

Image stitching

Digital Visual Effects

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with slides by Richard Szeliski, Steve Seitz, Matthew Brown and Vaclav Hlavac

Image stitching

- Stitching = alignment + blending

↑
geometrical
registration

↑
photometric
registration



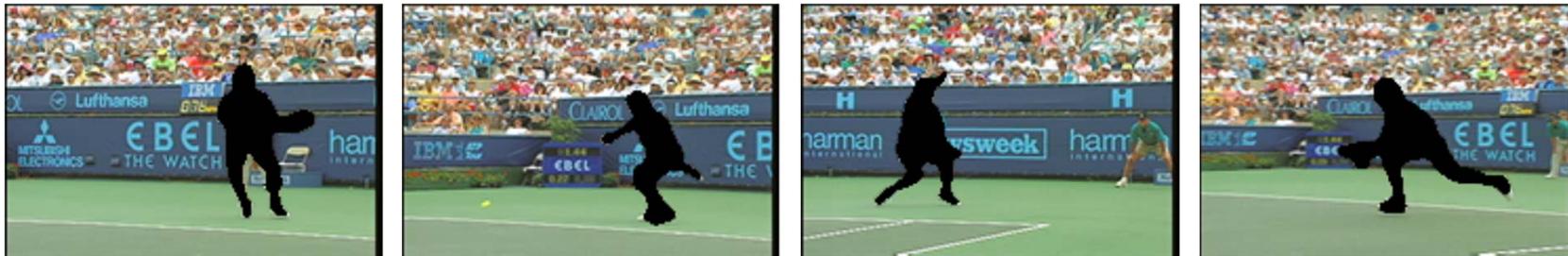
Applications of image stitching

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

Video summarization



Video compression



Object removal



input video

Object removal



background estimation

Panorama creation



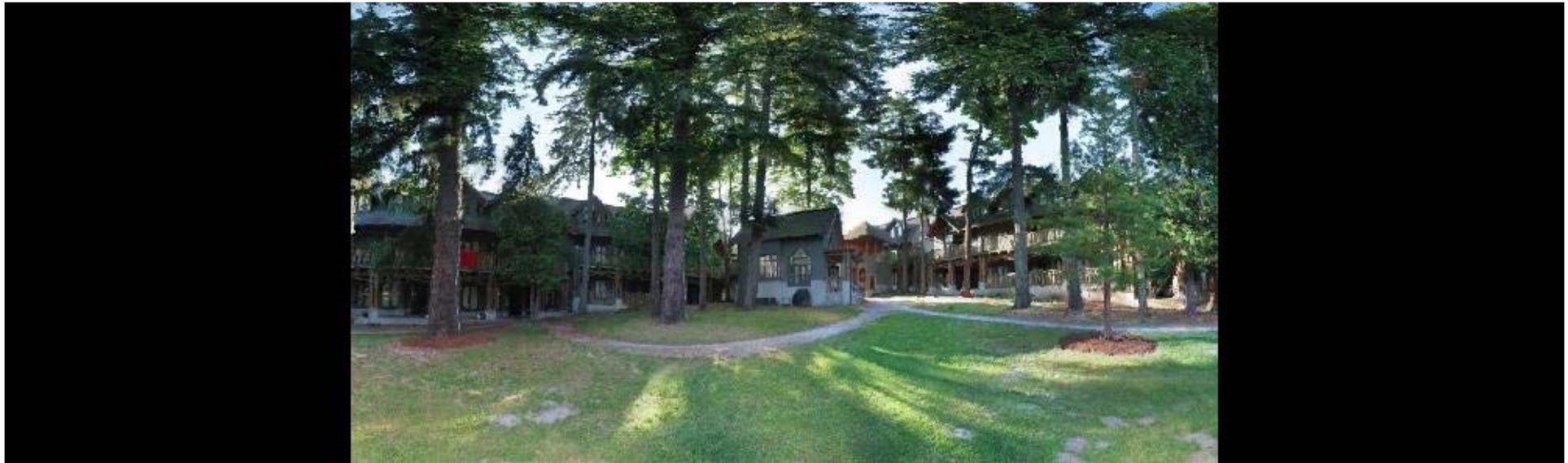
Why panorama?

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



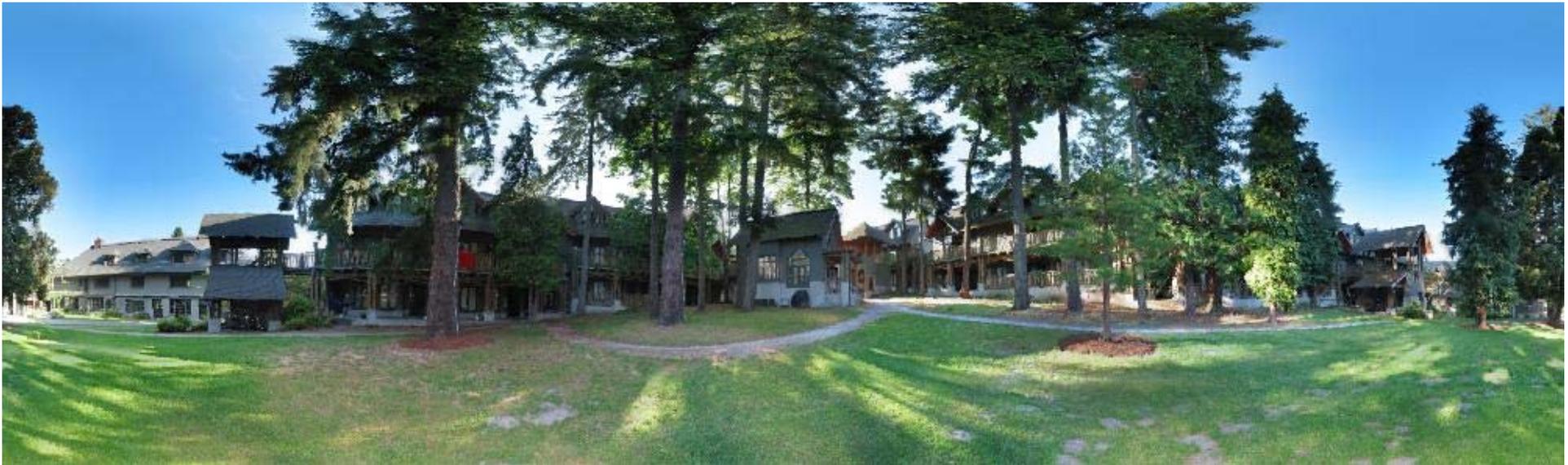
Why panorama?

- Are you getting the whole picture?
 - Compact Camera FOV = 50 x 35°
 - Human FOV = 200 x 135°



Why panorama?

- Are you getting the whole picture?
 - Compact Camera FOV = 50 x 35°
 - Human FOV = 200 x 135°
 - Panoramic Mosaic = 360 x 180°



Panorama examples

- Similar to HDR, it is a topic of computational photography, seeking ways to build a better camera using either hardware or 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>

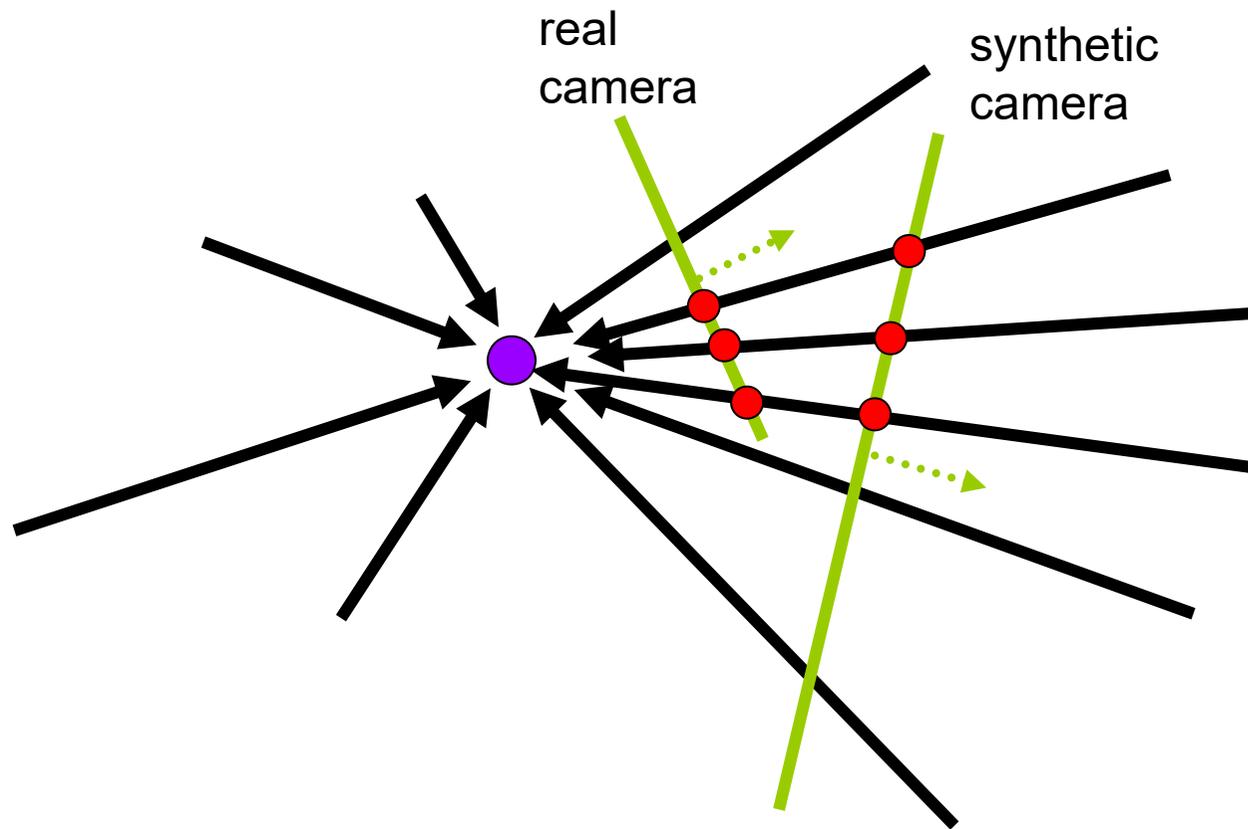
<http://www.360cities.net/>

<http://maps.google.com.tw/>

What can be globally aligned?

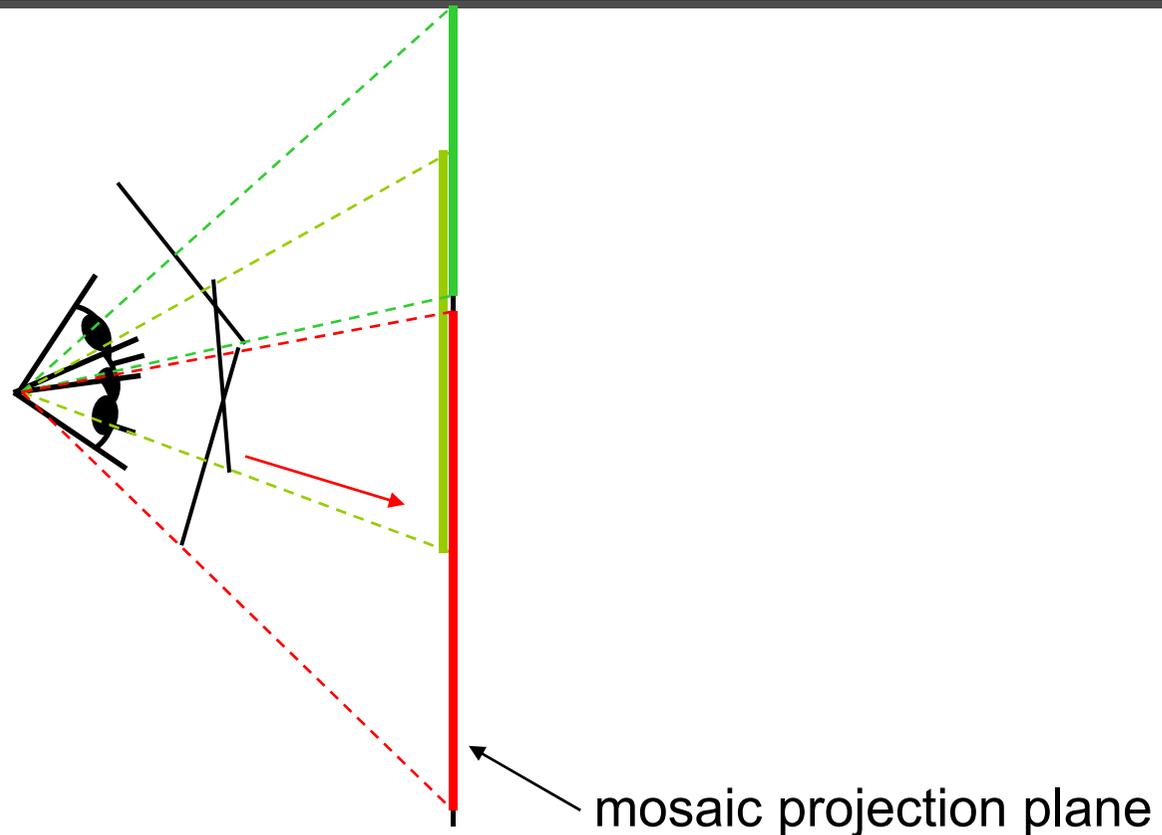
- In image stitching, we seek for a matrix to globally warp one image into another. Are any two images of the same scene can be aligned this way?
 - Images captured with the same center of projection
 - A planar scene or far-away scene

