Image stitching

Digital Visual Effects

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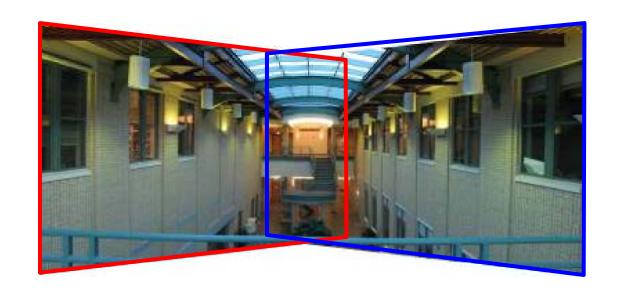
Stitching = alignment + blending

geometrical registration

photometric registration









Applications of image stitching

- Video stabilization
- Video summarization
- Video compression
- Video matting
- Panorama creation



Video summarization





Video compression













Object removal

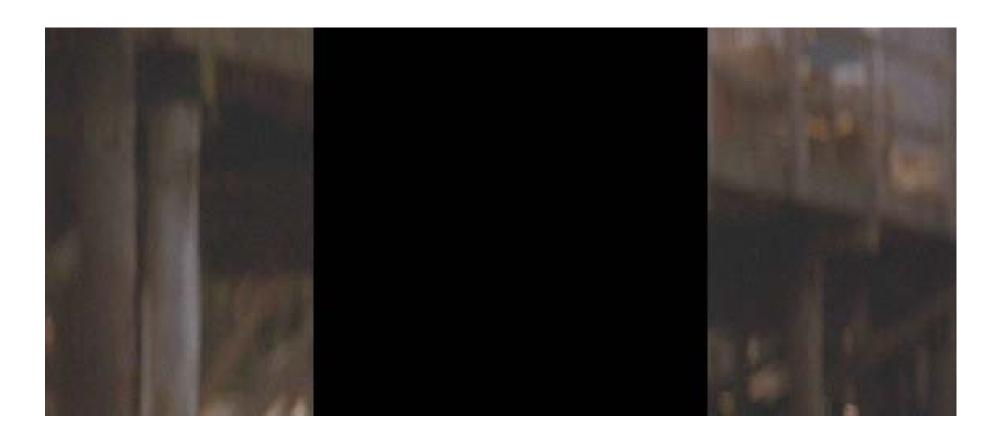




input video

Object removal





remove foreground







estimate background

Object removal





background estimation



Panorama creation











Why panorama?



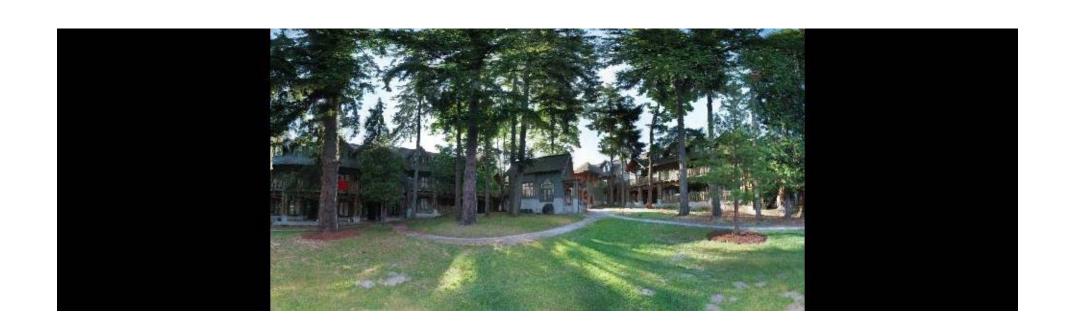
- Are you getting the whole picture?
 - Compact Camera FOV = 50 x 35°



Why panorama?



- Are you getting the whole picture?
 - Compact Camera FOV = 50 x 35°
 - Human FOV = $200 \times 135^{\circ}$



Why panorama?



Are you getting the whole picture?

- Compact Camera FOV = 50 x 35°

- Human FOV = $200 \times 135^{\circ}$

- Panoramic Mosaic = $360 \times 180^{\circ}$





Panorama examples

- Like HDR, it is a topic of computational photography, seeking ways to build a better camera mostly in software.
- Most consumer cameras have a panorama mode
- Mars:

http://www.panoramas.dk/fullscreen3/f2_mars97.html

• Earth:

http://www.panoramas.dk/new-year-2006/taipei.html

http://www.360cities.net/

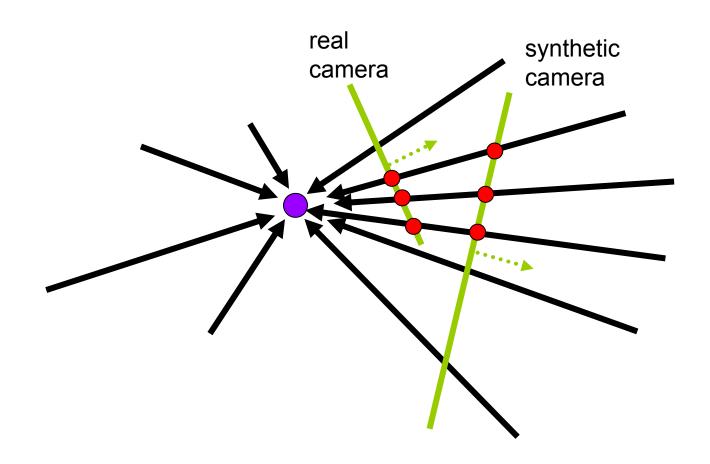


What can be globally aligned?

