

Announcements

- Project #3 is extend to 5/16
- Project #2 artifact voting results

Image-based modeling

Digital Visual Effects, Spring 2008

Yung-Yu Chuang

2008/5/6

with slides by Richard Szeliski, Steve Seitz and Alexei Efros

Honorable mention (7): 張子捷 游知浩



Honorable mention (9): 胡傳牲 高紹航



Third place (10): 吳懿倫 張哲瀚

DigiVFX



Second place (13): 賴韻芊 黃柏瑜

DigiVFX



First place (24): 游名揚 曾照涵

DigiVFX



Outline

DigiVFX

- Models from multiple (sparse) images
 - Structure from motion
 - Facade
- Models from single images
 - Tour into pictures
 - Single view metrology
 - Other approaches

Models from multiple images (Façade, Debevec *et. al.* 1996)

Facade

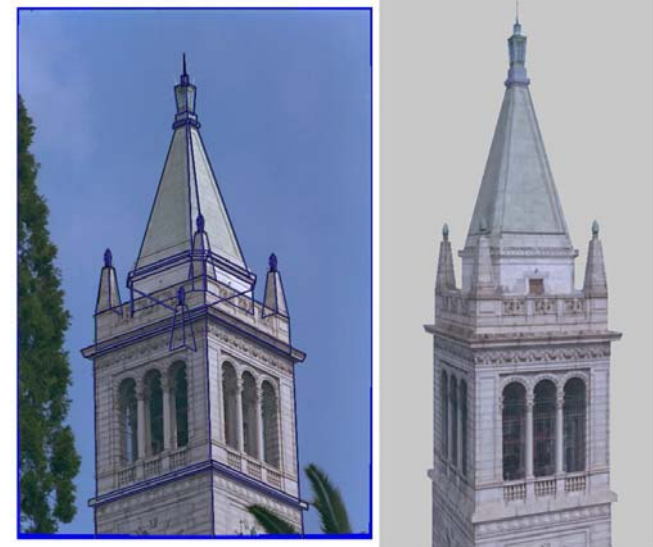
- Use a sparse set of images
- Calibrated camera (intrinsic only)
- Designed specifically for modeling architecture
- Use a set of blocks to approximate architecture

- Three components:
 - geometry reconstruction
 - texture mapping
 - model refinement

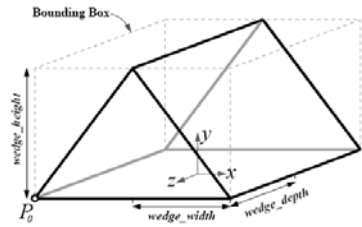
Idea



Idea

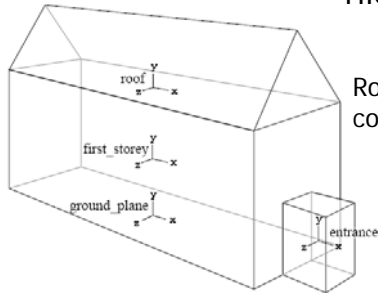


Geometric modeling

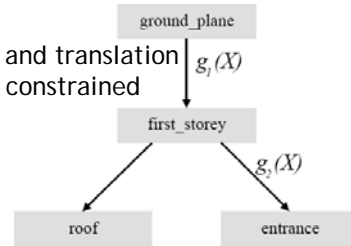


A block is a geometric primitive with a small set of parameters

Hierarchical modeling for a scene



Rotation and translation could be constrained



Reasons for block modeling

- Architectural scenes are well modeled by geometric primitives.
- Blocks provide a high level abstraction, easier to manage and add constraints.
- No need to infer surfaces from discrete features; blocks essentially provide prior models for architectures.
- Hierarchical block modeling effectively reduces the number of parameters for robustness and efficiency.

Reconstruction

minimize $\mathcal{O} = \sum Err_i$

$\mathbf{m} = R_j(\mathbf{v} \times (\mathbf{d} - \mathbf{t}_j))$
 $m_x x + m_y y - m_z f = 0$

Observed edge segment

predicted line:
 $m_x x + m_y y + m_z = 0$

$Err_i = \int_0^l h^2(s) ds$

Reconstruction

$Err_i = \int_0^l h^2(s) ds$

Observed edge segment

predicted line:
 $m_x x + m_y y + m_z = 0$

$h_1 = \frac{m_x x_1 + m_y y_1 + m_z}{\sqrt{m_x^2 + m_y^2}}$

$h_2 = \frac{m_x x_2 + m_y y_2 + m_z}{\sqrt{m_x^2 + m_y^2}}$

$h(s) = h_1 + s \frac{h_2 - h_1}{l}$

$Err_i = \int_0^l h^2(s) ds$
 $= \frac{l}{3} (h_1^2 + h_1 h_2 + h_2^2)$

Reconstruction

$$Err_i = \int_0^l h^2(s)ds = \frac{l}{3}(h_1^2+h_1h_2+h_2^2) = \mathbf{m}^T(A^TBA)\mathbf{m}$$