A pencil of rays contains all views



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

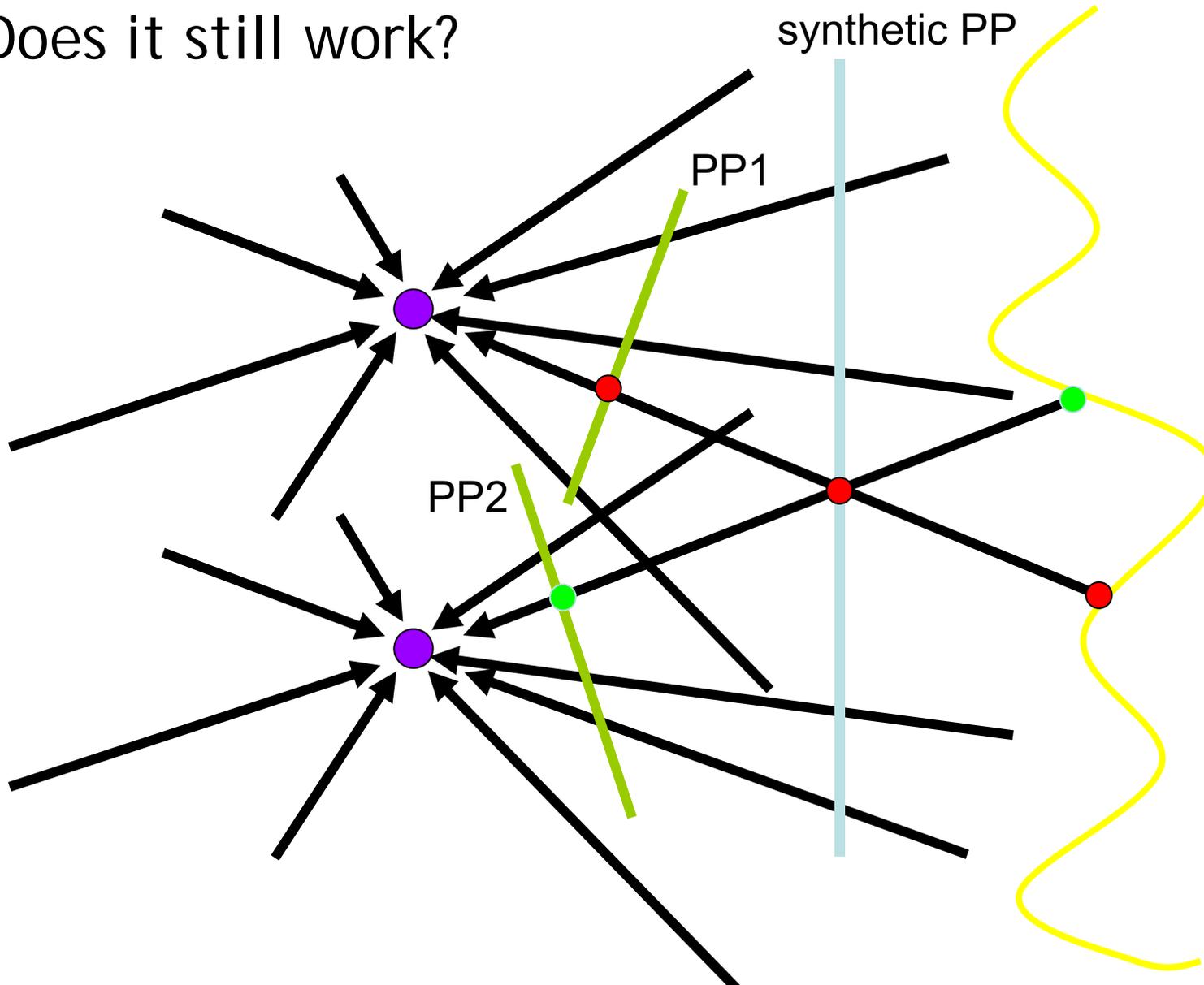
Mosaic as an image reprojection



- 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

- Does it still work?

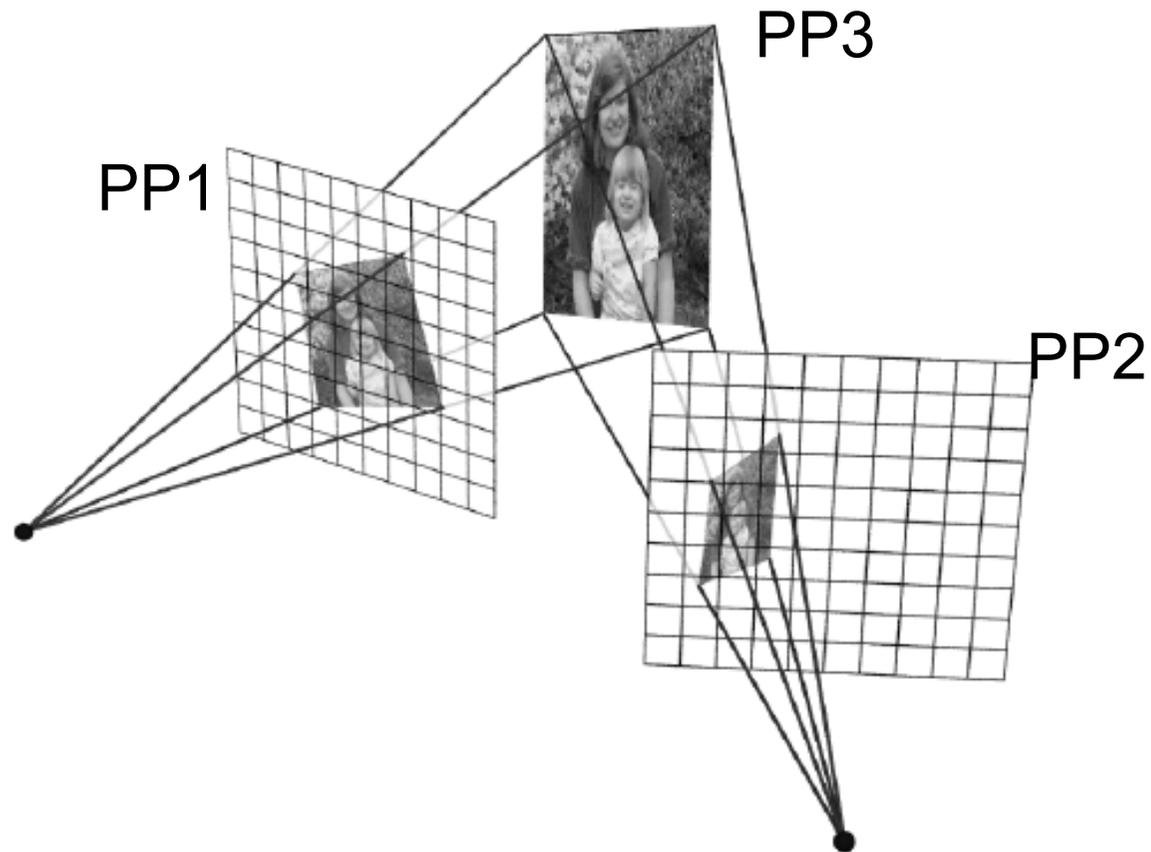


What cannot

- The scene with depth variations and the camera has movement



Planar scene (or a faraway one)

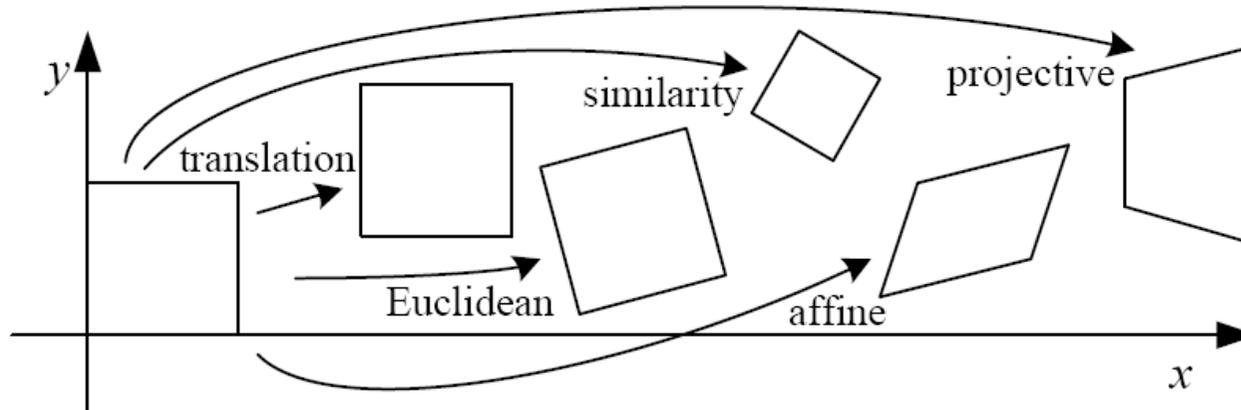


- PP3 is a projection plane of both centers of projection, so we are OK!
- This is how big aerial photographs are made

Motion models

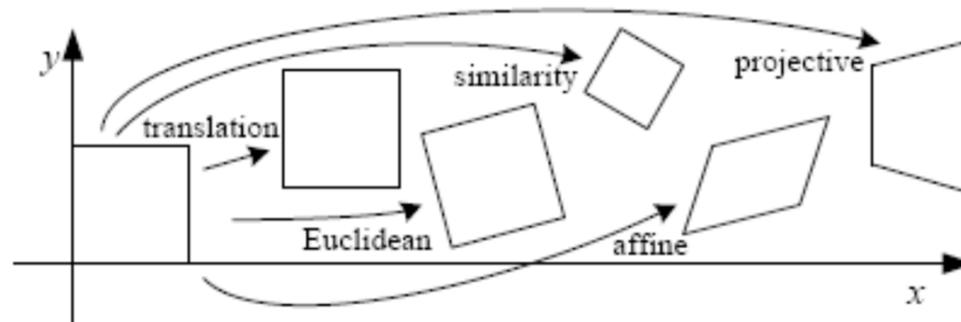
- Parametric models as the assumptions on the relation between two images.

2D Motion models

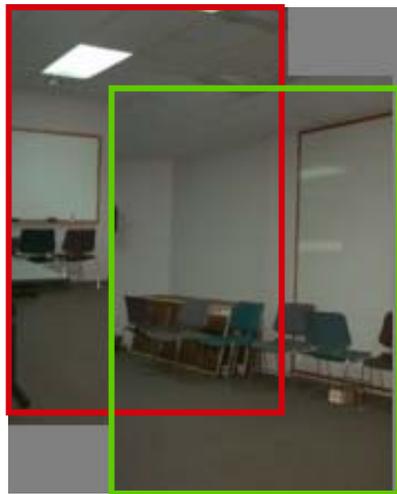


Name	Matrix	# D.O.F.	Preserves:	Icon
translation	$\begin{bmatrix} \mathbf{I} & \mathbf{t} \end{bmatrix}_{2 \times 3}$	2	orientation + ...	
rigid (Euclidean)	$\begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix}_{2 \times 3}$	3	lengths + ...	
similarity	$\begin{bmatrix} s\mathbf{R} & \mathbf{t} \end{bmatrix}_{2 \times 3}$	4	angles + ...	
affine	$\begin{bmatrix} \mathbf{A} \end{bmatrix}_{2 \times 3}$	6	parallelism + ...	
projective	$\begin{bmatrix} \tilde{\mathbf{H}} \end{bmatrix}_{3 \times 3}$	8	straight lines	

Motion models

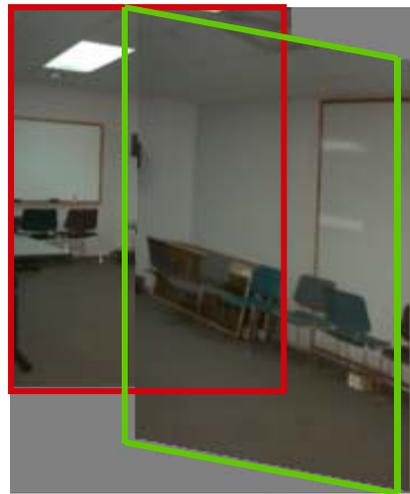


Translation



2 unknowns

Affine



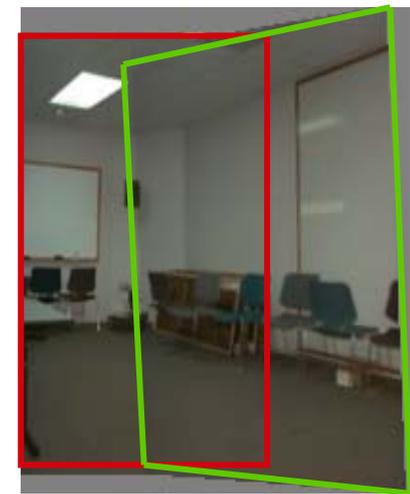
6 unknowns

Perspective



8 unknowns

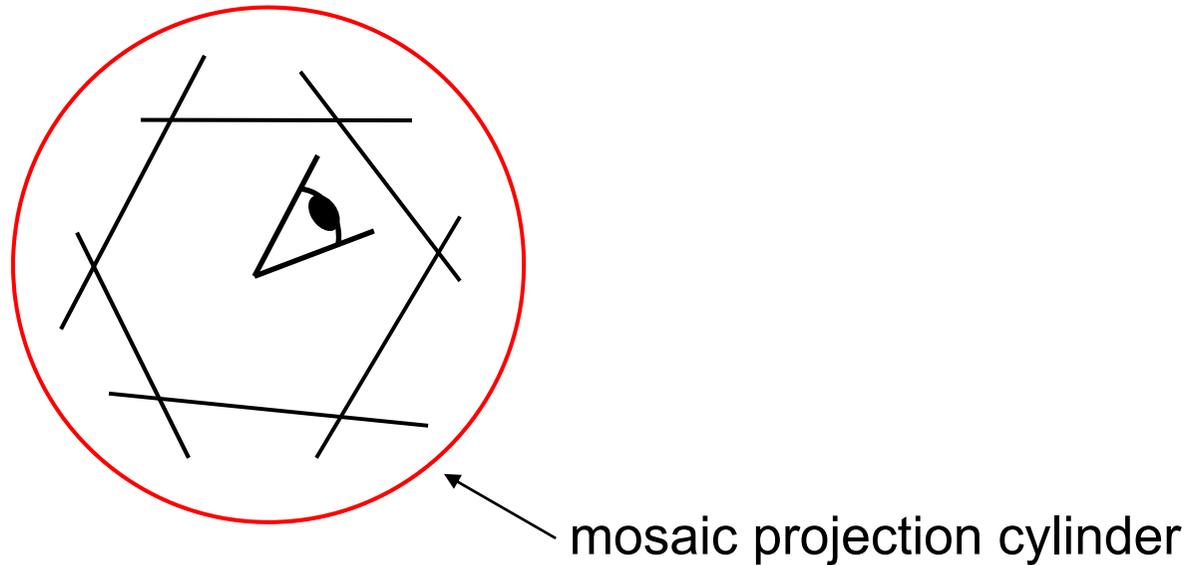
3D rotation



3 unknowns

A case study: cylindrical panorama

- What if you want a 360° field of view?



