

# Introduction to Digital Image Analysis

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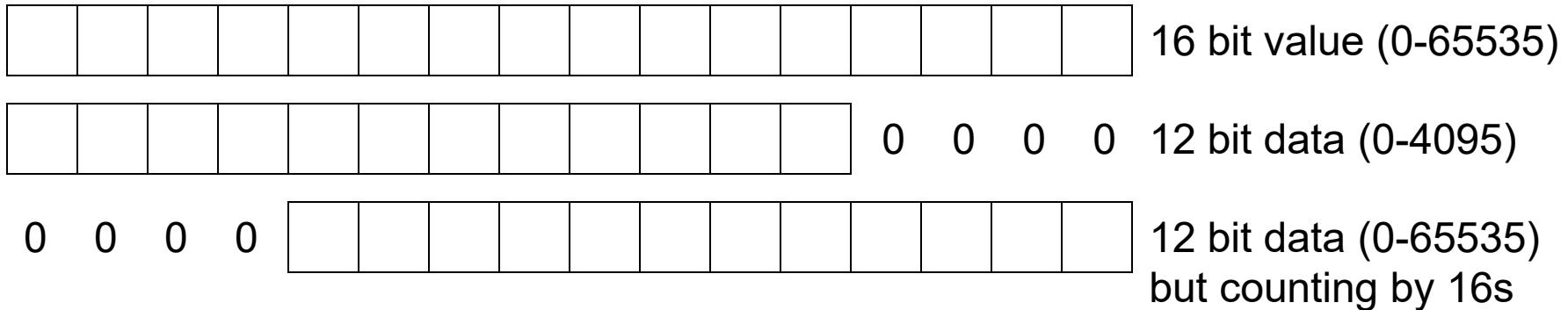
NIC

# Bitdepth

- Digital cameras have a specified bitdepth = number of gray levels they can record
- 8-bit  $\rightarrow 2^8 = 256$  gray levels
- 10-bit  $\rightarrow 2^{10} = 1024$  gray levels
- 12-bit  $\rightarrow 2^{12} = 4096$  gray levels
- 14-bit  $\rightarrow 2^{14} = 16384$  gray levels
- 16-bit  $\rightarrow 2^{16} = 65536$  gray levels

# Bitdepth and file formats

- Standard imaging formats, like tiff, are always 8 or 16 bit (because 8 bits = 1 byte)



# Color Images

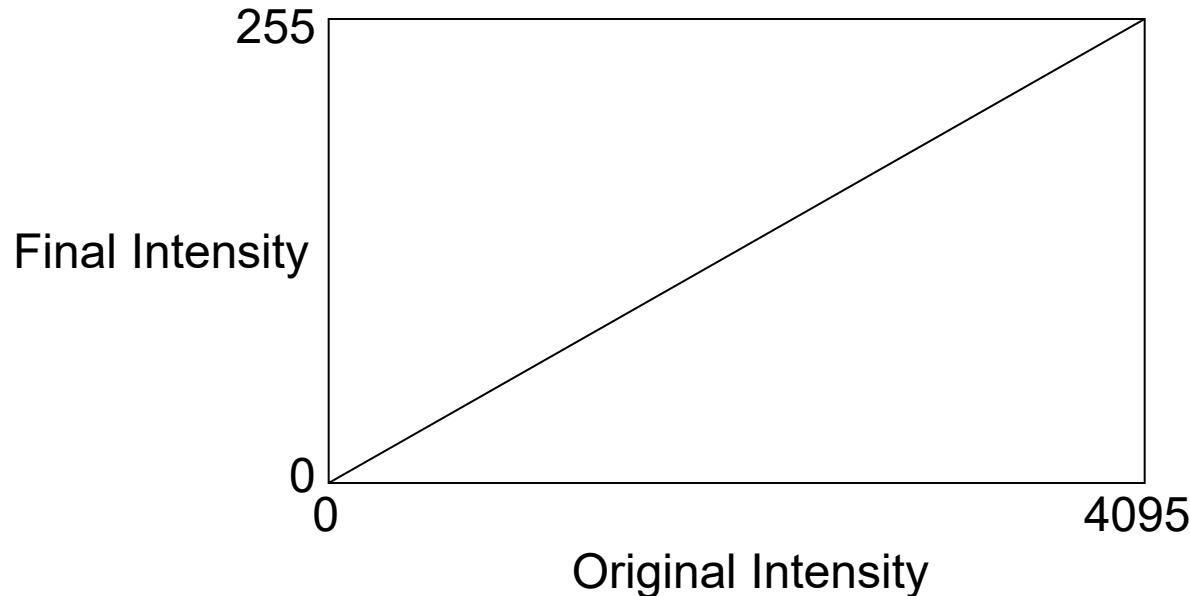
- Color images are made up of three gray scale images, one for each of red, green, and blue
- Can be 8 or 16 bits per channel

# File Formats

- Most portable: TIFF
  - 8 or 16-bit, lossless, supports grayscale or RGB
- OK: JPEG2000, custom formats (nd2, ids, zvi, lsm, etc.)
  - Lossless, supports full bitdepth
  - Custom formats often support multidimensional images
  - Not so portable
- Bad: Jpeg, GIF, BMP, etc.
  - Lossy and / or 8-bit

