
Various Perspectives (Mosaics and Panoramas)

Computational Photography

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Projects

List available now

Email to me: group, topic, why it is interesting

■ **until Thursday next week (24th of May)**

Project proposal (2 pages): 1st of June

Project idea presentation: 8th of June

Final Project presentation: 20th of July

Project report

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Mosaics and Panoramas

- basic idea
- registration
- resample
- blend

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Why Mosaic?

Are you getting the whole picture?

■ Compact Camera FOV = 50 x 35°



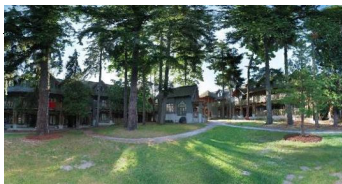
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Why Mosaic?

Are you getting the whole picture?

- Compact Camera FOV = 50 x 35°
- Human FOV = 200 x 135°



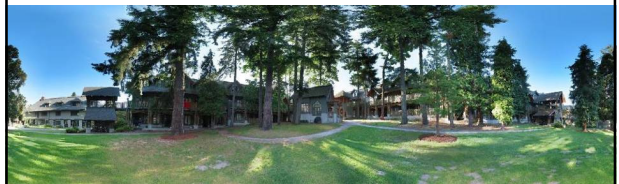
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Why Mosaic?

Are you getting the whole picture?

- Compact Camera FOV = 50 x 35°
- Human FOV = 200 x 135°
- Panoramic Mosaic = 360 x 180°



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Single vs. Multiple Viewpoint

Single-viewpoint

- Necessary for creating pure perspective images.
- Many vision algorithms assume pinhole cameras.
- Images that aren't perspective images look distorted.

Multi-viewpoint

- Cross-slit panoramas, etc.
- necessary for scenes which cannot be captured from a single viewpoint

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Omnidirectional (Catadioptric) Cameras



O-360

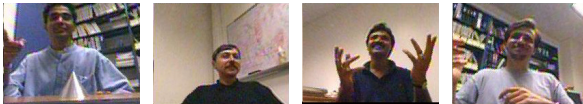


EyeSee360

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Images of an Omnidirectional Camera

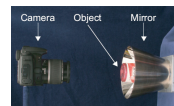


images: CAVE lab

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Catadioptric System – Full Texture

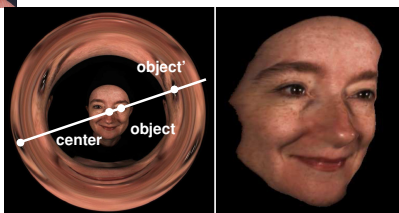
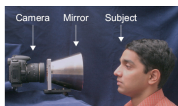


[Kuthirummal 2006]

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Catadioptric System – Stereo



epipolar line along the diameter

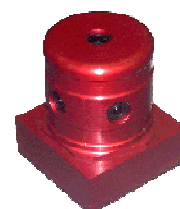
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Multi-camera, Single-viewpoint ?



Immersive Media "Dodeca2000"



PointGrey Ladybug

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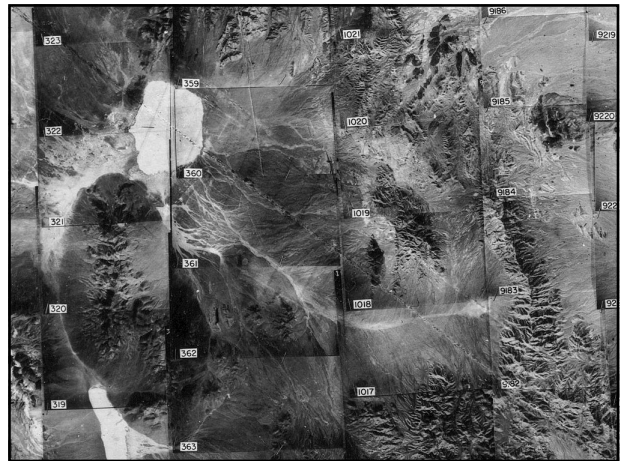
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Image Mosaicing

- Register multiple images
- Blend

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Single Center of Projection

- Take a sequence of images from the same position
 - Rotate the camera about its optical center
- Compute transformation between second image and first
- Transform the second image to overlap with the first
- Blend the two together to create a mosaic
- If there are more images, repeat

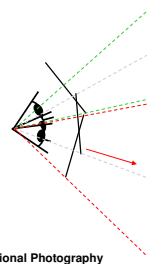
...why don't we need the 3D geometry?

