

Challenges on 3D shape acquisition for VR/AR/MR systems

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Oxford Brookes University, Wheatley

Hiroshi Kawasaki (Saitama University)
joint research with
Katsushi Ikeuchi (University of Tokyo) and
Ryo Furukawa (Hiroshima city University)

My research areas

Reconstruct real-world in a computer

- Acquisition
 - Geometry – Active scanning method
 - Photometry – BTF (BRDF) sampling machines
- Modeling
 - Geometry – Registration, integration and hole-filling
 - Photometry – BTF (BRDF) analysis and compression
- Photo-realistic rendering
 - Image based rendering and model based rendering
- Application
 - VR, ITS, CAD, digital archiving and Web

Current research projects

- City modeling project (ITS) granted by motor company, map company and government
- Active 3D scanning granted by MOF
- Texture acquisition and analysis granted by SCOPE
- Digital archiving project (Ikeuchi Lab.)

Research and project matrix

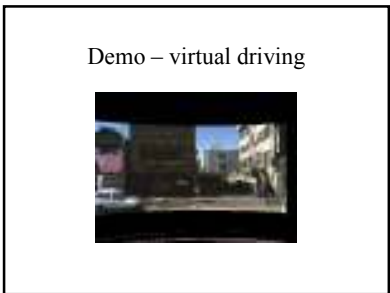
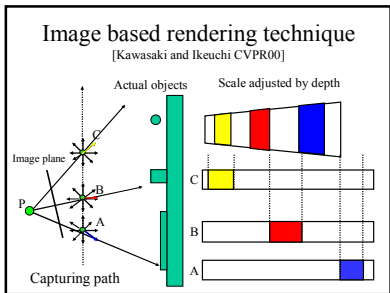
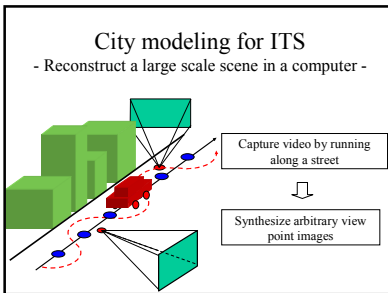
	Acquisition		Modeling		Rendering
	Geometry	Photometry	Geometry	Photometry	
City modeling (ITS)	△	◎		△	◎
Active 3D scanner	◎		◎		
Texture analysis		◎		◎	◎
Digital archiving	◎	△	◎	△	

Research and project matrix

	Acquisition		Modeling		Rendering
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Digital archiving	◎	△	◎	△	

Research and project matrix

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Texture analysis		◎		◎	◎
Digital archiving	◎	△	◎	△	



Several problems

- Quality (depend on number of images)
- No texture for **upper bound**

Problem -- input images

• configure multiple cameras

Simple solution → More Cameras

Captured Images

Fused Image

Still problem

Inconsistency of Optical Center

Consistency in Spatio-temporal Space

[H.Kawasaki and K.Ikeuchi ACCV 2003]

- × Inconsistent in space
- Consistent in Spatio-temporal Space

Image Fusion without Distortions

Time=T1 Time=T2 Time=T3

How to calibrate spatio-temporal parameter?

Spatio-temporal calibration

All cameras are parallel to moving direction

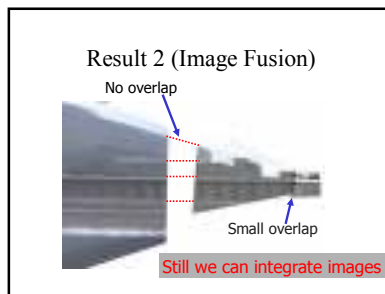
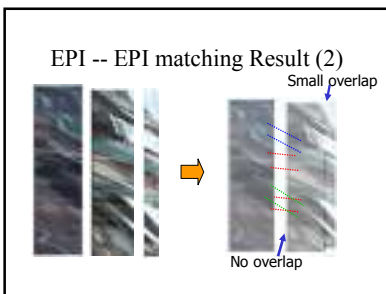
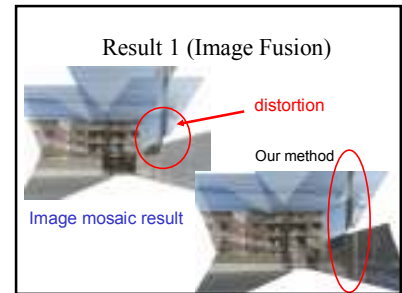
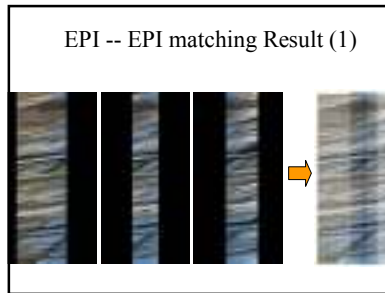
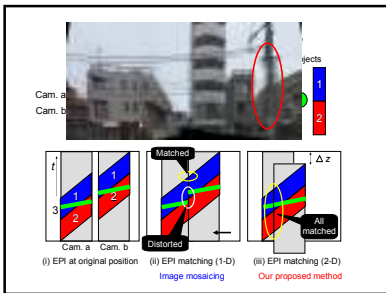
Camera configuration [US patented '07]