Cylindrical panoramas



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

applet

- <http://graphics.stanford.edu/courses/cs178/applets/projection.html>

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!

Taking pictures

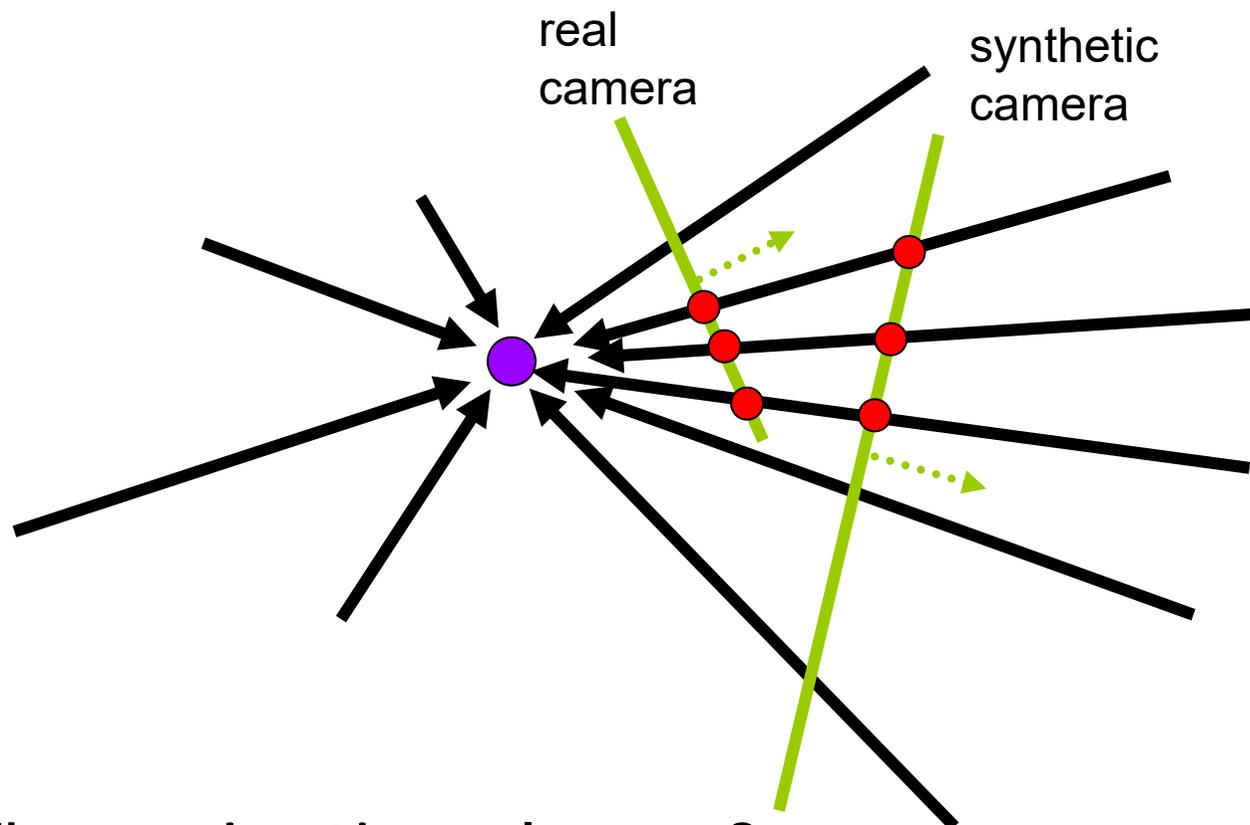


Kaidan panoramic tripod head

Translation model

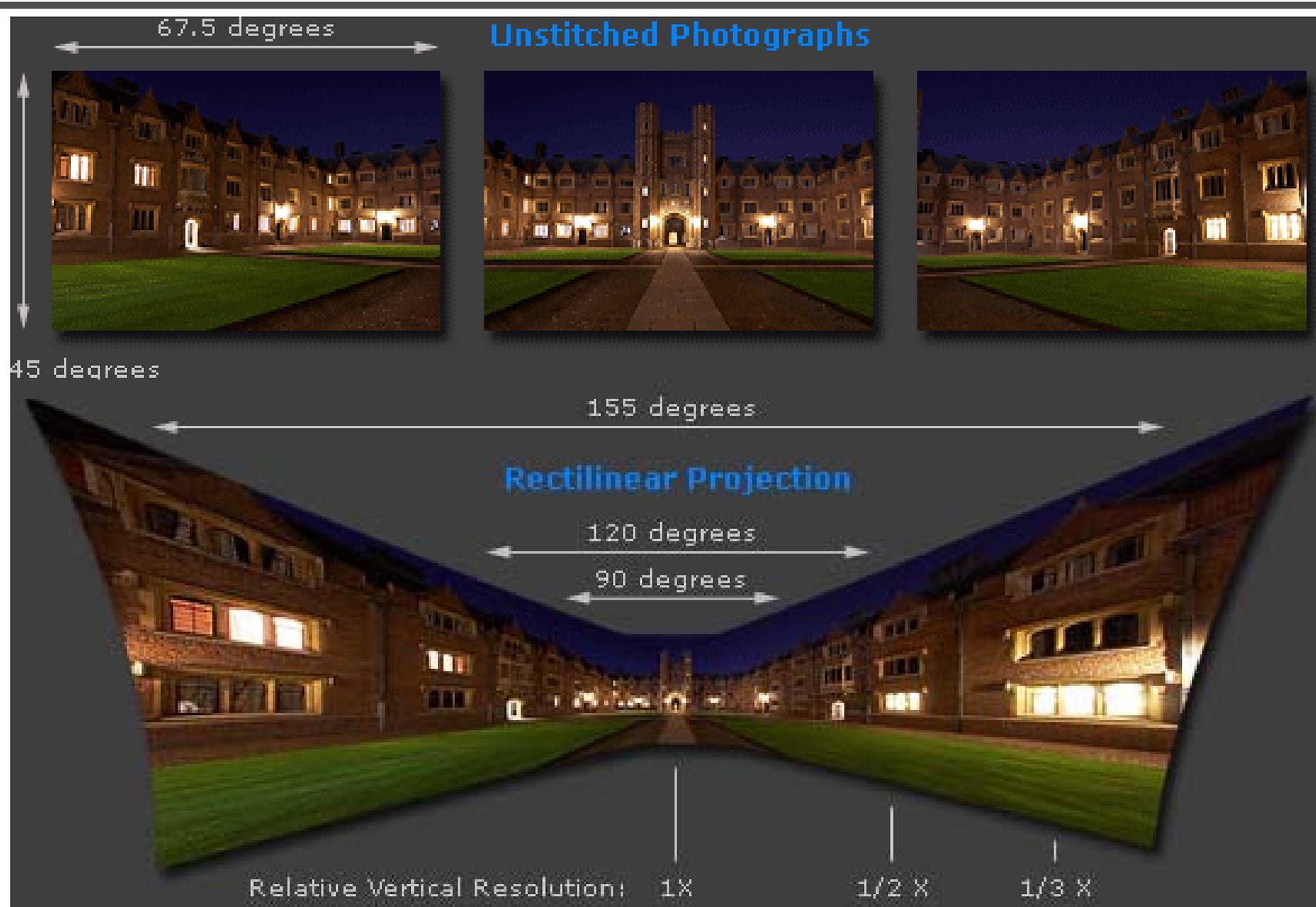


Where should the synthetic camera be

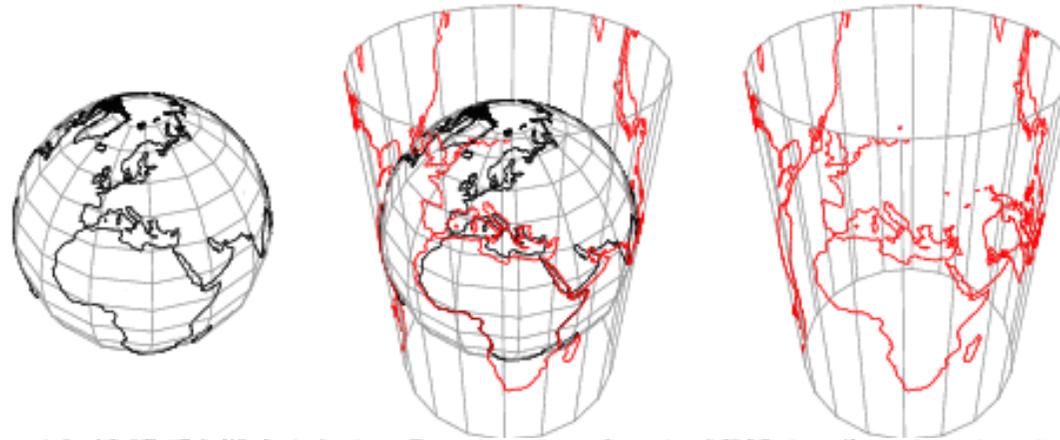


- The projection plane of some camera
- Onto a cylinder

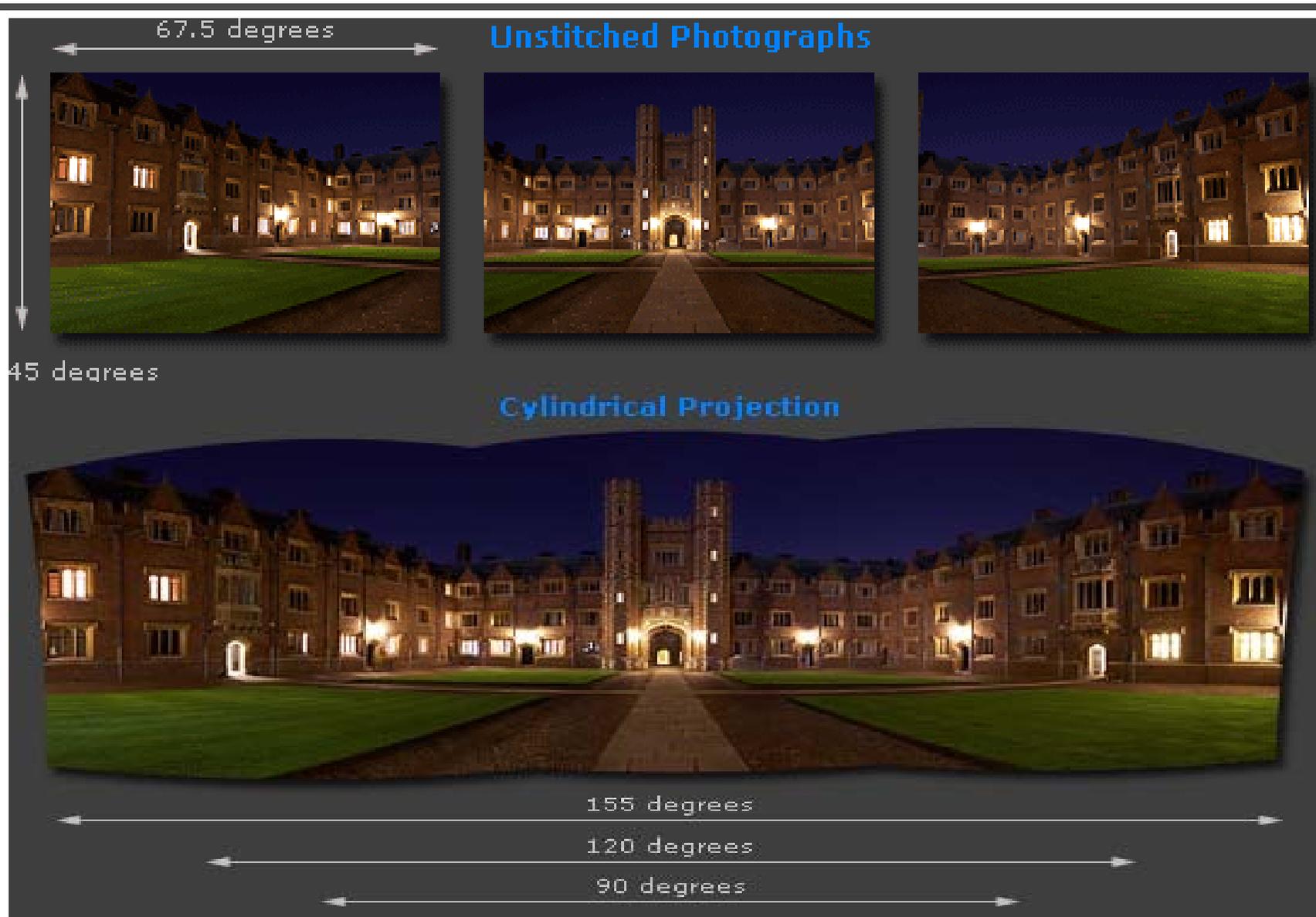
Cylindrical projection



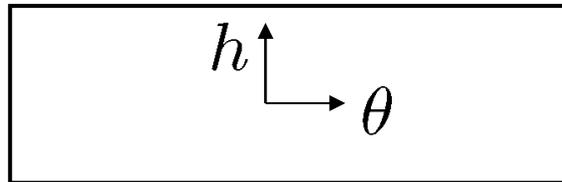
Cylindrical projection



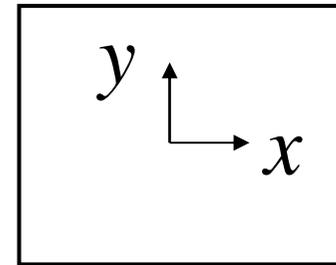
Cylindrical projection



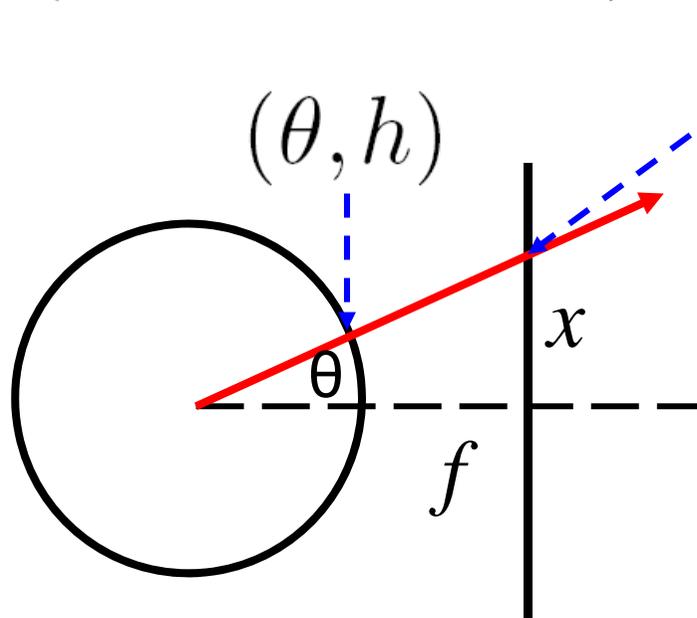
Cylindrical projection



unwrapped cylinder

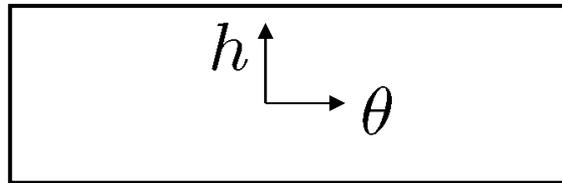


$$(\sin \theta, h, \cos \theta) \propto (x, y, f)$$

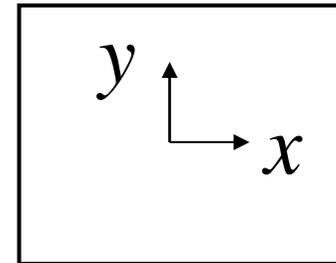


$$\theta = \tan^{-1} \frac{x}{f}$$

