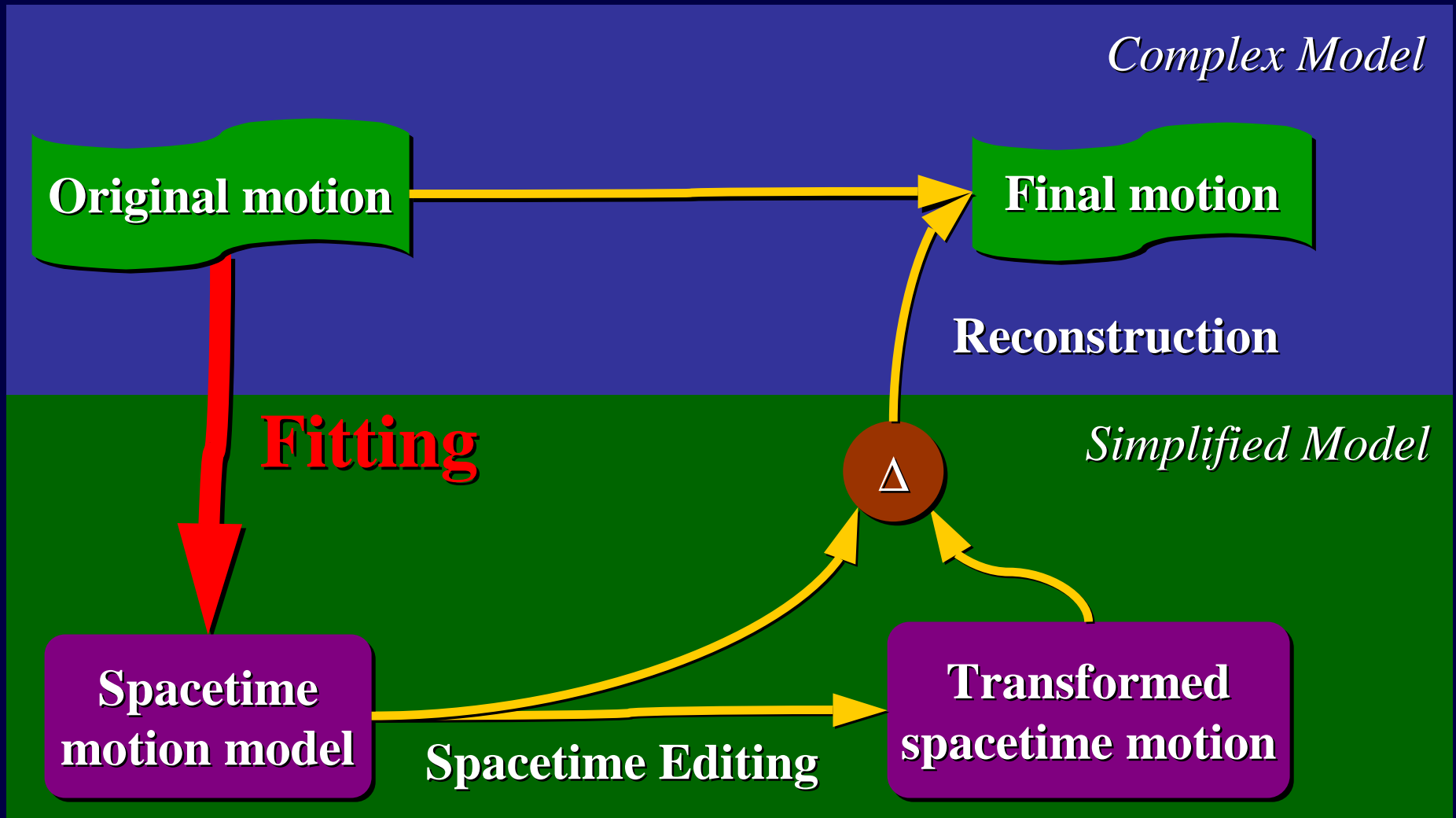
# The New Approach

- Transform existing motion
- Spacetime constraints formulation
- Simplified character representation
- Get the best of both worlds:
  - Expressiveness of captured data
  - Controllability of the spacetime model