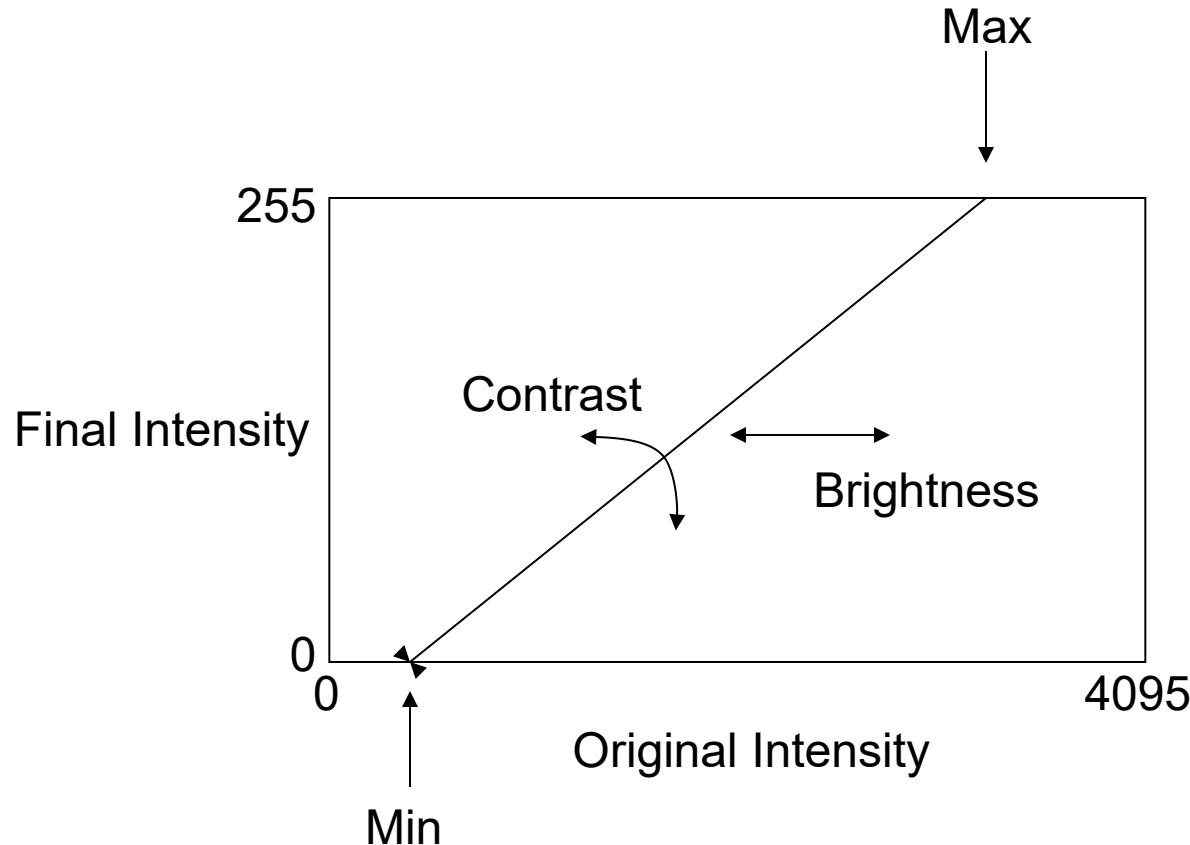
# Intensity scaling

- Computer screens are 8-bit
- Publishers also want 8-bit files

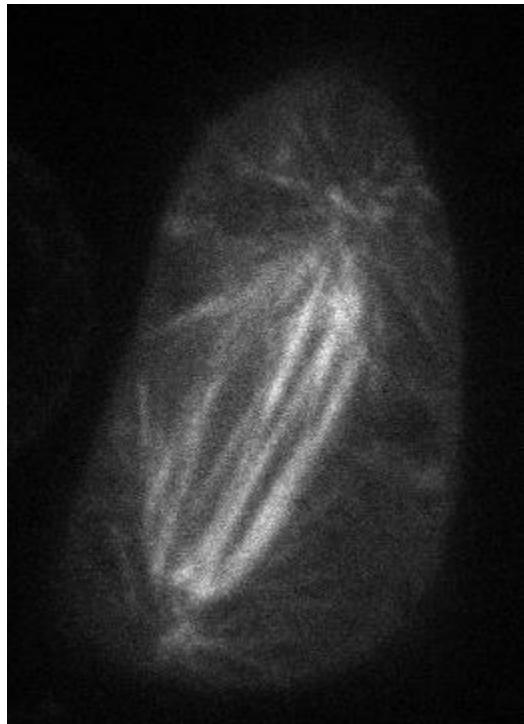


You lose information in this process –  
values 4080-4095 all end up as 255

# Intensity scaling



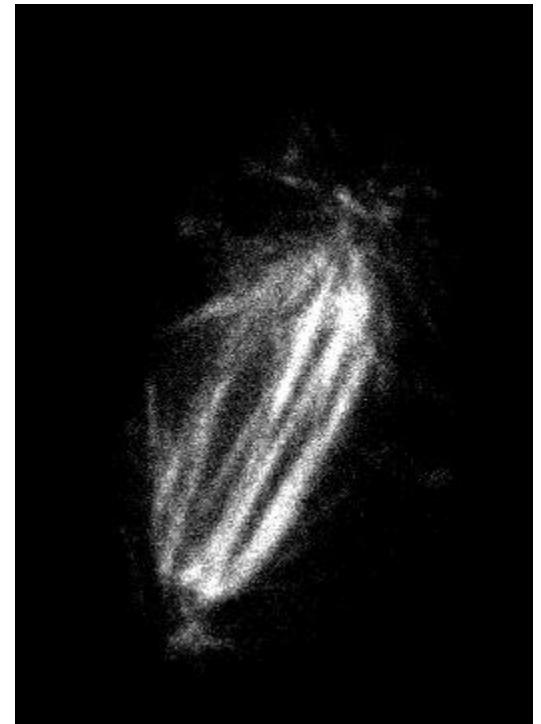
# Effect of scaling



Scaled to min / max  
(874 / 25438)



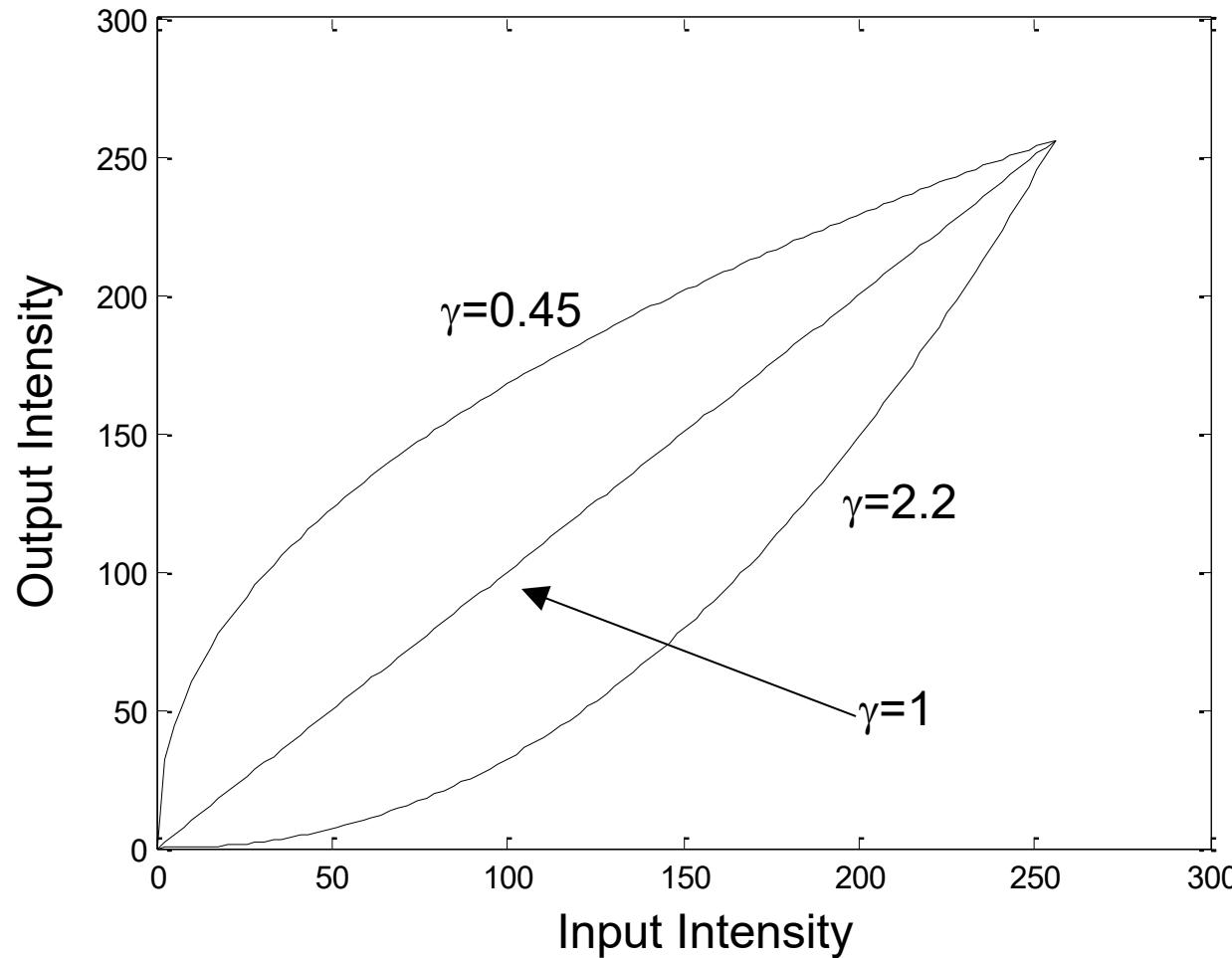
(874 / 19200)



(6400 / 18432)

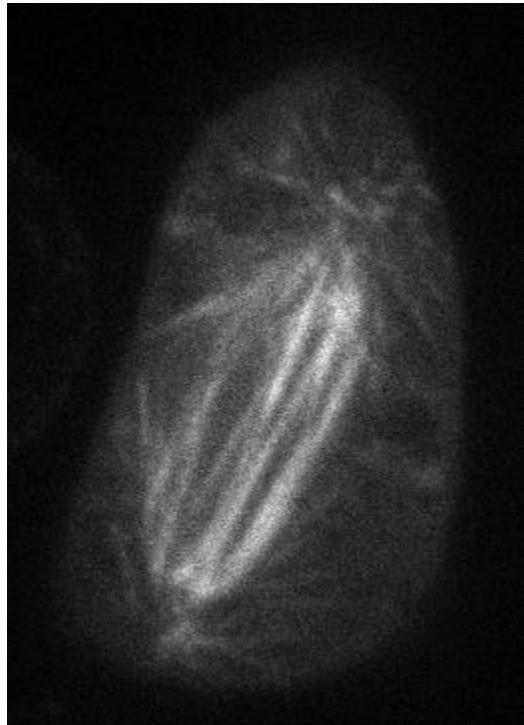
Drosophila S2 cell with mCherry-tubulin (Nico Stuurman)

