

Introducción a la Fotografía 3D

UBA/FCEN Marzo 27 – Abril 12 2013

Clase 6 : Martes Abril 9

Gabriel Taubin

Brown University



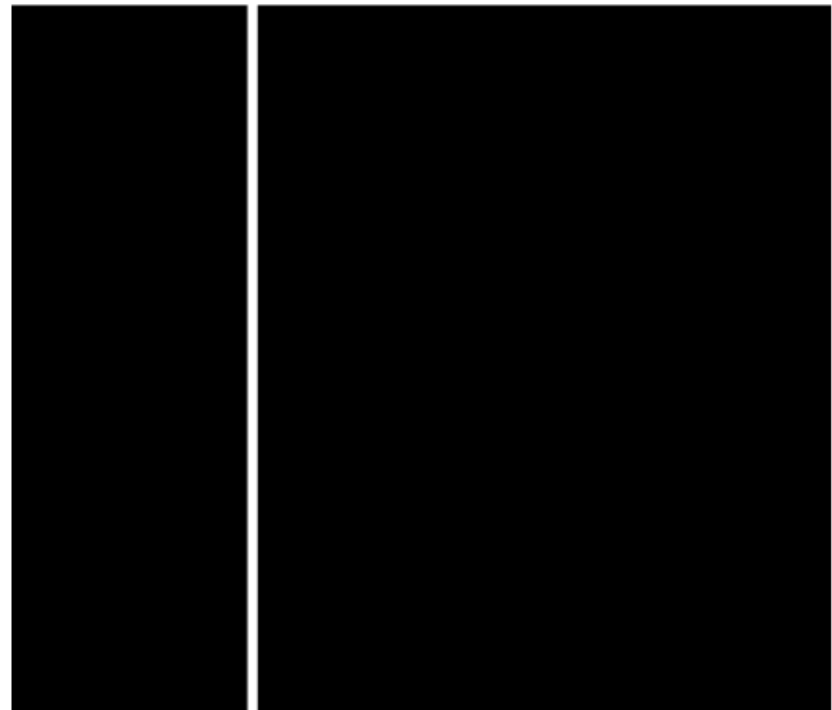
BROWN

Course Schedule

➤ *Structured Lighting*

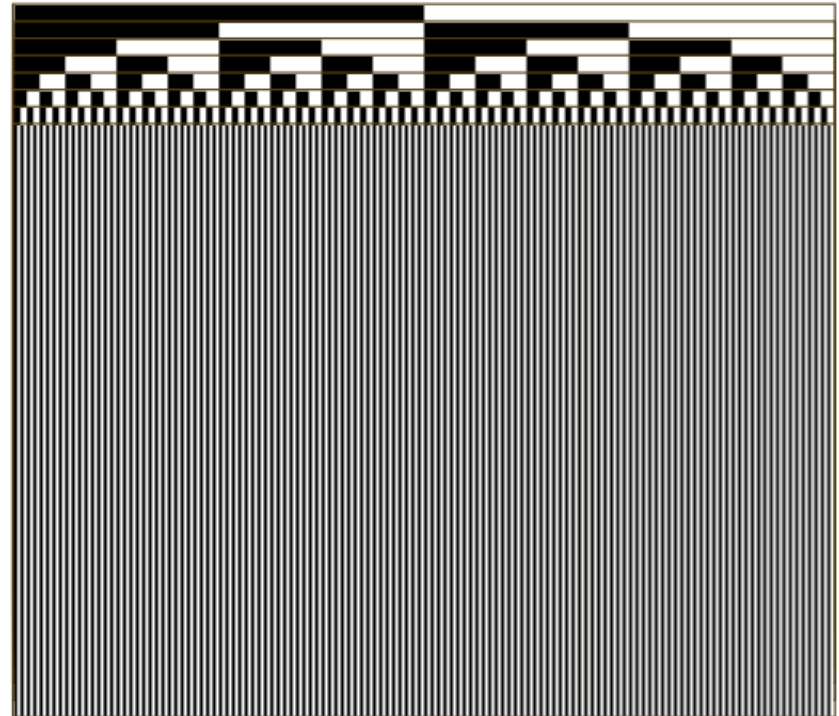
- Projector Calibration and Structured Light Reconstruction

