

# Image Warping / Morphing

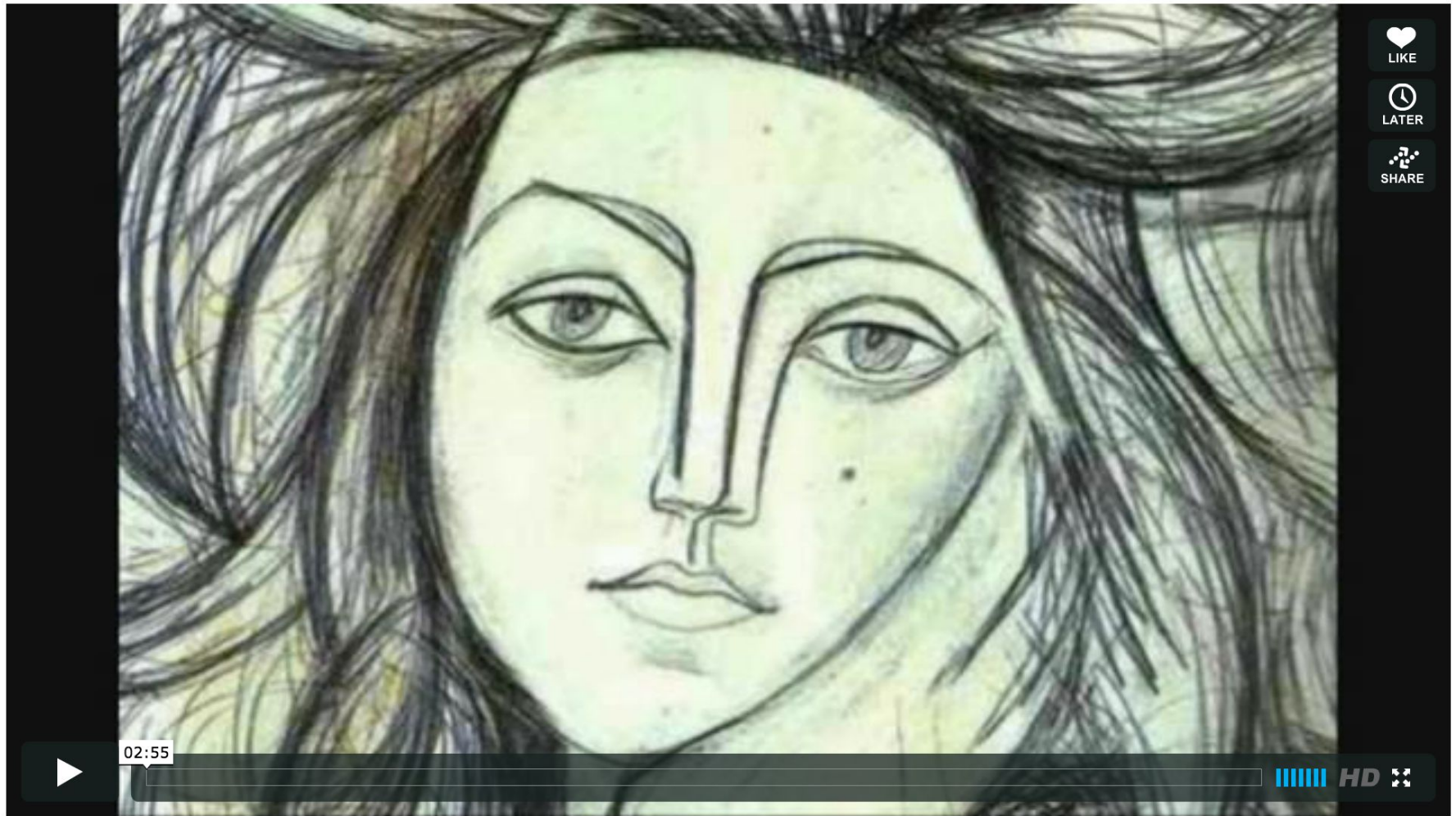


[Wolberg 1996, Recent Advances in Image Morphing]

## Computational Photography

### Connelly Barnes

# Morphing Video: Women in Art



- <http://www.vimeo.com/1456037>

# Terminator 2 Morphing (1991)

[Terminator 2 Clip \(YouTube\)](#)

# Image Warping in Biology

- D'Arcy Thompson

[http://en.wikipedia.org/wiki/D'Arcy\\_Thompson](http://en.wikipedia.org/wiki/D'Arcy_Thompson)

- Importance of shape and structure in evolution

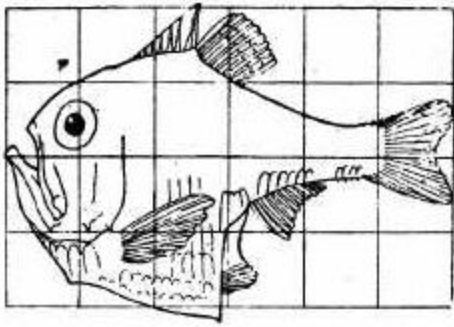
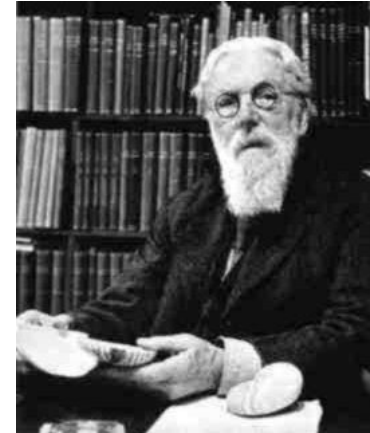
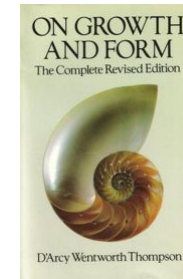


Fig. 517. *Argyropelecus Olfersi*.

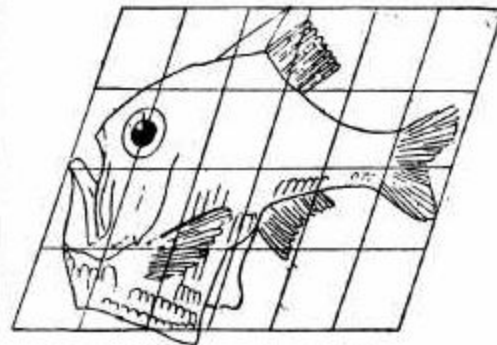
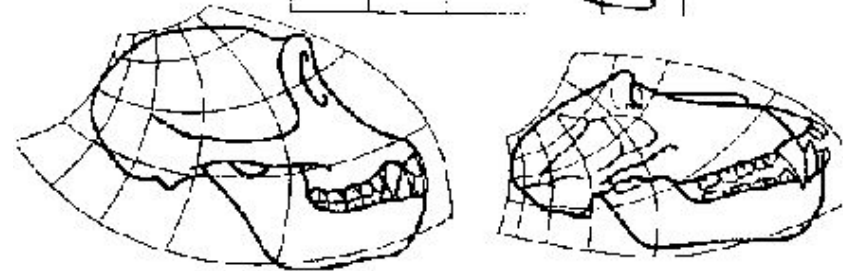
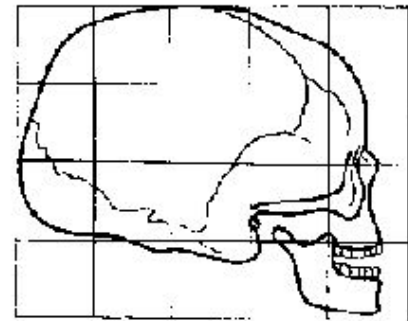


Fig. 518. *Sternoptyx diaphana*.