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Image Reprojection

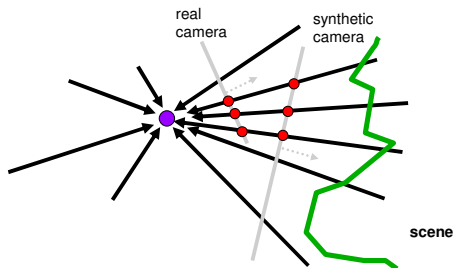
- The images are reprojected onto a common plane
- The mosaic is formed on this plane
- Mosaic is a *synthetic wide-angle camera*



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A pencil of rays contains all views



Can generate any synthetic camera view as long as it has the **same center of projection!**

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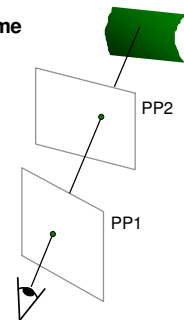
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Image reprojection

How to relate two images from the same camera center?

Images contain the same information along the same ray.

Use 2D image wrap instead of ray tracing.

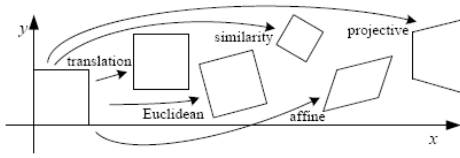


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Taxonomy of Projective Transformations

$$\begin{pmatrix} x'_1 \\ x'_2 \\ x'_3 \end{pmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$



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Taxonomy of Projective Transformations

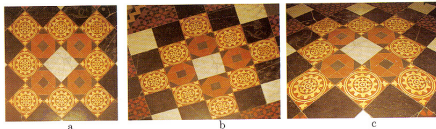
$$\begin{pmatrix} x'_1 \\ x'_2 \\ x'_3 \end{pmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

Group	Matrix	Distortion	Invariant properties
Projective 8 dof	$\begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix}$		Concurrency, collinearity, order of contact; intersection (1 pt contact); tangency (2 pt contact); inflections (3 pt contact with line); tangent discontinuities and cusps. cross ratio (ratio of ratio of lengths).
Affine 6 dof	$\begin{bmatrix} a_{11} & a_{12} & t_x \\ a_{21} & a_{22} & t_y \\ 0 & 0 & 1 \end{bmatrix}$		Parallelism, ratio of areas, ratio of lengths on collinear or parallel lines (e.g. midpoints), linear combinations of vectors (e.g. centroids). The line at infinity, l_∞ .
Similarity 4 dof	$\begin{bmatrix} sr_{11} & sr_{12} & t_x \\ sr_{21} & sr_{22} & t_y \\ 0 & 0 & 1 \end{bmatrix}$		Ratio of lengths, angle. The circular points, I, J (see section 1.7.3).
Euclidean 3 dof	$\begin{bmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ 0 & 0 & 1 \end{bmatrix}$		Length, area

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Distortions under Central Projection



- **Similarity**: circle remains circle, square remains square
 ⇒ line orientation is preserved
- **Affine**: circle becomes ellipse, square becomes rhombus
 ⇒ parallel lines remain parallel
- **Projective**: imaged object size depends on distance from camera
 ⇒ parallel lines converge

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Homography

A: Projective – mapping between any two PPs with the same center of projection

- rectangle should map to arbitrary quadrilateral
- parallel lines aren't
- but must preserve straight lines
- same as: project, rotate, reproject

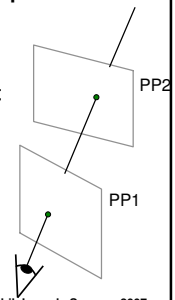
called **Homography**

$$\begin{bmatrix} wx' \\ wy' \\ w \end{bmatrix} = \begin{bmatrix} * & * & * \\ * & * & * \\ * & * & * \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

$$\mathbf{p}' = \mathbf{H} \mathbf{p}$$

To apply a homography **H**

- Compute $\mathbf{p}' = \mathbf{H} \mathbf{p}$ (regular matrix multiply)
- Convert \mathbf{p}' from homogeneous to image coordinates



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Removing Projective Distortion



$$\begin{pmatrix} x'_1 \\ x'_2 \\ x'_3 \end{pmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

Projective transformation in inhomogeneous form

$$x' = \frac{x'_1}{x'_3} = \frac{h_{11}x + h_{12}y + h_{13}}{h_{31}x + h_{32}y + h_{33}} \quad y' = \frac{x'_2}{x'_3} = \frac{h_{21}x + h_{22}y + h_{23}}{h_{31}x + h_{32}y + h_{33}}$$