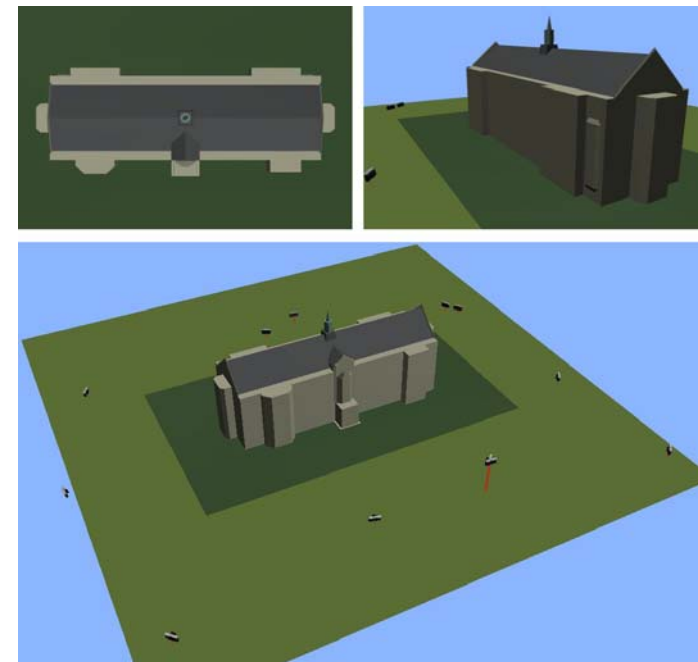
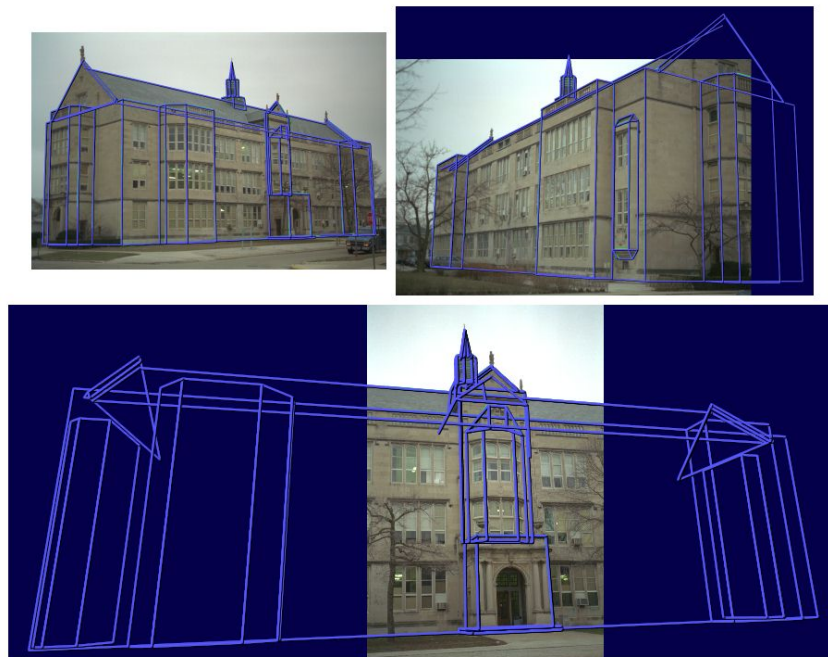
$$\mathbf{m} = (m_x, m_y, m_z)^T \quad \mathbf{m} = R_j(\mathbf{v} \times (\mathbf{d} - t_j))$$

$$A = \begin{pmatrix} x_1 & y_1 & 1 \\ x_2 & y_2 & 1 \end{pmatrix} \quad \text{nonlinear w.r.t. camera and model}$$

$$B = \frac{l}{3(m_x^2 + m_y^2)} \begin{pmatrix} 1 & 0.5 \\ 0.5 & 1 \end{pmatrix}$$