Skulls of a human, a chimpanzee and a baboon and transformations between them

# Cambrian Explosion



*Opabinia*



*Hallucigenia*



*Pikaia*



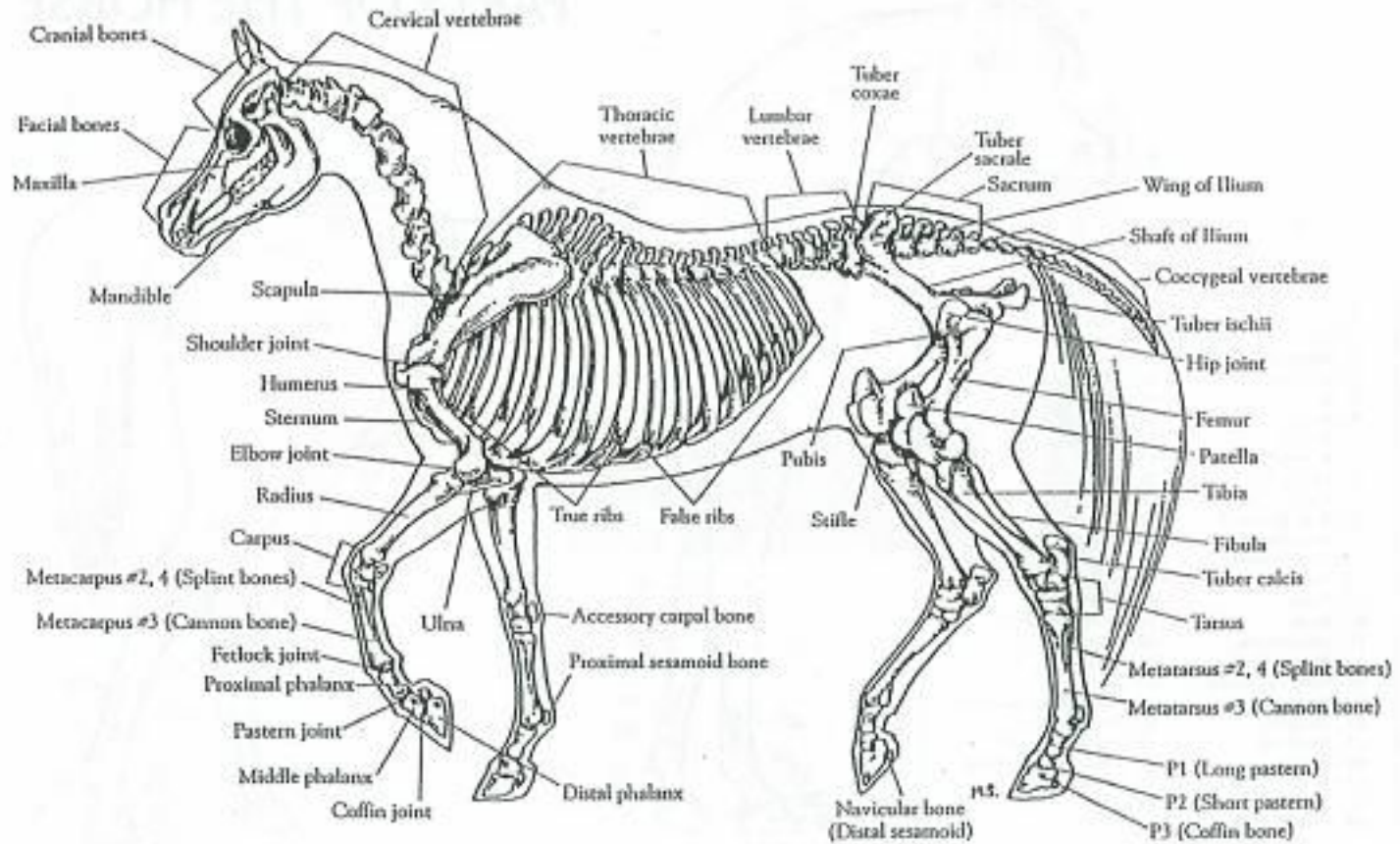
*Marrella*



*Aysheaia*

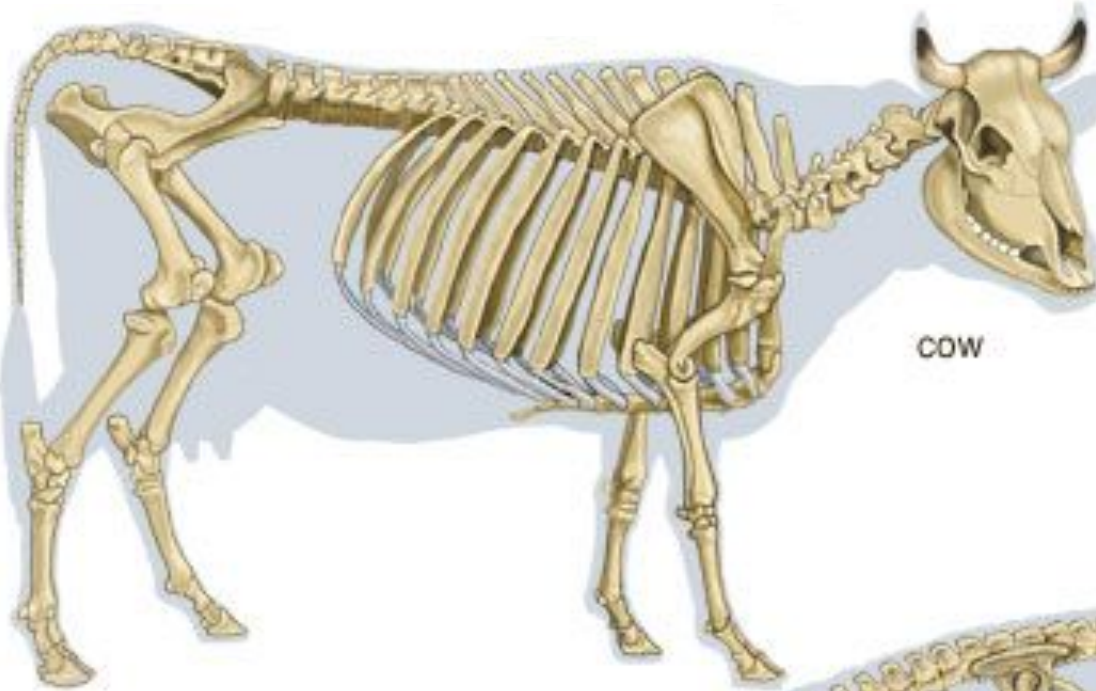
# Skeletons

## SKELTON OF THE HORSE



ARABIAN HORSE ASSOCIATION  
 10805 East Bethany Drive  
 Aurora, CO 80014-2605  
 phone (303) 696-4500 • fax (303) 696-4599  
[www.ArabianHorses.org](http://www.ArabianHorses.org)

# Skeletons



COW

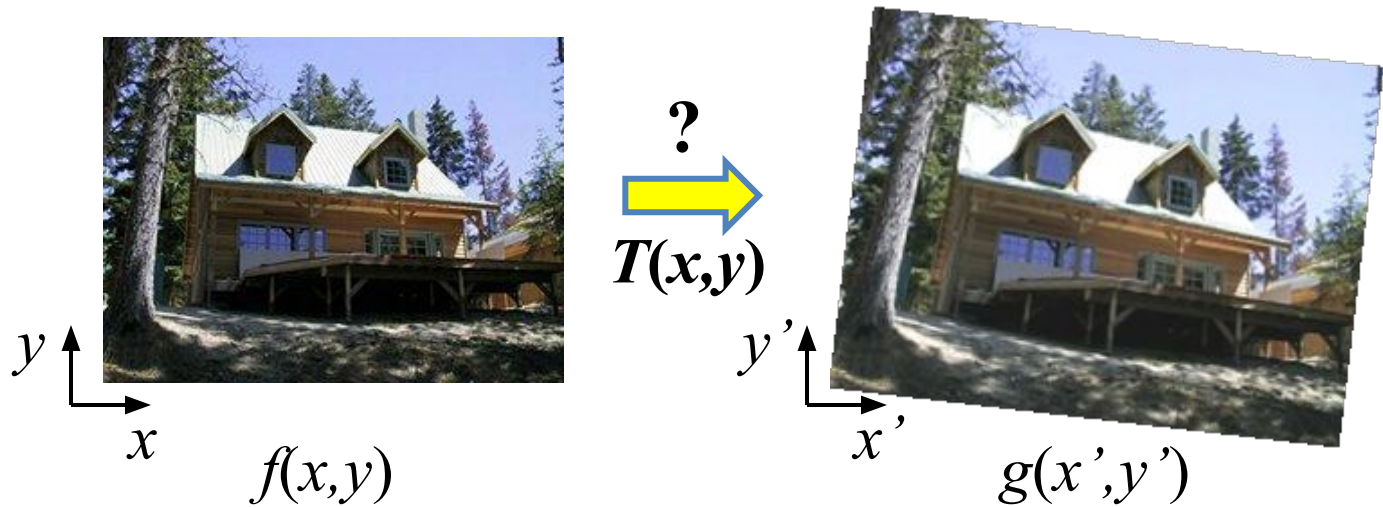


chicken



crocodile

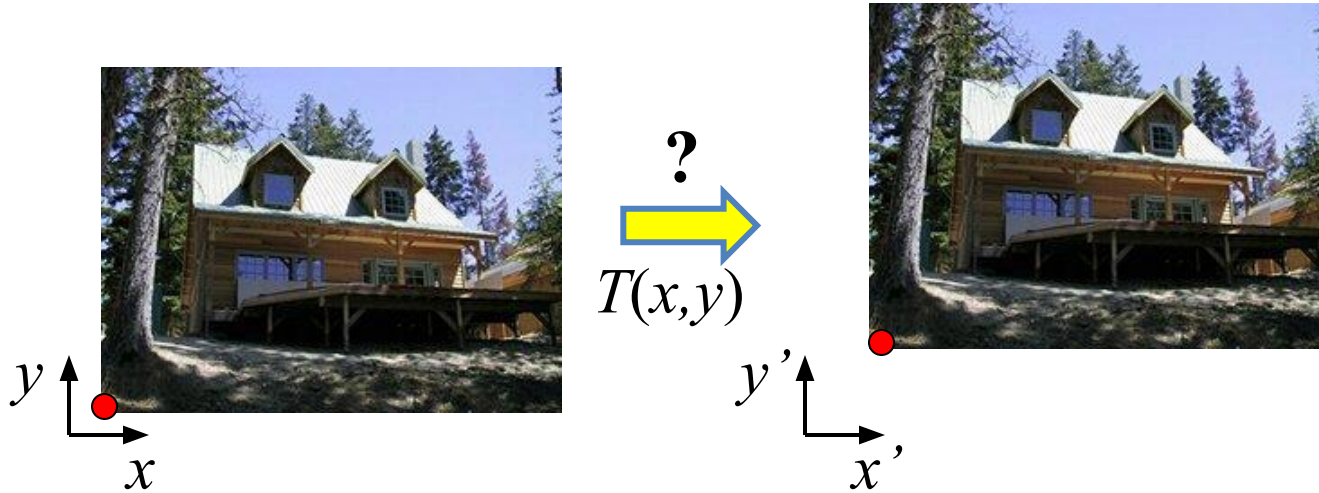
# Recovering Transformations



- What if we know  $f$  and  $g$  and want to recover the transform  $T$ ?
  - e.g. better align photographs you've taken
  - willing to let user provide correspondences
    - How many do we need?