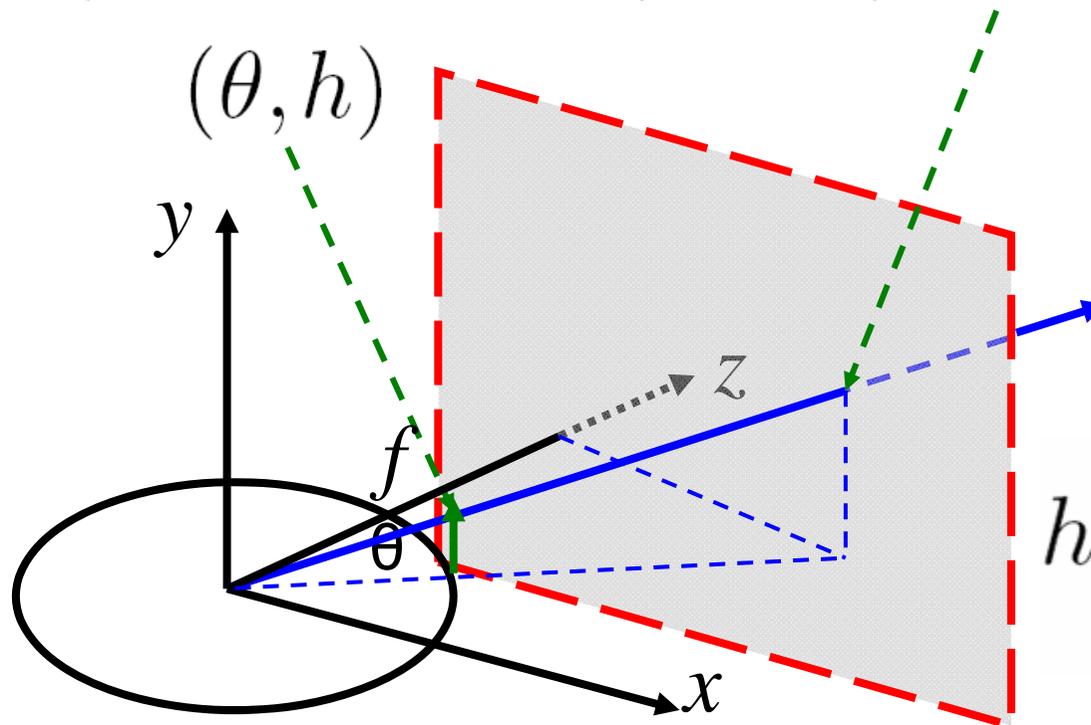
Cylindrical projection



unwrapped cylinder

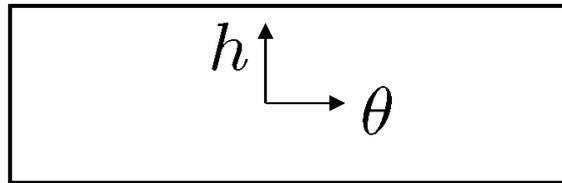


$$(\sin \theta, h, \cos \theta) \propto (x, y, f)$$

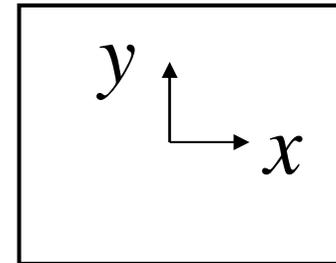


$$h = \frac{y}{\sqrt{x^2 + f^2}}$$

Cylindrical projection



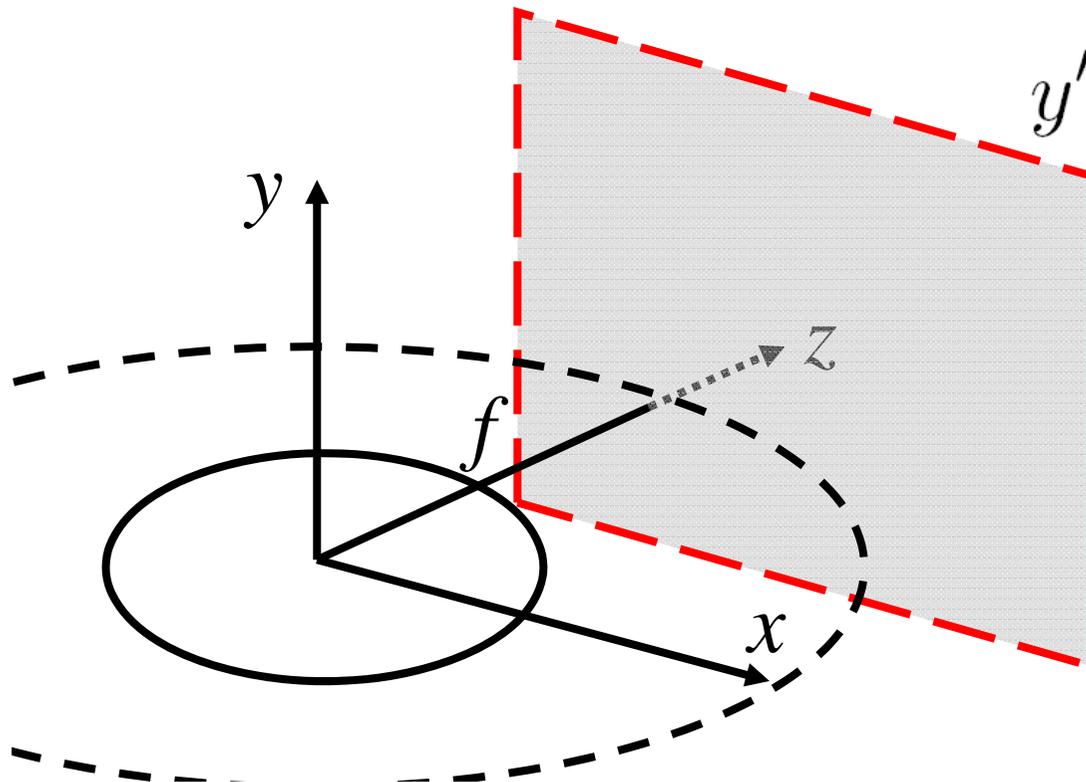
unwrapped cylinder



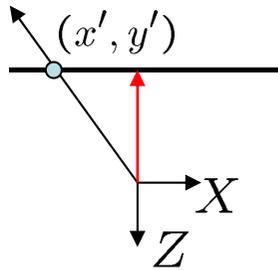
$$x' = s\theta = s \tan^{-1} \frac{x}{f}$$

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

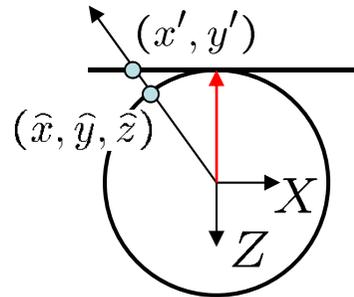
$s=f$ gives less distortion



Cylindrical reprojection



top-down view



Focal length – the dirty secret...

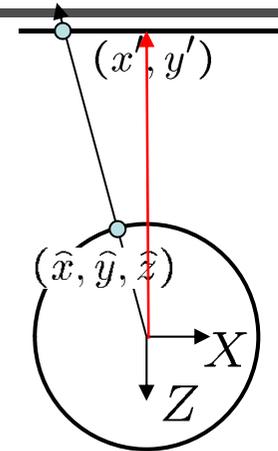
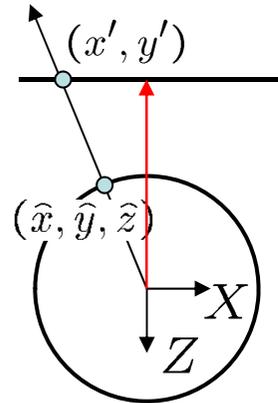


Image 384x300



f = 180 (pixels)

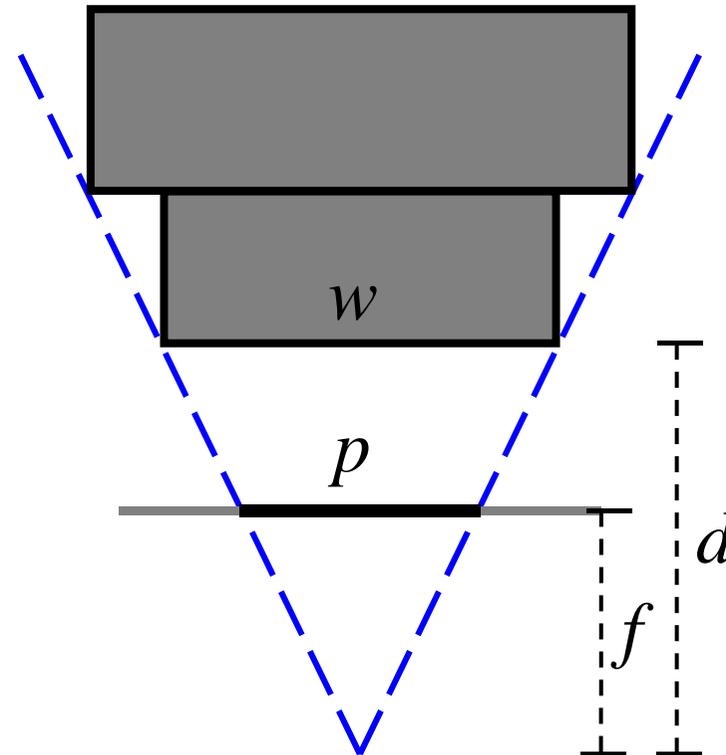
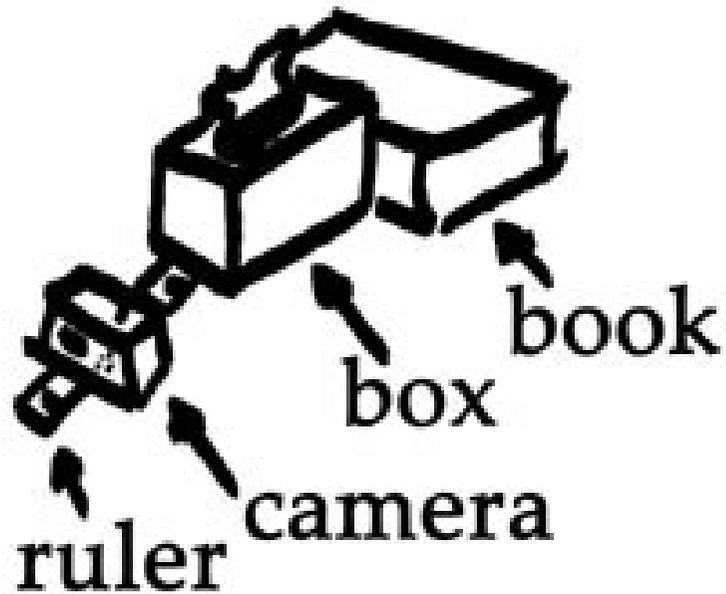


f = 280



f = 380

A simple method for estimating f

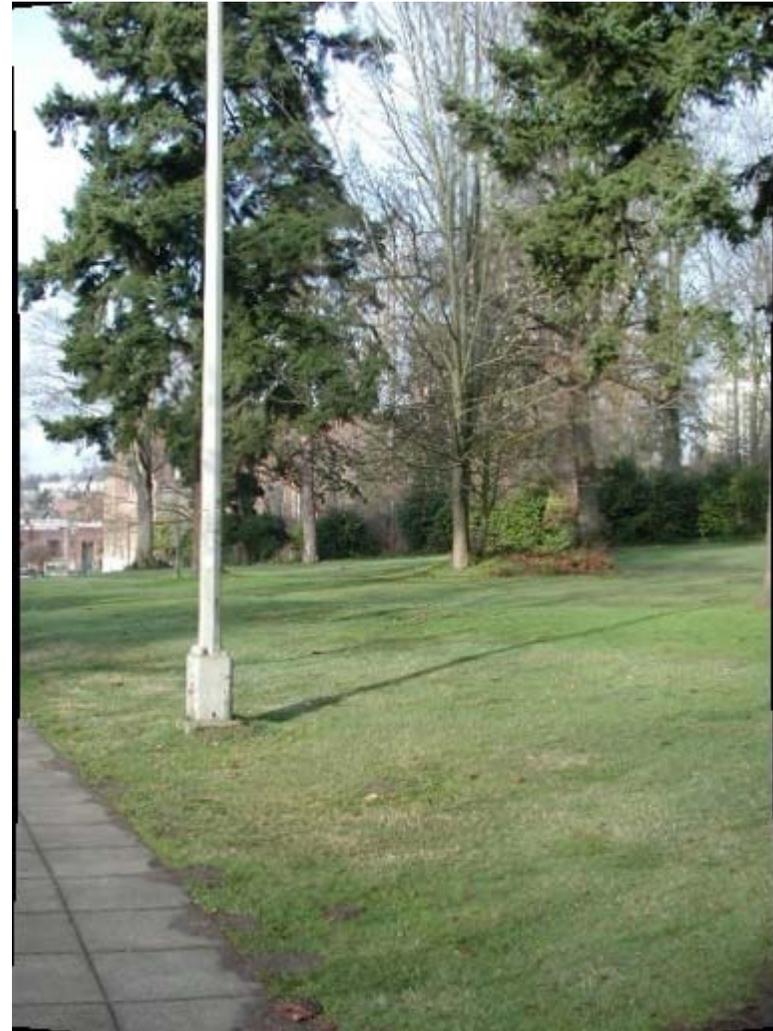


Or, you can use other software, such as AutoStich, to help.