- In image stitching, we seek for a matrix to globally warp one image into another. Are any two images of the same scene can be aligned this way?
 - Images captured with the same center of projection
 - A planar scene or far-away scene



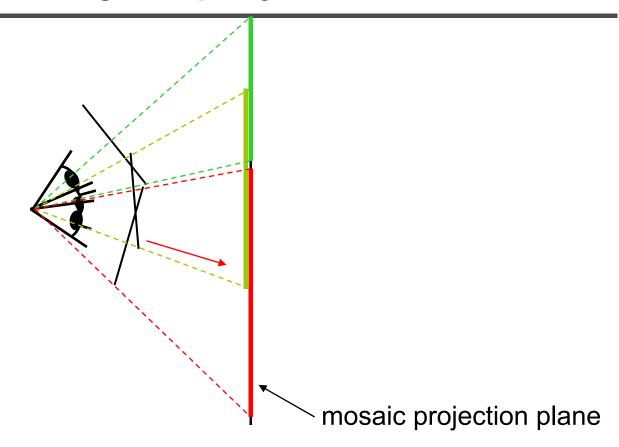
A pencil of rays contains all views



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



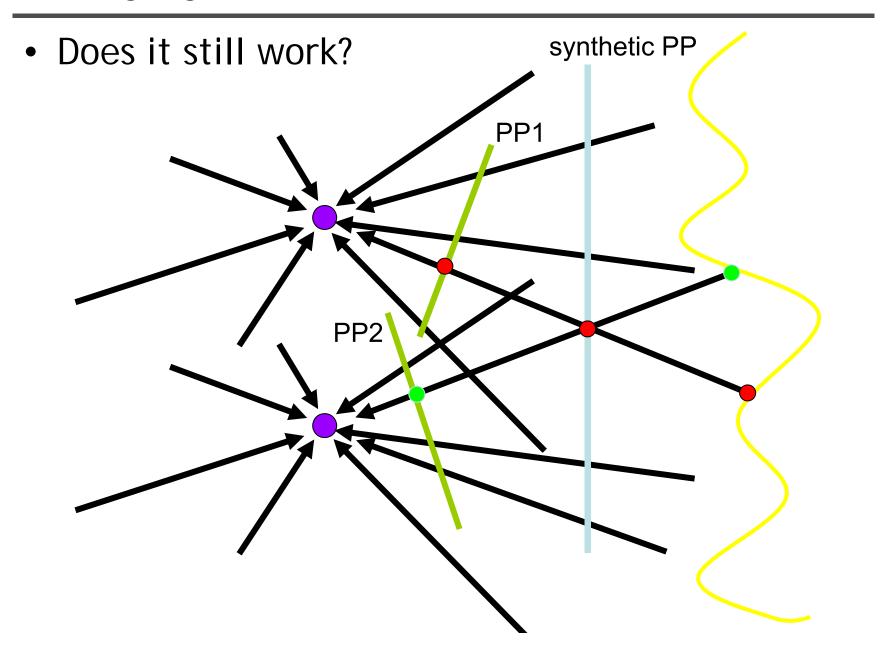
Mosaic as an image reprojection



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

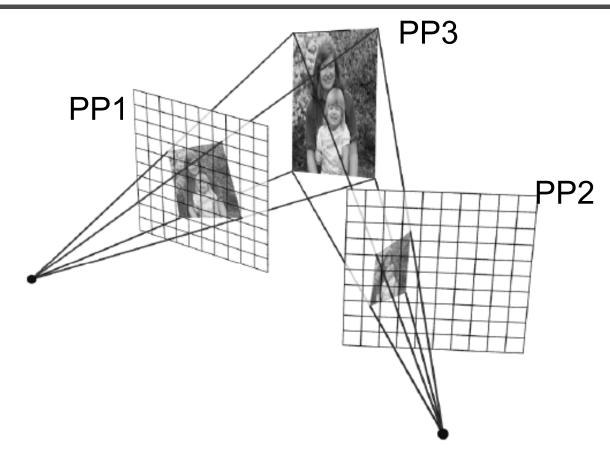


Changing camera center





Planar scene (or a faraway one)



- PP3 is a projection plane of both centers of projection, so we are OK!
- This is how big aerial photographs are made

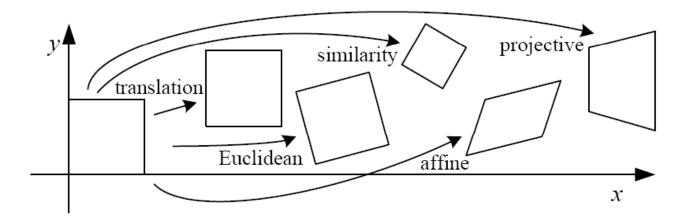
Motion models



• Parametric models as the assumptions on the relation between two images.

2D Motion models

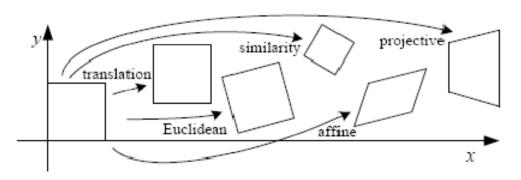




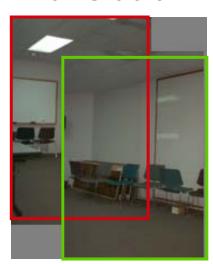
Name	Matrix	# D.O.F.	Preserves:	Icon
translation	$egin{bmatrix} ig[egin{array}{c c} ig[egin{array}{c c} I & t \end{bmatrix}_{2 imes 3} \end{array}$	2	orientation $+\cdots$	
rigid (Euclidean)	$igg igg[m{R} igg m{t} igg]_{2 imes 3}$	3	lengths + · · ·	\Diamond
similarity	$\boxed{\left[\begin{array}{c c} sR \mid t\end{array}\right]_{2\times 3}}$	4	$angles + \cdots$	\Diamond
affine	$\left[egin{array}{c} oldsymbol{A} \end{array} ight]_{2 imes 3}$	6	parallelism + · · ·	
projective	$\left[egin{array}{c} ilde{H} \end{array} ight]_{3 imes 3}$	8	straight lines	

Motion models

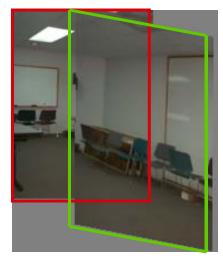




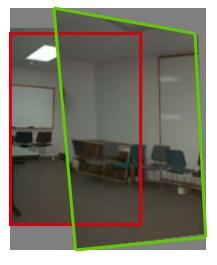
Translation



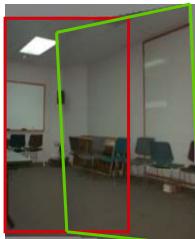
Affine



Perspective 3D rotation



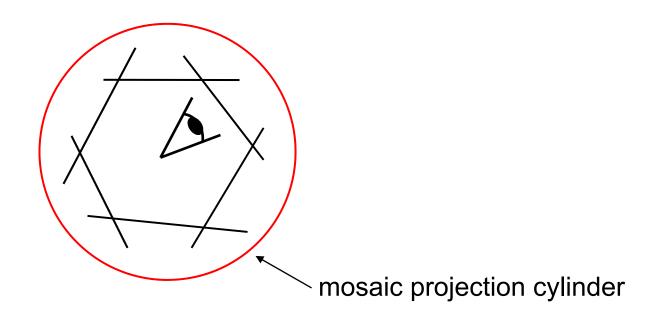
2 unknowns 6 unknowns 8 unknowns 3 unknowns





A case study: cylindrical panorama

What if you want a 360° field of view?









