

Structured-Light Based Acquisition (Part 2)

CS635 Spring 2010

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Acquiring Dynamic Scenes

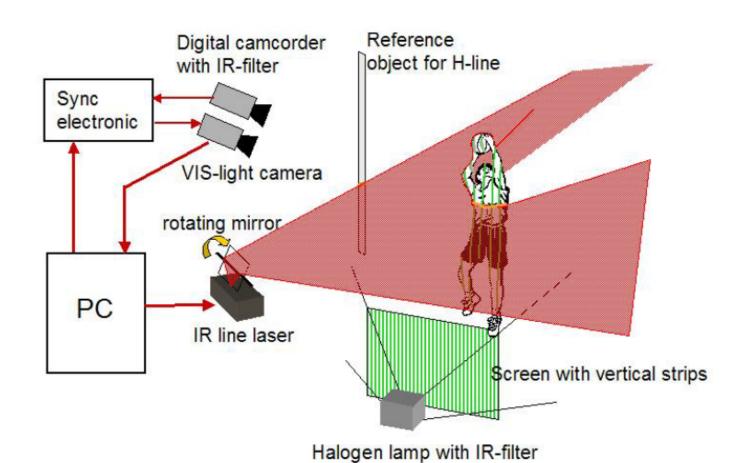
- Scene: object (or camera) is moving and/or object is deforming
- Acquisition: capture as much information as possible in one to a few frames
 - By exploiting coherence
 - By exploiting several "channels" of information (e.g., color, infrared, etc...)



Acquiring Dynamic Scenes

- "Capturing 2½D Depth and Texture of Time-Varying Scenes Using Structured Infrared Light", Frueh and Zakhor, PROCAMS 2005
 - Use single-frames and infrared illumination...
- "Rapid Shape Acquisition Using Color Structured Light and Multi-pass Dynamic Programming", Zhang et al., 3DPVT 2002
 - Use single-frames and colored patterns...
- "Fast 3D Scanning with Automatic Motion Compensation", Weise et al., CVPR
 2007
 - Use phase shifting and motion compensation over a few frames...
- "Real-time 3D Model Acquisition", Rusinkiewicz et al., SIGGRAPH 2002
 - Use patterns producing local correspondences over a few frames and merge...

Capturing 2½D Depth and Texture of Time-Varying Scenes Using Structured Infrared Light

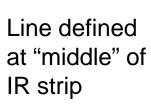




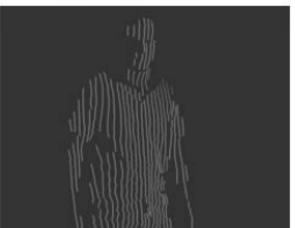
V-lines











How do you know which line is which? Ideas?

H-lines



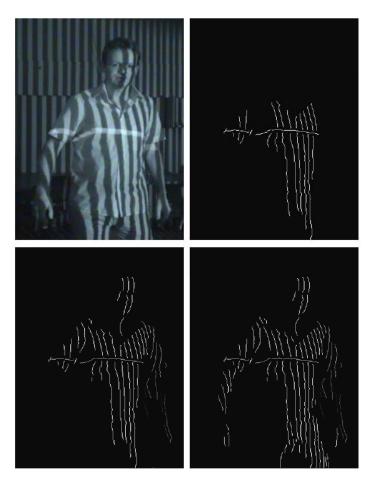
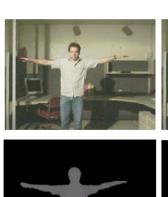


Figure 6: Reconstructing the depth along V-lines. (a) IR frame; (b) V-lines from intra-frame tracking only; (c) V-lines with additional forward inter-frame tracking, (d) final result after V-lines with both forward and backward inter-frame tracking, and line counting.

H-line sweeps up/down at 2Hz and enables an ordering of (a subset of) the V-lines and thus permits their correspondence













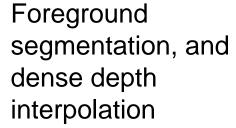
Grab color image



















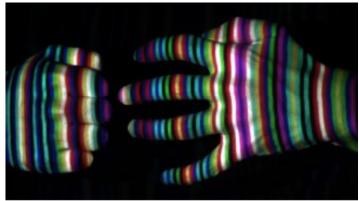
Put it all together...

IR camera at 30Hz, color camera at 10Hz (probably faster today...)



 Recall: how do we correspond lines?





- Use color transitions to define features
- Define lines at the transitions from color A to color B





- What is a notable problem?
- Resolution. Why?





