

## Robotics and Animatronics in Disney

Lecture 5: Adapting Human Motion Data to Different Kinematics



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## Goals

- Introduce a technique for mapping human motions to non-humanoid characters
- Introduce a technique for adapting human motions to different environmental constraints
- Discuss what features must be preserved to maintain naturalness



## Extending the Horizon

- Subject and environment specific
  - Characters/robots with different mass properties?
- **Mostly human subjects**
  - **Non-humanoid characters?**
- **Specific to environment**
  - **Different environment?**



## Animating Non-Humanoid Characters



[Yamane, Ariki, Hodgins 2010]

## Animating Non-Humanoid Characters

Non-humanoid characters



Optimization: can't incorporate styles  
Keyframing: rely on animator's talent



[Witkin and Kass 1988]



## Our Approach

Use human motion capture data: leverage **actor's** talent

- Perform natural and expressive motions
- Imagine and imitate **character's motion style**

Challenge: mapping human motion to characters

- Nonlinear
- Small training data set



### Example: Happy Lamp

human motion in the character's style

output of the system

### Related Work

#### Motion retargeting

- Assume same topology [Choi and Ko 2000] [Mozani et al. 2000] [Shin et al. 2001] [Tak and Ko 2005]

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### Related Work

#### Motion retargeting for different topologies

- Soda can example in [Gleicher 1998]: explicitly specify corresponding body parts
- Linear blending [Park and Shin 2004]
- Retargeting to user-created models [Hecker et al. 2008]: requires extensive annotation on the motions
- Retargeting mesh deformations [Baran et al. 2009]

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### Related Work

#### Learning from artists' input

- Style transfer [Ikemoto et al. 2009]: requires example animations for the new character
- Transfer cartoon styles [Bregler et al. 2002]: primarily for 2D transformations

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### Pipeline Overview

Motion capture (2 hours)

Actor selects key poses (18 hours)

Actor creates character's key poses

Static mapping (70 minutes)

Dynamics optimization (Time required to create 7 minutes of animations)

