



# **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

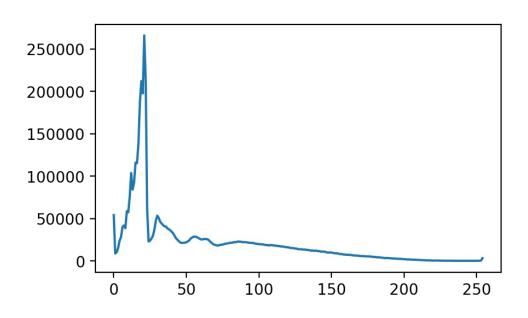
Data Type	Range	No. Values
uint8	0, 255	256
int8	-128, 127	256
uint16	0, 2 <sup>16</sup>	$2^{16}$
int16	$-2^{15}$ , $2^{15}$	$2^{16}$
float16	$\sim -2^{16}$ , $\sim 2^{16}$	>>2 <sup>16</sup>

```
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

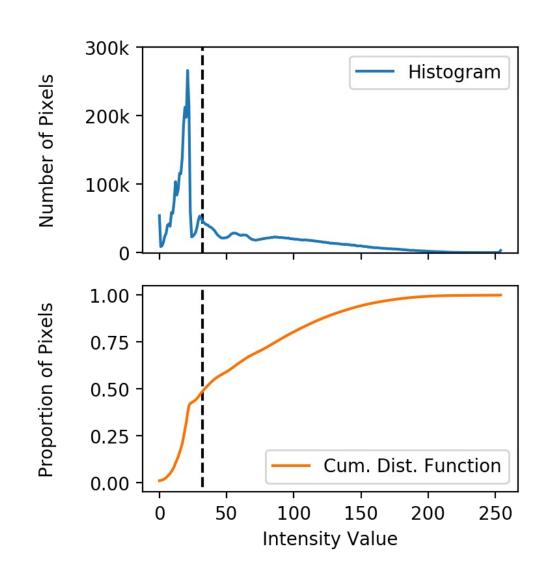
- **Histograms**: count number of pixels at each intensity value.
- Implemented in scipy.ndimage
  - higher-dimensional arrays
  - masked data
- Advanced techniques and functionality in scikit-image.





### 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

