

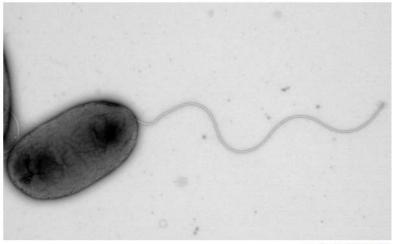
## Important take-home messages

- Raw Data Sacrosanct
- Save uncompressed
- Keep a back-up in a separate location
- Only change in ways valid for quantification
- GIGO: Measurements not useful without well planned experiments with proper controls

There are two types of operations we can do on images to change the way they appear:

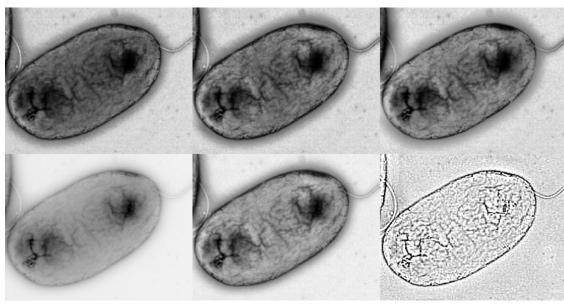
- 1. Things that do not change the raw data.
- 2. Things that do change the raw data.

www.nytimes.com/2013/08/15/science/watching-bacteria-evolve-with-predictable-results.html?\_r=0 
MATTER
Watching Bacteria Evolve, With Predictable Results





By CARL ZIMMER Published: August 15, 2013



Photoshop is good at keeping the underlying raw data under adjustment layers. This is especially good for making figures where different experimental conditions may all have the same adjustments applied and changed for different venues (e.g. printed journal, video projection, or web browser).

You can get PhotoShop from IT and use on campus where you have an Internet connection to the keyserver.

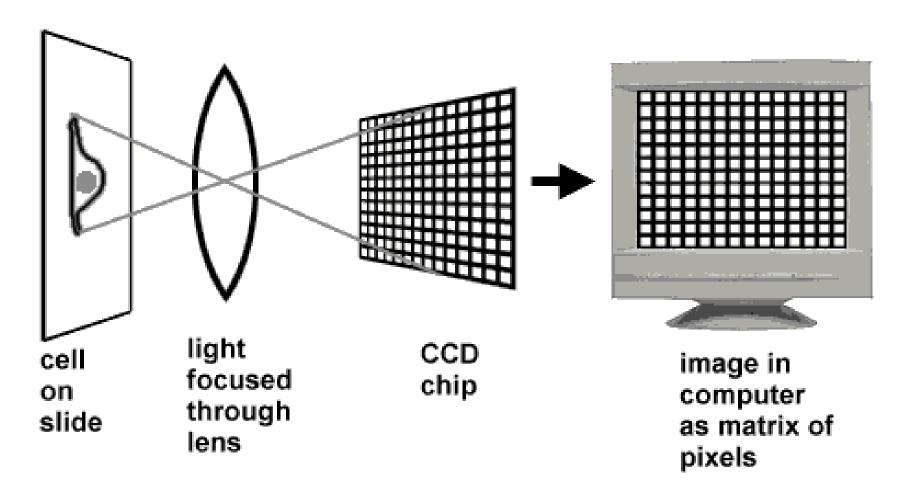
This class uses ImageJ specifically for quantification as well as for examples of mathematical operations of images.

#### Other similar software:

- Image Pro
- Slidebook
- Nikon Elements
- Metamorph
- Volocity
- Imarus
- MatLab
- IDL and lots more...



# What is a digital image?



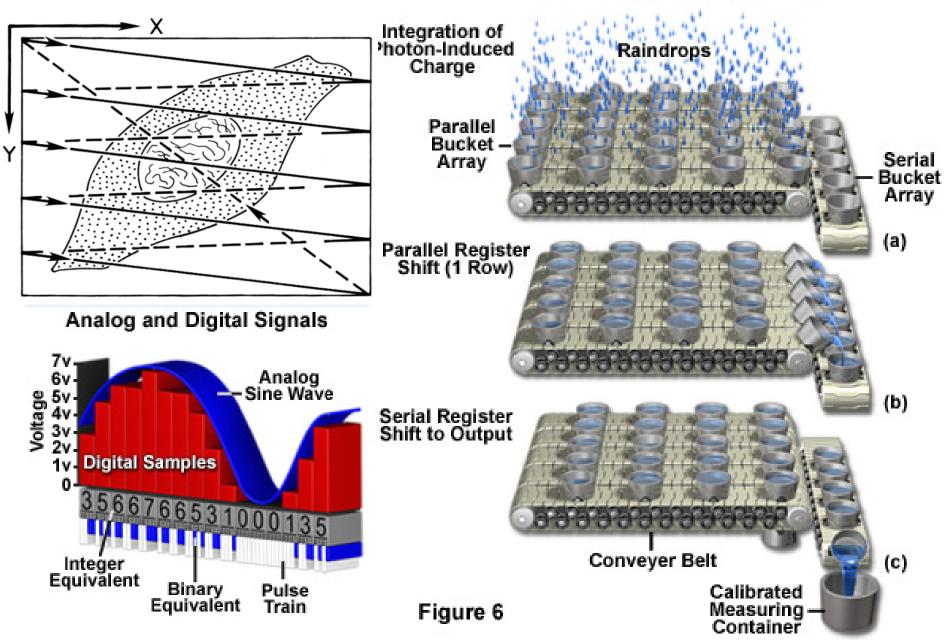
## How do we get images?

Most commonly CCD (or CMOS) camera

Raster scanning device, i.e. laser scanning confocal

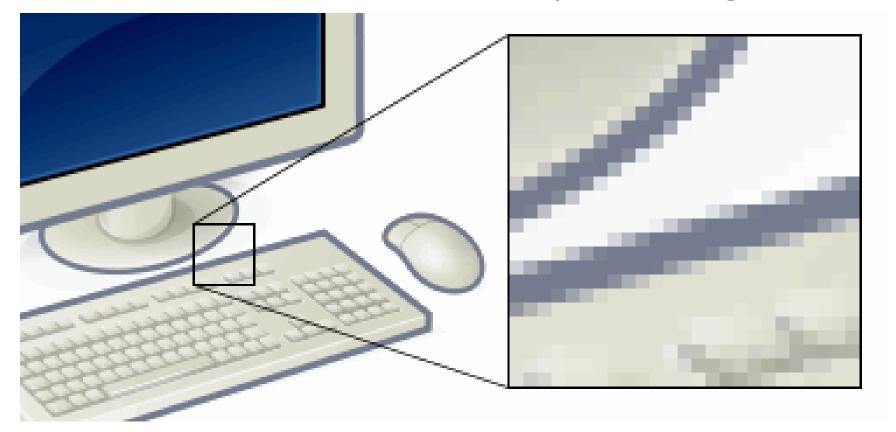
(Film & video are obsolete)

#### Bucket Brigade CCD Analogy



## Pixel: "Picture Element"

## Smallest unit of a 2D computer image



Each pixel is a discrete number which (usually) is intensity or luminosity.

### Examples what the numbers represent:

- amount of a protein at a location
- activity of a protein
- bound or free Ca<sup>++</sup>
- temperature
- population (# of objects at a location)
- decay time of a fluorescent molecule

# Voxel

Each pixel or voxel is a discrete number which (usually) is intensity or luminosity.

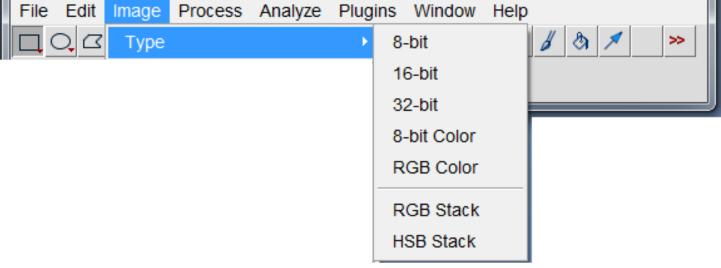
Examples what the numbers represent:

- activity of a protein
- bound or free Ca<sup>++</sup>
- temperature
- population (# of objects at a location)

Each pixel is a discrete number which (usually) is intensity or luminosity.

Typical integers for digital microscopy:

- 0 − 255
- $\bullet$  0 4095
- $\bullet$  0 16383
- 0 − 65535 From black to white or from no signal to saturation.



Each pixel is a discrete number which (usually) is intensity or luminosity.

Typical integers for digital microscopy:

• 8-bit 0 - 255

• 12-bit 0 − 4095

• 14-bit 0 − 16383

• 16-bit 0 – 65535

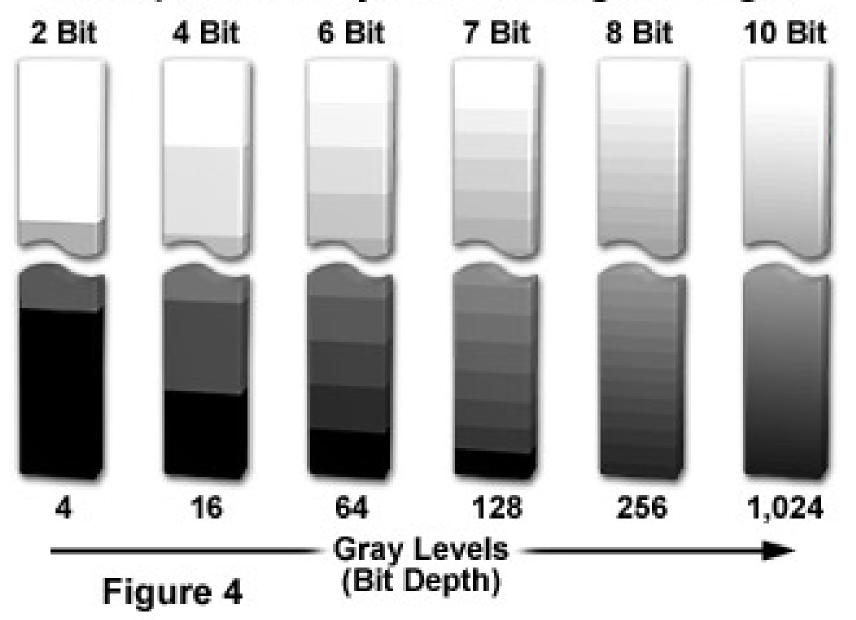
From black to white or

from no signal to

saturation.

We will talk about 32 bit floating point later...

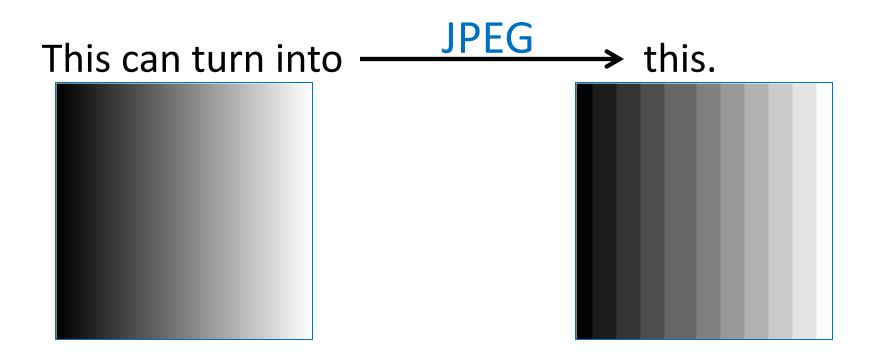
## Bit Depth and Gray Levels in Digital Images



If a scene varies greatly from dark to light, the full dynamic range may need more #s.

saturated
not enough signal

## Don't Compress Raw Data!



To paraphrase Monty Python, Every pixel is sacred.