Structured Lighting: Swept-Planes Revisited



- Swept-plane scanning recovers 3D depth using ray-plane intersection
- Use a data projector to replace manually-swept laser/shadow planes
- How to assign correspondence from projector planes to camera pixels?
- Solution: Project a spatially- and temporally-encoded image sequence
- **What is the optimal image sequence to project?**

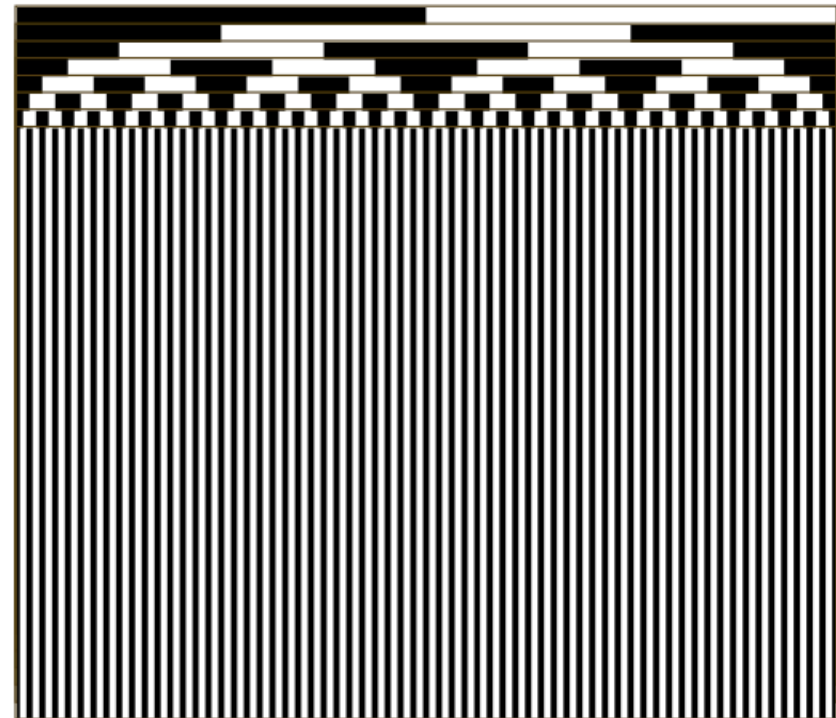
Structured Lighting: Binary Codes



Binary Image Sequence [Posdamer and Altschuler 1982]

- Each image is a bit-plane of the binary code for projector row/column
- Minimum of 10 images to encode 1024 columns or 768 rows
- In practice, 20 images are used to encode 1024 columns or 768 rows
- Projector and camera(s) must be synchronized

Structured Lighting: Gray Codes



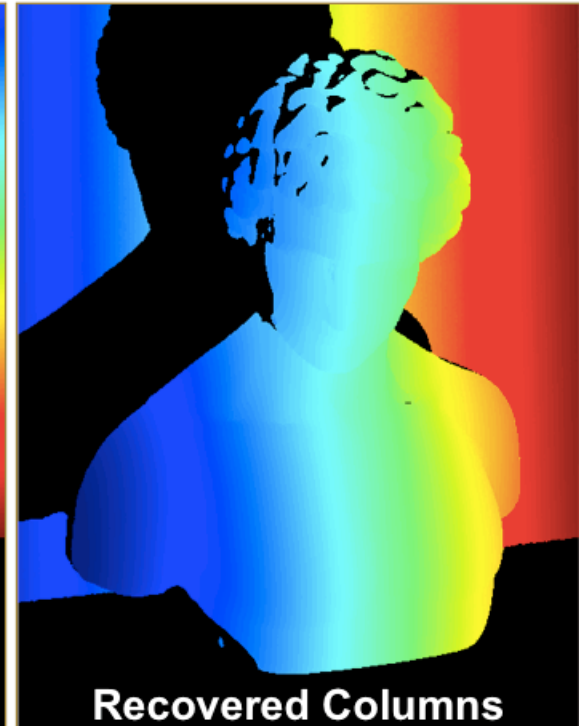
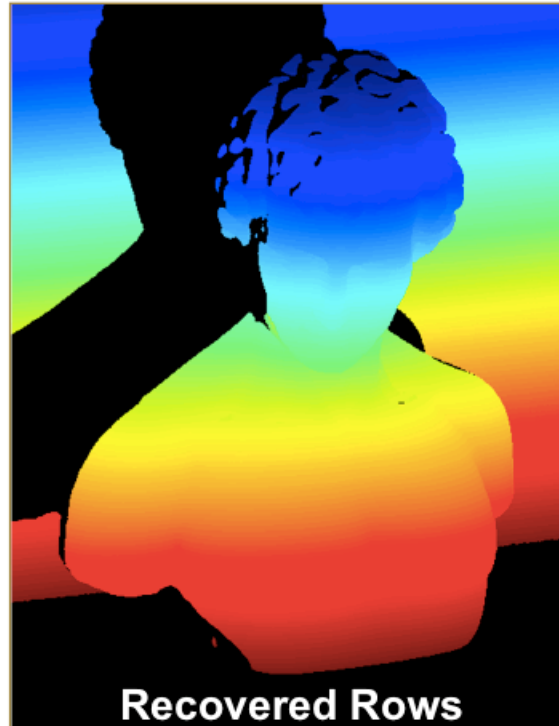
Gray Code Image Sequence [Inokuchi 1984]

