

# COMP 204

## Introduction to image analysis with scikit-image (part two)

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# Grayscaleing

Many image processing algorithms assume a 2D matrix

- ▶ not an image with a third dimension of color

To bring the image into two dimensions

- ▶ we need to summarize the three colors into a single value
- ▶ this process is more commonly know as **grayscaleing**
- ▶ where the resulting image only holds intensities of gray
  - ▶ with values between 0 and 1

skimage submodule **color** has useful functions for this task

- ▶ API

`http://scikit-image.org/docs/dev/api/skimage.color.html`

# Grayscale

Goal: Create a grayscale version of a color image (see next slide)

```
1 import skimage.io as io
2 import skimage.color as color
3 import matplotlib.pyplot as plt
4 from skimage.color import rgb2gray
5
6 # read image into memory
7 image = io.imread("monkey.jpg")
8 # convert to grayscale
9 gray_image = rgb2gray(image)
10
11 print(image[0,0]) # prints [255,255,255]
12 print(gray_image[0,0]) # prints 1.0
13 plt.imshow(gray_image)
14 plt.show()
15 io.imsave("monkey_grayscale.jpg", gray_image)
```



# Binary image

Goal: Produce a black-and-white version of a color image (see next slide).

```
1 import skimage.io as io
2 import skimage.color as color
3 import matplotlib.pyplot as plt
4 from skimage.color import rgb2gray
5 import numpy as np
6
7 image = io.imread("monkey.jpg")
8 gray_image = rgb2gray(image)
9
10 # this creates a new array,
11 # with 1's everywhere gray_image > 0.5, and 0 elsewhere
12 black_and_white = np.where(gray_image > 0.5, 255, 0)
13
14 plt.imshow(black_and_white)
15 plt.show()
16 io.imsave("monkey_black_and_white.jpg", black_and_white)
```



# Blurring an image

Goal: Reduce the resolution of an image by blurring it, e.g. to reduce fine-level "noise" (unwanted details).



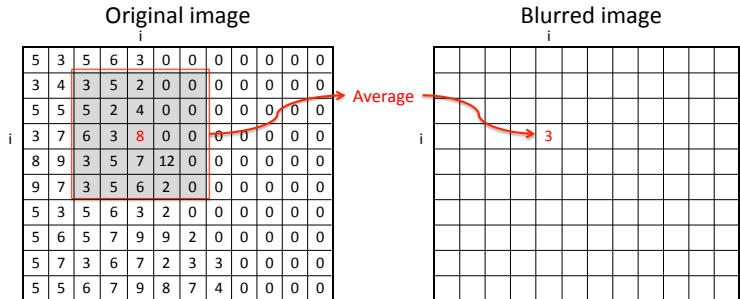
to



# Blurring an image

Blurring is achieved by replacing each pixel by the average value of the pixels in a small window centered on it.

Example, window of size 5:





## Blurring an image

```
1 def blur(image, filter_size):
2     n_row, n_col, colors = image.shape
3     blurred=np.zeros((n_row, n_col, colors),dtype=np.uint8)
4     half_size=int(filter_size/2)
5     for i in range(n_row):
6         for j in range(n_col):
7             # define the boundaries of window around (i,j)
8             left=max(0,j-half_size)
9             right=min(j+half_size, n_col)
10            top=max(0,i-half_size)
11            bot=min(n_row, i+half_size)
12            # calculate average of RGB values in window
13            blurred[i, j] = \
14                image[bot:top, left:right, :].mean(axis=(0,1))
15    return blurred_image
```

- ▶ `image[ bottom:top, left:right , ,:]` corresponds to the sub-image ranging from rows bottom to top-1 and columns left to right-1, and all 3 color dimensions.
- ▶ `means(axis=(0,1))` states that we want to take an average over dimension 0 (rows) and dimension 1 (columns) but not dimension 2 (RGB). This returns that a 1d ndarray containing the average red, green, and blue values in the subimage.