- Only have three color channels (R,G,B) and can only robustly differentiate "strong" color changes
- This reduces the number of colors to use, and
- Often results in ambiguity in the color coding

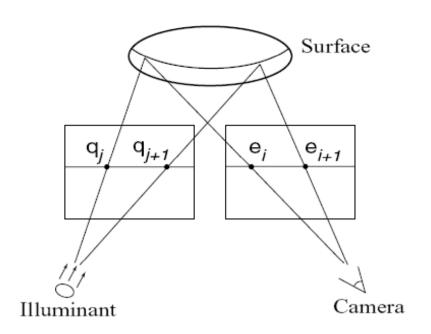
Challenges

— Given a color code, how to do "best" correspond the stripes?

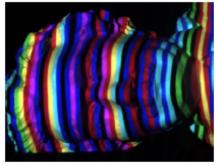
— With the above in mind, how do we design a good color code?

- Challenges
 - Given a color code, how to do "best" correspond the stripes?
 - With the above in mind, how do we design a good color code?

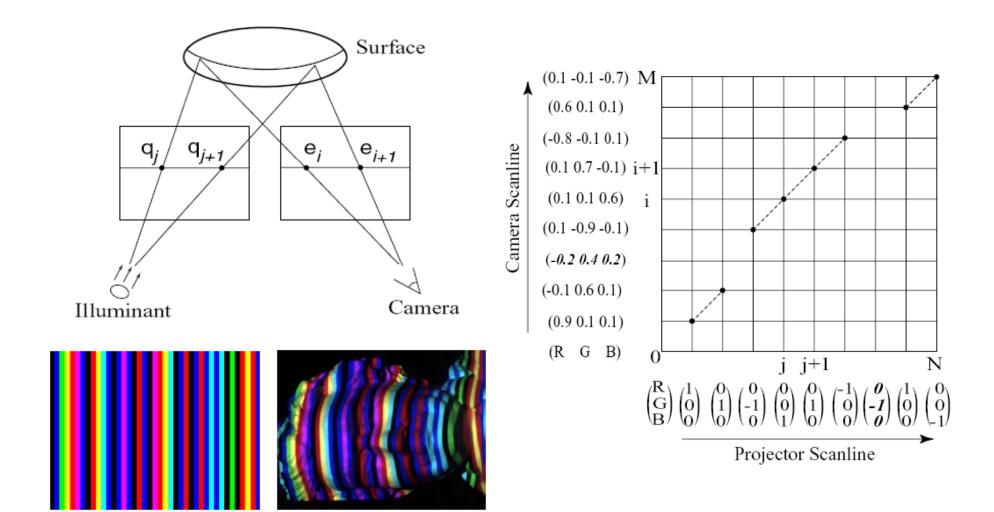
- Solution
 - Dynamic Programming

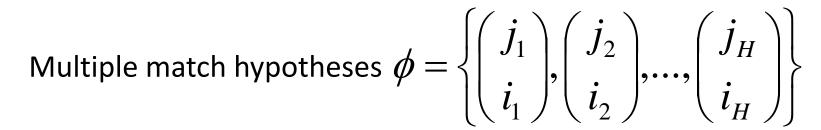






(rectified images)





Similarity score (of color) between edge e_i and transition q_j is $S(q_j, e_i)$

Score of the entire match sequence $f(\phi) = \sum_{k=1}^{n} s(q_{j_k}, e_{i_k})$

Dynamic programming objective is: $\underset{\phi}{\operatorname{arg}} \max(f(\phi))$

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However, the space all possible ϕ is very large: O(M $^{
m N}$)

Solution?

Assume monotonicity (of the depth ordering):

$$i_1 \leq i_2 \leq \ldots \leq i_H$$

Great! But this monotonicity does **not** hold in what situation?

Occlusions! Oh well...

But it holds for individual fragments, which we can combine

Dynamic programming objective is: $\underset{\phi}{\arg\max}(f(\phi))$ Let optimal ϕ be called ϕ^*

$$f(\phi^*_{ji}) = \begin{cases} 0 & \text{if } j=0 \text{ or } i=0 \\ & \int_{0}^{\infty} \int_{0}^{\infty} \frac{f(\phi^*_{j-1,i-1}) + s(q_j, e_i)}{f(\phi^*_{j-1,i})} \\ & \int_{0}^{\infty} \frac{f(\phi^*_{j-1,i-1}) + s(q_j, e_i)}{f(\phi^*_{j,i-1})} \end{cases}$$

f found through a recursive search and some optimizations to further reduce the search space (e.g., assume at most small depth changes from one column to another)

Challenges

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How do we design a good color code?