- Each image is a bit-plane of the Gray code for each projector row/column
- Requires same number of images as a binary image sequence, but has better performance in practice

Bin2Gray(B,G)

```
1  G ← B
2  for i ← n-1 downto 0
3    G[i] ← B[i+1] xor B[i]
```

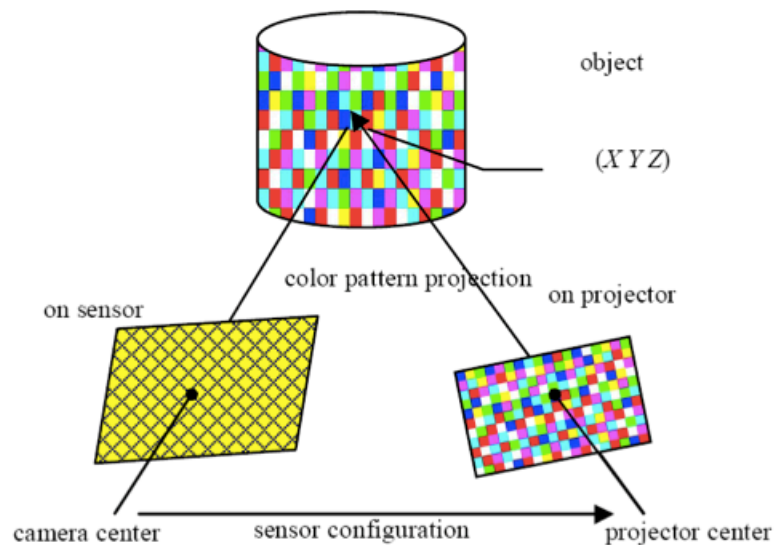
Gray Codes: Decoding Performance



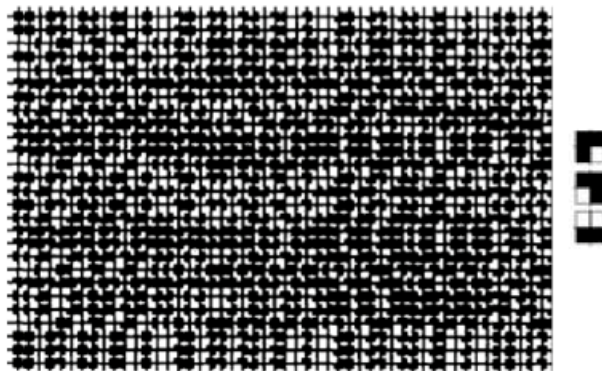
3D Reconstruction using Structured Light [Inokuchi 1984]

- Implemented using a total of 42 images (2 to measure dynamic range, 20 to encode rows, 20 to encode columns)
- Individual bits assigned by detecting if bit-plane (or its inverse) is brighter
- Decoding algorithm: Gray code \rightarrow binary code \rightarrow integer row/column index