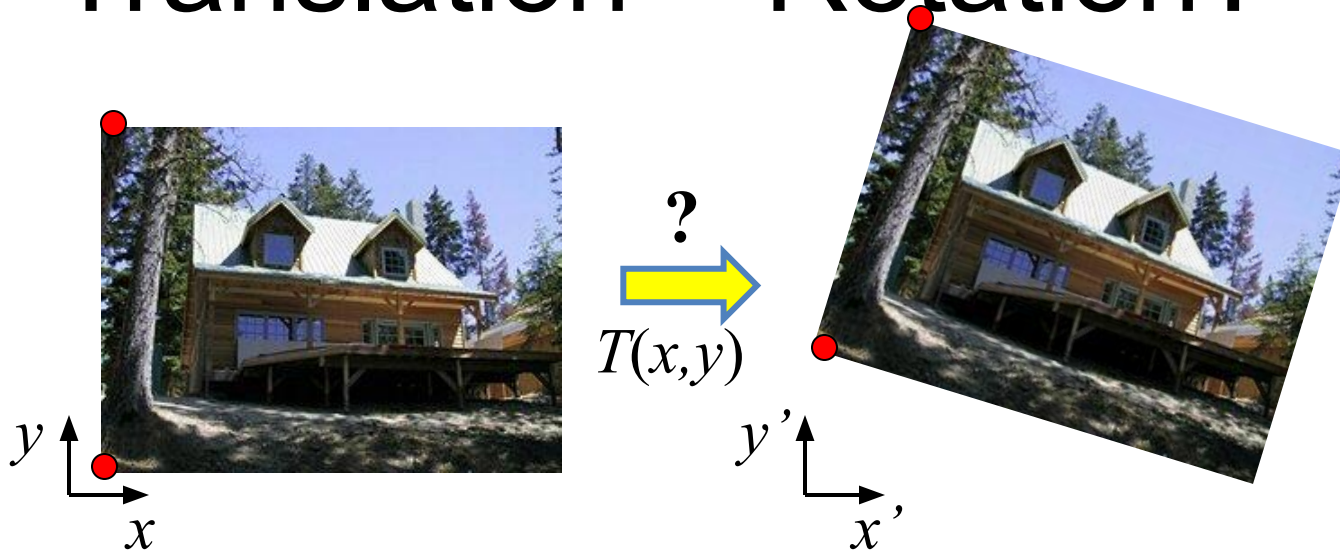
# Translation: # correspondences?



- How many correspondences needed for translation?
- How many Degrees of Freedom?
- What is the transformation matrix?

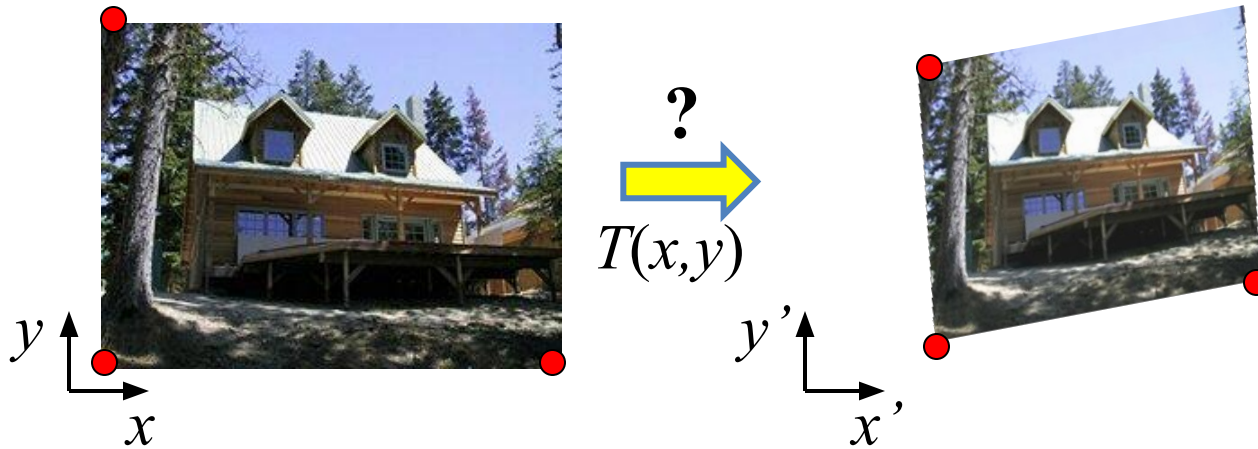
$$\mathbf{M} = \begin{bmatrix} 1 & 0 & p'_x - p_x \\ 0 & 1 & p'_y - p_y \\ 0 & 0 & 1 \end{bmatrix}$$

# Translation + Rotation?



- How many correspondences needed for translation+rotation?
- How many DOF?

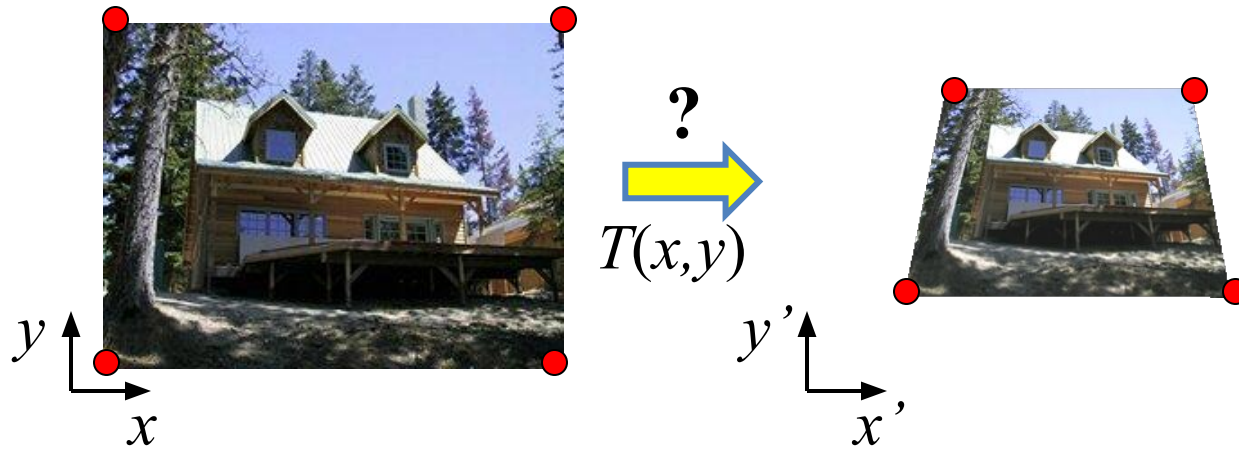
# Affine: # correspondences?



- How many correspondences needed for affine transform?
- How many DOF?

$$T(x, y) = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

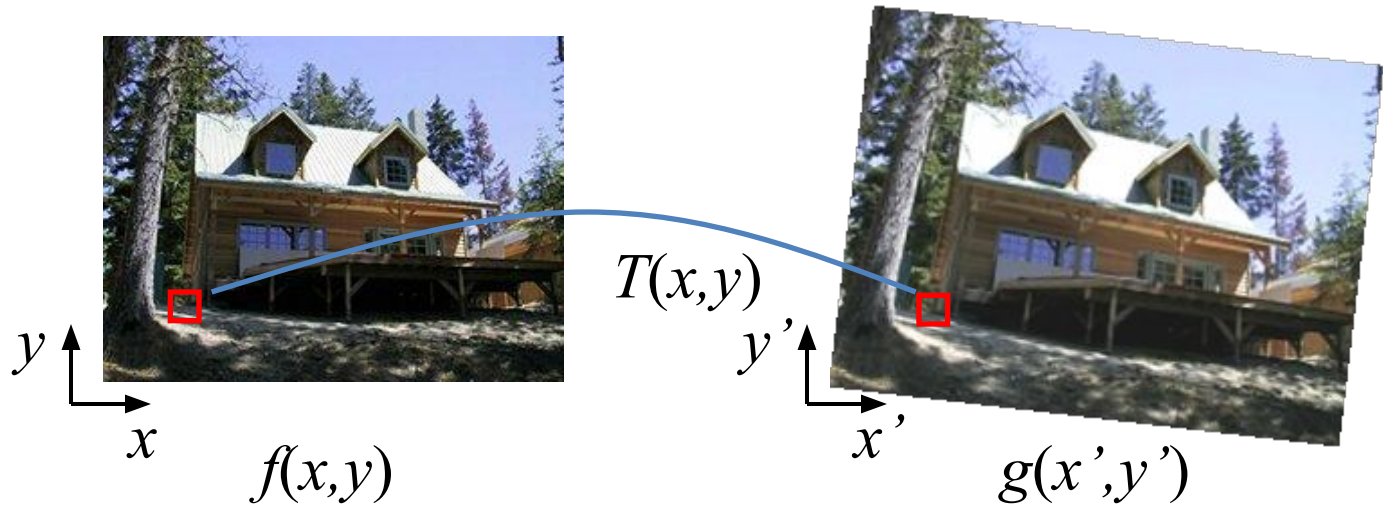
# Projective / Homography



- How many correspondences needed for projective? How many DOF?

