



BIOMEDICAL IMAGE ANALYSIS IN PYTHON

Intensity Values

Stephen Bailey Instructor



Pixels and voxels

- Pixels are 2D picture elements
- Voxels are 3D volume elements
- Two properties: intensity and location





Data types and image size

Array's data type controls range of possible intensities

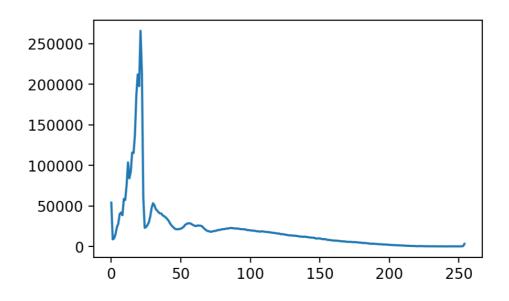
Data Type	Range	No. Values
uint8	0, 255	256
int8	-128, 127	256
uint16	0, 2 ¹⁶	2^{16}
int16	-2 ¹⁵ , 2	2 ¹⁶
float16	~-2 ¹⁶ , ~2 ¹⁶	>>216

```
import imageio
im=imageio.imread('foot-xray.jpg')
im.dtype
    dtype('uint8')
im.size
    153600
im_int64 = im.astype(np.uint64)
im_int64.size
    1228800
```

Histograms

- **Histograms**: count number of pixels at each intensity value.
- Implemented in scipy.ndimage
 - higher-dimensional arrays
 - masked data
- Advanced techniques and functionality in scikit-image.

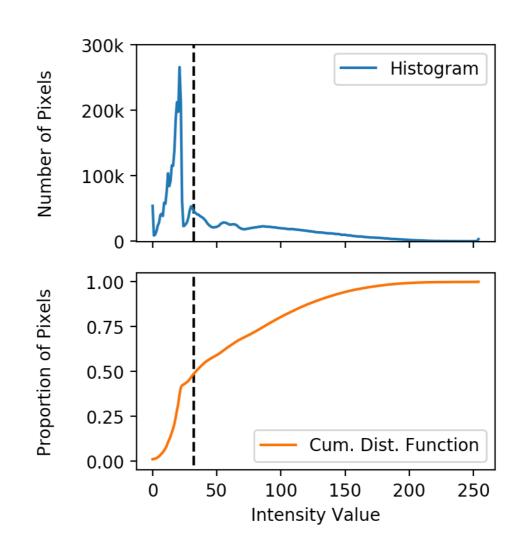
```
plt.plot(hist)
plt.show()
```





Equalization

- Distributions often skewed toward low intensities (background values).
- Equalization: redistribute values to optimize full intensity range.
- Cumulative distribution
 function: (CDF) shows
 proportion of pixels in range.





Equalization

```
fig, axes = plt.subplots(2, 1)
axes[0].imshow(im)
axes[1].imshow(im_equalized)
plt.show()
```









Let's practice!