Additional Structured Lighting Patterns



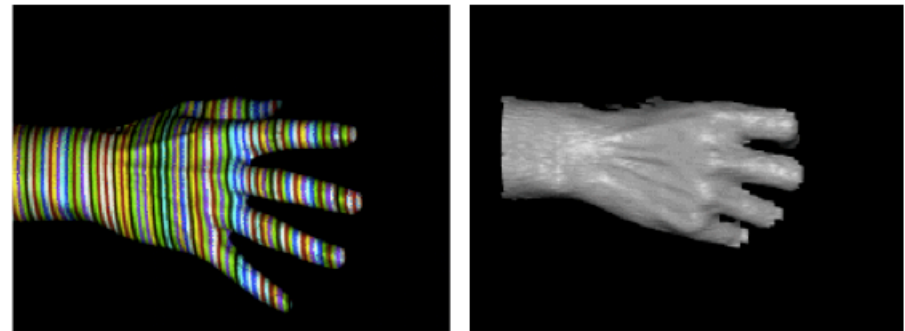
Spatial encoding strategies [Chen et al. 2007]



Pseudorandom and M-arrays [Griffin 1992]



“Single-shot” patterns (N-arrays, grids, random, etc.)



De Bruijn sequences [Zhang et al. 2002]



Phase-shifting [Zhang et al. 2004]

Assembling Your Own Scanner



Parts List [\$600-\$800, but only \$300 without mounting hardware]

- [1] Camera (Logitech QuickCam Pro 9000) [\$70]
- [1] Projector (InFocus LP 70+) [\$200-\$400 from eBay]
- [1] Manfrotto 190XB 3 Section Aluminum Tripod [\$130]
- [1] Manfrotto 486RC2 Compact Ball Head [\$75]
- [1] Custom Aluminum Rail (with 3/8 inch holes for mounting ball heads) [NA]
- [1] Custom Aluminum Tripod Adaptor Plate (for mounting projector to ball head) [NA]
- [2] Manfrotto 484 Mini Ball Heads [\$45 each]
- [1] Sunpak 620-500C Versipod (with adjustable, locking ball head) [\$15]
- [1] ~11x17 inch foamcore board (with an affixed calibration chessboard) [\$10]

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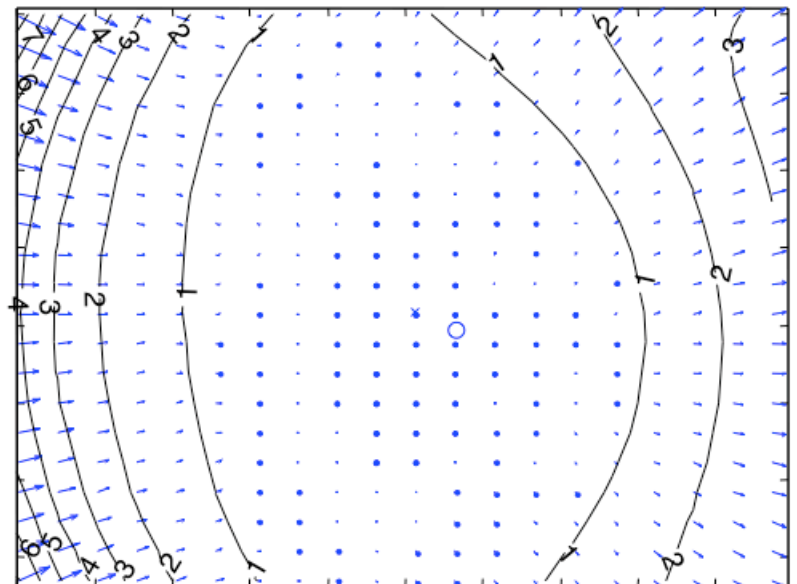
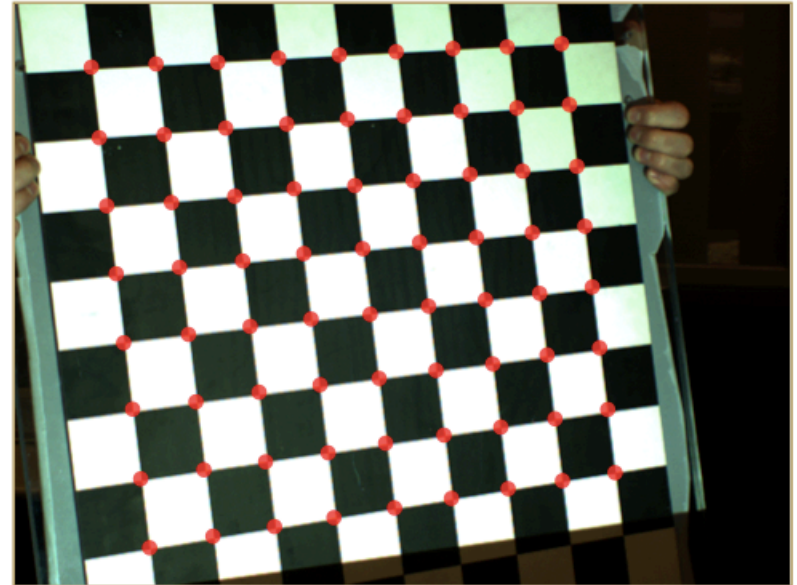
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Developing Your Own Software