# Original image



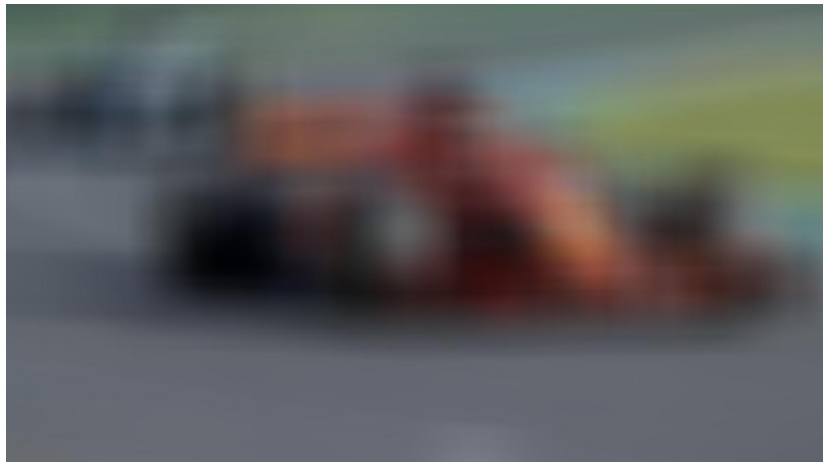
Window size = 5



Window size = 21



Window size = 101



## Running time issues

Note: When our window size is large (say 101), blurring the image is slow ( $> 1$  minute). Why?

- ▶ Our image is  $674 \times 1200$  pixels.
- ▶ For each pixel in the image, we need to calculate the average of the  $101 \times 101$  pixels around it, and for each of the three colors!
- ▶ The total number of operations is proportional to  $674 \times 1200 \times 101 \times 101 = 25$  Billion operations!

SkImage has many built-in blurring functions (called filters) with faster implementations:

<http://scikit-image.org/docs/dev/api/skimage.filters.html>

# Edge detection

Goal: Identify regions of the image that contain sharp changes in colors/intensities.

Why? Useful for

- ▶ delineating objects (image segmentation)
- ▶ recognizing them (object recognition)
- ▶ etc.

# Edge detection





# Edge detection



# Edge detection

What's an edge in an image?

Horizontal edge at row  $i$ :  $image(i - 1, j)$  is very different from  $image(i + 1, j)$

Vertical

edge at column  $j$ :  $image(i, j - 1)$  is very different from  $image(i, j + 1)$

Idea:

For each position  $(i, j)$  and each color (RGB), calculate

$change\_hor = image(i-1, j, color) - image(i+1, j, color)$

$change\_vert = image(i, j-1, color) - image(i, j+1, color)$

$edge\_image(i, j, color) = \sqrt{change\_hor^2 + change\_vert^2}$

# Edge detection

```
1 def detect_edges(image):
2     n_row, n_col, colors = image.shape
3     edge_image = np.zeros( (n_row, n_col, 3), dtype=np.uint8)
4     for i in range(1, n_row-1):
5         for j in range(1, n_col-1):
6             for c in range(3):
7
8                 # conversion to int needed to accommodate
9                 # for potentially negative values
10                d_r=int(image[i-1,j,c])-int(image[i+1,j,c])
11                d_c=int(image[i,j-1,c])-int(image[i,j+1,c])
12                grad = math.sqrt(d_r**2+d_c**2)
13
14                # limit value to 255
15                edge_image[i,j,c]=np.uint8(min(255, grad))
16     return edge_image
```