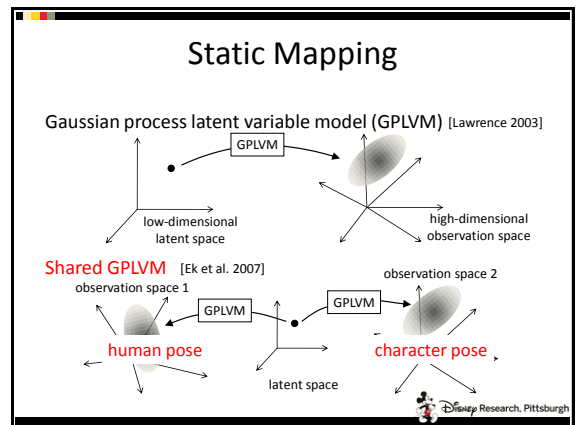
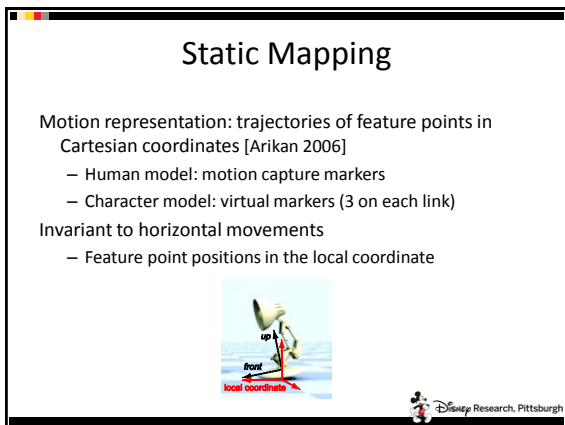
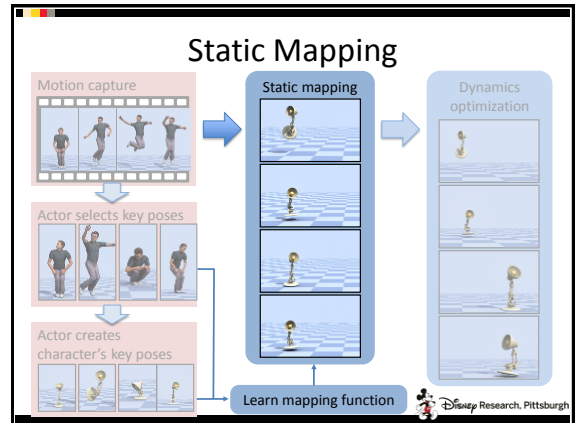
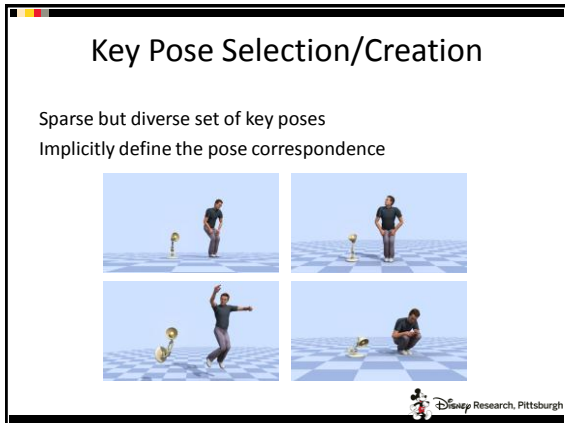
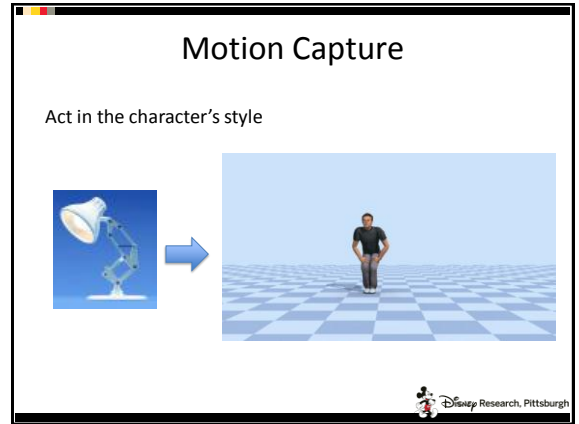
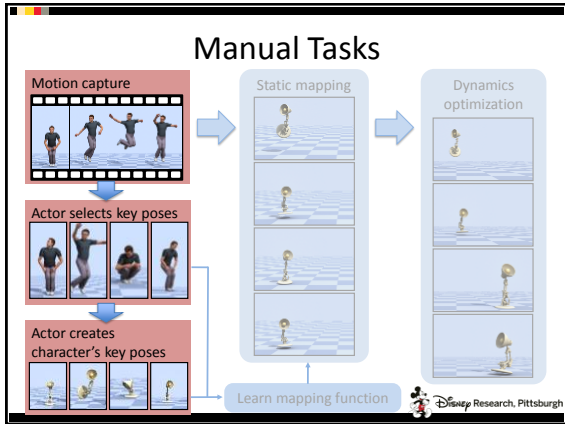
Learn mapping function

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### Features

- Sparse key poses (average interval 4.0s)
  - Faster than complete keyframing
- Leverage actor's talent
  - Recorded motion: style, naturalness and expressiveness
  - Key poses: body correspondence

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## Why Shared GPLVM?