Input images



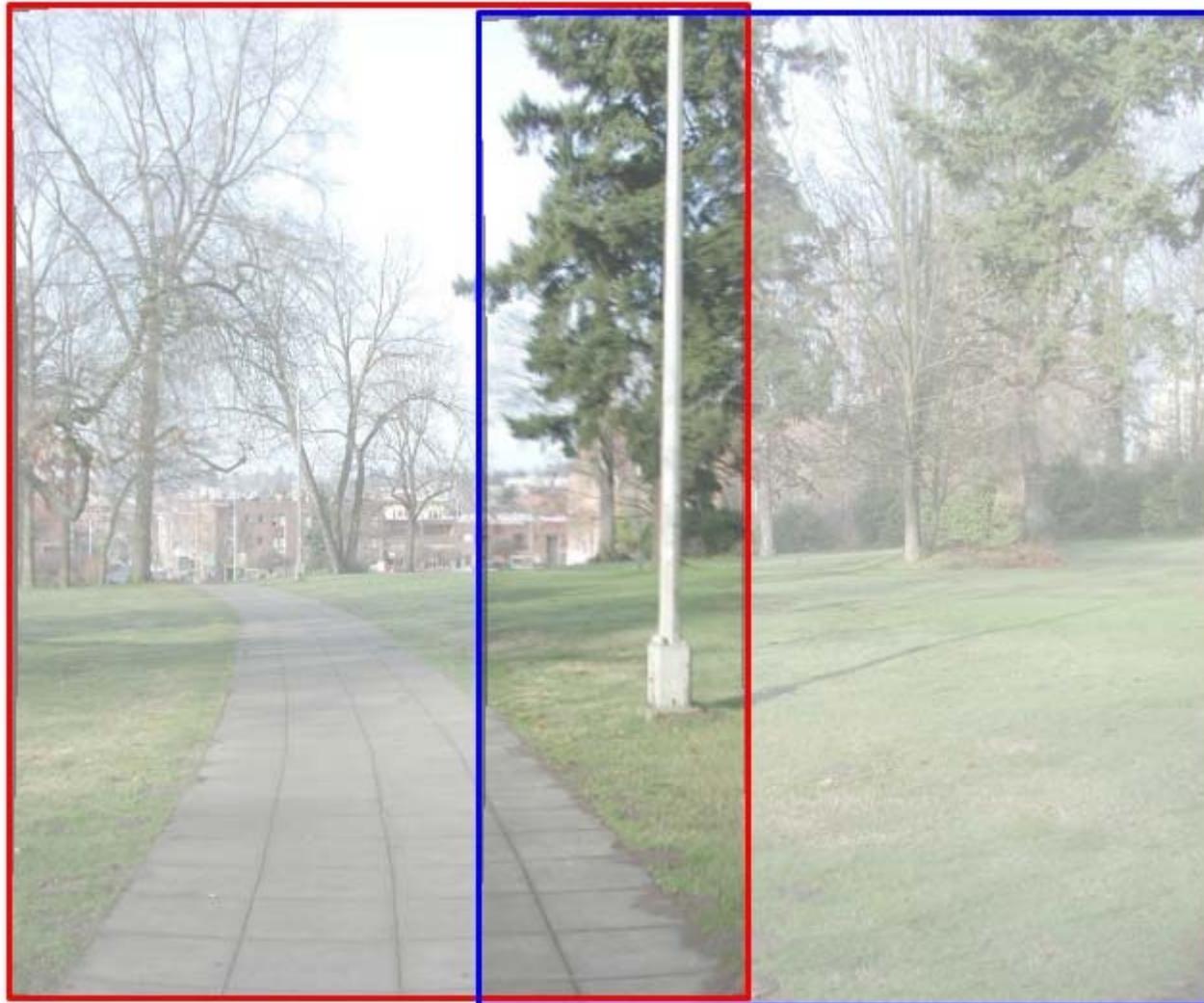
Cylindrical warping



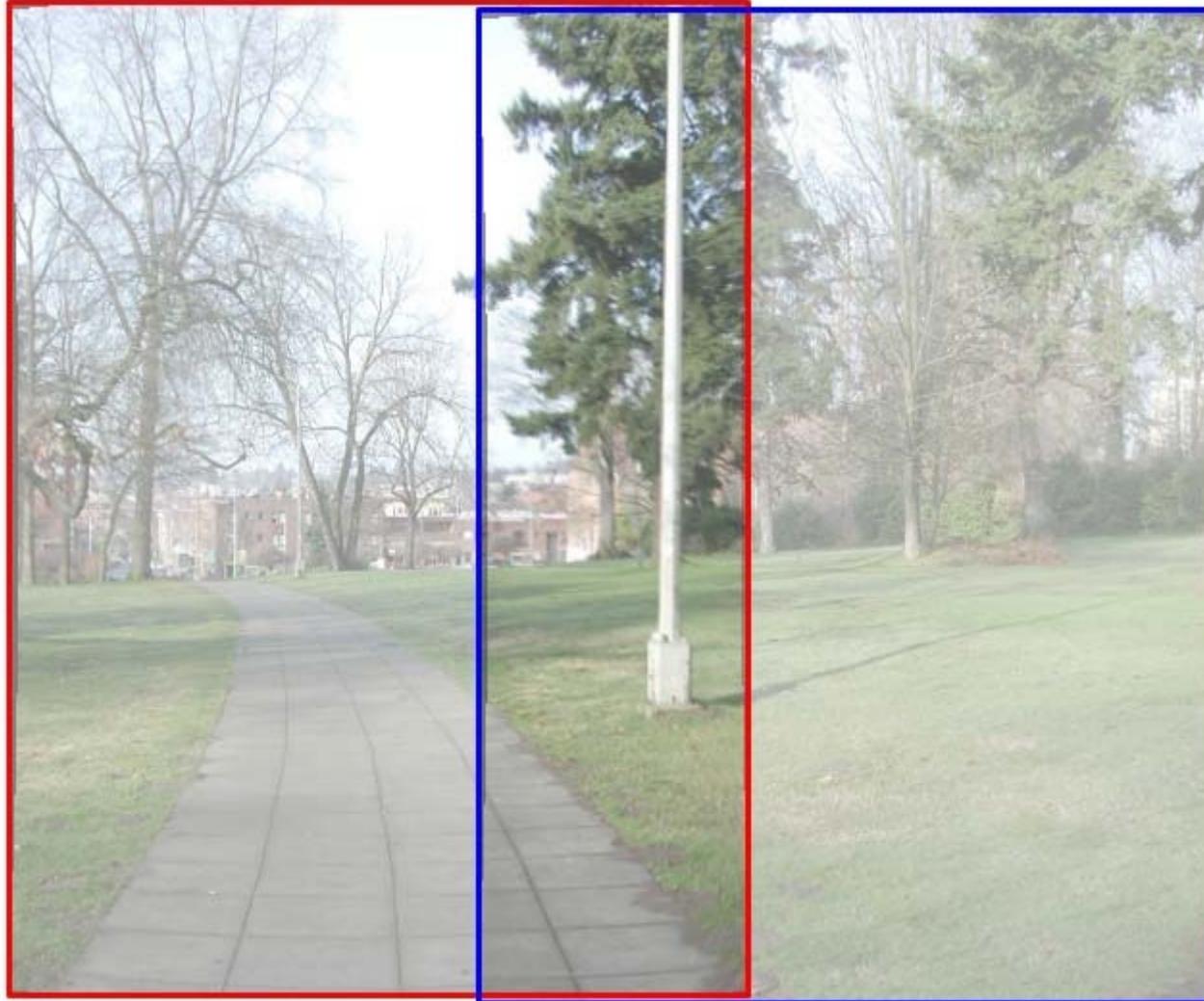
Blending

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

Blending



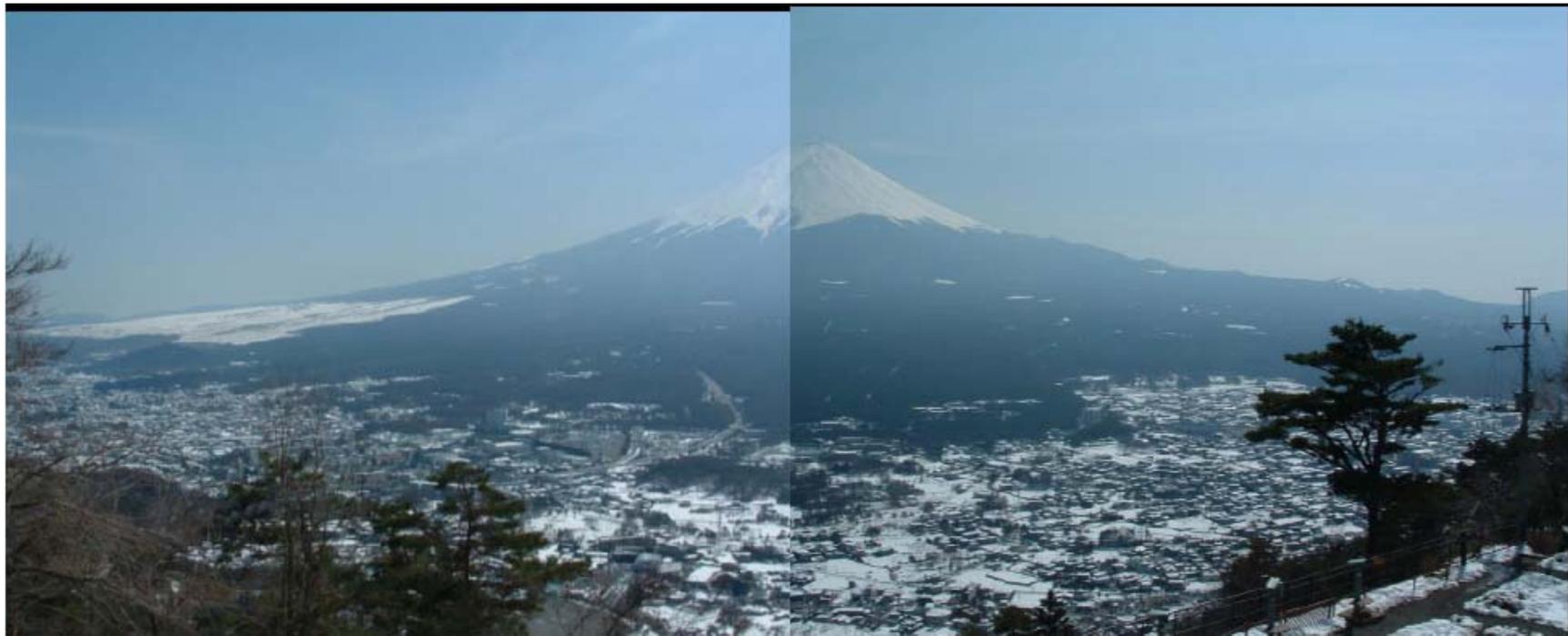
Blending



Blending



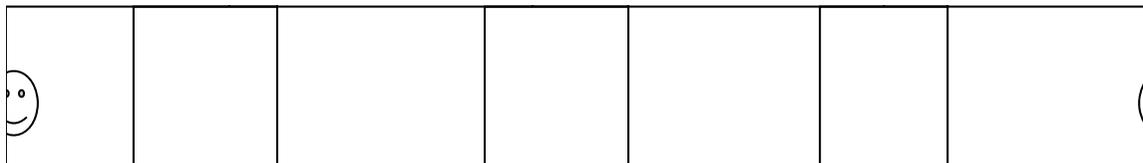
Gradient-domain stitching



Gradient-domain stitching

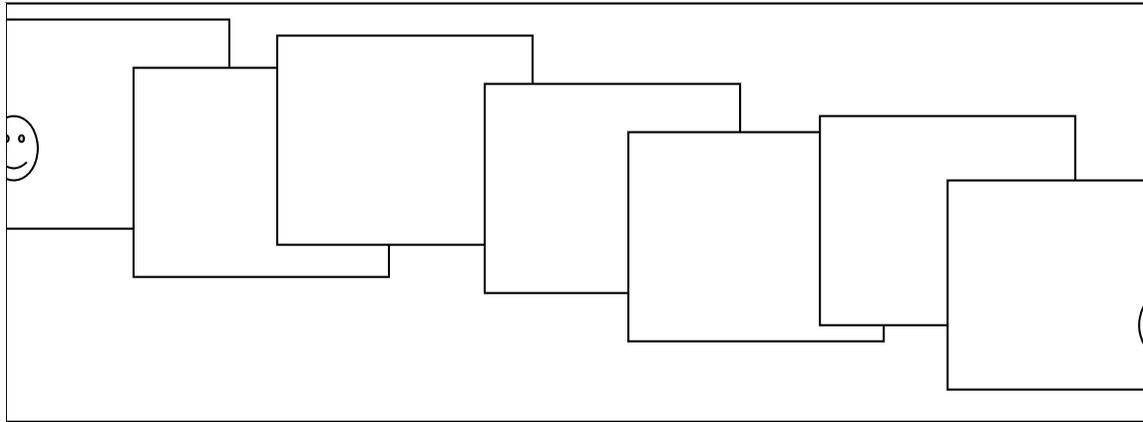


Assembling the panorama



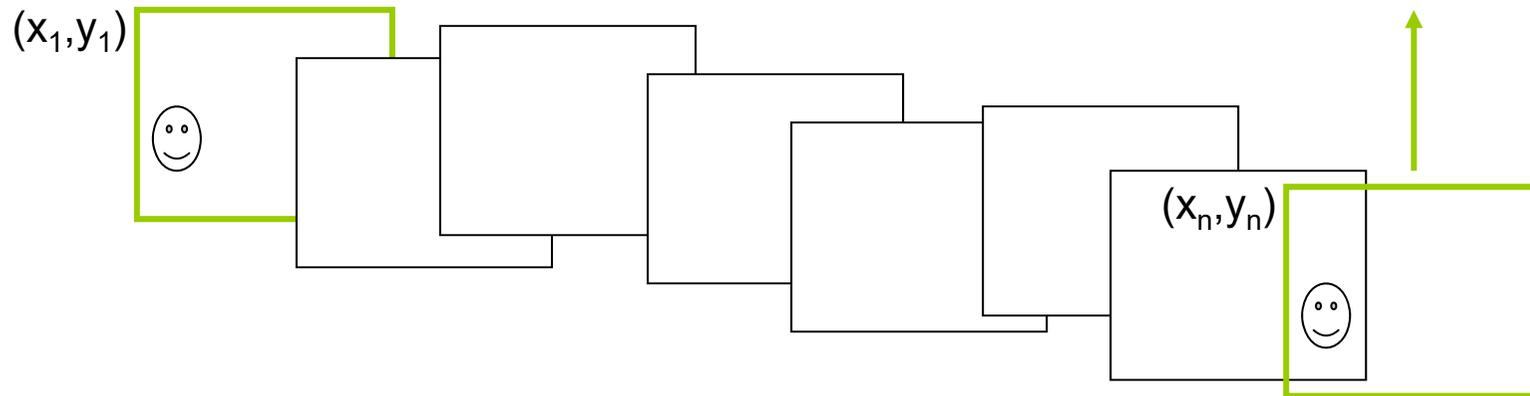
- Stitch pairs together, blend, then crop