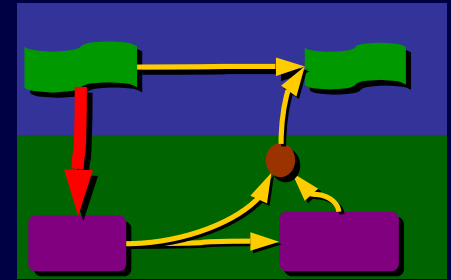
# Outline



# Outline

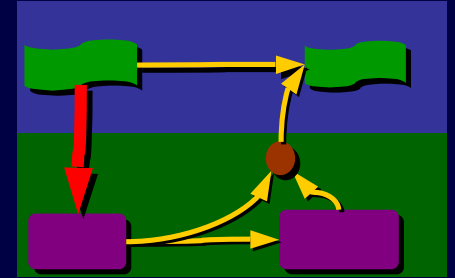


# Model Fitting

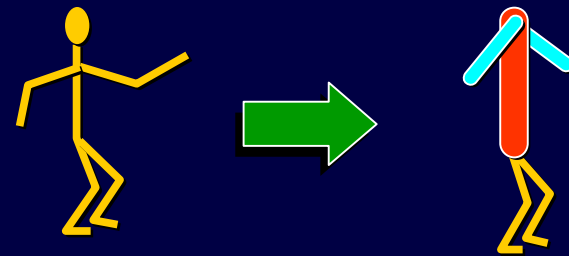


- Two phases:
  - Simplify character kinematics
  - Use input motion to construct a spacetime motion model

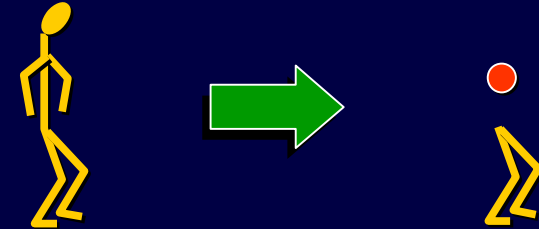
# Simplified Kinematics



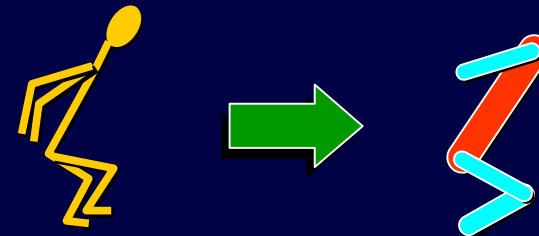
- Remove irrelevant DOFs



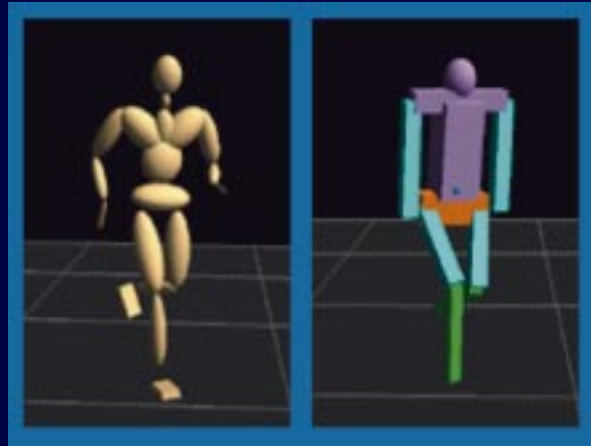
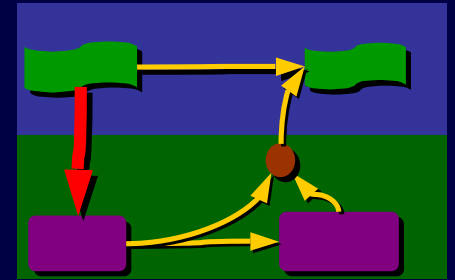
- Reduce *passive* body structure to mass points



