

## Animation by Example



Michael Gleicher  
and the UW Graphics Group  
University of Wisconsin- Madison  
[www.cs.wisc.edu/~gleicher](http://www.cs.wisc.edu/~gleicher)  
[www.cs.wisc.edu/graphics](http://www.cs.wisc.edu/graphics)

## The Challenge

- High Quality, Expressive Motion
  - Need motion capture (examples)
- Flexible, long-running, controllable
  - Need synthesis
- Synthesis from Examples!



## Survey of Techniques

### Flexibility:

- Link motions to make sequences
- Blend motions to gain control

### Use Databases of Examples:

- Find related motions in databases
- Combine data for interactive systems



## Survey of Projects

- Motion Graphs
  - Link motions to make long sequences
- Snap Together Motion
  - Synthesis for interactive systems
- Match Webs
  - Find related motions in a database
- Registration Curves + Parametric Families
  - Combine motions to make spaces
- Plus some others...



Work with Lucas Kovar, Hyun Joon Shin, ...

## Idea: Put Clips Together

- New motions from pieces of old ones!
- Good news:
  - Keeps the qualities of the original (with care)
  - Can create long and novel "streams" (keep putting clips together)
- Challenges:
  - How to connect clips?
  - How to decide what clips to connect?



## Connecting Clips Transition Generation

- Transitions between motions can be hard
- Simple method work *sometimes*
  - Blends between aligned motions
  - Cleanup footskate artifacts
- Just need to know when is "sometime"



## What is Similar?

- Factor out invariances and measure

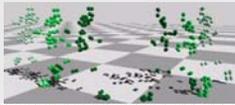
1) Initial frames



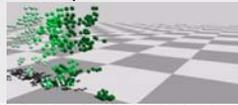
2) Extract windows



3) Convert to point clouds

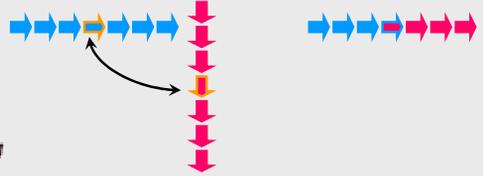


4) Align point clouds and sum squared distances



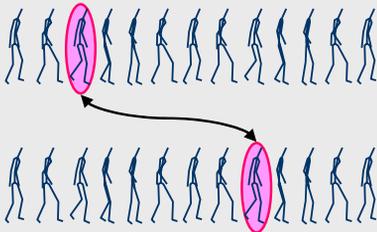
## An easy point to miss: Motions are Made Similar

- "Undo" the differences from invariances when assembling
- Rigidly transform motions to connect



## Building a Motion Graph

- Find Matching States in Motions

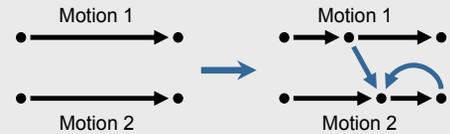


## Motion Graphs

Kovar, Gleicher, Pighin '02

Start with a database of motions, each with type and constraint information.

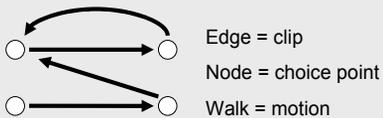
Goal: add transitions at opportune points.



Other Motion Graph-like projects elsewhere  
Differ in details, and attention to detail

## Motion Graphs

Idea: automatically add transitions within a motion database

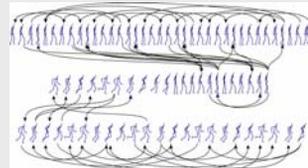


Quality: restrict transitions

Control: build walks that meet constraints

## Automatic Graph Construction

- Find many matches (opportunistic)



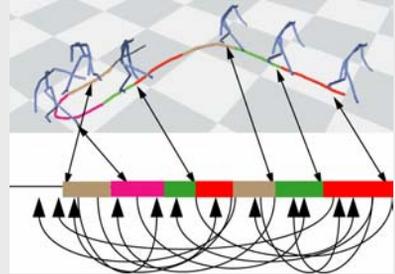
- Good: Automatic
- Good: Lots of choices

## Using a motion graph

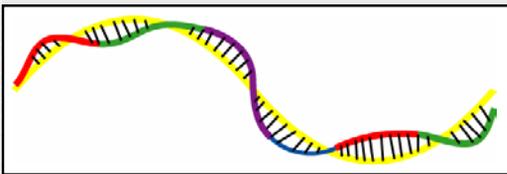
- Any walk on the graph is a valid motion
- Generate walks to meet goals
  - Random walks (screen savers)
  - Search to meet constraints
- Other Motion Graph like projects elsewhere
  - Differ in details, and attention to detail



## An example: Building a Motion Graph



## An example: Using a Motion Graph



- Given a path
- Find a motion that minimizes distance
- Combinatorial optimization

Video:  
mographs.avi



## Why is this good?

- Search the graphs for motions
- Look ahead avoids getting stuck
- Cleanup motions as generated
- Plan "around" missing transitions
- Optimization gets close as possible

Not OK for Interactive Apps!