Problem: Drift



- Error accumulation
 - small errors accumulate over time

Problem: Drift



- Solution

- 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”

- copy of first image

End-to-end alignment and crop



Rectangling panoramas



(a) input panorama



(b) image completion



(c) cropping



(d) our content-aware warping

[video](#)

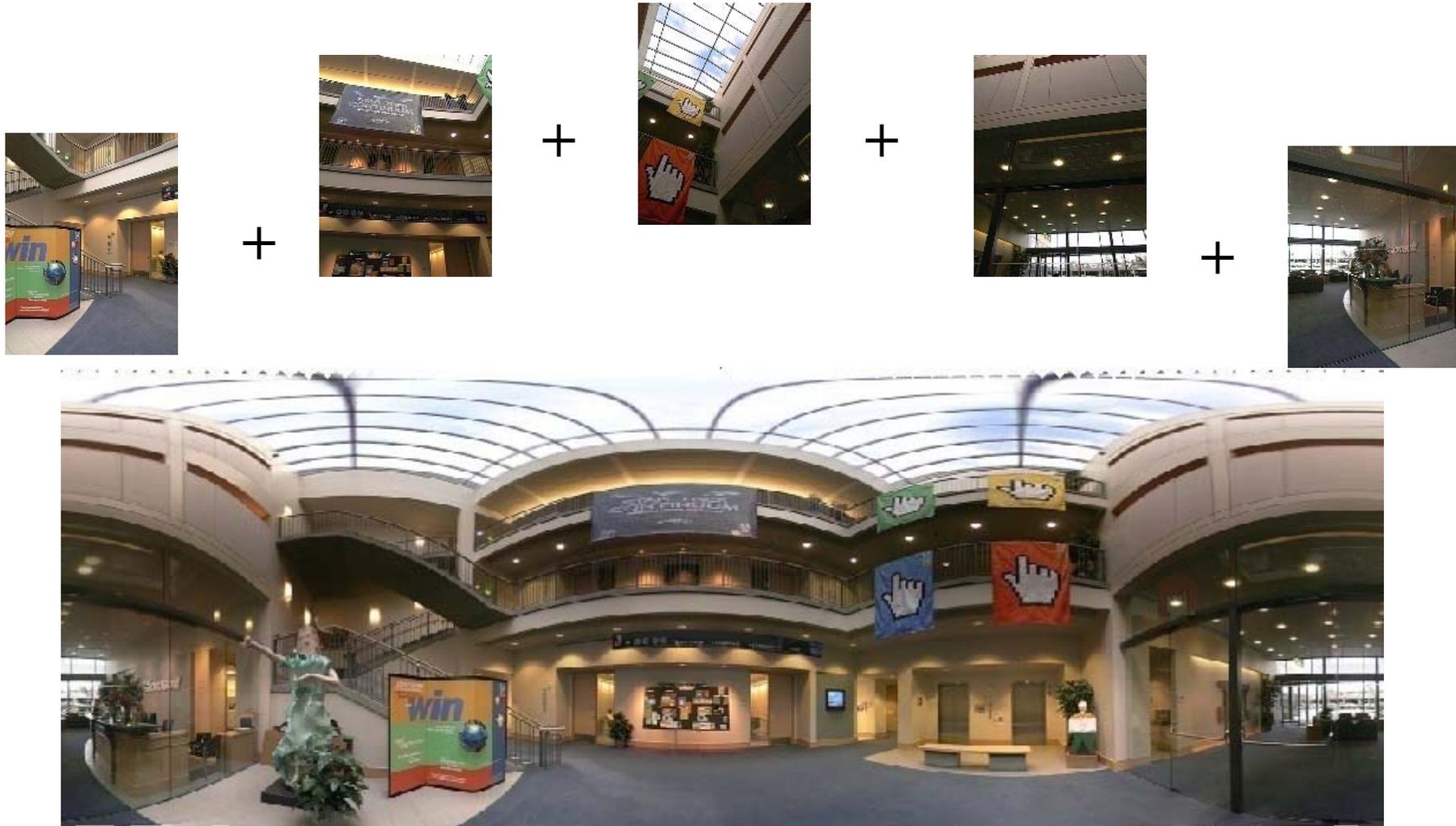
Rectangling panoramas



Rectangling panoramas

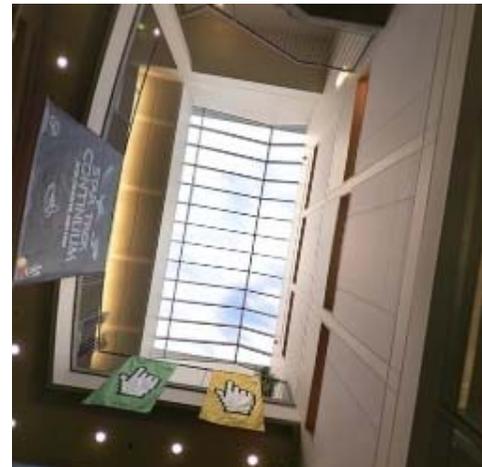


Viewer: panorama



example: <http://www.cs.washington.edu/education/courses/cse590ss/01wi/projects/project1/students/dougz/index.html>

Viewer: texture mapped model



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

365-GB panorama (biggest on the earth)



**Mont Blanc / Canon 70D / 70,000 images / [video](#) [web](#)
2-week shooting / 2-month processing [London](#)**

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

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \quad \begin{aligned} x_1 m_{11} + y_1 m_{12} &= x'_1 \\ x_1 m_{21} + y_1 m_{22} &= y'_1 \end{aligned}$$

- Overdetermined system

$$\begin{pmatrix} x_1 & y_1 & 0 & 0 \\ 0 & 0 & x_1 & y_1 \\ x_2 & y_2 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots \\ x_n & y_n & 0 & 0 \\ 0 & 0 & x_n & y_n \end{pmatrix} \begin{pmatrix} m_{11} \\ m_{12} \\ m_{21} \\ m_{22} \end{pmatrix} = \begin{pmatrix} x'_1 \\ y'_1 \\ x'_2 \\ \vdots \\ x'_n \\ y'_n \end{pmatrix}$$

Normal equation

Given an overdetermined system

$$\mathbf{Ax} = \mathbf{b}$$

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

$$\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$$

Why?

Normal equation

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

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

$n \times m$, n equations, m variables

Normal equation

$$\mathbf{Ax} - \mathbf{b} = \begin{bmatrix} \sum_{j=1}^m a_{1j} x_j \\ \vdots \\ \sum_{j=1}^m a_{ij} x_j \\ \vdots \\ \sum_{j=1}^m a_{nj} x_j \end{bmatrix} - \begin{bmatrix} b_1 \\ \vdots \\ b_i \\ \vdots \\ b_n \end{bmatrix} = \begin{bmatrix} \left(\sum_{j=1}^m a_{1j} x_j \right) - b_1 \\ \vdots \\ \left(\sum_{j=1}^m a_{ij} x_j \right) - b_i \\ \vdots \\ \left(\sum_{j=1}^m a_{nj} x_j \right) - b_n \end{bmatrix}$$
$$E(\mathbf{x}) = (\mathbf{Ax} - \mathbf{b})^2 = \sum_{i=1}^n \left[\left(\sum_{j=1}^m a_{ij} x_j \right) - b_i \right]^2$$

Normal equation

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

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

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

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

Normal equation

$$(\mathbf{Ax} - \mathbf{b})^2$$

Normal equation

$$\begin{aligned} & (\mathbf{Ax} - \mathbf{b})^2 \\ &= (\mathbf{Ax} - \mathbf{b})^T (\mathbf{Ax} - \mathbf{b}) \\ &= \left((\mathbf{Ax})^T - \mathbf{b}^T \right) (\mathbf{Ax} - \mathbf{b}) \\ &= \left(\mathbf{x}^T \mathbf{A}^T - \mathbf{b}^T \right) (\mathbf{Ax} - \mathbf{b}) \\ &= \mathbf{x}^T \mathbf{A}^T \mathbf{Ax} - \mathbf{b}^T \mathbf{Ax} - \mathbf{x}^T \mathbf{A}^T \mathbf{b} + \mathbf{b}^T \mathbf{b} \\ &= \mathbf{x}^T \mathbf{A}^T \mathbf{Ax} - \left(\mathbf{A}^T \mathbf{b} \right)^T \mathbf{x} - \left(\mathbf{A}^T \mathbf{b} \right)^T \mathbf{x} + \mathbf{b}^T \mathbf{b} \end{aligned}$$

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

Determine pairwise alignment

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

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

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

- 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 majority of them are generated from a model with parameters Θ , try to recover Θ .

RANSAC algorithm

Run k times:

(1) draw n samples randomly

(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 c

Output Θ with the largest c

← How many times?

← How big?

Smaller is better

← How to define?

Depends on the problem.