Key Functions

- Camera and projector calibration



Developing Your Own Software

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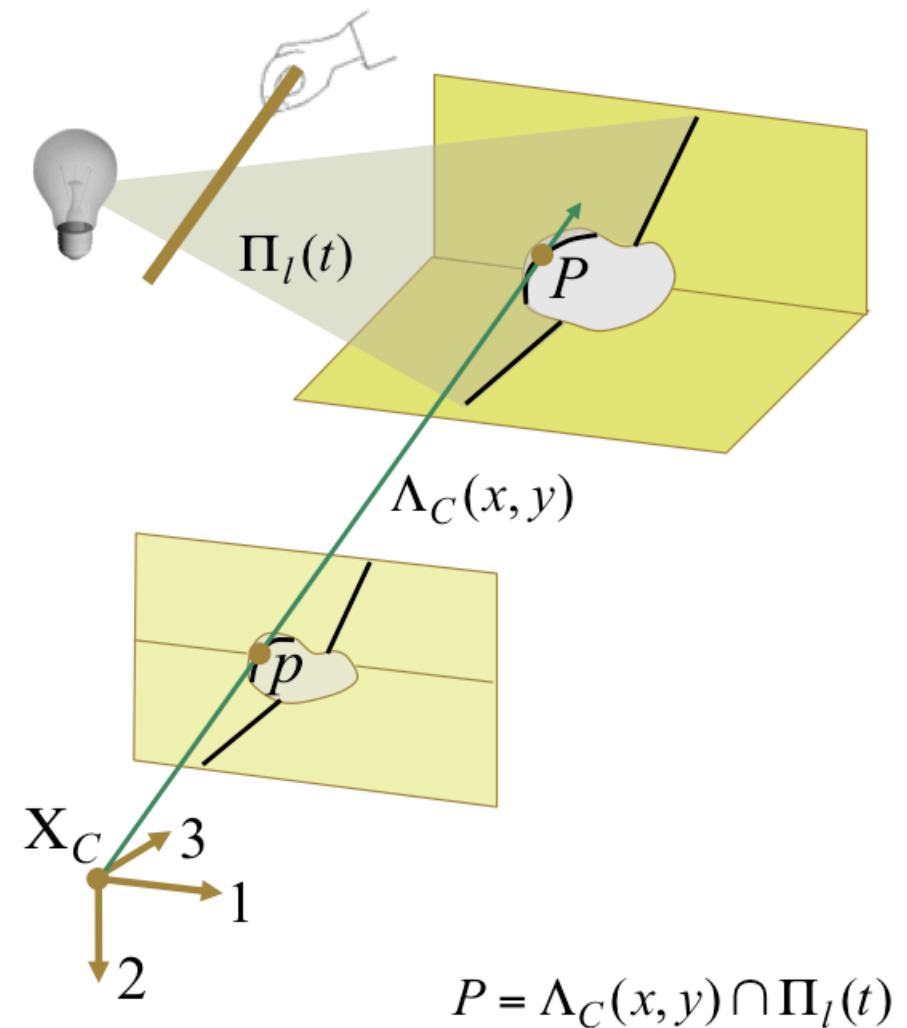
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Developing Your Own Software

Key Functions

- Camera and projector calibration
- Camera and projector interfaces
- Geometric operations