Steps

- Reproject each image onto a cylinder
- Blend
- Output the resulting mosaic



Cylindrical panorama

- 1. Take pictures on a tripod (or handheld)
- 2. Warp to cylindrical coordinate
- 3. Compute pairwise alignments
- 4. Fix up the end-to-end alignment
- 5. Blending
- 6. Crop the result and import into a viewer

It is required to do radial distortion correction for better stitching results!









Kaidan panoramic tripod head

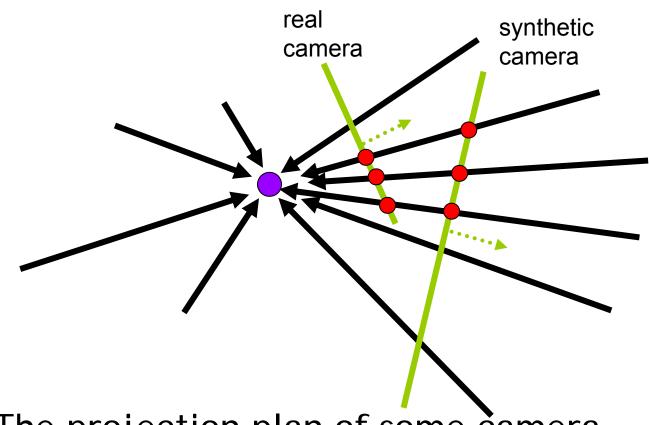
Translation model





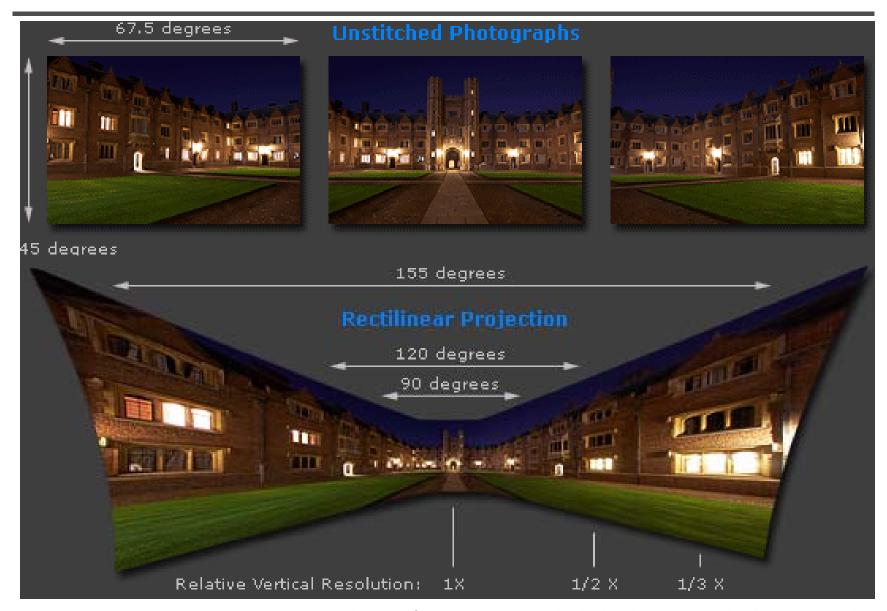
Try to align this in PaintShop Pro

Where should the synthetic camera be



- The projection plan of some camera
- Onto a cylinder

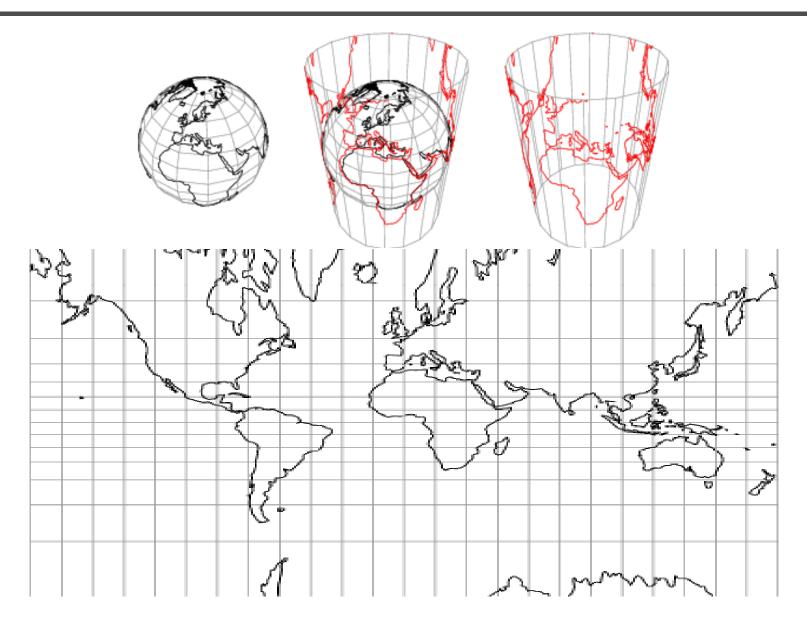