# Gamma correction

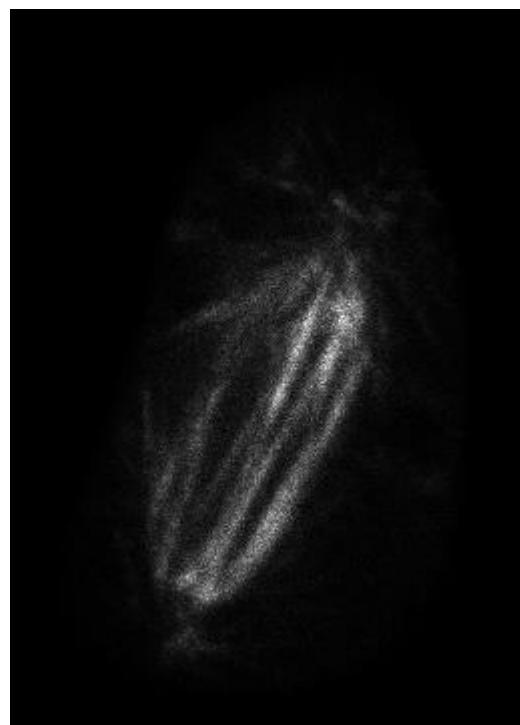


Other contrast stretching transforms....

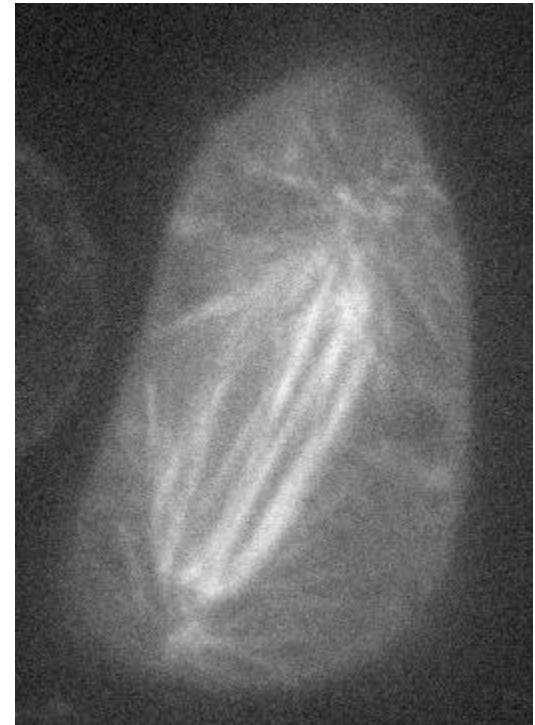
# Effect of gamma



Scaled to min / max  
(874 / 25438),  $\gamma = 1$



Scaled to min / max  
(874 / 25438),  $\gamma = 2.16$



Scaled to min / max  
(874 / 25438),  $\gamma = 0.45$

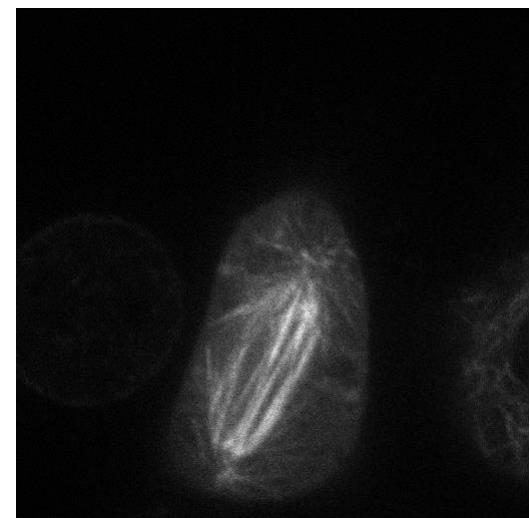
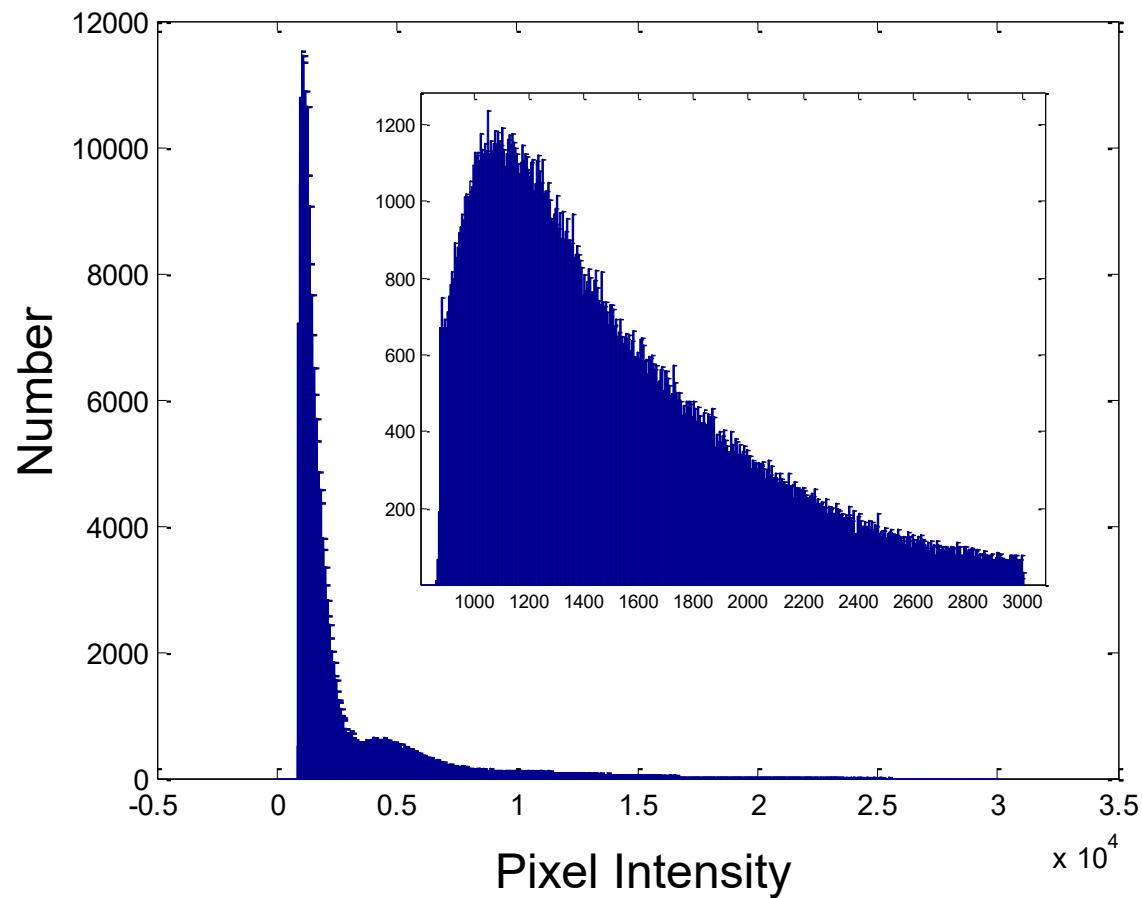
# What are acceptable image manipulations?