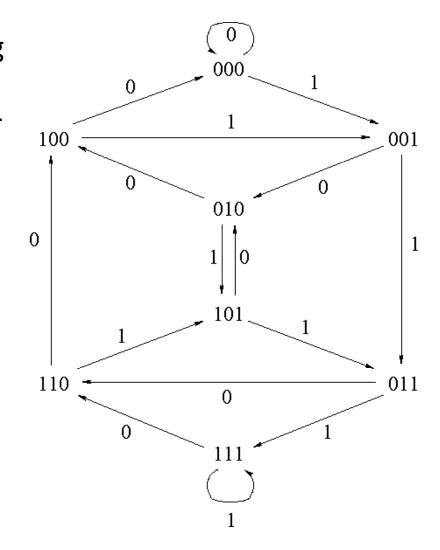
- De Bruijn sequence B(k,n)
 - is a cyclic sequence of a given alphabet A with size k for which every possible subsequence of length n in A appears as a sequence of consecutive characters exactly once
- B(k,n) has length k^n
- Example: $A = \{0, 1\}$
 - -B(2,3) = 000101111 or 11101000



De Bruijn sequence B(k,n)

 Can also be constructed by taking a Hamiltonian path of an ndimensional De Bruijn graph over k symbols; e.g.,

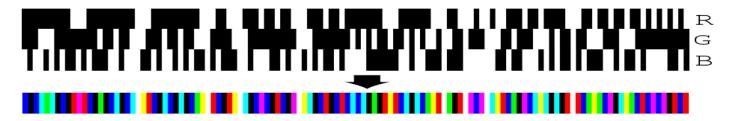
(Hamiltonian path means each vertex is visited once)



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Color Sequence

- Colors = $\{000, 100, 110, ..., 111\}$ total of 8-1=7 because 000 is useless
- Color sequence is created by $p_{j+1}=p_j XOR d_j$
 - XOR'ing effectively "flips bits" using d_i
 - $-p_0$ is a chosen initial color (e.g., 100)
- Want 3 letters sequences d_i to be unique
- In practice about 125 stripes is sufficient
- Thus, a B(5,3) is adequate

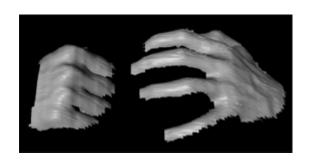


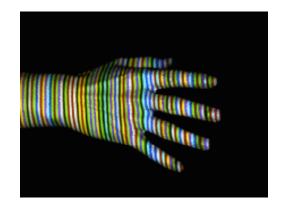


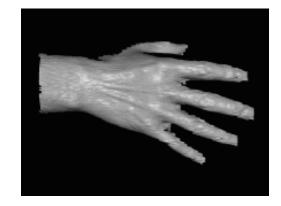
Examples











Fast 3D Scanning with Automatic Motion Compensation

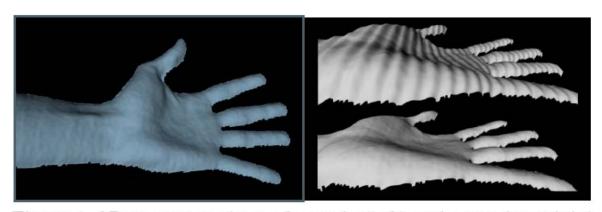


Figure 1. 3D reconstructions of a static (left) and a moving (right) hand. Motion compensation (bottom right) removes the ripples from the reconstructed surface (top right).

- Higher resolution/quality than previous method
- Uses phase-shifting and motion-compensation

Fast 3D Scanning with Automatic Motion Compensation

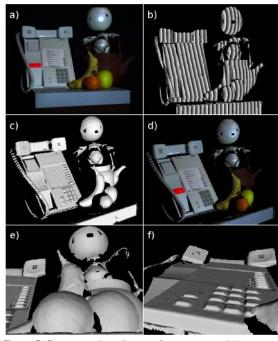


Figure 7. Reconstruction of a complex scene containing several objects (phone, teapot, figure, fruit): a) texture image, b) reconstructed phase, c) geometry, d) textured geometry, e)+f) close-ups



Figure 8. Reconstruction of a waving cloth. Motion correction correctly removes the ripples (right).



Figure 9. Reconstruction of a person speaking.

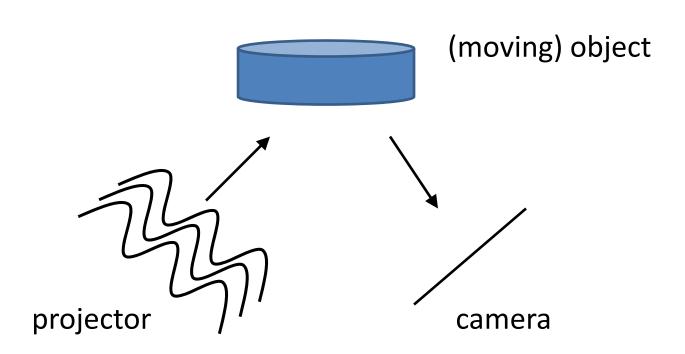


Figure 10. Reconstruction of moving hands in front of the torso. On the right with motion compensation.