BIOMEDICAL IMAGE ANALYSIS IN PYTHON

Masks

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Masks

Raw image

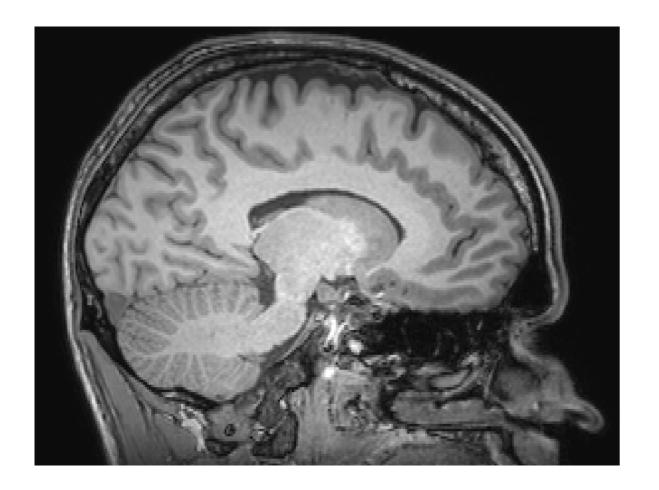


Image mask





Creating masks

Logical operations result in True / False at each pixel

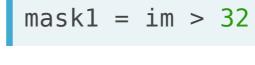
Sample Operations

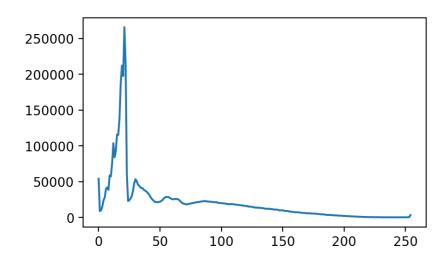
Operation	Example	
Greater	im > 0	
Equal to	im == 1	
X and Y	(im > 0) & (im < 5)	
X or Y	(im > 10) (im < 5)	

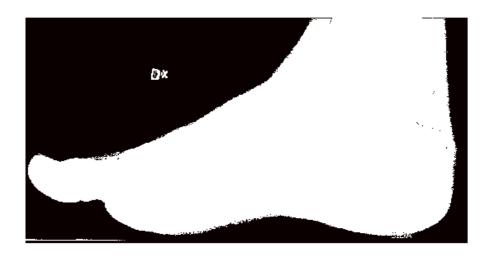


Creating masks

hist=ndi.histogram(im, 0, 255, 256)



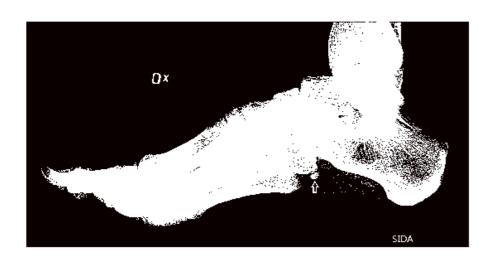






Creating masks

mask2 = im > 64



mask3 = mask1 & ~mask2



Applying masks

np.where(condition, x, y): control
what data passes through the
mask.

```
import numpy as np
im_bone = np.where(im > 64, im, 0)
```

```
plt.imshow(im_bone, cmap='gray')
plt.axis('off')
plt.show()
```

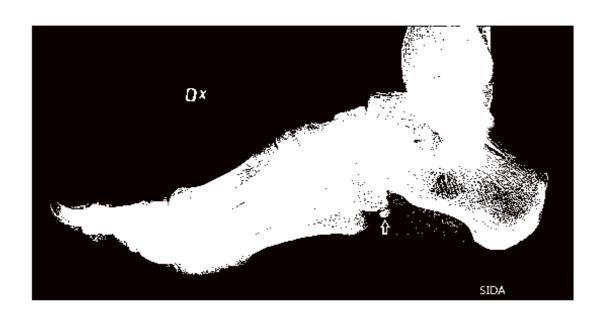




Tuning masks

```
m = np.where(im > 64, 1, 0)
```

ndi.binary_dilation(m,iterations=5)





Tuning masks

ndi.binary_erosion(m,iterations=5)







Let's practice!



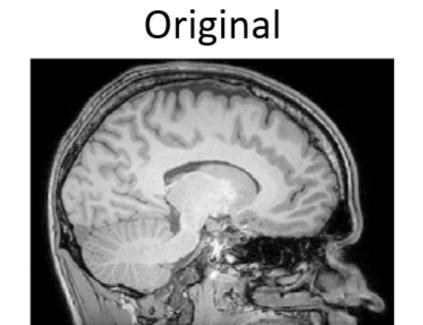
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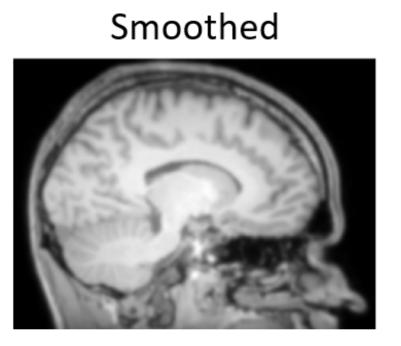
Filters