Need different tradeoffs



## What about interactive?

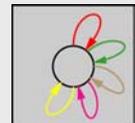
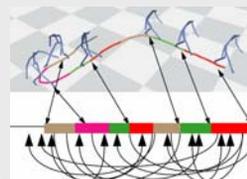
- Different set of tradeoffs!
- Runtime must be:
  - Responsive
  - Low overhead
- Willing to sacrifice quality to get



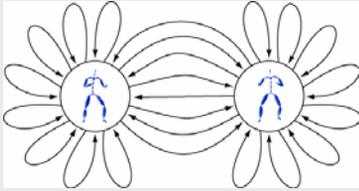
## Contrived Graph Structure?

Search: Look ahead to get where you need to go

React: Always lots of choices. Something close to need.

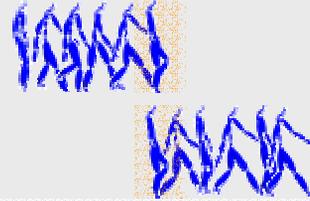


## Gamers use these



## Snapable Motions

- What if motions matched exactly?
  - Match both state and derivatives
  - Match reasonably at a larger scale



## Make motions match exactly

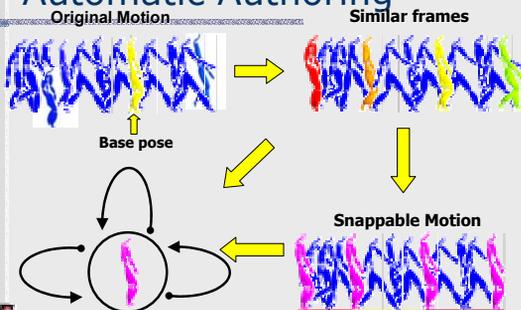
- Add in displacement maps
- Bumps we add to motions
- Modify motions to common pose
- Compute changes at author time



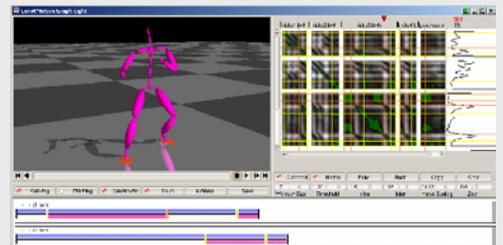
## Semi-Automatic Graph Construction

- Pick set of *match frames*
  - User selects
  - System picks "best" one
- Modify motions to build hub node
- Check graph and transitions

## Automatic Authoring

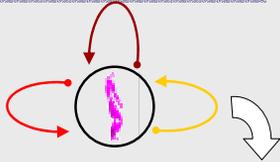


## Building the Motion Graphs



Video:  
stm-intro

## Runtime

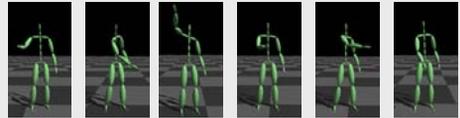


Synthesized Motion

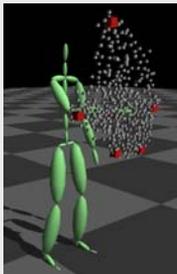


## Limitations of Motion Graphs

- Graphs provide discrete choices
- Use pieces of the database
- Can't capture ALL examples
- Synthesize new motions between example by blending

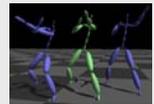


## Motions Between examples

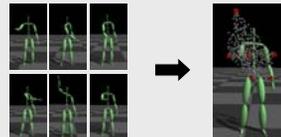


## Parameterized Motions

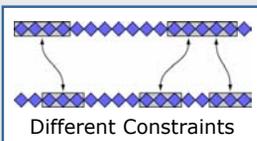
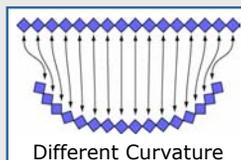
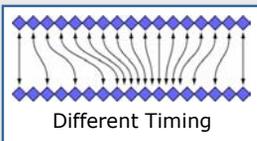
Blend captured motions to make new ones.



Create a *natural* parameterization for intuitive access to these new motions.



## Blending requires similar motions



video:  
regCurves

## Registration Curves

- Encode the relationships between *similar* motions

(video of pair blending apps)

If we have a big database...

- How do we find similar motions?
- How do we use several examples?

## Adapting to Large Data Sets

Previous: small, "contrived" data sets (e.g., Rose et al. '98, '02).