Figure 11. Online reconstruction of hand gestures.

What is (standard) phase shifting





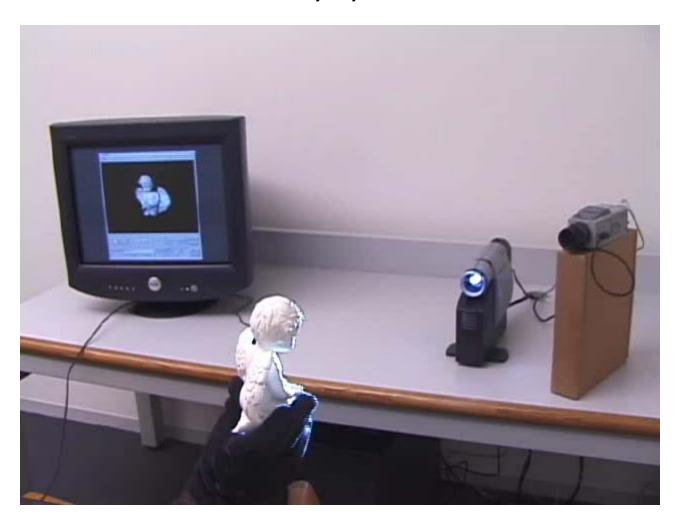
Motion Compensation

 Since phase shifting assumes a static scene, correlation-based stereo is used to compensate for motion

 An additional modification is proposed to handle discontinuities (which also plague standard phase shifting)