- JCB has the best guidelines
  - [http://jcb.rupress.org/misc/ifora.shtml#image\\_aquisition](http://jcb.rupress.org/misc/ifora.shtml#image_aquisition)
  - <http://jcb.rupress.org/cgi/content/full/166/1/1>
- Brightness and contrast adjustments ok, so long as done over whole image and don't obscure or eliminate background
- Nonlinear adjustments (like gamma) must be disclosed
- No cutting and pasting of regions within an image (e.g. individual cells)

# Background correction

- Cameras have a non-zero offset
- There can also be background fluorescence due to media autofluorescence, etc.
- Want to correct for this by background subtraction
  - Camera dark image
  - Estimate background from image

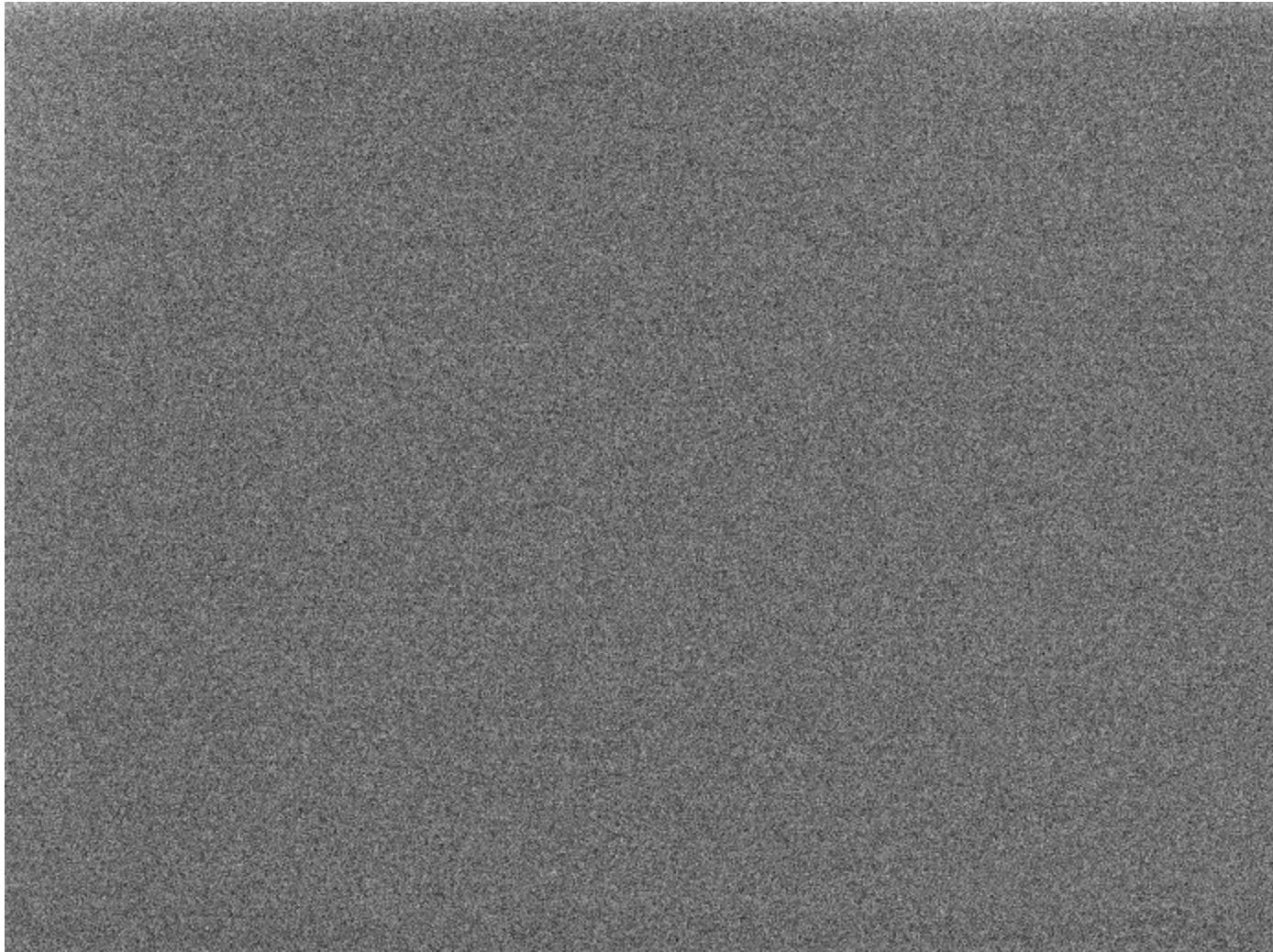
# Estimating background from image



# Dark image

- Acquired with no light going to the camera
  - Allows you to measure instrument background
  - Can detect what's real background autofluorescence

# Dark image



# Shading correction

- Measure and correct for nonuniformity in illumination and detection
- Image a uniform fluorescent sample

# Shading correction



# Correction procedure

$$I_{\text{meas}} = I_{\text{true}} * \text{Shading} + \text{Dark}$$

$$I_{\text{true}} = (I_{\text{meas}} - \text{Dark}) / \text{Shading}$$

Good to do on all images

## More formally...

$$I(x,y) = F(x,y) * E(x,y) * D(x,y) + D(x,y)$$



If no photobleaching, then Shading =  $E(x,y)*D(x,y)$  is a good approximation

If photobleaching...

$$I(x,y) = (F(x,y) / k^* E(x,y)) * [1 - \exp(-k^* E(x,y) * t)] * D(x,y) + \\ \text{Dark}(x,y)$$

Yuck

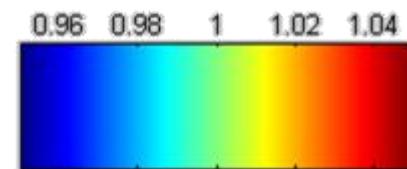
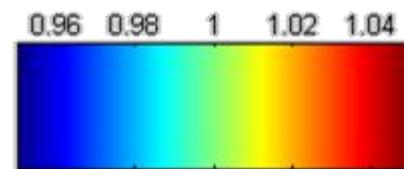
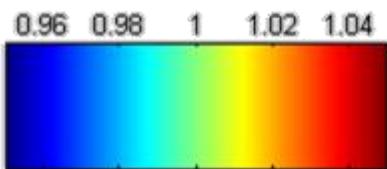
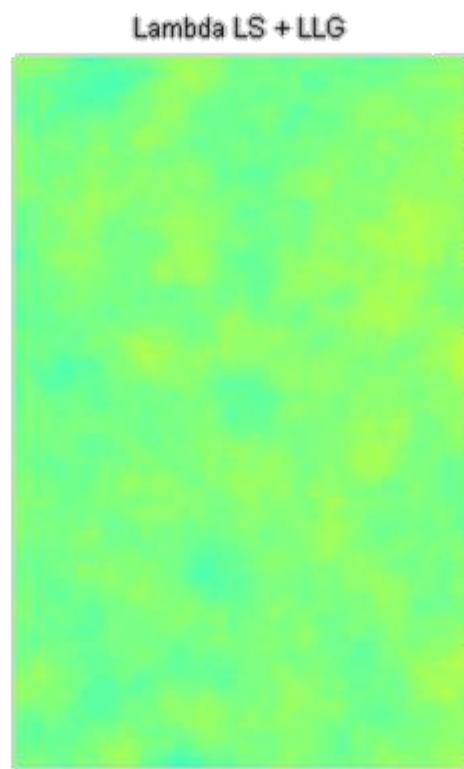
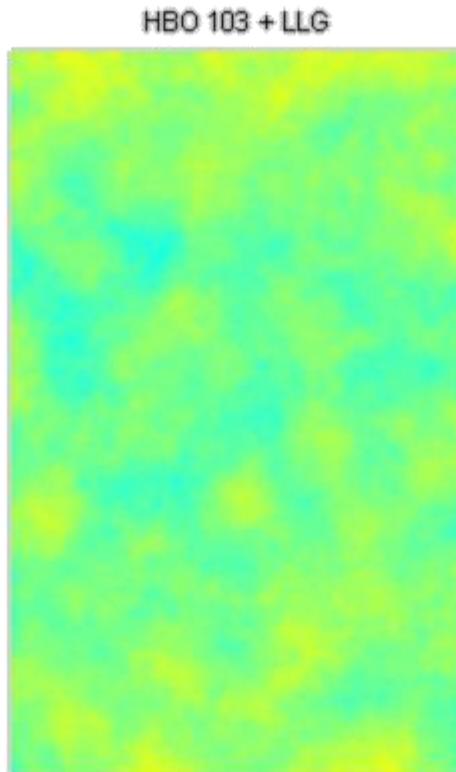
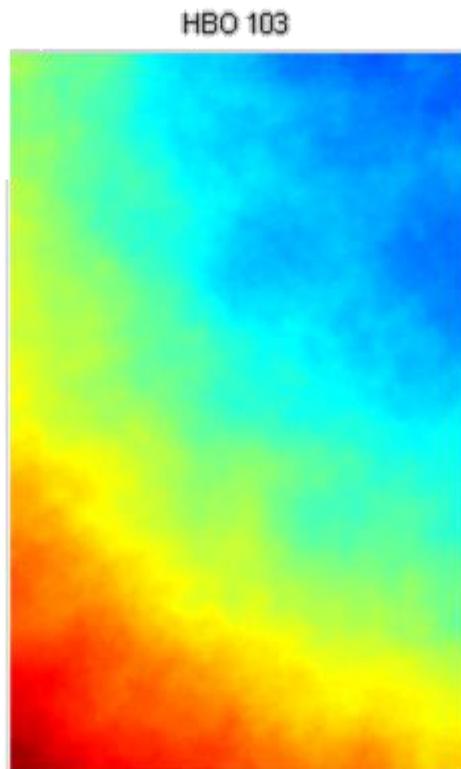
If you have significant photobleaching and significant excitation non-uniformity, shading correction breaks down

Can actually lead to contrast inversion (!)