Spatio-temporal calibration

→ use Epipolar Plane Image (EPI)

Angle of inclination represents a depth of object

$mV \propto D$



Hybrid rendering technique

[Kawasaki et al. CVPR '01]

- Back ground – image based rendering
- Near object – model based rendering (e.g. road surface, sign board)

Real-time rendering of IBR

- Multiple billboards for rendering
- target 60Hz

Wall

Update texture

Time: t Time: $t+1$



Registration between map and image

- Using GPS and manual modification

Panoramic images

Result – driving simulator

[Kawasaki et al ACM ISMAR'05]

Current research topics

- Accurate registration of camera position using images (SfM for omni-cameras)
- Real-time rendering using GPU
- Noise removal (reflection on windows, pedestrian and cars)

Research and project matrix

	Acquisition		Modeling		Rendering
	Geometry	Photometry	Geometry	Photometry	
City modeling (ITS)	△	⊙		△	⊙
Active 3D scanner	⊙		⊙		
Texture analysis		⊙		⊙	⊙
Digital archiving	⊙	△	⊙	△	

Why active 3D scanner?

- Passive system
 - ⊙ Input → only images
 - ⊖ Unstable
 - ⊖ Sparse 3D points
- Active system
 - ⊙ Stable and high accuracy
 - ⊙ Dense 3D points
 - ⊖ System → expensive and heavy

Because...

- Passive system
 - ⊙ Input → only images
 - ⊖ Unstable
 - ⊖ Sparse 3D points
- Active system
 - ⊙ Stable and high accuracy
 - ⊙ Dense 3D points
 - ⊖ System → expensive and heavy

Required for VR and other actual systems!

Active 3D scanner

Minolta (light sectioning)

Cyrax (time of flight)

Z+F (phase-based)

Active Stereo 3D scanner

- ⊙ Stable
- ⊙ Dense 3D points
- ⊖ Complicated and expensive system (lasers, mechanical actuator and laser sensor)

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Eliminate them by computer vision techniques!

Active Stereo 3D scanner

- ⊕ Stable
- ⊕ Dense 3D points
- ⊖ ~~Complicated and expensive system (lasers, mechanical actuator and laser sensor)~~

Eliminate them by **computer vision techniques!**

↓

Hand-held 3D scanner

What is a hand-held 3D scanner?

Line-Laser Projector

Single or multiple cameras
For simplicity → Single camera

Hand-held scan by single camera

- Light sectioning method (triangulation)
- Require a laser plane parameters

Laser plane ($ax+by+cz=1$)

Laser line

Single Camera

Previous methods

- Online calibration
 - Frame [Chu et al. 3DIM01]
 - Planes [David 2006 <http://www.rob.cs.tu-bs.de/news/david>]

A calibration frame

Our first system

[Furukawa and Kawasaki 3DIM03]

- Online calibration
 - Attach LEDs

LED Markers

Captured image with laser and LEDs are in a image

Capturing sequence and precision

Histogram of error

Error of distance[m]	Frequency
25000	0
26000	0
27000	0
28000	0
29000	0
30000	0
31000	0
32000	0
33000	0
34000	0
35000	0
36000	0
37000	0
38000	0
39000	0
40000	0
41000	0
42000	0
43000	0
44000	0
45000	0
46000	0
47000	0
48000	0
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66000	0
67000	0
68000	0
69000	0
70000	0
71000	0
72000	0
73000	0
74000	0
75000	0
76000	0
77000	0
78000	0
79000	0
80000	0
81000	0
82000	0
83000	0
84000	0
85000	0
86000	0
87000	0
88000	0
89000	0
90000	0
91000	0
92000	0
93000	0
94000	0
95000	0
96000	0
97000	0
98000	0
99000	0
100000	0

$\sigma=0.00017(m)$

Experiments of shape acquisition

Human face

Entire shape acquisition – rotation table

Laser pointer

Target object

Projected plane

LED markers

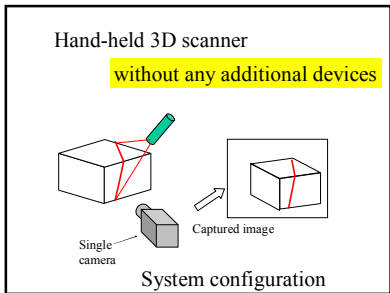
Video camera

Rotation table results

Hand-held scanner with single cam.

- ⊗ Simple configuration
- ⊗ No precision devices
- ⊗ Online-calibration
 - △ Frames or planes are required [david'06]
 - △ LED markers required [kawasaki'03]

Can we eliminate all additional devices ?