```
import scipy.ndimage as ndi
hist = ndi.histogram(im, min=0,
                         \max=255,
                         bins=256)
cdf = hist.cumsum() / hist.sum()
cdf.shape
    (256,)
im equalized = cdf[im] * 255
fig, axes = plt.subplots(2, 1)
axes[0].imshow(im)
axes[1].imshow(im_equalized)
plt.show()
```









# Let's practice!





### Masks

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## Masks

Raw image

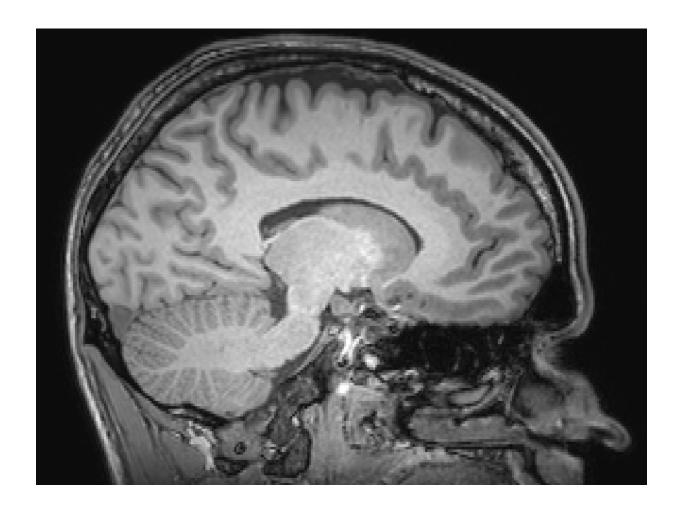


Image mask





# Creating masks

Logical operations result in True / False at each pixel

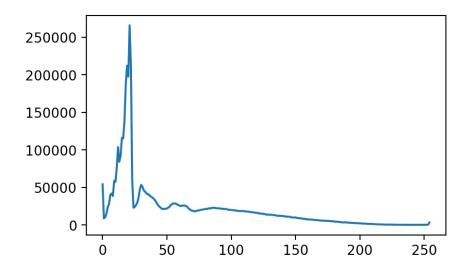
#### **Sample Operations**

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

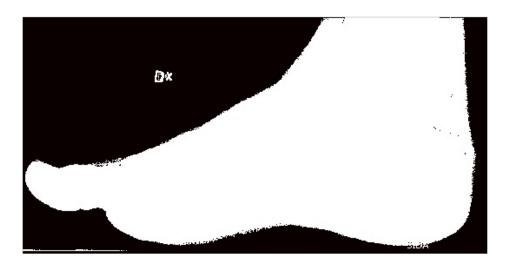


# Creating masks

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



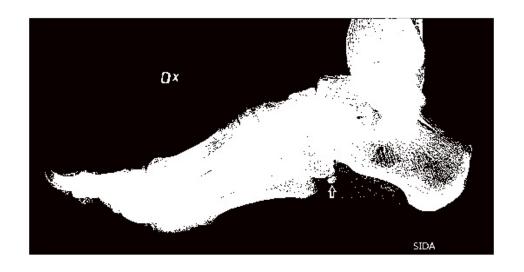
mask1 = im > 32





# Creating masks

$$mask2 = im > 64$$



mask3 = mask1 & ~mask2



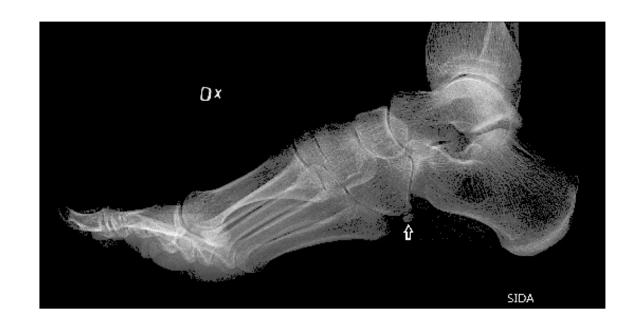


# Applying masks

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

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

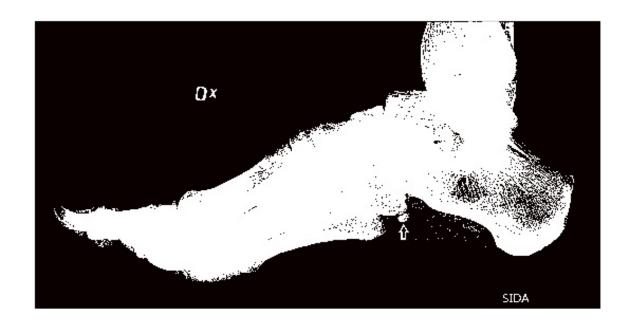
plt.imshow(im_bone, cmap='gray')
plt.axis('off')
plt.show()
```





# 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!



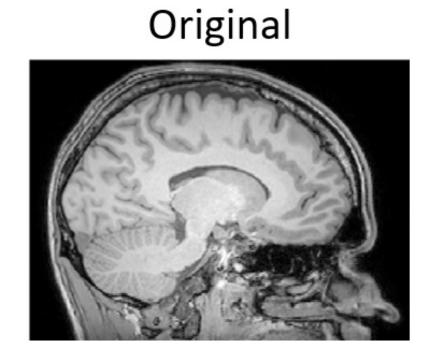


# **Filters**

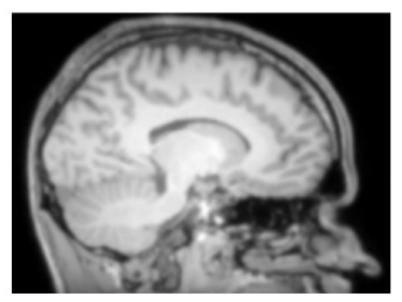
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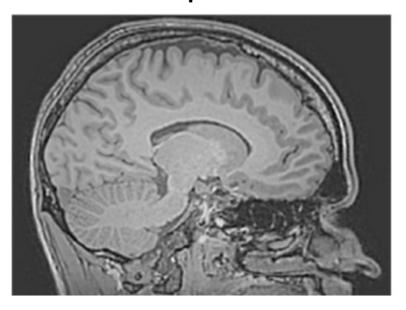
## Filters