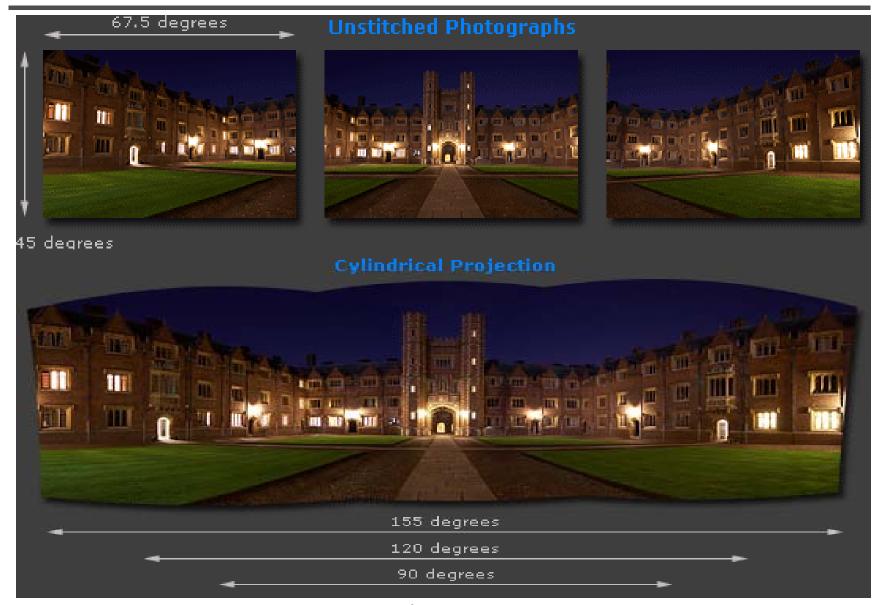
Adopted from http://www.cambridgeincolour.com/tutorials/image-projections.htm

DigiVFX

Cylindrical projection

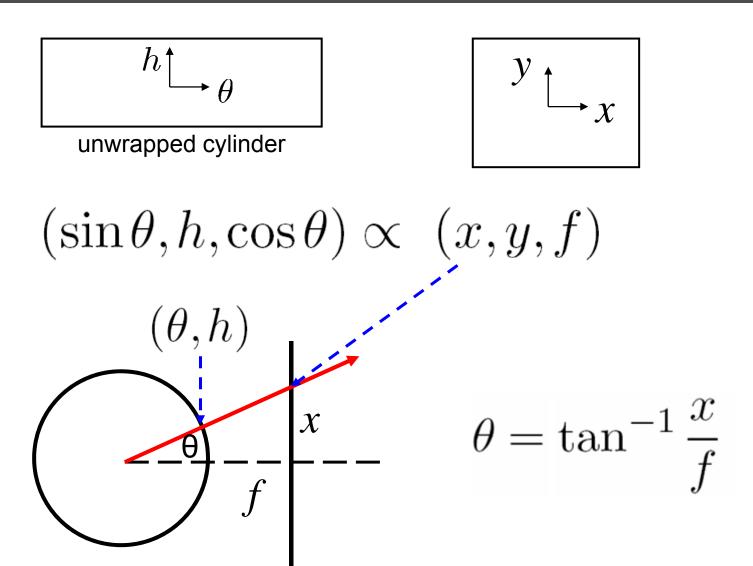




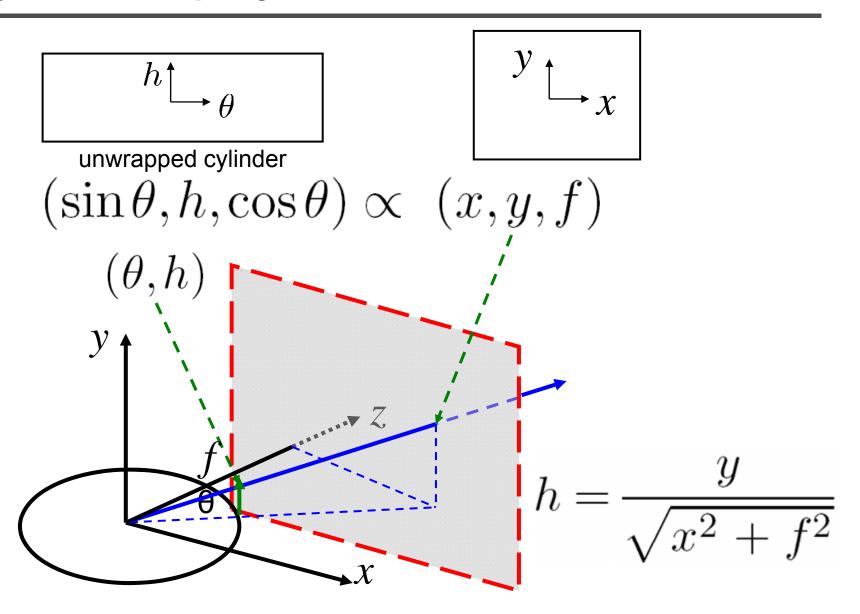


Adopted from http://www.cambridgeincolour.com/tutorials/image-projections.htm

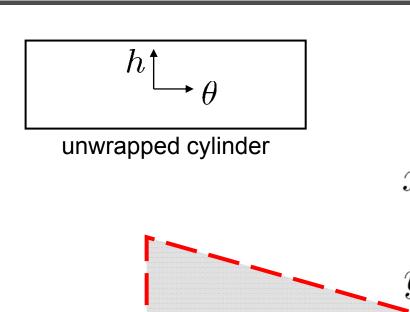


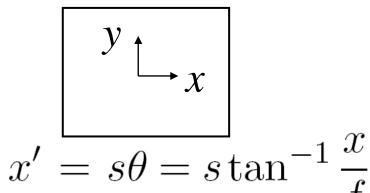








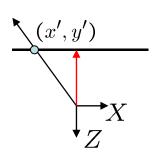


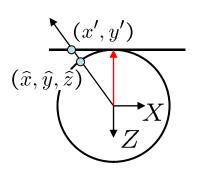


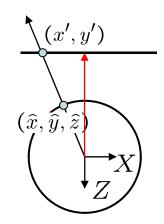
$$= sh = s \frac{y}{\sqrt{x^2 + f^2}}$$

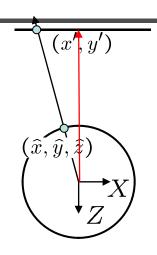
s=f gives less distortion











top-down view

Focal length – the dirty secret...