## **Compression Loses Spatial Information**

(exercise to see this and learn stacks and calculator)

- Open all the JPG images \exercises and demos\JPG compression\
- Image > Stacks > Images to Stack...
- Play through stack, note name in upper left
- Open original TIFF image
- Process > Image Calculator... to see Difference Each pixel (or voxel) has a unique location in X, Y, C, Z, T and may be operated on by a pixel at the same location in another image.

## Common Color Spaces

- RGB (Red Green Blue)
- HSV (Hue Saturation Value)
- Overlayed channels (grayscale images with LUTs (Look Up Tables) assigned
- CMYK (Cyan Magenta Yellow Black)

# What a color picture is, at least with your camera.



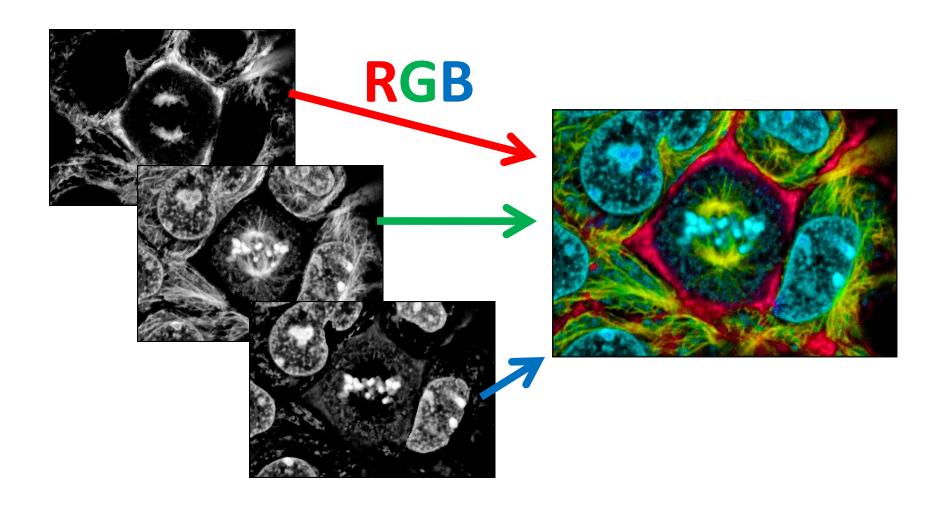
RED GREEN BLUE



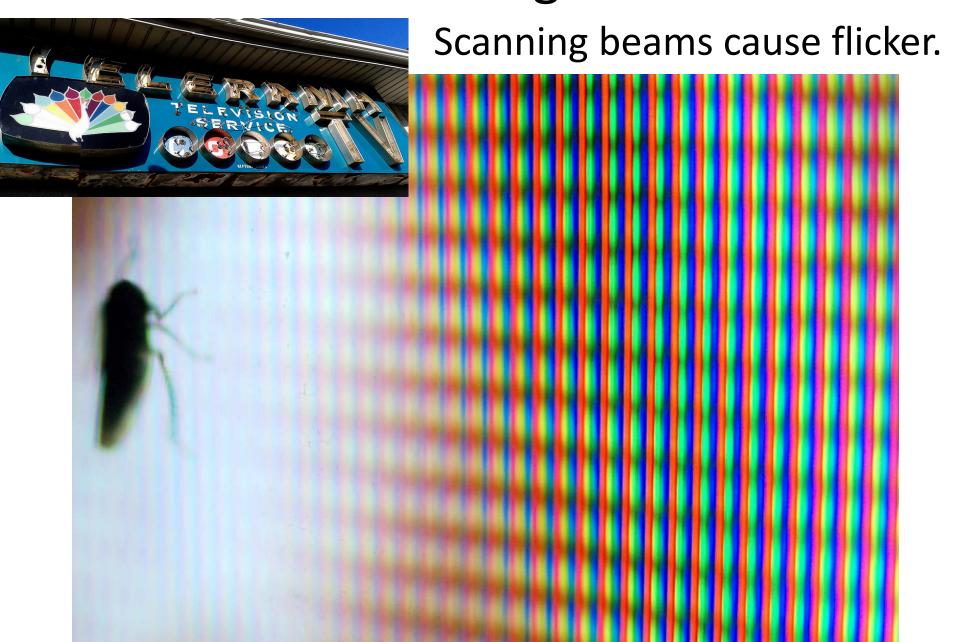


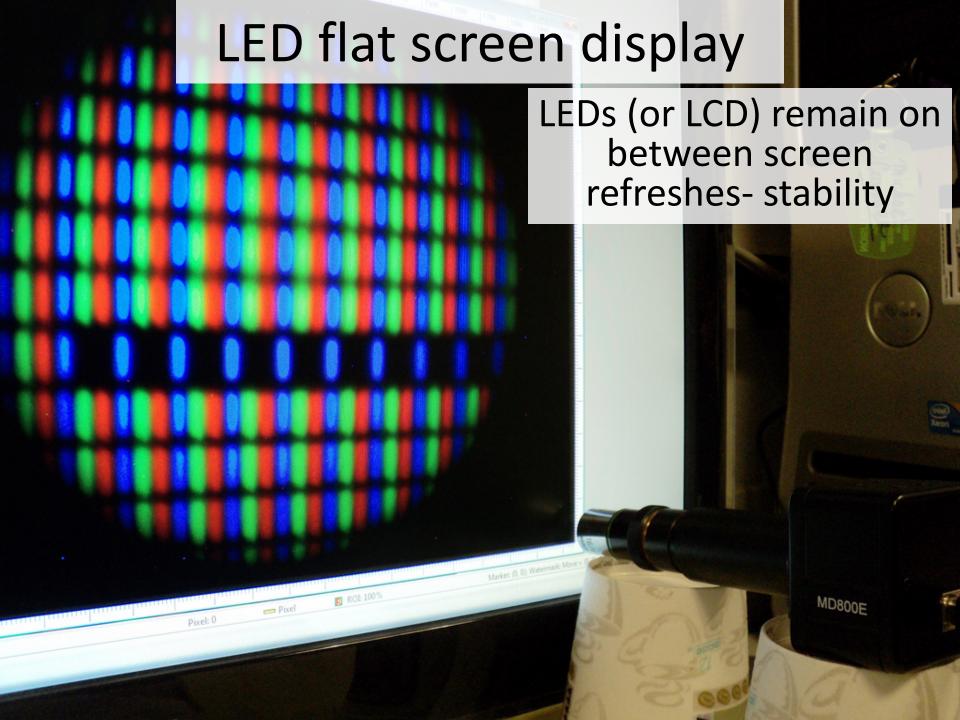


# Color images are (usually) channel mixtures of intensities

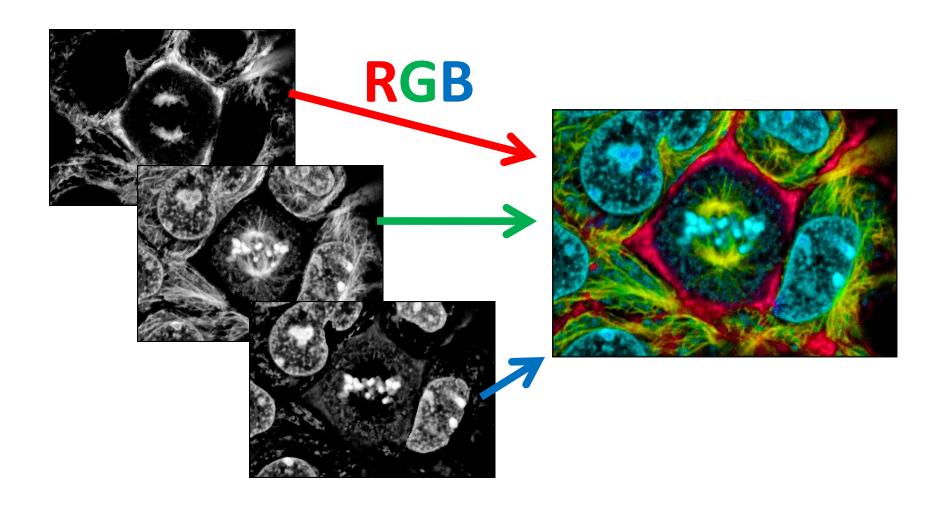


# Color TV with high res comb

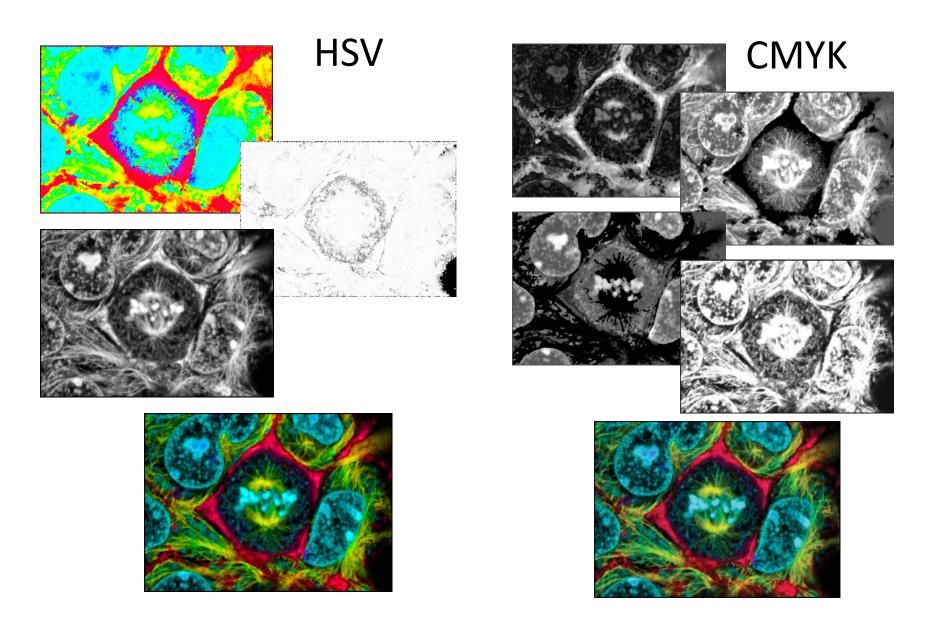




# Color images are (usually) channel mixtures of intensities



## **Less Common**



## Channels

• LUTs (Look Up Tables)

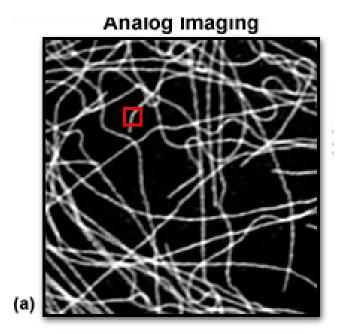
## Exercise with color spaces

- Import confocal images as multichannel (hyperstack)
- Change colors per channel
- Manipulate and measure in channels
- Convert to RGB
- Make an instant multi-panel figure using Montage command
- Add scale bar compare in multichannel and RGB modes

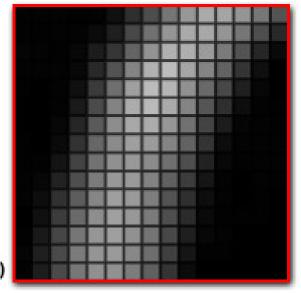
# Images are 2D matrices of numbers

The matrix itself is just numbers. To know more about these numbers, there must be additional tags which are usually stored in a header.