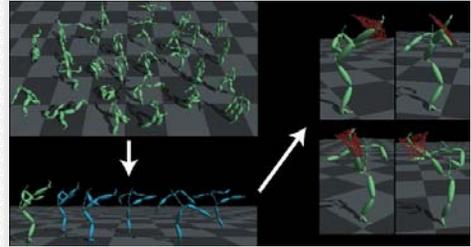
(Kovar and Gleicher '04): Adapt parameterized motions to large data sets

- Automatically find and extract examples
- Automated blending (K&G '03)
- Accurate and stable parameterization

Input: database + one example + parameterization function



## Motion Families



Database- Controllable Clip



## Motion Families

- Match Webs
  - Search for similar motions
- Registration
  - Align motions for blending
- Parameterization
  - Define useful controls
- Sampling
  - Improve nearest neighbor interpolation



## Finding Motions

Example motions are buried in longer motions.



Strategy: search for motion segments similar to a query.



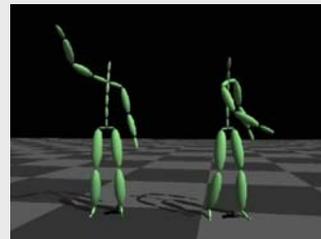
## Why It Is Hard to Find Motions

- Motions can be different lengths.
  - reach middle |-----| reach high |-----|
- Complicated distance metrics
- Logically similar ≠ numerically similar.

$$D(F, F') = \min_{\tau} \sum_i |p_i - T(\theta_i, x_i, y_i)|$$

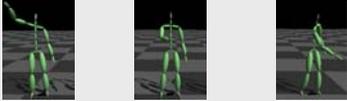


## Similar?



## Search Strategy

Find "close" matches and use as new queries.



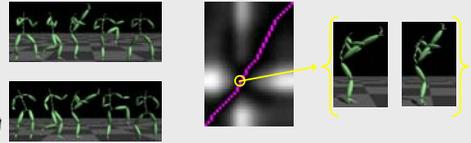
One search may involve many queries.

Precompute potential matches for interactivity.

## Computing Distance Between Motions

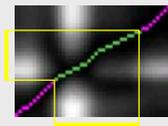
Distance between corresponding frames (in the best time warp)

- Factors out timing differences
- Allows arbitrary distance metrics for frames

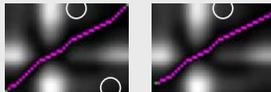


## Precomputing Matches: Insights

Any subset of an optimal path is optimal.

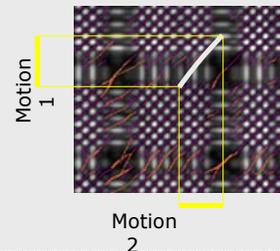


Optimal paths are redundant under endpoint perturbation.



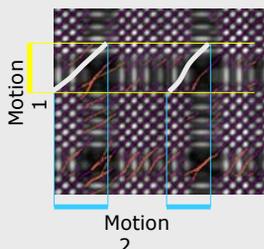
## Precomputing Matches: Match Webs

Compute a grid of distances between pairs of frames and find long, locally optimal paths.



## Precomputing Matches: Methods

At run time, intersect queries with the match web to find matches.



## Search Results

- 37,000 frame data set with ten different kinds of motions.
- 50 minutes to compute match web
- 21MB on disk
- All searches (up to 97 matches) in  $\leq 0.5s$
- Manual verification of accuracy

## Natural Parameterizations

Blend weights offer poor controls

We need more natural parameters.

$$g(\mathbf{M}) = \mathbf{p}$$

↑ motion     ↑ parameters

reaching	hand position at apex
turning	change in hip orientation
jumping	max height of center of mass



## From Parameters to Blend Weights

It is easy to map blend weights to parameters.

$$f(\mathbf{w}) = g(w_1 \mathbf{M}_1 \oplus \dots \oplus w_n \mathbf{M}_n) = \mathbf{p}$$

↑ blend weights     ⊕ blend     ↑ parameters

But we want  $\mathbf{w} = f^{-1}(\mathbf{p})$  !

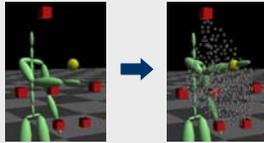
This has no closed form solution!



## Building Parameterizations

Given samples  $(\mathbf{p}, \mathbf{w})$ , we can approximate  $f^{-1}$  with  $k$  nearest neighbor interpolation.

Accuracy:  
create new blends to get additional



Require "reasonable":  $\sum_i w_i = 1$

$$-\epsilon \leq w_i \leq 1 + \epsilon$$

Video:  
families



## A Driving Application



## Thanks!

- To the UW graphics gang.
- Animation research at UW is sponsored by the National Science Foundation, Microsoft, and the Wisconsin University and Industrial Relations program.
- House of Moves, IBM, Alias/Wavefront, Discreet, Pixar and Intel have given us stuff.
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- And to all our friends in the business who have given us data and inspiration.