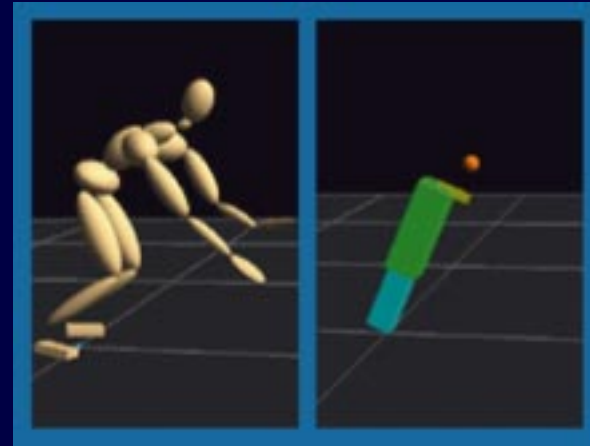
- Exploit symmetric movement of limbs



# Simplified Kinematics

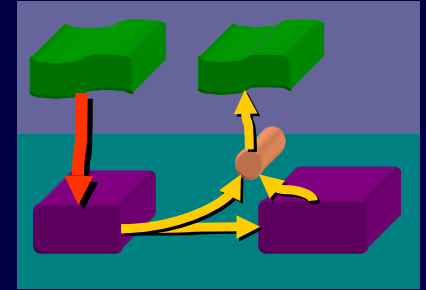


Human Run



Human Jump

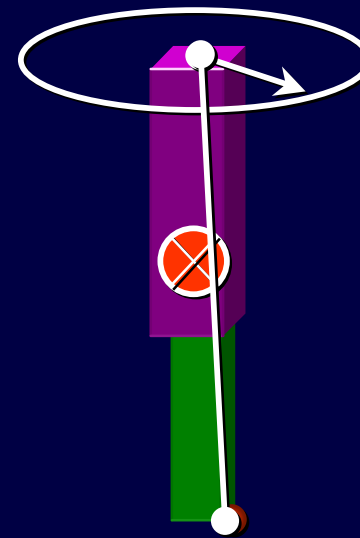
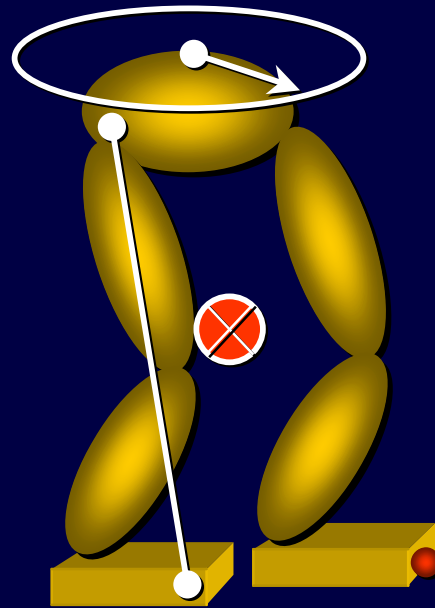
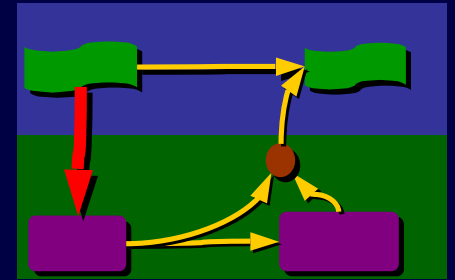
# Motion Fitting



- **Handle** – a property that correlates the original and simplified model
- Must have enough handles to fully determine simplified model configuration

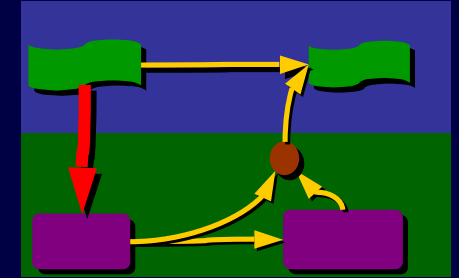


# Handle Examples

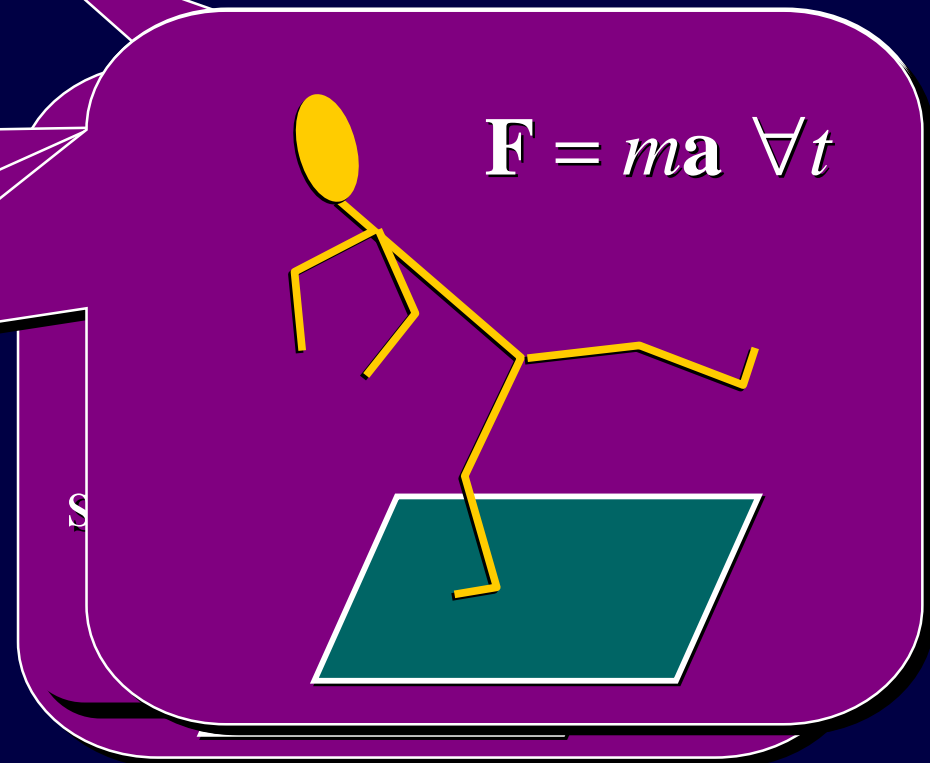


Distance Handle

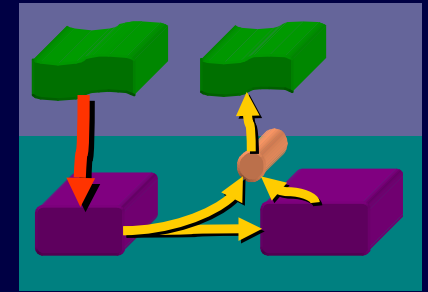
# Motion Synthesis As Constrained Optimization



