# Lecture 3: Barycentric Coordinates and Image Interpolation 

Mark Hasegawa-Johnson

University of Illinois
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Application: Animating a still image

Steps 2 and 3: Draw and Move Triangles

Step 4: Find the mapping between original and moved pixels: Barycentric coordinates

Step 5: Find the color of the source pixel: Bilinear interpolation

Conclusion

## Outline

Application: Animating a still image

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## Conclusion

## Strategy

1. Use affine projection to rotate, scale, and shear the $X R M B$ points so that they match the shape of the MRI as well as possible.
2. Draw triangles on the MRI so that every pixel is inside a triangle.
3. Move the triangles.
4. Map each integer pixel in the target image to a real-valued pixel in the original image
5. Use bilinear interpolation to find the color of the pixel

Step 1 (Last time): Use affine projection to map XRMB to MRI


## Steps 2 and 3: Draw triangles, then move them



Steps 4 and 5: Find the color of each pixel


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## Triangulation

A triangulation of a set of points $=a$ set of triangles, connecting those points, that covers the convex hull of those points.
There are many very cool algorithms that will automatically triangulate a set of points.


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Triangulation for the machine problem is provided for you


Once you have drawn the triangles, they move along with the points


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Source image
The source image is divided into non-overlapping triangles, $X_{k}$.


## Target image

In the target image, those triangles have moved to new locations, $Y_{k}$.



## Problem Statement: Moving pixels

- The size of the image we want to construct is $m \times n$.
- Consider a particular augmented target pixel, $y=\left[\begin{array}{c}y_{1} \\ y_{2} \\ 1\end{array}\right]$, where $0 \leq y_{1} \leq m-1$ and $0 \leq y_{2} \leq n-1$ are both integers.
- In order to find the color of target pixel $y$, we want to find out which source pixel, $x$, was moved to that location. Assume that pixels only move around - they don't change color while they move.


## Problem Statement: Piece-wise affine transform

- Suppose that $y \in Y_{k}$, the $k^{\text {th }}$ target triangle.
- We know therefore that $x \in X_{k}$. But there are lots of pixels inside $X_{k}$. Which one is it?
- Let's assume that, within each triangle, the pixels move according to an affine transform. In other words, if $y \in Y_{k}$, and if we already knew $A_{k}$, then we could find $x$ by solving:

$$
y=A_{k} x
$$

where

$$
y=\left[\begin{array}{c}
y_{1} \\
y_{2} \\
1
\end{array}\right], \quad x=\left[\begin{array}{c}
x_{1} \\
x_{2} \\
1
\end{array}\right], \quad A_{k}=\left[\begin{array}{ccc}
a_{1,1} & a_{1,2} & a_{1,3} \\
a_{2,1} & a_{2,2} & a_{2,3} \\
0 & 0 & 1
\end{array}\right]
$$

## Piece-wise affine transform

$$
\text { target point: } y=\left[\begin{array}{c}
y_{1} \\
y_{2} \\
1
\end{array}\right], \quad \text { source point: } x=\left[\begin{array}{c}
x_{1} \\
x_{2} \\
1
\end{array}\right]
$$

Definition: if $y$ is in the $k^{\text {th }}$ triangle in the output image, then we want to use the $k^{\text {th }}$ affine transform:

$$
y=A_{k} x, \quad x=A_{k}^{-1} y
$$

If it is known that $x=A_{k}^{-1} y$ for some unknown affine transform matrix $A_{k}$,
then
the method of barycentric coordinates finds $x$ without ever finding $A_{k}$.