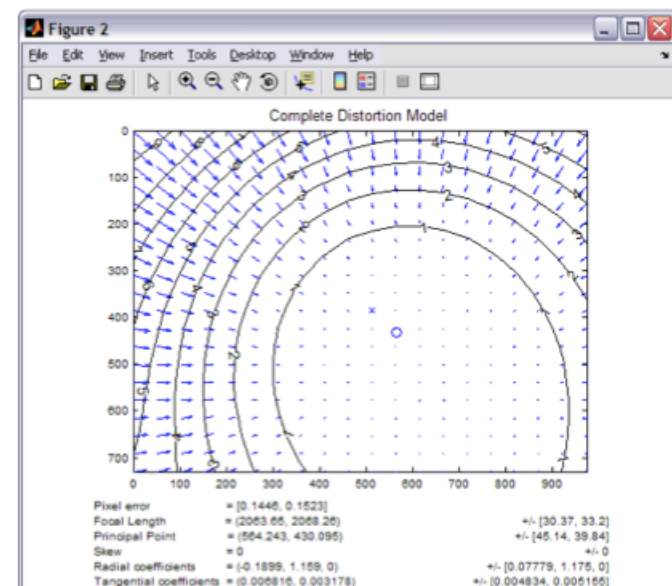
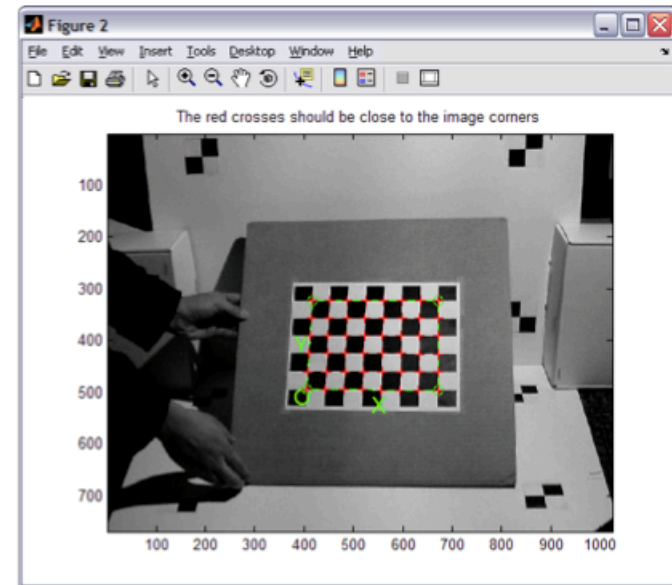
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Matlab

- Camera Calibration Toolbox
(updated to support projector calibration)



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- Projector-Camera Calibration Toolbox
(just released, supports full calibration)



<http://code.google.com/p/procamcalib/>

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- Image Acquisition Toolbox, etc.
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Developing Your Own Software

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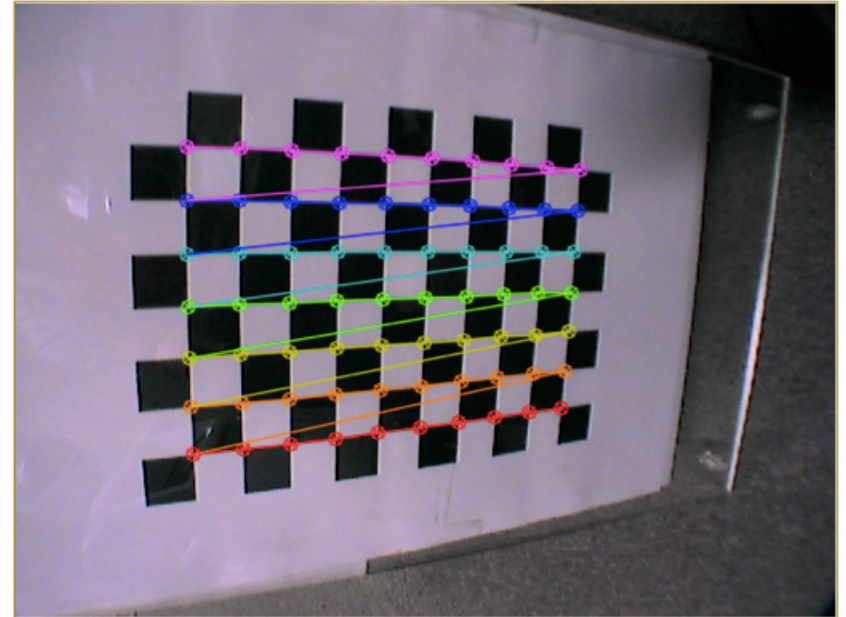
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OpenCV

- Supports camera calibration
- Updated to support projector calibration
(course source)



<http://opencv.willowgarage.com/wiki/>

Controlling a Projector for Structured Lighting

Operating System

- A projector is just another display...