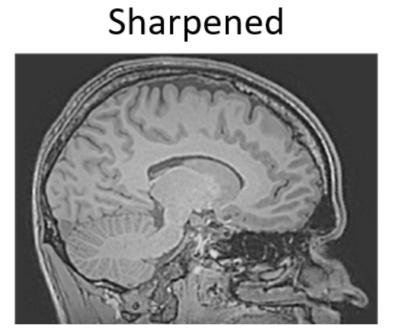
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Filters

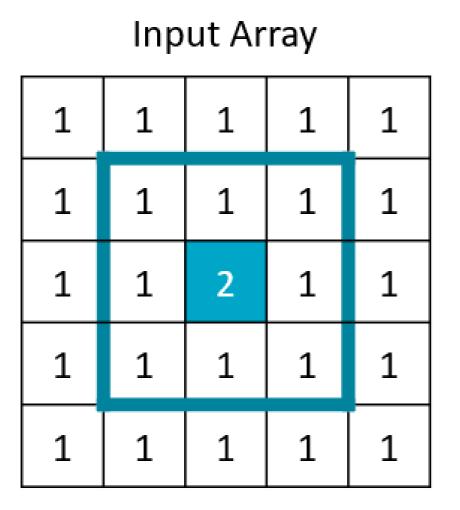




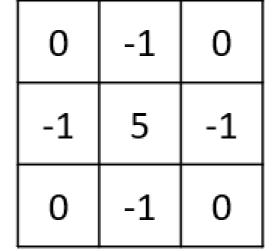




Convolution with a sharpening filter



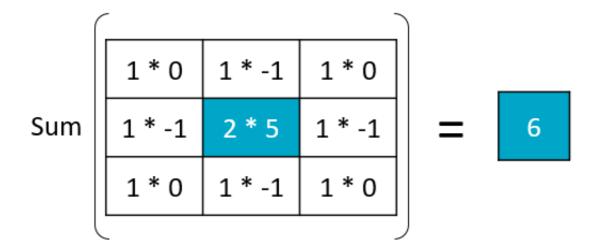
Filter Weights / Kernel



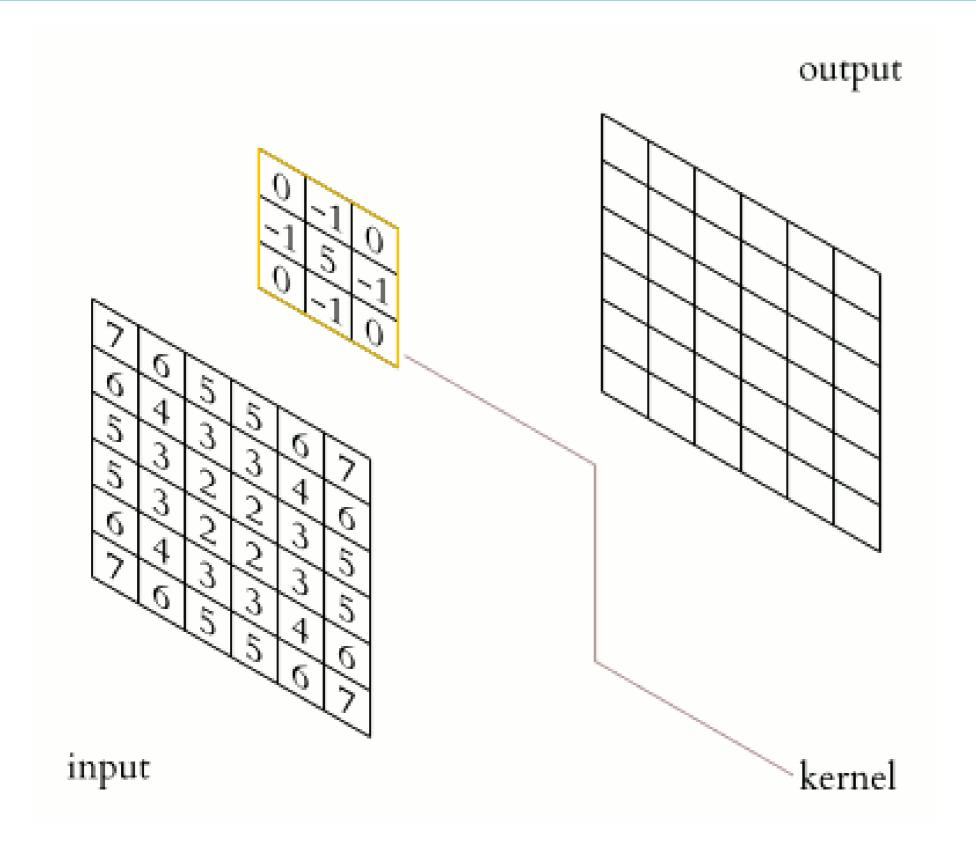
*

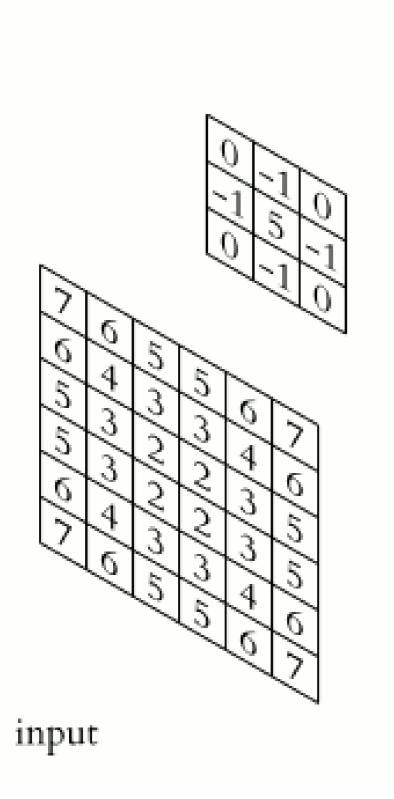


Convolution with a sharpening filter









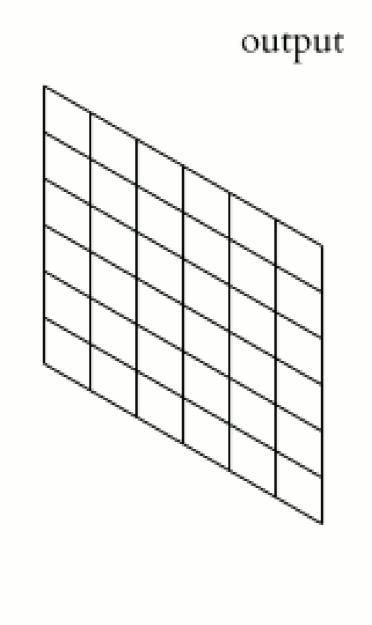


Image convolution

```
fig, axes = plt.subplots(2, 1)
axes[0].imshow(im, cmap='gray')
axes[1].imshow(im_filt,cmap='gray')
plt.imshow()
```







Filtering functions

scipy.ndimage.filters includes:

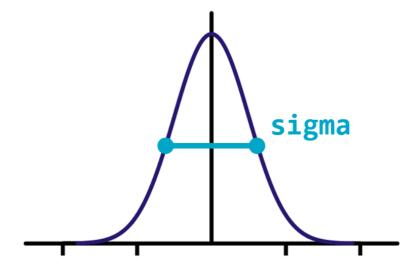
- median_filter()
- uniform_filter()
- maximum_filter()
- percentile_filter()

ndi.median_filter(im, size=10)

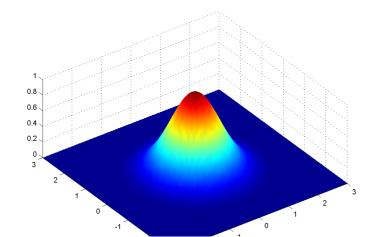


Gaussian filtering

Gaussian distribution in 1 dimension



Gaussian distribution in 2 dimensions



ndi.gaussian_filter(im, sigma=5)



ndi.gaussian_filter(im, sigma=10)







Let's practice!





