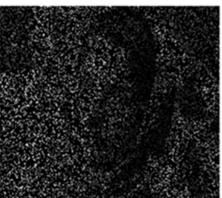
Image Processing

It is a technique used for the analysis of an image in order to manipulate it and extract information from it.



Corrupted image, 6.4119dB



Original image



Restored image, 26.8036dB



What is the purpose of image processing?

- Visualization Observe the objects that are not visible.
- Image Sharpening and Restoration To create a better image.
- Image Retrieval Seek for the image of interest.
- Measurement of Pattern Measures various objects in an image.
- Image Recognition Distinguish the objects in an image.

Python can do this all!

Is there anything that Python can't do. You have to be kidding me!

OpenCV is a **Open-source computer vision** library available for multiple platforms including C++. But first let's take a sneak peek at code of C++ I used for simplistic image processing task.

```
void searchForMovement(Mat thresholdImage, Mat &cameraFeed){
     bool objectDetected=false:
     Mat temp;
     thresholdImage.copyTo(temp);
     vector< vector<Point> > contours;
     vector<Vec4i> hierarchy;
     findContours(temp,contours,hierarchy,CV_RETR_EXTERNAL,CV_CHAIN_APPROX_SIMPLE );// retrieves external contours
     if(contours.size()>0)objectDetected=true;
     else objectDetected = false;
     if(objectDetected){
           vector< vector<Point> > largestContourVec;
          largestContourVec.push_back(contours.at(contours.size()-1));
           //make a bounding rectangle around the largest contour then find its centroid //this will be the object's final estimated position.
          objectBoundingRectangle = boundingRect(largestContourVec.at(0));
          int xpos = objectBoundingRectangle.x+objectBoundingRectangle.width/2;
int ypos = objectBoundingRectangle.y+objectBoundingRectangle.height/2;
          theObject[0] = xpos , theObject[1] = ypos;
     int x = theObject[0];
     int y = theObject[1];
     circle(cameraFeed,Point(x,y),20,Scalar(0,255,0),2);
line(cameraFeed,Point(x,y),Point(x,y-25),Scalar(0,255,0),2);
line(cameraFeed,Point(x,y),Point(x,y+25),Scalar(0,255,0),2);
     line(cameraFeed,Point(x,y),Point(x-25,y),Scalar(0,255,0),2);
line(cameraFeed,Point(x,y),Point(x+25,y),Scalar(0,255,0),2);
putText(cameraFeed,"Tracking object at (" + intToString(x)+","+intToString(y)+")",Point(x,y),1,1,Scalar(255,0,0),2);
```

Too many words, the hell with this shit! It literally took me two months to actually learn to code this, with clear concepts of course. Plus it's tedious to set up the libraries and do the required configurations before you can write the actual code. If you do really want to check that out though, I have maintained a journal containing core code concepts for C++ to use with OpenCV.

Moving forward with Python

Yes, people may claim why Python when much faster language exists that can process images way faster. I ask them, even if it is faster to debug them or write the code and not get stuck with the grammar of it?

Disclaimer: If you are hating C++ right at this instant, Python is built on C++. So



C++ is going nowhere!

Installing OpenCV on Python

OpenCV doesn't have an official pre-compiled package solution available on either Windows, Mac, or on Linux.



But don't you worry, this site: https://www.lfd.uci.edu/~gohlke/pythonlibs/#opencv) has the solution. Some people are heavenly and they will go straight to <code>Jana'ah</code>. These people have compiled the package for Windows and provided the complete solution. The download will be a <code>.whl</code> file. Choose appropriate version according to your Python version and your processor (32-bit or 64-bit).

