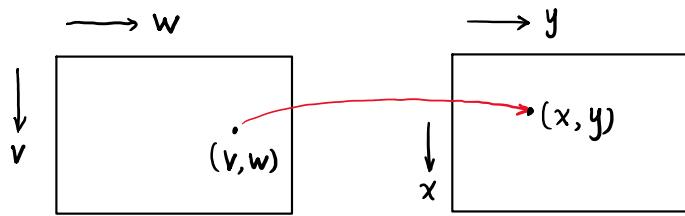


## Lecture 9

### Geometric Spatial Transformation



$$(x, y) = T(v, w)$$

↑  
transformation

$$\underbrace{\begin{bmatrix} x & y & 1 \end{bmatrix}}_{1 \times 3} = \underbrace{\begin{bmatrix} v & w & 1 \end{bmatrix}}_{1 \times 3} \cdot \underbrace{T}_{3 \times 3} = \begin{bmatrix} v & w & 1 \end{bmatrix} \begin{bmatrix} t_{11} & t_{12} & 0 \\ t_{21} & t_{22} & 0 \\ t_{31} & t_{32} & 1 \end{bmatrix}$$

Translation:

$$T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 20 & 20 & 1 \end{bmatrix}$$

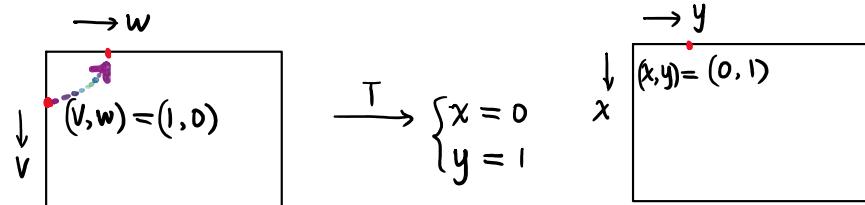
$$\underbrace{\begin{bmatrix} x & y & 1 \end{bmatrix}}_{1 \times 3} = \underbrace{\begin{bmatrix} v & w & 1 \end{bmatrix}}_{1 \times 3} \cdot \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 20 & 20 & 1 \end{bmatrix} = \begin{bmatrix} v+20 & w+20 & 1 \end{bmatrix}$$

Rotation:

$$\begin{cases} x = v \cos \theta - w \sin \theta \\ y = v \sin \theta + w \cos \theta \end{cases} \Rightarrow \underbrace{\begin{bmatrix} x & y & 1 \end{bmatrix}}_{1 \times 3} = \underbrace{\begin{bmatrix} v & w & 1 \end{bmatrix}}_{1 \times 3} \cdot \begin{bmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\text{e.g., } \theta = \frac{\pi}{2}$$

$$\begin{cases} x = -w \\ y = v \end{cases}$$



Counter-clockwise rotation by  $90^\circ$

## imwarp

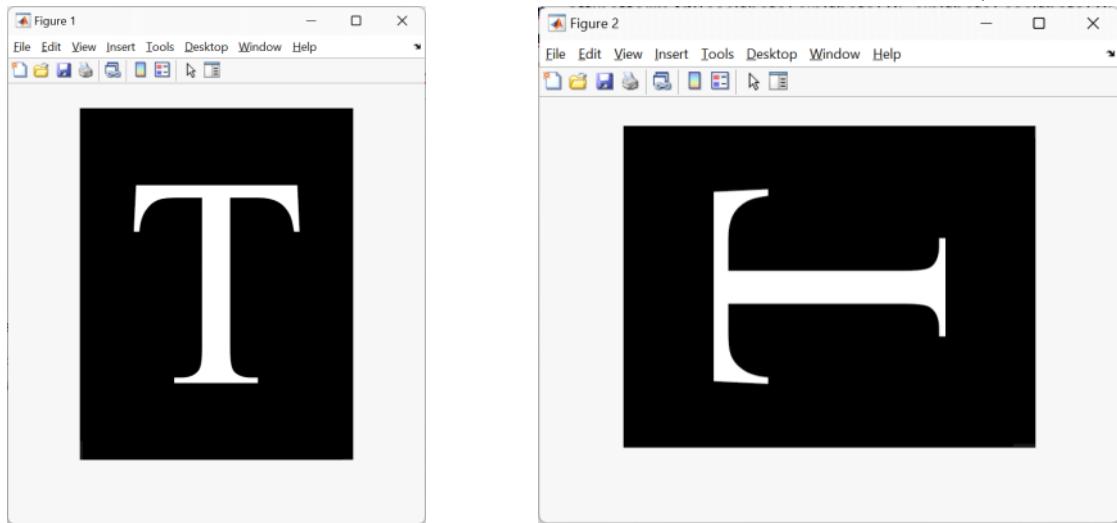
Apply geometric transformation to image

## affinetform2d

2-D affine geometric transformation  
Since R2022b

```
>> theta = pi/2;  
>> I = imread('Fig0236(a)(letter_T).tif');  
>> tform = affinetform2d([cos(theta) sin(theta) 0; -sin(theta) cos(theta) 0; 0 0 1]);  
>> J = imwarp(I, tform);  
>> figure; imshow(J)
```

counter-clockwise rotation by 90°



Horizontal Shear

$$\begin{aligned}x &= v \\y &= 0.5v + w\end{aligned} \Rightarrow T = \begin{bmatrix} 1 & 0.5 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

