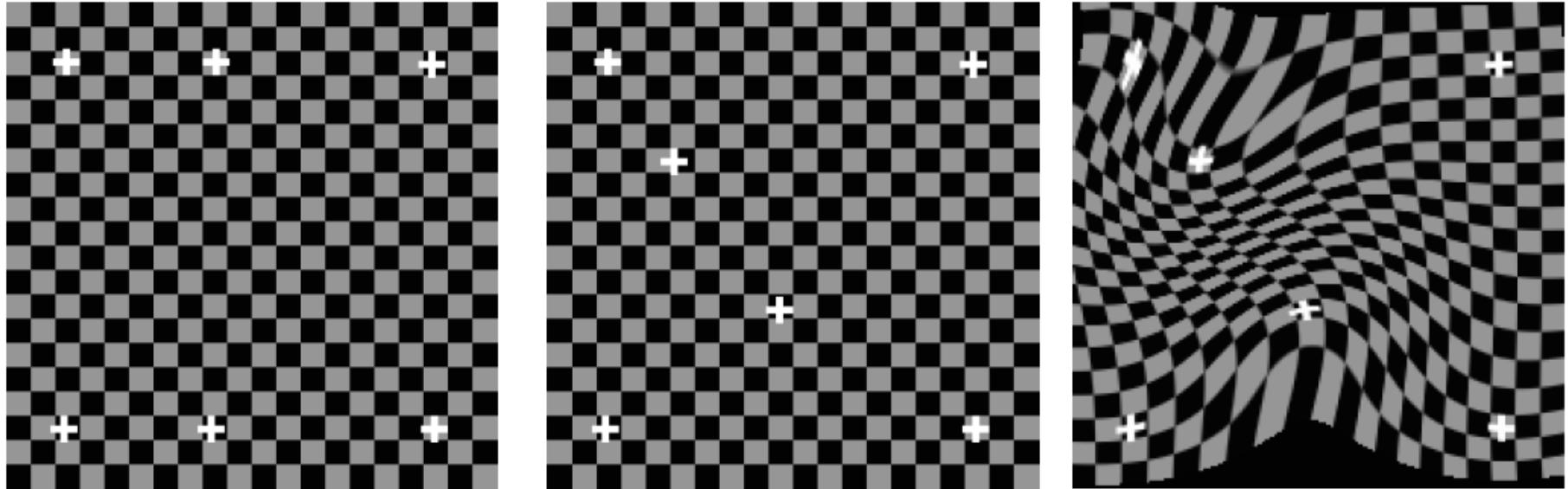


## Chapter 5.2 Geometric Transformations II

Sources:  
Slides Ross Whitaker (Utah)  
and G. Gerig

# Geometric Transformations

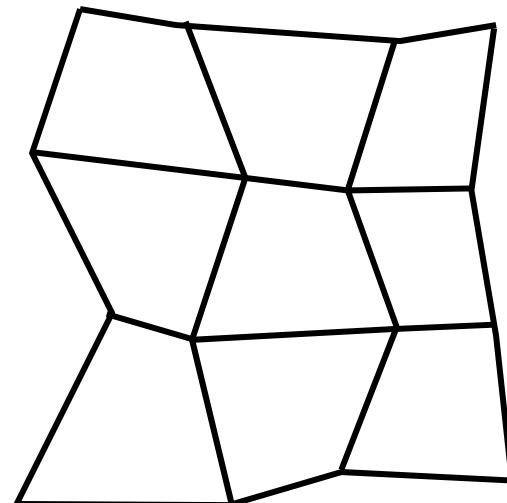
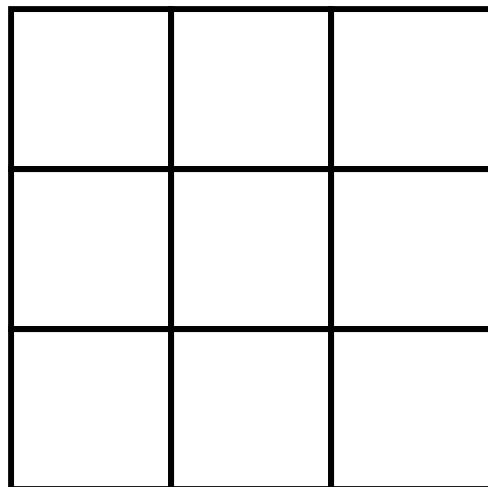
- Greyscale transformations -> operate on range/output
- Geometric transformations -> operate on image domain
  - Coordinate transformations
  - Moving image content from one place to another
- Two parts:
  - Define transformation
  - Resample greyscale image in new coordinates



## SPECIFYING “WARPS” VIA SPARSE SET OF LANDMARKS

# Specifying Warps – Another Strategy

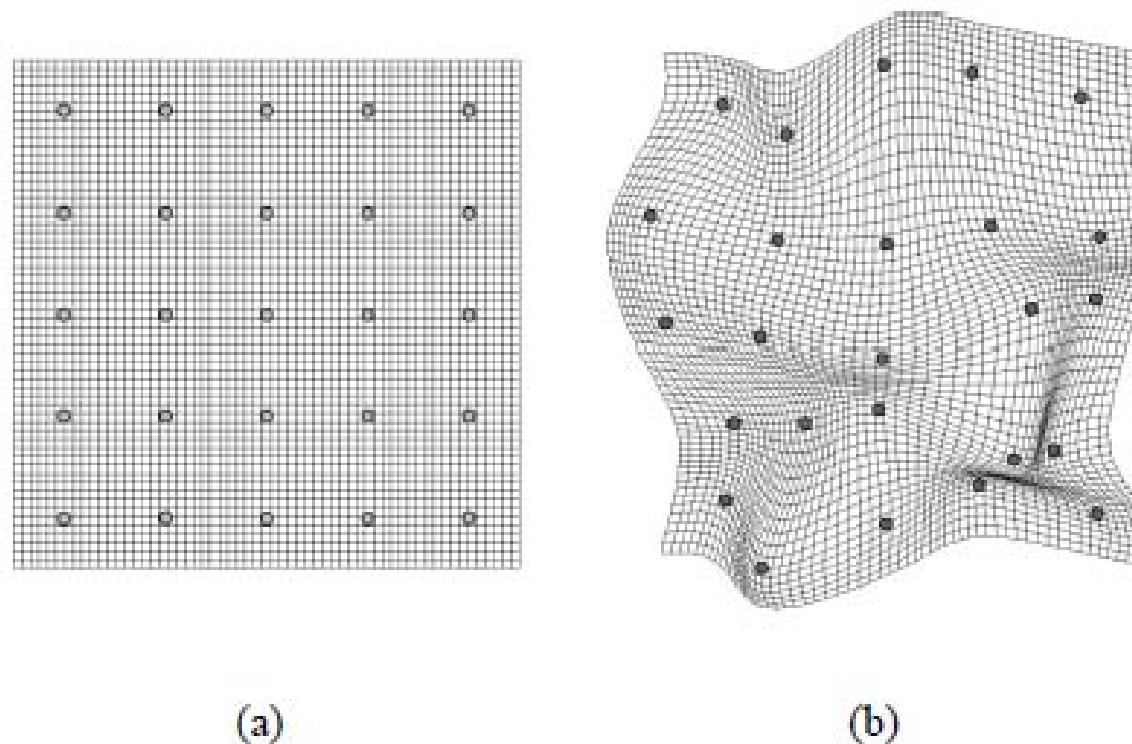
- Let the # DOFs in the warp equal the # of control points ( $x1/2$ )
  - Interpolate with some grid-based interpolation
    - E.g. bilinear, splines



# Landmarks Not On Grid

- Landmark positions driven by application
- Interpolate transformation at unorganized correspondences
  - *Scattered data interpolation*
- How do we do scattered data interpolation?
  - Idea: use kernels!
- *Radial basis functions*
  - Radially symmetric functions of distance to landmark