Results

3 of 12 photographs



Texture mapping

DigiVFX



Texture mapping in real world

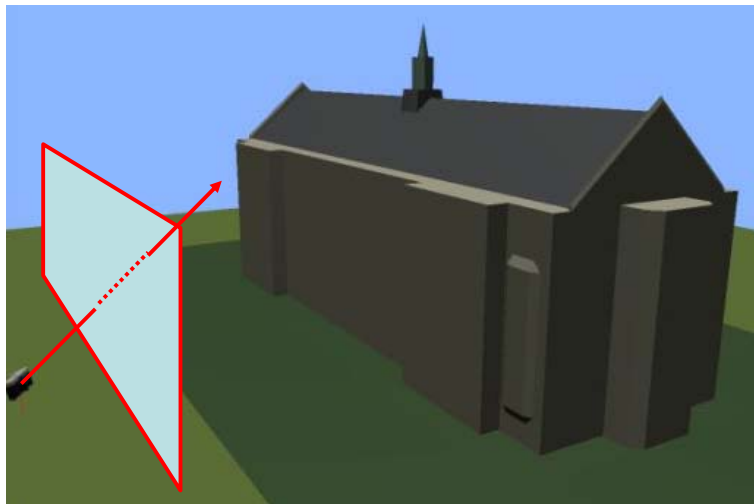
DigiVFX



[Demo movie](#)
Michael Naimark,
San Francisco Museum of
Modern Art, 1984

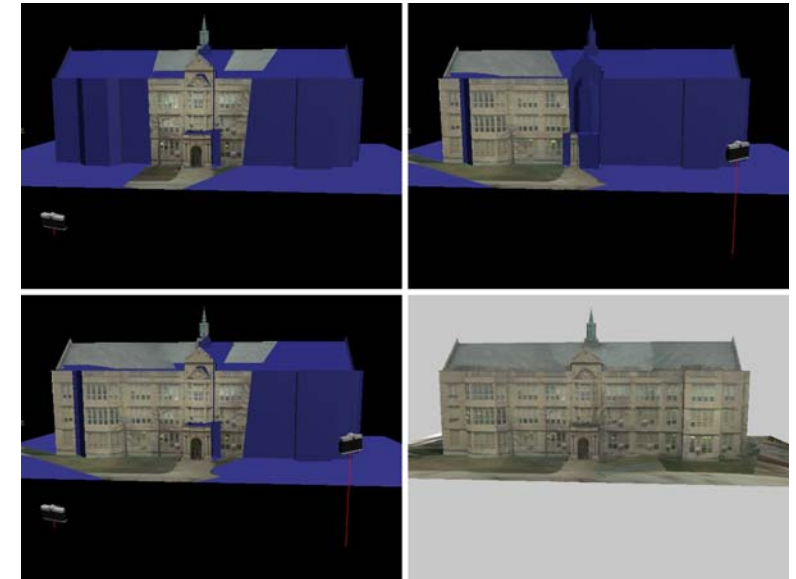
Texture mapping

DigiVFX



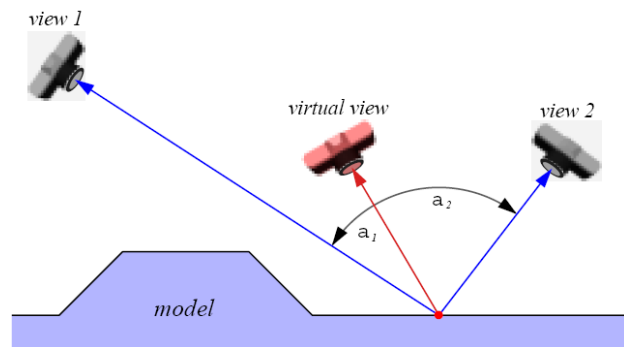
Texture mapping

DigiVFX



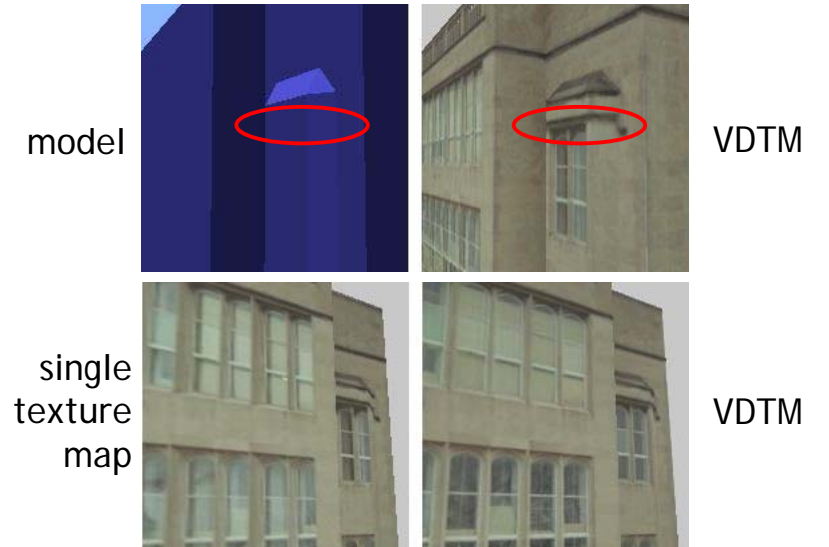
View-dependent texture mapping

DigiVFX



View-dependent texture mapping

DigiVFX



View-dependent texture mapping

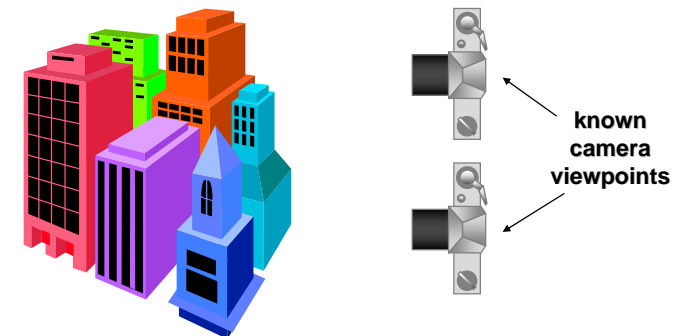
DigiVFX



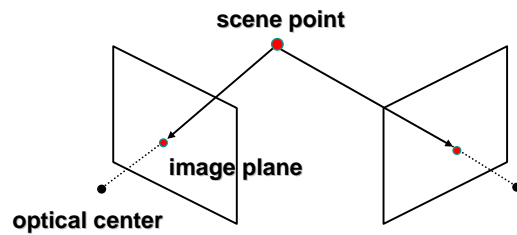
Model-based stereo

DigiVFX

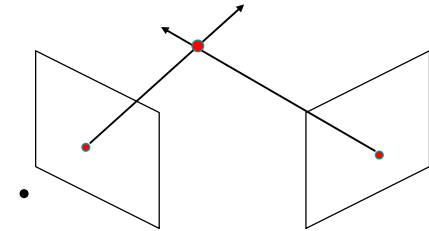
- Use stereo to refine the geometry



Stereo



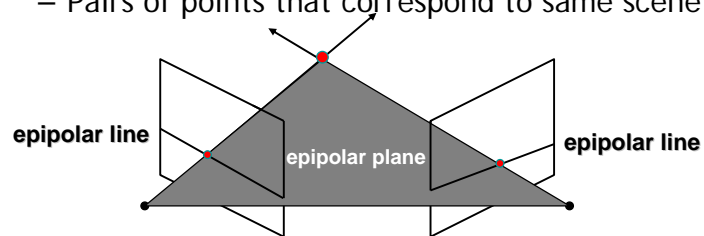
Stereo



- Basic Principle: Triangulation
 - Gives reconstruction as intersection of two rays
 - Requires
 - calibration
 - *point correspondence*

Stereo correspondence

- Determine Pixel Correspondence
 - Pairs of points that correspond to same scene point



- Epipolar Constraint
 - Reduces correspondence problem to 1D search along *conjugate epipolar lines*

Finding correspondences

- apply feature matching criterion (e.g., correlation or Lucas-Kanade) at *all* pixels simultaneously
- search only over epipolar lines (much fewer candidate positions)



Image registration (revisited)

- How do we determine correspondences?
 - *block matching* or *SSD* (sum squared differences)

$$E(x, y; d) = \sum_{(x', y') \in N(x, y)} [I_L(x' + d, y') - I_R(x', y')]^2$$

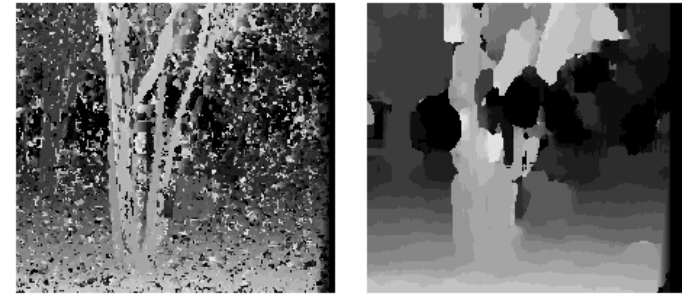
d is the *disparity* (horizontal motion)



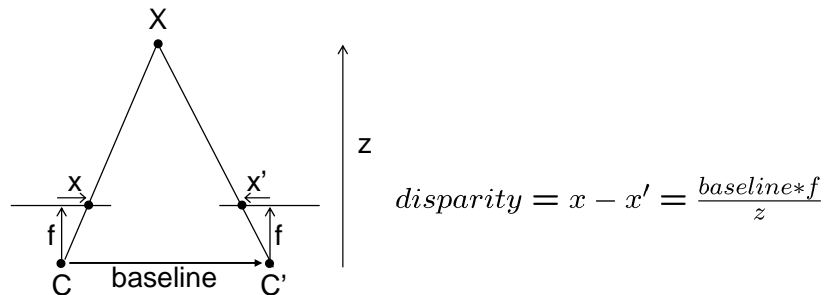
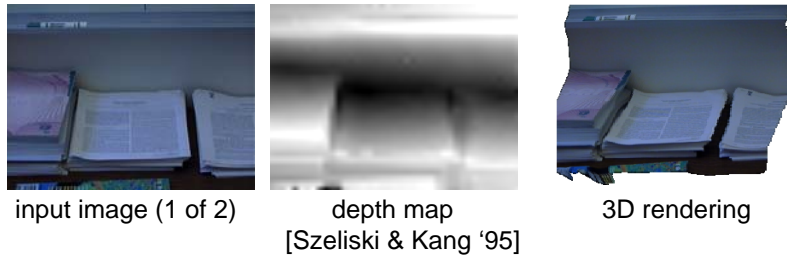
- How big should the neighborhood be?