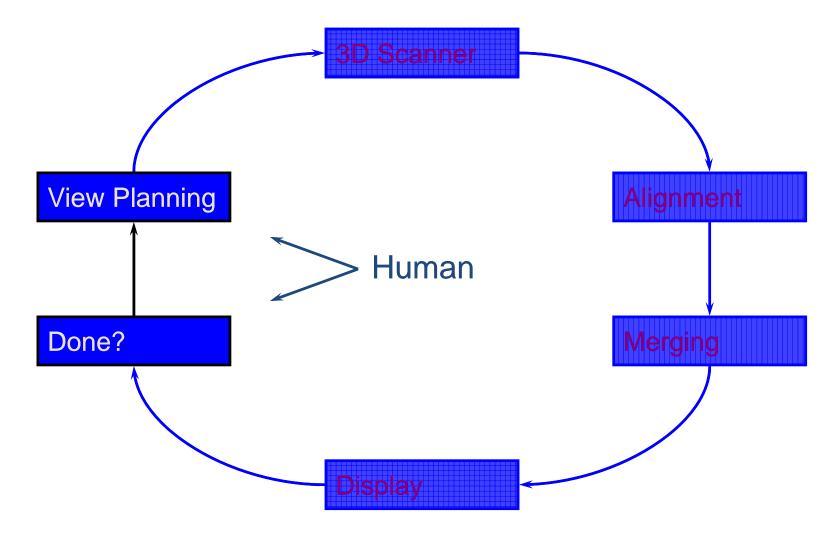
Real-Time 3D Model Acquisition

(slides and videos of this section by Syzmon Rusinkiewicz @ Princeton



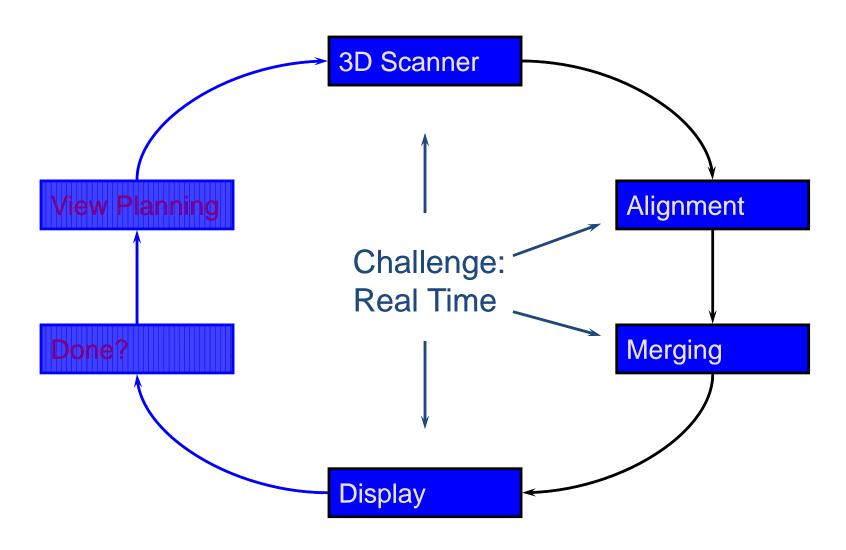
Real-Time 3D Model Acquisition Pipeline





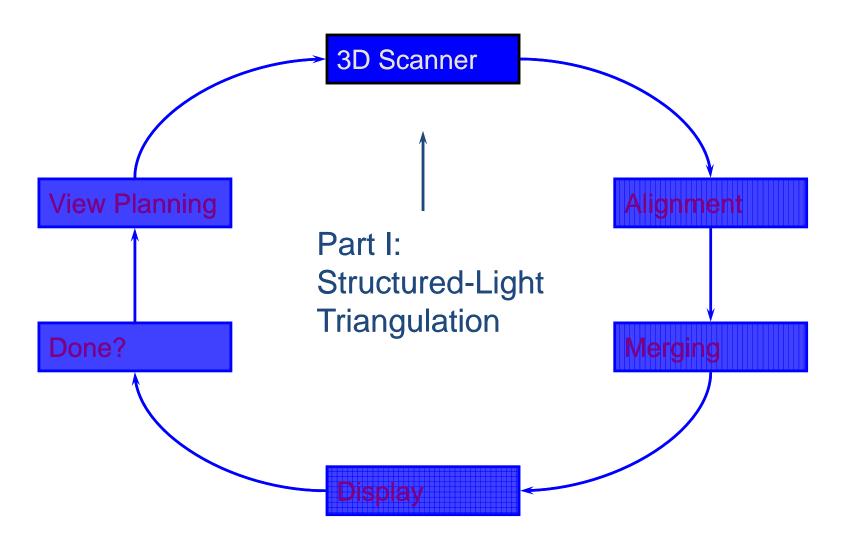
Real-Time 3D Model Acquisition _ Pipeline





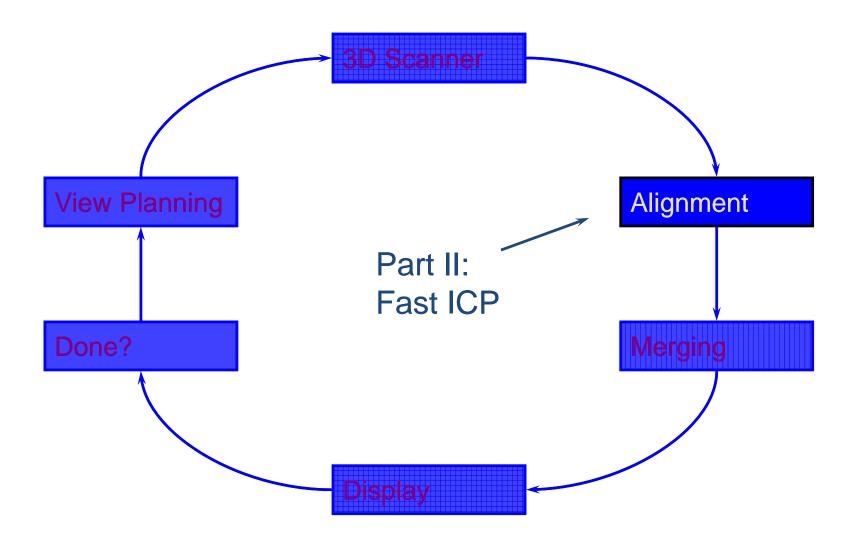
Real-Time 3D Model Acquisition _ Pipeline