# Concept



**Figure 1. Warping a 2D mesh with RBFs: a) original mesh; b) mesh after warping.**

# Concept

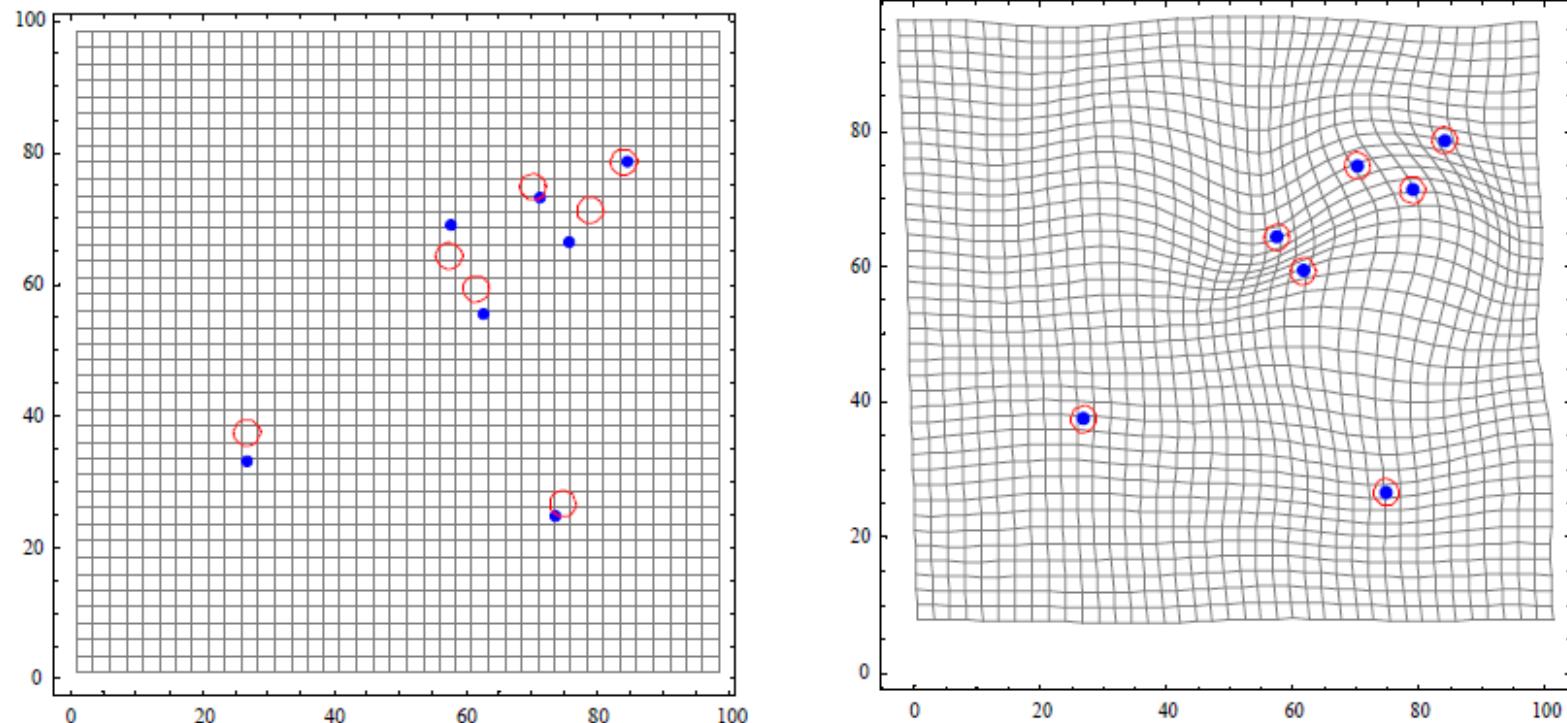


Fig. 5 Radial basis interpolation of a regular grid, based on the random motion of 7 landmarks.

**Warping a Neuro-Anatomy Atlas on 3D MRI Data with Radial Basis Functions** H.E. Bennink, J.M. Korbeeck, B.J. Janssen, B.M. ter Haar Romeny

# Concept

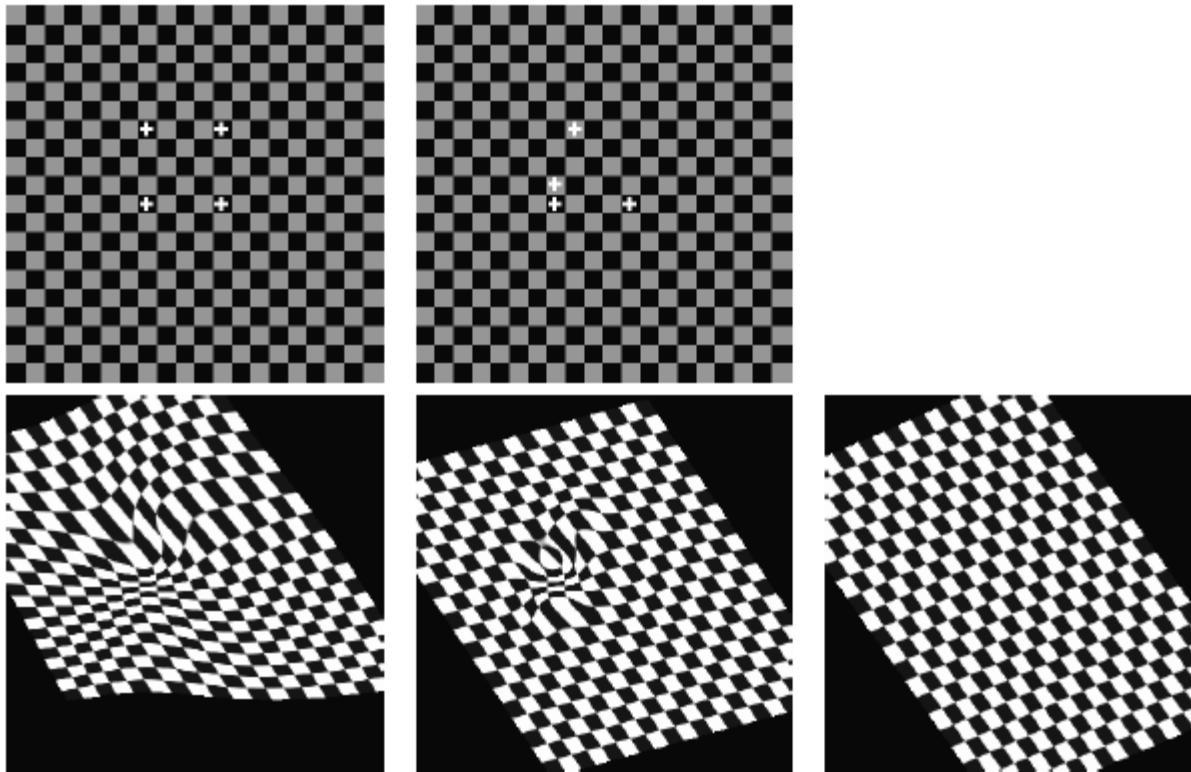
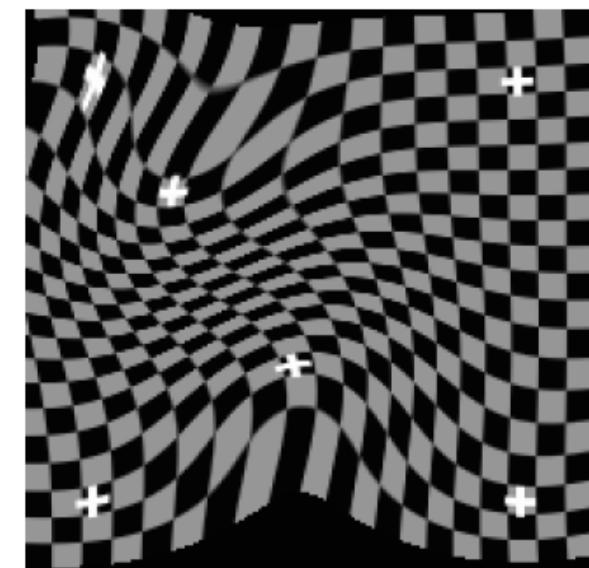
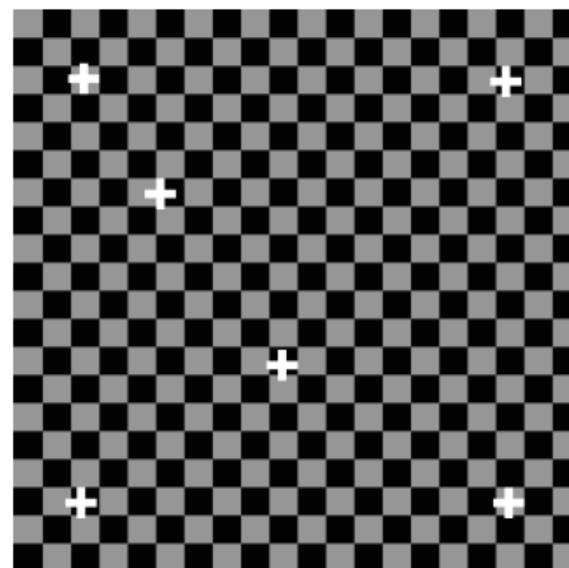
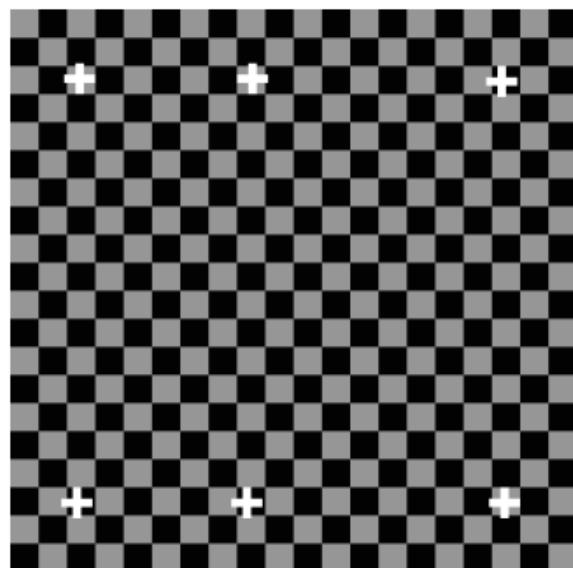


Figure 8: Generalizing affine mappings in different ways. **Top:** Position of source and target anchor points. **Bottom** (left to right): thin-plate warp, Gaussian warp, affine least-square warp ( $\lambda = \infty$ ). In all cases the mapping can be well approximated by an affine mapping far away from the anchors. In the thin-plate case this affine map is different at different regions, unlike the Gaussian case in which the same affine component appearing in the definition of the mapping dominates the transformation in all areas away from the anchors.

