CSE 107: Lab 03: Image Resizing Angelo Fatali LAB: R 7:30-10:20pm Liang, Haolin October 24, 2022

Abstract:

The point of this lab is to understand the difference between bilinear interpolation and nearest neighbor when resizing an image. To do this we can downsample and upsample using both methods and see the difference. We then can use the root mean square error (RMSE) value to determine which is better mathematically.

Qualitative Results:



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



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



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



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

Quantitative Results:

	Nearest neighbor interpolation	Bilinear interpolation
Downsample then upsample	22.746414	16.882088
Upsample then downsample	0.000000	5.449098

Table 1: RMSE values between the original image and then the down/upsampled and up/downsampled versions for both nearest neighbor and bilinear interpolation.

Questions:

- 1. 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?
- The downsampled image that used nearest neighbor interpolation is much more blocky is pixelated. The bilinear, while still a little pixelated, is much better when compared to the nearest neighbor. This is because nearest neighbor simply gets the pixel value based off the closest distance the other pixels are around it. While bilinear will get the average of the other pixels and get a much better estimate of pixel value.
- 2. 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?
- The down then upsampled image using nearest neighbor resulted in a very pixelated but crisp image. The bilinear result was less pixelated but seemed blurry, almost like it was out of focus. I personally like the bilinear result more since we get more of an idea of what each pixel is, rather than super pixelated. The RMSE value agrees with me as 22 (NN result) is greater than 16 (Bilinear result).
- 3. 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?
- The down then upsampled image using nearest neighbor resulted in an identical image to the original. The bilinear result was pretty good but seemed a little blurry. I personally like the nearest neighbor result more since we got an identical image to the original. The RMSE value agrees with me as 0 (NN result) is less than 5 (Bilinear result).
- 4. 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?
- Since nearest neighbor, when up scaling, is getting the closest pixel value to the other pixels, we get a very pixelated up scale image. However, when you take that same up scaled image and then down sample it to the original size, we actually don't get any pixel values to be different since the closest pixels are simply the original pixel values.
- 5. What was the most difficult part of this assignment?
- I kept having a ton of index out of range errors and had to figure out where I needed to add a -1 and where I didn't. Other than that it was fairly easy.

test_myresize.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.show()
# 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<sup>T</sup>)
# 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.show()
# 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)
```

MyImageFunctions.py

```
import numpy as np
import math
# Write the functionmyimresize() that takes as input
# A numpy matrix representing a grayscale image.
# The number of rows in the resized image.
# The number of columns in the resized image.
# A string with values 'nearest' or 'bilinear'.
# and outputs the resized numpy matrix image.
# Based off the TA's sample file
def myImageResize(inImage pixels, M, N, interpolation method ):
    Minput, Ninput = inImage pixels.shape
    # create a new output matrix
    out = np.zeros(shape=(M, N))
    for m in range(M+1): # int
        for n in range(N+1): # int
            # loop all the pixels from the out
            # 1. we estimate the index of input image -> estimate row and col
                                                      # float number,
            m_{inter} = (((m-0.5)/M) * Minput) + 0.5
            n_{inter} = (((n-0.5)/N) * Ninput) + 0.5
            if interpolation_method == 'nearest':
                # use nn
                # round()
                m_inter = round(m_inter)
                n_inter = round(n_inter)
                # print("m_inter: ", m_inter)
# print("n_inter: ", n_inter)
                out[m-1, n-1] = inImage_pixels[m_inter-1, n_inter-1]
            elif interpolation method == 'bilinear':
                # use the bilinear
                # find the 4 points
                # use the if else from the hw
                # m_inter -> m1, m2 which are the nearest row index
                if(m_inter == int(m_inter)):
                    m1 = m inter-1
                    m2 = m_inter-1
                else:
                    if m_inter < 1:</pre>
                        m1, m2 = 1, 2
                    elif m_inter > Minput-1:
                        m1, m2 = Minput-2, Minput-1
                        m1 = math.floor(m inter-1)
                        m2 = math.ceil(m_inter-1)
                # n_inter -> n1, n2 which are the nearest col index
                if(n inter == int(n inter)):
                    n1 = n_{inter-1}
                    n2 = n inter-1
                else:
```

```
if n_inter < 1:</pre>
                        \bar{n}1, \bar{n}2 = 1, 2
                     elif n_inter > Ninput-1:
                        n1, n2 = Ninput-2, Ninput-1
                        n1 = math.floor(n_inter-1)
                        n2 = math.ceil(n_inter-1)
                p1 = inImage_pixels[m1, n1]
                p2 = inImage_pixels[m1, n2]
                p3 = inImage_pixels[m2, n1]
                p4 = inImage_pixels[m2, n2]
                p5 = 0
                p5 =
mybilinear(m1,n1,p1,m1,n2,p2,m2,n1,p3,m2,n2,p4,m_inter-1,n_inter-1,p5)
                out [m-1, n-1] = p5
    return out
def mybilinear(x1,y1,p1,x2,y2,p2,x3,y3,p3,x4,y4,p4,x5,y5,p5):
    # use eq from hw
    # apply eq on p1 and p3 -> t1
    # apply eq on p2 and p4 -> t2
                  t1 and t2 -> p5
    p5\_prime = (p3-p1)*((x5-x1)/(x3-x1))+p1
    p5_{doubleprime} = (p4-p2)*((x5-x2)/(x4-x2))+p2
    p5 = (p5\_doubleprime-p5\_prime)*((y5-y1)/(y2-y1))+p5\_prime
    return p5
def myRMSE(first_im_pixels, second_im_pixels):
    total = 0
    M,N = first_im_pixels.shape
    for m in range(0,M):
        for n in range(0,N):
            total += (first_im_pixels[m,n] - second_im_pixels[m,n])**2
    # now you have sum
    total = total / (M * N)
    return math.sqrt(total)
```