Neighborhood size

- Smaller neighborhood: more details
- Larger neighborhood: fewer isolated mistakes



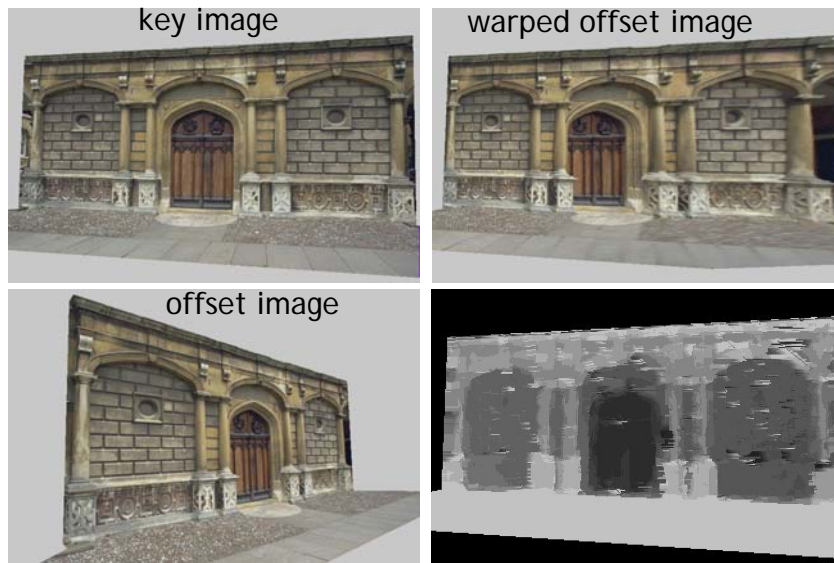
Depth from disparity



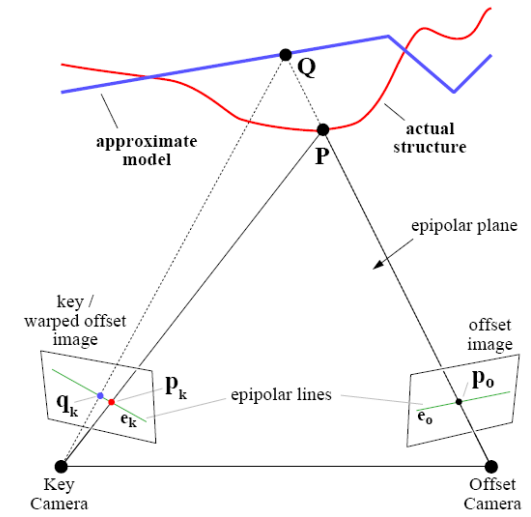
Stereo reconstruction pipeline

- Steps
 - Calibrate cameras
 - Rectify images
 - Compute disparity
 - Estimate depth
- What will cause errors?
 - Camera calibration errors
 - Poor image resolution
 - Occlusions
 - Violations of brightness constancy (specular reflections)
 - Large motions
 - Low-contrast image regions