Smoothed

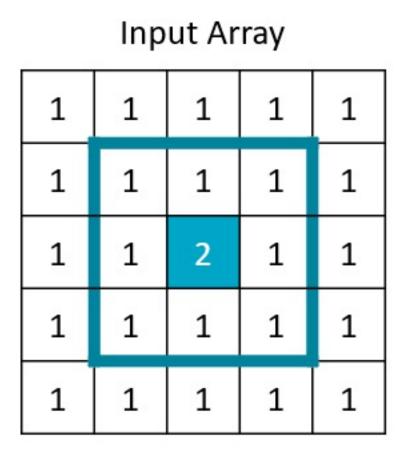


Sharpened





## Convolution with a sharpening filter

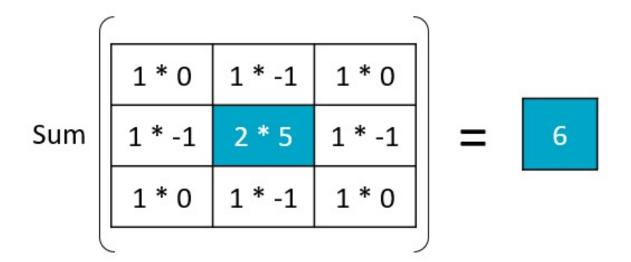


Filter Weights / Kernel

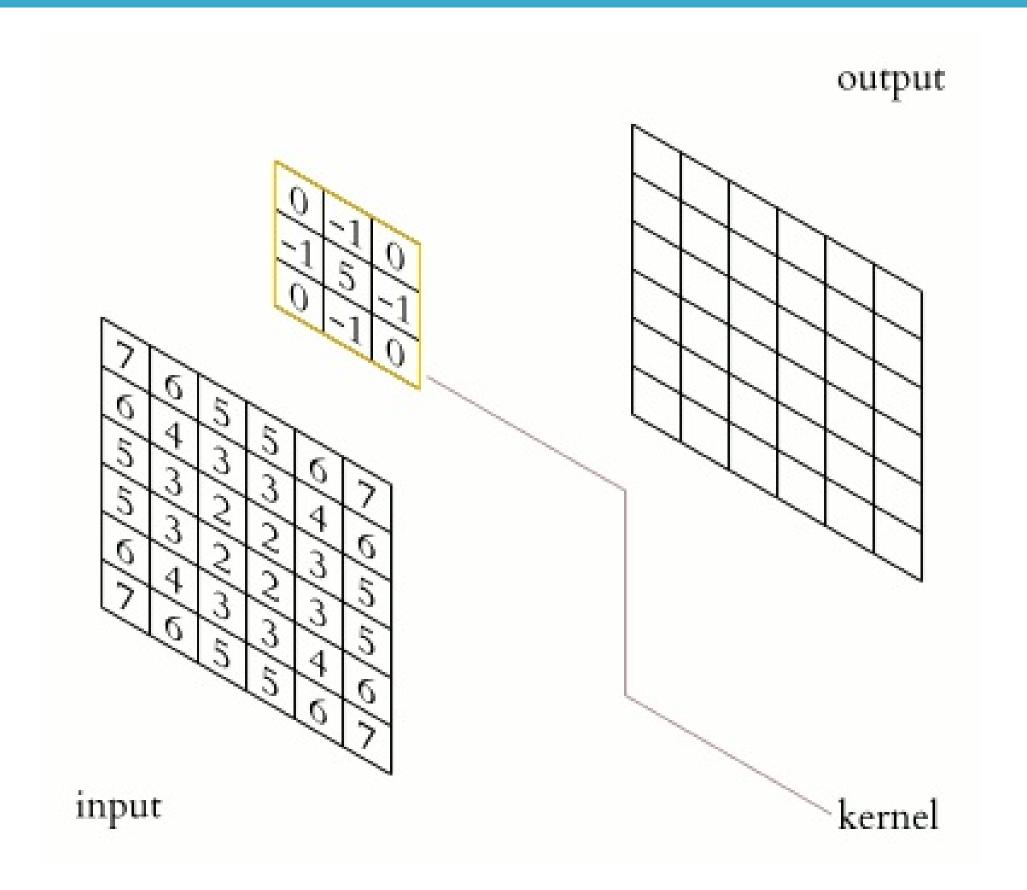
0	-1	0
-1	5	-1
0	-1	0

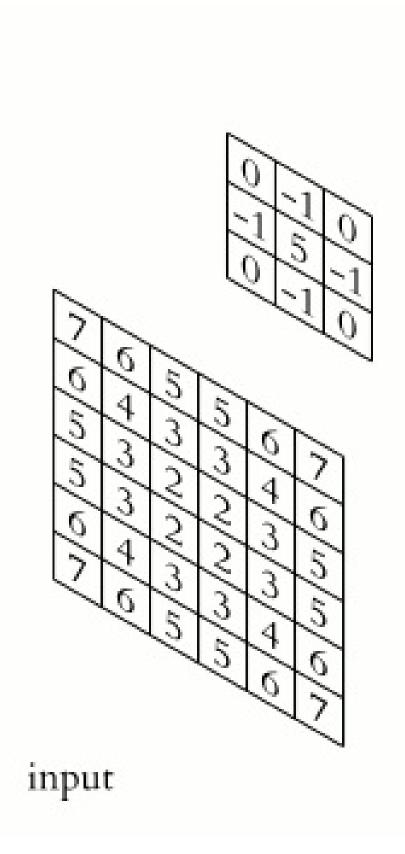


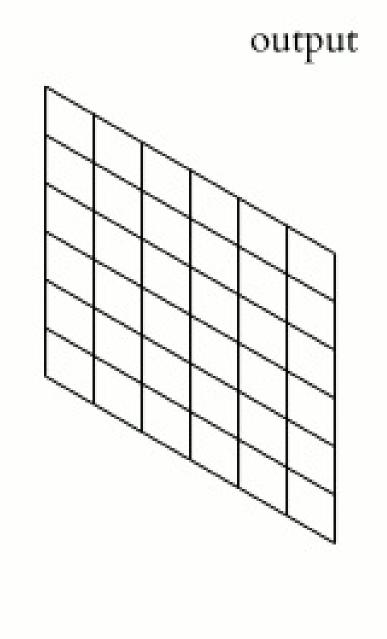
## Convolution with a sharpening filter













### Image convolution







# Filtering functions

scipy.ndimage.filters includes:

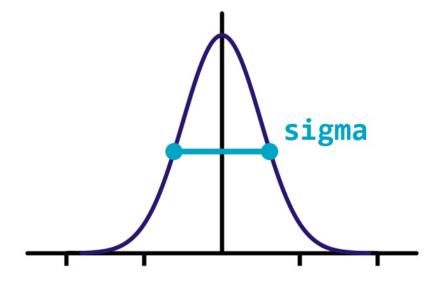
- median\_filter()
- uniform\_filter()
- maximum\_filter()
- percentile\_filter()

ndi.median\_filter(im, size=10)

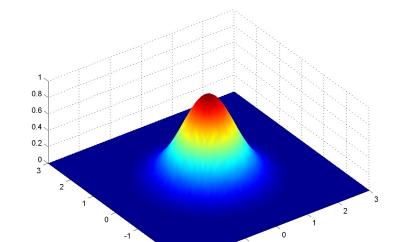


# Gaussian filtering

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!