Open CMD and go to directory where the downloaded file is present and install in this way:

• pip install "opencv python-3.4.3-cp36-cp36m-win amd64.whl"

And now let's code

With few functions always in mind:

- imread(directory)
- imshow(image)
- cvtColor(image, conversion_from_and_to)
- imwrite(directory, image)
- VideoCapture(address)

```
In [1]: import cv2
import numpy as np
import matplotlib.pyplot as plt
% matplotlib inline
```

```
In [69]: # Read the image
   im = cv2.imread('samples/1 (1).png')
   # Display the original image
   cv2.imshow('original',im)
   cv2.waitKey(0)
```

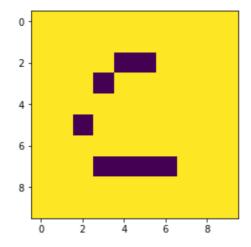
Out[69]: -1

Processing the original image

original -> BGR to gray -> resizing -> display

```
In [67]: # Let's change the RGB image (or BGR as said in OpenCV) to Grayscale
im = cv2.cvtColor(im, cv2.CoLOR_BGR2GRAY)
# Now if you want to resize the image to 10x10
im = cv2.resize(im, (10,10))
# Display the processed image
#cv2.imshow('gray',im)
#cv2.waitKey(0)
plt.imshow(im)
```

Out[67]: <matplotlib.image.AxesImage at 0x2313b270860>

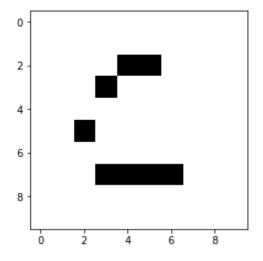


Print the normalized pixel values

Normalization can be done by dividing the pixel values by the higest number possible which in the case of images in 255. **Notice how matrix values printed below is exact representation of the image**

- 0 -> Black
- 1 -> White

Compare the image with the pixel values



```
[[1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 0. 0. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 0. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
```

A video is just a collection of images

When a set of 60 images are displayed in a second, a video is born. A 10 second video basically contains 60x10 = 600 images. In OpenCV you can process the video in a very similar fashion like that of still images.

```
In [70]: # Read the video or open the webcam.
         cap = cv2.VideoCapture(0)
         # Read the first frame, then the second, then the third, and so on.
         # You will require a loop that can read new frame in every iteration and display it
         while(cap.isOpened()):
             # Read the frame
             _, frame = cap.read()
             cv2.imshow('original',frame)
             gray = cv2.cvtColor(frame, cv2.COLOR BGR2GRAY)
             cv2.imshow('gray', gray)
             hsv = cv2.cvtColor(frame, cv2.COLOR_BGR2HSV)
             cv2.imshow('hsv', hsv)
             # In order to end the process or video forcefully, press 'q'
             if cv2.waitKey(1) & 0xff == ord('q'):
         # Release the memory. If you don't do this, it will eat up your memory and program will cr
         ash ultimately
         cap.release()
         # Close all the opened windows because either the video has been finished or you forcefull
         y ended the video
         cv2.destroyAllWindows()
```

Notice the while block: The process taking place in the body is exactly the same as that of the still images.

Concept Alert!

There is a **don't care variable** in the code above represented by an underscore. Sometimes you don't care about the variable name because they are not used anywhere in the code so you just let Python take care of it and name it anything it want.

Another thing you might have seen is this:

```
• a, b = somefunc()
```

This simply means that if a function returns two values, you can say that first value goes into 'a' and the second goes into 'b'.

```
In [71]: def twonum():
    return 2,3

_, b = twonum()
print(b)
```