Model-based stereo



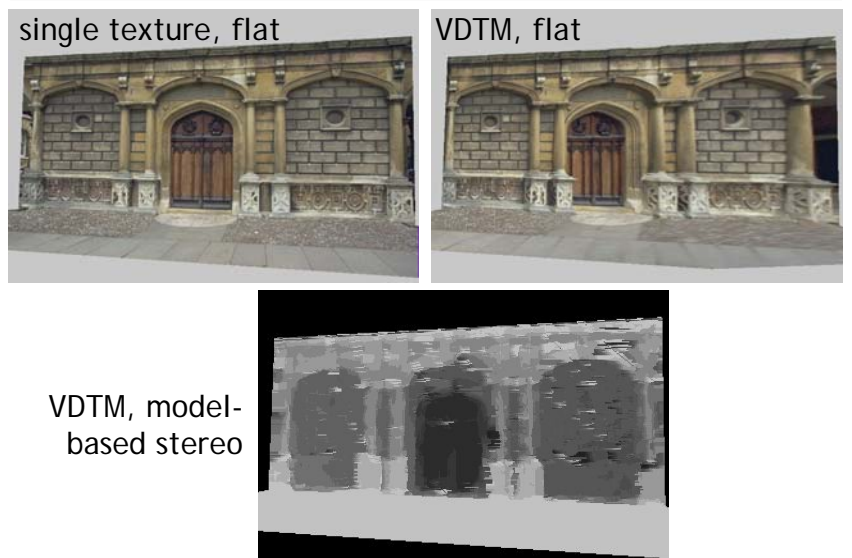
Epipolar geometry



Results

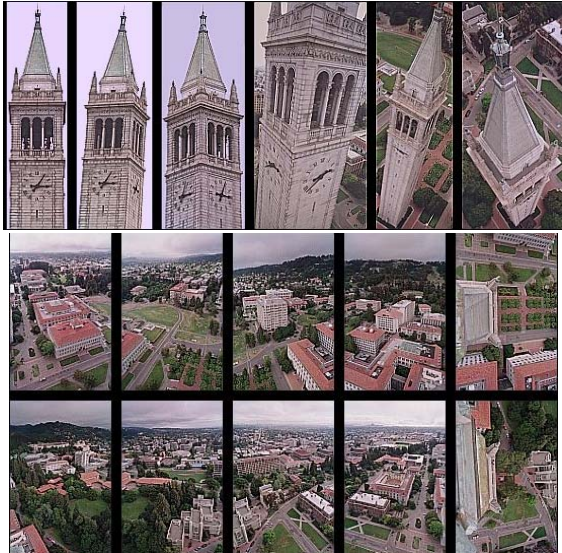


Comparisons



Final results

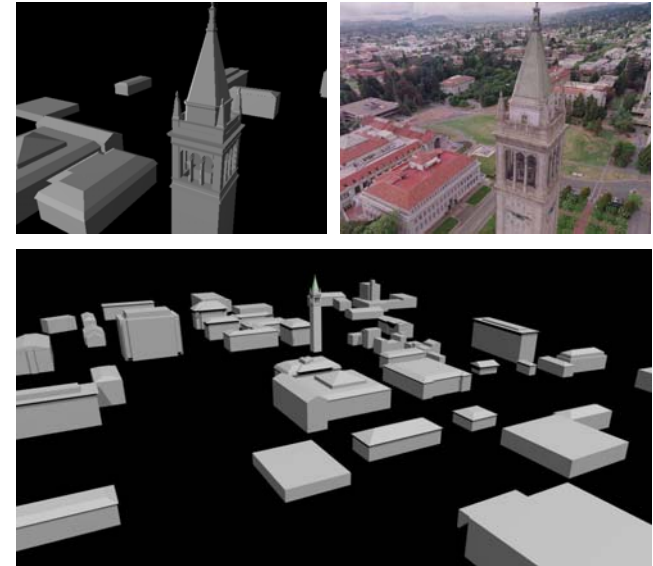
DigiVFX



[Kite photography](#)

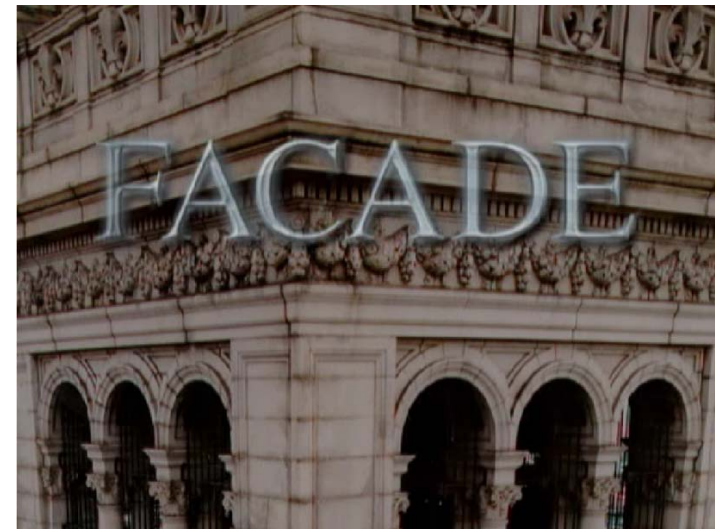
Final results

DigiVFX



Results

DigiVFX



Results

DigiVFX



Commercial packages

DigiVFX

- [REALVIZ ImageModeler](#)



The Matrix

DigiVFX

Cinefex #79, October 1999.

Since the bullet-time rig would be visible in shots featuring a 360-degree sweep of the characters, it was employed only for the shooting of the foreground subject – namely, the actors or their stunt doubles – necessitating a different approach for the backgrounds. Shot separately, the backgrounds used a virtual cinematography process that allowed a 360-degree environment to be constructed in the computer from stills taken on set. This approach for generating the backgrounds was based on the Berkeley Tower flyover, a novel image-based rendering technique presented at Siggraph '97 by George Borshukov and Paul Debevec, a researcher at UC Berkeley. The technique employed twenty stills of that town's college campus to create a virtual environment through which the camera could travel. "Instead of reinventing the background in traditional CG fashion – painting textures, shooting orthographic views of the set, and then proceeding to texture replication – we generated a completely free, high-resolution camera move that would have been impossible to achieve using traditional CG," Borshukov said, "and we did it working from just a handful of stills."

The Matrix

DigiVFX

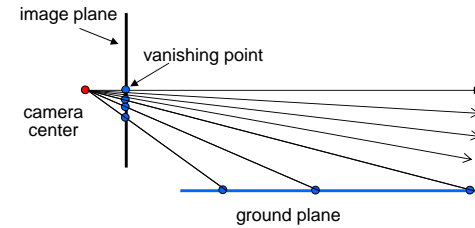
- *Academy Awards for Scientific and Technical achievement for 2000*
To George Borshukov, Kim Libreri and Dan Piponi for the development of a system for image-based rendering allowing choreographed camera movements through computer graphic reconstructed sets.

This was used in The Matrix and Mission Impossible II; See The Matrix Disc #2 for more details



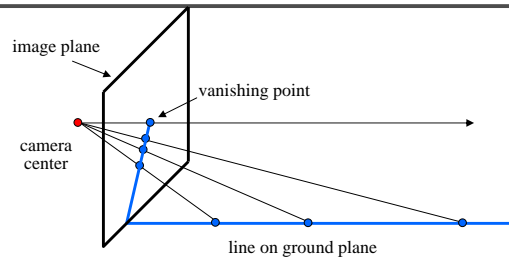
Models from single images

Vanishing points

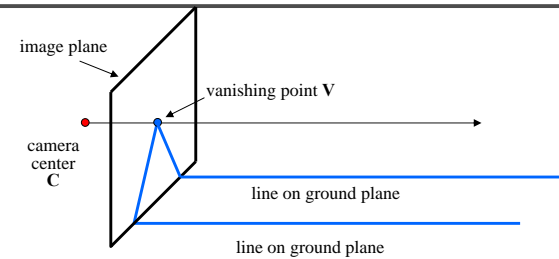


- Vanishing point
 - projection of a point at infinity

Vanishing points (2D)

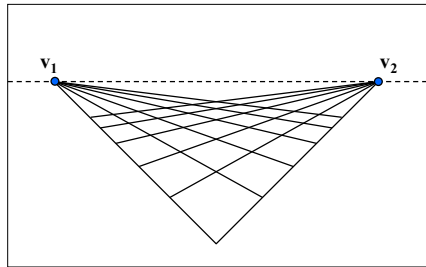


Vanishing points



- Properties
 - Any two parallel lines have the same vanishing point v
 - The ray from C through v is parallel to the lines
 - An image may have more than one vanishing point

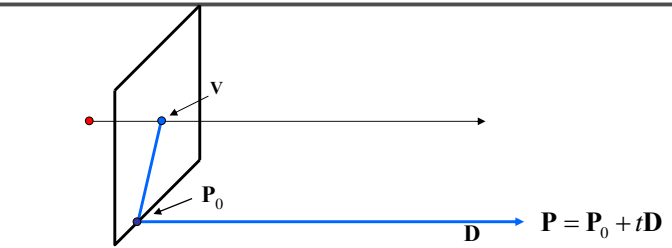
Vanishing lines



• Multiple Vanishing Points

- Any set of parallel lines on the plane define a vanishing point
- The union of all of these vanishing points is the *horizon line*
 - also called *vanishing line*
- Note that different planes define different vanishing lines