# RBF Formulation

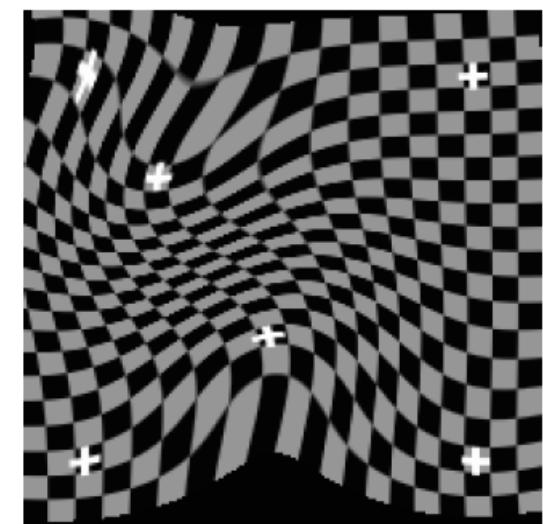
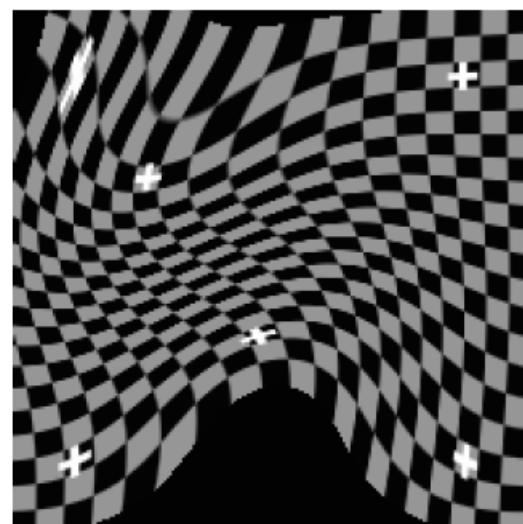
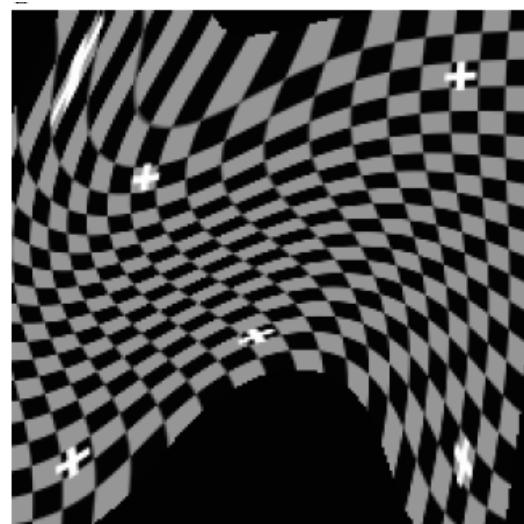
- Please see Gerig handouts for formulation

# RBF Warp – Example

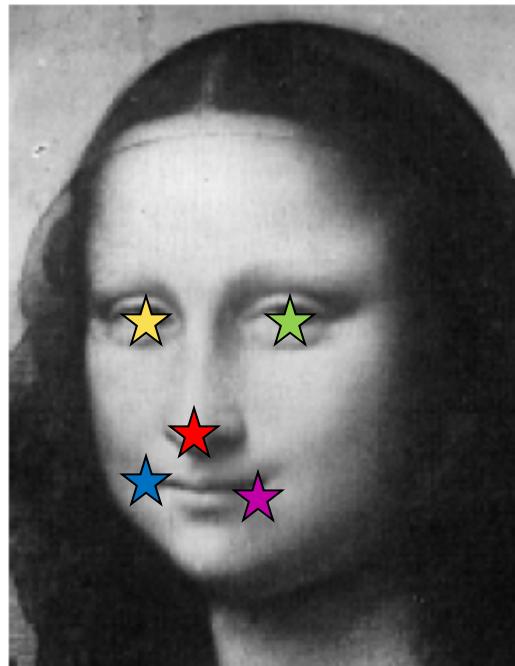


# What Kernel Should We Use

- Gaussian
  - Variance is free parameter – controls smoothness of warp



# RBFs – Aligning Faces



Mona Lisa – Target

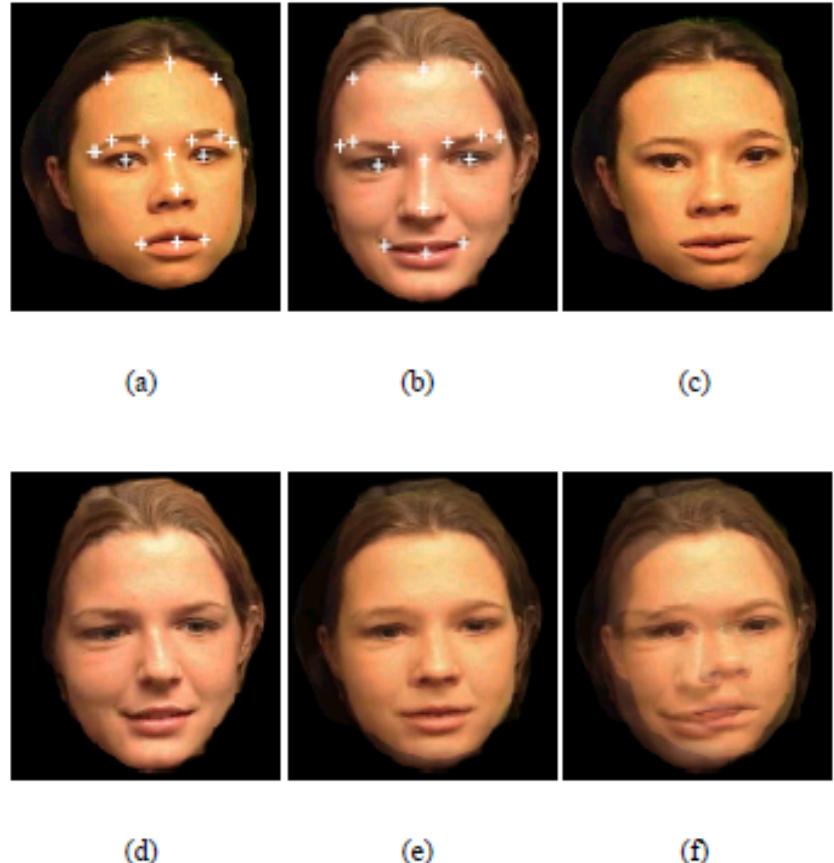


Venus – Source



Venus – Warped

# Symmetry?



**Figure 2. Image metamorphosis with RBFs:**  
a) source image  $I_0$ ; b) destination image  $I_1$ ;  
c) forward warping  $I_0$  with  $d_{0 \rightarrow 1}$ ; d) backward  
warping  $I_1$  with  $d_{1 \rightarrow 0}$ ; e) result of morphing  
between  $I_0$  and  $I_1$ ; f) cross-dissolved image.

**Image-based Talking Heads using Radial Basis Functions** James D. Edge and Steve Maddock

# Symmetry?

What can we say about symmetry:  $A \rightarrow B$   
and  $B \rightarrow A$  ?

# Application



**Figure 4. Synthesized viseme transitions. Central column contains transitional frames between the source and destination visemes.**

- Modeling of lip motion in speech with few landmarks.
- Synthesis via motion of landmarks.

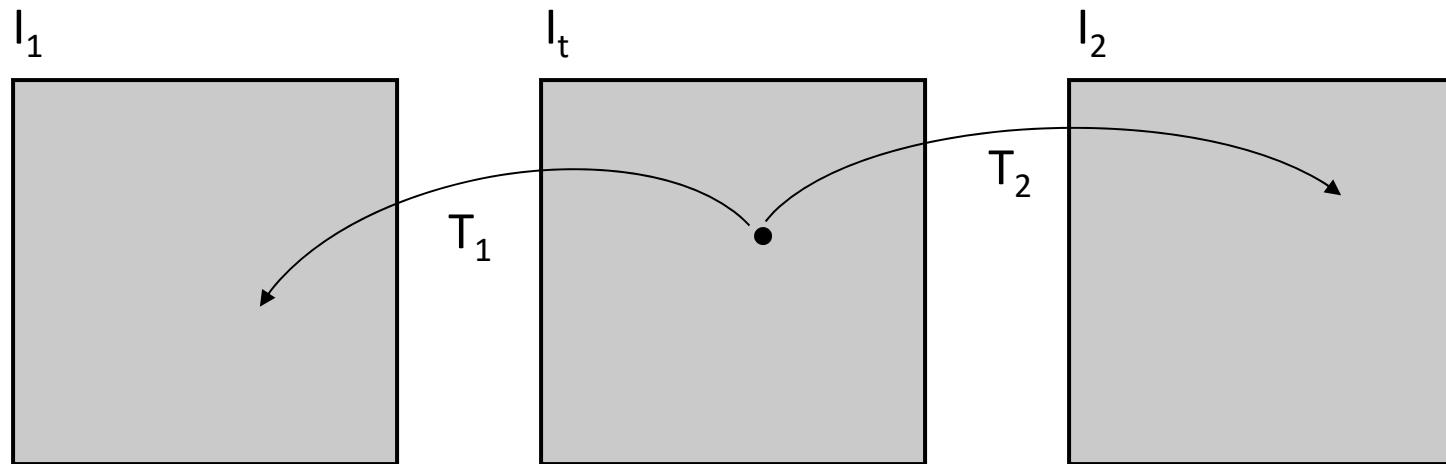
# Application: Image Morphing

- Combine shape and intensity with time parameter  $t$ 
  - Just blending with amounts  $t$  produces “fade”
$$I(t) = (1 - t)I_1 + tI_2$$
  - Use control points with interpolation in  $t$ 
$$\bar{c}(t) = (1 - t)\bar{c}_1 + t\bar{c}_2$$
  - Use  $c_1, c(t)$  landmarks to define  $T_1$ , and  $c_2, c(t)$  landmarks to define  $T_2$