How to determine k

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

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

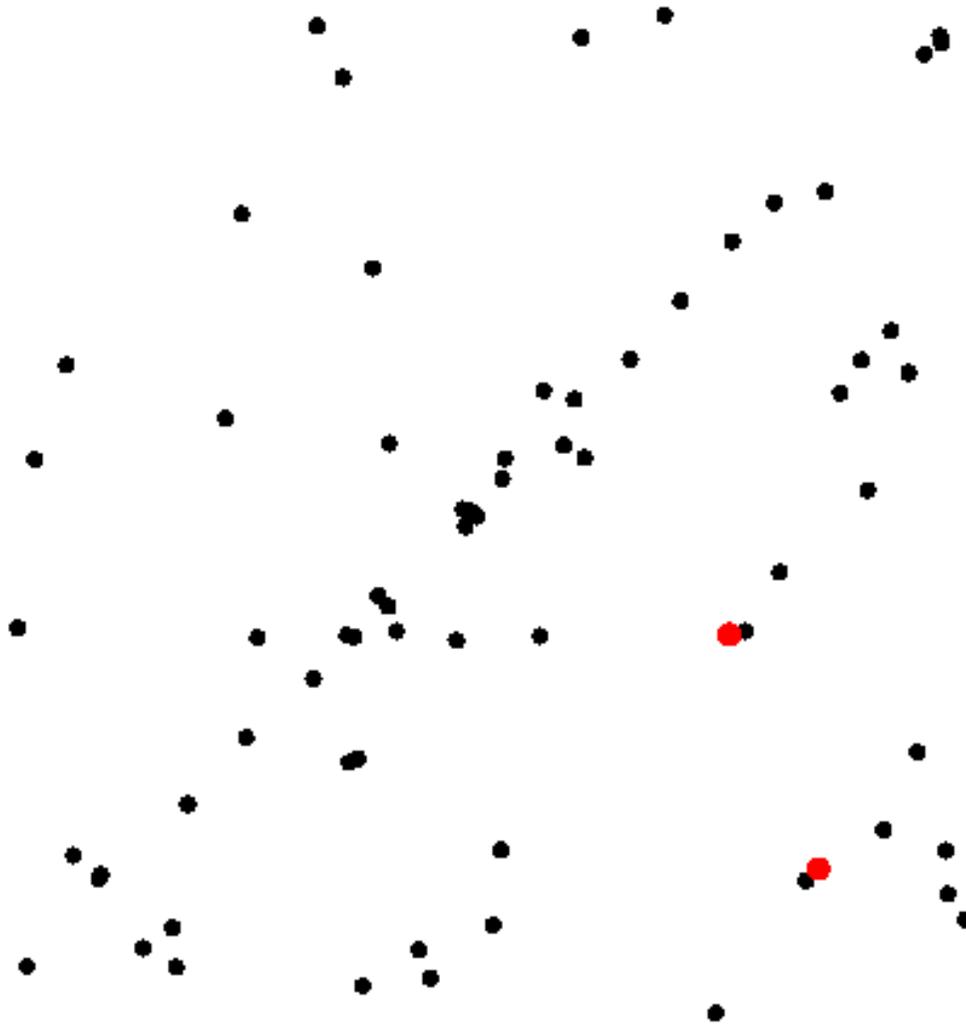
for $P=0.99$

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

Example: line fitting

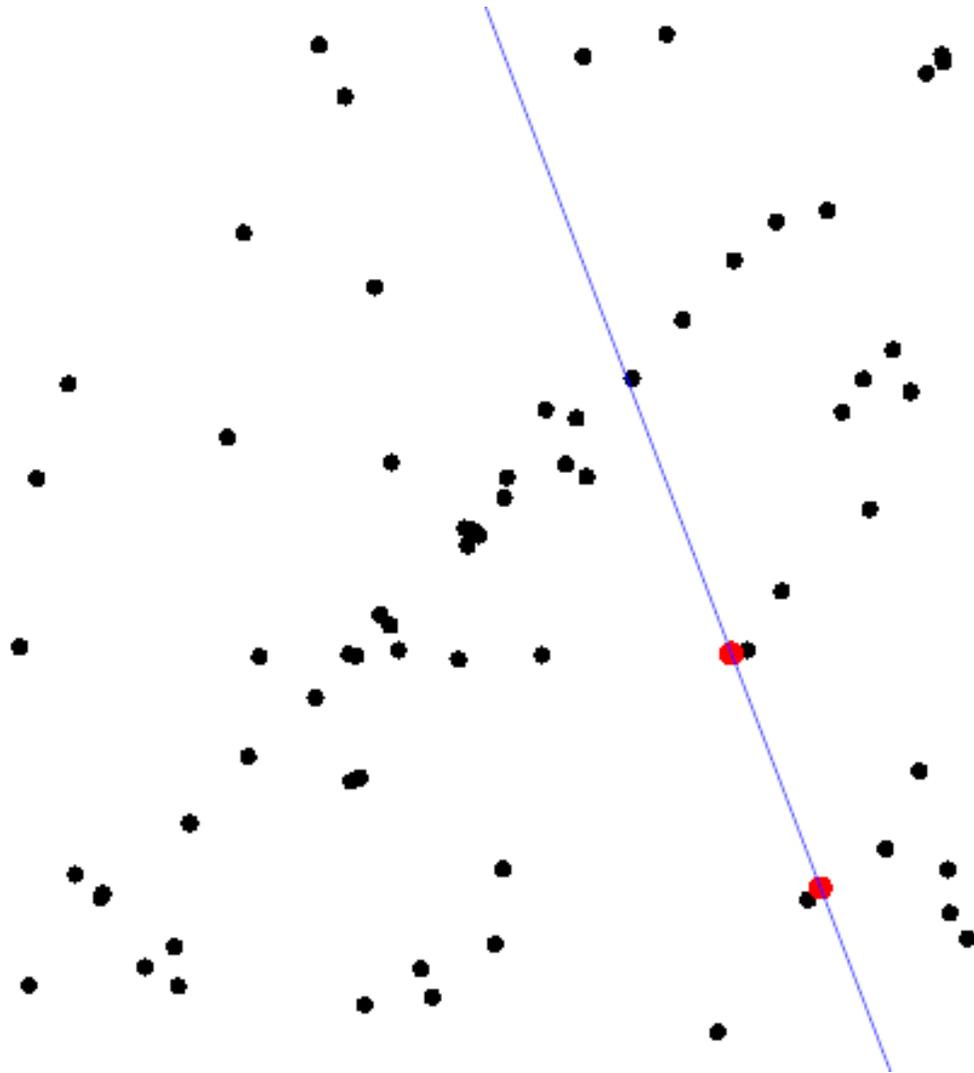


Example: line fitting

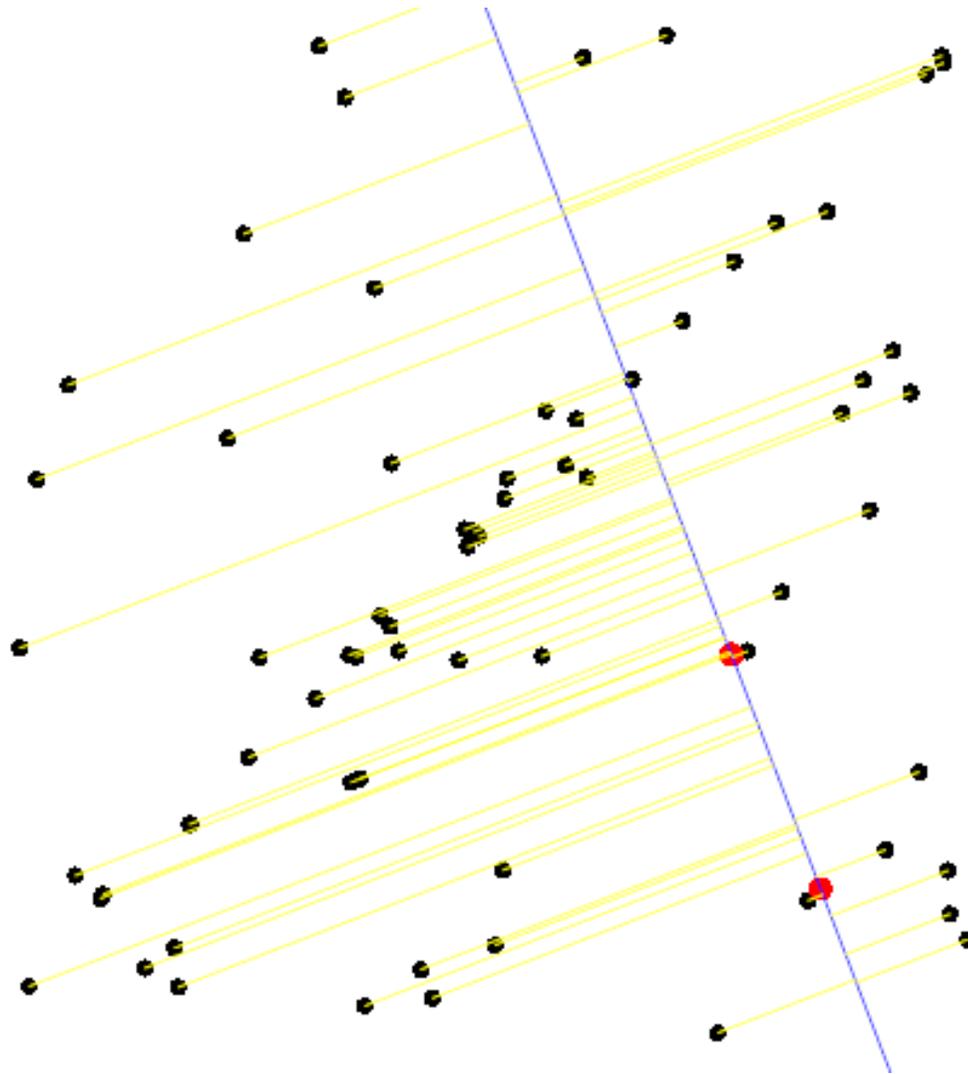


$n=2$

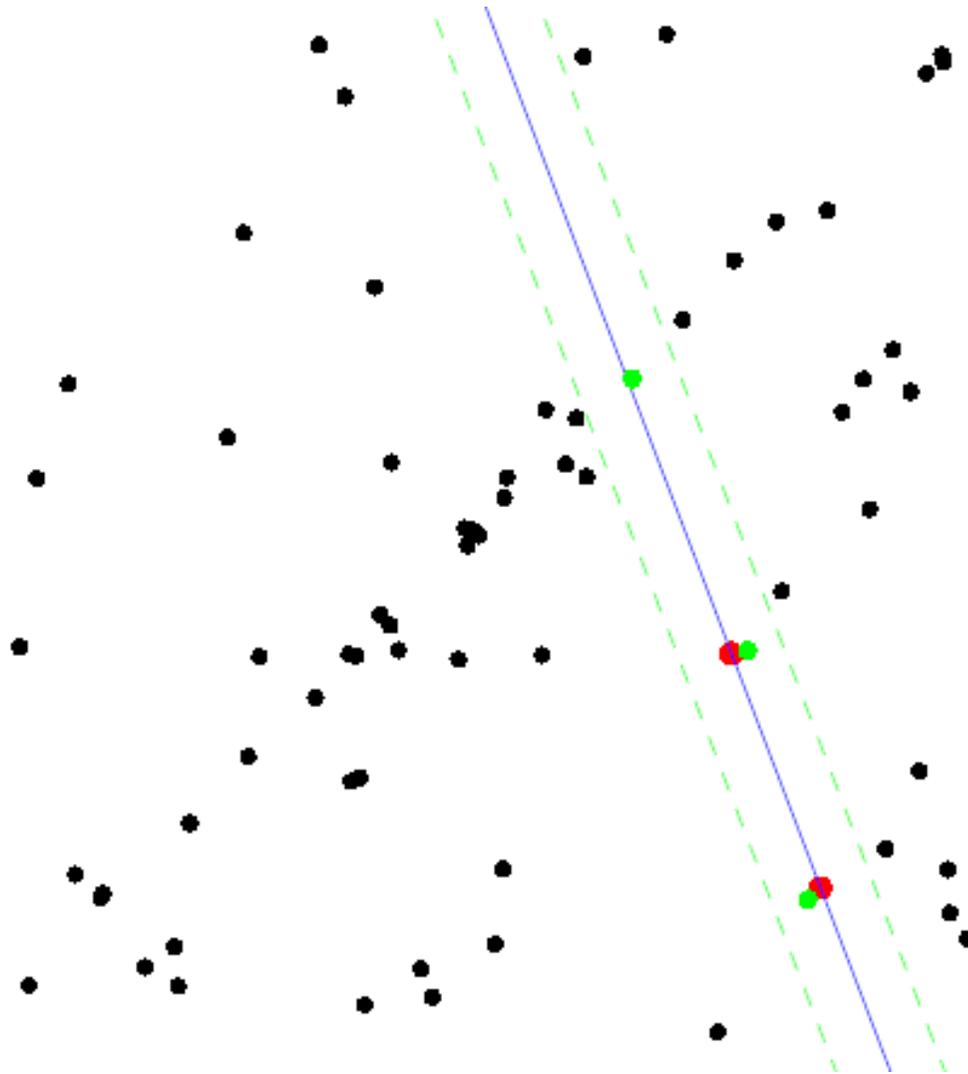
Model fitting



Measure distances

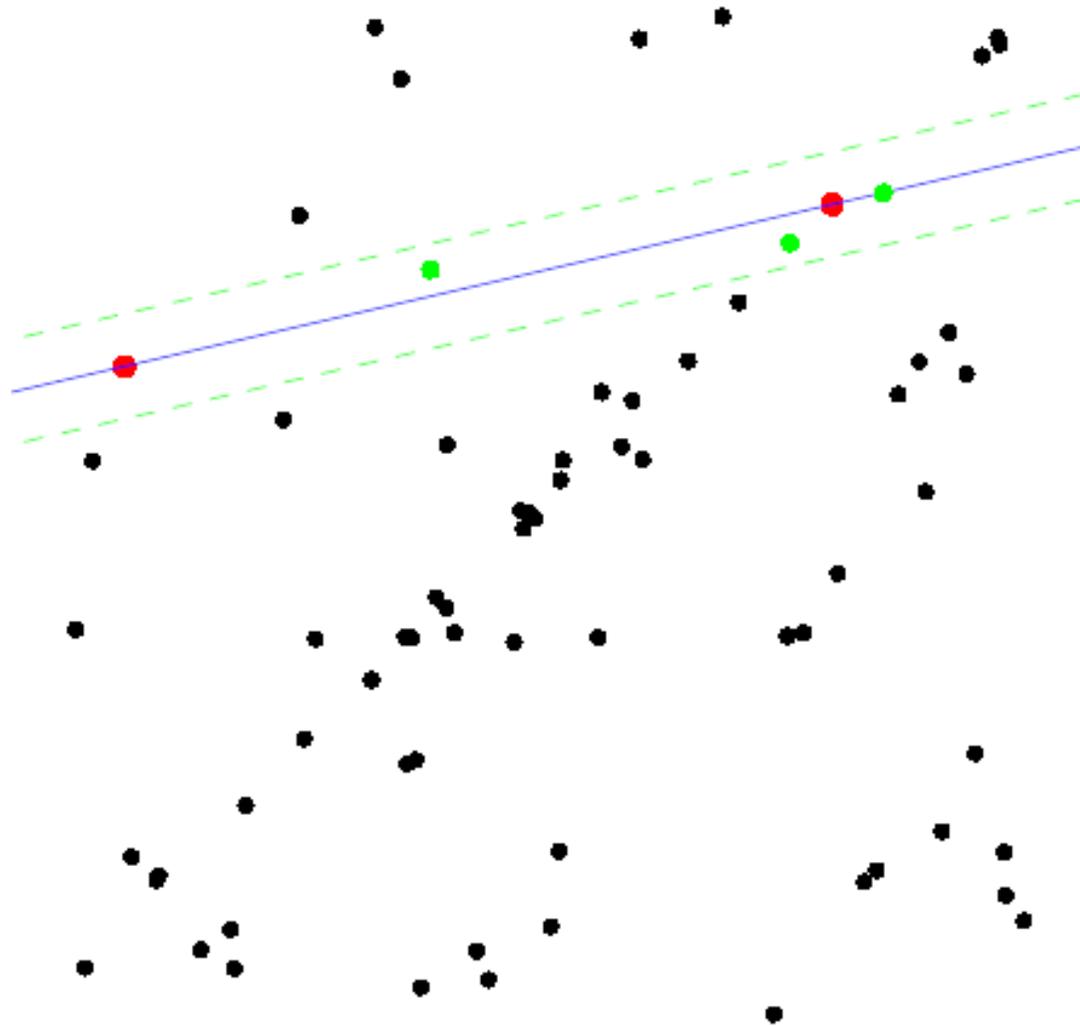


Count inliers



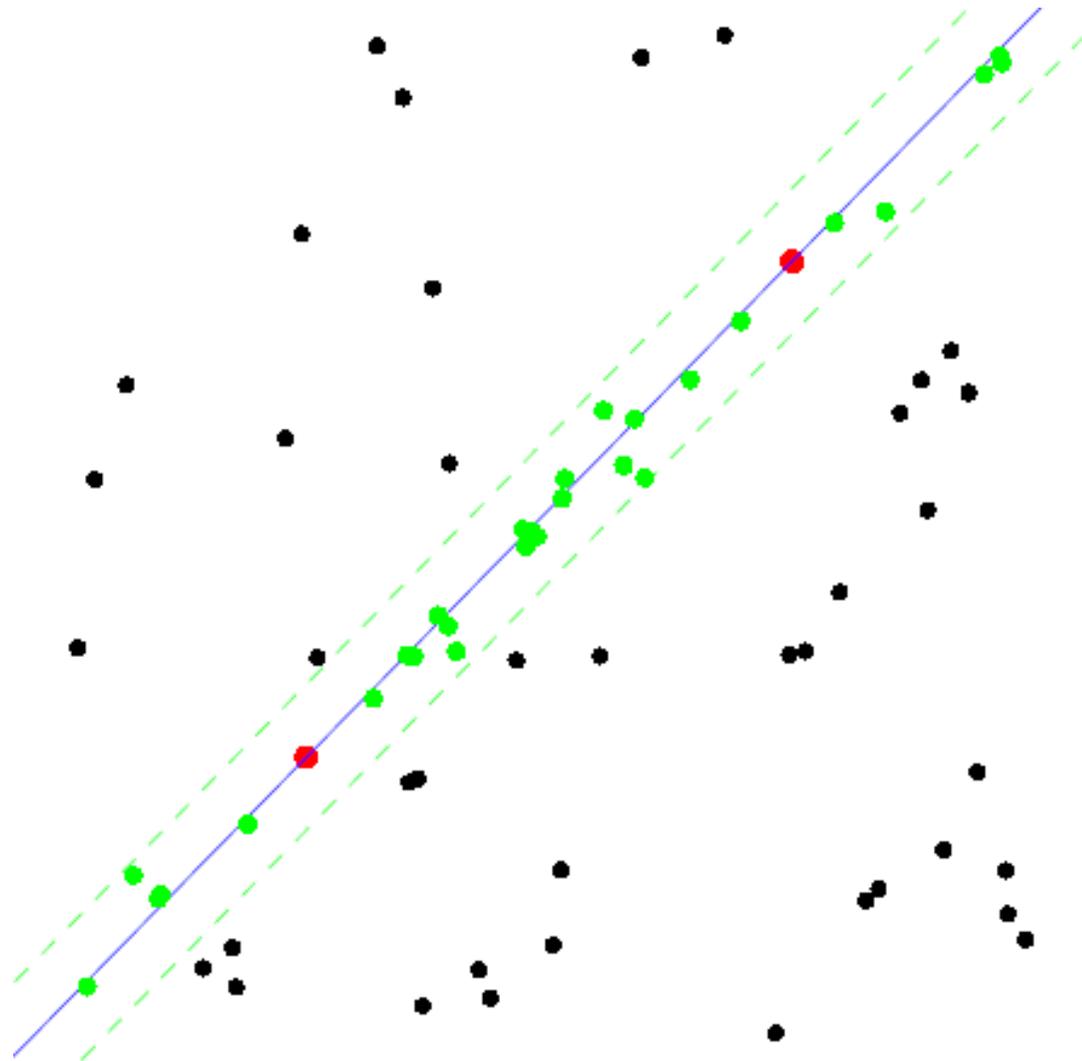
$$c=3$$

Another trial



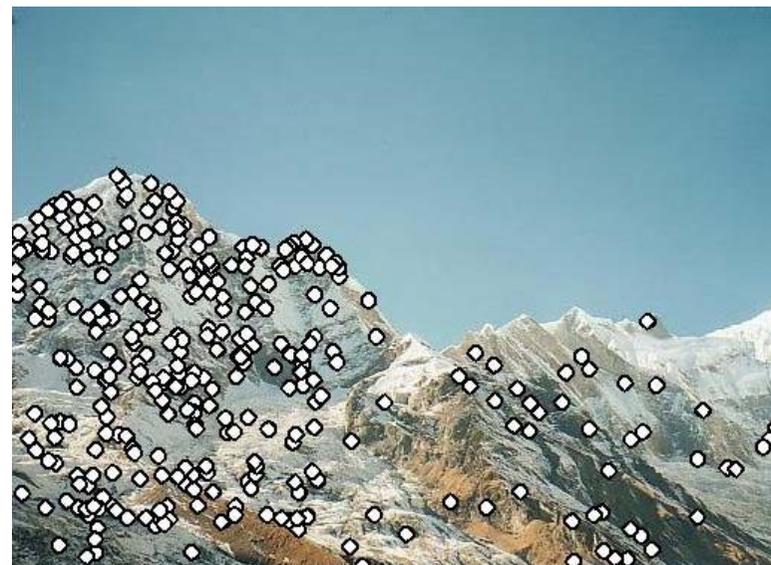
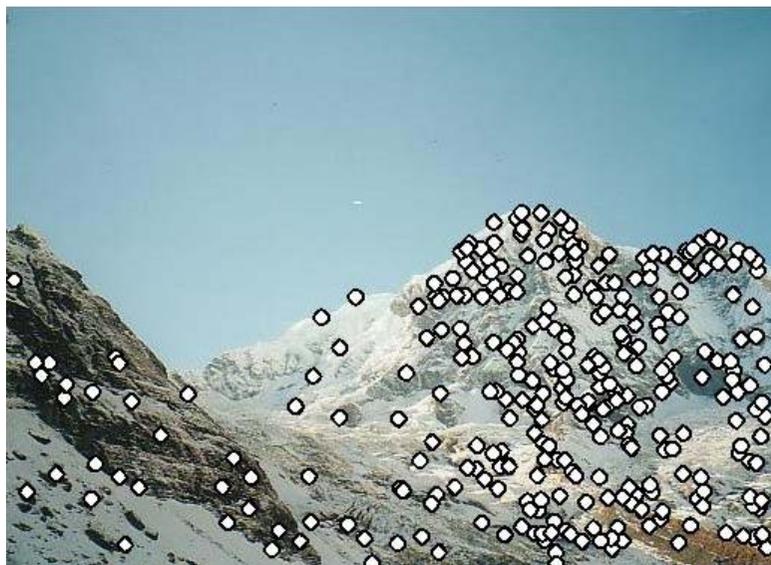
$$c=3$$

The best model