# Measuring Excitation and Detection Distributions

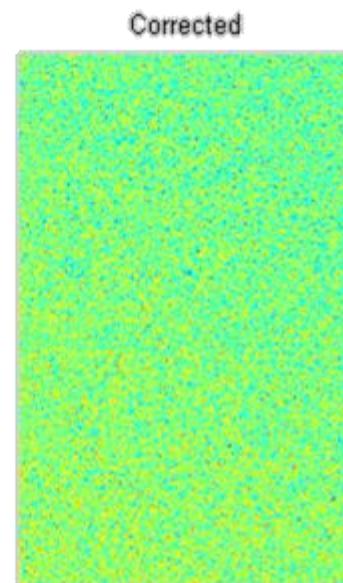
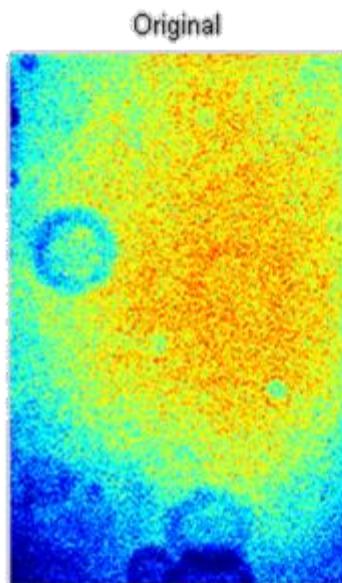
- Take photobleaching time-series on uniform thin-film of fluorophore
- Photobleaching rate at each pixel gives intensity
- Variance not explained by intensity distribution gives detector sensitivity distribution

# Illumination Uniformity



# Correcting

- Once excitation distribution is relatively flat, correction for detector distribution is once again linear.



# Back to digital image analysis

- Filtering

1 1 1

1 **1** 1

Averaging / Smoothing

1 1 1

0 1 2 1 0

1 6 10 6 1

2 10 **16** 10 2 Gaussian smoothing

1 6 10 6 1

0 1 2 1 0

# How this works

1	1	1
---	---	---

$$1 \quad \textcolor{red}{1} \quad 1 \quad (\mathbf{10+11+22+13+8+10+20+20+15})/9 = 14$$

1	1	1
---	---	---

Multiply corresponding pixels and sum

10	11	22	5	7
----	----	----	---	---

13	14	10	5	24
----	----	----	---	----

20	20	15	23	14
----	----	----	----	----

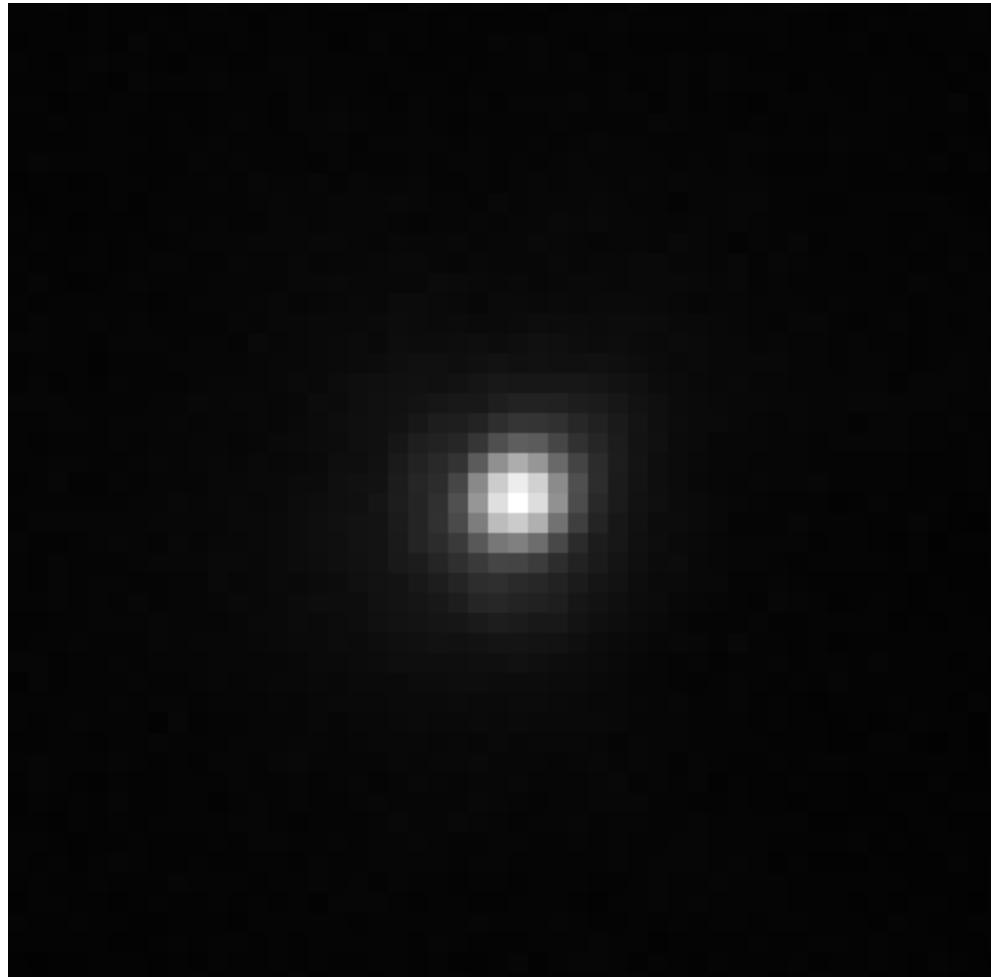
0	3	17	15	8
---	---	----	----	---

7	11	6	15	12
---	----	---	----	----

# Why smooth?

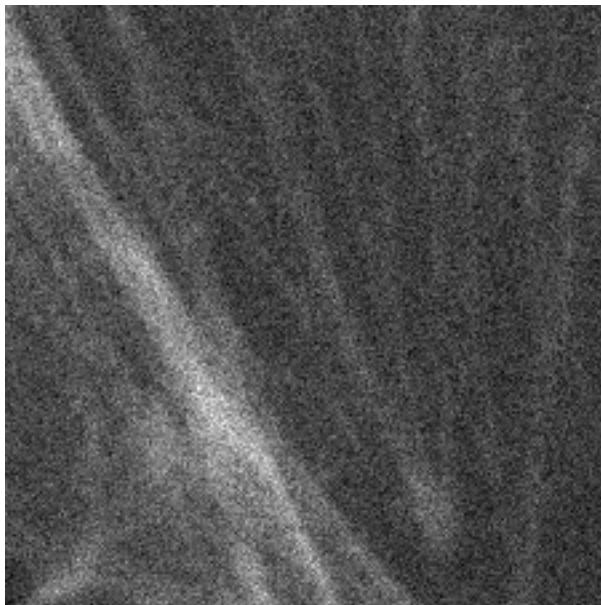
- If your image is sampled appropriately (at Nyquist) the point spread function will be spread out over multiple pixels
- Properly exploiting this redundancy requires deconvolution
- But smoothing helps
- Also reduces single pixel noise artifacts that can't be real