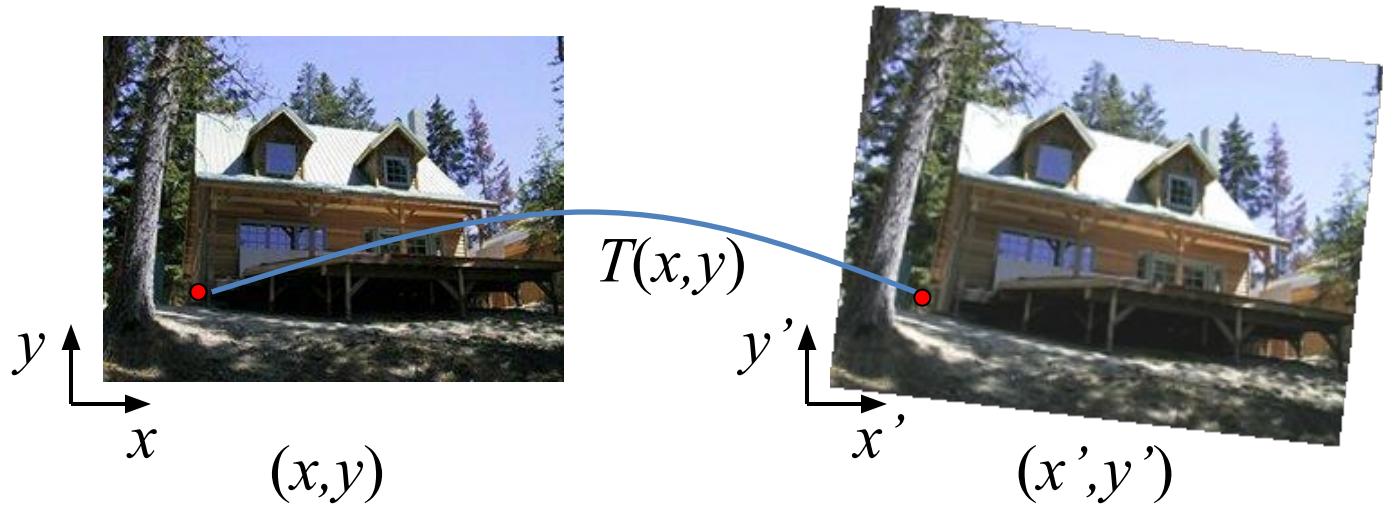
$$T(x, y) = h \left( \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} \right) \quad h(x, y, z) = (x / z, y / z)$$

# Image Warping



- Given a coordinate transform  $(x',y') = T(x,y)$  and a source image  $f(x,y)$ , how do we compute a transformed image  $g(x',y') = f(T(x,y))$ ?

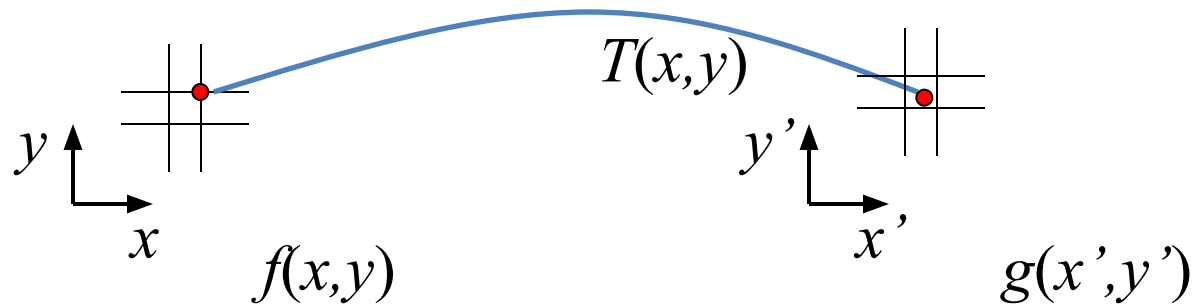
# Forward warping



- Send each pixel  $(x, y)$  to its corresponding location

$(x', y') = T(x, y)$  in the second image

# Forward warping



Q: what if pixel lands “between” two pixels?

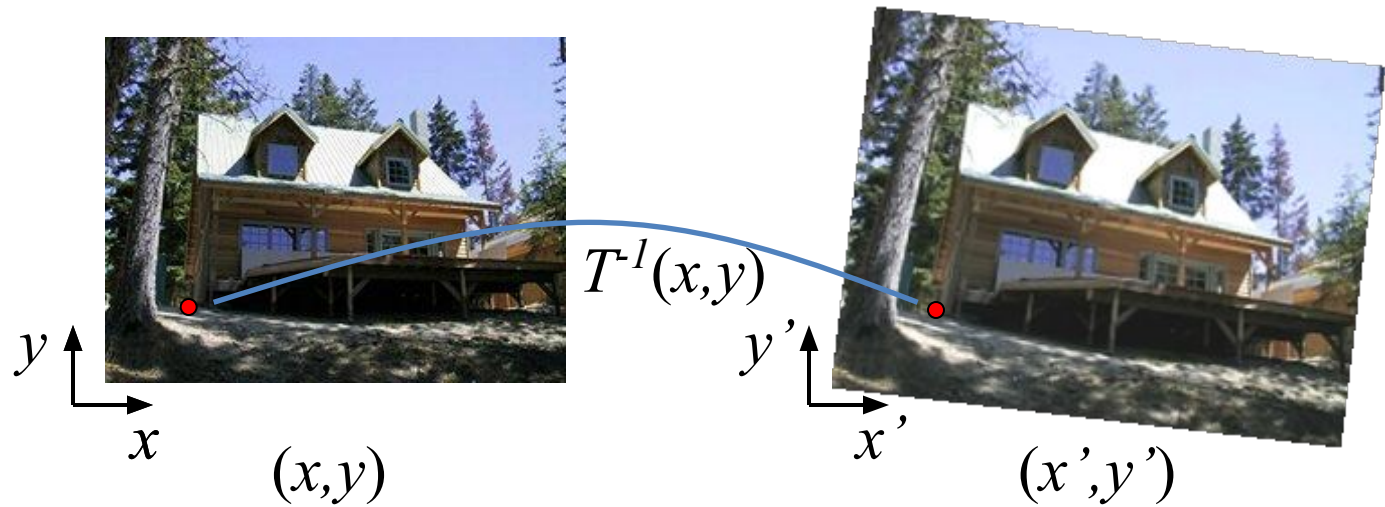
A: distribute color among neighboring pixels ( $x',y'$ )

- Known as “splatting”

- Can also interpolate points in target image:

[griddata](#) (Matlab), [scipy.interpolate.griddata](#) (Python)

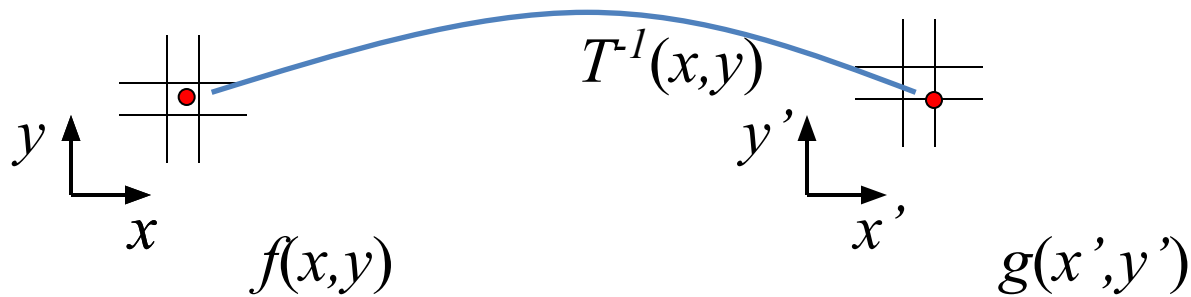
# Inverse warping



- Get each pixel color  $g(x', y')$  from its corresponding location  
 $(x, y) = T^{-1}(x', y')$  in the first image



# Inverse warping



Q: what if pixel comes from “between” two pixels?

A: *Interpolate* color value from neighbors

– nearest neighbor, bilinear, Gaussian, bicubic

– See [interp2](#) (Matlab),

[scipy.interpolate.interp2d](#) (Python)

# Forward vs. inverse warping

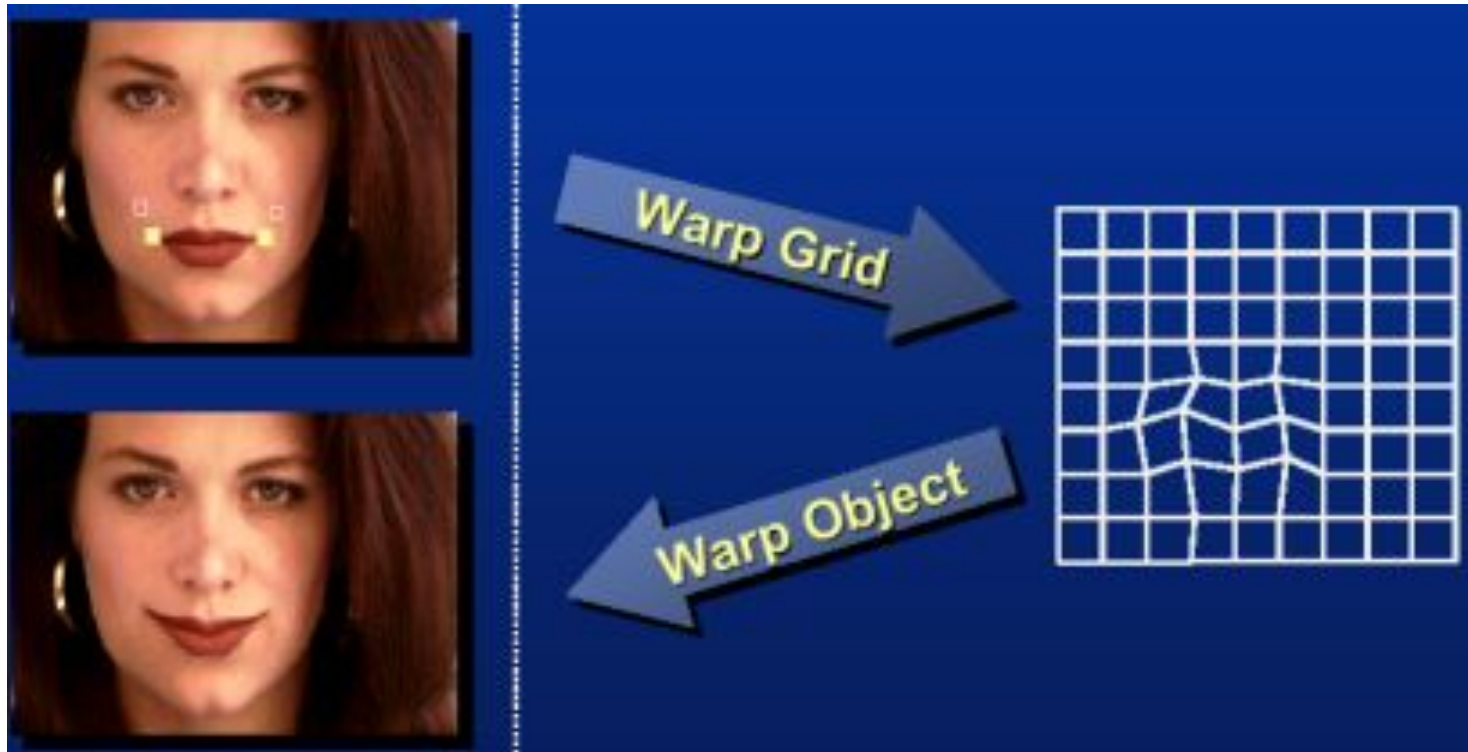
- Q: Which is better?