Question 1

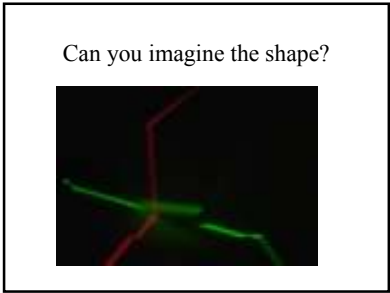
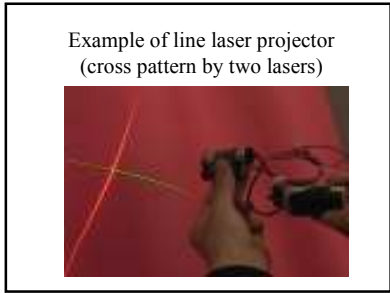
- Can we reconstruct shape from the following image?

NO!

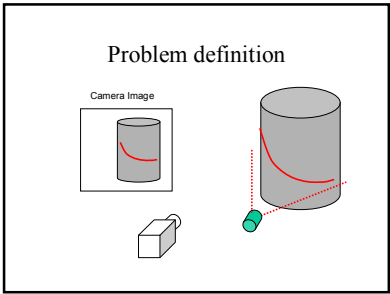
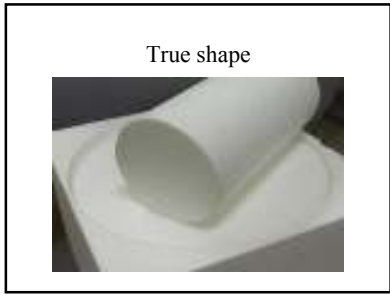
Question 2

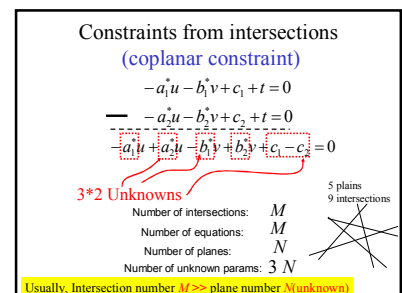
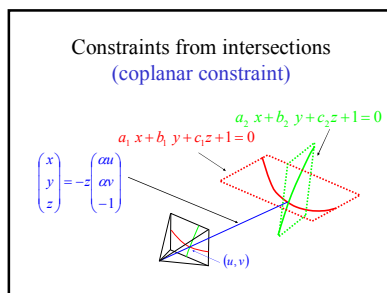
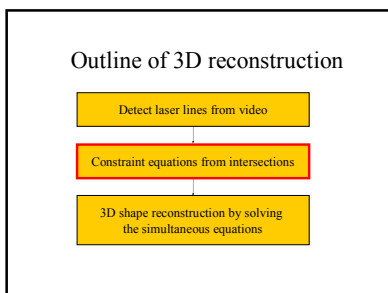
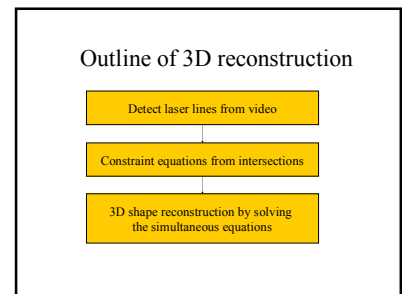
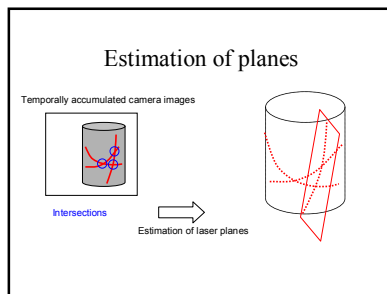
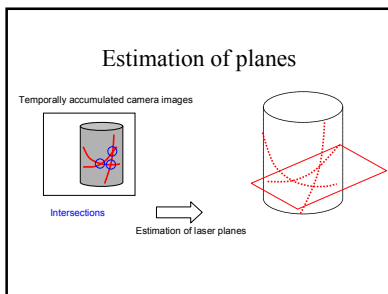
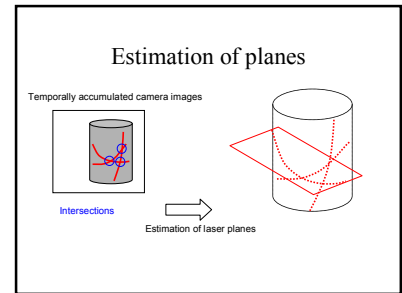
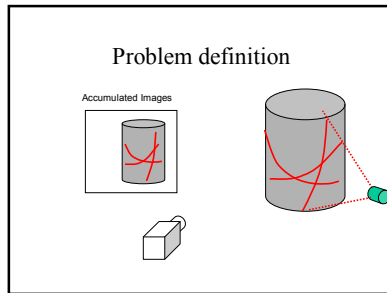
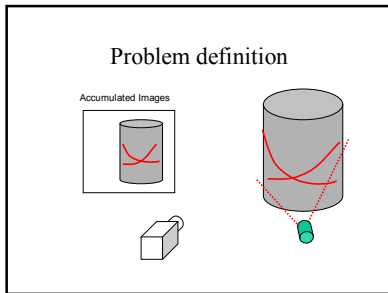
- How about this?