#### Images are 2D matrices of numbers



#### **Digital Sampling**



#### **Pixel Quantization**

2	3	3	2	3	16	47	78	118	152	170	165	149	119	83
3	2:	3	3	11	3:1	72	113	150	170	168	149	118	78	41
5	3	2	7	28	51	100	147	177	175	155	123	82	38	10
3	3	3	8	41	83	132	177	188	165	132	98	52	15	0
2	0	3	16	59	111	160	188	181	145	105	72	34	7	0
0	0	3	26	75	131	17.2	186	170	132	85	47	13	3	2
0	0	7	38	90	141	168	172	150	113	67	28	5	3	2
2	0	10	51	983	139	159	155	132	98	52	13	3	3	3
2	2	20	67	114	144	152	139	111	78	34	8	2	3	3
3	7	34	78	123	152	954	131	96	60	21	3	0	3	2
3	13	49	93-	134	159	154	126	83	47	13	3.	0	2	0
0	16	60	106	142	160	152	123	78	43	10	2	2	2	.0
2	25	70	118	150	159	145	116	72	34	7	3	0	0	0
5	31	74	123	149	154	136	110	64	21	2	2	0	0	0
3	29	74	119	144	145	124	96	52	11	0	0	0	0	0

(c)

To know more about these numbers, there must be additional tags which are usually stored in a

							•							
¹²³hëa¹der. ¹	109	51	218	221	147	195	13	45	153	253	88			
ricaaci.	244	204	106	0	102	177	86	135	152	244	215			
	105	98	247	213	79	18	241	208	58	<b>1</b> 36	91			
	10	84	196	134	194	76	211	191	9	110	79			
	83	<b>1</b> 64	147	<b>114</b>	213	92	184	28	38	213	209			
	217	88	250	128	118	235	54	<b>164</b>	137	167	26			
<del></del> 1	103	38			146	31	78	220	136	163	174			
There may	V100	15	<b>O</b> 20	<u> </u>	<b>a</b>	<u>d6</u>	81	36	<b>S S 9</b>	252	51			
batuuaaa	107	176	252	34	93	<b>4</b> 9	208	189	80	4	157			
between	26	87	22	78	43	36	214	249	107	250	17			
	164	127	25	116	60	155	46	134	125	128	30	1		
	213	39	180	56	121	19	6	60	132	77	85	5	7	

# Exercise: Look at metadata in our confocal images.

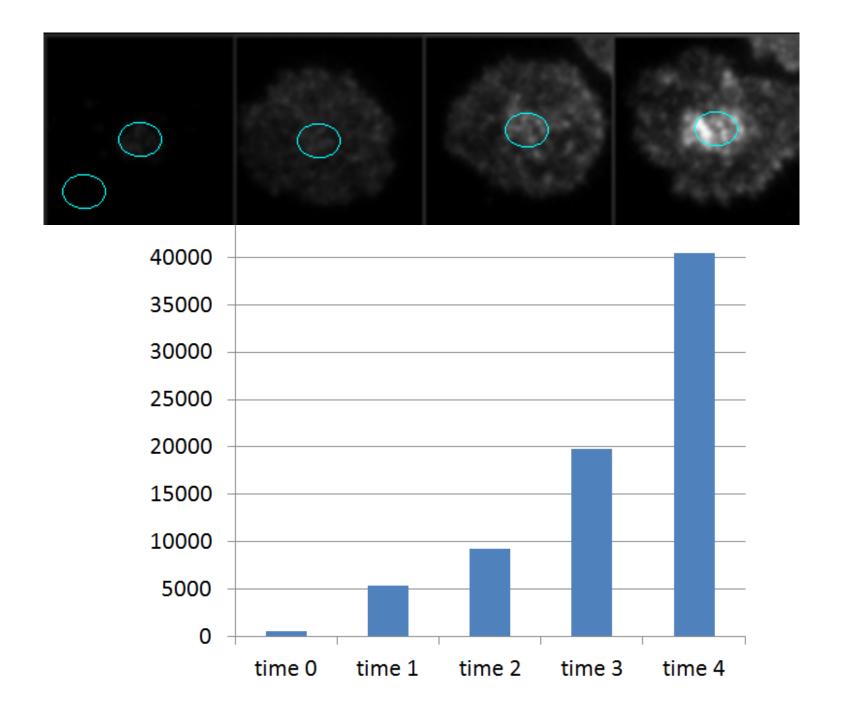
- Using LOCI plugins, import native format files with show metadata and XLM boxes clicked.
- Image > Properties...
- Image > Show Info...

## Images are 2D matrices of numbers

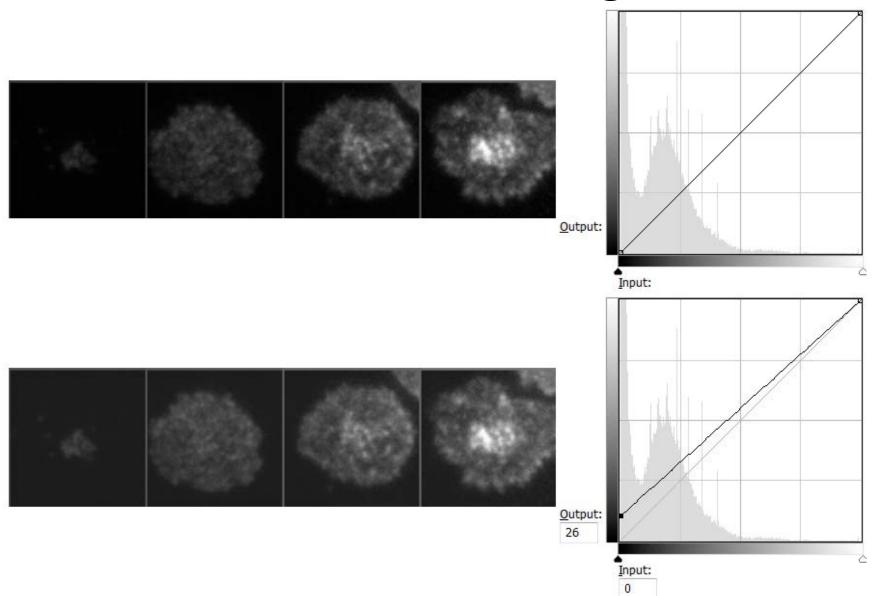
We may quantify these numbers.

Manipulate the numbers.

Operations that change the way images look without changing the raw data and operations that change the raw data.



# We have difficulty seeing subtleties in the darks of images

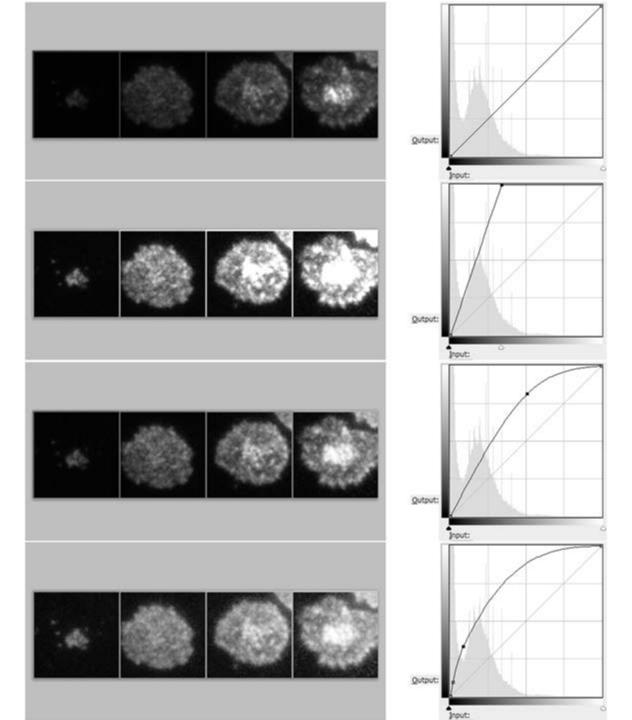


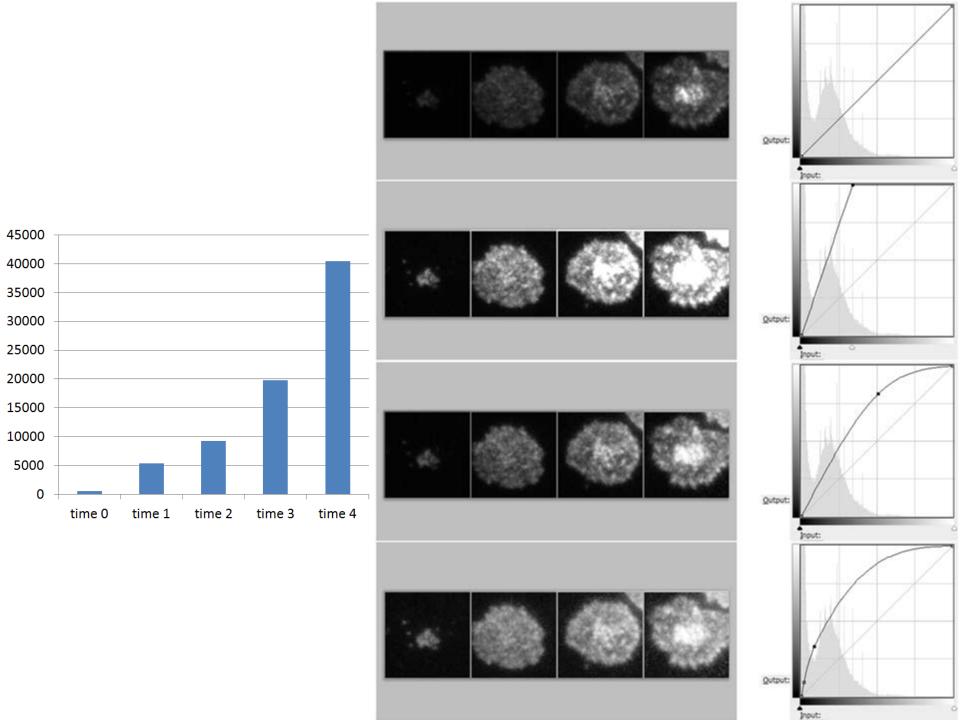
#### What is gamma?

It is a non-linear contrast adjustment.

#### We use it

- when the dynamic range of data is too large to display linearly;
- to selectively boost contrast in a narrow range;
- because we (our eyes and brains) do not discriminate darks and lights linearly.