# Forward vs. inverse warping

- Q: Which is better?
- A: Usually inverse – eliminates holes
  - However, it requires an invertible warp function
  - Not always possible

# How to Obtain Warp Field?

- Move control points to specify a spline warp
- Spline produces a smooth vector field  $T(x, y)$

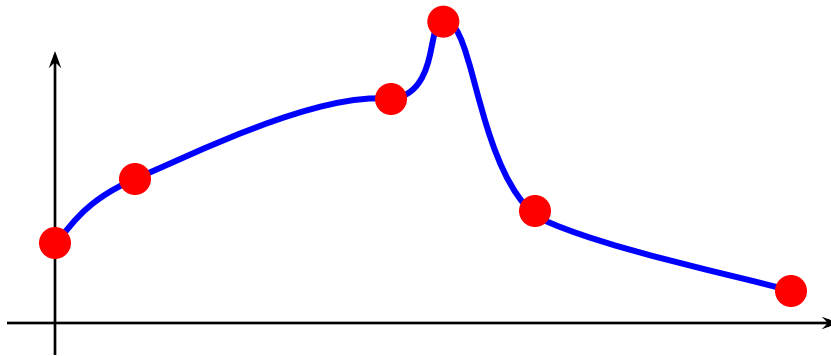


# Warp as Interpolation

- **We are looking for a warping field**
  - A function that given a 2D point, returns a warped 2D point
- **We have a sparse number of correspondences**
  - These specify values of the warping field
- **This is an interpolation problem**
  - Given sparse data, find smooth function

# Interpolation in 1D

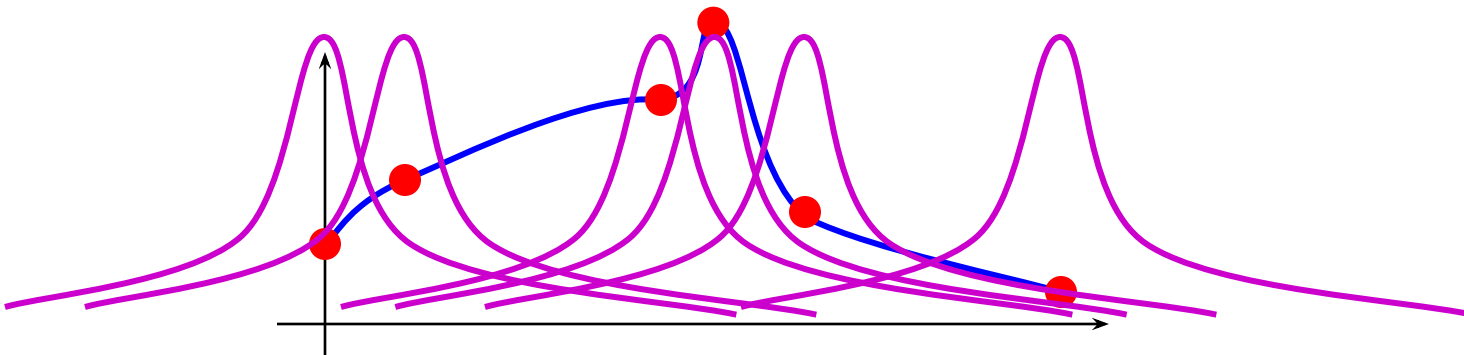
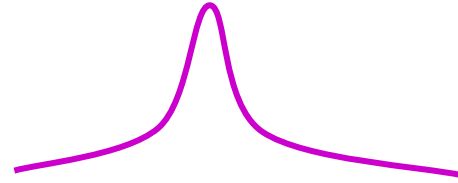
- We are looking for a function  $f$
- We have  $N$  data points:  $x_i, y_i$ 
  - Scattered: spacing between  $x_i$  is non-uniform
- We want  $f$  so that
  - For each  $i, f(x_i)=y_i$
  - $f$  is smooth
- Depending on notion of smoothness, different  $f$



# Radial Basis Functions (RBF)

- Place a smooth kernel  $R$  centered on each data point  $x_i$

$$f(z) = \sum \alpha_i R(\|z - x_i\|)$$



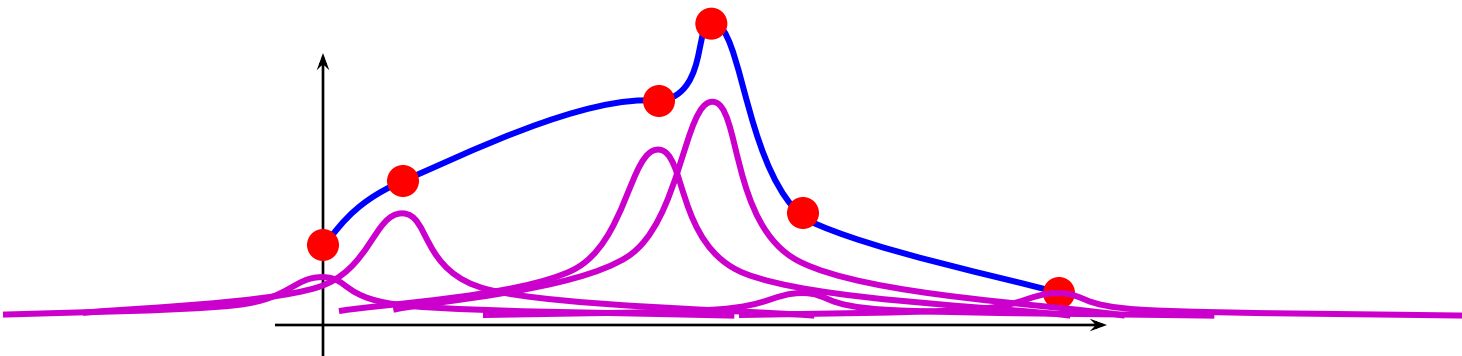
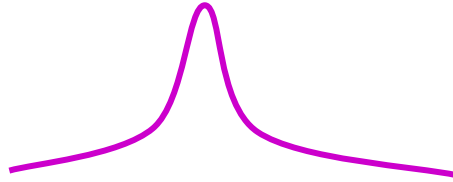
# Radial Basis Functions (RBF)

- Place a smooth kernel  $R$  centered on each data point  $x_i$

$$f(z) = \sum \alpha_i R(\|z - x_i\|)$$

- Find weights  $\alpha_i$  to make sure we interpolate the data

for each  $i$ ,  $f(x_i) = y_i$





# Radial Basis Function Kernels

Linear

$$R(r) = r$$

Cubic

$$R(r) = r^3$$

Quintic

$$R(r) = r^5$$

Thin plate

$$R(r) = r^2 \log r$$

Inverse

$$R(r) = 1 / \sqrt{(r / w)^2 + 1}$$

Multiquadratic

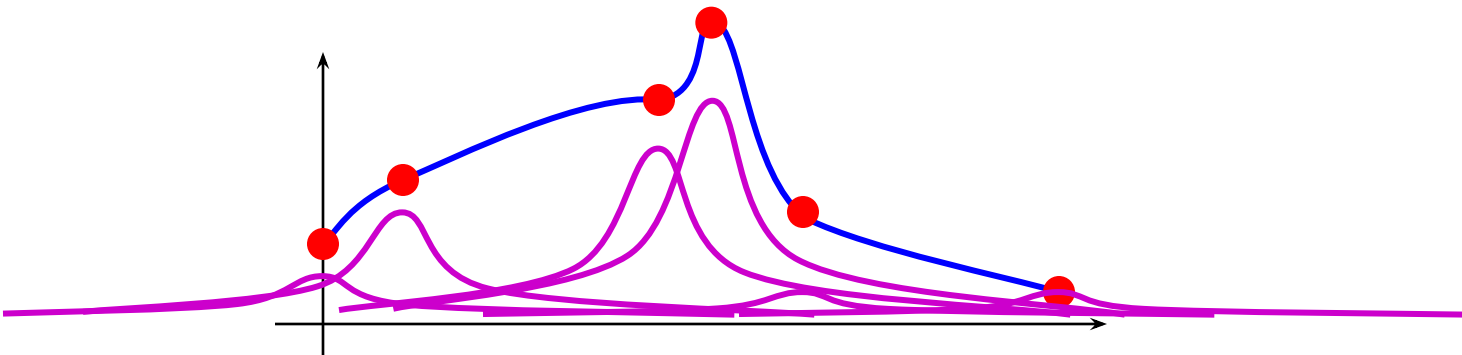
$$R(r) = \sqrt{(r / w)^2 + 1}$$

# Solve RBF Interpolation Problem

$$f(z) = \sum \alpha_i R(\|z - x_i\|)$$

For each  $j$ , 
$$\sum \alpha_i R(\|x_j - x_i\|) = y_j$$

- In 1D:  $N$  equations,  $N$  unknowns, linear solver.
- In  $n$ -D: Denote  $\alpha_i, \mathbf{x}_i, \mathbf{y}_i \in \mathbf{R}^m$   
Solve  $Nm$  equations in  $Nm$  unknowns  $\alpha_i$ .