Maybe?



Is it possible to reconstruct the 3D shape?





Matrix form

$$\begin{pmatrix} a_1 & b_1 & c_1 & 0 & 0 & 0 & 0 & 0 & 0 \\ u_1 & v_1 & 1 & -u_2 & -v_2 & -1 & 0 & 0 & 0 \\ u_1 & v_1 & 1 & 0 & 0 & 0 & u_2 & v_2 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \\ a_2 \\ b_2 \\ c_2 \\ a_3 \\ b_3 \\ c_3 \end{pmatrix} = \mathbf{0}$$

$\mathbf{Lx} = \mathbf{0}$

$\mathbf{L} : 3N * M$ matrix (Intersection num M , Plane num N)

Reconstruction from coplanarity

$\mathbf{Lx} = \mathbf{0}$

Solution \mathbf{x} has 4 degrees of freedom (Projective reconstruction)

The 4 DOFs → Found in other research areas.
e.g. Polyhedra analyses in single view reconstruction
Generalized Bas-Relief Ambiguity in photometric stereo

Upgrade to Euclidean solution

- Metric constraints from laser planes

Laser projector with 2 line lasers

Laser Planes

known angle (= 90 degree)

Formulation of metric constraints

Constraints from orthogonality

$$ax + by + cz + 1 = 0 \quad dx + ey + fz + 1 = 0$$

$$ad + be + cf = 0$$

90 degree

Outline of 3D reconstruction

- Detect laser lines from video
- Projective reconstruction from intersections
- Euclid upgrade using metric constraints

Eliminate 4dofs

Sparse data sets

Outline of dense 3D reconstruction

- Detect laser lines from video
- Projective reconstruction from intersections
- Euclid upgrade using metric constraints
- Dense shape reconstruction

Experiments

- Simulation data
- Real data

Real data 2

- Cross line laser

Real data 2 – result

Proposed method

- Only require a line laser and a single camera
 - General solution for "Shape from Coplanarity"
 - Any other applications?

Other applications

- Shape from cast shadow

Shape from cast shadow

Other applications

- Single view reconstruction

Other applications

- Single view reconstruction

Summary of the proposed method

- Temporal accumulation
 - ○ Self-calibration for hand held 3D scanner
 - △ take times

→ Can we make enough intersections at one time?

Self-calibration for 3D scanner

- Use many laser projectors

Self-calibration for Coded Structured light

- Use many laser projectors

More projectors to pixel resolution

Coded structured light


Self-calibration for Coded Structured light

[H.Kawasaki and R.Furukawa, 3DIM2005]

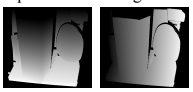
- No mechanical devices
- No pre-calibration → Flexible configuration
- High accuracy
- Dense sampling

Structured light example

- Projecting patterns → two directions

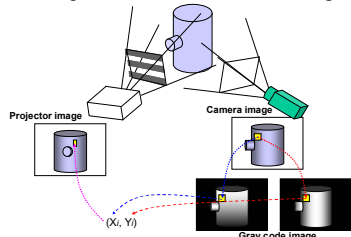


- Acquired coded images



vertical horizontal

Correspondences from decoded images



Projector image Camera image

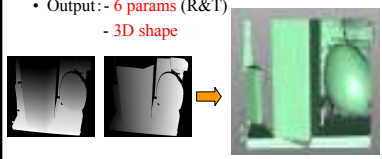
(X_c, Y_c) Gray code image

Self-calibration for Coded Structured light

- problem definition -

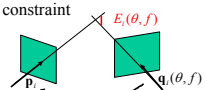
- Input : - camera params (focal length, etc.)
- two index images

- Output : - 6 params (R&T)
- 3D shape



Non-linear optimization

- Epipolar constraint



Extrinsic parameter $\theta = (t, \alpha, \beta, \gamma)$

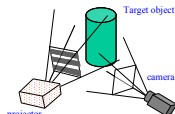
t : Translation α, β, γ : Rotation (Euler angles)

$$E_i(\theta, f) = |t \cdot N(\mathbf{p}_i \times \mathbf{q}(\alpha, \beta, \gamma, f))|$$

Using $F(\theta, f) = \sum_i \{E_i(\theta, f)\}^2$ as a minimizing function.