Image 384x300

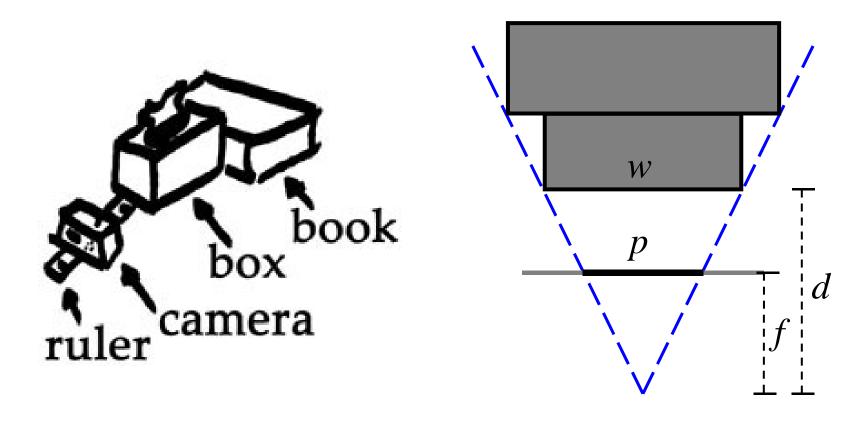
f = 180 (pixels)

f = 280

f = 380



A simple method for estimating f



Or, you can use other software, such as AutoStich, to help.

Input images

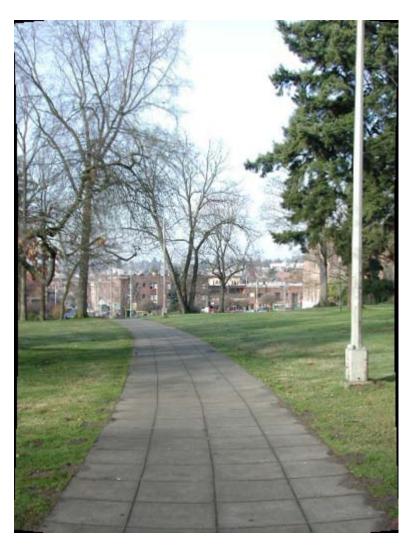














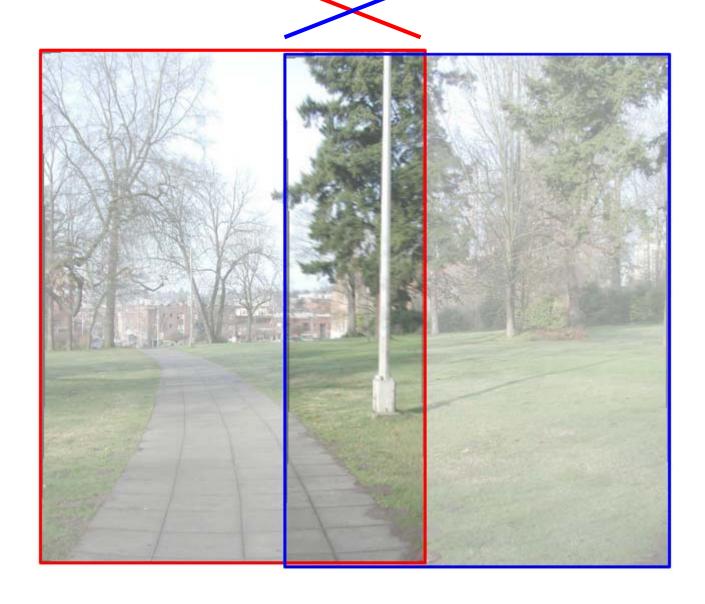


 Why blending: parallax, lens distortion, scene motion, exposure difference







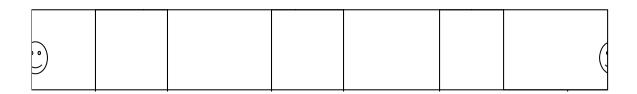








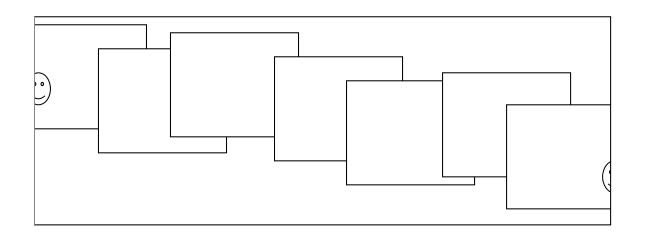
Assembling the panorama



Stitch pairs together, blend, then crop

Problem: Drift

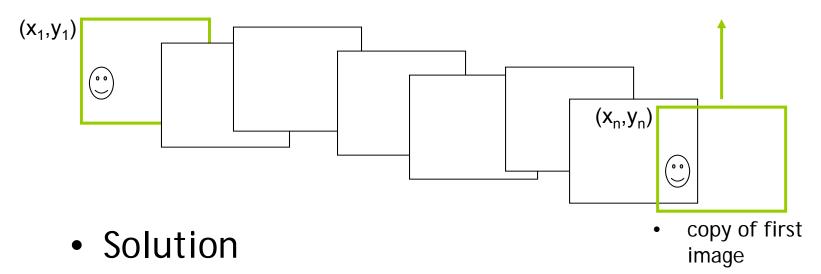




- Error accumulation
 - small errors accumulate over time

Problem: Drift





- add another copy of first image at the end
- there are a bunch of ways to solve this problem
 - add displacement of $(y_1 y_n)/(n 1)$ to each image after the first
 - compute a global warp: y' = y + ax
 - run a big optimization problem, incorporating this constraint
 - best solution, but more complicated
 - known as "bundle adjustment"



End-to-end alignment and crop





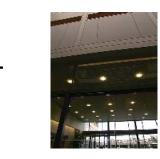
Viewer: panorama















 $\textbf{example:} \ \underline{\textbf{http://www.cs.washington.edu/education/courses/cse590ss/01wi/projects/project1/students/dougz/index.html} \\$



Viewer: texture mapped model



example: http://www.panoramas.dk/

Cylindrical panorama



- 1. Take pictures on a tripod (or handheld)
- 2. Warp to cylindrical coordinate
- 3. Compute pairwise alignments
- 4. Fix up the end-to-end alignment
- 5. Blending
- 6. Crop the result and import into a viewer



Determine pairwise alignment?

- Feature-based methods: only use feature points to estimate parameters
- We will study the "Recognising panorama" paper published in ICCV 2003
- Run SIFT (or other feature algorithms) for each image, find feature matches.