# RBF Summary

- Interpolates “scattered data”, or data defined only at a few sparse locations.
- Basis functions have infinite extent...
- Python: [scipy.interpolate.Rbf](#)
- MATLAB: Google [“matlab rbf interpolation”](#)  
(3rd party code)

# Applying a warp: use inverse

- Forward warp:

- For each pixel in **input** image

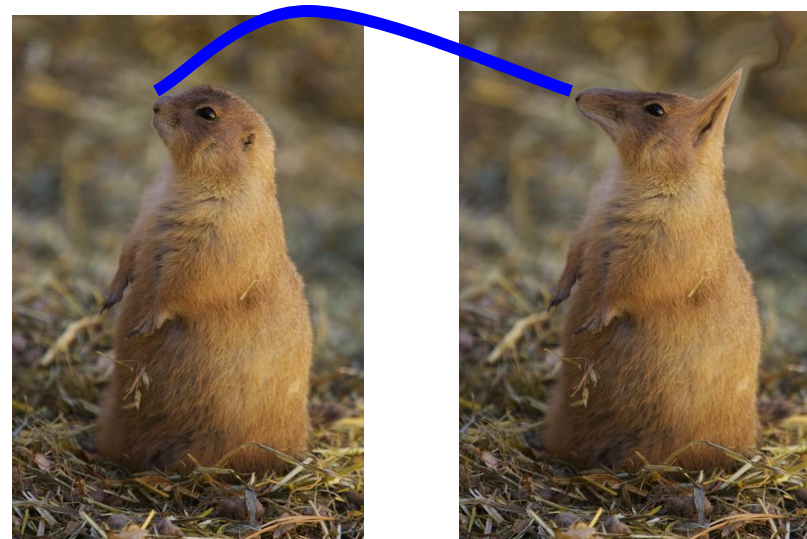
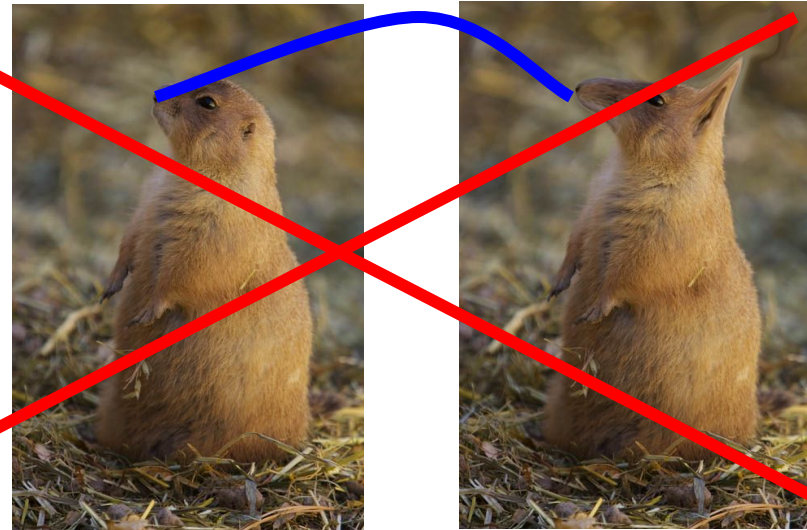
- Paste color **to warped** location in output

- Problem: gaps

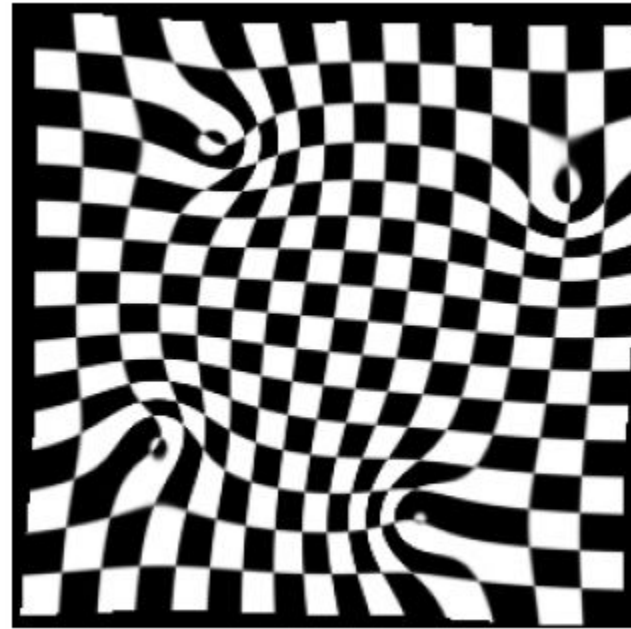
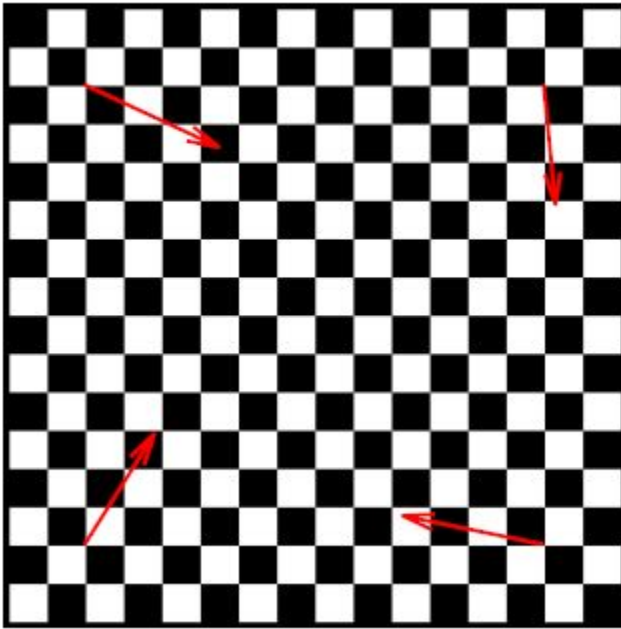
- Inverse warp

- For each pixel in **output** image

- Lookup color **from inverse-warped** location

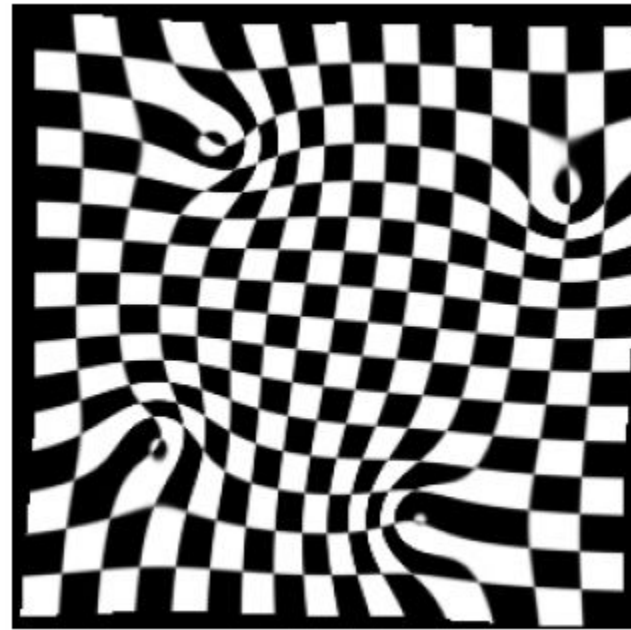
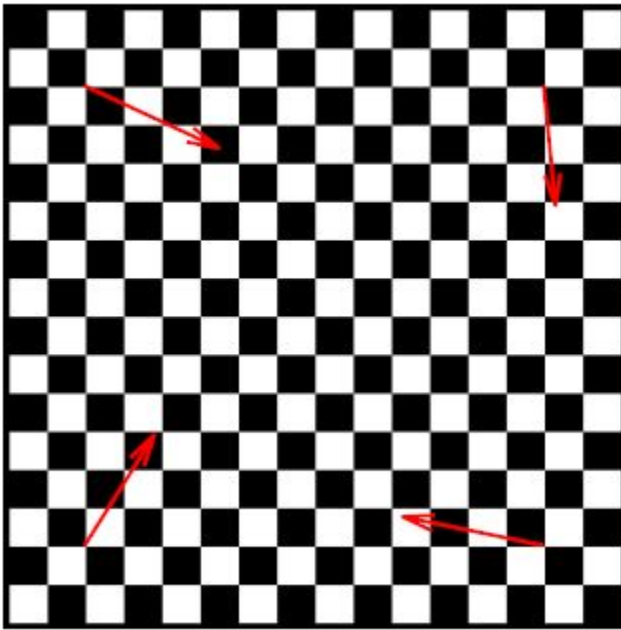


