Compact Real-Time Modeling of Seated Humans by Video Sprite Sequence Quantization

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Motivation

 Inadequate support for distributed multi-user computer graphics applications

- □ Distance learning
- Teleconferencing
- Distributed virtual environments



Problems

Modeling

Real-time modeling of participants is challenging (depth acquisition)

Bandwidth

Low bandwidth between acquisition and rendering sites (1-3 Mbps for commodity connectivity such as DSL or cable modem)



Modeling: video sprites [SHADE 1996]

- □ Inexpensive—\$100 webcam
- Robust and efficient—use favorable background
- □ Effective—realism of video frames
- Approximate—limited range of desired views
- Data-intensive— 640x480 @10fps video sprite require 1-5 Mbps





MPEG compression high quality – 280 Kbits / sprite





MPEG compression low quality – 53 Kbits / sprite





Our result – 16 bits / sprite



Modeling: video sprites [SHADE 1996]



Webcam

Video frame

Video sprite



- Modeling: video sprites [SHADE 1996]
- Bandwidth reduction: leverage limited number of representative body poses for seated participant
 - □ Neutral, raising right hand, chin on hand ...





- Modeling: video sprites [SHADE 1996]
- Bandwidth reduction: leverage limited number of representative body poses for seated participant
 Pixel-level differences do not necessarily imply semantic differences



Input sprite Database sprite difference Input sprite Database sprite difference



- Modeling: video sprites
 [SHADE 1996]
- Bandwidth reduction: leverage limited number of representative body poses for seated participant
 - Not all frame regions have same importance

REC





- Modeling: video sprites [SHADE 1996]
- Bandwidth reduction: leverage limited number of representative body poses for seated participant
 Video compression doesn't!



Results Preview





Input Database 640 x 480, 7fps



System Overview

Acquisition ModuleRendering Module



Acquisition Module





Rendering Module





Database Construction

- Transform frame into sprite using background subtraction
- Segment the raw video footage into sequence by pauses between poses

Off-Line Database Construction



Database Construction

Construct a binary tree of depth 3 on sprite sequence shape

- Shape is classified for each sprite sequence
- Tree is constructed recursively from the set of sprite sequence based on shape





Shape classification



yes-left

yes-right

yes-left & yes-right no-left & no-right

- Shape is a bitmap with pixel value 1 if it is hit by any sprite in the sequence, 0 otherwise
- Shoulder lines (blue) and central line (yellow) define the three regions
- Shape of sprite sequence is defined as left, right, left + right, others.



Tree construction

- Assignment of sprite sequence according to which region its shape crosses into
- A tree leaf stores an array of sprite sequence



- 2550 raw frames
- 32 sequences
- 1592 sprites
- 4 leaves (top -> bottom)
 - 10 sequences
 - □ 4 sequences
 - 10 sequences
 - 8 sequences



Tree construction



- 1650 raw frames
- 28 sequences
- 1044 sprites
- 4 leaves (left -> right)
 - □ 7 sequences
 - □ 7 sequences
 - □ 7 sequences
 - □ 7 sequences



Real-time Sprite Matching

- For each input sprite, down-sample and blur the sprite
- Traverse the tree based on sprite shape until appropriate leaf of the tree is found
- Discard leaf sprite sequences with unmatched shape
- Linear searching the remaining leaf sprite sequences to get the best match based on color + shape

Sprites Comparison in Linear Searching





- Align the two sprites by 2D translation
- Comparison on shape and color using subset of pixels
- Sprites in the same group have higher searching priority (coherence)



Head Region Processing

In high fidelity mode, head region of the input sprite is detected, compressed, sent to the rendering site, and composited with the matching database sprite





Head Region Detection and Composition

- Detection of bounding box of head region
- Composition
 - Neck alignment
 - □ Blending in transitional area below the head region



database





input





Head Region Detection and Composition

- Detection of bounding box of head region
- Composition
 - Neck alignment
 - Blending in transitional area below the head region
 - Using k-most recent frames to estimate the frame-to-frame changes of the position of the head.



database





input



composite



K = 1



Head Region Detection and Composition





Results

Test on four different subjects
640x480 Video raw footage frames at 30fps



Performance at acquisition site

- Searching time on 1000 downsampled 160x120 frames (100 MB)
 - □ Brute force linear searching: 500 ms
 - □ Binary tree searching all sprites at the current leaf: 250 ms
 - □ Use coherence: 120ms
- Database uploading time for 2000 compressed sprites (95% quality factor)
 - □ 360 kbps uploading speed
 - 20/10/5 minutes for 640x480/320x240/160x120 resolutions



Performance at rendering site

- Size of the stored database depends on the resolution of the frame, compression quality factor and the complexity of the texture
- The renderer can easily handle 20 databases of 2000 640x480 sprites (1.2 GB)



Memory size of a 640x480 sprite as a function of the compression quality factor



Performance at rendering site

- The decompression time depends little on compression quality factor
- Decompression time determines the number of databases to be loaded
- Average decompression time for 95% quality factor
 - 12/3.5/1.0 ms for
 640x480/320x240/160x120
 resolutions



Total frame time and decompression time of 640x480 sprite as a function of the compression quality factor



Simulated Renderer



Simulated renderer with 30 databases of 2000 160x120 sprites (540 MB) at 25fps



Simulated Renderer



Simulated renderer with 6 databases of 2000 640x480 sprites (360 MB) at 10 fps



Conclusion

Our method

- Models seated human in real time for teleconferencing and distance learning application
- Drastically reduces the transmission data by sending 16 bits index per sprite
- Makes best use of MTU if sending the index with the audio data



Future Work

- Apply our method in the context of an actual distance learning system
- Use as analytical tool for video archive
 - □ Video summarization
 - □ Video abstraction



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