Computing vanishing points



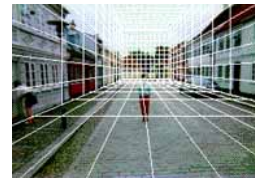
$$P_t = \begin{bmatrix} P_x + tD_x \\ P_y + tD_y \\ P_z + tD_z \\ 1 \end{bmatrix} \cong \begin{bmatrix} P_x / t + D_x \\ P_y / t + D_y \\ P_z / t + D_z \\ 1/t \end{bmatrix} \quad t \rightarrow \infty \quad P_\infty \cong \begin{bmatrix} D_x \\ D_y \\ D_z \\ 0 \end{bmatrix}$$

• Properties $v = \Pi P_\infty$

- P_∞ is a point at *infinity*, v is its projection
- They depend only on line *direction*
- Parallel lines $P_0 + tD$, $P_1 + tD$ intersect at P_∞

Tour into pictures

- Create a 3D “theatre stage” of five billboards



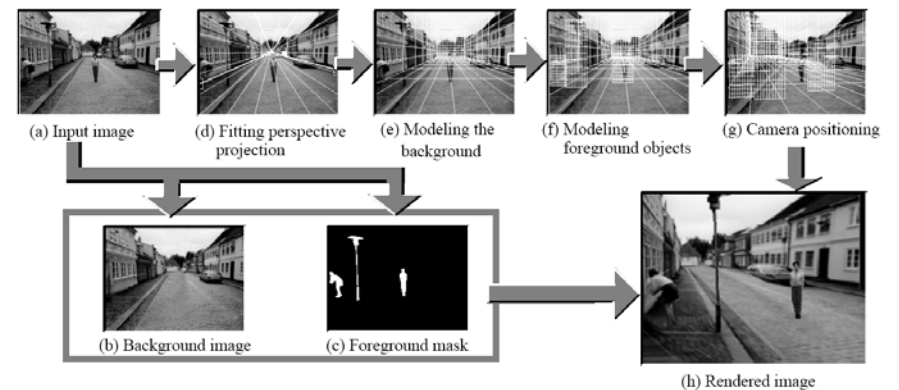
- Specify foreground objects through bounding polygons



- Use camera transformations to navigate through the scene



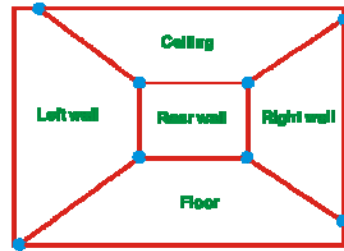
Tour into pictures



The idea

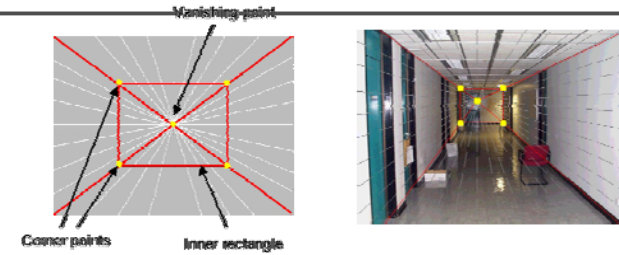
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- Many scenes (especially paintings), can be represented as an axis-aligned box volume (i.e. a stage)
- Key assumptions:
 - All walls of volume are orthogonal
 - Camera view plane is parallel to back of volume
 - Camera up is normal to volume bottom
 - Volume bottom is $y=0$
- Can use the vanishing point to fit the box to the particular Scene!



Fitting the box volume

DigiVFX



- User controls the inner box and the vanishing point placement (6 DOF)

Foreground Objects

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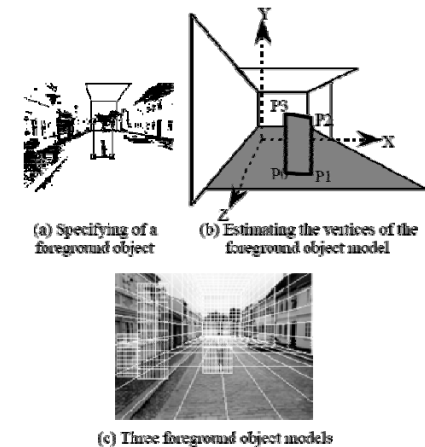
- Use separate billboard for each
- For this to work, three separate images used:
 - Original image.
 - Mask to isolate desired foreground images.
 - Background with objects removed



Foreground Objects

DigiVFX

- Add vertical rectangles for each foreground object
- Can compute 3D coordinates P_0 , P_1 since they are on known plane.
- P_2 , P_3 can be computed as before (similar triangles)



Example

DigiVFX



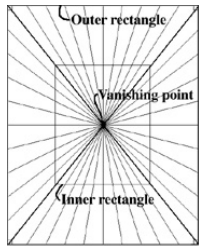
(a) Input image



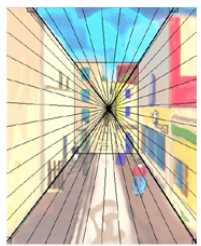
(b) Background



(c) Foreground mask



(a) Initial state

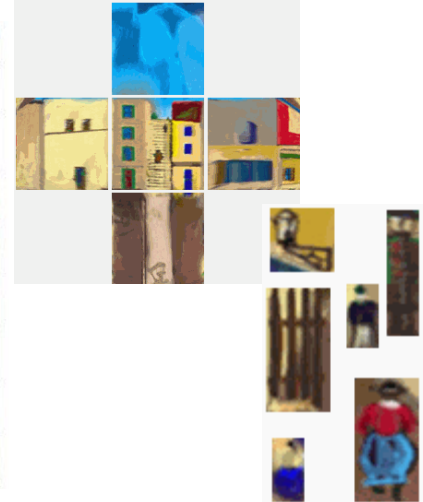
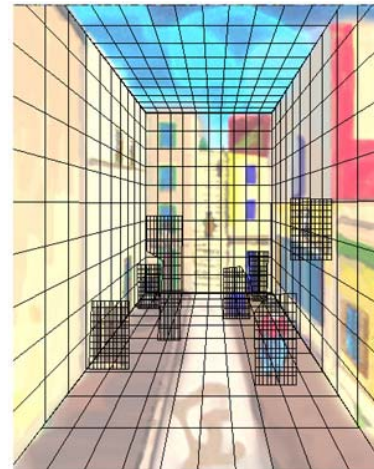