Determine pairwise alignment

- p'=Mp, where M is a transformation matrix, p and p' are feature matches
- It is possible to use more complicated models such as affine or perspective
- For example, assume M is a 2x2 matrix

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

Find M with the least square error

$$\sum_{i=1}^{n} (Mp - p')^2$$



Determine pairwise alignment

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \qquad \begin{aligned} x_1 m_{11} + y_1 m_{12} &= x_1' \\ x_1 m_{21} + y_1 m_{22} &= y_1' \end{aligned}$$

Overdetermined system

$$\begin{pmatrix} x_1 & y_1 & 0 & 0 \\ 0 & 0 & x_1 & y_1 \\ x_2 & y_2 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots \\ x_n & y_n & 0 & 0 \\ 0 & 0 & x_n & y_n \end{pmatrix} \begin{pmatrix} m_{11} \\ m_{12} \\ m_{21} \\ m_{22} \end{pmatrix} = \begin{pmatrix} x_1 \\ y_1 \\ x_2 \\ \vdots \\ x_n \\ y_n \end{pmatrix}$$





Given an overdetermined system

$$Ax = b$$

the normal equation is that which minimizes the sum of the square differences between left and right sides

$$\mathbf{A}^{\mathrm{T}}\mathbf{A}\mathbf{x} = \mathbf{A}^{\mathrm{T}}\mathbf{b}$$

Why?



$$E(\mathbf{x}) = (\mathbf{A}\mathbf{x} - \mathbf{b})^2$$

$$\begin{bmatrix} a_{11} & \dots & a_{1m} \\ \vdots & & \vdots \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nm} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ \vdots \\ b_n \end{bmatrix}$$

nxm, n equations, m variables



$$\mathbf{Ax} - \mathbf{b} = \begin{bmatrix} \sum_{j=1}^{m} a_{1j} x_{j} \\ \vdots \\ \sum_{j=1}^{m} a_{ij} x_{j} \\ \vdots \\ \sum_{j=1}^{m} a_{nj} x_{j} \end{bmatrix} - \begin{bmatrix} b_{1} \\ \vdots \\ b_{i} \\ \vdots \\ b_{n} \end{bmatrix} = \begin{bmatrix} \sum_{j=1}^{m} a_{1j} x_{j} \\ \vdots \\ \sum_{j=1}^{m} a_{nj} x_{j} \\ \vdots \\ \sum_{j=1}^{m} a_{nj} x_{j} \end{bmatrix} - b_{i}$$

$$E(\mathbf{x}) = (\mathbf{A}\mathbf{x} - \mathbf{b})^2 = \sum_{i=1}^n \left[\left(\sum_{j=1}^m a_{ij} x_j \right) - b_i \right]^2$$



$$E(\mathbf{x}) = (\mathbf{A}\mathbf{x} - \mathbf{b})^2 = \sum_{i=1}^n \left[\left(\sum_{j=1}^m a_{ij} x_j \right) - b_i \right]^2$$

$$0 = \frac{\partial E}{\partial x_1} = \sum_{i=1}^n 2 \left[\left(\sum_{j=1}^m a_{ij} x_j \right) - b_i \right] a_{i1}$$

$$= 2 \sum_{i=1}^n a_{i1} \sum_{j=1}^m a_{ij} x_j - 2 \sum_{i=1}^n a_{i1} b_i$$

$$0 = \frac{\partial E}{\partial \mathbf{x}} = 2(\mathbf{A}^{\mathsf{T}} \mathbf{A} \mathbf{x} - \mathbf{A}^{\mathsf{T}} \mathbf{b}) \longrightarrow \mathbf{A}^{\mathsf{T}} \mathbf{A} \mathbf{x} = \mathbf{A}^{\mathsf{T}} \mathbf{b}$$



$$(\mathbf{A}\mathbf{x} - \mathbf{b})^{2}$$

$$= (\mathbf{A}\mathbf{x} - \mathbf{b})^{T} (\mathbf{A}\mathbf{x} - \mathbf{b})$$

$$= ((\mathbf{A}\mathbf{x})^{T} - \mathbf{b}^{T}) (\mathbf{A}\mathbf{x} - \mathbf{b})$$

$$= (\mathbf{x}^{T}\mathbf{A}^{T} - \mathbf{b}^{T}) (\mathbf{A}\mathbf{x} - \mathbf{b})$$

$$= \mathbf{x}^{T}\mathbf{A}^{T}\mathbf{A}\mathbf{x} - \mathbf{b}^{T}\mathbf{A}\mathbf{x} - \mathbf{x}^{T}\mathbf{A}^{T}\mathbf{b} + \mathbf{b}^{T}\mathbf{b}$$

$$= \mathbf{x}^{T}\mathbf{A}^{T}\mathbf{A}\mathbf{x} - (\mathbf{A}^{T}\mathbf{b})^{T}\mathbf{x} - (\mathbf{A}^{T}\mathbf{b})^{T}\mathbf{x} + \mathbf{b}^{T}\mathbf{b}$$

$$\frac{\partial E}{\partial \mathbf{x}} = 2\mathbf{A}^{T}\mathbf{A}\mathbf{x} - 2\mathbf{A}^{T}\mathbf{b}$$



Determine pairwise alignment

- p'=Mp, where M is a transformation matrix, p and p' are feature matches
- For translation model, it is easier.

$$E = \sum_{i=1}^{n} \left[\left(m_1 + x_i - x_i^{'} \right)^2 + \left(m_2 + y_i - y_i^{'} \right)^2 \right]$$

$$0 = \frac{\partial E}{\partial m_1}$$

What if the match is false? Avoid impact of outliers.

RANSAC



- RANSAC = Random Sample Consensus
- An algorithm for robust fitting of models in the presence of many data outliers
- Compare to robust statistics
- Given N data points x_i , assume that mjority of them are generated from a model with parameters Θ , try to recover Θ .





Run(k times:) How many times?