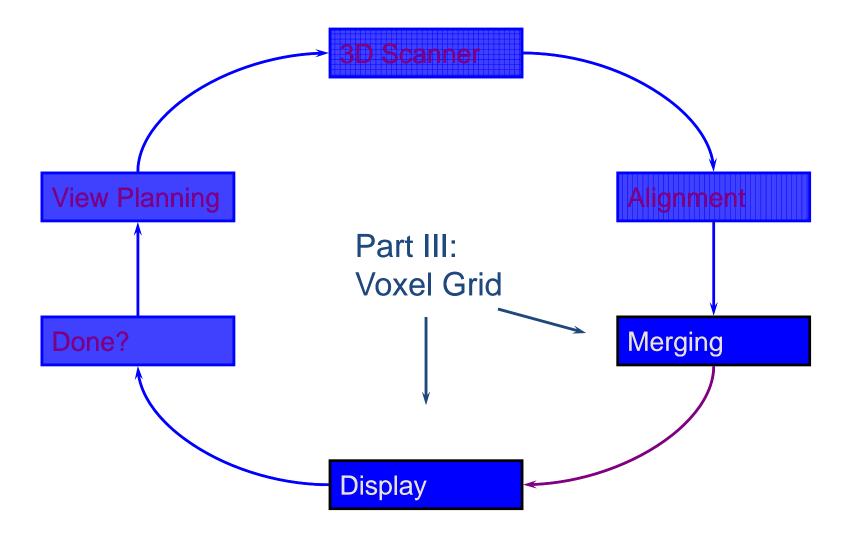
Real-Time 3D Model Acquisition Pipeline





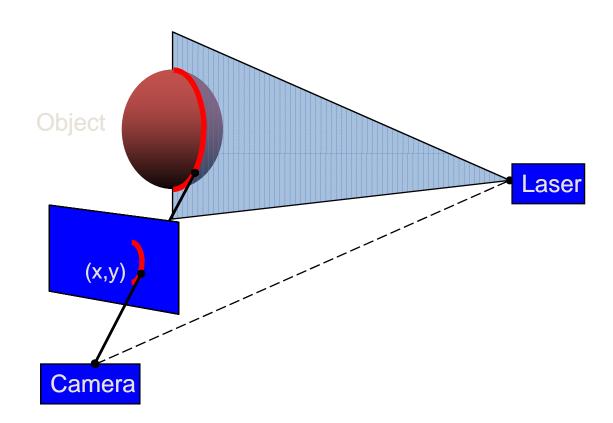
Real-Time 3D Model Acquisition Pipeline







Triangulation



• Depth from ray-plane triangulation



Triangulation

- Faster acquisition: project multiple stripes
- Correspondence problem: which stripe is which?



Codes for Moving Scenes

- Assign time codes to stripe boundaries
- Perform frame-to-frame tracking of corresponding boundaries
 - Propagate illumination history

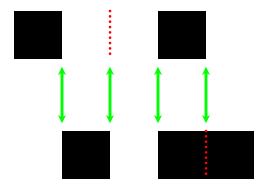
Illumination history = (WB),(BW),(WB) Code

[Hall-Holt & Rusinkiewicz, ICCV 2001]



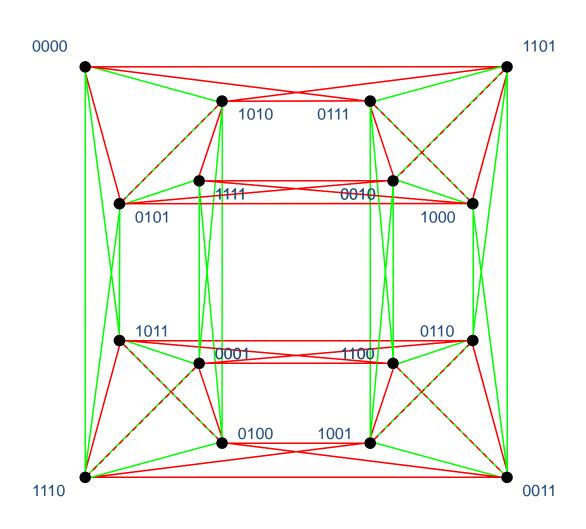
Designing a Code

- Want many "features" to track: lots of black/white edges at each frame
- Try to minimize ghosts WW or BB "boundaries" that can't be seen directly





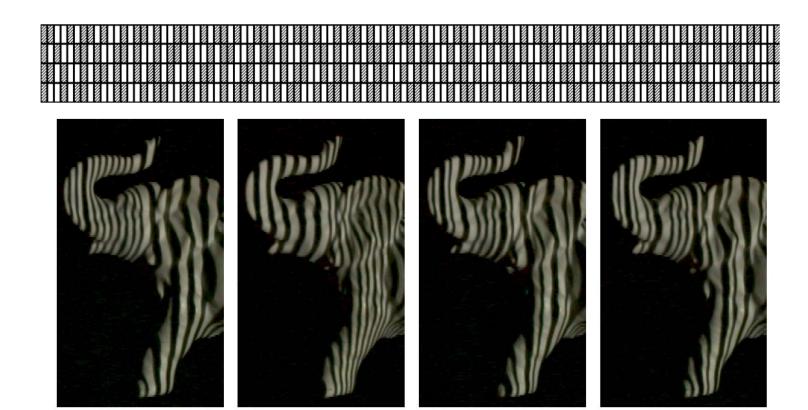
Designing a Code



[Hall-Holt & Rusinkiewicz, ICCV 2001]