4 general point correspondences $(x, y \rightarrow x', y')$ on the planar facade lead to eight linear equations of the type

$$x' (h_{31}x + h_{32}y + h_{33}) = h_{11}x + h_{12}y + h_{13}$$

$$y' (h_{31}x + h_{32}y + h_{33}) = h_{21}x + h_{22}y + h_{23}$$

Sufficient to solve for **H** up to multiplicative factor

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The Direct Linear Transform (DLT) Algorithm



Given: 4 2D point correspondences

$$\mathbf{x}_i = \begin{pmatrix} x_i \\ y_i \\ 1 \end{pmatrix} \leftrightarrow \mathbf{x}'_i = \begin{pmatrix} x'_i \\ y'_i \\ 1 \end{pmatrix}$$

Objective: estimate the projective transform matrix **H**

$$\mathbf{x}' = \mathbf{H} \mathbf{x}$$

$$\begin{pmatrix} x'_1 \\ x'_2 \\ x'_3 \end{pmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

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The DLT Algorithm II

Estimating matrix H from point correspondences is equivalent to

$$\mathbf{x}' = H\mathbf{x}, \quad \mathbf{x}_i = \begin{pmatrix} x_i \\ y_i \\ w_i \end{pmatrix} \Leftrightarrow \mathbf{x}_i' = \begin{pmatrix} x_i' \\ y_i' \\ w_i' \end{pmatrix} \quad \mathbf{x}_i' \times H\mathbf{x}_i = 0$$

might have different length but are collinear

gives

$$\mathbf{x}_i' \times H\mathbf{x}_i = \begin{pmatrix} y_i' h^3 x_i - w_i' h^2 x_i \\ w_i' h^3 x_i - x_i' h^3 x_i \\ x_i' h^2 x_i - y_i' h^3 x_i \end{pmatrix}$$

Re-ordering into \mathbf{h} vector

$$\begin{bmatrix} 0^T & -w_i' x_i^T & y_i' x_i^T \\ w_i' x_i^T & 0^T & -x_i' x_i^T \\ -y_i' x_i^T & x_i' x_i^T & 0^T \end{bmatrix} \begin{pmatrix} h^1 \\ h^2 \\ h^3 \end{pmatrix} = 0.$$

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The DLT Algorithm III

$$\begin{bmatrix} 0^T & -w_i' x_i^T & y_i' x_i^T \\ w_i' x_i^T & 0^T & -x_i' x_i^T \\ -y_i' x_i^T & x_i' x_i^T & 0^T \end{bmatrix} \begin{pmatrix} h^1 \\ h^2 \\ h^3 \end{pmatrix} = 0, \quad \mathbf{x}_i = \begin{pmatrix} x_i \\ y_i \\ w_i \end{pmatrix} \Leftrightarrow \mathbf{x}_i' = \begin{pmatrix} x_i' \\ y_i' \\ w_i' \end{pmatrix}$$

Only rows 1 and 2 are linearly independent \Rightarrow omit row 3

$$\begin{bmatrix} 0^T & -w_i' x_i^T & y_i' x_i^T \\ w_i' x_i^T & 0^T & -x_i' x_i^T \end{bmatrix} \begin{pmatrix} h^1 \\ h^2 \\ h^3 \end{pmatrix} = 0, \quad A_i \mathbf{h} = 0$$

Inhomogeneous solution: set one matrix entry equal to 1 (e.g. h^3)

$$\begin{bmatrix} 0 & 0 & 0 & -x_i w_i' & -y_i w_i' & -w_i w_i' & x_i y_i' & y_i y_i' \\ x_i w_i' & y_i w_i' & w_i w_i' & 0 & 0 & 0 & -x_i x_i' & -y_i x_i' \end{bmatrix} \mathbf{h} = \begin{pmatrix} -w_i y_i' \\ w_i x_i' \end{pmatrix}$$

Solve by Gaussian elimination or least-squares techniques

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Estimating Homographies

Objective

Given $n \geq 4$ 2D to 2D point correspondences $\{\mathbf{x}_i \leftrightarrow \mathbf{x}_i'\}$, determine the 2D homography matrix H such that $\mathbf{x}_i' = H\mathbf{x}_i$.