# Image Morphing

- Create from blend of two warped images

$$I_t(\bar{x}) = (1 - t)I_1(T_1(\bar{x})) + tI_2(T_2(\bar{x}))$$



# Image Morphing



# Application: Image Templates/Atlases

- Build image templates that capture statistics of class of images
  - Accounts for shape and intensity
  - Mean and variability
- Purpose
  - Establish common coordinate system (for comparisons)
  - Understand how a particular case compares to the general population

# Templates – Formulation

- N landmarks over M different subjects/samples

Images

$$I^j(\bar{x}) \quad \bar{c}_i^j \quad \begin{pmatrix} \bar{c}_1^1 & \dots & \bar{c}_N^1 \\ \vdots & & \vdots \\ \bar{c}_1^M & \dots & \bar{c}_N^M \end{pmatrix}$$

Mean of correspondences (template)

$$\hat{c}_i = \frac{1}{M} \sum_{j=1}^M \bar{c}_i^j$$

Transformations from mean to subjects

$$\bar{c}_i^j = T^j(\hat{c}_i)$$

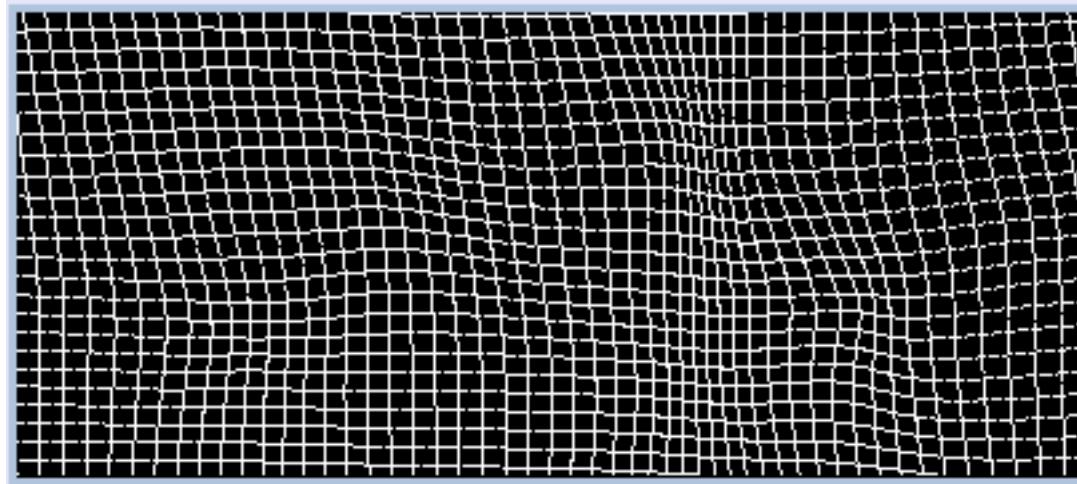
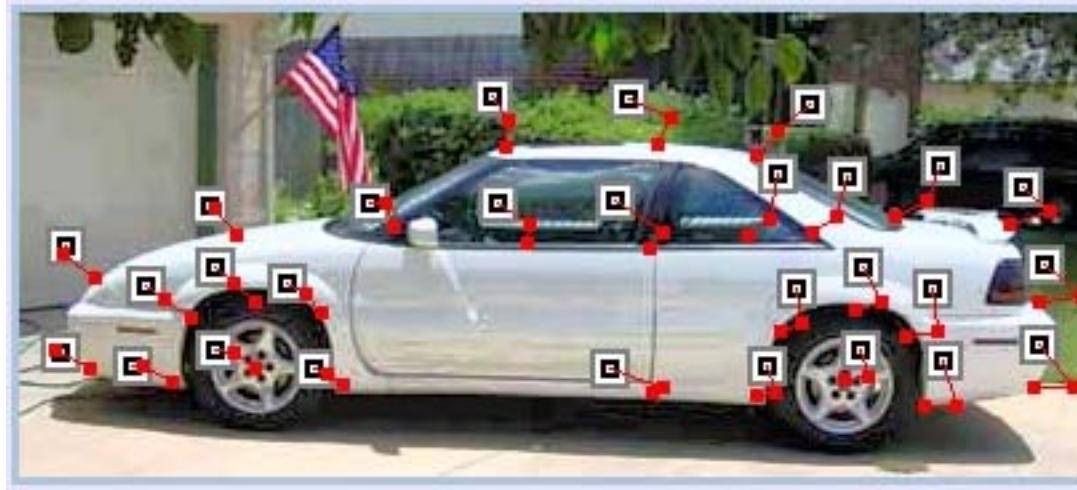
Templated image

$$\hat{I}(\bar{x}) = \frac{1}{M} \sum_j I^j(T^j(\bar{x}))$$

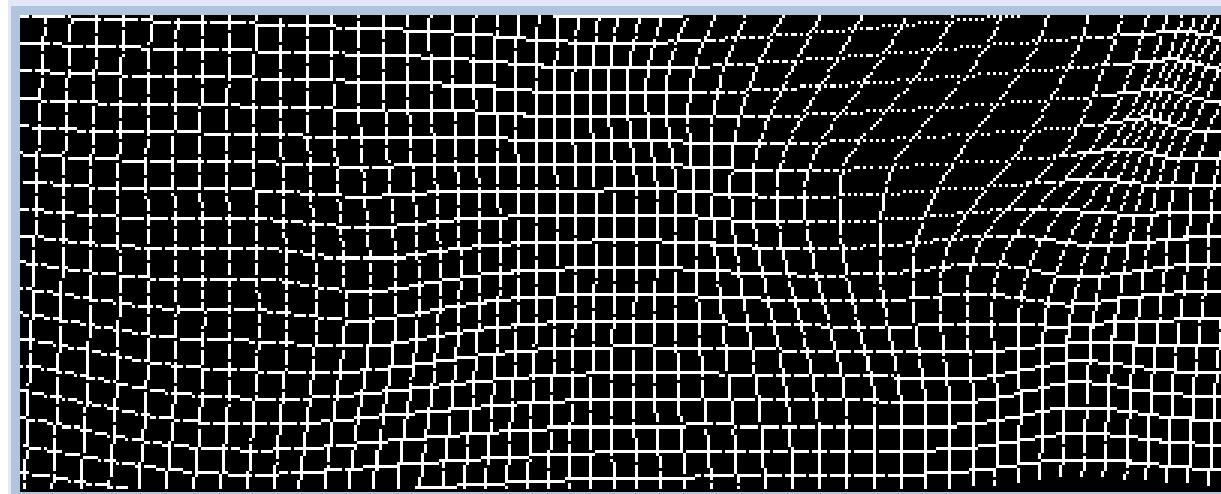
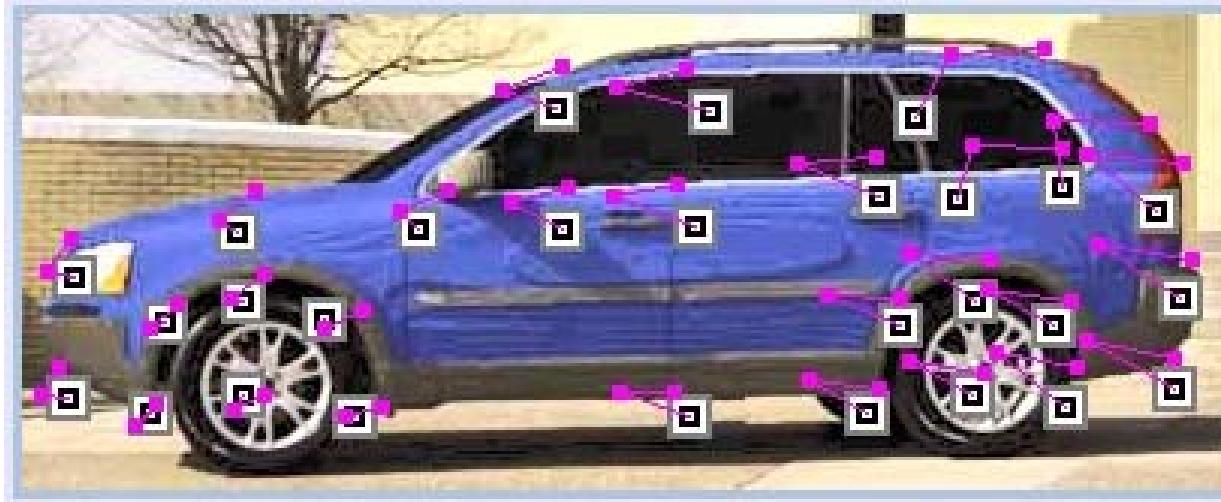
# Cars