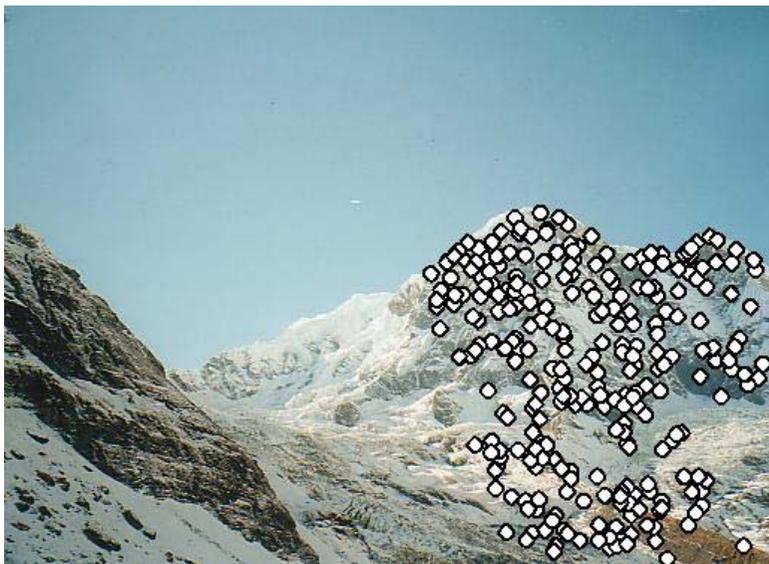


$c=15$

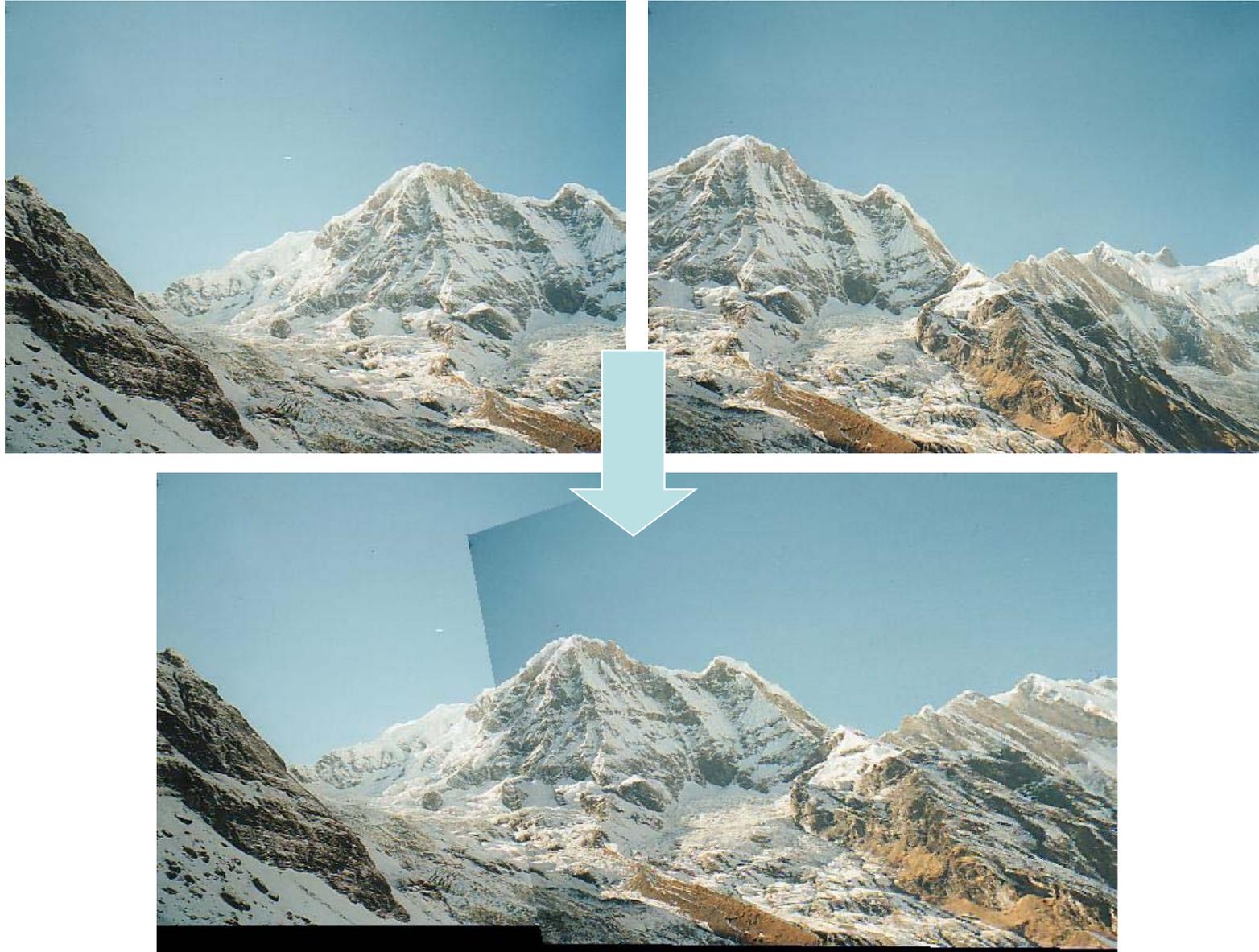
RANSAC for Homography



RANSAC for Homography

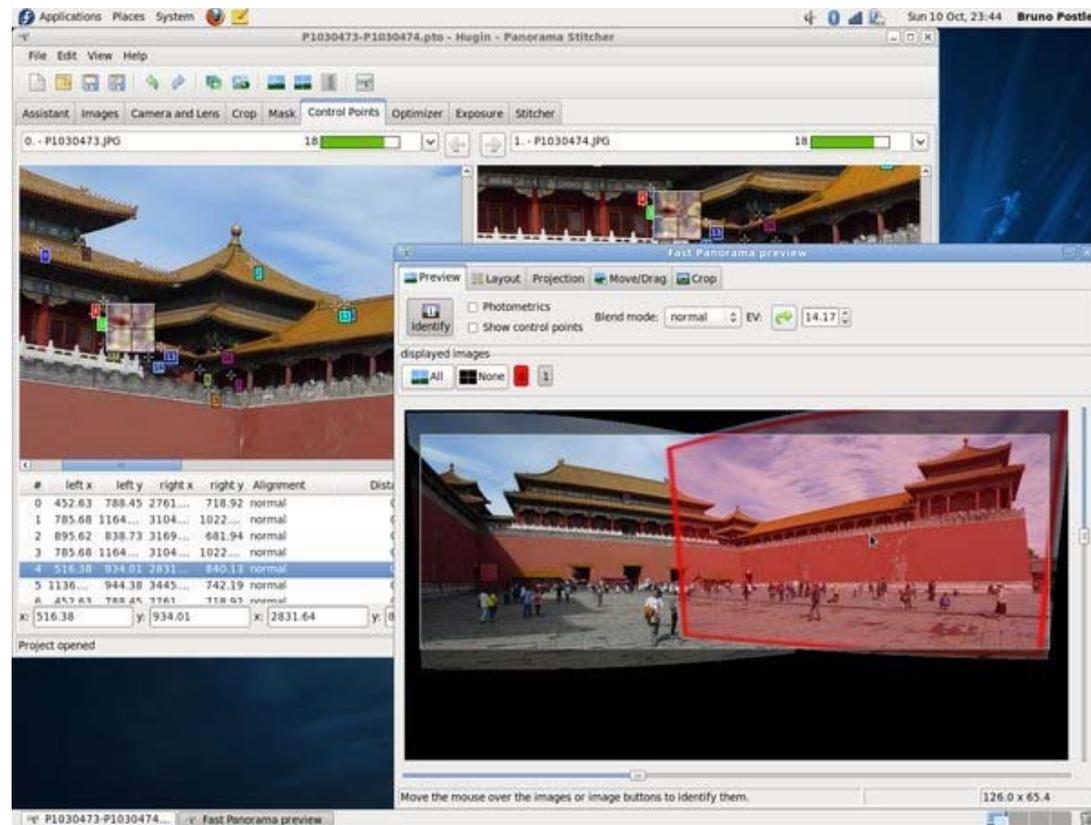


RANSAC for Homography



Tools for image stitching

- Hugin
- Image Composite Editor
- AutoStitch
- Google photo
- ...



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)



Reference

- Richard Szeliski, [Image Alignment and Stitching: A Tutorial](#), *Foundations and Trends in Computer Graphics and Computer Vision*, 2(1):1-104, December 2006.
- R. Szeliski and H.-Y. Shum. [Creating full view panoramic image mosaics and texture-mapped models](#), SIGGRAPH 1997, pp251-258.
- M. Brown, D. G. Lowe, [Recognising Panoramas](#), ICCV 2003.