# Example



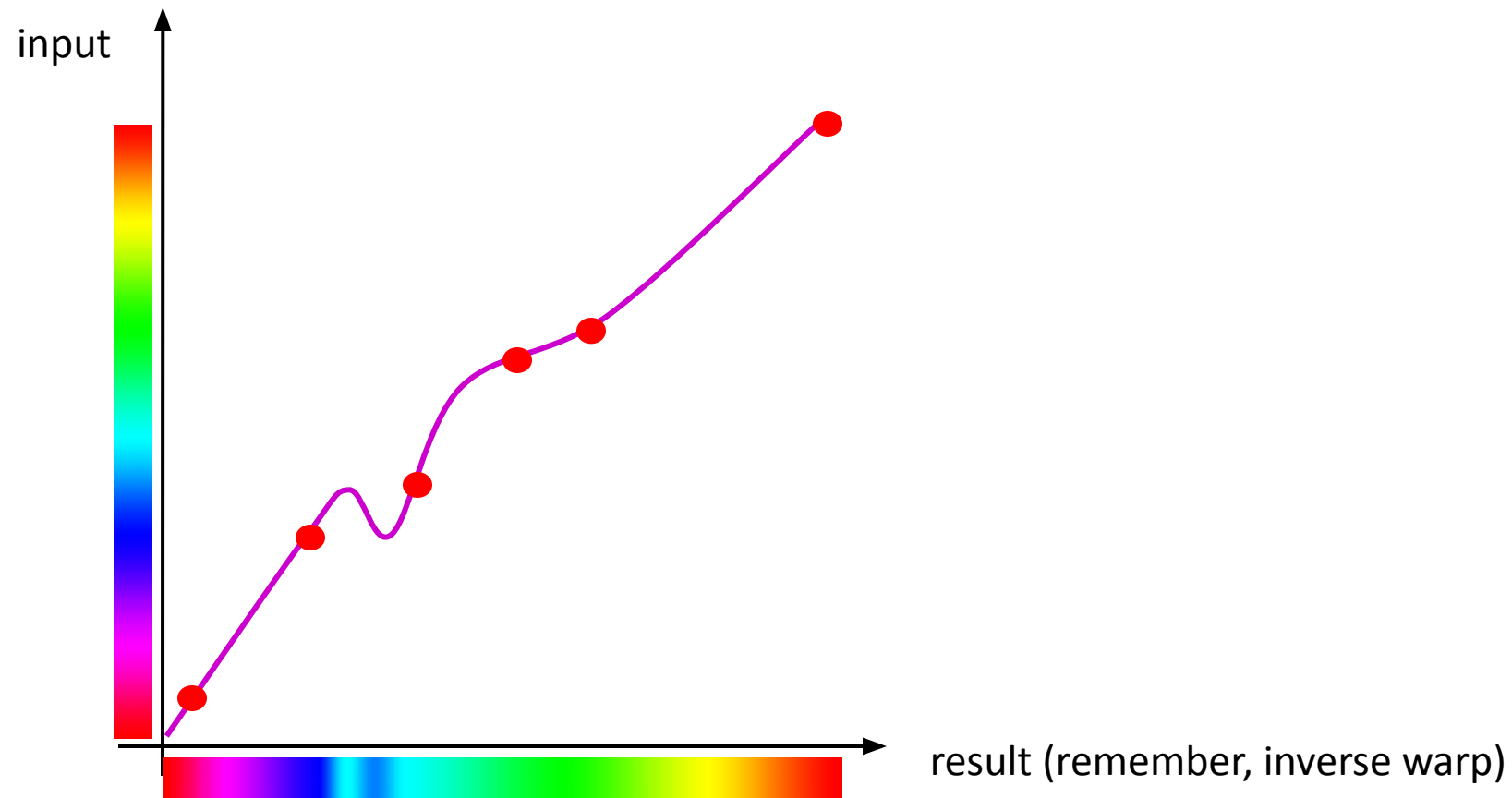
# Example

- Fold problems
  - Oh well...



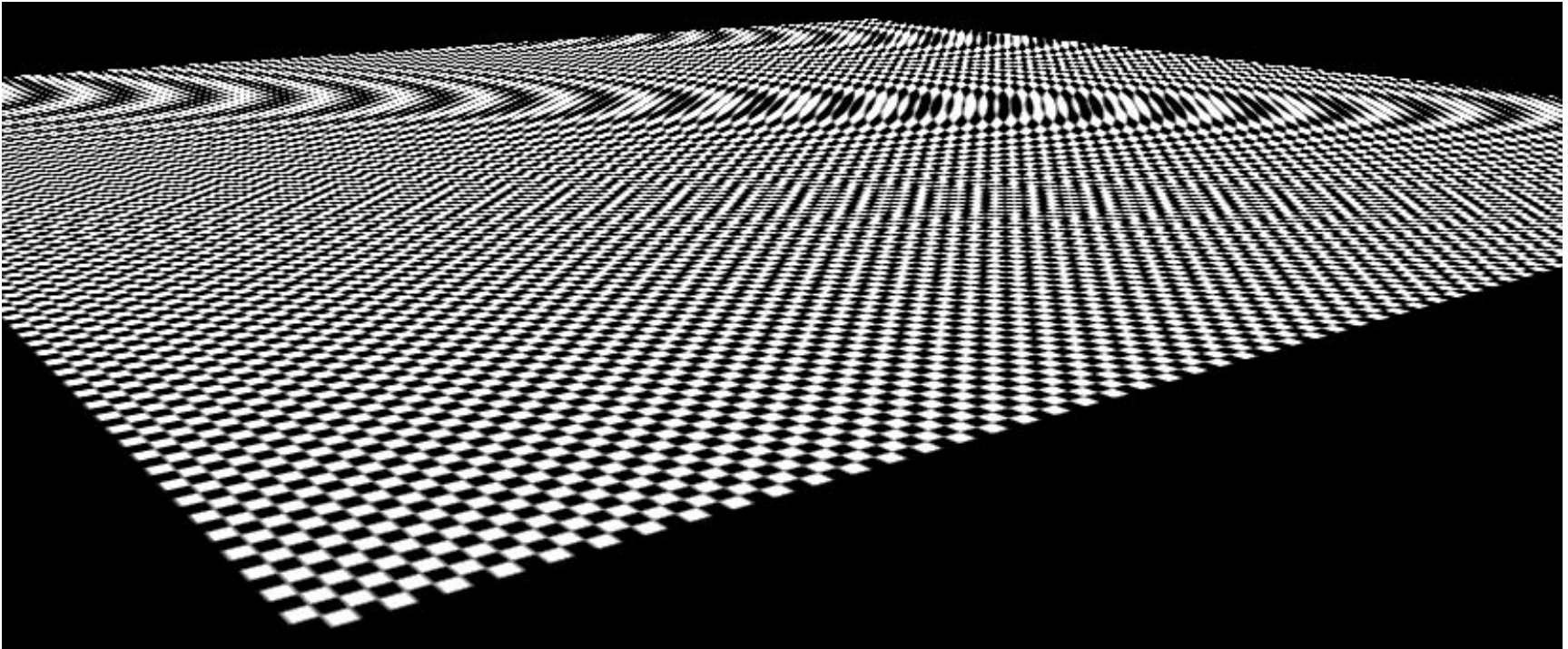
# 1D equivalent of folds

- No guarantee that our 1D RBF is monotonic



# Aliasing Issues with Warping

- *Aliasing* can happen if warps are extreme. This is especially noticeable during animation.



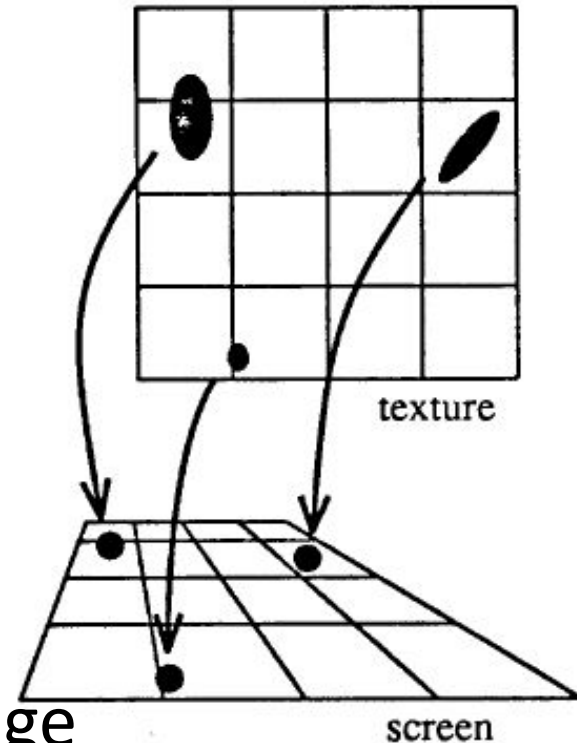


# Aliasing Solution

- Use an ellipsoidal Gaussian:

$$G(x, y) = G_{\sigma_1}(x)G_{\sigma_2}(y)$$

- “Elliptical Weighted Average” (EWA)
- Filter is deformed based on warping.
- For inverse warping, each output (warped) pixel does a weighted average of nearby pixels against the filter.
- Can approximate with circular Gaussian.