### Gaussian process

Sparse training data set

### Latent variables

Human motions are confined to a low-dimensional space  
[Safonova et al. 2004]

### Shared

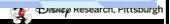
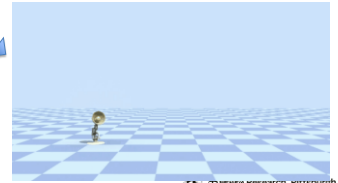
Common structure in human and character motions



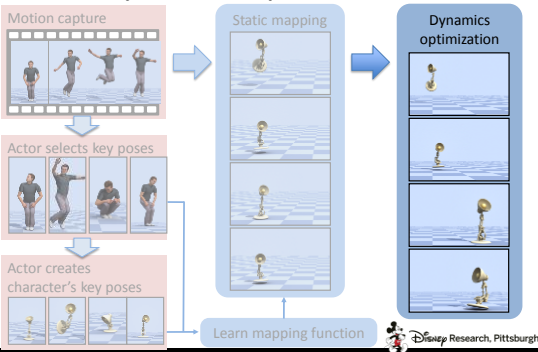
## Static Mapping



Direct mapping using  
shared GPLVM



## Dynamics Optimization



## Dynamics Optimization

Horizontal movement

Contacts

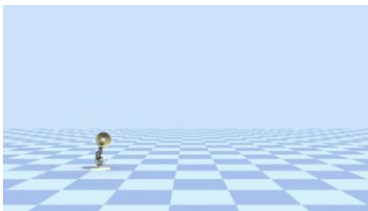
Physical plausibility



## Horizontal Movement

removed in static mapping

match linear and angular momenta to human motion



## Contacts / Physical Plausibility

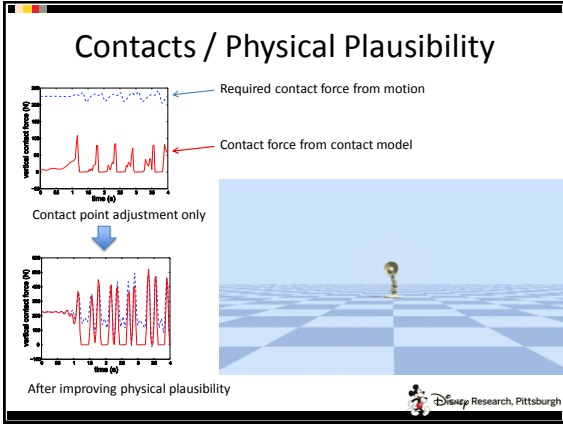
Contacts

- Correct contact points based on contact information of the original human motion and recalculate IK

Physical plausibility

- Modify joint trajectories so that required and actual contact forces match
- Penalty-based model for contact force computation
- Typically vertical translation only





### Results: Characters

Lamp      Penguin      Squirrel

- Topology
- Hop instead of walk
- Short limbs
- Locomotion style
- Relatively human-like
- Walk on four legs

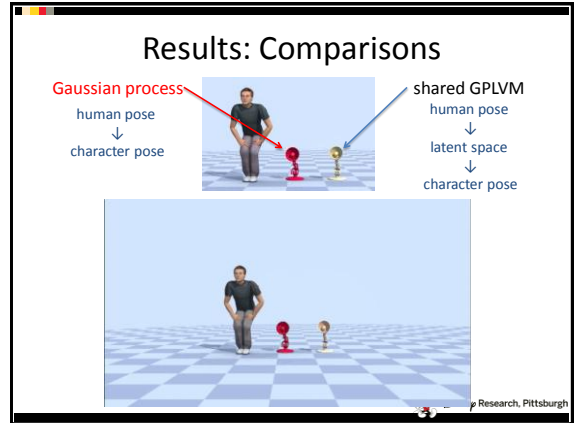
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### Results: Data

motion	lamp		penguin		squirrel	
	length	key poses	length	key poses	length	key poses
anger	18.6	16	14.3	3	22.9	14
disgust	34.4	1	23.3	7	20.0	3
fear	29.7	2	25.2	4	28.5	18
happiness	20.1	7	23.7	11	25.8	9
sadness	19.5	3	29.6	4	26.7	3
surprise	10.7	1	26.1	4	19.2	5
dance	14.7	0	25.6	0	26.1	6
total	148	30	168	33	169	58