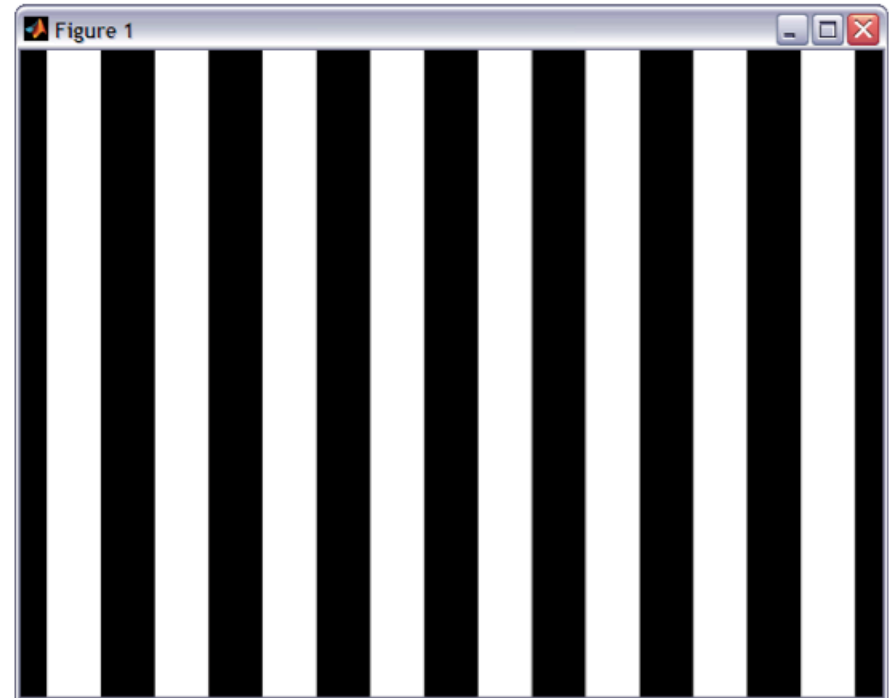
Controlling a Projector for Structured Lighting

Operating System

- A projector is just another display...

Matlab

- Fullscreen display not fully supported (cannot remove window decorations)



```
imagesc(P{1}(:, :, 5));  
axis image off;  
set(gca, 'Pos', [0 0 1 1]);  
set(gcf, 'MenuBar', 'none');  
set(gcf, 'Pos', [-1023 283 1024 768]);
```

Controlling a Projector for Structured Lighting

Operating System

- A projector is just another display...

Matlab

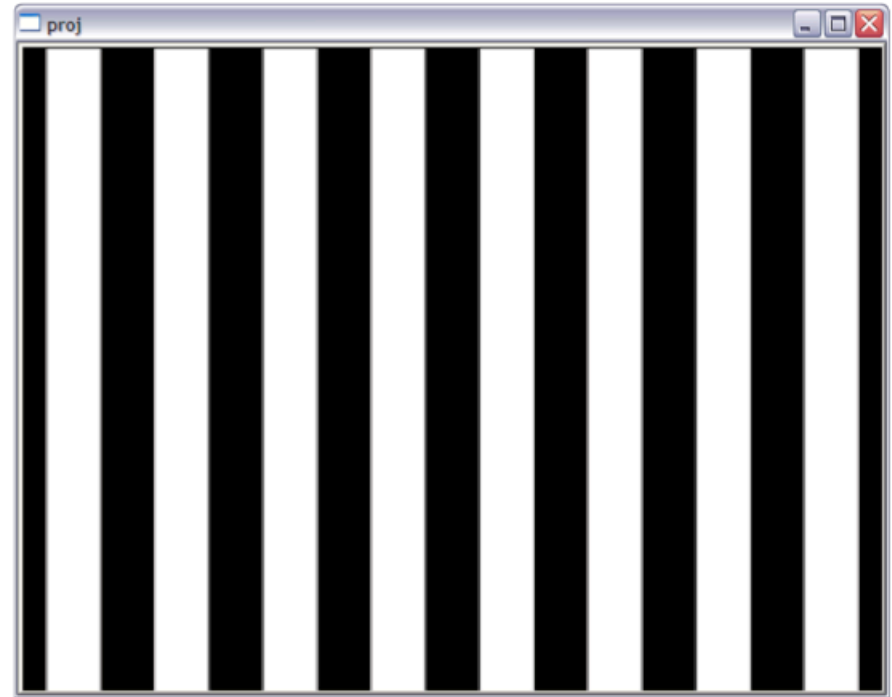
- Fullscreen display not fully supported (cannot remove window decorations)
- Use Java to control displays (slow, multiple display problems)
- Use the *Psychophysics Toolbox* (*Screen.m* wraps OpenGL functions)