- Body, muscle and force DOFs:  $\mathbf{q}(t)$
- Constraints:
  - Pose  $\mathbf{C}_p$
  - Mechanical  $\mathbf{C}_m$
  - Dynamics  $\mathbf{C}_d$
- Objective  $E(\mathbf{q}(t))$



# Spacetime Model Fitting

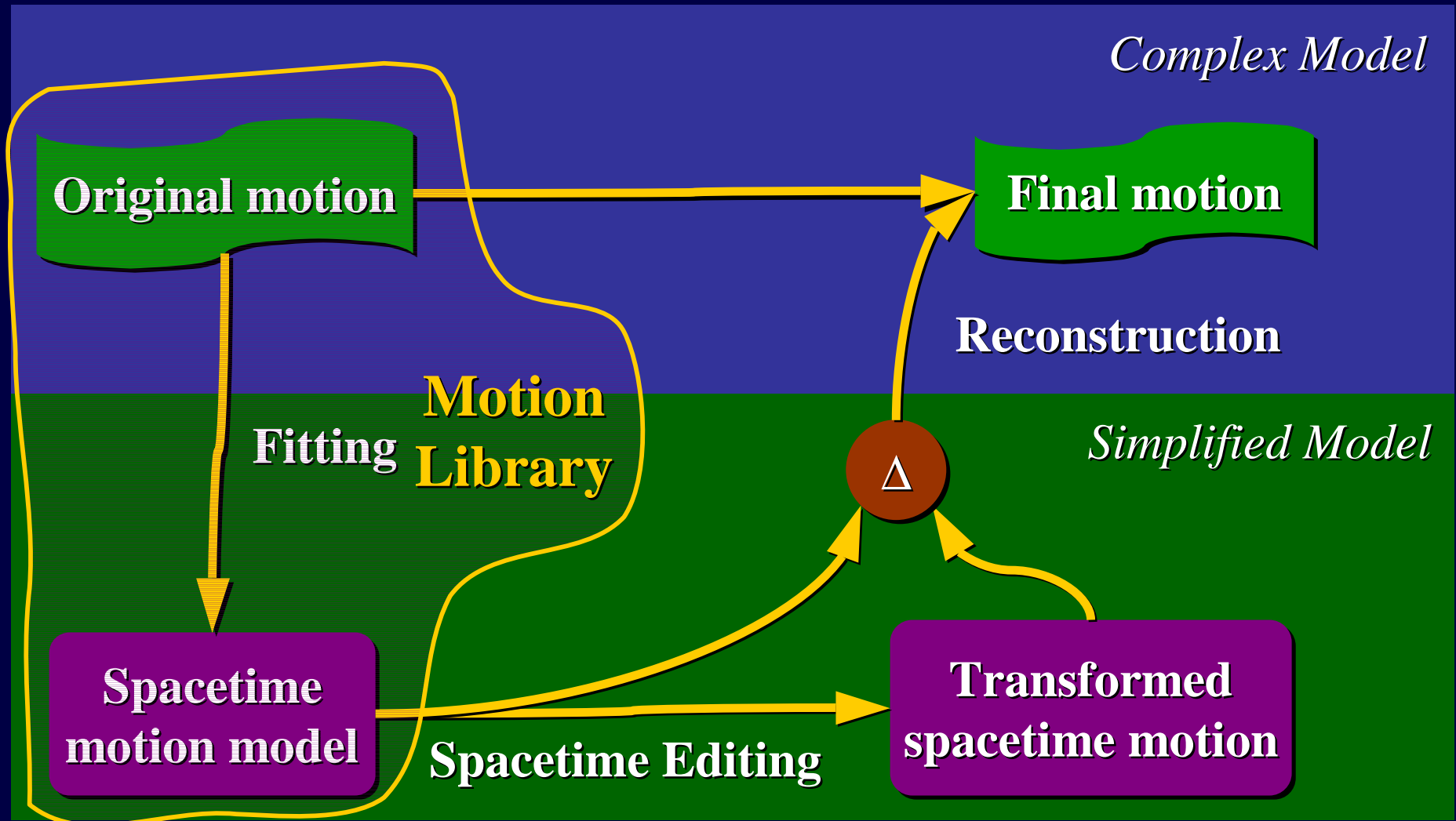


- Biological data: mass distribution, muscles
- Use *handles* to create “best-guess” motion
- Specify constraints essential for given motion (e.g. foot placements)
- Use simple objective: smooth muscles