# Actual PSF

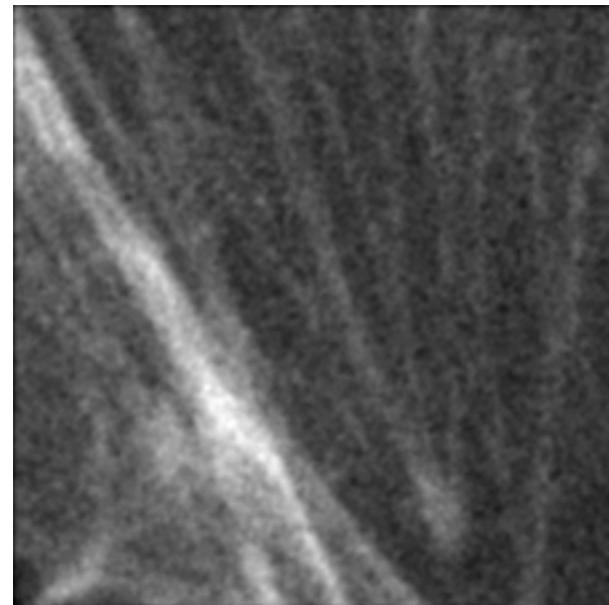


# Why smooth?

- Averages redundancy and suppresses noise

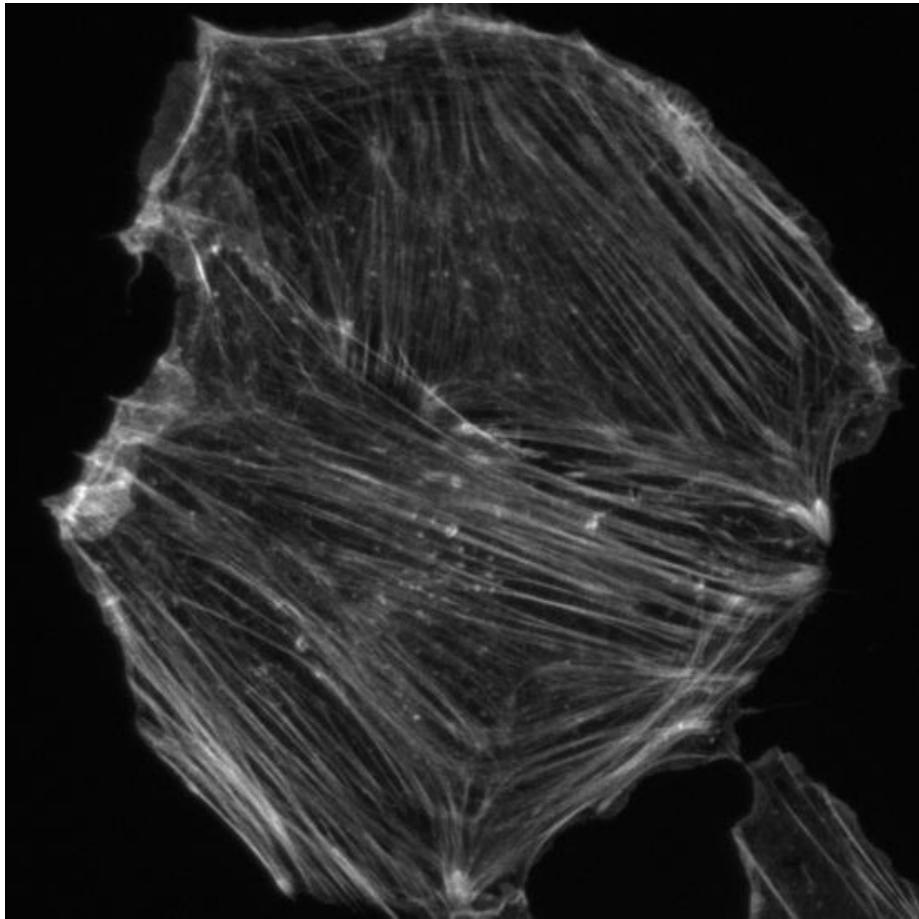


10 photons/pixel average  
5 e- read noise

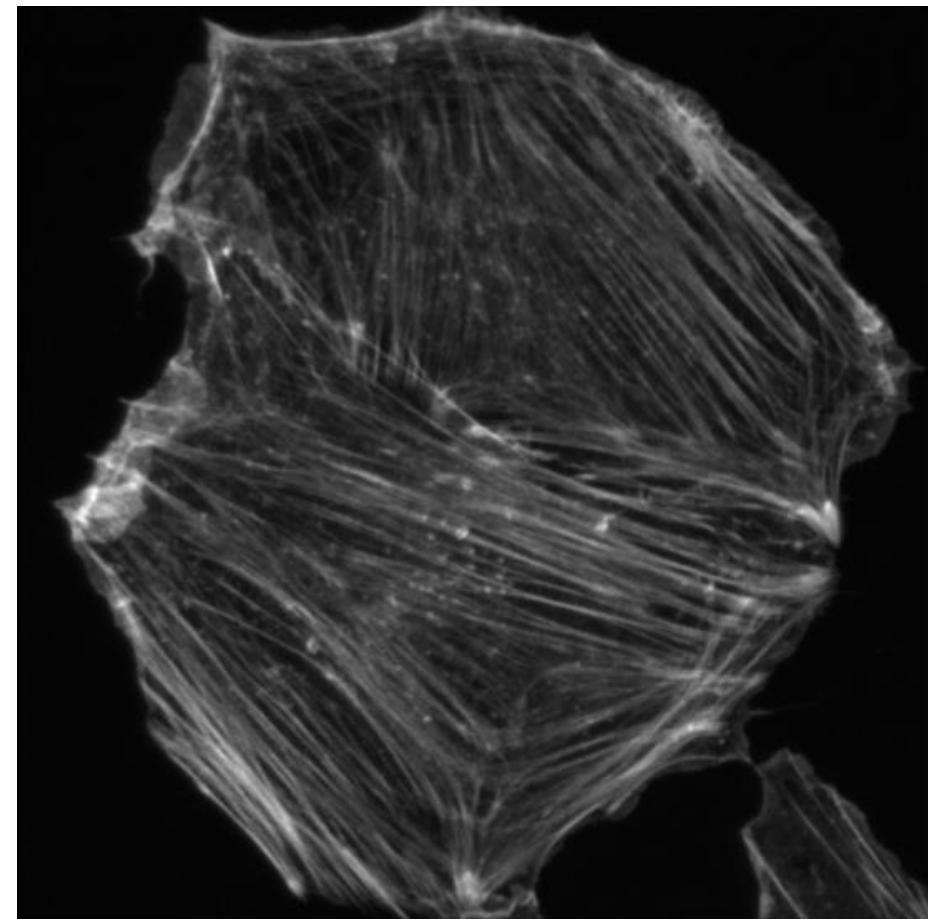


Gaussian smoothing  
filter,  $\sigma = 1$  pixel

# Smoothing



Original