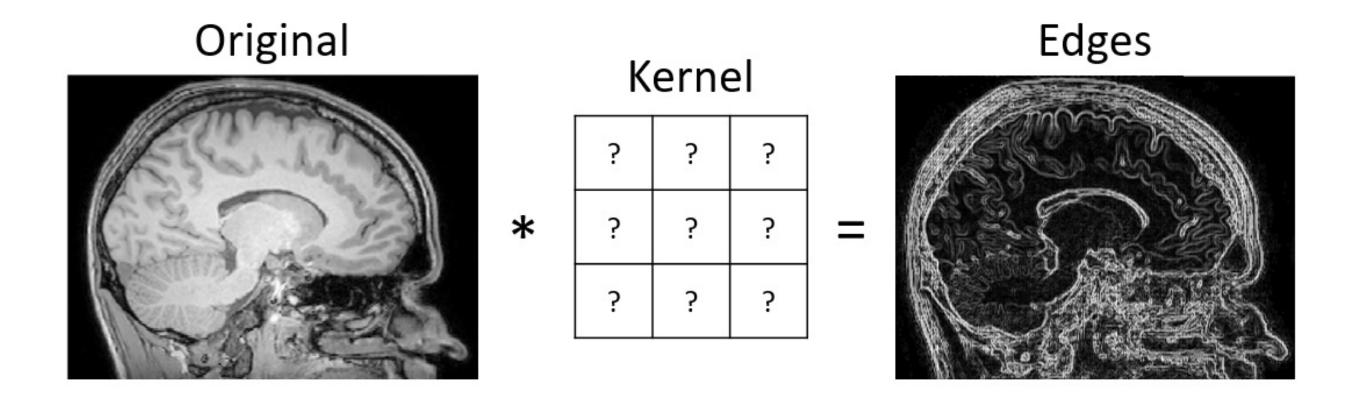


### **Feature Detection**

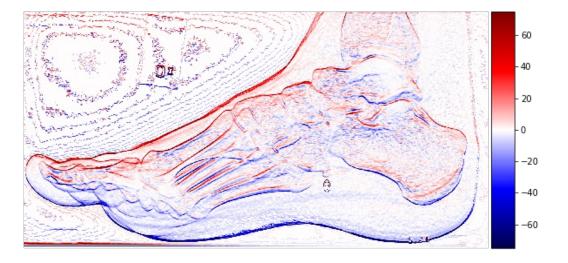
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# Edges: sharp changes in intensity



## Edge detection





### Sobel filters

Sobel (H)

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

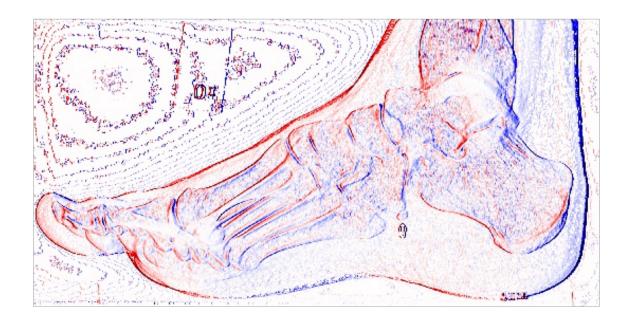
Sobel (V)

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

### Sobel filters

ndi.sobel(im, axis=0)

ndi.sobel(im, axis=1)



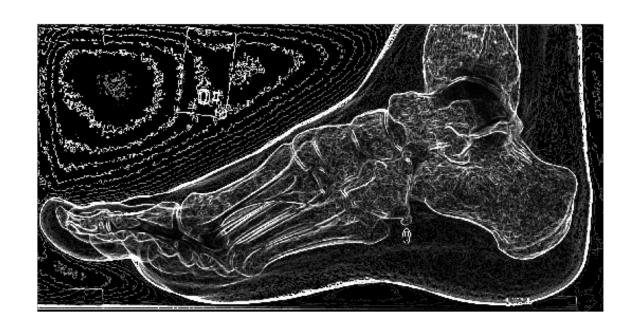


# Sobel filter magnitude

Combine horizontal and vertical edge data by calculating distance:

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

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







# Let's practice!