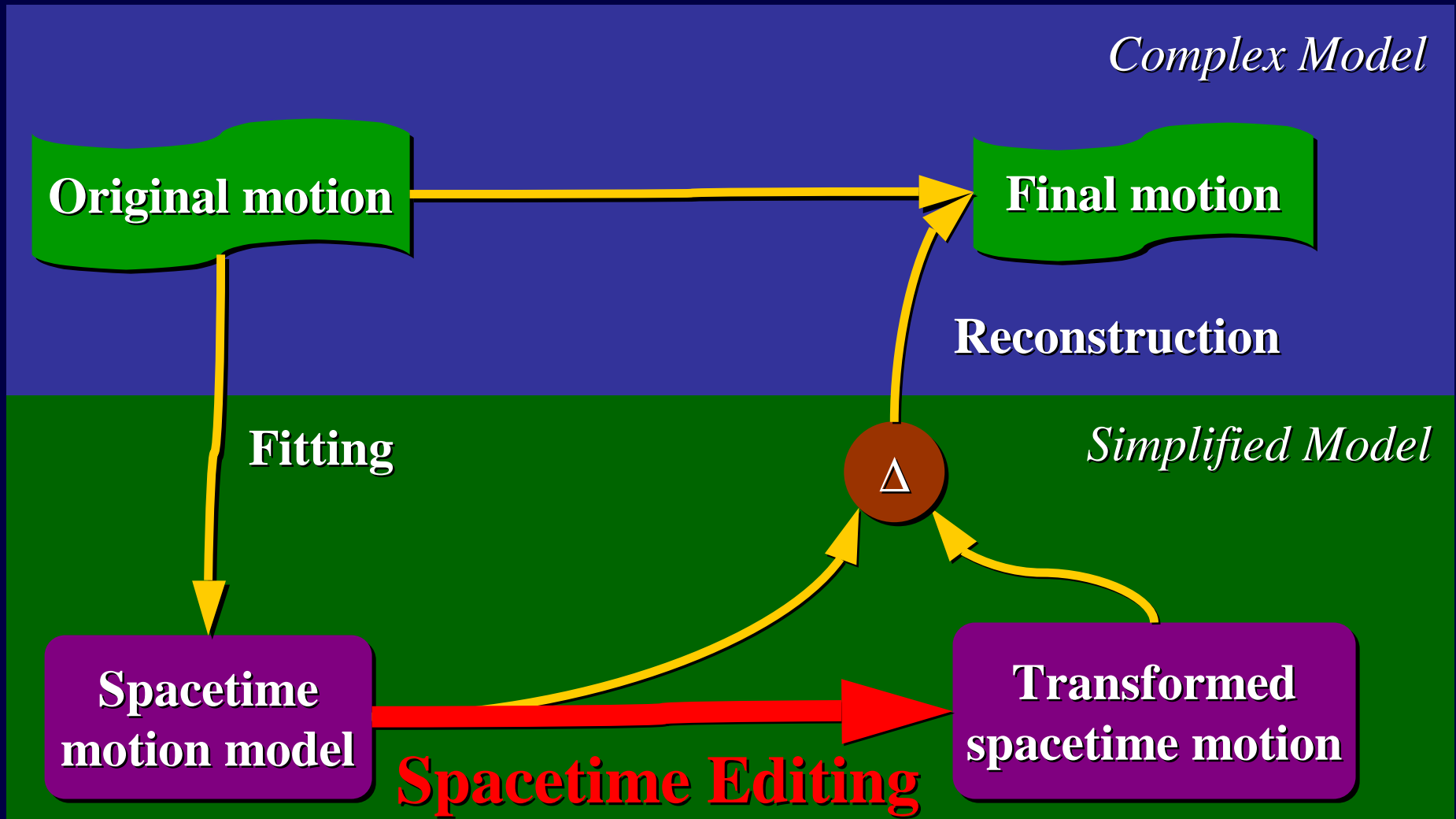
$$E(\mathbf{q}) = \ddot{\mathbf{q}}^2$$



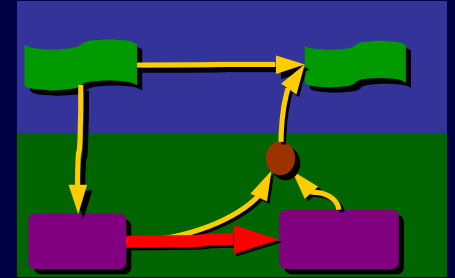
# Outline



# Outline

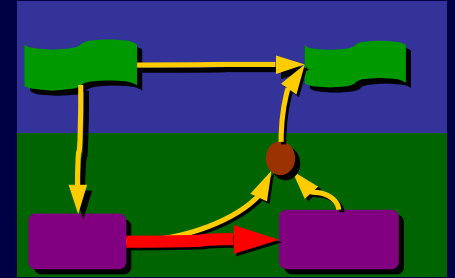


# Spacetime Editing



- Change pose and environment constraints
  - Foot placement and timing
  - Introduce a new obstacle
- Change the objective function
  - Minimize floor impact forces
  - Make dynamic balance more important