Motion length (s)/number of key poses

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## Contributions

New approach that leverages actor's talent

- Natural and expressive motions in the character's style
- Human-character pose correspondence
- Faster than keyframing

Techniques

- Shared GPLVM for mapping nonlinear pose spaces
- Physics-based optimization for fixing artifacts



## Adapting to New Environments



[Yamane, Kuffner, Hodgins 2004]

## Constrained Motions

- Manipulating an object
- Opening a door / drawer
- Cooperative lifting
- Difficulty in synthesis
  - Hard constraints (hands, feet)
  - Object / body / environment collisions
  - Naturalness / style



## Motivation

	Model-driven Simulate kinematics / dynamics model	Data-driven Connect postures from a database
Good	Physically correct Easy to include constraints	Natural
Bad	Difficult to obtain natural motions	Difficult to handle new scenario

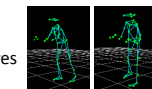
Insight: combine the two approaches



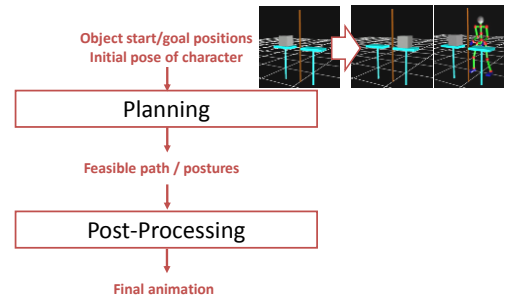
## Our Approach

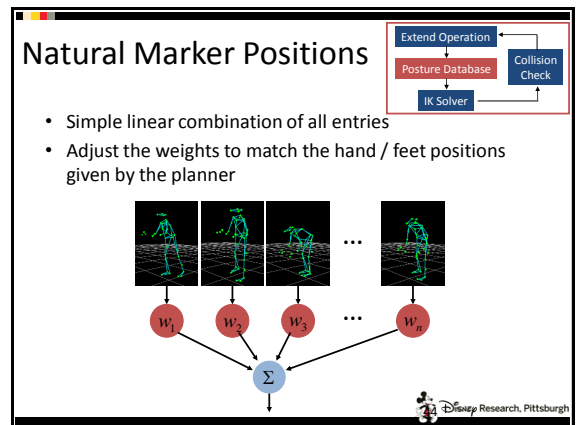