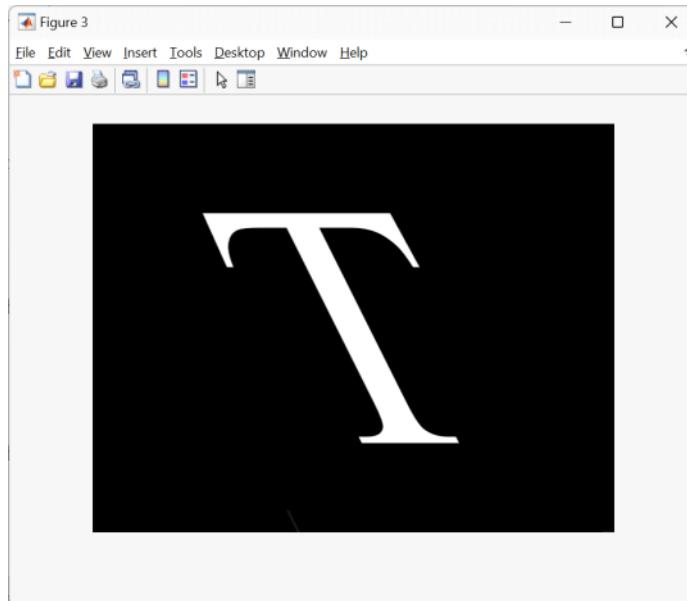
$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} v \\ w \\ 1 \end{bmatrix} \begin{bmatrix} t_{11} & t_{12} & 0 \\ t_{21} & t_{22} & 0 \\ t_{31} & t_{32} & 1 \end{bmatrix}$$

```
>> T = [1 0.5 0; 0 1 0; 0 0 1]
```

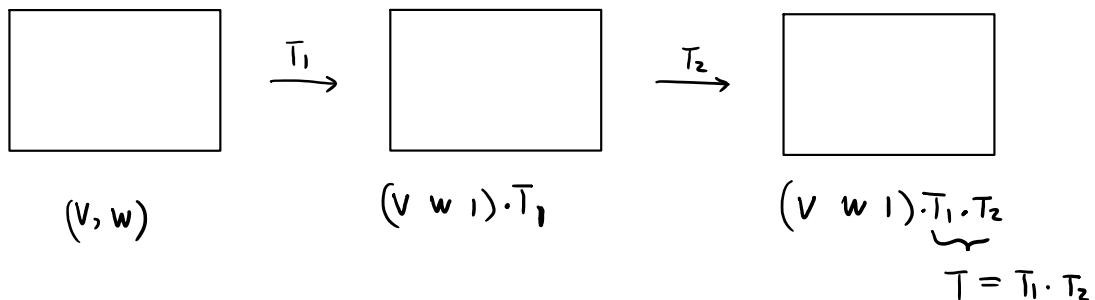
```
T =
```

```
1.0000 0.5000 0  
0 1.0000 0  
0 0 1.0000
```

```
>> tform = affinetform2d(T);  
>> J = imwarp(I, tform);  
>> figure; imshow(J)
```



## Sequence of transformations

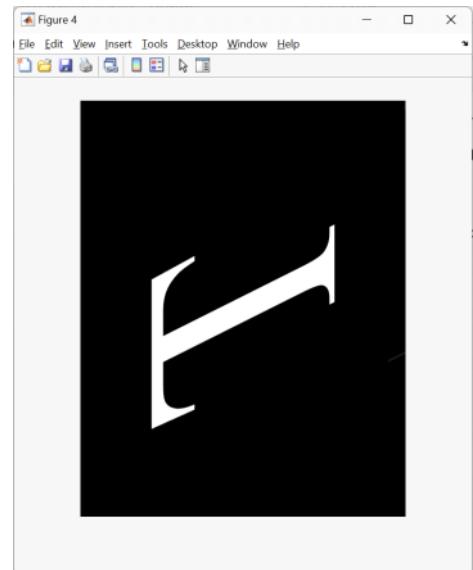


```
>> T1 = [cos(theta) sin(theta) 0; -sin(theta) cos(theta) 0; 0 0 1];
>> T2 = [1 0.5 0; 0 1 0; 0 0 1];
>> T = T1*T2
```