# Spacetime Editing

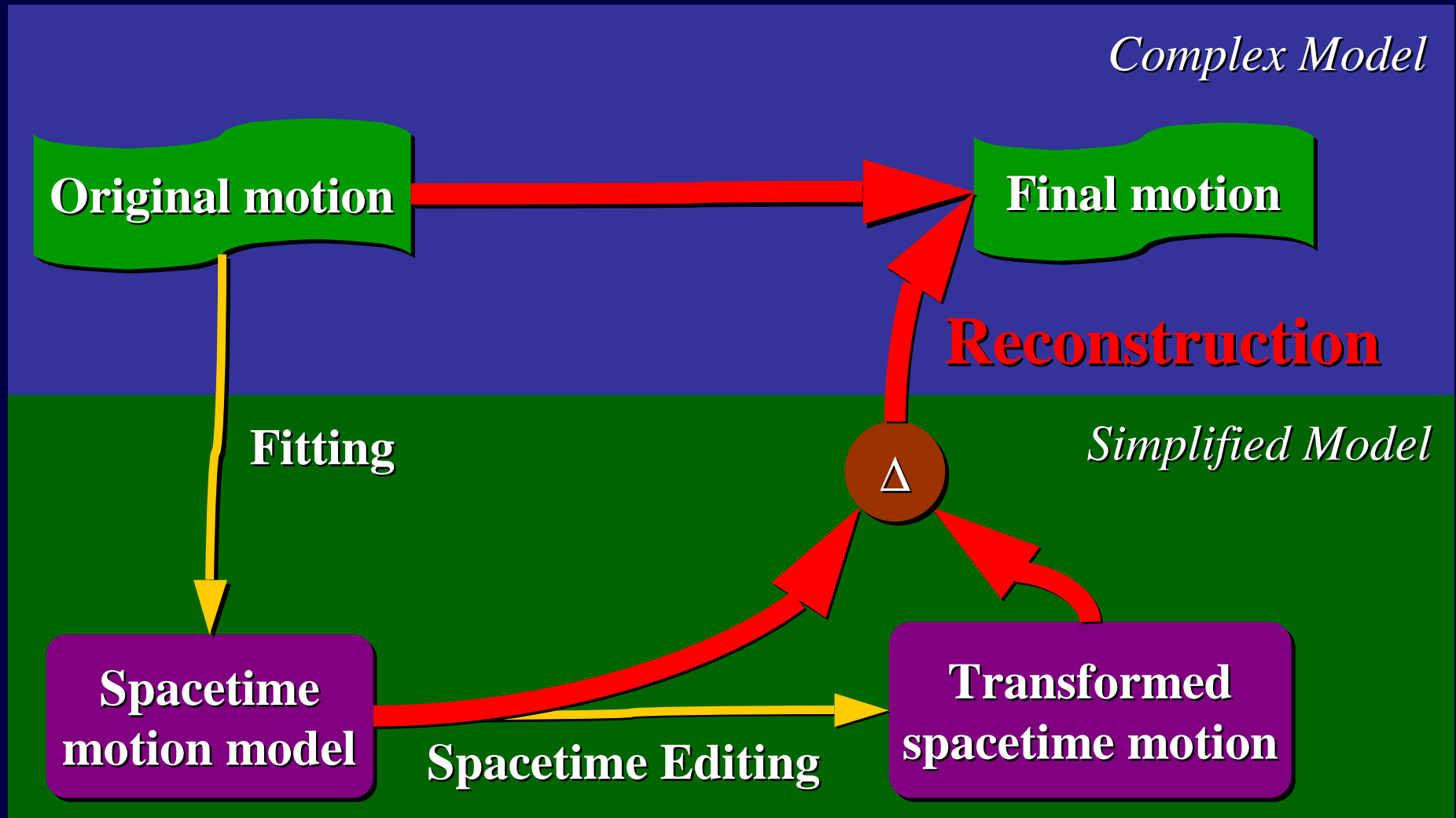


- Change explicit character parameters
  - Short leg
  - Redistribute mass
  - Modify muscle characteristic
  - Gravity

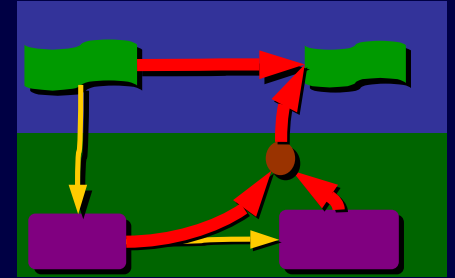




# Outline



# Motion Reconstruction

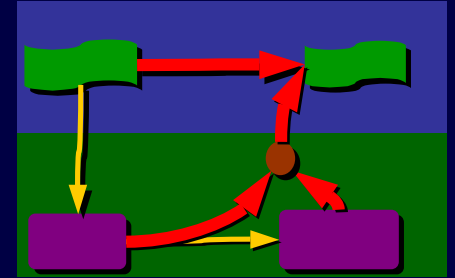


- Three different handle sets
  - Original motion handles  $\mathbf{h}(\mathbf{q}_o)$
  - Spacetime fit handles  $\mathbf{h}(\mathbf{q}_s)$
  - Transformed spacetime handles  $\mathbf{h}(\mathbf{q}_t)$
- Compute final motion handles