- (1) draw n samples randomly Smaller is better
- (2) fit parameters Θ with these n samples
- (3) for each of other N-n points, calculate its distance to the fitted model, count the number of inlier points cOutput Θ with the largest c

How to define? Depends on the problem.





p: probability of real inliers

P: probability of success after k trials

$$P = 1 - (1 - p^{n})^{k}$$
n samples are all inliers
a failure

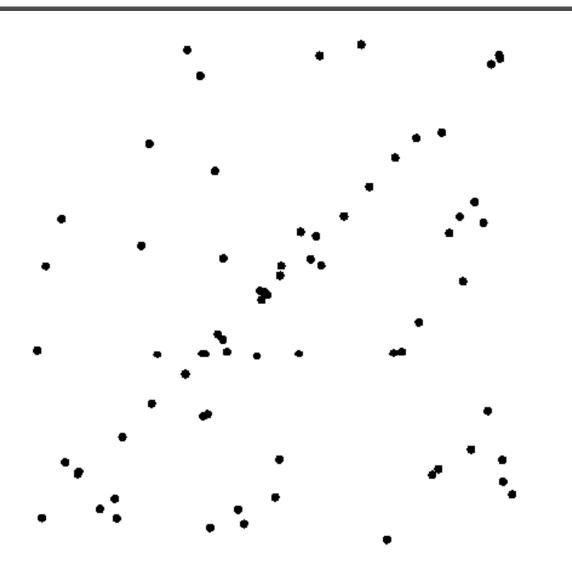
failure after k trials

$$k = \frac{\log(1-P)}{\log(1-p^n)}$$
 for $P = 0.99$

n	p	k
3	0.5	35
6	0.6	97
6	0.5	293

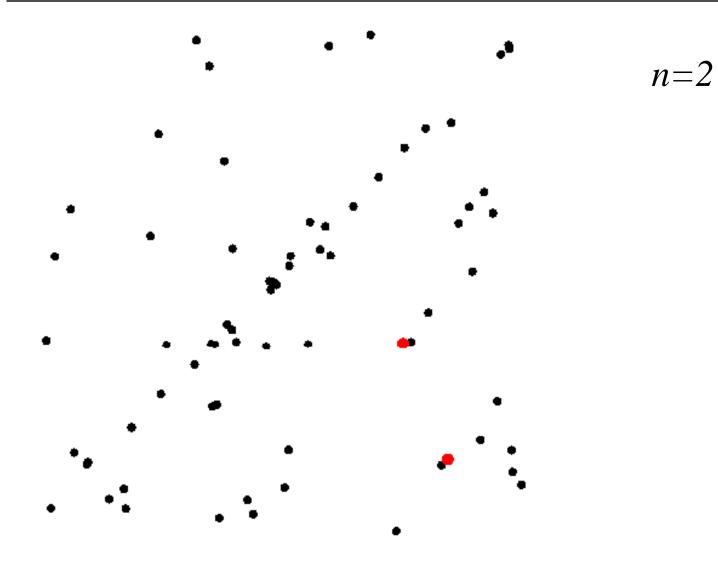


Example: line fitting



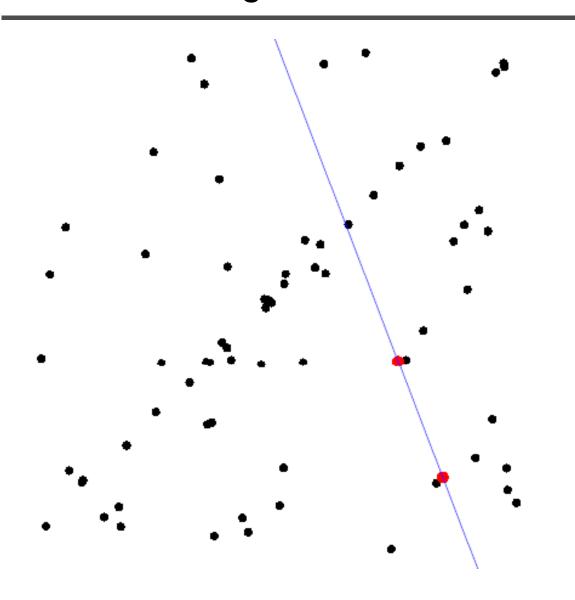