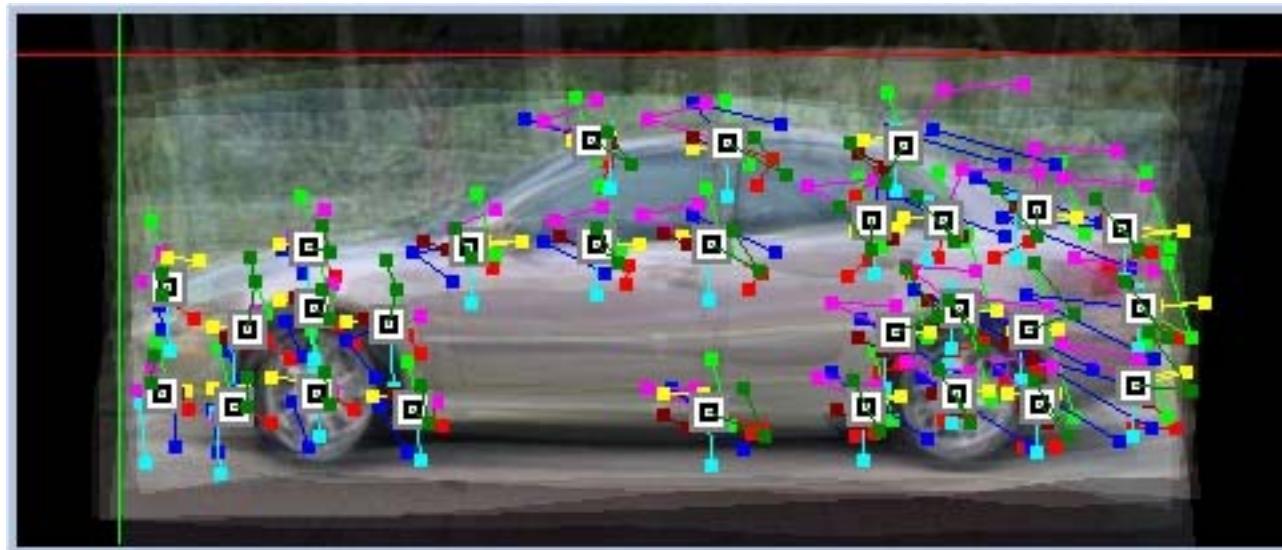
# Car Landmarks and Warp



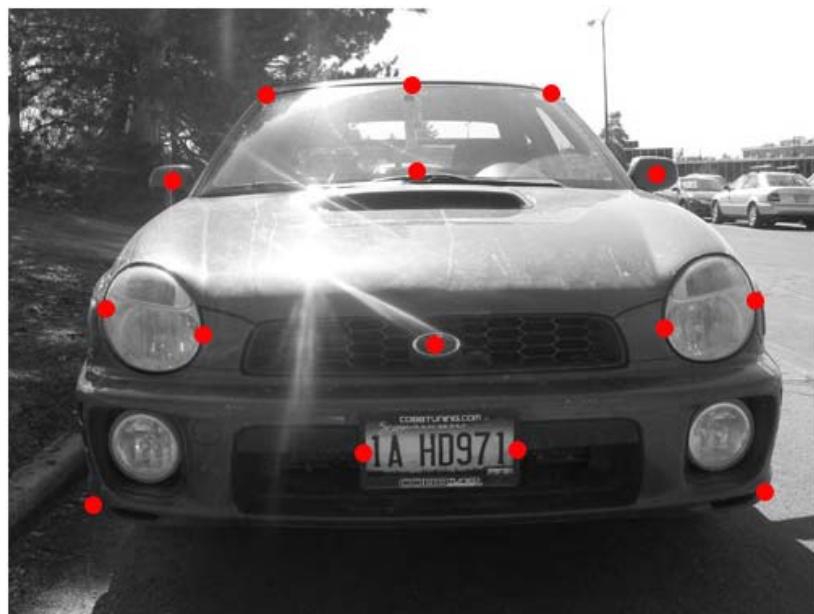
# Car Landmarks and Warp



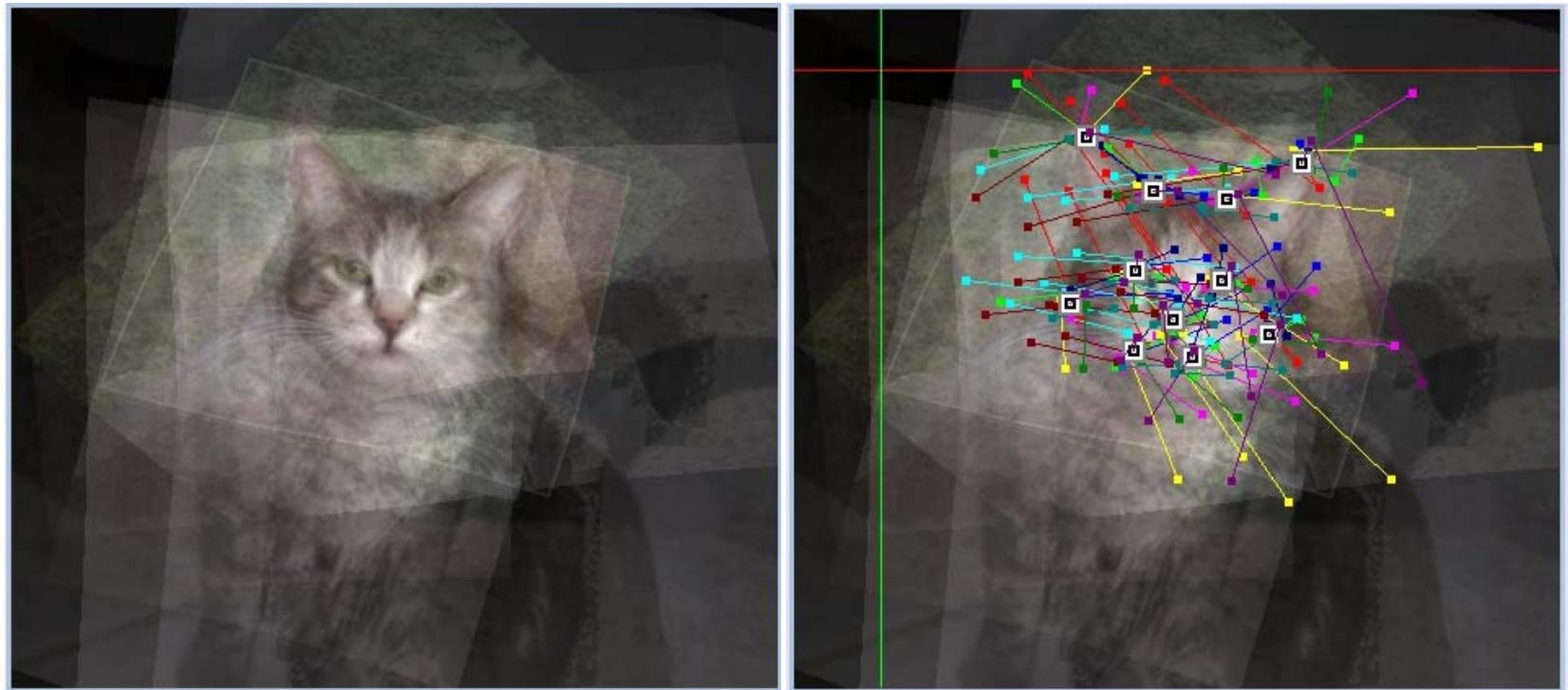
# Car Mean



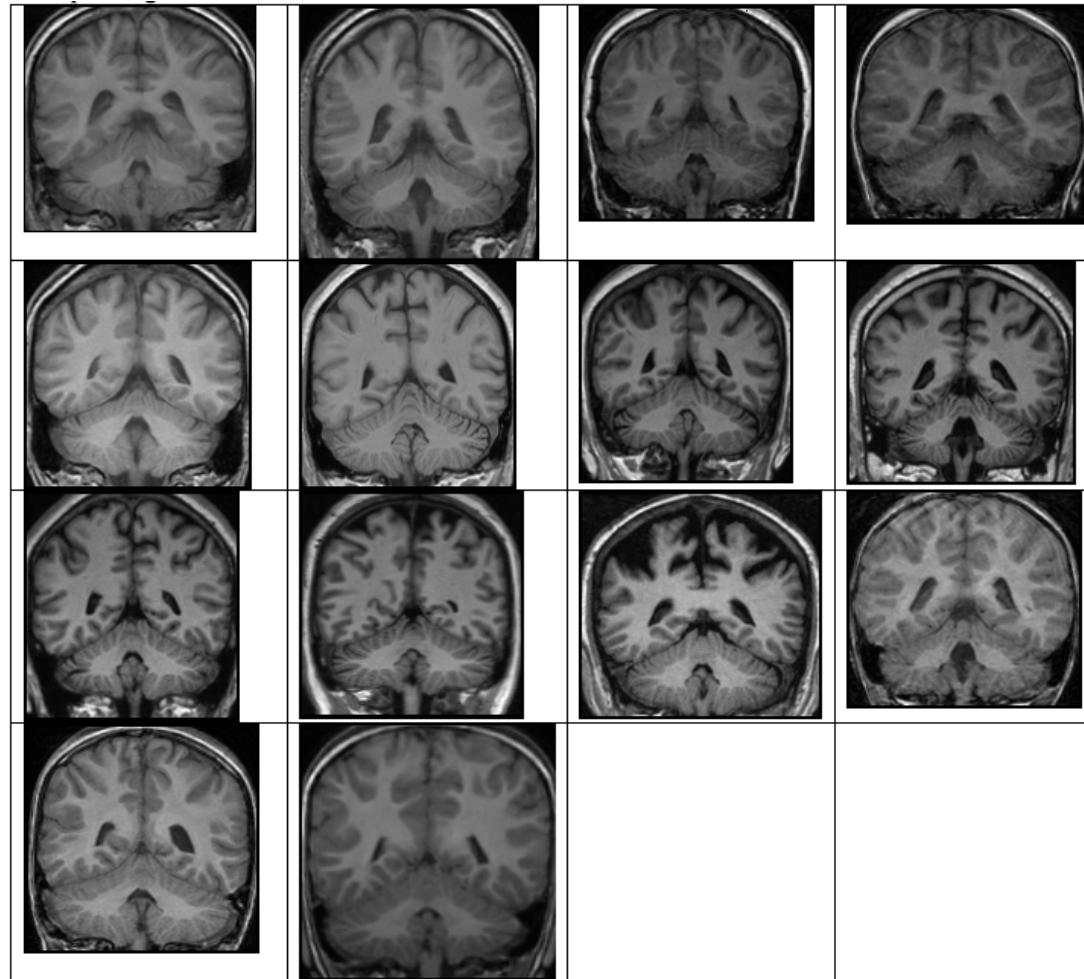
# Cars



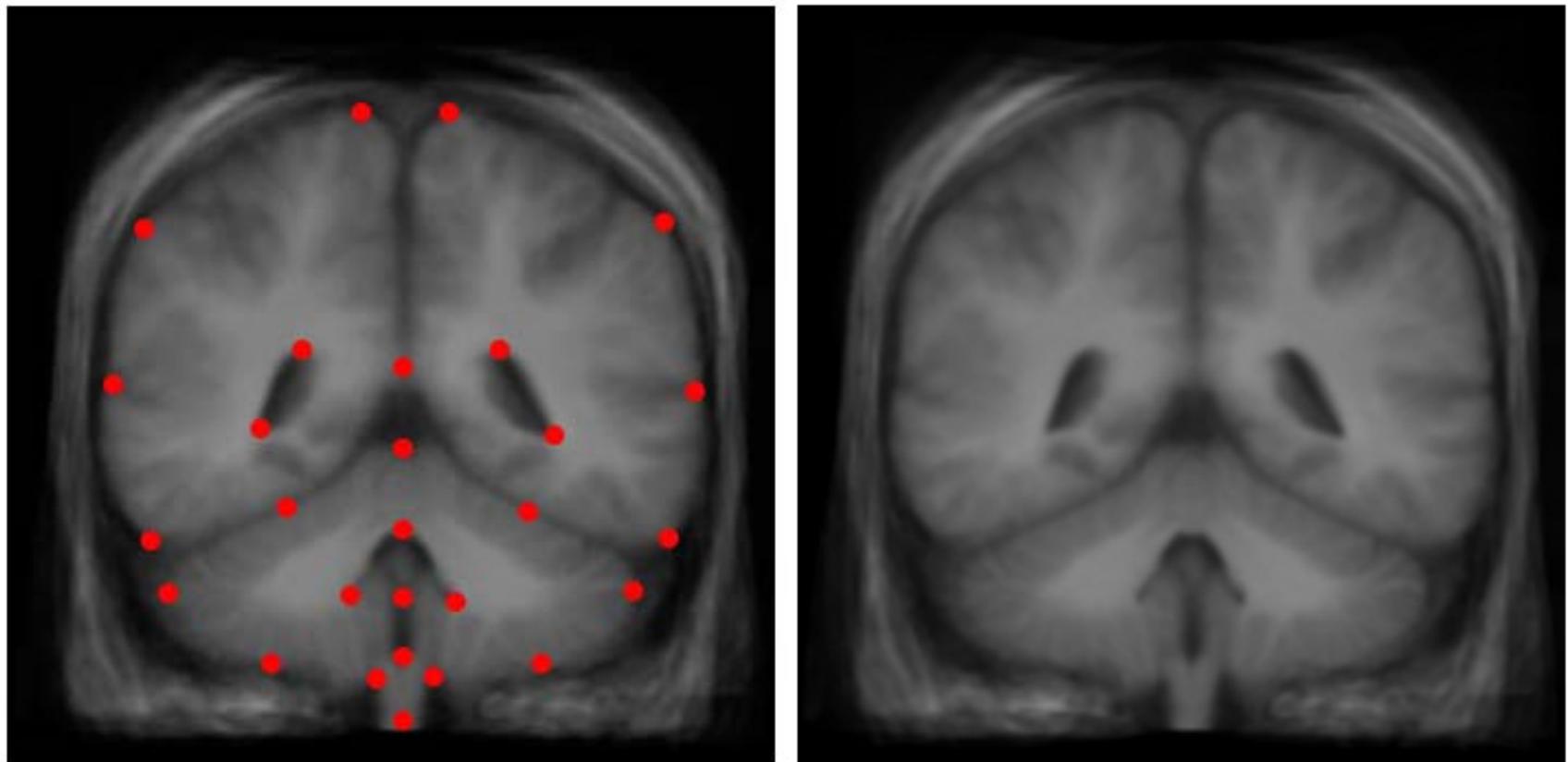