## Edge detection on monkey image

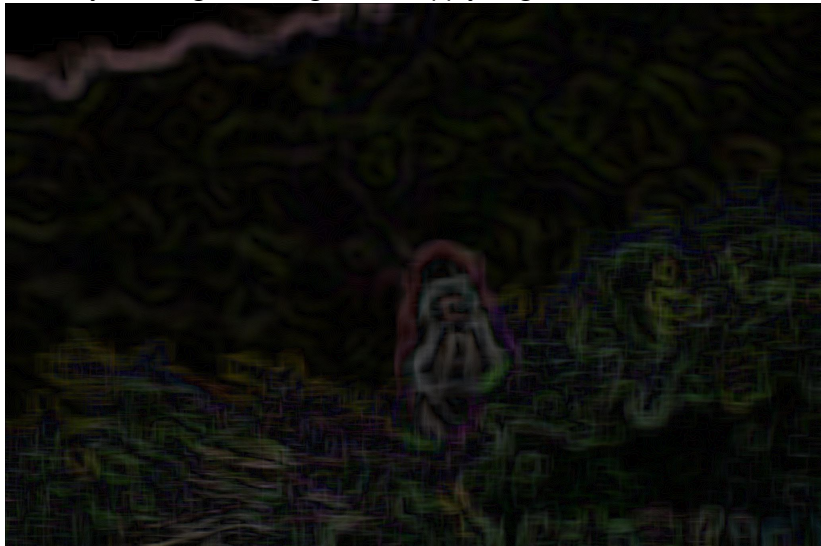


Not so great if our goal is to find the monkey in the image!

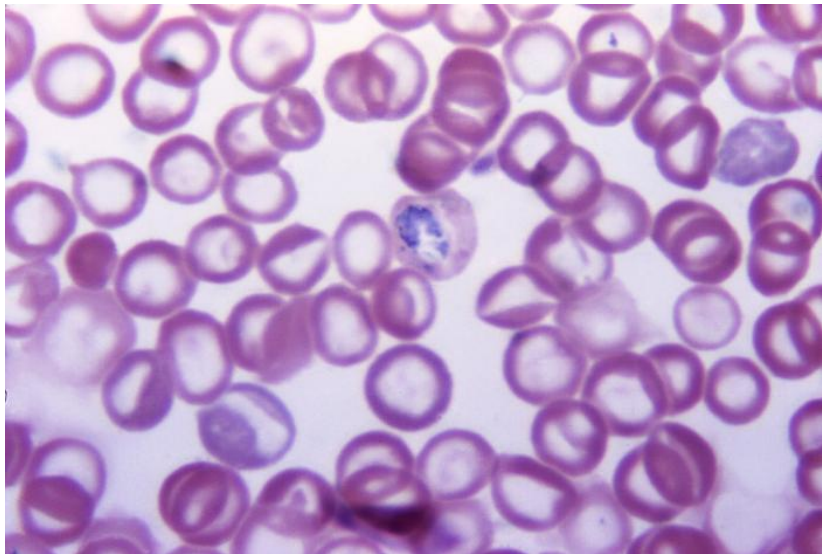
## Blurring + Edge detection

To smooth out fine details like leaves:

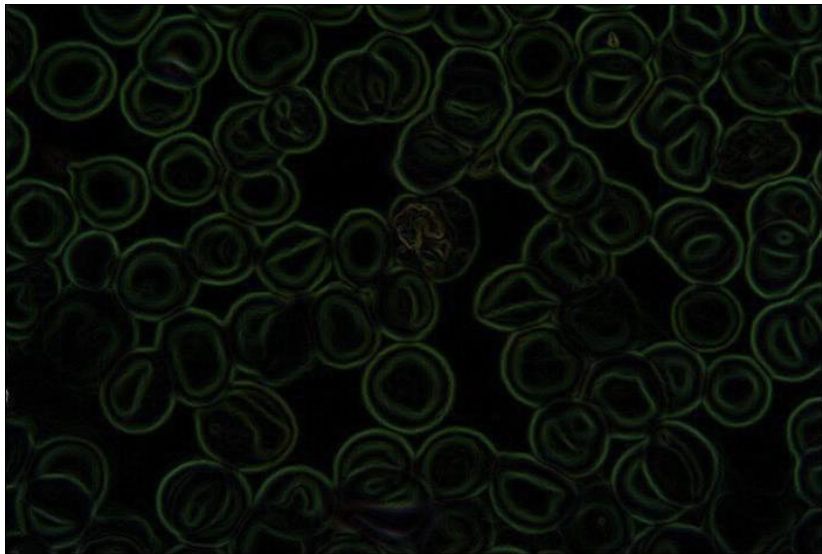
Start by blurring the image, then apply edge detection.



## Analysis of microscopy images



## Edge detection



# Edge detection

Skimage has many edge detection algorithms:

[http://scikit-image.org/docs/0.5/auto\\_examples/plot\\_canny.html](http://scikit-image.org/docs/0.5/auto_examples/plot_canny.html)