Algorithm

- (i) **Normalization of \mathbf{x} :** Compute a similarity transformation T , consisting of a translation and scaling, that takes points \mathbf{x}_i to a new set of points $\tilde{\mathbf{x}}_i$ such that the centroid of the points $\tilde{\mathbf{x}}_i$ is the coordinate origin $(0,0)^T$, and their average distance from the origin is $\sqrt{2}$.
- (ii) **Normalization of \mathbf{x}' :** Compute a similar transformation T' for the points in the second image, transforming points \mathbf{x}_i' to $\tilde{\mathbf{x}}_i'$.
- (iii) **DLT:** Apply algorithm to the correspondences $\tilde{\mathbf{x}}_i \leftrightarrow \tilde{\mathbf{x}}_i'$ to obtain a homography \tilde{H} .
- (iv) **Denormalization:** Set $H = T'^{-1} \tilde{H} T$.

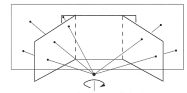
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Panoramic Mosaicing

Rotation about camera center: homography

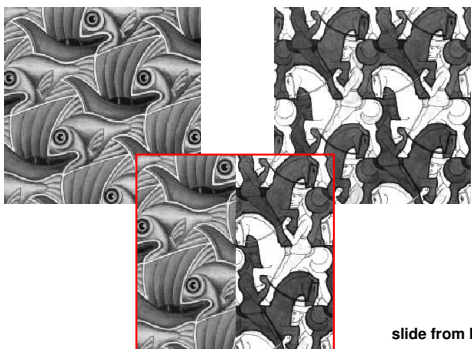
- choose one image as reference
 - compute homography to map neighboring image to reference image plane
 - projectively warp image, add to reference plane
 - repeat for all images
- \Rightarrow bow tie shape



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Image Blending

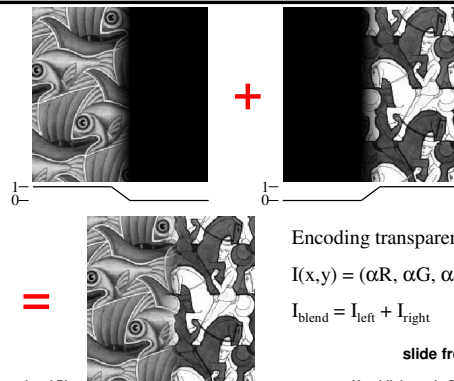


slide from Efros

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Feathering



$$I(x,y) = (\alpha R, \alpha G, \alpha B, \alpha)$$

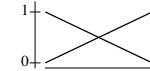
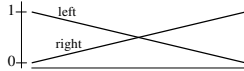
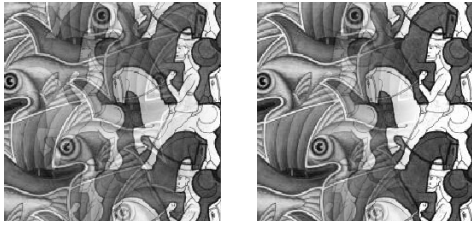
$$I_{\text{blend}} = I_{\text{left}} + I_{\text{right}}$$

slide from Efros

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Affect of Window Size

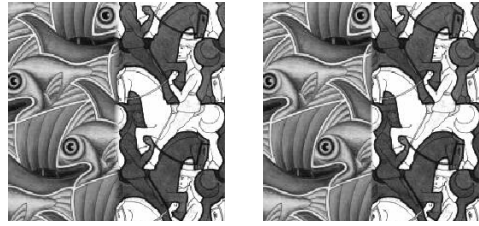


slide from Efros

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Affect of Window Size

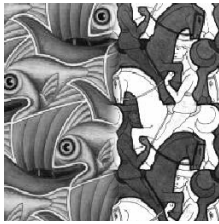


slide from Efros

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Good Window Size



“Optimal” Window: smooth but not ghosted

slide from Efros

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What is the Optimal Window?

To avoid seams

- window \gg size of largest prominent feature

To avoid ghosting

- window $\leq 2 \times$ size of smallest prominent feature

Natural to cast this in the *Fourier domain*

- largest frequency $\leq 2 \times$ size of smallest frequency
- do blending in different frequency bands

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What if the Frequency Spread is Wide



Idea (Burt and Adelson)

- Compute $F_{\text{left}} = \text{FFT}(I_{\text{left}})$, $F_{\text{right}} = \text{FFT}(I_{\text{right}})$
- Decompose Fourier image into octaves (bands)
 - $F_{\text{left}} = F_{\text{left}}^1 + F_{\text{left}}^2 + \dots$
- Feather corresponding octaves F_{left}^i with F_{right}^i
- Can compute inverse FFT and feather in spatial domain
- Sum feathered octave images in frequency domain

Better implemented in *spatial domain*

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What does blurring take away?



original

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What does blurring take away?



smoothed (5x5 Gaussian)