Feature of the method

- Single camera and a projector
- Freely movable (both camera and projector)
- High speed scan
- High accuracy



Target object camera projector

Extended techniques

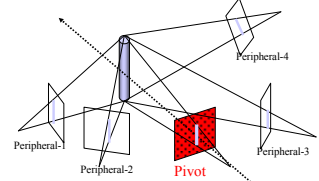
- Wide range reconstruction by pivot scanning
- Simultaneous reconstruction method

Extended techniques

- Wide range reconstruction by pivot scanning
- Simultaneous reconstruction method

Pivot scanning

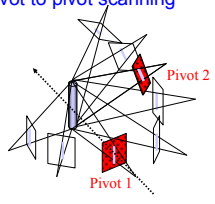
Fix **pivot** device and move **peripheral** device arbitrarily



Peripheral-1 Peripheral-2 Pivot Peripheral-3 Peripheral-4

More wide view scanning


Pivot to pivot scanning



Pivot 1 Pivot 2

Result – pivot to pivot

- No alignment algorithm applied



1+2 result

Demo movie

- Pivot scan

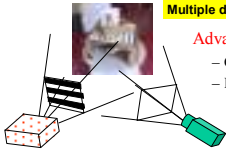


Extended techniques

- Wide range reconstruction by pivot scanning
- Simultaneous reconstruction method

Simultaneous reconstruction

- Capture multiple scenes
- 3D reconstruction simultaneously



Multiple depths for single pixel

Advantage


- Consistent scaling
- Improving result
- Redundant input

Demo movie


- Simultaneous scan



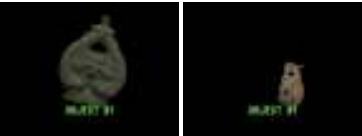
Results (1)



Results (2)



Final results



With

- Fast mesh integration [Furukawa and Kawasaki 3DIM '05]
- seamless texture [Inose, Kawasaki et al. '06 '07]

Live Demo

Self calibrating structured light

- Only a projector and a camera
- △ (Only several patterns, but) still take times

For actual VR systems, moving object should be captured

→ Can we eliminate several patterns to just "one"?

Previous method

- Spatial encoding method
 - Single pattern
 - × Low resolution
 - × Unstable

Proposed method

Simple grid pattern

- Only red and blue → Stable for image processing
- Dense as long as tractable → High resolution

Assume Pro-cams are precalibrated.

Epipolar constraint

Multiple candidates

Epipolar line

Proposed method

- ~~Epipolar constraint~~
- Coplanar constraint

[Kawasaki and Furukawa 3DIM'03, 3DIM'05, ACCV '07]

Algorithm overview

```

  graph LR
    Input[Input image] --> Pattern[Pattern detection]
    Pattern --> Curve[Curve detection]
    Curve --> Intersection[Intersection detection]
    Intersection --> Solution[Solution of correspondences]
    Solution --> Reconstruction[Dense reconstruction By light sectioning method]
  
```

Shape from coplanarity

Coplanarity constraint (1)

All points on a detected curves are on the undetermined common plane. $a_1x+b_1y+c_1z+1=0$

All planes go through the common line L_1 !

Planes for vertical pattern $d_1P_x+e_1P_y+f_1P_z+1=0$
 $d_1Q_x+e_1Q_y+f_1Q_z=0$

Correspondences are unknown

Known (calibrated) vertical planes

$L_1(Q_x, Q_y, Q_z)$

$L_2(P_x, P_y, P_z)$

Coplanarity constraint (2)

For horizontal patterns $d_1x+e_1y+f_1z+1=0$

All planes go through the common line L_1 !

Planes for horizontal pattern $d_1P_x+e_1P_y+f_1P_z+1=0$
 $d_1Q_x+e_1Q_y+f_1Q_z=0$

Correspondences are unknown

Known (calibrated) horizontal planes

$L_1(R_x, R_y, R_z)$

$L_2(P_x, P_y, P_z)$

Intersection constraint (3)

Intersection of curves on a captured image

→ Constraint between an undetermined vertical plane and an undetermined horizontal plane

Line of sight $(x,y,z)=(su, sv, -s)$

$a_1x+b_1y+c_1z+1=0$

$d_1x+e_1y+f_1z+1=0$

$ua_1 - u'd_1 + v'b_1 - v'e_1 + c_1 - f_1 = 0$

Intersection (u, v)

Example of formulation

vertical

$$\begin{aligned} v_1 \cdot o_p &= -1 \\ v_1 \cdot l_1 &= 0 \\ v_2 \cdot o_p &= -1 \\ v_2 \cdot l_1 &= 0 \\ v_3 \cdot o_p &= -1 \\ v_3 \cdot l_1 &= 0 \end{aligned}$$

horizontal

$$\begin{aligned} h_1 \cdot o_p &= -1 \\ h_1 \cdot l_1 &= 0 \\ h_2 \cdot o_p &= -1 \\ h_2 \cdot l_1 &= 0 \\ h_3 \cdot o_p &= -1 \\ h_3 \cdot l_1 &= 0 \end{aligned}$$

intersection

$$\begin{aligned} (v_1 - h_1) \cdot u_{1,1} &= 0 \\ (v_1 - h_2) \cdot u_{1,2} &= 0 \\ (v_1 - h_3) \cdot u_{1,3} &= 0 \\ (v_2 - h_1) \cdot u_{2,1} &= 0 \\ (v_2 - h_2) \cdot u_{2,2} &= 0 \\ (v_2 - h_3) \cdot u_{2,3} &= 0 \\ (v_3 - h_1) \cdot u_{3,1} &= 0 \\ (v_3 - h_2) \cdot u_{3,3} &= 0 \end{aligned}$$

