Vectorization and Level Operations Lab

All of the programming assignments are to be done in Python using additional libraries specified in the assignments. There are many libraries available, some of which we will be using, and you are welcome to use them with one exception: if the library or a function within it performs the specific function you are asked to code, you may not use that other than perhaps as a reference to compare against. All of the code you submit must be your own. You are welcome to turn in a completed jupyter notebook.

Preliminaries

Before we begin the lab, let's review how to import packages in Python and introduce the ones you will be using.

Python has multiple ways to import packages to use in the notebook. The keywords you can use are **import**, **from**, and **as**. Run the following cell by clicking "Run" or using shift + enter.

```
import math
In [35]:
          print(math.pi)
          from math import pi
          print(pi)
          from math import pi as taco
          print(taco)
          from math import * #Import everything from the package
          print(e)
         3.141592653589793
         3.141592653589793
         3.141592653589793
         2.718281828459045
```

The two main packages we will use in this and future labs are called Numpy and Matplotlib. These are the "Numerical Python" and "Matrix Plotting Library" packages respectively. These two libraries provide tons of functionality related to matrix processing and image input and output.

The conventional import method for these two packages looks like this:

```
In [36]:
         import numpy as np
          import matplotlib.pyplot as plt
```

Now we can reference these two packages simply as "np" and "plt". If either of these packages aren't importing, make sure to quit your jupyter notebook and run pip install numpy or pip install matplotlib in your terminal.

> Let's look at Numpy first. Numpy allows you store matrices natively and do matrix operations in a single step.

Let's look at a simple matrix.

```
In [37]: import numpy as np
          a = np.matrix([[1, 2],[3,4]])
          print(a)
         [[1 2]
          [3 4]]
         print(a[0,1]) #Row, Column zero-based indexing
In [38]:
         print(a[:,1]) #Grab all rows and the second column
In [39]:
          print()
          print(a[0,:]) #Grab everything in the first row
         [[2]
          [4]]
         [[1 2]]
In [40]:
         print(np.multiply(a,a)) #Element-wise multiply
          print()
          print(np.matmul(a,a)) #True matrix multiply
          [[ 1 4]
          [ 9 16]]
         [[ 7 10]
          [15 22]]
```

There are plenty of Numpy operations that we are not covering, but we will see more of these as we go throughout the class. For more details, go to https://docs.scipy.org/doc/numpy/reference/. Also, make sure to check out the Useful Numpy Commands page on Canvas.

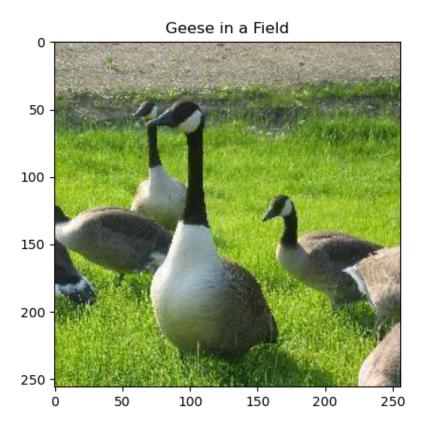
Now let's look at how we can import and display images with Matplotlib.

Importing is easy. All we need to do is use the imread() function.

```
In [41]: import matplotlib.pyplot as plt
         geese = plt.imread('geese.jpg')
         print(geese.shape)
         (256, 256, 3)
```

Grabbing the .shape property shows us that we have loaded an images that is 256 rows by 256 columns and has 3 color channels. Let's look at what that picture looks like. We use the imshow() and show() functions to do so.

```
plt.imshow(geese)
In [43]:
          plt.show()
```



Now we can see the picture, but why do we need to use two seperate commands to display? Because you may choose to modify the display properties before you show it. For example, maybe you want to add a title.

```
In [45]:
         plt.imshow(geese) # This Loads the image into the buffer
          plt.title("Geese in a Field") # This modifies the plotting properties
         plt.show() # This outputs the final displayed image
```

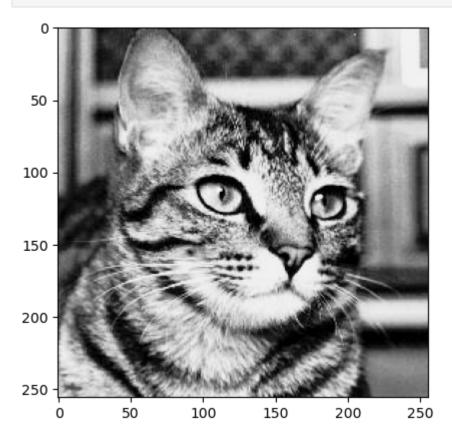
There is also a plt.savefig() function you can use to save images to your harddrive.