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High-Pass Filter



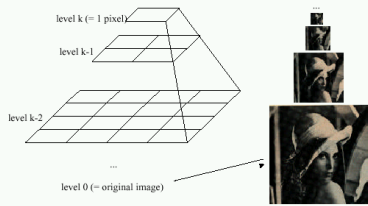
smoothed - original

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Image Pyramids

Idea: Represent $N \times N$ image as a "pyramid" of $1 \times 1, 2 \times 2, 4 \times 4, \dots, 2^k \times 2^k$ images (assuming $N=2^n$)

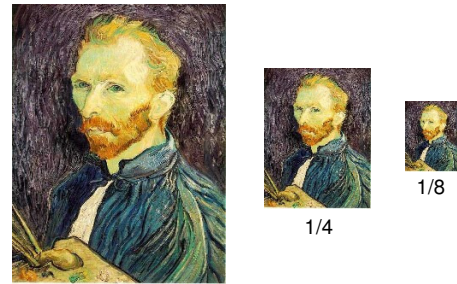


mipmap or precursor of wavelets

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Image Sub-sampling

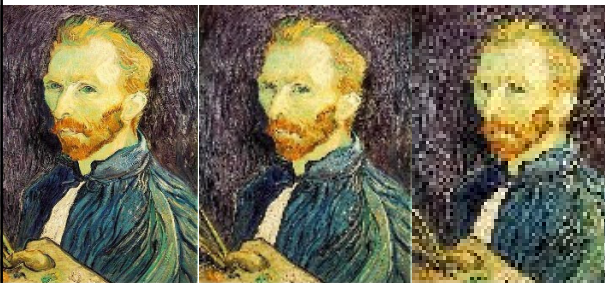


Throw away every other row and column to create a $1/2$ size image

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Image Sub-sampling



1/2

1/4 (2x zoom)

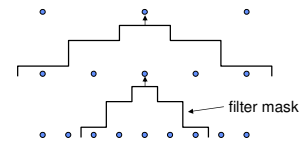
1/8 (4x zoom)

Why does this look so bad?

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Gaussian Pyramid Construction



Repeat

- Filter
- Subsample

Until minimum resolution reached

Whole pyramid is only $4/3$ the size of the original image!

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Gaussian pre-filtering



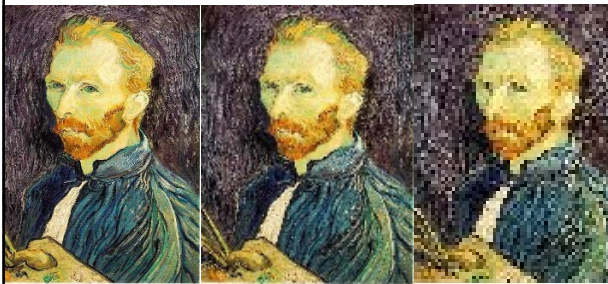
Gaussian 1/2
Solution: filter the image, then subsample
 ■ Filter size should double for each 1/2 size reduction.
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Subsampling with Gaussian pre-filtering



Gaussian 1/2 G 1/4 G 1/8
Solution: filter the image, then subsample
 ■ Filter size should double for each 1/2 size reduction.
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Compare with...



1/2 1/4 (2x zoom) 1/8 (4x zoom)
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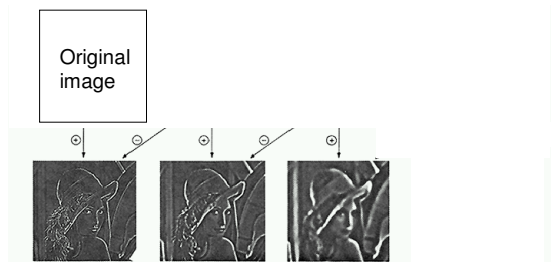
Band-pass filtering

Gaussian Pyramid (low-pass images)



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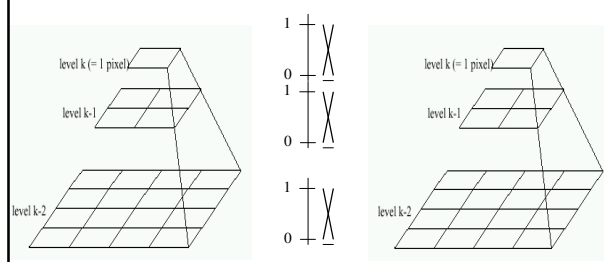
Laplacian Pyramid



How can we reconstruct (collapse) this pyramid into the original image?