All equations are **linear**
 → can be solved by **simultaneous equations**

Simultaneous equ. in a matrix form

$$\begin{bmatrix} o_p & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & o_p & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ u_{1,1} & 0 & 0 & -u_{1,1} & 0 & 0 \\ u_{1,2} & 0 & 0 & 0 & -u_{1,2} & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & u_{3,3} & 0 & 0 & -u_{3,3} \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ \vdots \\ -1 \\ 0 \\ \vdots \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Vertical
 L_v
Horizontal
 L_h
Intersection

Solution

$$\begin{bmatrix} o_p & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ u_{1,1} & 0 & 0 & -u_{1,1} & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & u_{3,3} & 0 & 0 & -u_{3,3} \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ \vdots \\ -1 \\ 0 \\ \vdots \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ \vdots \\ -1 \\ 0 \\ \vdots \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Solution

$$\begin{bmatrix} o_p & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ u_{1,1} & 0 & 0 & -u_{1,1} & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & u_{3,3} & 0 & 0 & -u_{3,3} \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ \vdots \\ -1 \\ 0 \\ \vdots \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

1 dof remains, because of (kind of) **scaling**

Algorithm overview

```

  graph LR
    Input[Input image] --> PD[Pattern detection]
    PD --> CD[Curve detection]
    CD --> ID[Intersection detection]
    ID --> S1[Solution of plane with 1DOF]
    S1 --> EID[Elimination of 1DOF]
    EID --> Output[Shape from coplanarity]
  
```

Algorithm overview

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  graph LR
    Input[Input image] --> PD[Pattern detection]
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    S1 --> EID[Elimination of 1DOF]
    EID --> Output[Shape from coplanarity]
    Output --> DR[Dense reconstruction By light sectioning method]
    