## Barycentric Coordinates

Barycentric coordinates turns the problem on its head. Suppose $y$ is in a triangle with corners at $y_{1}, y_{2}$, and $y_{3}$. That means that

$$
y=\beta_{1} y_{1}+\beta_{2} y_{2}+\beta_{3} y_{3}
$$

where

$$
0 \leq \beta_{1}, \beta_{2}, \beta_{3} \leq 1
$$

and

$$
\beta_{1}+\beta_{2}+\beta_{3}=1
$$



## Barycentric Coordinates

Suppose that all three of the corners are transformed by some affine transform $A$, thus

$$
x_{1}=A y_{1}, \quad x_{2}=A y_{2}, \quad x_{3}=A y_{3}
$$

Then if

$$
\text { If: } y=\beta_{1} y_{1}+\beta_{2} y_{2}+\beta_{3} y_{3}
$$

Then:

$$
\begin{aligned}
x & =A y \\
& =\beta_{1} A y_{1}+\beta_{2} A y_{2}+\beta_{3} A y_{3} \\
& =\beta_{1} x_{1}+\beta_{2} x_{2}+\beta_{3} x_{3}
\end{aligned}
$$

In other words, once we know $\beta$, we no longer need to find $A$. We only need to know where the corners of the triangle have moved.

## Barycentric Coordinates

If

$$
y=\beta_{1} y_{1}+\beta_{2} y_{2}+\beta_{3} y_{3}
$$

Then

$$
x=\beta_{1} x_{1}+\beta_{2} x_{2}+\beta_{3} x_{3}
$$



## How to find Barycentric Coordinates

But how do you find $\beta_{1}, \beta_{2}$, and $\beta_{3}$ ?

$$
\left[\begin{array}{c}
y_{1} \\
y_{2} \\
1
\end{array}\right]=\beta_{1} y_{1}+\beta_{2} y_{2}+\beta_{3} y_{3}=\left[\begin{array}{ccc}
y_{1,1} & y_{1,2} & y_{1,3} \\
y_{2,1} & y_{2,2} & y_{2,3} \\
1 & 1 & 1
\end{array}\right]\left[\begin{array}{l}
\beta_{1} \\
\beta_{2} \\
\beta_{3}
\end{array}\right]
$$

Write this as:

$$
y=Y \beta
$$

Therefore

$$
\beta=Y^{-1} y
$$

This always works: the matrix $Y$ is always invertible, unless all three of the points $y_{1}, y_{2}$, and $y_{3}$ are on a straight line.

## How do you find out which triangle the point is in?

- Suppose we have $K$ different triangles, each of which is characterized by a $3 \times 3$ matrix of its corners

$$
Y_{k}=\left[y_{1}^{(k)}, y_{2}^{(k)}, y_{3}^{(k)}\right]
$$

where $y_{m}^{(k)}$ is the $m^{\text {th }}$ corner of the $k^{\text {th }}$ triangle.

- Notice that, for any point $y$, for ANY triangle $Y_{k}$, we can find

$$
\beta=Y_{k}^{-1} y
$$

- However, the coefficients $\beta_{1}, \beta_{2}$, and $\beta_{3}$ will all be between 0 and 1 if and only if the point $y$ is inside the triangle $Y_{k}$. Otherwise, some of the elements of $\beta$ must be negative.


## The Method of Barycentric Coordinates

To construct the animated output image frame $J\left[y_{2}, y_{1}\right]$, we do the following things:

- First, for each of the reference triangles $X_{k}$ in the input image $I\left(x_{2}, x_{1}\right)$, decide where that triangle should move to. Call the new triangle location $Y_{k}$.
- Second, for each integer output pixel $\left(y_{1}, y_{2}\right)$ :
- For each of the triangles, find $\beta=Y_{k}^{-1} y$.
- Choose the triangle for which all of the elements of $\beta$ are $0 \leq \beta_{m} \leq 1$.
- Find $x=X_{k} \beta$.
- Find the color of pixel $I\left(x_{2}, x_{1}\right)$ in the input image.
- Set $J\left[y_{2}, y_{1}\right]=I\left(x_{2}, x_{1}\right)$.


