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## **Image and Video Processing Assignment - 6**

```
clc;
clear all;
close all;
```

#### **Affine Transformation**

Affine transformation are linear matrix operations on an image through which an image can be scaled, rotated or sheared. These transformations are linear and so does not affect the shapes in a image. It is usually represented as:

```
[u,v] = T([x,y])
(u,v) -> Transformed Coordinates
(x, y) -> Original Coordinates
```

### **Code for Affine Transformation**

```
for i=1:row
        for j=1:col
            transformed coords = round(T * [i; j; 1]);
            x_min = min(transformed_coords(1), x_min);
            x_max = max(transformed_coords(1), x_max);
            y_min = min(transformed_coords(2), y_min);
            y max = max(transformed coords(2), y max);
        end
    end
    out = zeros(x_max-x_min, y_max-y_min);
    % If the minimum value of any coordinate turns out to be zero we
need
    % to shift it to bring it to 1
   x_{shift} = (1 + abs(x_{min}))*(x_{min}<0);
   y_shift = (1 + abs(y_min))*(y_min<0);
    for i=1:row
        for j=1:col
            transformed_coords = round(T * [i; j; 1]);
            out(transformed_coords(1) + x_shift, transformed_coords(2)
 + y \text{ shift}) = imq(i, j);
        end
    end
    out = mat2gray(out);
end
```

### **Example: Scaling**

Scaling means to enlarge or compress an image. The transformation matrix for scaling is defined as:

$$T = \begin{pmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

a and b are scaling parameters in x and y direction. If a and b are greater than 1 image is enlarged otherwise the image is shrunk.

```
% Read an image
orig_img = imread('cameraman.tif');
img = double(orig_img);

% Create transformation matrix to enlarge
T = [4, 0, 0; 0, 2, 0; 0, 0, 1];
enlarged = affine_transformation(img, T);

% Create transformation matrix to shrink
T = [0.5, 0, 0; 0, 1, 0; 0, 0, 1];
```

```
shrinked = affine_transformation(img, T);

figure('Name', 'Affine Transformation: Scaling');
subplot(131)
imshow(orig_img);
title('Original Image');

subplot(132)
imshow(enlarged);
title('Enlarged image');

subplot(133)
imshow(shrinked);
title('Shrinked image');
```

Original Image



Enlarged image



Shrinked image



# **Example: Rotation**

To rotate an image through a particular angle we need to define the transformation matrix in the following manner:

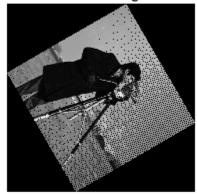
$$T = \begin{pmatrix} \cos(\theta) & \sin(\theta) & 0 \\ -\sin(\theta) & \cos(\theta) & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

% Create transformation matrix to rotate through angle theta

**Original Image** 



Rotated image



# **Example: Rotation and Scaling**

In this example we will show how can we use succesive transformations

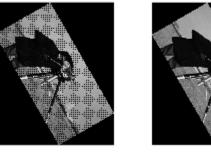
```
% Create transformation matrix to shrink
T1 = [0.5, 0, 0; 0, 1, 0; 0, 0, 1];
enlarged = affine_transformation(img, T1);
% Now, we will rotate the enlarged image
```

```
rot_and_scaled = affine_transformation(enlarged, T);
% We can show that this entire two step process can be represented by
 just
% one transformation matrix which is the product of the two matrices
T_new = T * T1;
transformed = affine_transformation(img, T_new);
figure('Name', 'Scaling and Rotation');
subplot(131)
imshow(orig_img);
title('Original Image');
subplot(132)
imshow(rot_and_scaled);
title('Scaled and Rotated image');
subplot(133)
imshow(transformed);
title('Scaled and Rotated with new transformtion');
```

Original Image



Scaled and Rotate signate and Rotated with new transformti



### **Conclusion**

Through this experiment we learnt about affine transformation. These operations are linear in nature and act on the coordinates of the pixels so they only affect the geometry of the image rather than the image

quality. Affine transformations are very simple to implement and have widespread use in resizing images, comparing rotated images, fixing orientations etc. We also noticed that any complex affine transformation can be broken down into simpler transformations. One more important point to notice is when we enlarge an image through affine transformation we need to use some interpolation technique to fill in the gaps between the pixels.

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