(b) Specification result



Example

DigiVFX



gITip

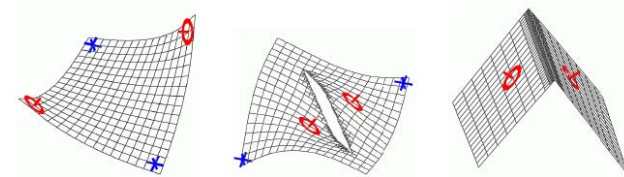
DigiVFX

- <http://www.cs.ust.hk/~cpegnel/gITIP/>



Zhang *et. al.* CVPR 2001

DigiVFX



Methods	Iteration 0	Iteration 200	Iteration 1200	Iteration 2500	Iteration 9500
No hierarchical transformation					

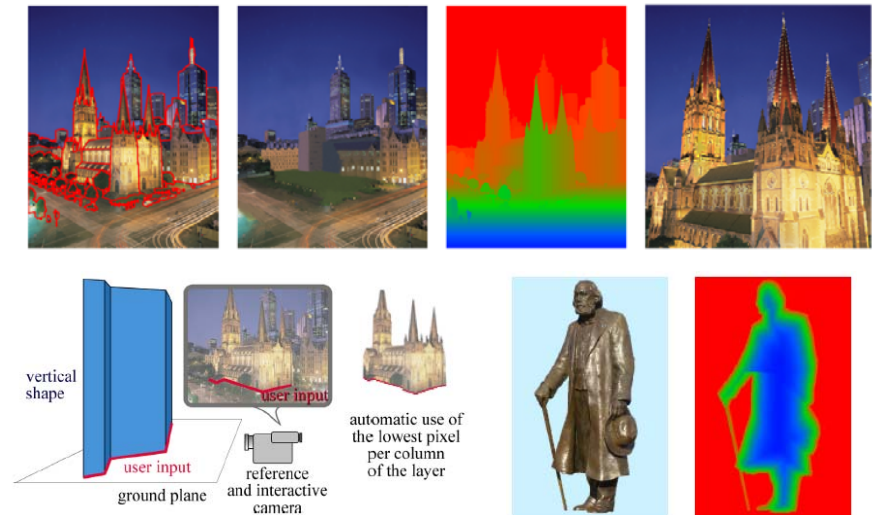
Zhang et. al. CVPR 2001

DigiVFX

original image	constraints	3D wireframe	novel view

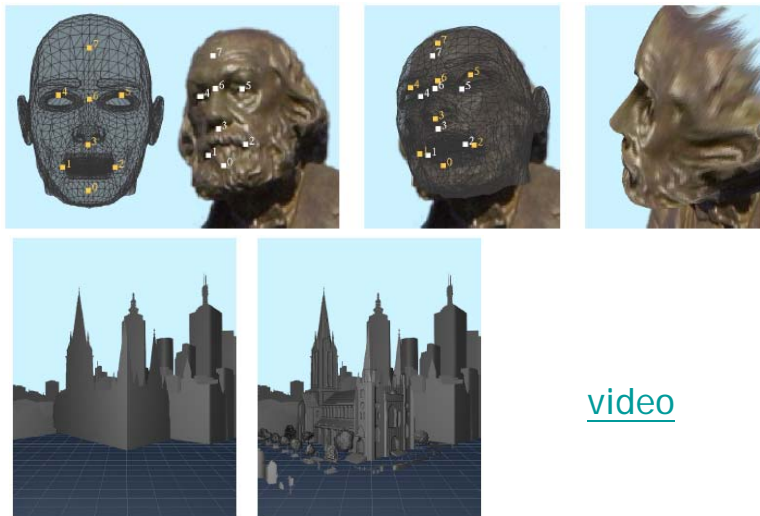
Oh et. al. SIGGRAPH 2001

DigiVFX



Oh et. al. SIGGRAPH 2001

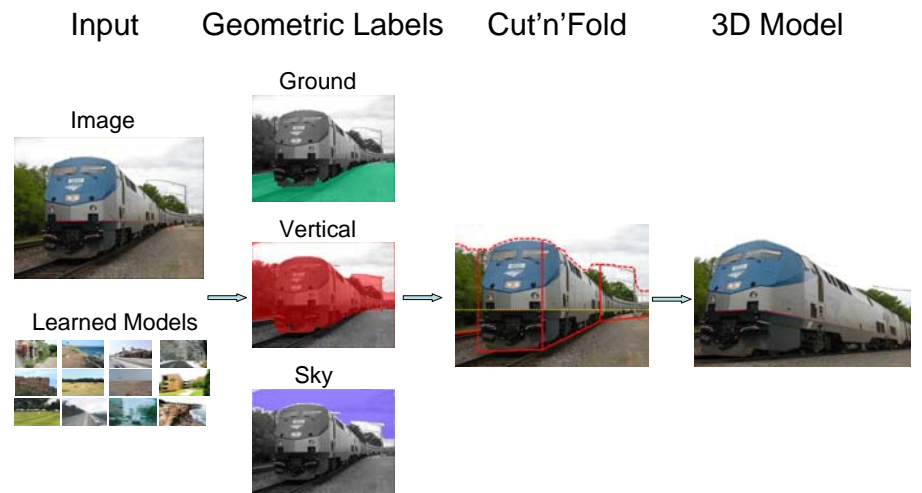
DigiVFX



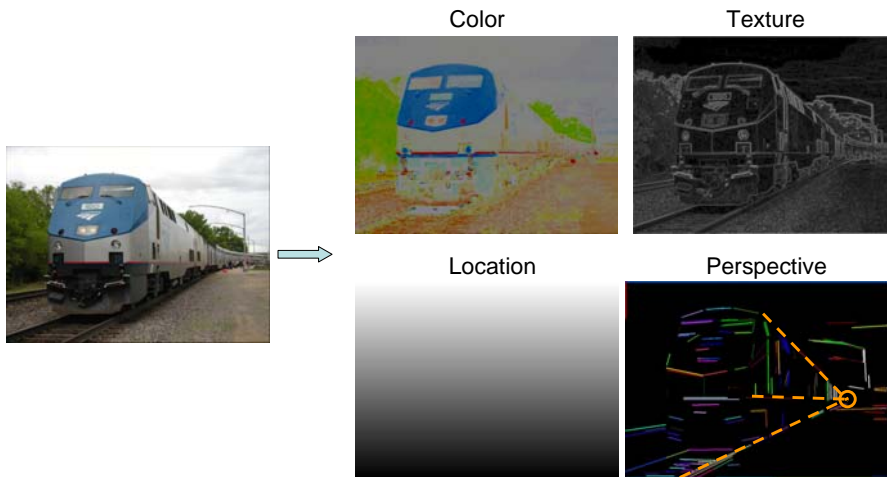
[video](#)