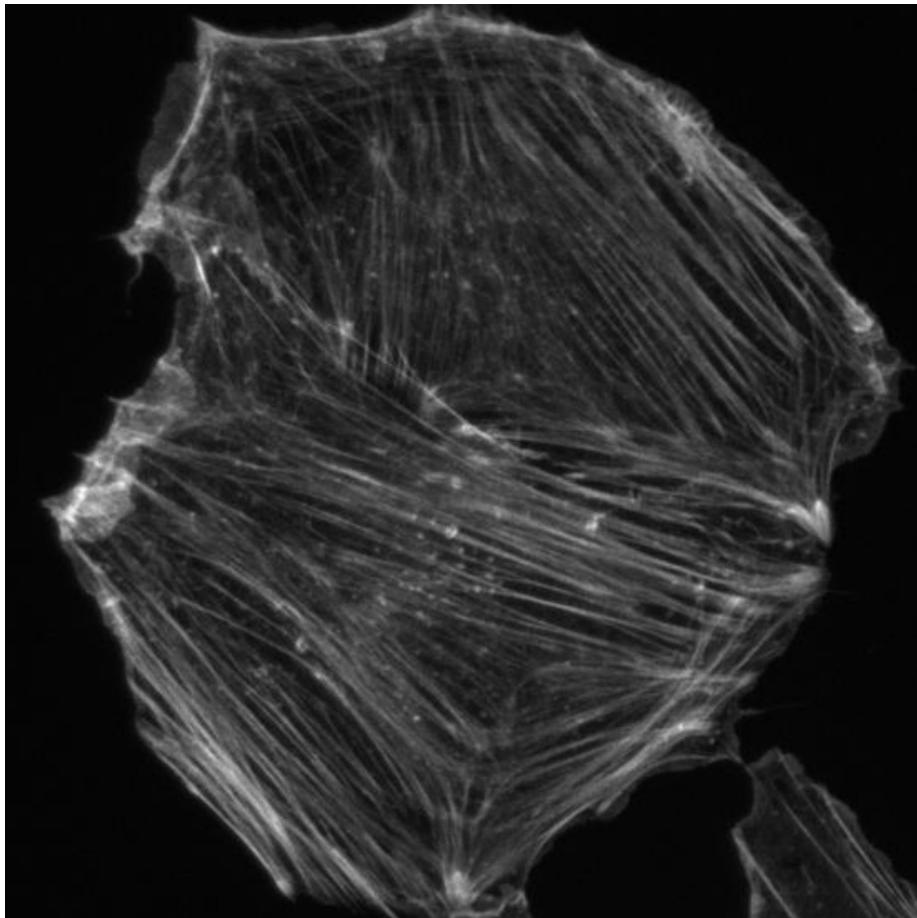
Gaussian filtered,  $\sigma = 0.6$

# Other filters

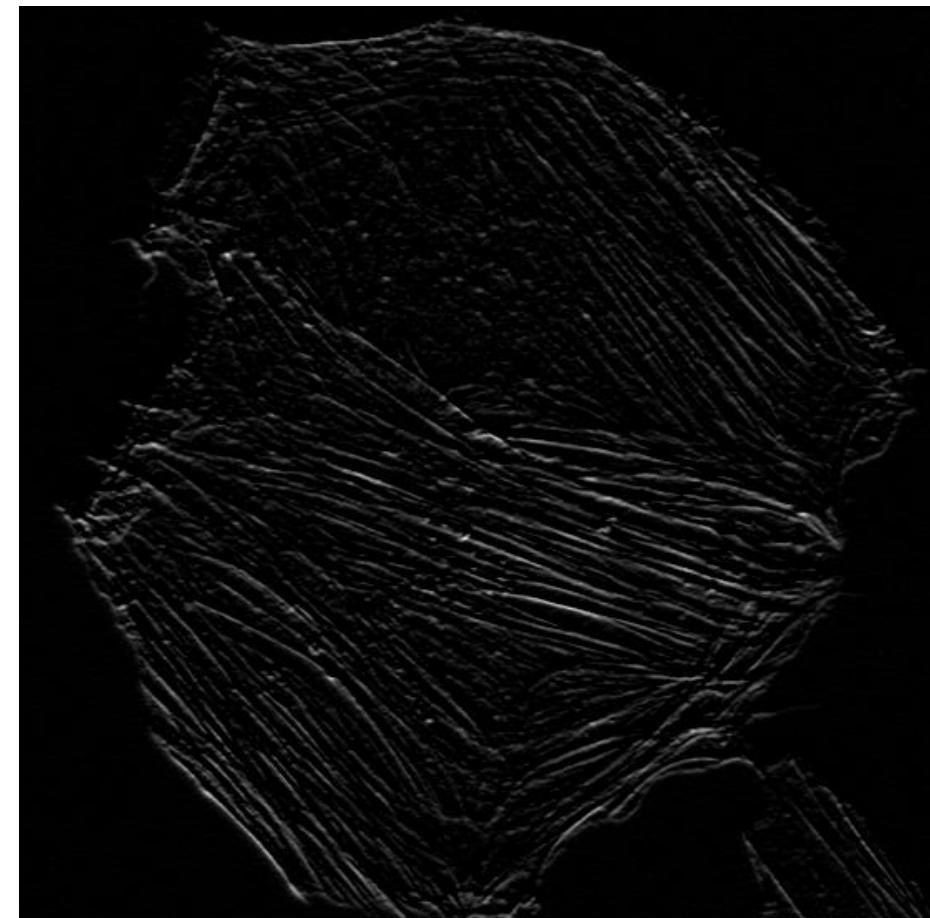
- Edge Detection

$$\begin{array}{ccc} 1 & 1 & 1 \\ 0 & \textcolor{red}{0} & 0 \\ -1 & -1 & -1 \end{array} \quad \begin{array}{ccc} 1 & 2 & 1 \\ 0 & \textcolor{red}{0} & 0 \\ -1 & -2 & -1 \end{array}$$