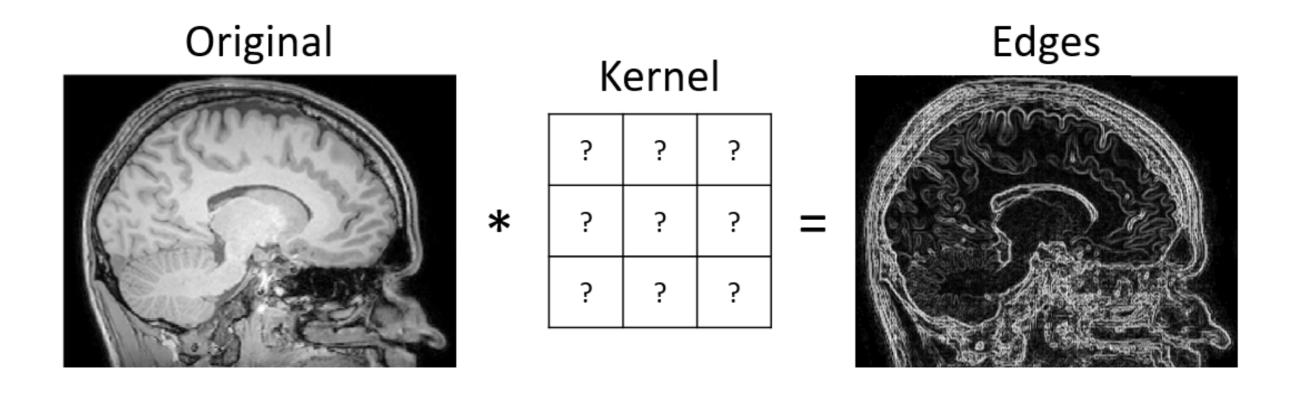
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Feature Detection

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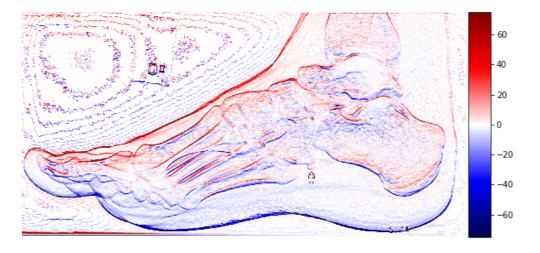


Edges: sharp changes in intensity





Edge detection





Sobel filters

Sobel (H)

1	2	1
0	0	0
-1	-2	-1

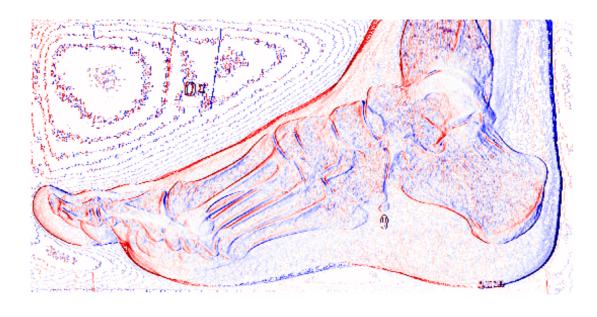
Sobel (V)

1	0	-1
2	0	-2
1	0	-1

Sobel filters

ndi.sobel(im, axis=0)

ndi.sobel(im, axis=1)

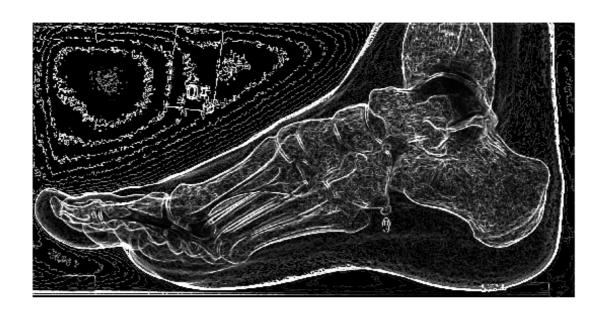


Sobel filter magnitude

Combine horizontal and vertical edge data by calculating distance:

$$z=\sqrt{x^2+y^2}$$

```
plt.imshow(edges, cmap='gray')
```







Let's practice!