# Cats



# Brains



# Brain Template



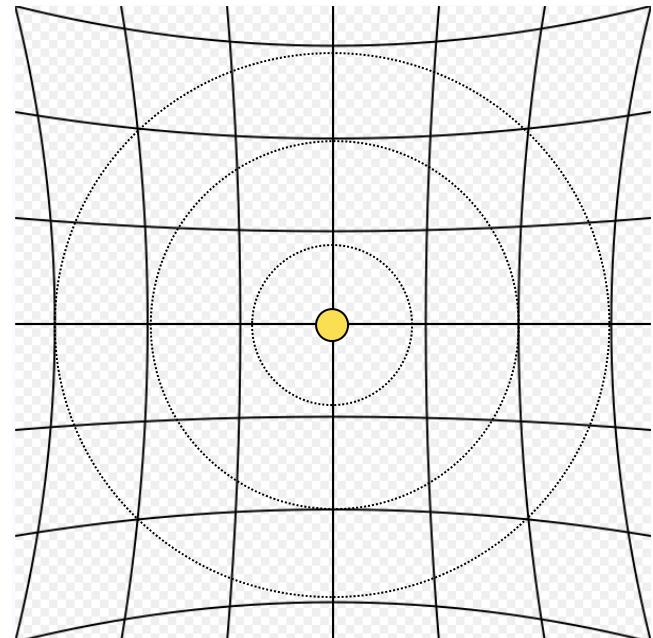
# APPLICATIONS

# Warping Application: Lens Distortion

- Radial transformation – lenses are generally circularly symmetric

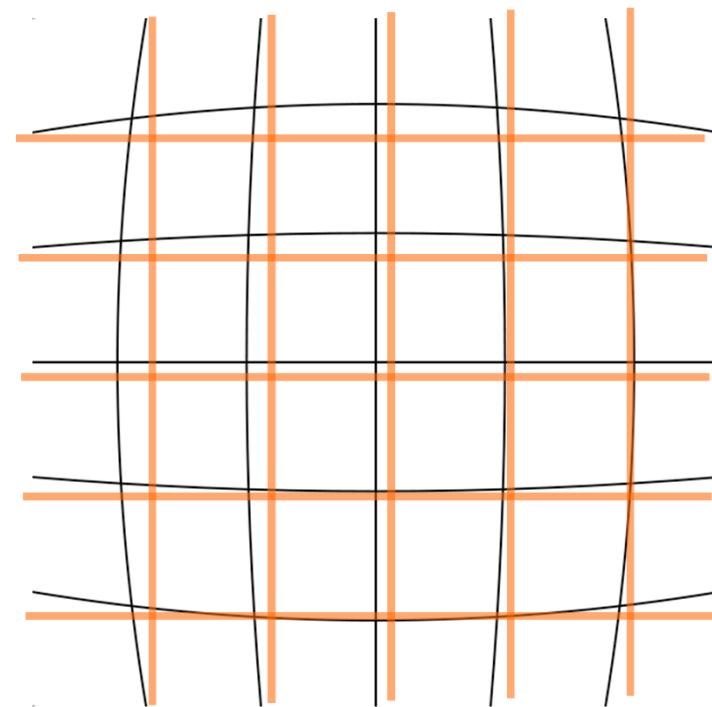
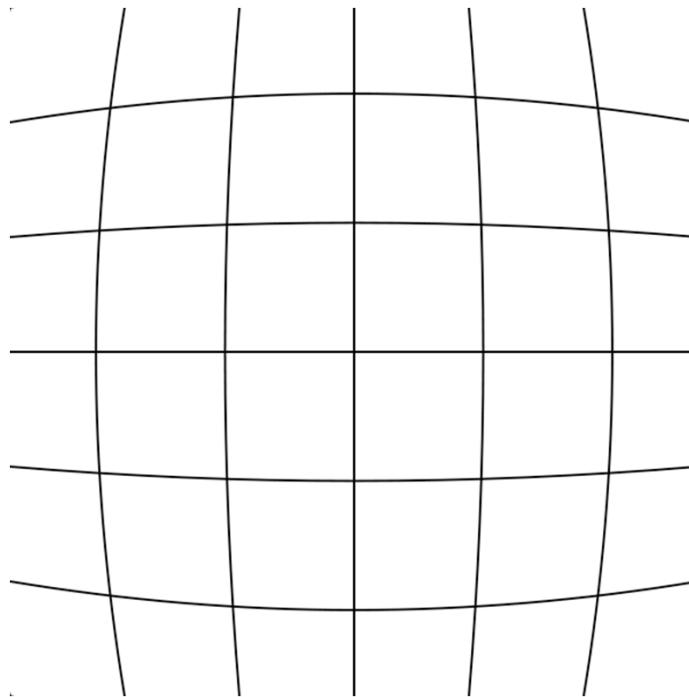
- Optical center is known
  - Model of transformation:

$$\bar{x}' = \bar{x} (1 + k_1 r^2 + k_2 r^4 + k_3 r^6 + \dots)$$



# Correspondences

- Take picture of known grid – crossings

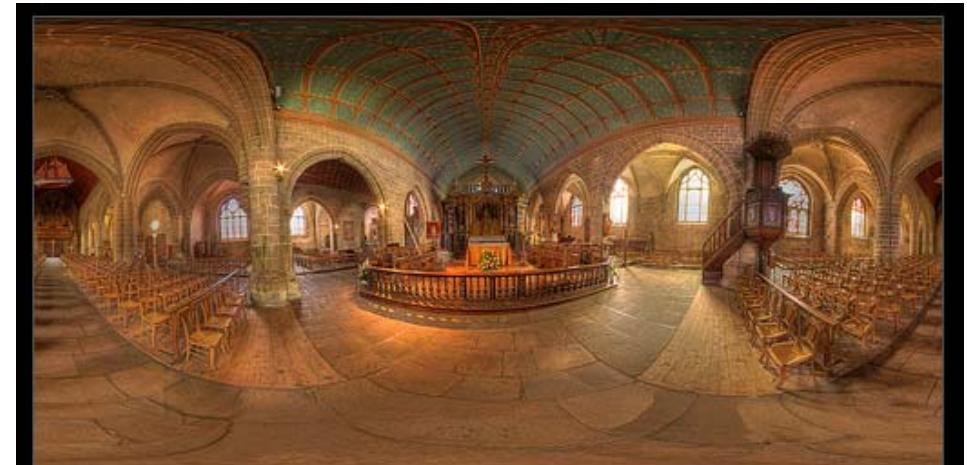


- Measure set of landmark pairs  $\rightarrow$  Estimate transformation, correct images

# Image Mosaicing

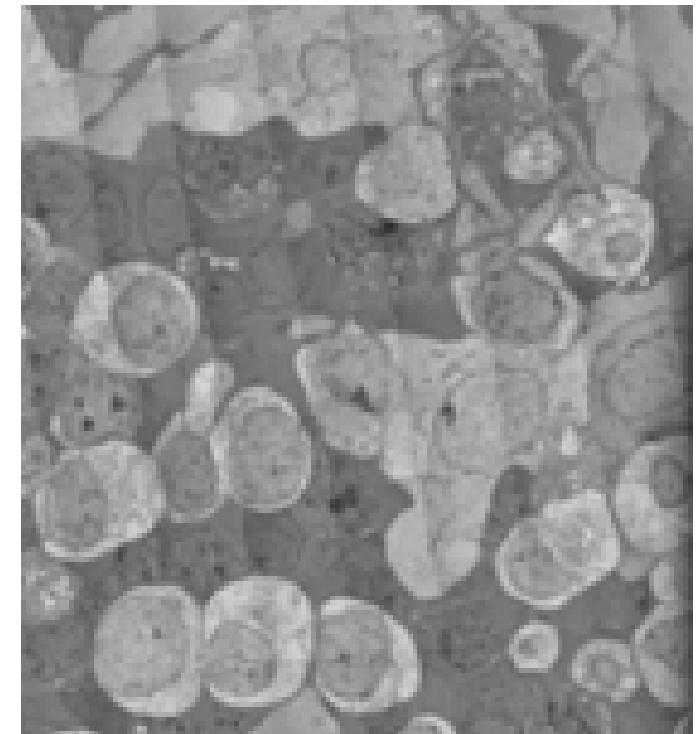
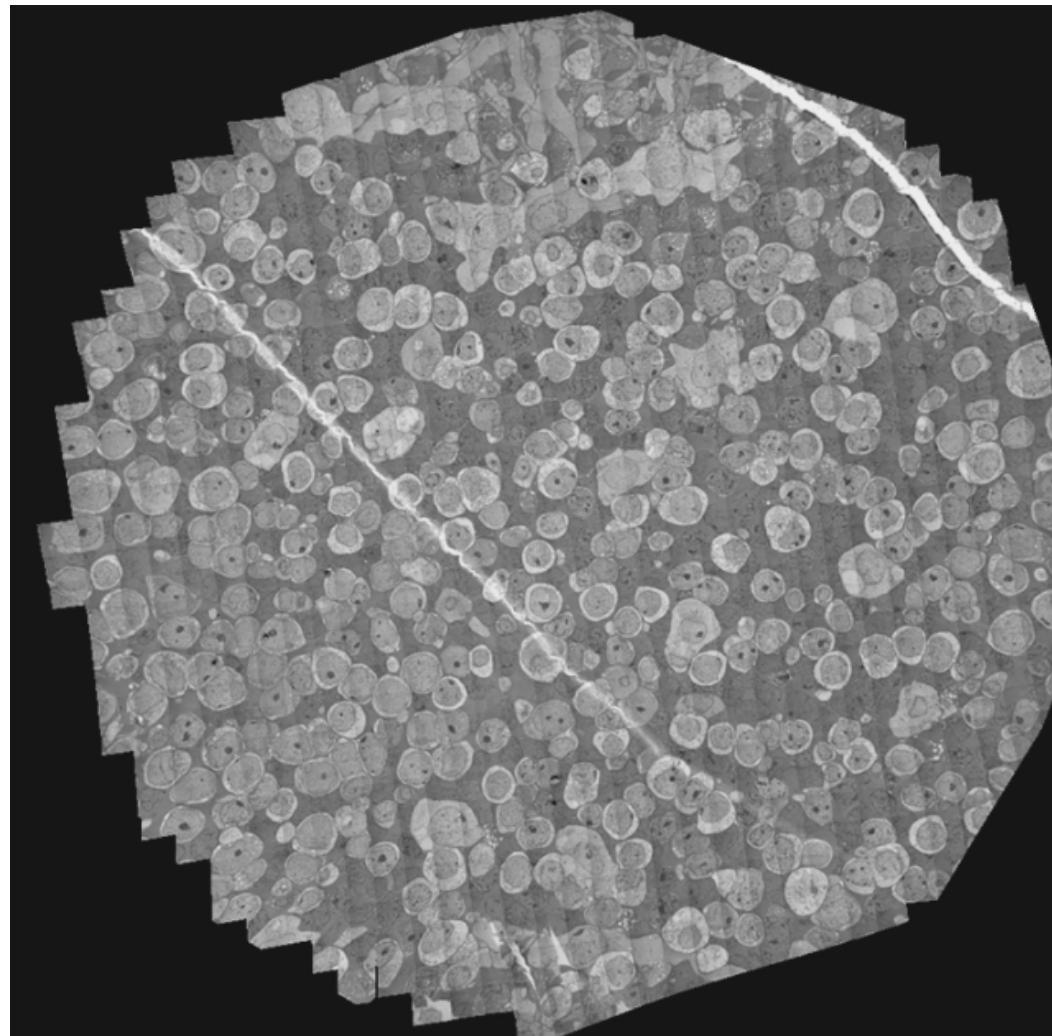
- Piecing together images to create a larger mosaic
- Doing it the old fashioned way
  - Paper pictures and tape
  - Things don't line up
  - Translation is not enough
- Need some kind of warp
- Constraints
  - Warping/matching two regions of two different images only works when...