    subgraph External
      GIP[Geometric information from Pro-cams params.] --> CPI[Calibrated plane information]
      CPI --> M[matching]
      M --> S1
    end
  
```

Elimination of 1DOF

Matching between **undetermined plane set** represented by 1 parameter and **calibrated plane set**

Energy function

Angle differences

$$E(s) = \sum_{k=1}^n \min_{l=1,2,3} \{D(\mathbf{h}_l(s), \mathbf{H}_k)\}$$

$D(\mathbf{a}, \mathbf{b})$: Angle between a and b

$$\hat{s} = \arg \min_s E(s)$$

Energy function

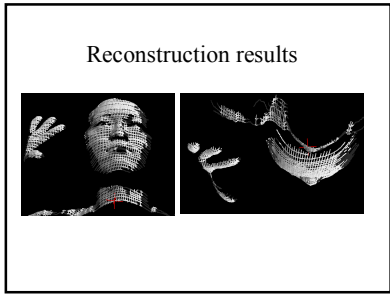
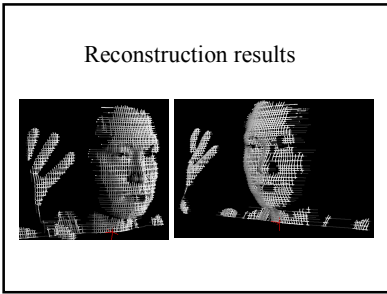
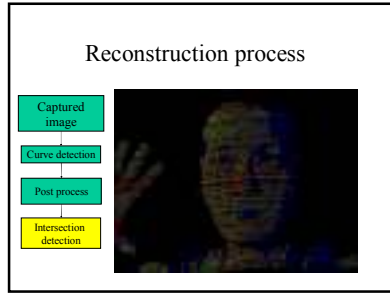
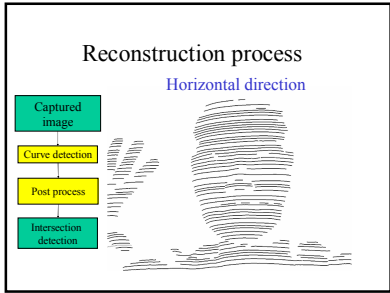
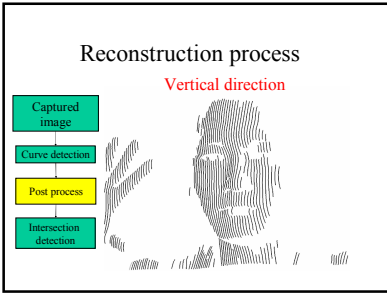
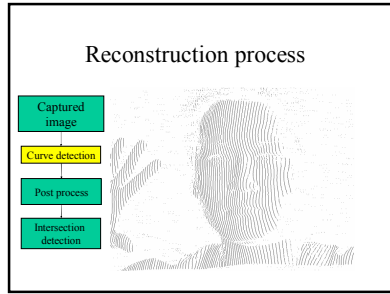
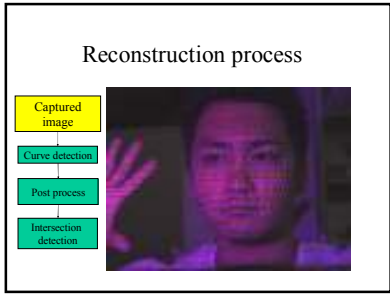
Exact match!

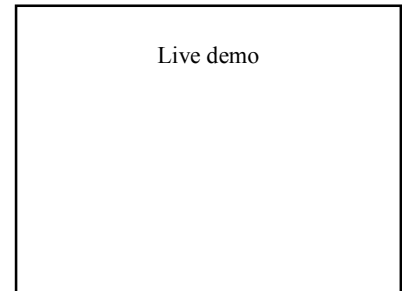
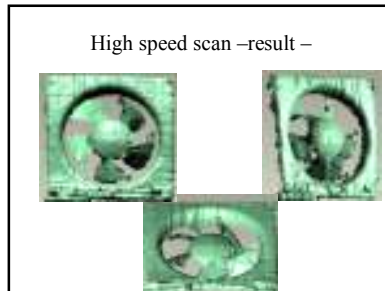
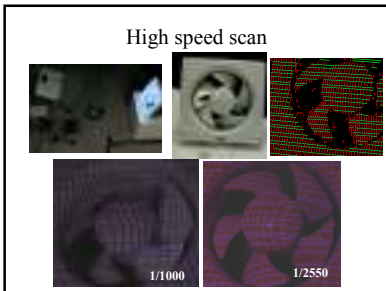
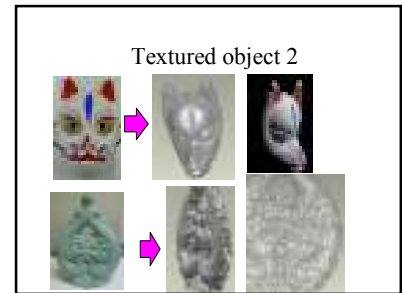
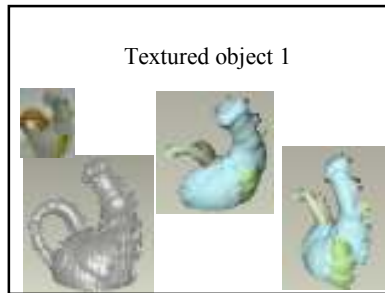
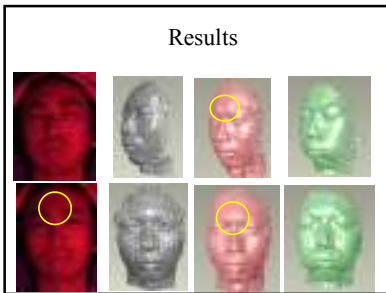
$$E(s) = \sum_{k=1}^n \min_{l=1,2,3} \{D(\mathbf{h}_l(s), \mathbf{H}_k)\}$$

$D(\mathbf{a}, \mathbf{b})$: Angle between a and b

$$\hat{s} = \arg \min_s E(s)$$

Experiments





- ### What's next?
- Infrared projector for VR/MR systems
 - Recognition human behavior by 3D video
 - Ultra high-speed 3D scan

- ### Thanks.
- Any questions or comments?
 - Web sites:
 - <http://www.cgv.ics.saitama-u.ac.jp>
 - e-mail:
 - kawasaki@cgv.ics.saitama-u.ac.jp



Research and project matrix

	Acquisition		Modeling		Rendering
	Geometry	Photometry	Geometry	Photometry	
City modeling (ITS)	△	⊙		△	⊙
Active 3D scanner	⊙		⊙		
Texture analysis		⊙		⊙	⊙
Digital archiving	⊙	△	⊙	△	

Texture acquisition

- Surface reflectance of the object
 - Light dependency $L(\theta, \phi)$
 - Diffuse
 - Specularity
 - View dependency $V(\theta, \phi)$
 - Velvet
 - Fur

4D function ($L(\theta, \phi), V(\theta, \phi)$)
 → Bidirectional Reflectance Distribution Function (BRDF)

Texture acquisition device

- Single camera (1D)
- Light stage v.1-v.3 (2D)
 - [Devebec SIGGRAPH01]