And with that, you now have the basic tools you need to work with images. For this lab, we will only use grayscale operations to simplify our practice of level operations. The test cases will be provided for each part of the lab, but you can also use the following code to load and plot grayscale images:

```
import matplotlib.pyplot as plt
In [1]:
        import numpy as np
        %matplotlib inline
         cat = plt.imread('cat.jpg')
         cat = np.matrix(cat,dtype=np.int32)
```

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```
plt.imshow(cat,cmap="Greys_r", vmin=0)
plt.show()
```



Programming Exercises

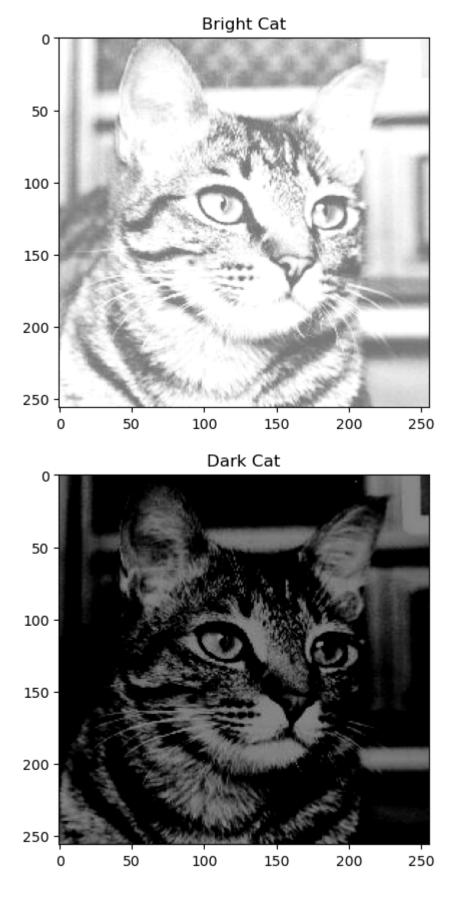
The purpose of this lab is to learn how to use the vectorization features of Numpy. You are not allowed to use a for or while loop for any part of this lab. Any use of a for or while loop will cost points.

Implement each of the following level operations as Python functions.

Function 1: Brightness Adjust

Takes in a grayscale image and returns the brightened version of that image according to a passed in parameter. Use a max image value of 255.

```
def brightAdjust(image, c):
In [3]:
            return np.clip(image + c,0,255)
In [4]: #Test Cases
        bright cat = brightAdjust(cat, 100)
        plt.imshow(bright_cat, cmap="Greys_r",vmin=0, vmax=255);plt.title("Bright Cat");plt.sh
        dark_cat = brightAdjust(cat, -100)
        plt.imshow(dark_cat, cmap="Greys_r",vmin=0, vmax=255);plt.title("Dark Cat");plt.show()
```



Function 2: Contrast Adjustment

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def contrastAdjust(image,c):

In [5]:

Takes in a grayscale image and returns the contrasted version of that image according to a passed in parameter. Use a max image value of 255.

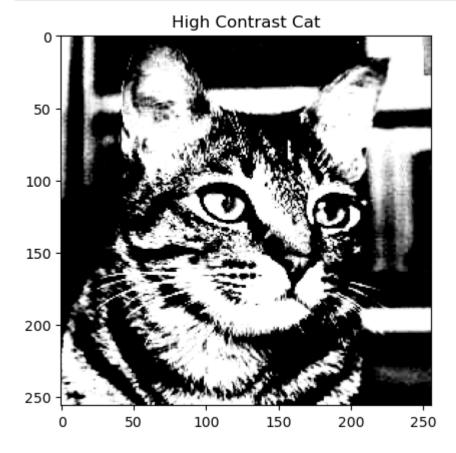
Also, rather than a straight linear operation, we will use a mapping similar to what Photoshop does. In particular, the contrast will be in the range [-100,100] where 0 denotes no change, -100 denotes complete loss of contrast, and 100 denotes maximum enhancement (8x multiplier). If c is the contrast parameter, then the level operation applied is:

$$s = \left(rac{c+100}{100}
ight)^4 (r-128) + 128$$

Make sure you work in floating point, not integers. Integer division would not be very acurate.