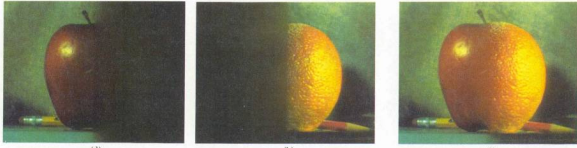
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Pyramid Blending



Left pyramid blend Right pyramid
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Pyramid Blending



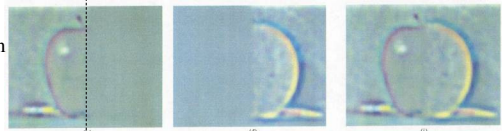
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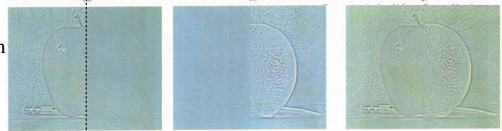
laplacian
level
4



laplacian
level
2



laplacian
level
0



left pyramid

right pyramid

blended pyramid

Simplification: Two-band Blending

Brown & Lowe, 2003

- Only use two bands: high freq. and low freq.
- Blends low freq. smoothly
- Blend high freq. with no smoothing: use binary mask



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2-band Blending



Low frequency ($\lambda > 2$ pixels)

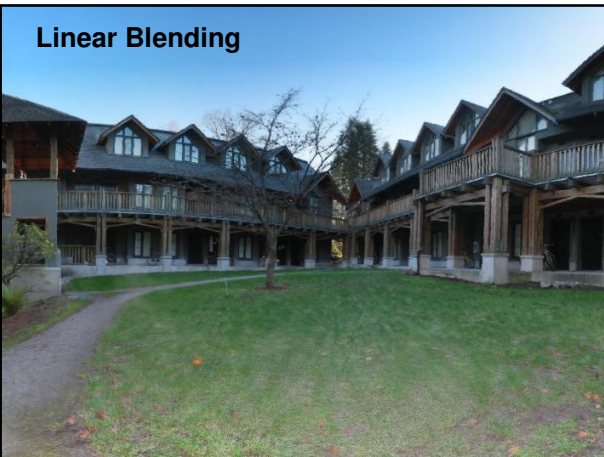


High frequency ($\lambda < 2$ pixels)

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Linear Blending



2-band Blending



Still Some Artifacts Left...

Ghosting—objects move in the scene.

Differing exposures between images.

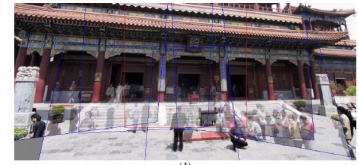
- Pyramid blending does not solve this.

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De-Ghosting

In regions with differences don't blend - crop.



[Uyttendaele et al. 2001]

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Gradient Domain Blending

In Pyramid Blending, we decomposed our image into 2nd derivatives (Laplacian) and a low-res image

Let us now look at 1st derivatives (gradients):

No need for low-res image

- captures everything (up to a constant)
- easy to deal with low-frequency differences

Idea:

- Differentiate
- Blend
- Reintegrate

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Gradient Domain Blending (2D)

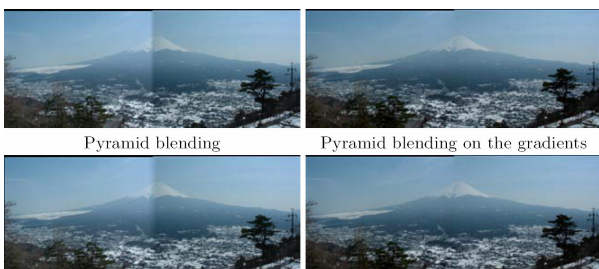
Trickier in 2D:

- Take partial derivatives dx and dy (the gradient field)
- Fiddle around with them (smooth, blend, feather, etc)
- Reintegrate
 - But now $\text{integral}(dx)$ might not equal $\text{integral}(dy)$
- Find the most agreeable solution
 - Equivalent to solving Poisson equation
 - Can use FFT, deconvolution, multigrid solvers, etc.

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Comparisons [Levin et al 2004]



Pyramid blending

Pyramid blending on the gradients

Feathering

GIST1

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Homography or not ?



- Coincidences between 3D points at different depths are preserved
 - Pure camera rotation about camera center
- ⇒ 2D Homography

- Different depths are imaged to different image positions
 - Camera rotates and translates
- ⇒ Motion Parallax, no Homography

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Multiperspective Panoramas

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Multiperspective Panoramas

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Aspect Ratio Distortion [Roman 2005]

Close and distant objects get distorted due to different perspective in x and y.

Aspect Ratio Distortion

Images with the original perspective don't suffer from this issue.
How to seamlessly combine multiple perspective images?