Example: line fitting





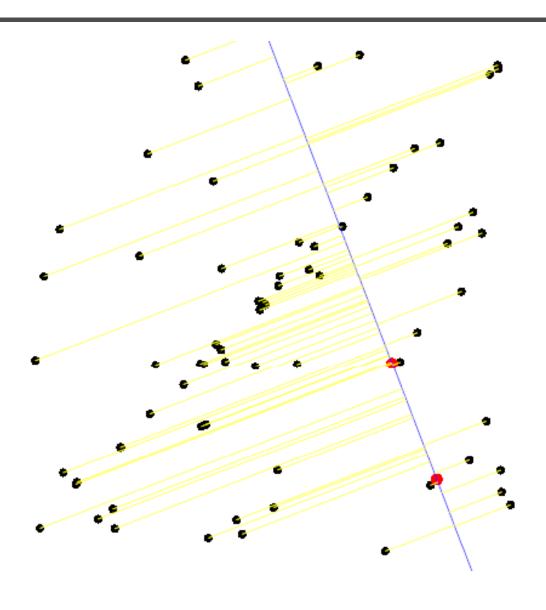
Model fitting





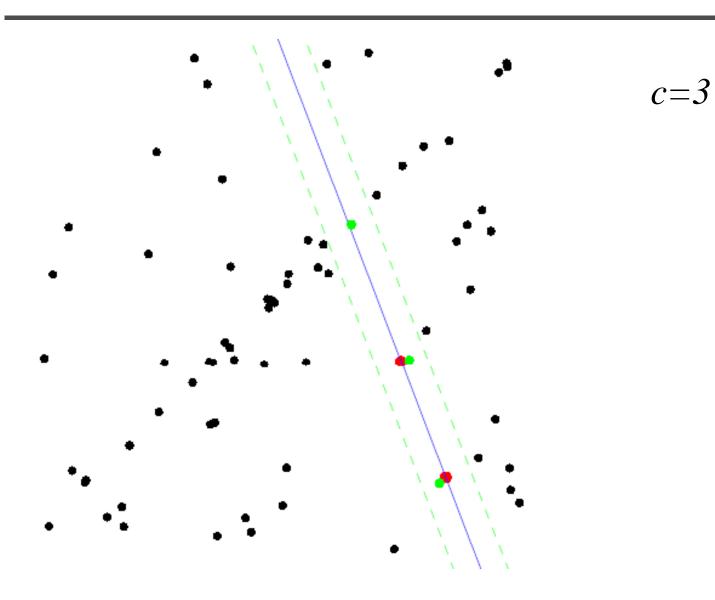
Measure distances





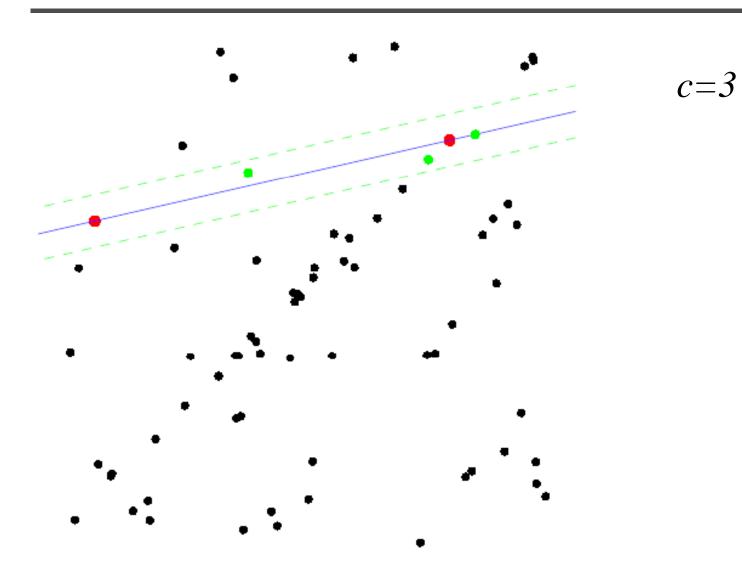
Count inliers





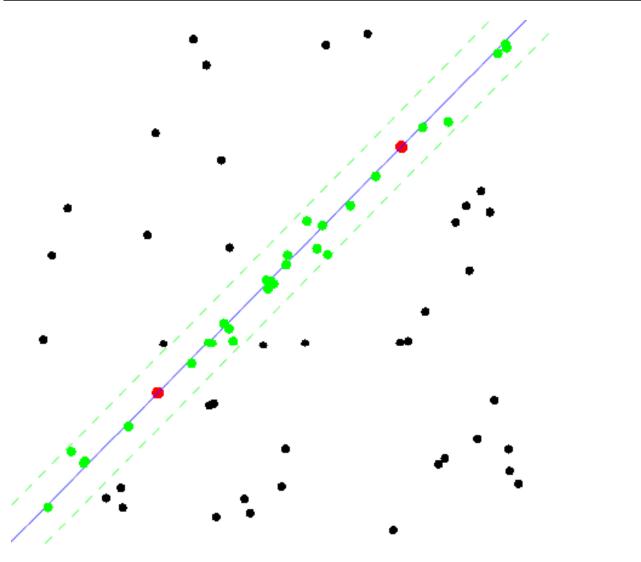
Another trial





The best model

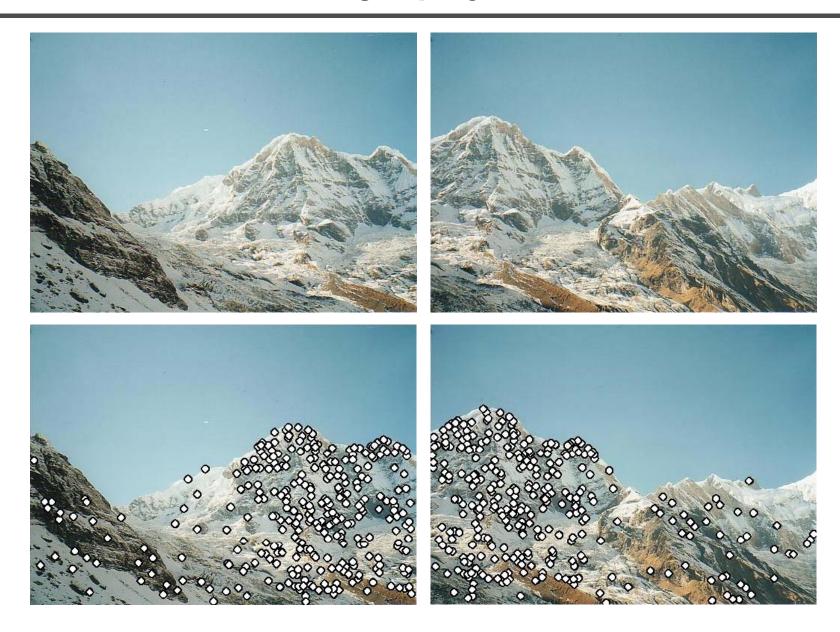




$$c = 15$$

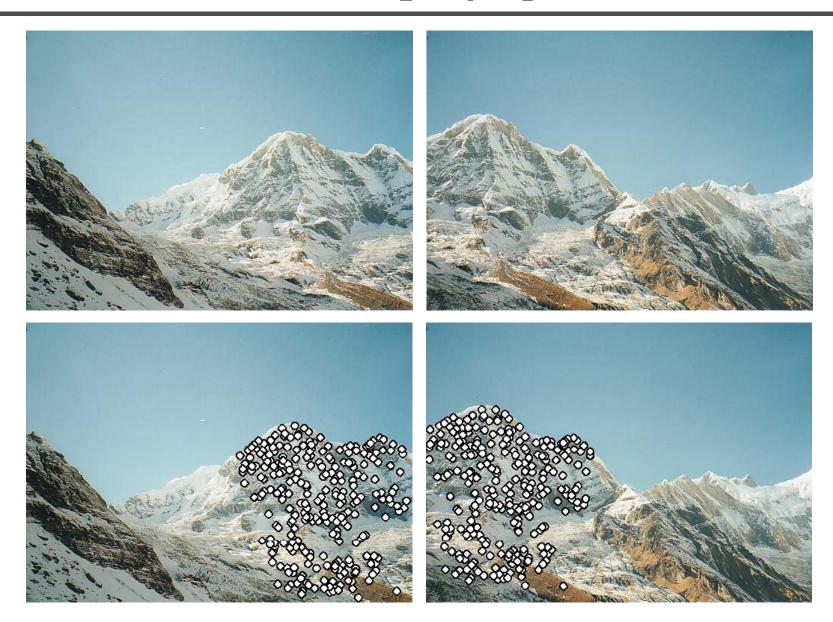


RANSAC for Homography





RANSAC for Homography





RANSAC for Homography









Applications of panorama in VFX

- Background plates
- Image-based lighting



Troy (image-based lighting)



http://www.cgnetworks.com/story_custom.php?story_id=2195&page=4



Spiderman 2 (background plate)