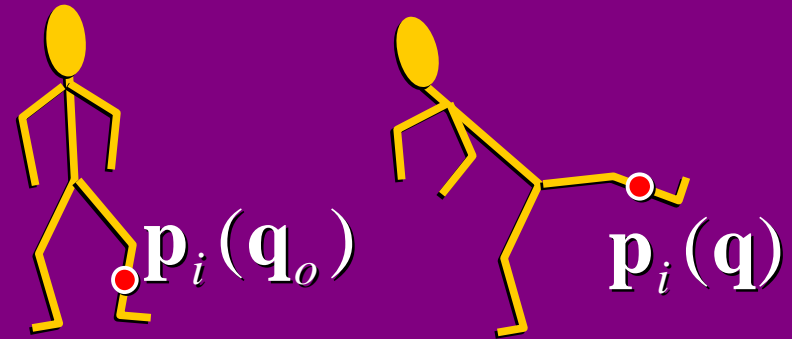
$$\mathbf{h}(\mathbf{q}_f) = \mathbf{h}(\mathbf{q}_o) + (\mathbf{h}(\mathbf{q}_t) - \mathbf{h}(\mathbf{q}_s))$$



# Minimum Displaced Mass Objective

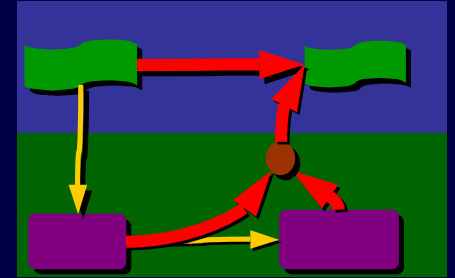


- $E_{dm}(\mathbf{q}_o, \mathbf{q})$  evaluates *total displaced mass* when moving a character from pose  $\mathbf{q}_o$  to pose  $\mathbf{q}$



$$E_{dm} = \iiint_i \mu_i (\mathbf{p}_i(\mathbf{q}_o) - \mathbf{p}_i(\mathbf{q}))^2 dx dy dz$$

# Reconstruction Algorithm



- For each time  $t$  solve

$$\begin{array}{ll} \underset{\mathbf{q}_f}{\text{minimize}} & E_{dm}(\mathbf{q}_o, \mathbf{q}_f) \\ \text{subject to} & \mathbf{h}(\mathbf{q}_f) = \mathbf{h}(\mathbf{q}_o) + (\mathbf{h}(\mathbf{q}_t) - \mathbf{h}(\mathbf{q}_s)) \end{array}$$