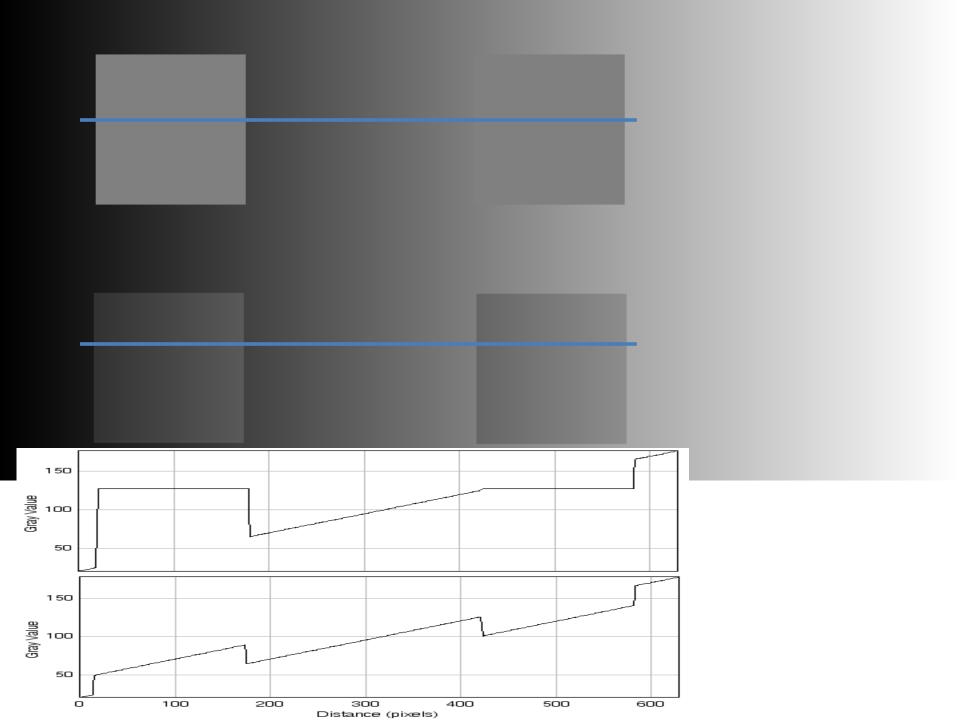


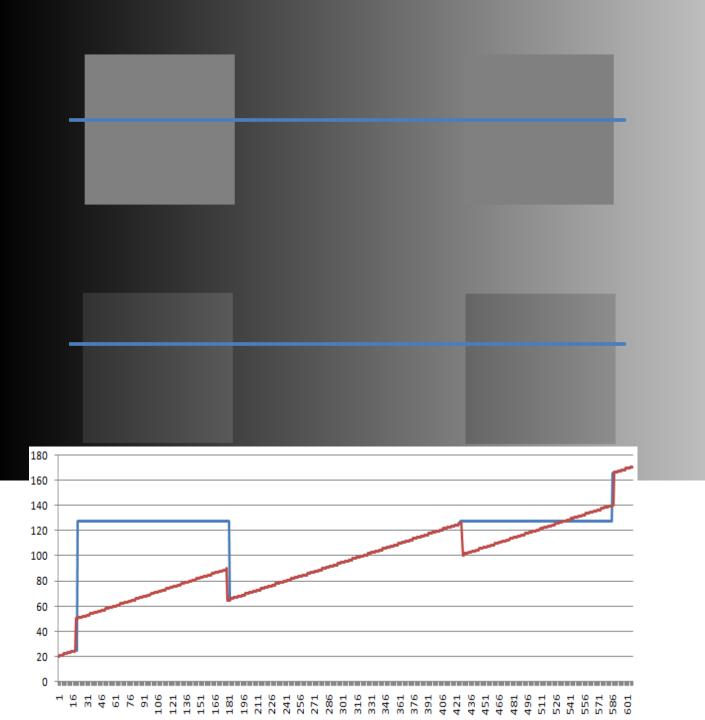
# Importance of Computer Image Analysis Because Often We Cannot Trust Our Eyes



#### Are these squares

- 1.) the same or different intensities and
- 2.) are they uniform or shifting (gradient) intensities?





#### Most Quantification of Intensity

Fluorescence intensity usually linear relationship to protein or other chemical.

Absorption intensity usually inverse log relationship and very sensitive to thickness of sample.

Most measurements are of fluorescence combined with area.

#### Area

Count the number of pixels.

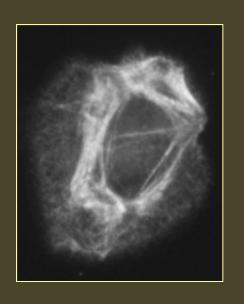
(In old days, cut out from photos and weighed or manually counted with overlaid grid.)

#### Intensity

Photometer in eyepiece of microscope.

Masked portions of image.

## Quantification of mass and concentration in cells by fluorescent microscopy and computer analysis



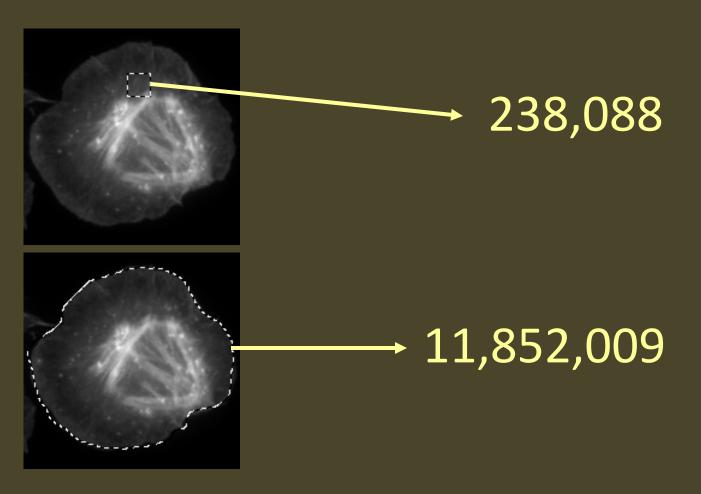
= 555

# There are two types of measurements we get from intensity information.

- 1. Mass (total amount of a species in a given area or volume)
- 2. Concentration (amount of mass per unit area or volume)

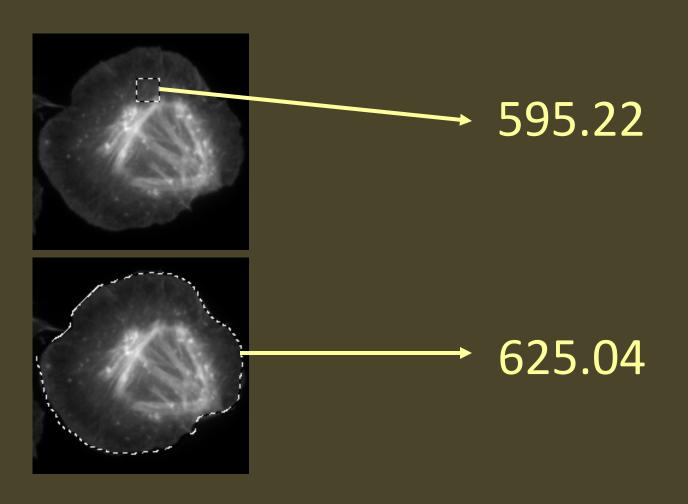
#### Mass

Mass = integrated intensity = sum of all pixel values in a given area



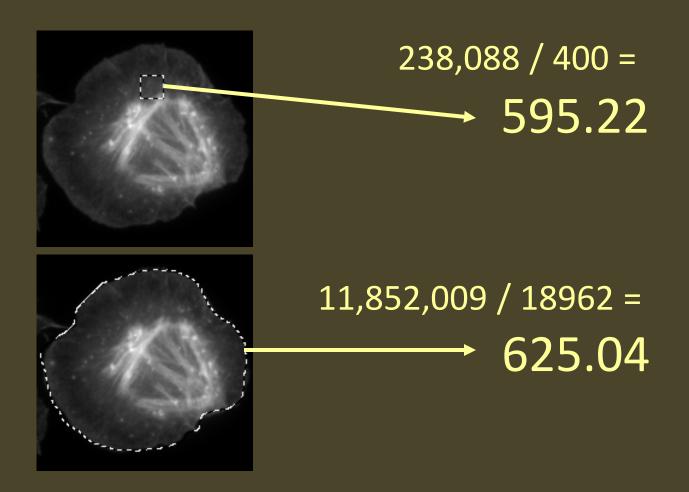
#### Concentration

Concentration = mass per unit area (# of pixels) = mean mass



#### Concentration

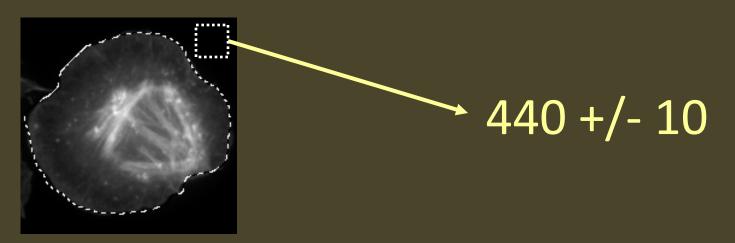
Concentration = mass per unit area (# of pixels) = average mass



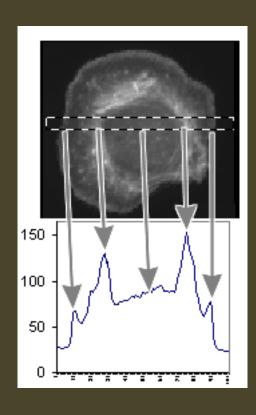
#### Whoops, we skipped a step!

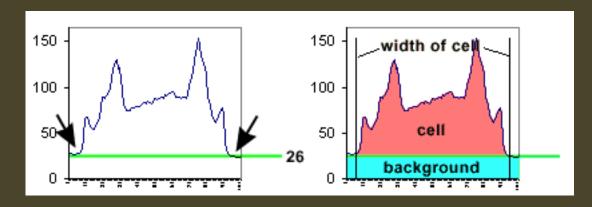
The previous measurements assumed an ideal situation where everything that was non-cell was pure black or zero.

But when we collect images, this isn't true!

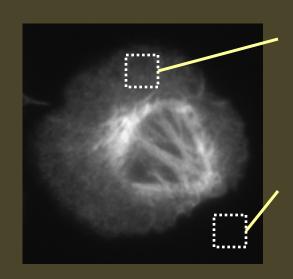


#### Area under the curve





#### Subtract the background

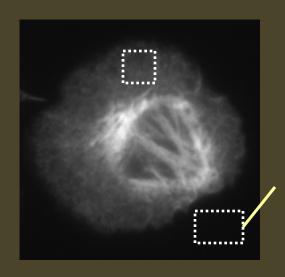


Let's say that this box is 10 X 10 pixels. Let's say its integrated intensity = 100,000.

Let's say that this box is 10 X 10 pixels. Let's say its integrated intensity = 44,000.

