CSE 107 Lab 03: Image Resizing

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Abstract:

The purpose of this lab is to demonstrate what nearest neighbor and bilinear interpolation can do to an image when upsampling and downsampling an image. This lab will also help compare between the two and display images after upsampling and downsampling an image. RMSE values will also be recorded to help understand both interpolations and see errors between images.

Quantitative Results:



Figure 1: Downsampled to size (100, 175) using nearest neighbor interpolation.



Figure 2: Downsampled to size (100, 175) using bilinear interpolation.



Figure 3: Upsampled to size (500, 625) using nearest neighbor interpolation.



Figure 4: Upsampled to size (500, 625) using bilinear interpolation.

Quantitative Results:

	Nearest neighbor interpolation	Bilinear interpolation
Downsample then upsample	22.746414	16.835790
Upsample then downsample	0.000000	5.405959

Questions:

Visually compare the two downsampled images, one using nearest neighbor interpolation and one using bilinear interpolation. How are they different and why based on what you know about the two interpolations?

When zooming in, it seems like bilinear interpolation looks more clearer, smoother and better quality than the nearest neighbor interpolation. This is because bilinear interpolation considers the weighted average to get its interpolated value. Nearest neighbor considers 1 pixel while bilinear considers 4 pixels, making it so much smoother.

Visually compare the two down then upsampled images, one using nearest neighbor interpolation and one using bilinear interpolation. How are they different? Which one looks better to you? Does this agree with the RMSE values?

When comparing the two down then upsampled images, bilinear interpolation is much more clearer and less chunky than nearest neighbor, so bilinear looks better to me. This does agree with the RMSE values because bilinear RMSE value is 16.835790 while nearest neighbor is 22.746414. The lower the value, the better the smoother image and performance.

Visually compare the two up then downsampled images, one using nearest neighbor interpolation and one using bilinear interpolation. How are they different? Which one looks better to you? Does this agree with the RMSE values?

When comparing the two up then downsampled image, the nearest neighbor looks a tiny bit clearer than the bilinear. They almost look exactly the same but it seems like nearest neighbor looks much clearer. This does agree with the RMSE values because nearest neighbor has an RMSE value of 0 while bilinear has an RMSE value of 5.405959. Because of these values, this does agree with the images because nearest neighbor looks a little bit more clearer.

If your image resizing is implemented correctly, you should get an RMSE value of zero between the original image and the up then downsampled one using nearest neighbor interpolation. Why is this the case?

This is the case because we are considering one pixel and scaling that up. That one pixel value can be scaled up to a certain number so that we have that pixel value in the same cell it would normally be in but scaled up. Because we can reverse this easily, this can be scaled down using the same pixel values and same cells, causing the RMSE to be the value of zero.

What was the most difficult part of this assignment?

The most difficult part of this lab was trying to figure out the bilinear interpolation function. This is because I got confused with the indexing and some of the equations that were needed to calculate mybilinear.

Code:

MyImageFunctions.py

```
from PIL import Image, ImageOps
# Import numpy
import numpy as np
from numpy import asarray
import math
def myRMSE(I1, I2):
   M, N = np.shape(I1)
    rmse = 0
    #Helps calculate for RMSE
    for m in range (M):
       for n in range(N):
            rmse += (I1[m,n] - I2[m,n])**2
    rmse = np.sqrt(rmse/(M*N))
    return rmse
def myImageResize( inputPixels, M, N, interpolationName ):
```

```
#Gets shape of original
   ro, co = inputPixels.shape
    #New shape of array
   resized = np.empty((M, N), float)
   #Helps estimate index of input image
   for row in range(M+1):
       for col in range(N+1):
            mt = (((row-0.5)/M) * ro) + 0.5
            nt = (((col-0.5)/N) * co) + 0.5
            if(interpolationName == "nearest"):
                        #Helps calculate for nearest neighbor. Uses round
                        mt = round(mt)
                        nt = round(nt)
                        resized[row-1][col-1] = inputPixels[mt-1][nt-1]
            elif(interpolationName == "bilinear"):
                        #Gets 4 points m1,m2,n1,n2. This is the bilinear
part of it.
                        if(mt == int(mt)):
                            m1 = mt-1
                            m2 = mt-1
                        else:
                            if(mt < 1 ):</pre>
                                m1 = 1-1
                                m2 = 2-1
                            elif(mt>ro-1):
                                m1 = ro-2
                                m2 = ro-1
                            else:
```

```
m1 = math.floor(mt-1)
                                m2 = math.ceil(mt-1)
                        if(nt == int(nt)):
                            n1 = nt-1
                            n2 = nt-1
                        else:
                            if(nt < 1 ):
                                n1 = 1-1
                                n2 = 2-1
                            elif(nt>co-1):
                                n1 = co-2
                                n2 = co-1
                            else:
                                n1 = math.floor(nt-1)
                                n2 = math.ceil(nt-1)
                        #Gets pixels from original image
                        p1 = inputPixels[m1][n1]
                        p2 = inputPixels[m1][n2]
                        p3 = inputPixels[m2][n1]
                        p4 = inputPixels[m2][n2]
                        x5 = mt -1
                        y5 = nt -1
                       #Goes into mybilinear function to help calculate
using all points
                        pq =
mybilinear(m1,n1,p1,m1,n2,p2,m2,n1,p3,m2,n2,p4,x5,y5)
                        resized[row-1][col-1] = pq
            #Returns new array
    return resized
def mybilinear(x1,y1,p1,x2,y2,p2,x3,y3,p3,x4,y4,p4,x5,y5):
    #Equation gotten from HW to help calculate for bilinear interpolation
```

```
pri1 = (p3-p1)*((x5-x1)/(x3-x1)) + p1
pri2 = (p4-p2)*((x5-x2)/(x4-x2)) + p2
p5 = (pri2-pri1) * ((y5-y1)/(y2-y1)) + pri1
return p5
```