Space-Time Boundary Code

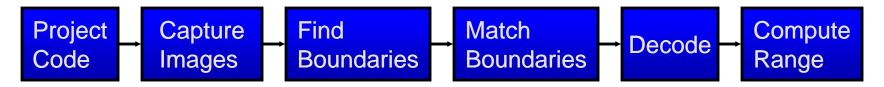






Implementation

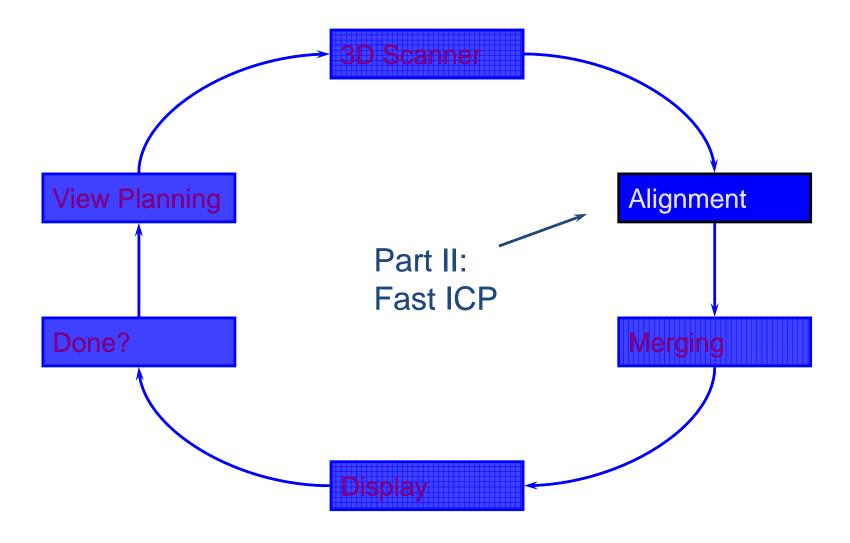
• Pipeline:



- DLP projector illuminates scene @ 60 Hz.
- Synchronized NTSC camera captures video
- Pipeline returns range images @ 60 Hz.

Real-Time 3D Model Acquisition Pipeline







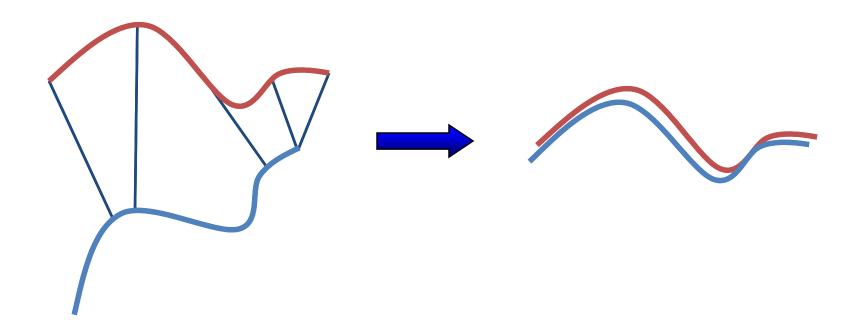
Aligning 3D Data

- This range scanner can be used for any moving objects
- For rigid objects, range images can be aligned to each other as object moves



Aligning 3D Data

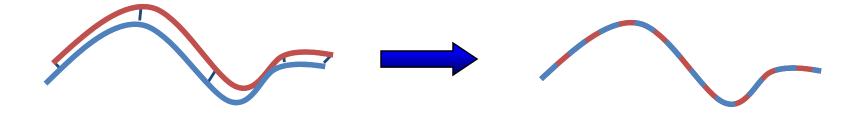
• ICP (Iterative Closest Points): for each point on one scan, minimize distance to closest point on other scan...





Aligning 3D Data

- ... and iterate to find alignment
 - Iterated Closest Points (ICP) [Besl & McKay 92]





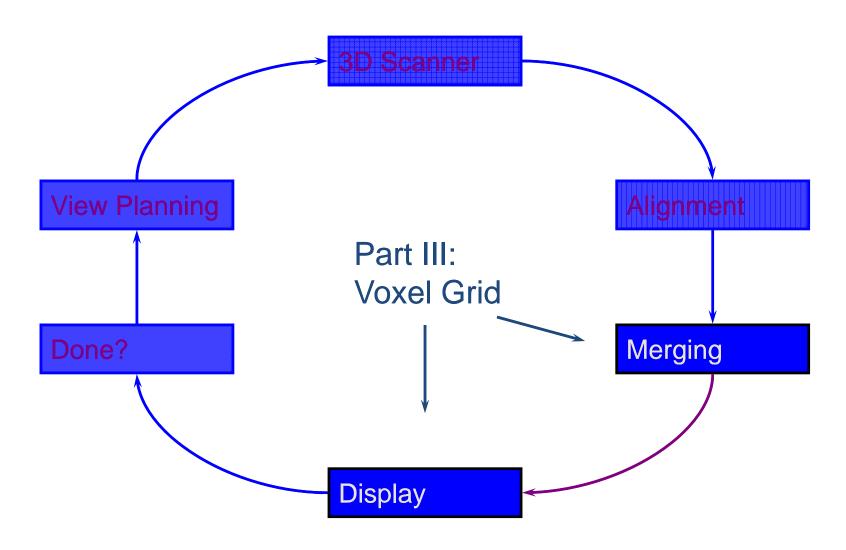
ICP in the Real-Time Pipeline

- Potential problem with ICP: local minima
 - In this pipeline, scans close together
 - Very likely to converge to correct (global)
 minimum
- Basic ICP algorithm too slow (~ seconds)
 - Point-to-plane minimization
 - Projection-based matching
 - With these tweaks, running time ~ milliseconds