The mass of the box within the cell is really 100,000 - 44,000 = 56,000

#### Subtract the background



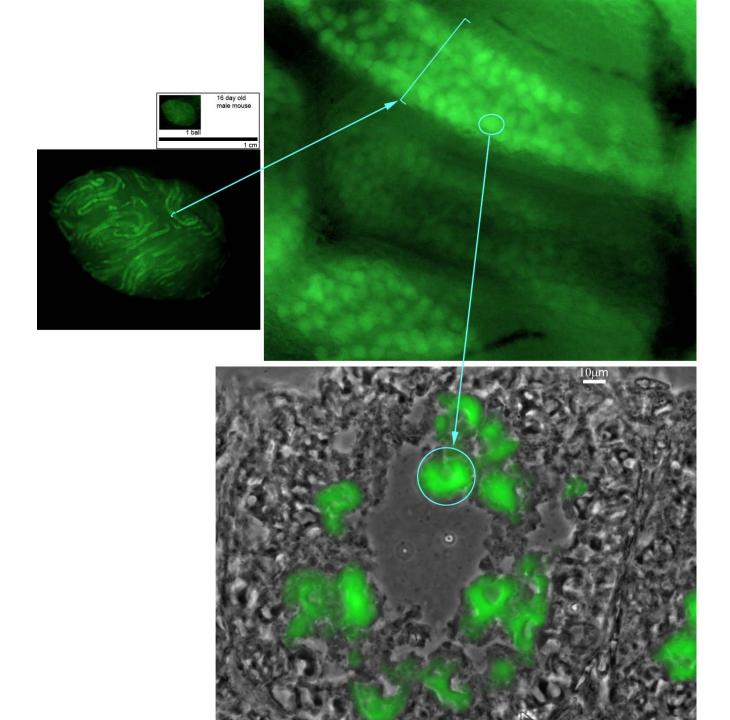
Because the background area may vary in size, it it simpler to always do the math based on the concentration.

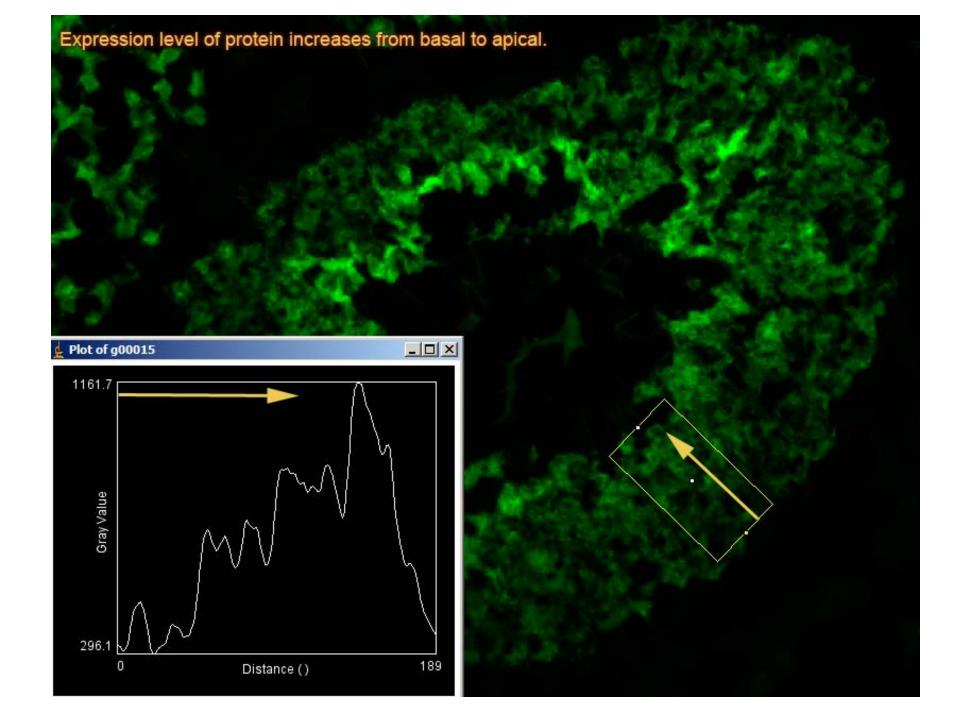
The size or shape of this box doesn't matter.

All we care about is that its

Mass / area = 440.

WT GFP+ 1<sub>cm</sub>

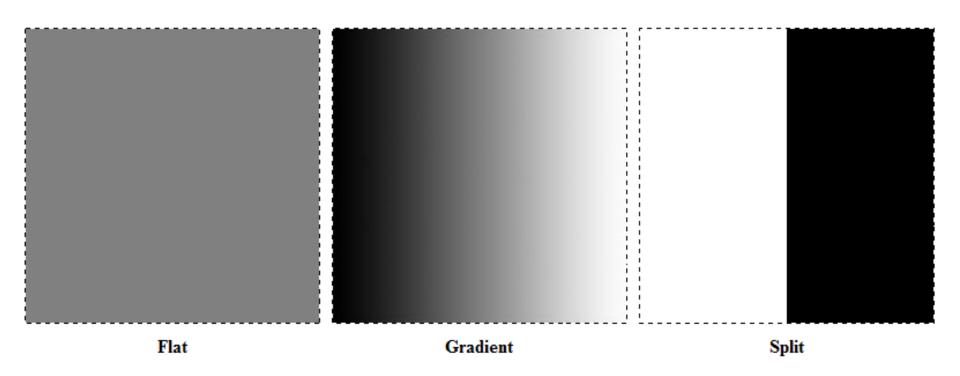




### Summary of Basic Intensity Measurements

- Background
- Range of intensities
- Maximum value not saturated

#### Distribution of Intensities

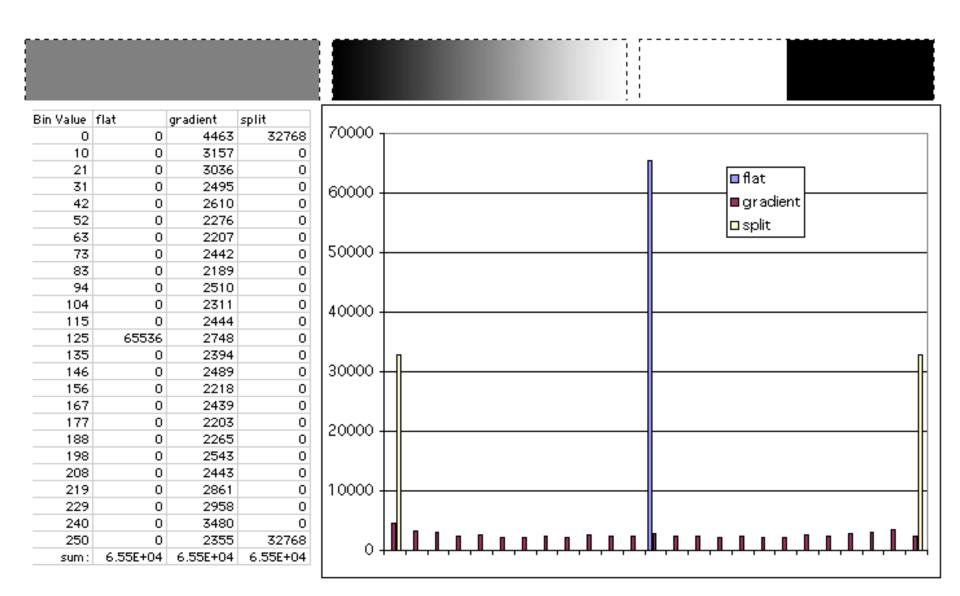


Images are 8 bit grayscale (values from 0 to 255). All three images have mean values of 128.

#### Distribution of Intensities

1		meuzi	urement re	2	
		Meas.#	Mean	Stand.Dev.	
	Ø	1.00	128.00	0	
	1	2.00	127.00	79.18	
	2	3.00	127.50	127.50	

#### Distribution of Intensities



### Automated Segmentation and Measuring

#### Noise

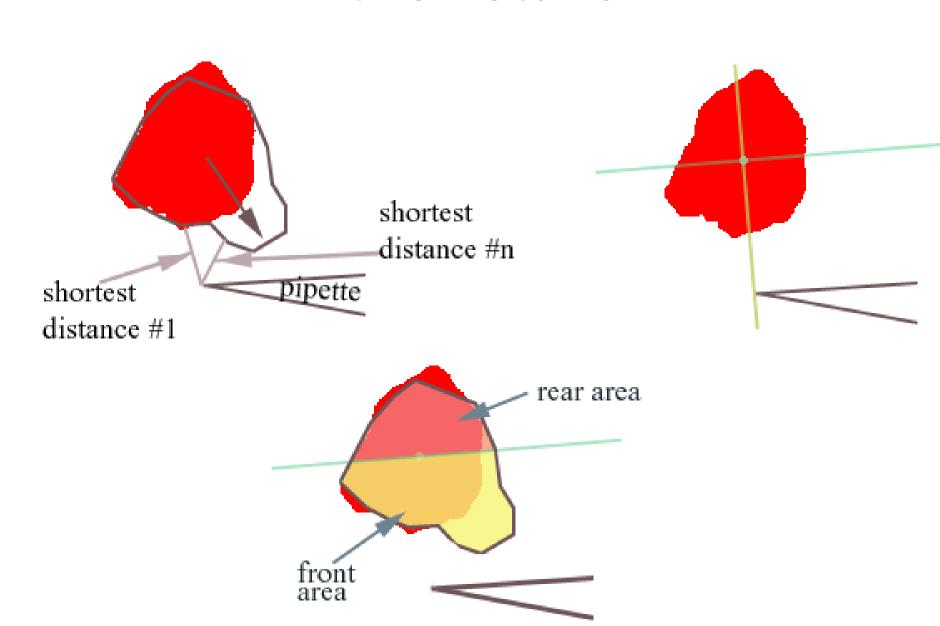
#### Flatfield correction

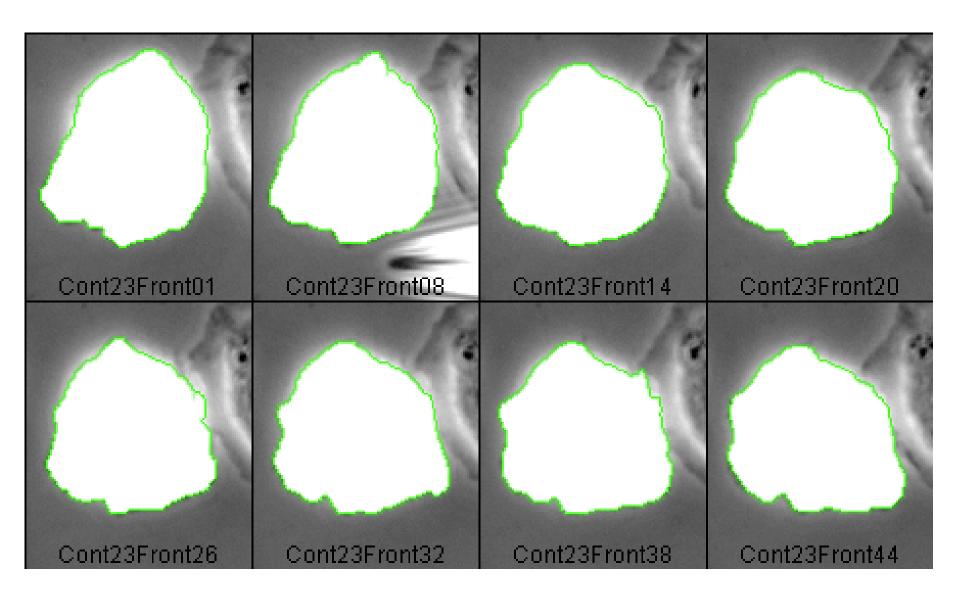
- Adjacent image
- Out of focus image
- Calculated or blurred image
   (does not eliminate high frequency noise, e.g.
   dirt in lightpath)
- Estimate with high pass filter (accentuates high frequency noise)