# Applications



Saint-Guénolé Church of Batz-sur-Mer Equirectangular 360° by Vincent Montibus

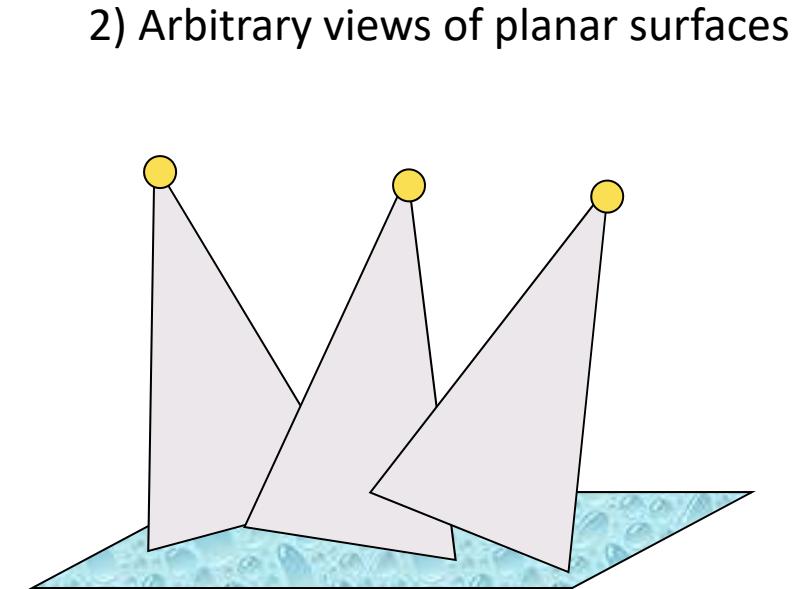
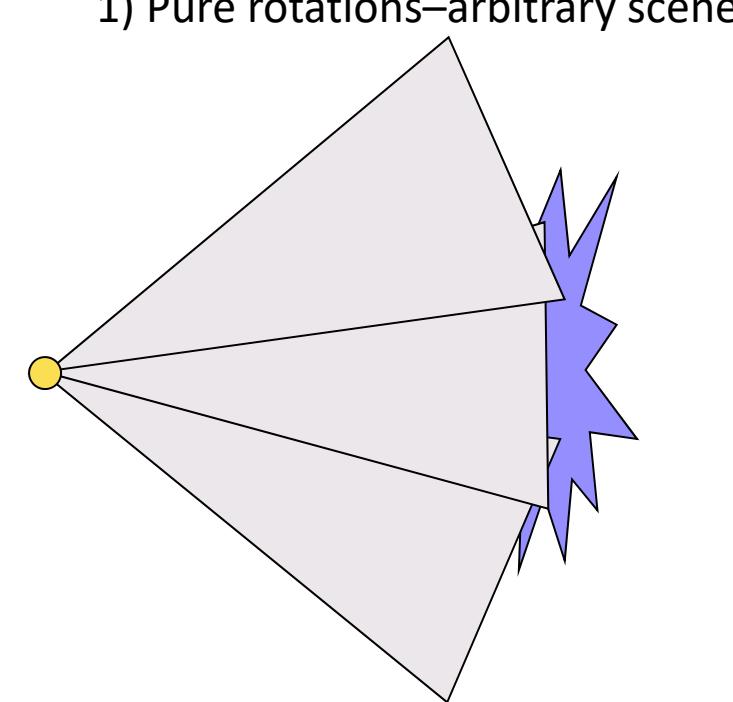
# Microscopy (Morane Eye Inst, UofU, T. Tasdizen et al.)





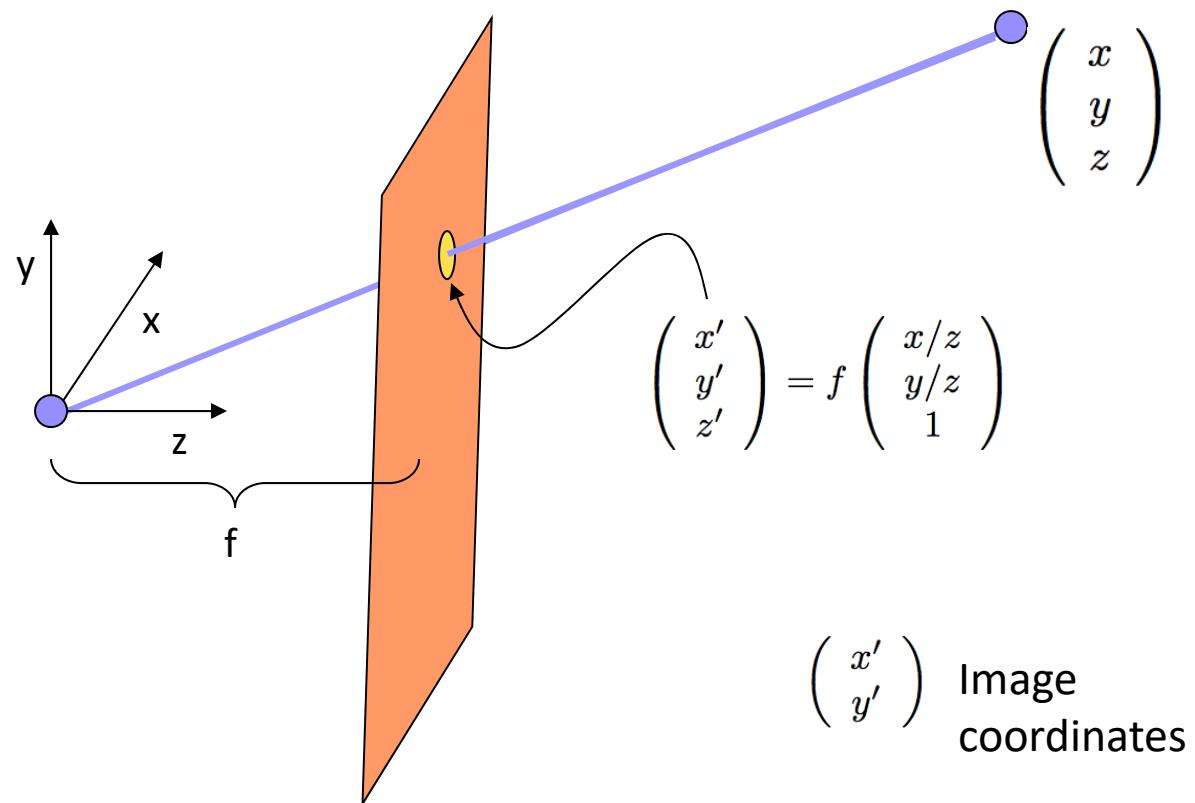
# Special Cases

- Nothing new in the scene is uncovered in one view vs another
  - No ray from the camera gets behind another



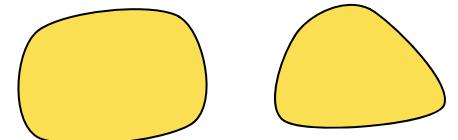
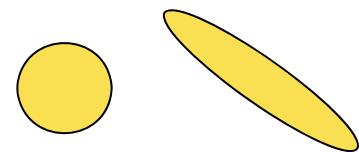
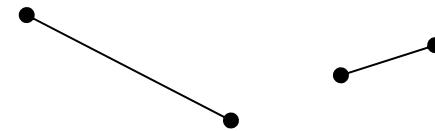
# 3D Perspective and Projection

- Camera model



# Perspective Projection Properties

- Lines to lines (linear)
- Conic sections to conic sections
- Convex shapes to convex shapes
- Foreshortening



# Image Homologies

- Images taken under cases 1,2 are perspectively equivalent to within a linear transformation
  - Projective relationships – equivalence is

$$\begin{pmatrix} a \\ b \\ c \end{pmatrix} \equiv \begin{pmatrix} d \\ e \\ f \end{pmatrix} \iff \begin{pmatrix} a/c \\ b/c \\ 1 \end{pmatrix} = \begin{pmatrix} d/f \\ e/f \\ 1 \end{pmatrix}$$

# Transforming Images To Make Mosaics

Linear transformation with matrix  $P$

$$\bar{x}^* = P\bar{x} \quad P = \begin{pmatrix} p_{11} & p_{12} & p_{13} \\ p_{21} & p_{22} & p_{23} \\ p_{31} & p_{32} & 1 \end{pmatrix} \quad \begin{aligned} x^* &= p_{11}x + p_{12}y + p_{13} \\ y^* &= p_{21}x + p_{22}y + p_{23} \\ z^* &= p_{31}x + p_{32}y + 1 \end{aligned}$$

Perspective equivalence

$$x' = \frac{p_{11}x + p_{12}y + p_{13}}{p_{31}x + p_{32}y + 1}$$

$$y' = \frac{p_{21}x + p_{22}y + p_{23}}{p_{31}x + p_{32}y + 1}$$

Multiply by denominator and reorganize terms

$$p_{31}xx' + p_{32}yx' - p_{11}x - p_{12}y - p_{13} = -x'$$

$$p_{31}xy' + p_{32}yy' - p_{21}x - p_{22}y - p_{23} = -y'$$

Linear system, solve for  $P$

$$\begin{pmatrix} -x_1 & -y_1 & -1 & 0 & 0 & 0 & x_1x'_1 & y_1x'_1 \\ -x_2 & -y_2 & -1 & 0 & 0 & 0 & x_2x'_2 & y_2x'_2 \\ & & & \vdots & & & & \\ -x_N & -y_N & -1 & 0 & 0 & 0 & x_Nx'_N & y_Nx'_2 \\ 0 & 0 & 0 & -x_1 & -y_1 & -1 & x_1y'_1 & y_1y'_1 \\ 0 & 0 & 0 & -x_2 & -y_2 & -1 & x_2y'_2 & y_2y'_2 \\ & & & \vdots & & & & \\ 0 & 0 & 0 & -x_N & -y_N & -1 & x_Ny'_N & y_Ny'_N \end{pmatrix} \begin{pmatrix} p_{11} \\ p_{12} \\ p_{13} \\ p_{21} \\ p_{22} \\ p_{23} \\ p_{31} \\ p_{32} \end{pmatrix} = \begin{pmatrix} -x'_1 \\ -x'_2 \\ \vdots \\ -x'_N \\ -y'_1 \\ -y'_2 \\ \vdots \\ -y'_N \end{pmatrix}$$

# Image Mosaicing



# 4 Correspondences



# 5 Correspondences



# 6 Correspondences



# Mosaicing Issues

- Need a canvas (adjust coordinates/origin)
- Blending at edges of images (avoid sharp transitions)
- Adjusting brightnesses
- Cascading transformations