Flip the frames

Sometimes during the training of the AI system, the data available is not enough. One way to increase the data is to make small changes to the existing data. This is known as **Data Augmentation**. In case of the images, you can flip them and add these generated images to the dataset.

```
In [8]:
    cap = cv2.VideoCapture(0)
    while(cap.isOpened()):
        ret, frame = cap.read()
    if ret==True:
        frame = cv2.flip(frame, 0)
        # Flip the frame
        cv2.imshow('frame',frame)

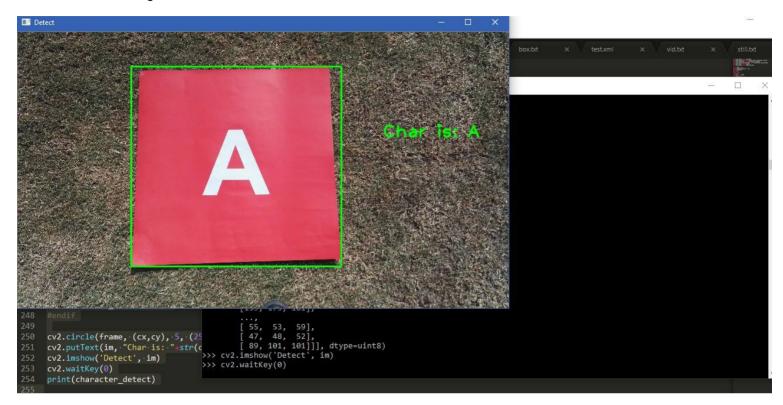
        # In order to end the process or video forcefully, press 'q'
        if cv2.waitKey(1) & 0xff == ord('q'):
            break
    else:
        break

cap.release()
cv2.destroyAllWindows()
```

Drawing on the images

Very often in **Object Localization** the computer draws a boundary around the image showing you where the object is.

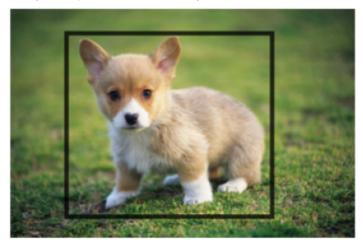
For Example: In the image below, Al system has not only accurately predicted the character but also shows where the character is in the image.

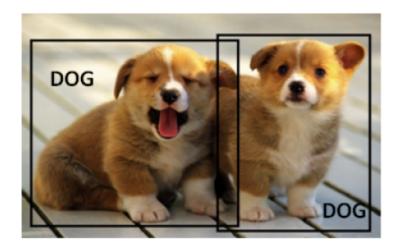


Further examples include

Complex image segmentation and localization, many of which is done by a simple function call in Python.









With that motivation

Let's learn how the computer draws on the image, shall we?

```
In [73]: # A 2D array of zeros as we saw means the image is black
    # The following code is to create a black image of size 100x100
    img = np.zeros((500,500,3), np.uint8)

# cv2.line(image, (x_init, y_init), (h, w), (b, g, r), line_width)
    img = cv2.line(img, (0,80), (500,80), (255,0,0), 2)
    img = cv2.line(img, (0,60), (260,500), (0,255,0), 2)
    img = cv2.line(img, (240,500), (500,60), (0,0,255), 2)

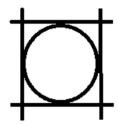
# cv2.rectangle(img, (x_2, y_1), (x_1, y_2), (b, g, r), line_width)
    img = cv2.rectangle(img, (350,200), (150,300), (255,255,255), 2)

# cv2.circle(img, (x_cen, y_cen), radius, (b, g, r), fill)
    img = cv2.circle(img, (250,250), 20, (255,0,255), -1)

# cv2.putText(img, text, (x, y), font, size, (b, g, r), boldness, 0)
    cv2.imshow('img', img)
    cv2.waitKey(0)
```

Out[73]: -1

Try drawing these







Masking

In order to extract a specific color from the image, a concept known as masking is used. This concept can be understood intuitively. Mask is simply the array of pixels which you want to consider only and all the other pixels are turned to zero.



Mask is a Binary Image!

But now you want to retain the color of the object as well. Therefore, we can apply AND operation on the mask and the original image. The pixels in the mask which were zero will turn the result of AND operation to become zero. Similarly the pixels in the mask which were ones will turn the result of AND operation to be unaffected.