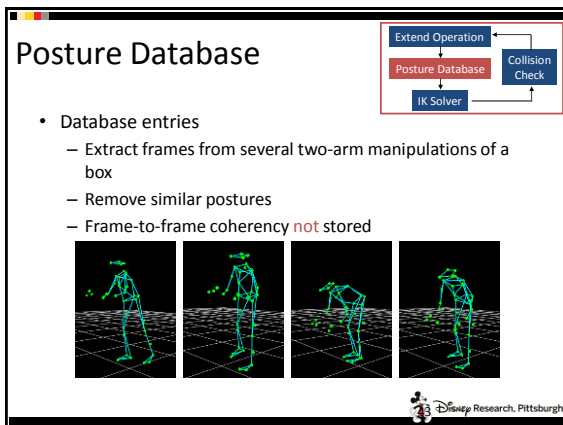
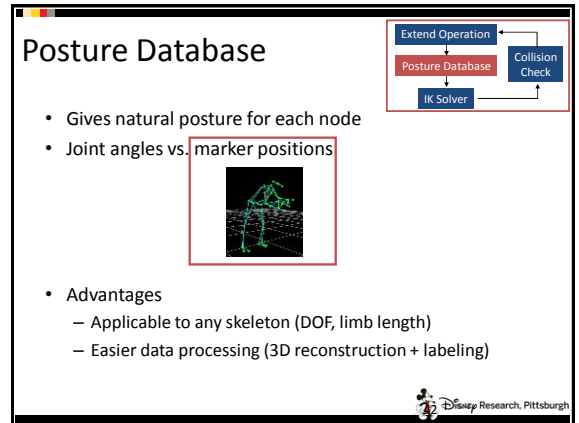
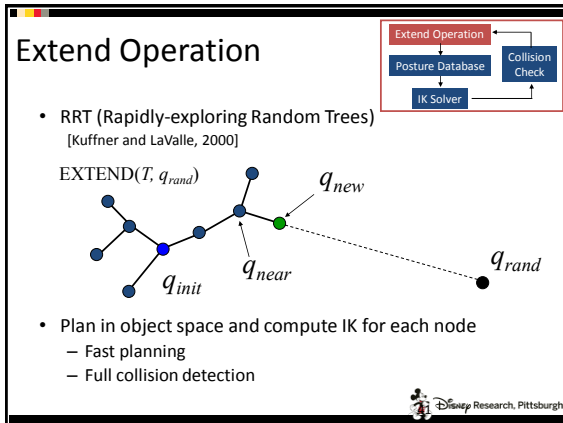
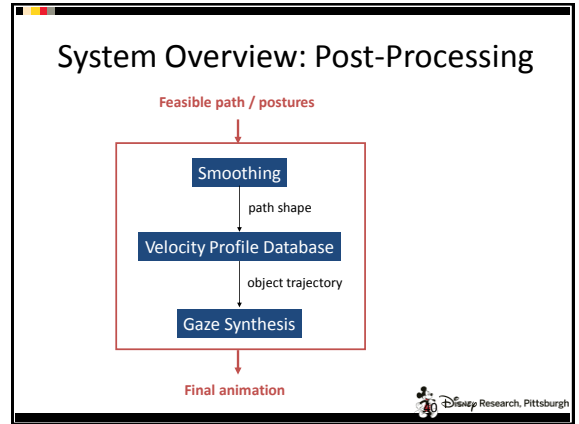
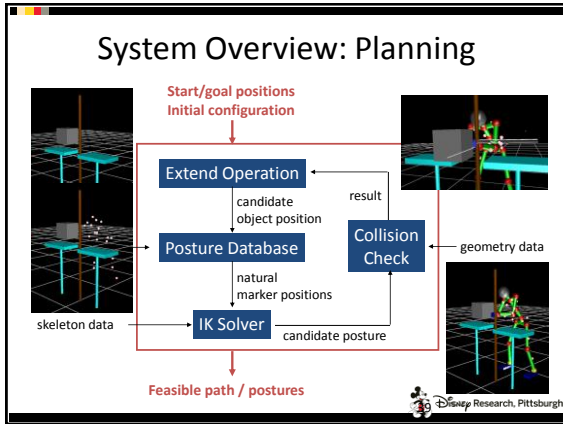
Focus on constrained, collision-free motions

- Use model for flexibility
  - IK: constraints, balance, different characters
  - Planning: environment, different geometry
- Use data for naturalness / style
  - Bias the IK solution toward natural postures
  - Different databases for different styles



## System Overview





## Computing the Weights

- Linear equation
- Pseudoinverse can be precomputed

$$\begin{matrix} \text{constraints} \downarrow & \xrightarrow{\text{database entries}} & & \\ \begin{pmatrix} c_{lhand}^1 & c_{lhand}^2 & \dots & c_{lhand}^{nf} \\ c_{rhand}^1 & c_{rhand}^2 & \dots & c_{rhand}^{nf} \\ c_{lfoot}^1 & c_{lfoot}^2 & \dots & c_{lfoot}^{nf} \\ c_{rfoot}^1 & c_{rfoot}^2 & \dots & c_{rfoot}^{nf} \end{pmatrix} & \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_{nf} \end{pmatrix} & = & \begin{pmatrix} p_{lhand} \\ p_{rhand} \\ p_{lfoot} \\ p_{rfoot} \end{pmatrix} \end{matrix}$$