#### Measuring Change in Shape

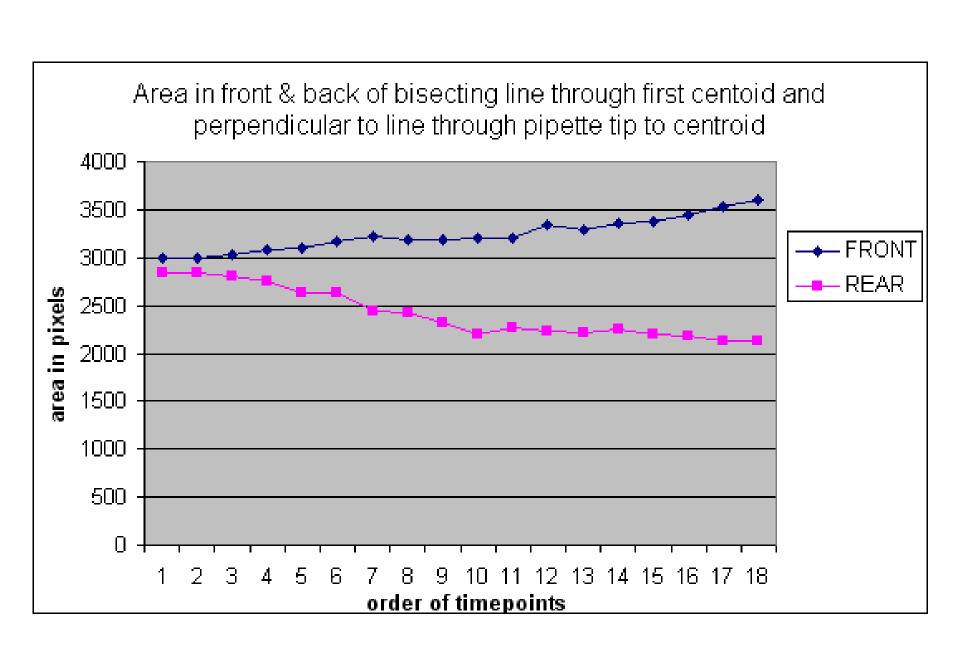
- Movement in a particular direction
- Ratio area to perimeter
- Ratio min and max axes

#### Chemotaxis



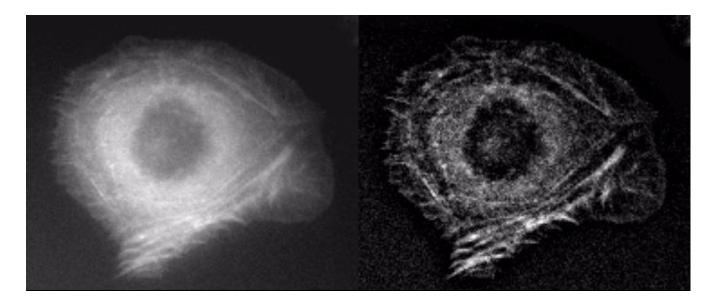


FRONT A	rea :	Χ	Υ	REAR	Area	Χ	Υ	FRONT/REAR
1	2994	107	111	1	2836	106	71	1.055712271
2	2994	107	111	2	2836	106	71	1.055712271
3	3024	108	111	3	2808	105	72	1.076923077
4	3078	108	111	4	2757	105	72	1.116430903
5	3092	109	111	5	2638	104	74	1.172100076
6	3168	109	112	6	2631	103	74	1.204104903
7	3221	110	112	7	2440	102	75	1.320081967
8	3187	110	112	8	2418	102	75	1.318031431
9	3193	109	112	9	2318	101	76	1.377480587
10	3209	110	112	10	2207	100	76	1.454009968
11	3205	111	112	11	2267	100	76	1.413762682
12	3341	111	112	12	2239	100	76	1.492184011
13	3290	111	112	13	2216	99	77	1.48465704
14	3355	111	113	14	2259	99	- 77	1.485170429
15	3382	111	113	15	2196	98	- 77	1.54007286
16	3440	113	113	16	2174	98	78	1.582336707
17	3541	113	113	17	2135	97	78	1.658548009
18	3606	115	113	18	2123	98	78	1.698539802



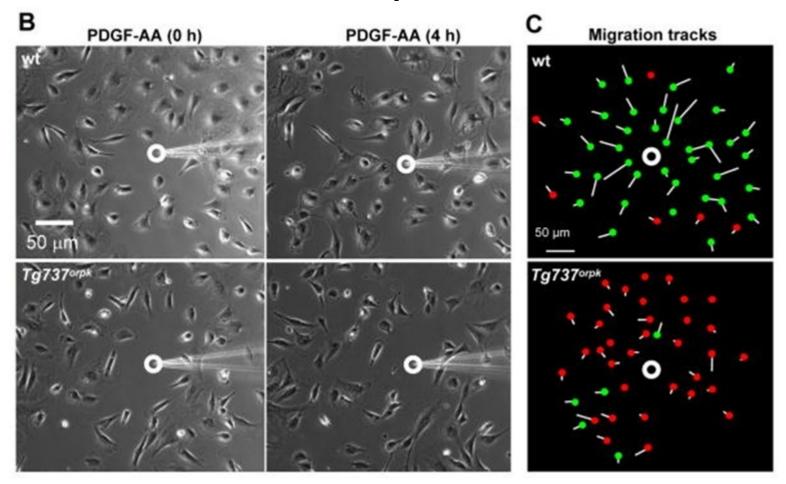
4-07-94
15:40:15

EGF Upshift Fusion



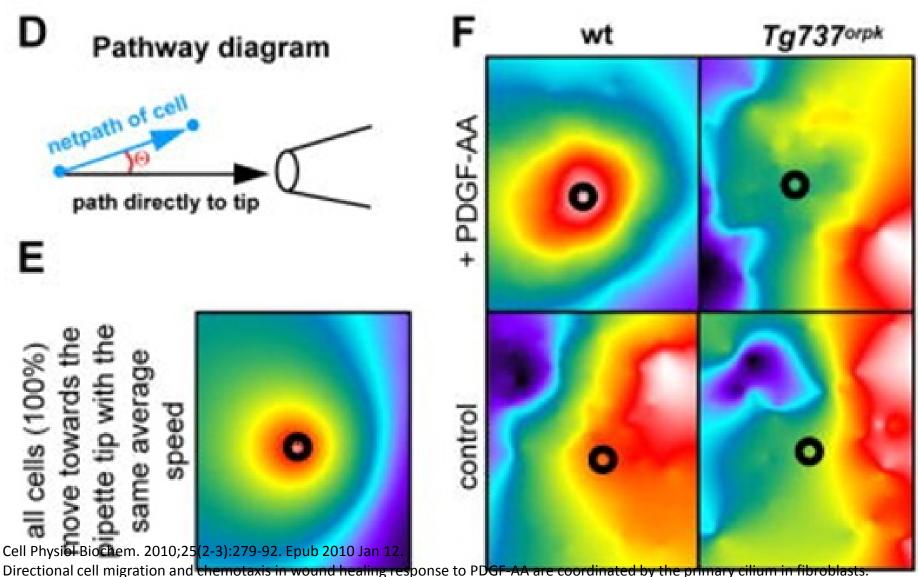
### EGF Kinetics graph here

### Chemotaxis quantification



Cell Physiol Biochem. 2010;25(2-3):279-92. Epub 2010 Jan 12.

Directional cell migration and chemotaxis in wound healing response to PDGF-AA are coordinated by the primary cilium in fibroblasts. Schneider L, Cammer M, Lehman J, Nielsen SK, Guerra CF, Veland IR, Stock C, Hoffmann EK, Yoder BK, Schwab A, Satir P, Christensen ST.

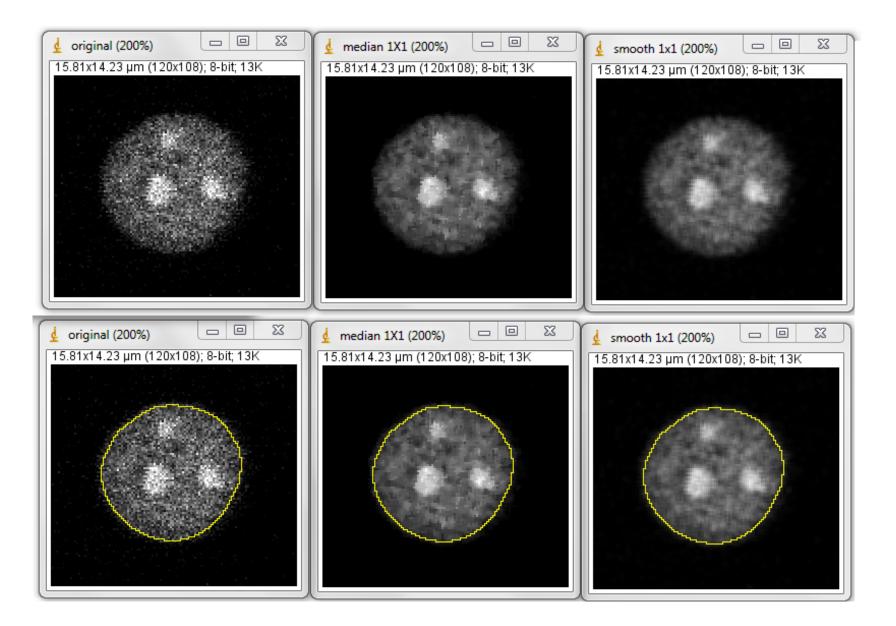


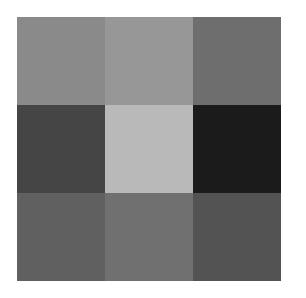
Schneider L, Cammer M, Lehman J, Nielsen SK, Guerra CF, Veland IR, Stock C, Hoffmann EK, Yoder BK, Schwab A, Satir P, Christensen ST.