OpenCV

- Cannot remove window decorations
- *cvNamedWindow/cvMoveWindow*

General

- Use OpenGL wrappers



```
cvNamedWindow("proj",CV_WINDOW_AUTOSIZE);  
IplImage* proj_frame =  
    cvCreateImage(cvSize(w,h),8,3);  
cvSet(proj_frame,cvScalar(0,0,0));  
cvShowImage("proj",proj_frame);  
cvMoveWindow("proj",-w-7,-33);  
cvWaitKey(1);
```

<http://opencv.willowgarage.com/wiki>

Course Schedule

- Structured Lighting
- ***Robust Pixel Classification***
- Projector Calibration / Structured Light Reconstruction
- Surface Reconstruction from Point Clouds
- Elementary Mesh Processing

Robust Pixel Classification



- **Robust pixel classification for 3d modeling with structured light**, by Yi Xu and Daniel G. Aliaga, in Proceedings of Graphics Interface 2007, GI '07, pages 233–240, New York, NY, USA, 2007. ACM.
- **Fast separation of direct and global components of a scene using high frequency illumination**, by Shree K. Nayar, Gurunandan Krishnan, Michael D. Grossberg, and Ramesh Raskar, in ACM Trans. Graph., 25(3):935–944, July 2006.

Robust Pixel Classification

- The intensity of a pixel can be decomposed into the direct component and indirect (or global) component.
- The direct component is due to light bouncing off the surface in a single reflection.
- The indirect component is due to multiple reflections (e.g. inter-reflections, subsurface, scattering etc.).

$$L(p) = \begin{cases} L_d(p) + \alpha L_g(p) & \text{if } p \text{ is on} \\ (1 - \alpha)L_g(p) & \text{if } p \text{ is off,} \end{cases}$$

- where $0 \leq \alpha \leq 1$ is a fraction of activated source pixels. For example, for a random binary pattern $\alpha = 1$, and for a white pattern $\alpha = 1$.

Robust Pixel Classification

- Estimate direct and global components for each pixel using the method by Nayar
- Use two images of the scene, one with the scene lit with high-frequency illumination, and the other lit with the complementary illumination.

$$L_d(p) = L_{max}(p) - \frac{\alpha}{1 - \alpha} L_{min}(p)$$

$$L_g(p) = \frac{1}{1 - \alpha} L_{min}(p),$$

Robust Pixel Classification

- Projecting the code pattern i and its inverse yields two values for each pixel.
- Classify p according to the rules

$$p \text{ is } \begin{cases} \text{uncertain} & \text{if } L_d(p) < m \\ \text{on} & \text{if } L_d(p) > L_g(p) \text{ and } L_i^+(p) > L_i^-(p) \\ \text{off} & \text{if } L_d(p) > L_g(p) \text{ and } L_i^+(p) < L_i^-(p) \\ \text{off} & \text{if } L_i^+(p) < L_d(p) \text{ and } L_i^-(p) > L_g(p) \\ \text{on} & \text{if } L_i^+(p) > L_g(p) \text{ and } L_i^-(p) < L_d(p) \\ \text{uncertain} & \text{otherwise,} \end{cases}$$