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## Conclusion

Integer Target Points
Now let's suppose that you've figured out the coordinate transform: for any given $J\left[y_{2}, y_{1}\right]$, you've figured out which pixel should be used to create it $\left(J\left[y_{2}, y_{1}\right]=I\left(x_{2}, x_{1}\right)\right)$.

$$
x=X_{k} \beta=X_{k} Y_{k}^{-1} y
$$

The Problem: Non-Integer Input Points
If $\left[y_{2}, y_{1}\right]$ are integers, then usually, $\left(x_{2}, x_{1}\right)$ are not integers.

## Image Interpolation

It is necessary to find $I(v, u)$ at non-integer values of $(v, u)$ by interpolating between the integer-valued pixels provided in the image file.
Given the pixels of $I[n, m]$ at integer values of $m$ and $n$, we can interpolate using an interpolation kernel $h(v, u)$ :

$$
I(v, u)=\sum_{m} \sum_{n} I[n, m] h(v-n, u-m)
$$

## Piece-Wise Constant Interpolation

$$
\begin{equation*}
I(v, u)=\sum_{m} \sum_{n} I[n, m] h(v-n, u-m) \tag{1}
\end{equation*}
$$

For example, suppose

$$
h(v, u)= \begin{cases}1 & 0 \leq u<1, \quad 0 \leq v<1 \\ 0 & \text { otherwise }\end{cases}
$$

Then Eq. (1) is the same as just truncating $u$ and $v$ to the next-lower integer, and outputting that number:

$$
I(v, u)=I[\lfloor v\rfloor,\lfloor u\rfloor]
$$

where $\lfloor u\rfloor$ means "the largest integer smaller than $u$ ".

## Example: Original Image

For example, let's downsample this image, and then try to recover it by image interpolation:


## Example: Downsampled Image

Here's the downsampled image:


## Example: Upsampled Image

Here it is after we upsample it back to the original resolution (insert 3 zeros between every pair of nonzero columns):


## Example: PWC Interpolation

Here is the piece-wise constant interpolated image:


## Bi-Linear Interpolation

$$
I(v, u)=\sum_{m} \sum_{n} I[n, m] h(v-n, u-m)
$$

For example, suppose

$$
h(v, u)=\max (0,(1-|u|)(1-|v|))
$$

Then Eq. (1) is the same as piece-wise linear interpolation among the four nearest pixels. This is called bilinear interpolation because it's linear in two directions.

$$
\begin{aligned}
m & =\lfloor u\rfloor, \quad e=u-m \\
n & =\lfloor v\rfloor, \quad f=v-m \\
I(v, u) & =(1-e)(1-f) I[n, m]+(1-e) f l[n, m+1] \\
& +e(1-f) I[n+1, m]+e f l[n+1, m+1]
\end{aligned}
$$

## Example: Upsampled Image

Here's the upsampled image again:


## Example: Bilinear Interpolation

Here it is after bilinear interpolation:


## PWC and PWL Interpolator Kernels

Bilinear interpolation uses a PWL interpolation kernel, which does not have the abrupt discontiuity of the PWC interpolator kernel.


## Sinc Interpolation

$$
I(v, u)=\sum_{m} \sum_{n} I[n, m] h(v-n, u-m)
$$

For example, suppose

$$
h(v, u)=\operatorname{sinc}(\pi u) \operatorname{sinc}(\pi v)
$$

Then Eq. (1) is an ideal band-limited sinc interpolation. It guarantees that the continuous-space image, $I(v, u)$, is exactly a band-limited $\mathrm{D} / \mathrm{A}$ reconstruction of the digital image $I[n, m]$.

## Sinc Interpolation

Here is the cat after sinc interpolation:


## Summary of interpolation methods

- PWC interpolation results in a blocky image
- Sinc interpolation results in a smooth image, and would be perfect if the input image was infinite in size, but since real-world images have edges, the sinc interpolation produces ripple artifacts
- Bilinear interpolation is a very efficient solution with good results
- Better results are available using deep-learning-based super-resolution neural nets, but only after the neural net has been trained for a few weeks!


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2. Draw triangles on the MRI so that every pixel is inside a triangle.
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