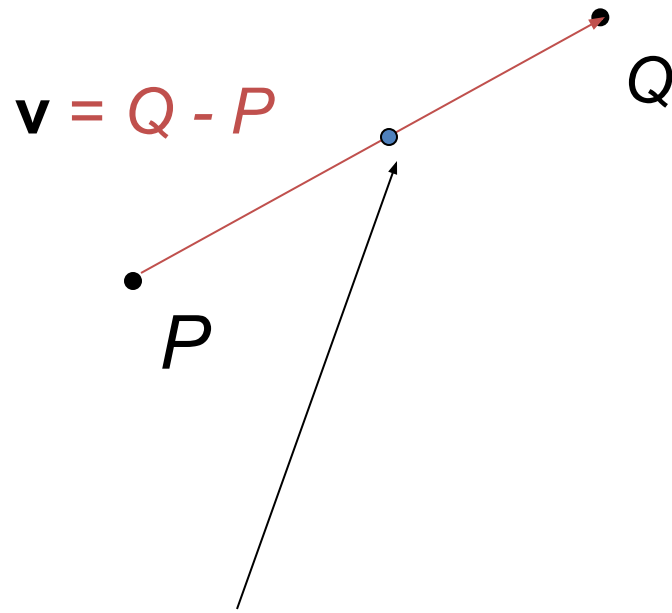
# Morphing = Object Averaging



- The aim is to find “an average” between two objects
  - Not an average of two images of objects...
  - ...but an image of the average object!
  - How can we make a smooth transition in time?
    - Do a “weighted average” over time  $t$
- How do we know what the average object looks like?
  - We haven't a clue!
  - But we can often fake something reasonable
    - Usually required user/artist input

# Linear Interpolation

How can we linearly transition between point  $P$  and point  $Q$ ?



$$P + t\mathbf{v}$$
$$= (1-t)P + tQ, \quad \text{e.g. } t = 0.5$$

- $P$  and  $Q$  can be anything:
  - points on a plane (2D) or in space (3D)
  - Colors in RGB or HSV (3D)
  - Whole images (m-by-n D)... etc.

# Idea #1: Cross-Dissolve



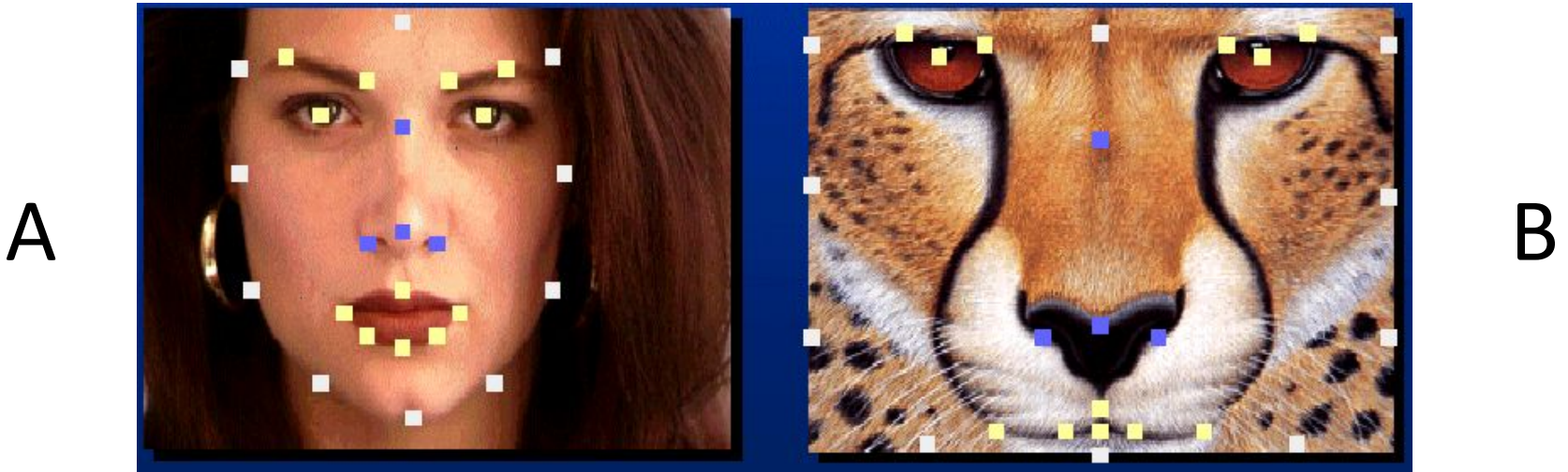
- Interpolate whole images:
- $\text{Image}_{\text{halfway}} = (1-t) \cdot \text{Image}_1 + t \cdot \text{Image}_2$
- This is called **cross-dissolve** in film industry
- But what if the images are not aligned?

# Idea #2: Align, then cross-dissolve



- Align first, then cross-dissolve
  - Alignment using global warp – picture still valid

# Full Morphing



- What if there is no simple global function that aligns two images?
- User specifies corresponding feature points
- Construct warp animations  $A \rightarrow B$  and  $B \rightarrow A$
- Cross dissolve these

# Full Morphing



Image A



# Full Morphing

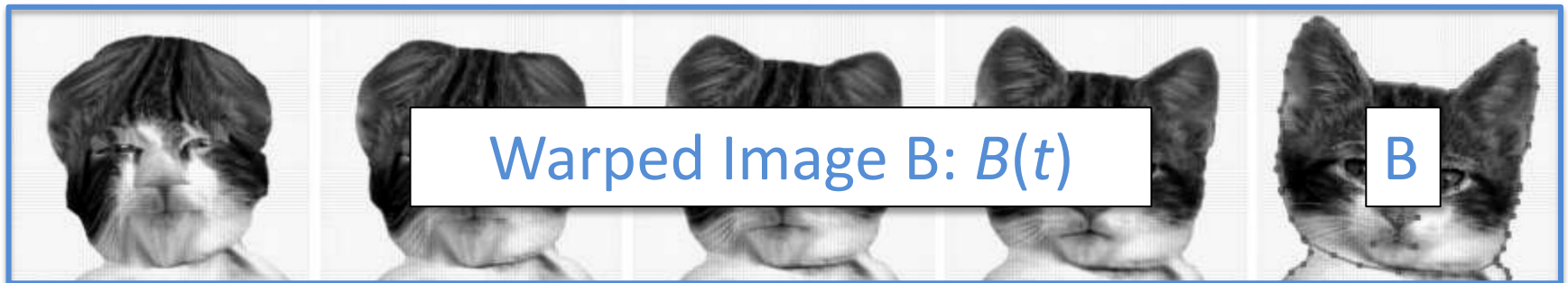
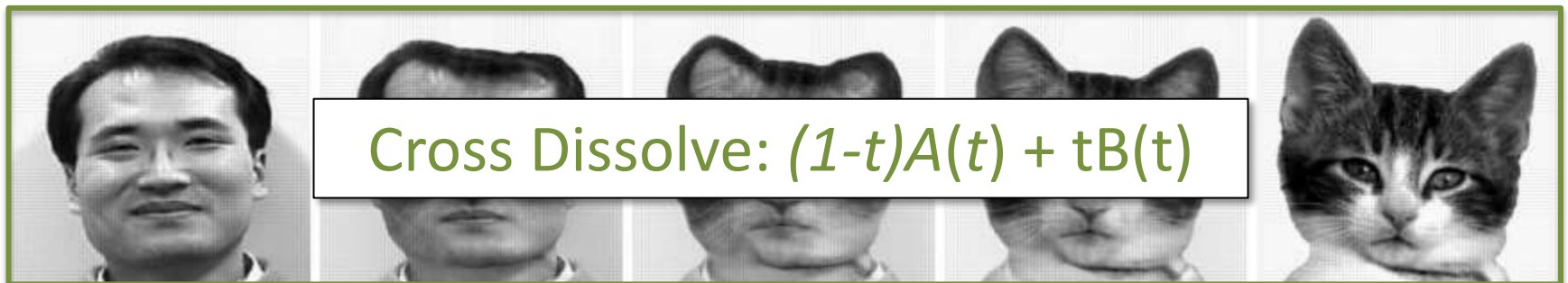
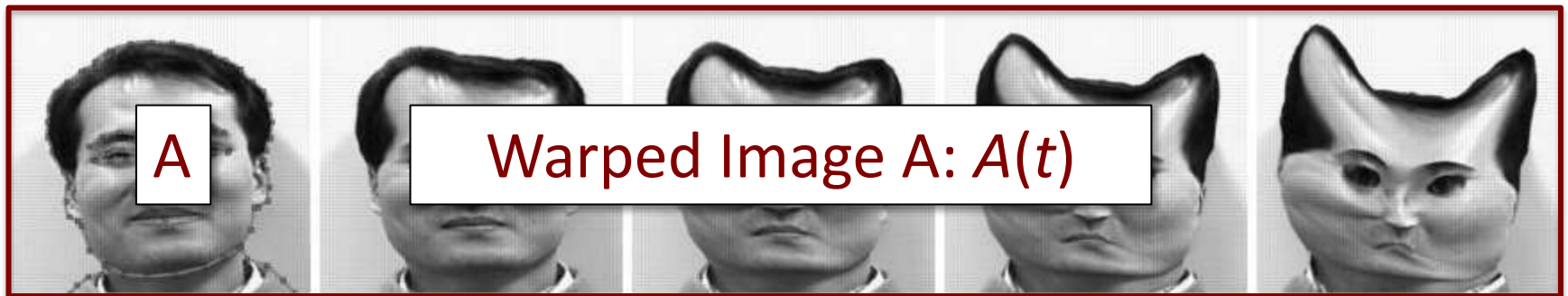
Image B



1. Find warping fields from user constraints (points or lines):
  - Warp field  $T_{AB}(x, y)$  that maps A pixel to B pixel
  - Warp field  $T_{BA}(x, y)$  that maps B pixel to A pixel
2. Make video  $A(t)$  that warps A over time to the shape of B
  - Start warp field at identity and linearly interpolate to  $T_{BA}$
  - Construct video  $B(t)$  that warps B over time to shape of A
3. Cross dissolve these two videos.



# Full Morphing



# Catman!



# Conclusion

- Illustrates general principle in graphics:
  - First register, then blend
- Avoids ghosting

[Michael Jackson - Black or White](#)