#### Important take-home messages

- Raw Data Sacrosanct
- Save uncompressed
- Keep a back-up in a separate location
- Do not change contrast
- Do not resize
- Don't change in other ways
- GIGO: Measurements not useful without well planned experiments with proper controls

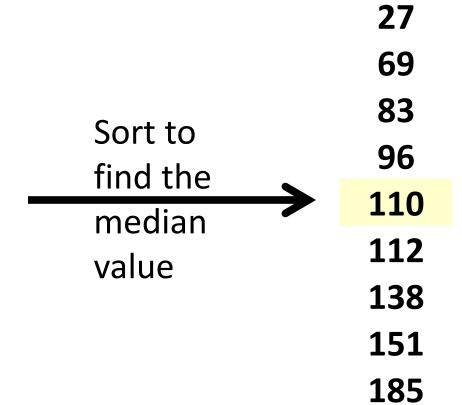
#### Two simple ways to filter noise





138	151	110
69	185	27
96	112	83

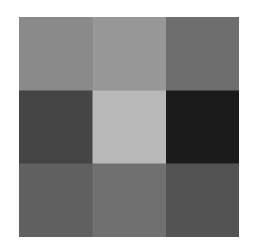
138	151	110
69	185	27
96	112	83

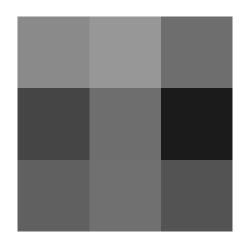


138	151	110
69	185	27
96	112	83

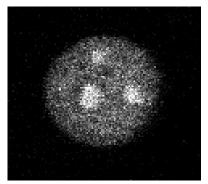
Replace with median value

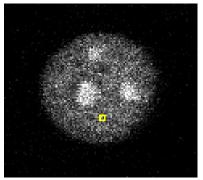
138	151	110
69	110	27
96	112	83

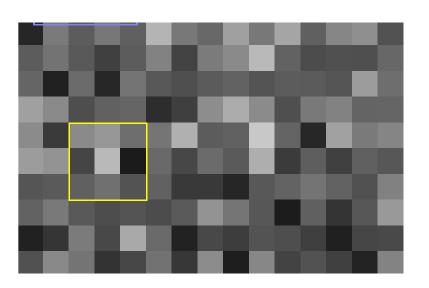




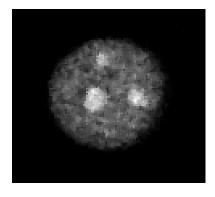
#### Original

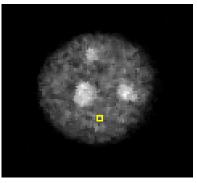


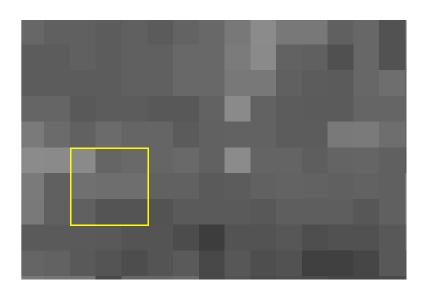




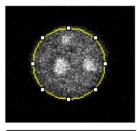
**Median 3X3** 

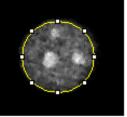


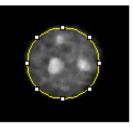


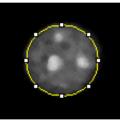


# Is this valid for intensity quantifications?



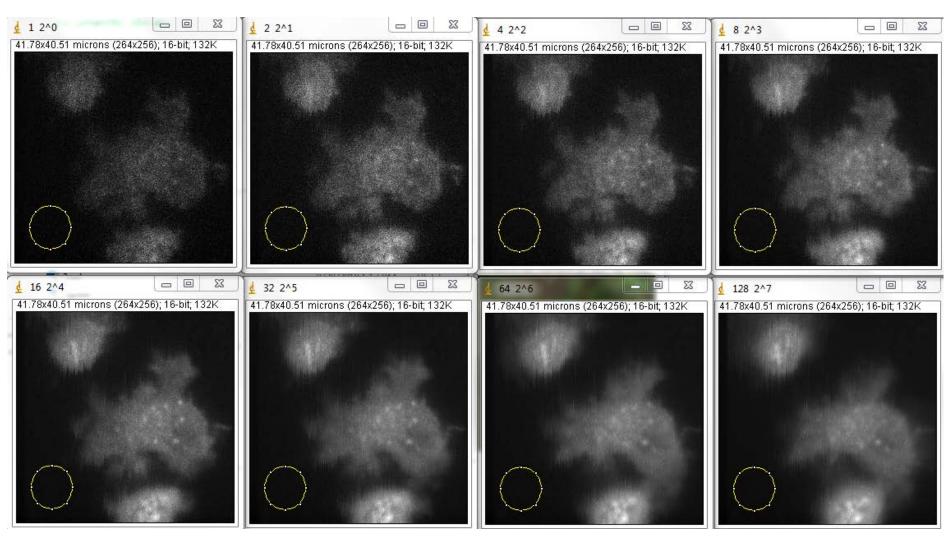


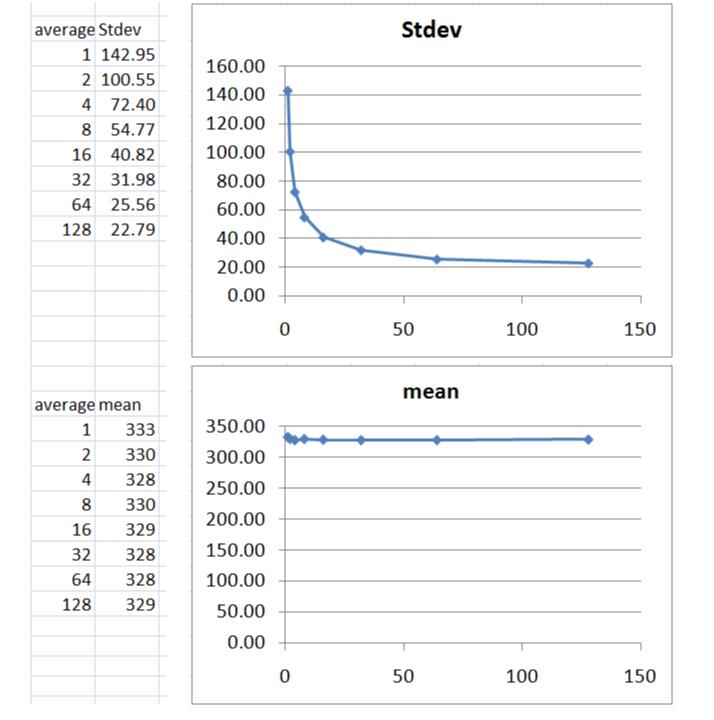




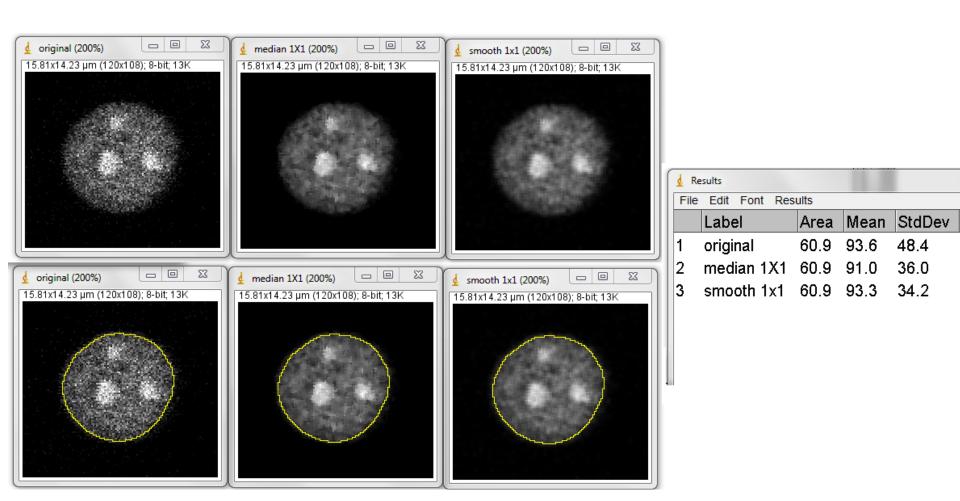
	Label	Area	Mean	StdDev
1	ORIGINAL	60.5	93.5	48.7
2	median 3X3	60.5	90.9	36.3
3	median 5X5	60.5	90.1	33.9
4	median 7X7	60.5	89.3	32.4

### Averaging serial frames

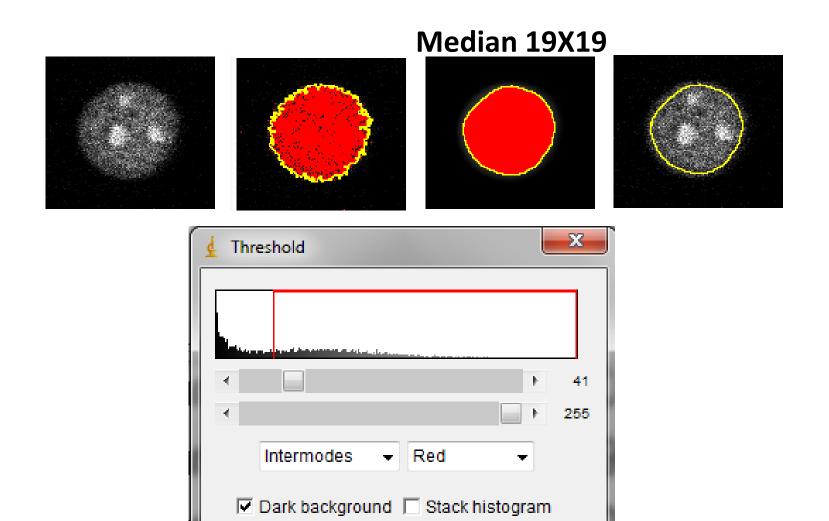




### Two simple ways to filter noise

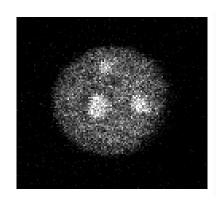


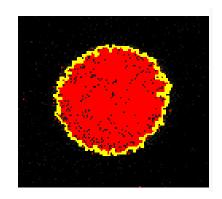
# Median filtering good for automated feature detection by intensity

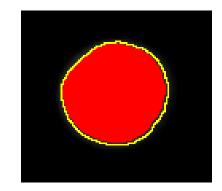


# Median good for removing noise for spatial measurements

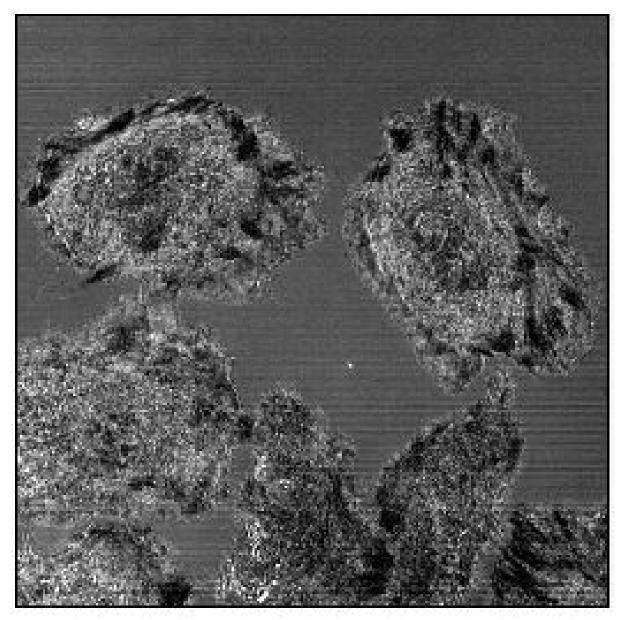
	Label	Area	Perim.	Circ.	Round
1	ORIGINAL	61.6	56.7	0.2	0.9
2	median	57.8	28.4	0.9	0.9

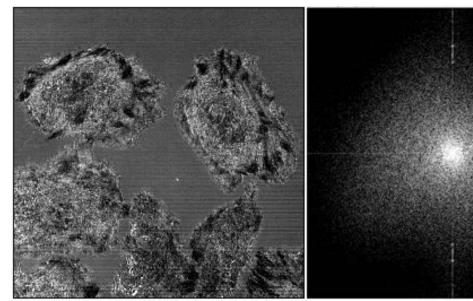






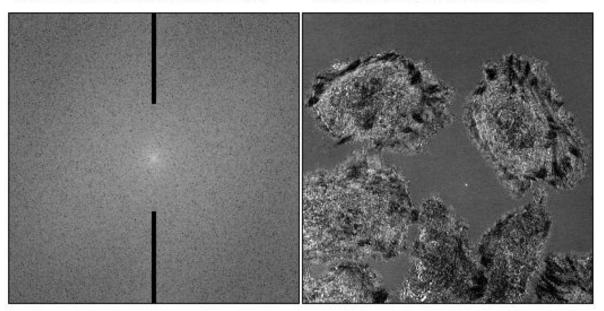
### Filtering periodic noise





The original image. Reflectance mode of the confocal using the 458 nm line of an Ar laser. Note the horizontal lines.