Background: Perspective Images

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Background: Pushbroom Images

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Cross-Slits Images

Cross-slits
Multi-
Perspective
Image

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Cross-Slits Images

Cross-slits
Multi-
Perspective
Image

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Cross-Slits Images

Cross-slits
Multi-
Perspective
Image

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Automatic Construction

What causes the distortion?

- difference between vertical and horizontal perspectives
- changes aspect ratio

How can it be reduced?

- quantify the distortion
- place picture surface
- select ray angles that minimize overall distortion

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Aspect Ratio – Perspective

$$h = D_0 \left(\frac{H}{D_0 - \Delta d} \right)$$

$$w = D_0 \left(\frac{W}{D_0 - \Delta d} \right)$$

picture surface
camera path
Perspective image

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Aspect Ratio – Cross-slits

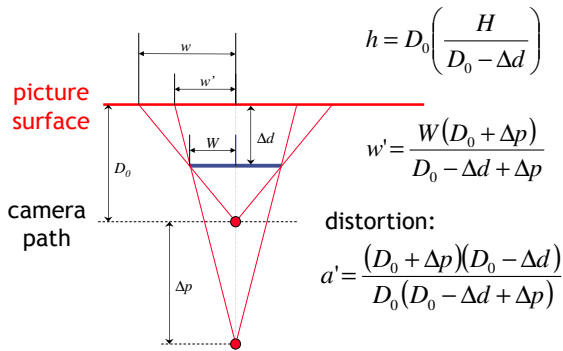
$$h = D_0 \left(\frac{H}{D_0 - \Delta d} \right)$$

picture surface
camera path
Crossed-slits image

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Aspect Ratio – Crossed-slits

[Zomet 03]

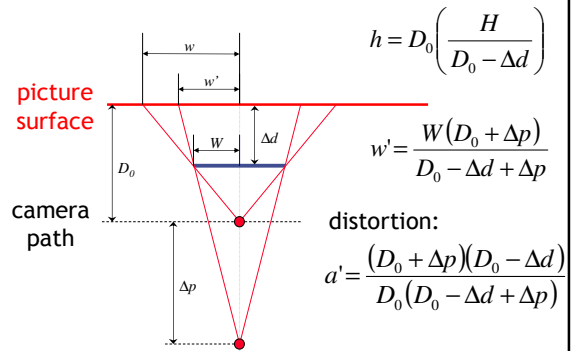


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Aspect Ratio – Crossed-slits

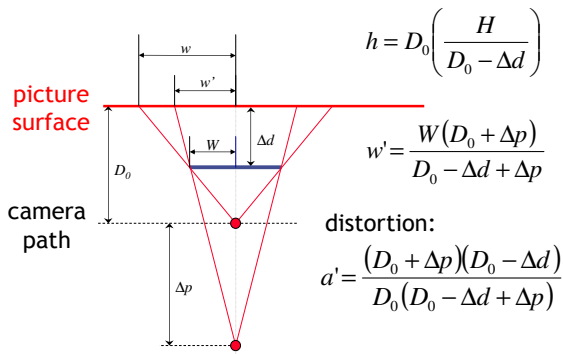
[Zomet 03]



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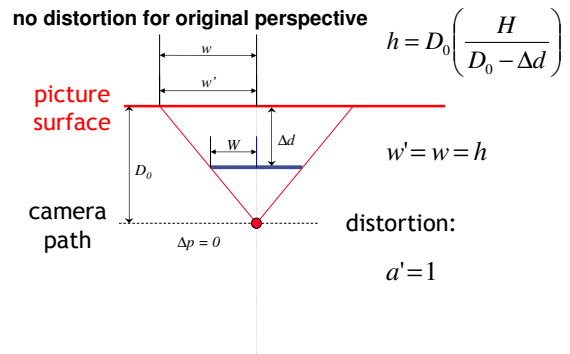
Special Case: $\Delta p = 0$



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Special Case: $\Delta p = 0$