It is better explained by the example below.

```
In [75]: img = cv2.imread('samples/C (60).jpg')
         im = cv2.resize(img, (500,500))
         cv2.imshow('image', im)
         # Always convert the image to HSV for masking
         hsv = cv2.cvtColor(im, cv2.COLOR_BGR2HSV)
         # Define the color you want for mask
         lower_red = np.array([100,130,150])
         upper red = np.array([255,255,255])
         # Apply the mask
         mask = cv2.inRange(hsv, lower_red, upper_red)
         cv2.imshow('mask', mask)
         # Do the AND operation on mask and original image
         res = cv2.bitwise_and(im, im, mask=mask)
         cv2.imshow('res', res)
         cv2.waitKey(0)
         cv2.destroyAllWindows()
```

But how to find out the masking values?

Trust me, you have to play with it and test it repeatedly on images containing similar color until you find that sweet spot.

Wasn't anything not impossible with Python?

So can we not resolve this trial and error thing to find the sweet spot and make an algorithm that makes it easy to do so. **Let's find out!**

Trackbars for images

```
In [61]: img = cv2.imread('samples/1 (3).jpg')
          im = cv2.resize(img, (500,500))
          hsv = cv2.cvtColor(im, cv2.COLOR BGR2HSV)
          # initialize by making a blank result image
          res = np.zeros((500,500,3), np.uint8)
          # creating a window on which image will be displayed
          cv2.imshow('image', im)
          def nothing(x):
              pass
          #initializing the values to be set using trackbars to be zero
          l_h, l_s, l_v = 0, 0, 0
          higher = 255
          # creating trackbars
          cv2.createTrackbar('h', 'image', l_h, higher, nothing)
          cv2.createTrackbar('s', 'image', 1_s, higher, nothing)
cv2.createTrackbar('v', 'image', 1_v, higher, nothing)
          # with every new value from trackbar, image is needed to be refreshed and redisplayed
          while(True):
              cv2.imshow('res', res)
              # get the trackbar positions and set them to L h, L s, L v respectively
              1_h = cv2.getTrackbarPos('h', 'image')
              l_s = cv2.getTrackbarPos('s', 'image')
              1_v = cv2.getTrackbarPos('v', 'image')
              lower_red = np.array([l_h,l_s,l_v])
              upper_red = np.array([255,255,255])
              mask = cv2.inRange(hsv, lower_red, upper_red)
              res = cv2.bitwise and(im, im, mask=mask)
              if cv2.waitKey(1) & 0xff == ord('q'):
                  print("hsv -> ",l_h,l_s,l_v)
                  break
          # save the image
          cv2.imwrite('samples/samples.jpg', res)
          cv2.destroyAllWindows()
```

hsv -> 39 119 135

Can we make trackbars for video

Sure we can, just a little more code and more copy paste from above!

```
In [76]: cap = cv2.VideoCapture(0)
          def nothing(x):
              pass
          #initializing the values to be set using trackbars to be zero
          l_h, l_s, l_v = 0,0,0
          higher = 255
          # creating trackbars
          cv2.namedWindow('image')
          cv2.createTrackbar('h', 'image', l_h, higher, nothing)
cv2.createTrackbar('s', 'image', l_s, higher, nothing)
          cv2.createTrackbar('v', 'image', l v, higher, nothing)
          while(cap.isOpened()):
              _,frame = cap.read()
              hsv = cv2.cvtColor(frame, cv2.COLOR_BGR2HSV)
              1_h = cv2.getTrackbarPos('h', 'image')
              l_s = cv2.getTrackbarPos('s', 'image')
              1_v = cv2.getTrackbarPos('v', 'image')
              lower_red = np.array([1_h,1_s,1_v])
              upper_red = np.array([255,255,255])
              mask = cv2.inRange(hsv, lower_red, upper_red)
              res = cv2.bitwise and(frame, frame, mask=mask)
              cv2.imshow('image', frame)
              cv2.imshow('res', res)
              if cv2.waitKey(1) & 0xff == ord('q'):
                   print("hsv -> ",l_h,l_s,l_v)
                   break
          cap.release()
          cv2.destroyAllWindows()
```

hsv -> 54 134 93

Image Shape

OpenCV gives you a handy facility to know the dimensions of your image.

Fin