# Edge Detection



Original



Horizontal edge detection

# Other filters

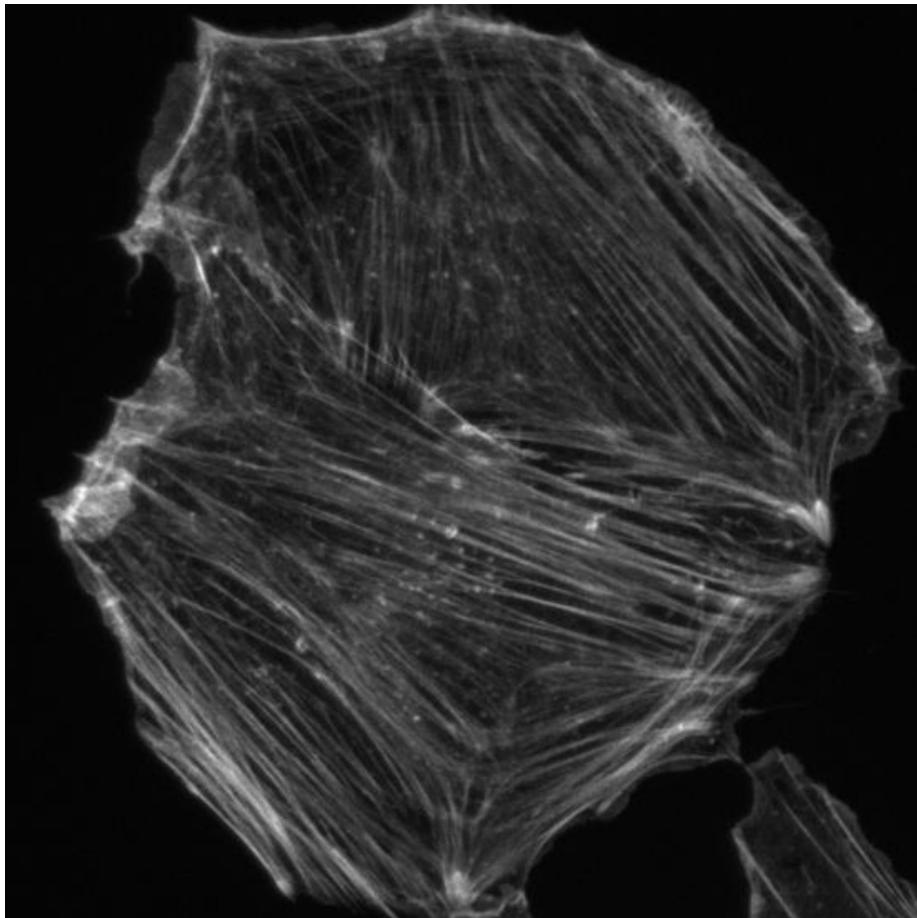
- Unsharp masking

-1      -4      -1

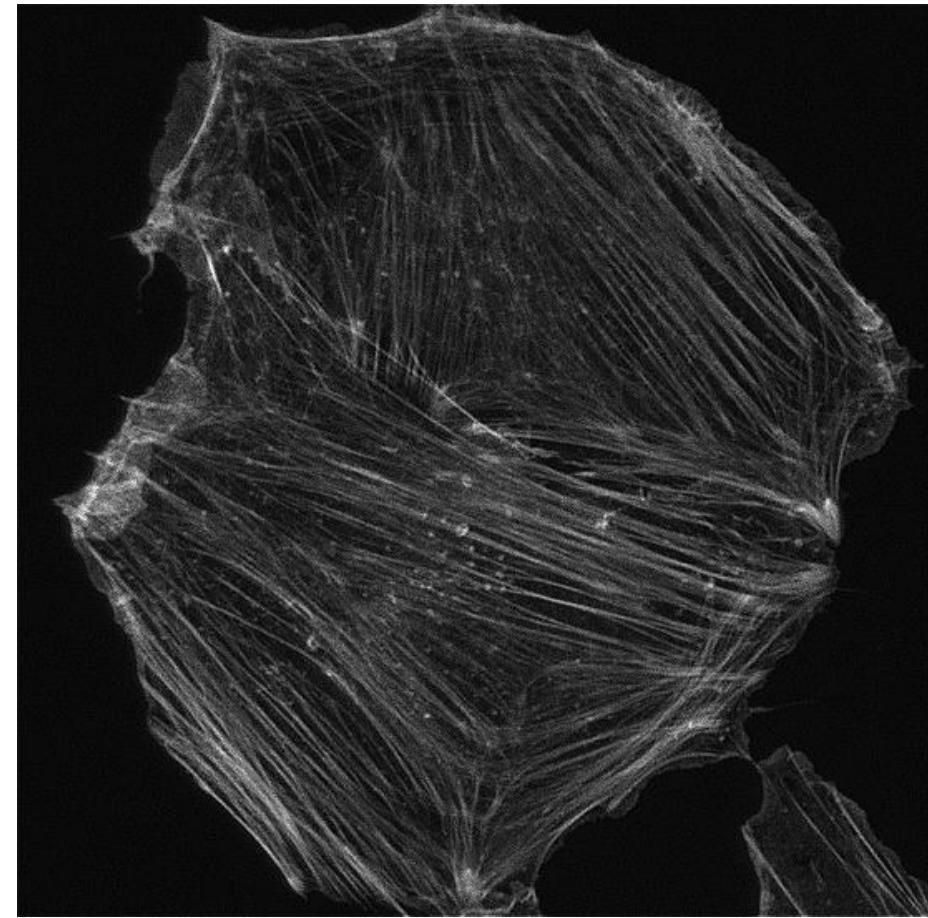
-4      **26**      -4

-1      -4      -1

# Unsharp Masking



Original

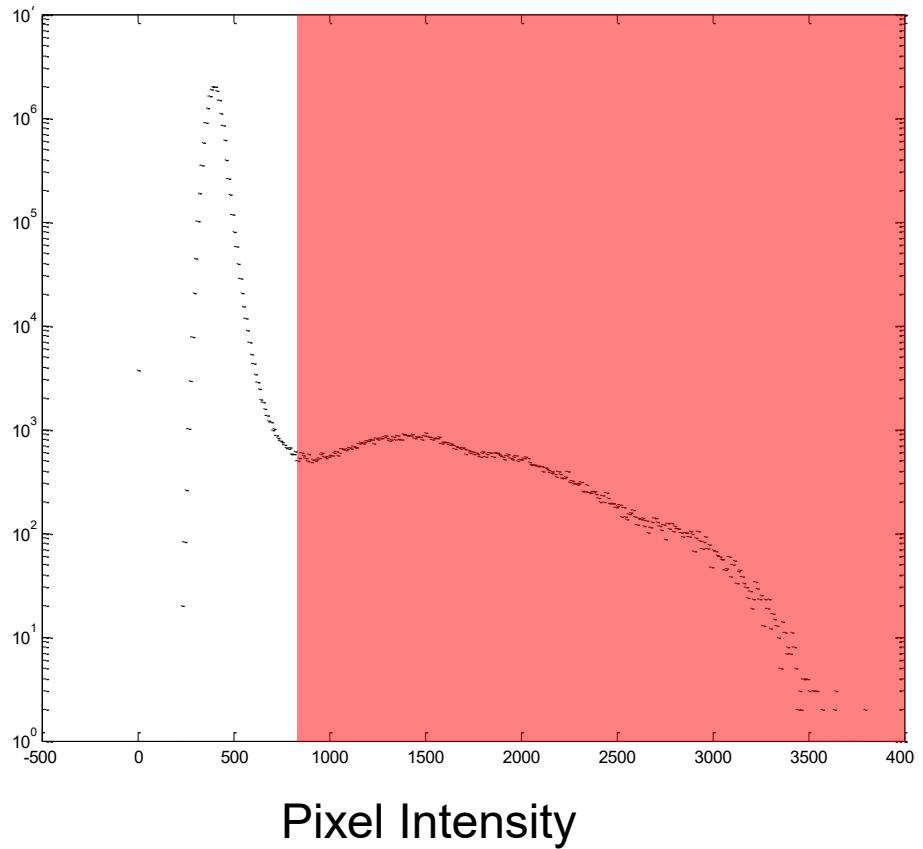
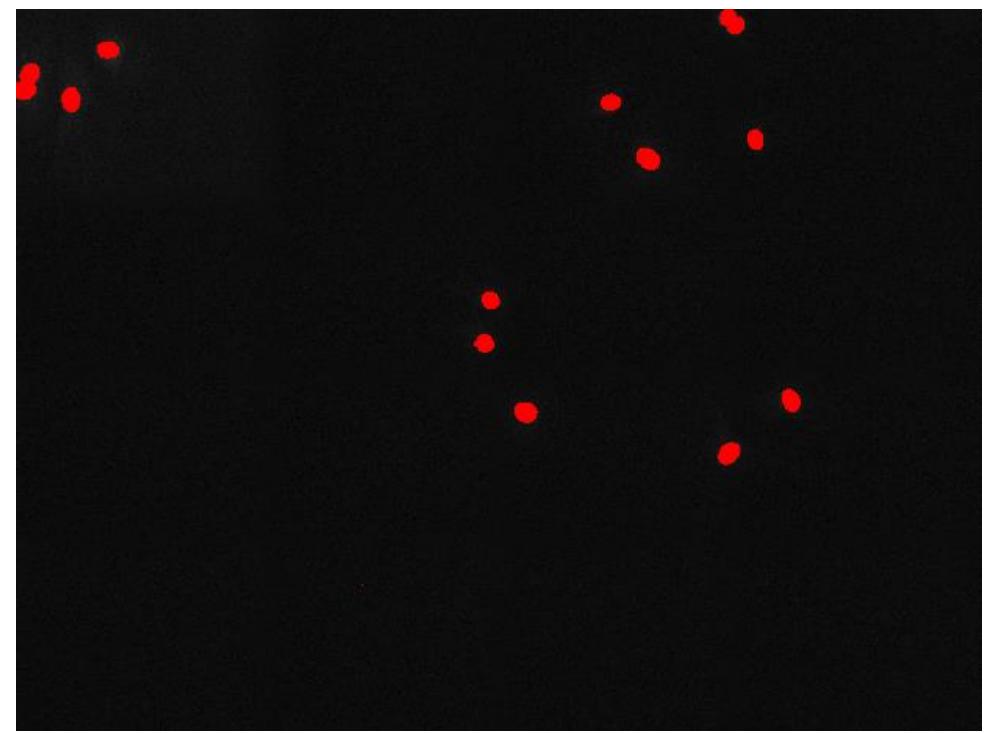


Unsharp masked

# Nonlinear filters

- Can do things like median filtering – replace center pixel by median value within box
- Good for smoothing while maintaining edges

# Thresholding



# One problem with this approach...

- It's biased towards brighter objects
- Ideally would use second channel to independently define objects to measure

# Binary images

- Thresholding gives you a binary image; 1 inside an object, 0 elsewhere
- This can be used to identify objects
- It can also be manipulated

# Binary operations: dilation and erosion

0	0	0	0	0	0	0
0	0	0	1	0	0	0
0	0	1	1	1	0	0
0	0	1	1	1	1	0
0	1	1	1	1	0	0
0	0	1	1	0	0	0
0	0	0	0	0	0	0

→ erode

0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	1	0	0	0
0	0	0	1	1	0	0
0	0	1	1	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0

# Binary operations: dilation and erosion

0	0	0	0	0	0	0
0	0	0	1	0	0	0
0	0	1	1	1	0	0
0	0	1	1	1	1	0
0	1	1	1	1	0	0
0	0	1	1	0	0	0
0	0	0	0	0	0	0

dilate →

0	0	0	1	0	0	0
0	0	1	1	1	0	0
0	1	1	1	1	1	0
0	1	1	1	1	1	1
1	1	1	1	1	1	0
0	1	1	1	1	0	0
0	0	1	1	0	0	0

# Other binary operations

- Sequential erosion and dilation – tends to smooth objects
- Hole filling
- Removing objects at borders

# Further reading

- Lectures available online: [nic.ucsf.edu/edu.html](http://nic.ucsf.edu/edu.html)
- Gonzalez, Woods, and Eddins: Digital Image Processing (Using Matlab)
- Pawley: Handbook of Biological Confocal Microscopy