Automatic popup

DigiVFX



Geometric cues

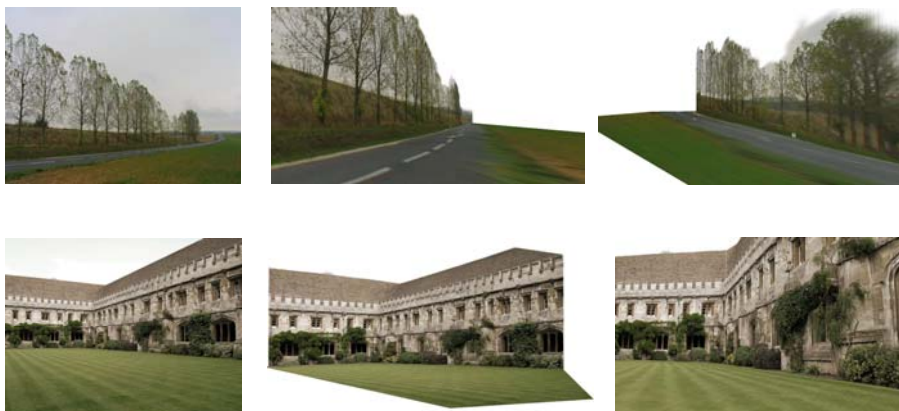


Automatic popup



Feature Descriptions	Num	Used
Color	15	15
C1. RGB values: mean	3	3
C2. HSV values: conversion from mean RGB values	3	3
C3. Hue: histogram (5 bins) and entropy	6	6
C4. Saturation: histogram (3 bins) and entropy	3	3
Texture	29	13
T1. DOOG Filters: mean abs response	12	3
T2. DOOG Filters: mean of variables in T1	1	0
T3. DOOG Filters: id of max of variables in T1	1	1
T4. DOOG Filters: (max - median) of variables in T1	1	1
T5. Textons: mean abs response	12	7
T6. Textons: max of variables in T5	1	0
T7. Textons: (max - median) of variables in T5	1	1
Location and Shape	12	10
L1. Location: normalized x and y, mean	2	2
L2. Location: norm. x and y, 10 th and 90 th percentile	4	4
L3. Location: norm. y wrt horizon, 10 th and 90 th pct	2	2
L4. Shape: number of superpixels in constellation	1	1
L5. Shape: number of sides of convex hull	1	0
L6. Shape: $\frac{\text{num pixels}}{\text{area}(\text{convex hull})}$	1	1
L7. Shape: whether the constellation region is contiguous	1	0
3D Geometry	35	28
G1. Long Lines: total number in constellation region	1	1
G2. Long Lines: % of nearly parallel pairs of lines	1	1
G3. Line Intersection: hist. over 12 orientations, entropy	13	11
G4. Line Intersection: % right of center	1	1
G5. Line Intersection: % above center	1	1
G6. Line Intersection: % far from center at 8 orientations	8	4
G7. Line Intersection: % very far from center at 8 orientations	8	5
G8. Texture gradient: x and y "edginess" (T2) center	2	2

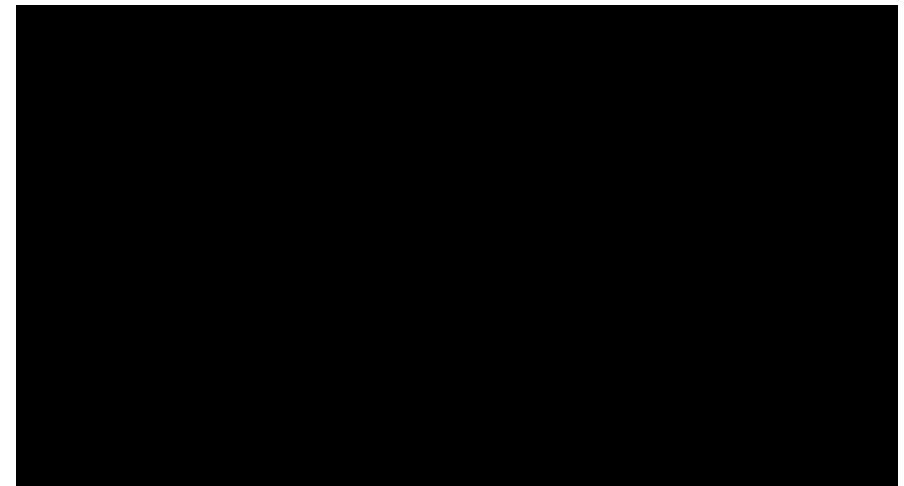
Results



Input Images

Automatic Photo Pop-up

Results



This approach works roughly for 35% of images.

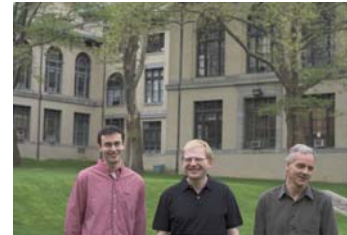
Failures

Labeling Errors



Failures

Foreground Objects



References

- P. Debevec, C. Taylor and J. Malik. [Modeling and Rendering Architecture from Photographs: A Hybrid Geometry- and Image-Based Approach](#), SIGGRAPH 1996.
- Y. Horry, K. Anjyo and K. Arai. [Tour Into the Picture: Using a Spidery Mesh Interface to Make Animation from a Single Image](#), SIGGRAPH 1997.
- L. Zhang, G. Dugas-Phocion, J.-S. Samson and S. Seitz. [Single View Modeling of Free-Form Scenes](#), CVPR 2001.
- B. Oh, M. Chen, J. Dorsey and F. Durand. [Image-Based Modeling and Photo Editing](#), SIGGRAPH 2001.
- D. Hoiem, A. Efros and M. Hebert. [Automatic Photo Pop-up](#), SIGGRAPH 2005.