Hand/foot positions in the database      Weights      Hand/foot positions specified by the planner

## IK Solver

- UTPposer
  - Flexible inverse kinematics (IK) [Yamane and Nakamura 2003]
  - Intuitive pin-and-drag interface
- Two priority levels
  - Higher: dragged link
  - Lower: pinned links
- Problem: redundancy
  - Selects one numerically stable solution for IK problem
  - Not always natural

## IK Solver

- Use the natural marker positions as soft constraints to bias the IK solution

$$\text{Minimize } \sum \|p_m - p_m^{ref}\|^2 + \|p_{com} - p_{com}^{ref}\|^2 \text{ subject to } \sum \|p_h - p_h^{ref}\|^2 = 0$$

- Simple interpolation of database entries is enough because
  - IK takes care of the hard constraints
  - Database entries and constraints are of the same dimension
- Generalizes a database for wider range of tasks, character, environment

## Collision Check

- External collision detection library
  - ColDet (<http://photoneffect.com/coldet/>)

## Details: Post-Processing

Feasible path / postures

↓

Smoothing

↓ path shape

Velocity Profile Database

↓ object trajectory

Gaze Synthesis

Final animation

## Smoothing

Optimize the path

- Select two arbitrary nodes on the planned path
- Connect the two nodes
- Solve IK on the new path and check collisions



## Velocity Profile

Observation: similar paths have similar velocity profiles  
[Atkeson and Hollerbach 1985]

no obstacle

one obstacle

two obstacles

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## Gaze Synthesis

- IK with gaze direction constraints
- Target direction given by the user  
e.g. object goal position

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## Results

Posture Database

- Posture database
  - 247 frames from four motions
- Velocity profile database
  - 21 sets of path + velocity profiles
- Total time
  - Initial setup < 10 min. (in 3DCG software package)
  - Planning: 1~5 min. (depends on complexity of environment)
  - Post-processing: ~1 min.
- cf. 1-2 hours of keyframing in Maya by a trained user

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## Results

- Character
  - adult
  - child
  - long-legged clown
  - long-armed gorilla
- Environment
  - shelf
  - closet
  - car trunk
- Task
  - cooperative manipulation
  - opening a desk drawer
  - opening a cabinet
  - one-handed manipulation

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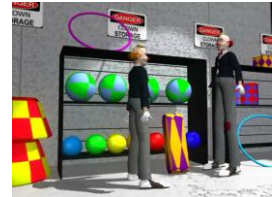
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(with specialized database)





### Results

Different styles for lifting with two different databases





bend back



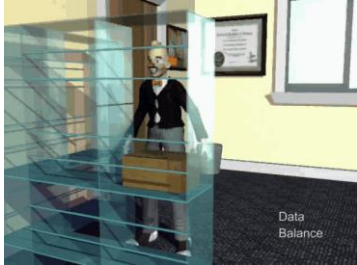


bend knee







### Evaluation: with/without Data




with data



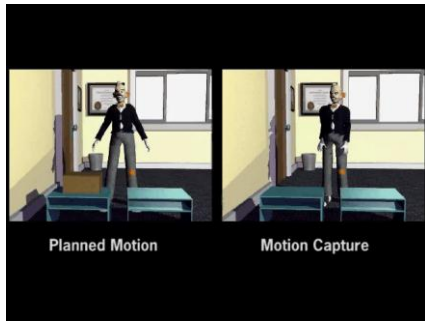
### Evaluation: with/without Data



without data



## Comparison with Motion Capture



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## Conclusion

- Combination of model- and data-driven approaches for synthesizing constrained motions
  - With emphasis on the model (IK / planning)
- Model generalizes a single database to various
  - Characters
  - Environments
  - tasks
- Data supports the model by adding
  - Naturalness
  - Style

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## Discussions

- Significantly different character / environment
  - Large change required
- How much change is acceptable to
  - Maintain the style
  - Look natural

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