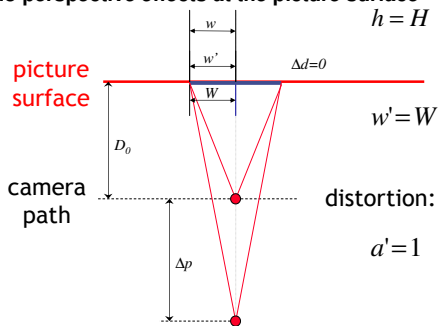


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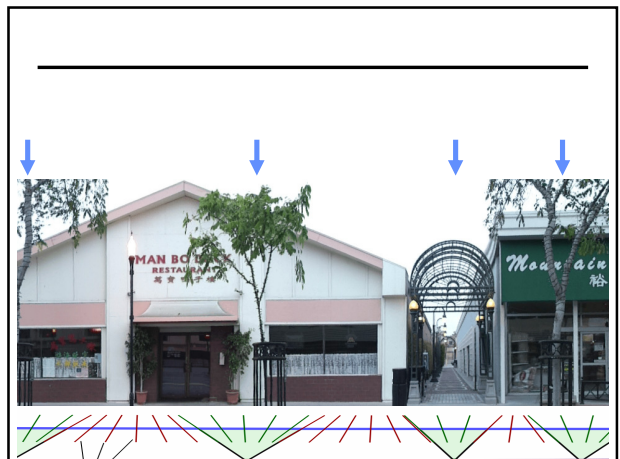
Special Case: $\Delta d = 0$

no perspective effects at the picture surface



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Automatic Multiperspective Panormamas

pushbroom



automatic perspective

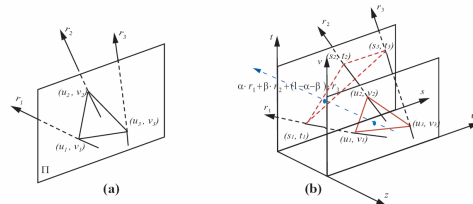
[Roman 2006]

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General Linear Cameras

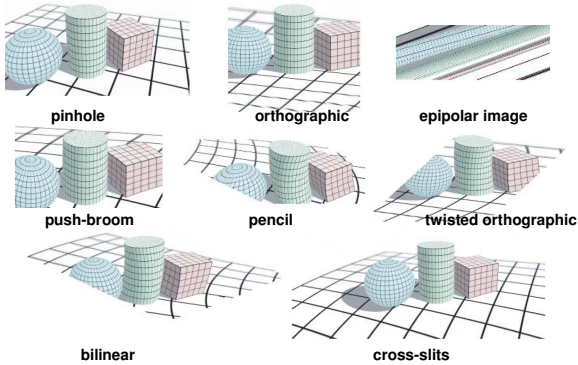
[Yu and McMillan 2004]



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General Linear Cameras

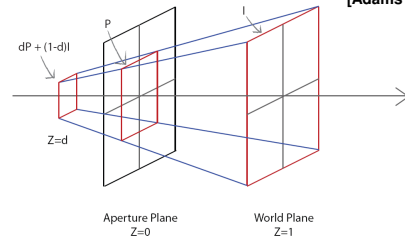


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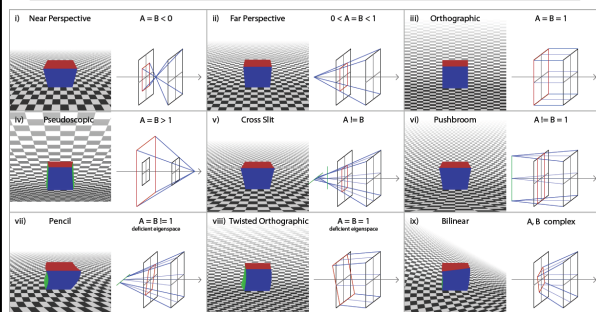
General Linear Cameras - Classification

[Adams 2007]



- rays described by: $(P(x, y), (x, y))$
- characteristic matrix: $P'_d = (1-d)P + dI$
 $P'_\lambda = P + \lambda I$ with $\lambda = \frac{d}{d-1}$

General Linear Cameras



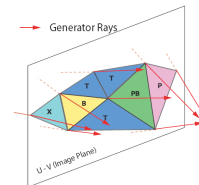
analyze the eigenvalues of P

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Multiperspective – Rendering Framework

[Yu and McMillan 2004b]



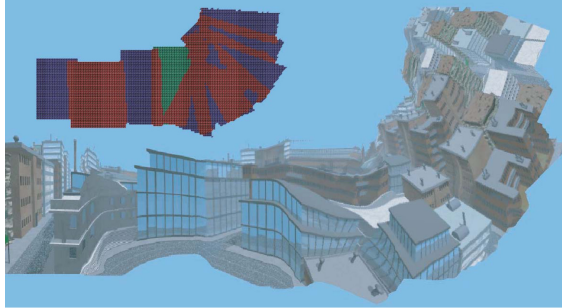
- specify perspective per triangle
- blend between neighboring triangles

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Multiperspective – Rendering Framework

[Yu and McMillan 2004b]



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Panoramas

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General Linear Cameras

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