test myimresize.py

```
# Import pillow
from PIL import Image, ImageOps
# Import numpy
import numpy as np
from numpy import asarray
# Read the image from file.
orig im = Image.open('Lab 03 image.tif')
# Show the original image.
orig_im.show()
# Create numpy matrix to access the pixel values.
# NOTE THAT WE WE ARE CREATING A FLOAT32 ARRAY SINCE WE WILL BE DOING
# FLOATING POINT OPERATIONS IN THIS LAB.
orig_im_pixels = asarray(orig_im, dtype=np.float32)
# Import myImageResize from MyImageFunctions
from MyImageFunctions import myImageResize
# Experiment 1: Downsample then upsample using nearest neighbor
interpolation.
# Create a downsampled numpy matrix using nearest neighbor interpolation.
downsampled im NN pixels = myImageResize(orig im pixels, 100, 175,
'nearest')
# Create an image from numpy matrix downsampled im NN pixels.
```

```
downsampled im NN =
Image.fromarray(np.uint8(downsampled im NN pixels.round()))
# Show the image.
downsampled im NN.show()
# Save the image.
downsampled im NN.save('downsampled NN.tif');
# Upsample the numpy matrix to the original size using nearest neighbor
interpolation.
down up sampled im NN pixels = myImageResize(downsampled im NN pixels,
400, 400, 'nearest')
# Create an image from numpy matrix down up sampled im NN pixels.
down up sampled im NN =
Image.fromarray(np.uint8(down up sampled im NN pixels.round()))
# Show the image.
down up sampled im NN.show()
# Import myRMSE from MyImageFunctions
from MyImageFunctions import myRMSE
# Compute RMSE between original numpy matrix and down then upsampled
nearest neighbor version.
down up NN RMSE = myRMSE( orig im pixels, down up sampled im NN pixels)
print('\nDownsample/upsample with myimresize using nearest neighbor
interpolation = %f' % down up NN RMSE)
# Experiment 2: Downsample then upsample using bilinear interpolation.
# Create a downsampled numpy matrix using bilinear interpolation.
downsampled im bilinear pixels = myImageResize(orig im pixels, 100, 175,
'bilinear')
 Create an image from numpy matrix downsampled im bilinear pixels.
```

```
downsampled im bilinear =
Image.fromarray(np.uint8(downsampled im bilinear pixels.round()))
# Show the image.
downsampled im bilinear.show()
# Save the image.
downsampled im bilinear.save('downsampled bilinear.tif');
# Upsample the numpy matrix to the original size using bilinear
interpolation.
down up sampled im bilinear pixels =
myImageResize(downsampled im bilinear pixels, 400, 400, 'bilinear')
# Create an image from numpy matrix down up sampled im bilinear pixels.
down up sampled im bilinear =
Image.fromarray(np.uint8(down up sampled im bilinear pixels.round()))
# Show the image.
down up sampled im bilinear.show()
# Compute RMSE between original numpy matrix and down then upsampled
bilinear version.
down up bilinear RMSE = myRMSE( orig im pixels,
down up sampled im bilinear pixels)
print('Downsample/upsample with myimresize using bilinear interpolation =
%f' % down up bilinear RMSE)
# Experiment 3: Upsample then downsample using nearest neighbor
interpolation.
# Create an upsampled numpy matrix using nearest neighbor interpolation.
upsampled im NN pixels = myImageResize(orig im pixels, 500, 625,
'nearest')
```

```
Create an image from numpy matrix upsampled im NN pixels.
upsampled im NN =
Image.fromarray(np.uint8(upsampled im NN pixels.round()))
# Show the image.
upsampled im NN.<mark>show()</mark>
# Save the image.
upsampled im NN.save('upsampled NN.tif');
# Downsample the numpy matrix to the original size using nearest neighbor
interpolation.
up down sampled im NN pixels = myImageResize(upsampled im NN pixels, 400,
400, 'nearest')
# Create an image from numpy matrix up down sampled im NN pixels.
up down sampled im NN =
Image.fromarray(np.uint8(up down sampled im NN pixels.round()))
# Show the image.
up down sampled im NN.show()
# Compute RMSE between original numpy matrix and down then upsampled
nearest neighbor version.
up down NN RMSE = myRMSE( orig im pixels, up down sampled im NN pixels)
print('\nUpsample/downsample with myimresize using nearest neighbor
interpolation = %f' % up down NN RMSE)
# Experiment 3: Upsample then downsample using bilinear interpolation.
# Create an upsampled numpy matrix using bilinear interpolation.
upsampled im bilinear pixels = myImageResize(orig im pixels, 500, 625,
'bilinear')
# Create an image from numpy matrix upsampled im bilinear pixels.
```

```
upsampled im bilinear =
Image.fromarray(np.uint8(upsampled_im_bilinear_pixels.round()))
# Show the image.
upsampled im bilinear.<mark>show()</mark>
# Save the image.
upsampled im bilinear.save('upsampled bilinear.tif');
# Downsample the numpy matrix to the original size using bilinear
interpolation.
up down sampled im bilinear pixels =
myImageResize(upsampled im bilinear pixels, 400, 400, 'bilinear')
# Create an image from numpy matrix up down sampled im bilinear pixels.
up down sampled im bilinear =
Image.fromarray(np.uint8(up down sampled im bilinear pixels.round()))
# Show the image.
up down sampled im bilinear.show()
# Compute RMSE between original numpy matrix and up then downsampled
bilinear version.
up down bilinear RMSE = myRMSE( orig im pixels,
up down sampled im bilinear pixels)
print('Upsample/downsample with myimresize using bilinear interpolation =
%f' % up down bilinear RMSE)
```