```
return np.clip(((((c + 100) / 100.0) ** 4) * (image - 128.0)) + 128.0,0, 255)
                                                                     #Test Cases
In [6]:
                                                                        high contrast cat = contrastAdjust(cat, 50)
                                                                         plt.imshow(high_contrast_cat, cmap="Greys_r",vmin=0, vmax=255);plt.title("High Contrast_cat, 
                                                                         low contrast cat = contrastAdjust(cat, -50)
```

plt.imshow(low_contrast_cat, cmap="Greys_r",vmin=0, vmax=255);plt.title("Low Contrast



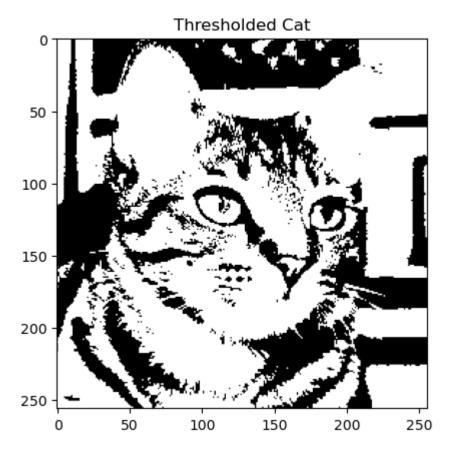
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Function 3: Thresholding

Takes in a grayscale image and returns the thresholded version of the image according to a passed in parameter. Every pixel that is higher than or equal to the parameter is 255, everything below is zero. (Hint: Use np.where)

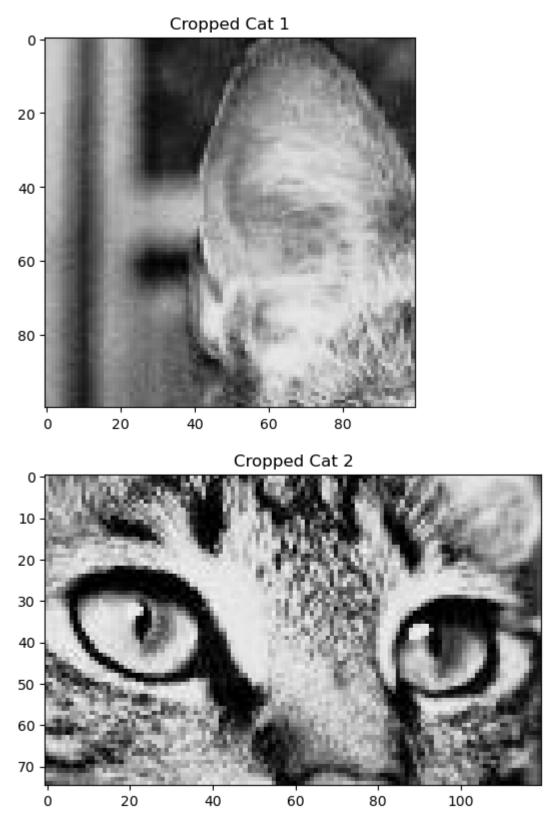
```
In [7]:
        def thresholder(image, c):
             return np.where(image >= c, 255, 0)
In [8]:
        #Test Cases
        thresholded cat = thresholder(cat, 80)
        plt.imshow(thresholded_cat, cmap="Greys_r",vmin=0, vmax=255);plt.title("Thresholded Cat")
```



Function 4: Cropping

Takes in a grayscale image, an x and y of a topleft pixel, a width, and a height and returns a cropped version of that image according to those parameters. Recall to use row-major indexing with numpy arrays and matrices, i.e. they are indexed image[row, column], which corresponds to image[y, x]. Vectors use [x,y,z], however.

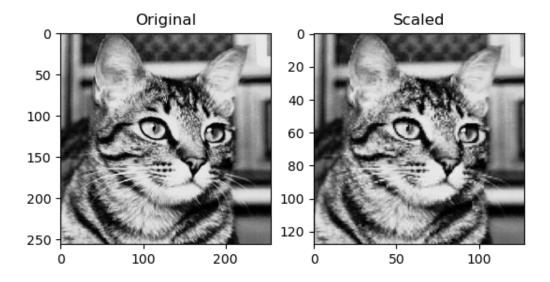
```
In [9]:
                                                    def cropper(image, width, height, x=0, y=0):
                                                                          return image[y:y+height, x:x+width]
In [10]:
                                                   #Test Cases
                                                    #This should show just the ear of the cat
                                                    cropped_cat1 = cropper(cat, 100, 100)
                                                    plt.imshow(cropped_cat1, cmap="Greys_r",vmin=0, vmax=255);plt.title("Cropped Cat 1");;
                                                    #This should show just the eyes of the cat
                                                     cropped_cat2 = cropper(cat, 120, 75, 90, 80)
                                                     plt.imshow(cropped_cat2, cmap="Greys_r", vmin=0, vmax=255); plt.title("Cropped Cat 2"); plt.imshow(cropped_cat2, cmap="Greys_r", vmin=0, vmax=255); plt.title("Cropped Cat 2"); plt.imshow(cropped_cat2, cmap="Greys_r", vmin=0, vmax=255); plt.title("Cropped_cat2"); plt.imshow(cropped_cat2, cmap="Greys_r", vmin=0, vmax=255); plt.title("Cropped_cat2"); plt.title("Croppe
```



Function 5: Scaling

Takes in a grayscale image and returns the same image with a resolution that is half the width and half the height of the original. (Hint: Think about what pixels you will want to grab to make that smaller image)

```
def scaler(image):
In [97]:
             return image[::2, ::2]
         #This line makes the image easier to see.
In [98]:
         %matplotlib notebook
          scaled_cat = scaler(cat)
         fig = plt.figure()
         ax1 = fig.add_subplot(1,2,1)
         ax1.imshow(cat, cmap="Greys_r",vmin=0, vmax=255); ax1.set_title("Original")
         ax2 = fig.add_subplot(1,2,2)
         ax2.imshow(scaled_cat, cmap="Greys_r",vmin=0, vmax=255); ax2.set_title("Scaled")
         plt.show()
```



Run the following line of code once you are done to return back to normal plotting functions.

```
%matplotlib inline
In [ ]:
```