- Polynomial texture map (4D)
 - [Malzender SIGGRAPH01]
- University of Bonn[EG04] (4D)
- CURET [Columbia Univ.] (4D)

Assume plane (2D) geometry

Surface Reflectance Sampler

[Aritaki, Kawasaki et.al. ICIP01]

- Full automatic 4D data capturing machine
 - Japan Patent Office #3601031

Surface Reflectance Sampler

- Texture of 3D object can be acquired

Rendering system

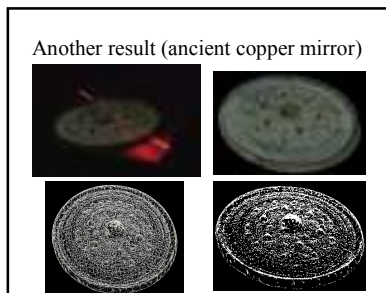
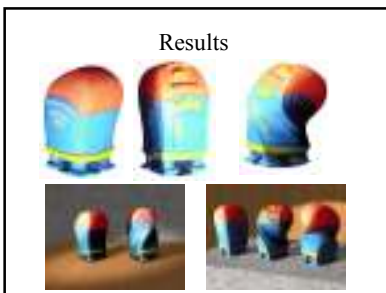
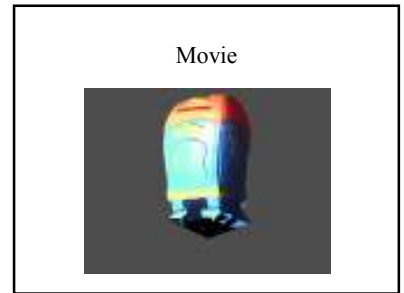
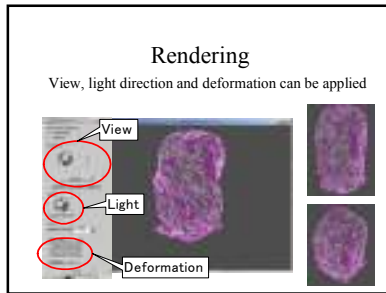
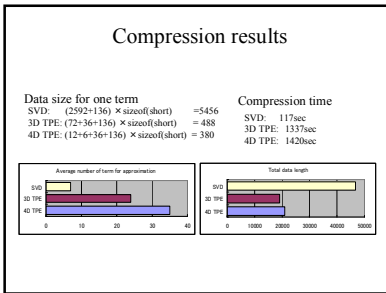
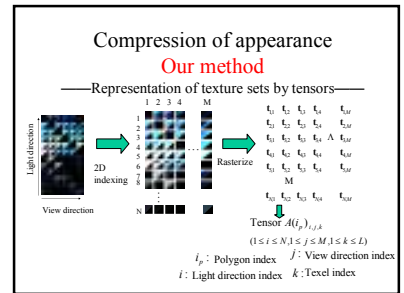
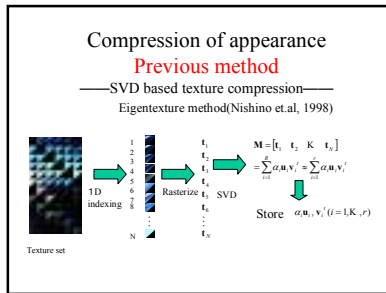
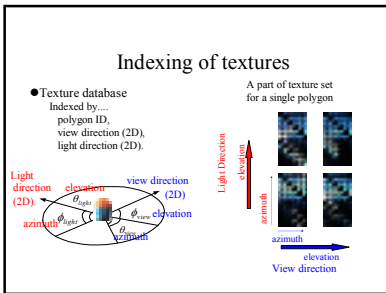
[Furukawa, Kawasaki et.al. EGSR03]

Capturing of real image samples

Restoration of geometry(Voxel Curving)

Acquisition of textures

Textures are obtained by projecting each polygon into calibrated sample images. Texture can be specified by view/light directions.



Real-time rendering results

- Using pixel and vertex shader

Discussion

- BTF – efficiently capture, compress and render 3D objects
- × **Intricate shaped** object is still difficult to render

Intricate Shaped Objects

complicated surface reflectance minute geometry

captured by @VespaPuff Academy

Solution → MBR + IBR

1. Rough surface model
2. Resampling
3. Multiresolution
4. A set of microfacets

Put view dependent texture on microfacets

Billboarding Microfacets

viewpoint voxel microfacet billboarding

Selection of Texture Images

$\theta_1 < \theta_2$ texture1 texture2 $\theta_1 > \theta_2$

microfacet

Blending of Texture Images

texture1 texture2 microfacet

$$\frac{\alpha_1}{\alpha_1 + \alpha_2} \text{texture1} + \frac{\alpha_2}{\alpha_1 + \alpha_2} \text{texture2}$$

Alpha Estimation

- Alpha value is used only before rendering.

Final Synthesis

Previous method Proposed method



Thanks.

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- Web sites:
 - <http://www.cgvis.ics.saitama-u.ac.jp>
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 - kawasaki@cgvis.ics.saitama-u.ac.jp