[Rusinkiewicz & Levoy, 3DIM 2001]

Real-Time 3D Model Acquisition Pipeline





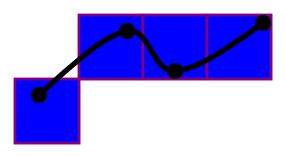


- Goal: visualize the model well enough to be able to see holes
- Cannot display all the scanned data accumulates linearly with time
- Standard high-quality merging methods:
 processing time ~ 1 minute per scan

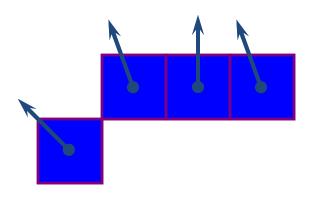




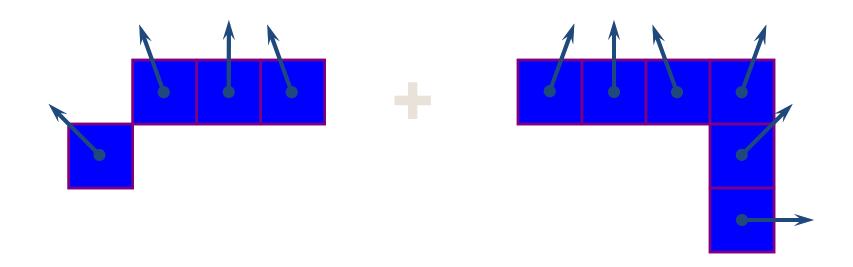




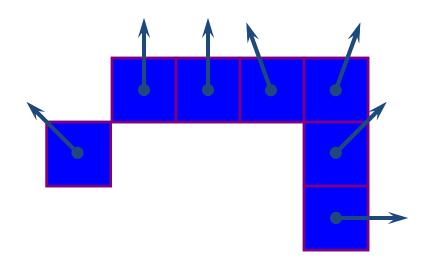












Point rendering, using accumulated normals for lighting



Example: Photograph





Result



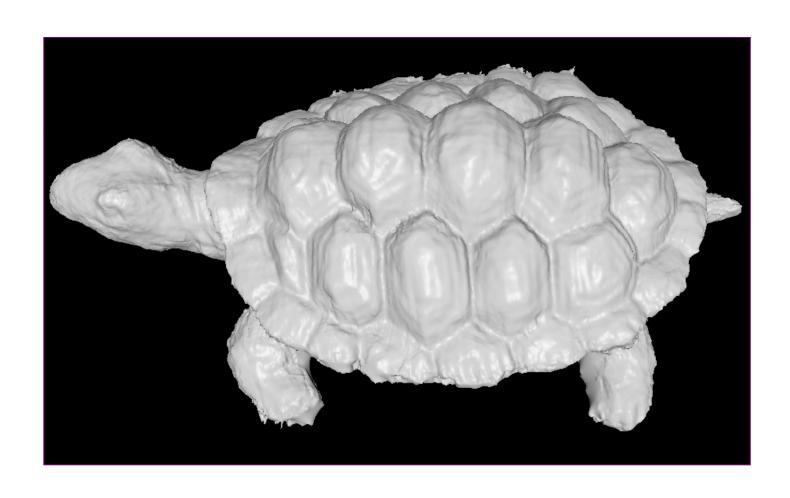


Postprocessing

- Real-time display
 - Quality/speed tradeoff
 - Goal: let user evaluate coverage, fill holes
- Offline postprocessing for high-quality models
 - Global registration
 - High-quality merging (e.g., using VRIP [Curless 96])



Postprocessed Model





Recapturing Alignment



PUR

Summary

- 3D model acquisition pipeline optimized for obtaining complete, hole-free models
- Use human's time most efficiently
- Pieces of pipeline selected for real-time use:
 - Structured-light scanner for moving objects
 - Fast ICP variant
 - Simple grid-based merging, point rendering

PUR

Limitations

- Prototype noisier than commercial systems
 - Could be made equivalent with careful engineering
 - Ultimate limitations on quality: focus, texture
- Scan-to-scan ICP not perfect ⇒ alignment drift
 - Due to noise, miscalibration, degenerate geometry
 - Reduced, but not eliminated, by "anchor scans"
 - Possibly combine ICP with separate trackers