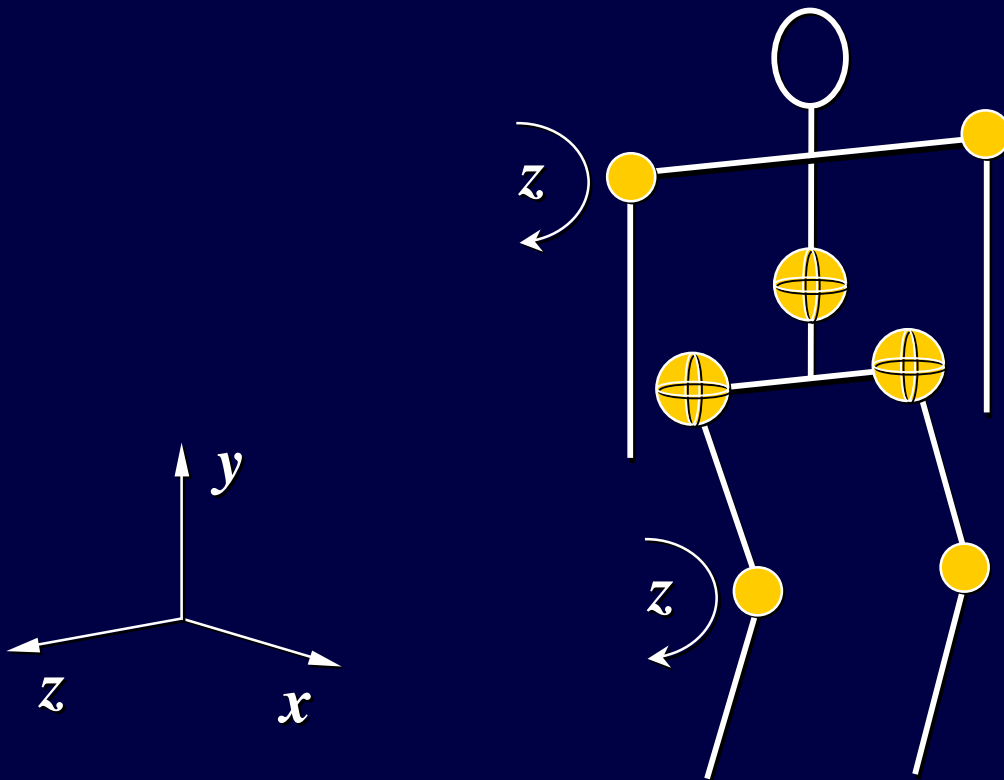


# Example: Human Run

- Original model has 59 DOFs
- Simplified model has **19** DOFs
- Optimizations are done on one gait cycle
- Each optimization completes within 2 minutes



# Biped



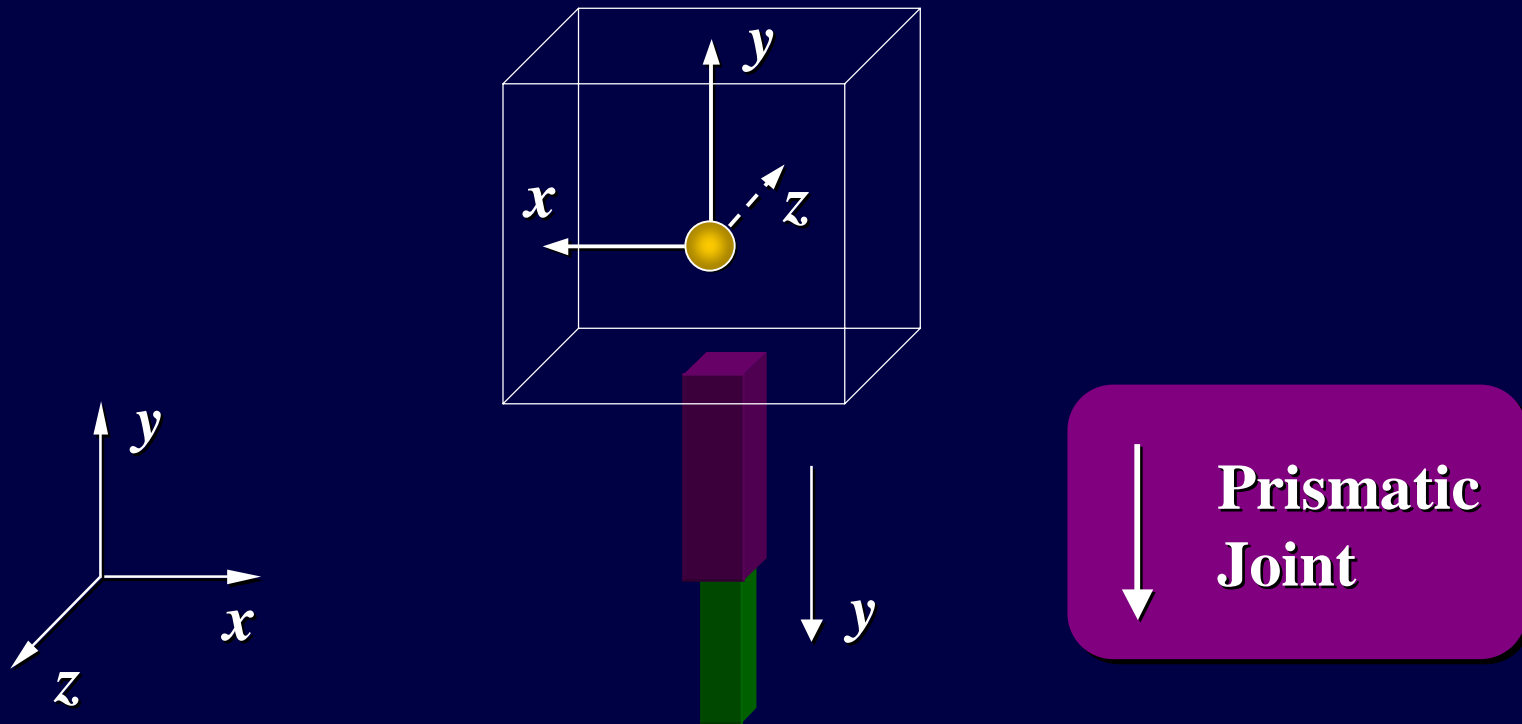
- Hinge Joint
- ⊕ Ball Joint



# Example: Human Broad Jump

- Original model has 59 DOFs
- Simplified model has **11** DOFs
- Entire upper body reduced to a mass point
- No joint angle DOFs

# Hopper





# Future Work

- Optimal robots
- Extracting style
- Motion retargeting
- Building motion libraries
- Digital actors