$T =$

```
0.0000 1.0000 0
-1.0000 -0.5000 0
0 0 1.0000
```

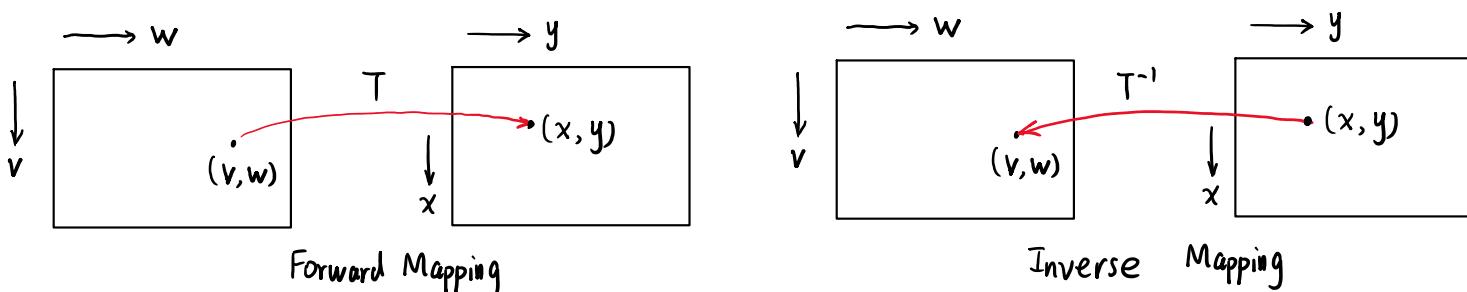
```
>> tform = affinetform2d(T);
J = imwarp(I, tform);
figure; imshow(J)
```



## Forward vs. Inverse Mapping

### Inverse mapping

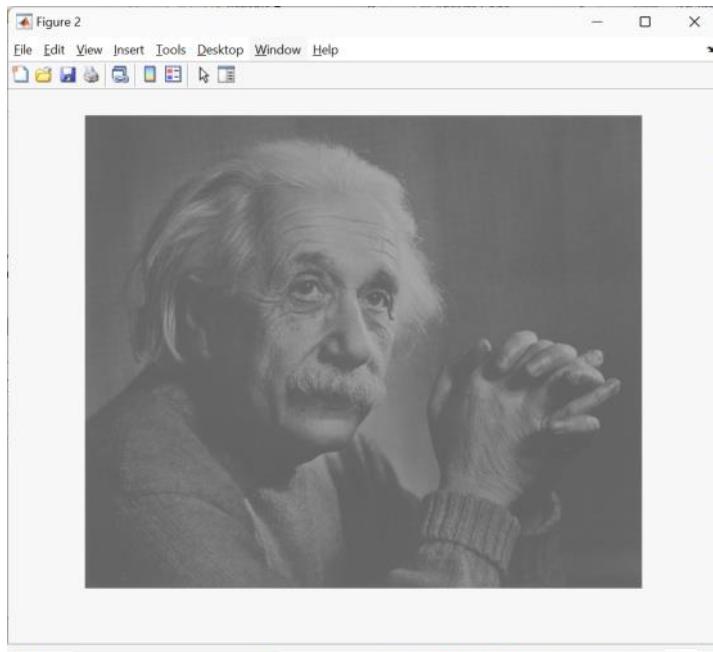
- Scanning the output pixel locations and, at each location,  $(x, y)$ , computes the corresponding location in the input image using  $(v, w) = T^{-1}(x, y)$
- It then **interpolates** (using one of the techniques discussed previously among the nearest input pixels to determine the intensity of the output pixel value).
- Inverse mappings are more efficient to implement than forward mappings and are used in numerous commercial implementations of spatial transformations



## Probabilistic Methods

[http://www.ece.uah.edu/~dwpan/course/ee604/code/ch2/fig2\\_41.m](http://www.ece.uah.edu/~dwpan/course/ee604/code/ch2/fig2_41.m)

```
% STD as a measure of intensity contrast
I = imread('Fig0241(a)(einstein low contrast).tif');
I_1d = reshape(I, 1, 679*800);
imhist(I);
std(double(I_1d))
I = imread('Fig0241(a)(einstein med contrast).tif');
I = imread('Fig0241(a)(einstein hig contrast).tif');
```



```
>> std(double(I_1d))
```

```
ans =
```

**14.2924**

```
>> I = imread('Fig0241(c)(einstein high contrast).tif');
>> I_1d = reshape(I, 1, 679*800);
imhist(I);
figure; imshow(I)
std(double(I_1d))
ans =
49.2428
```