The power spectrum calculated by ImageJ, contrast enhanced to show the bright spots that represent the X axis fluctuation.

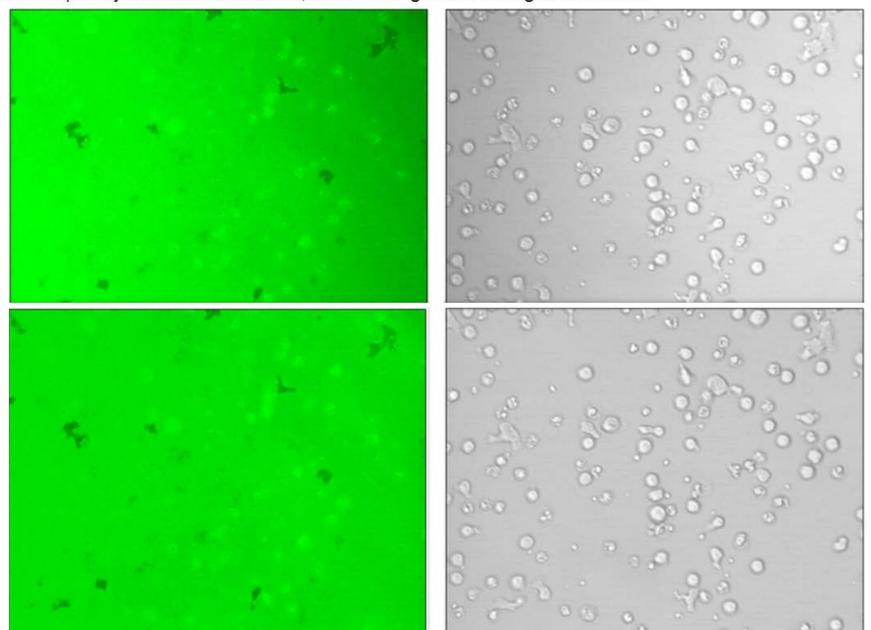


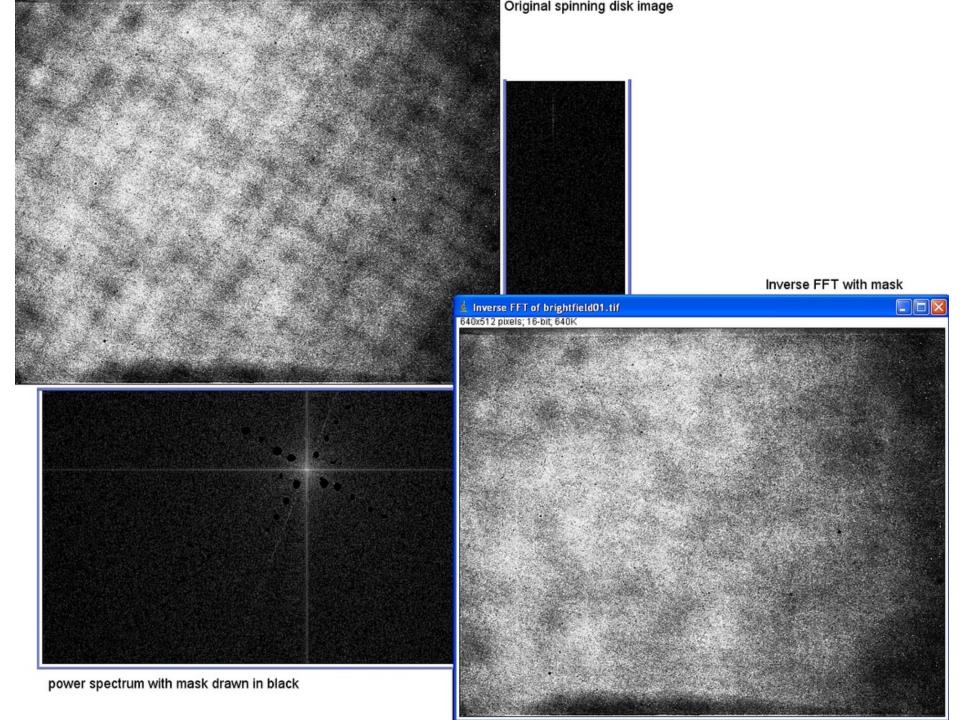
The power spectrum with masks drawn on it.

The inverse transform applying the masks.

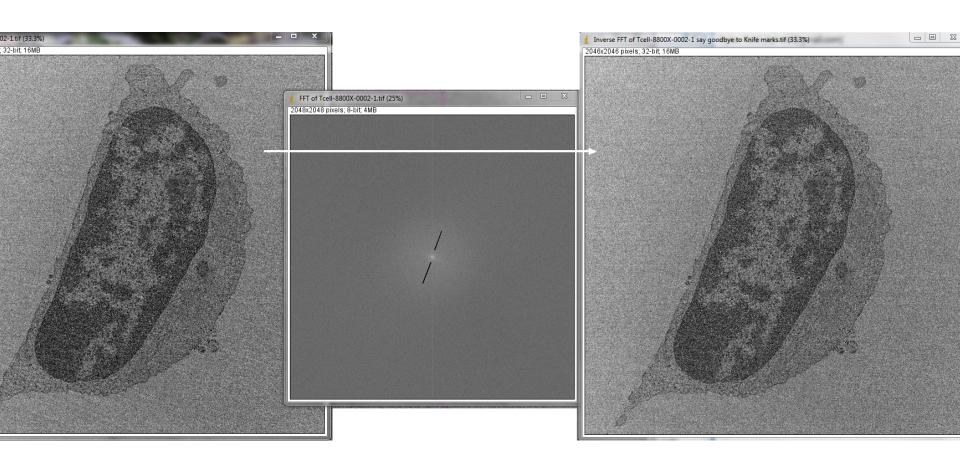


Here is an example where the raw data have uneven illumination from left to right. The raw data are 512X512 pixels. By applying a high pass filter from 128 to 0 pixels, the low frequency variation is removed; the resulting bottom images are flatter.

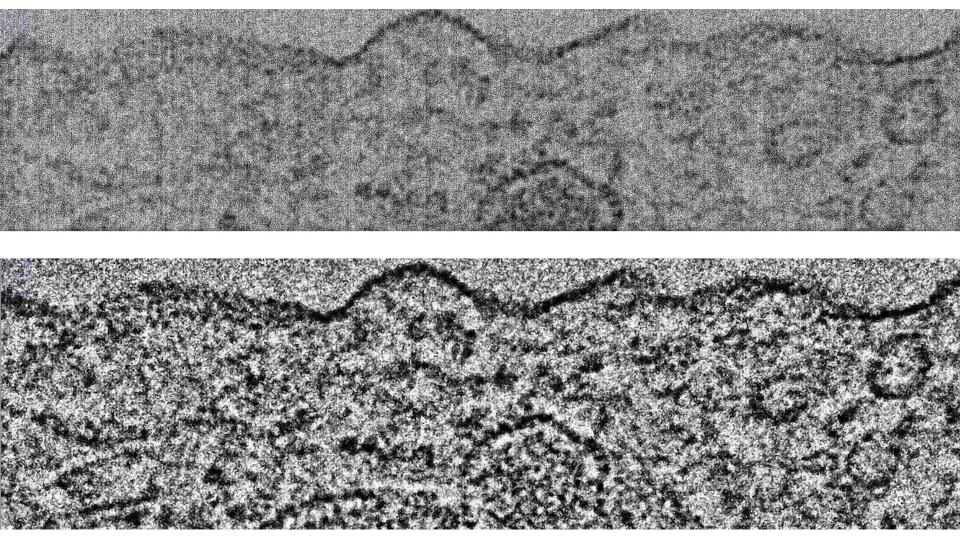




### Transmission Electron Micrograph Knife Marks Minimized



### Transmission Electron Micrograph Enhanced



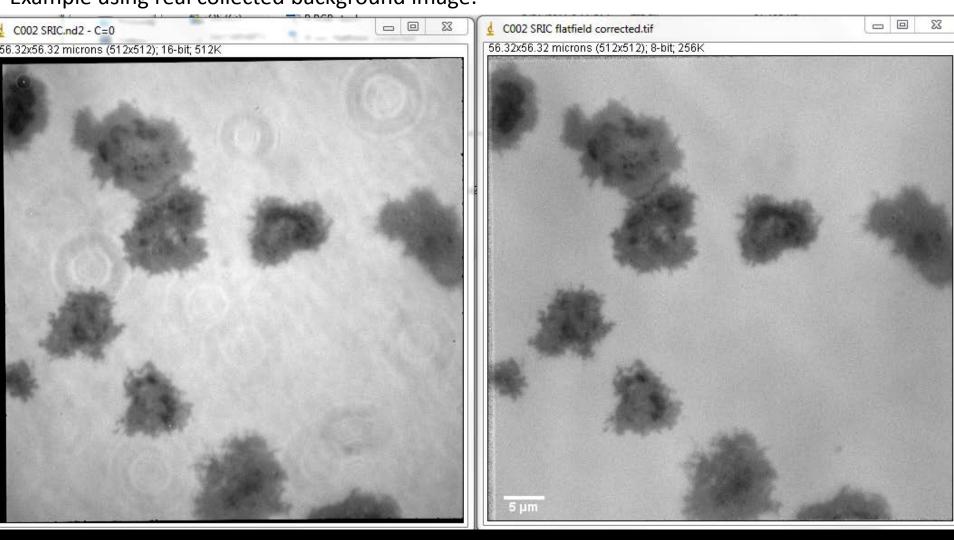
Tcell-66000X-0009 fftcustom outliers1x1dark median1X1 CLAHE144-245-3.tif

#### Flatfield Correction

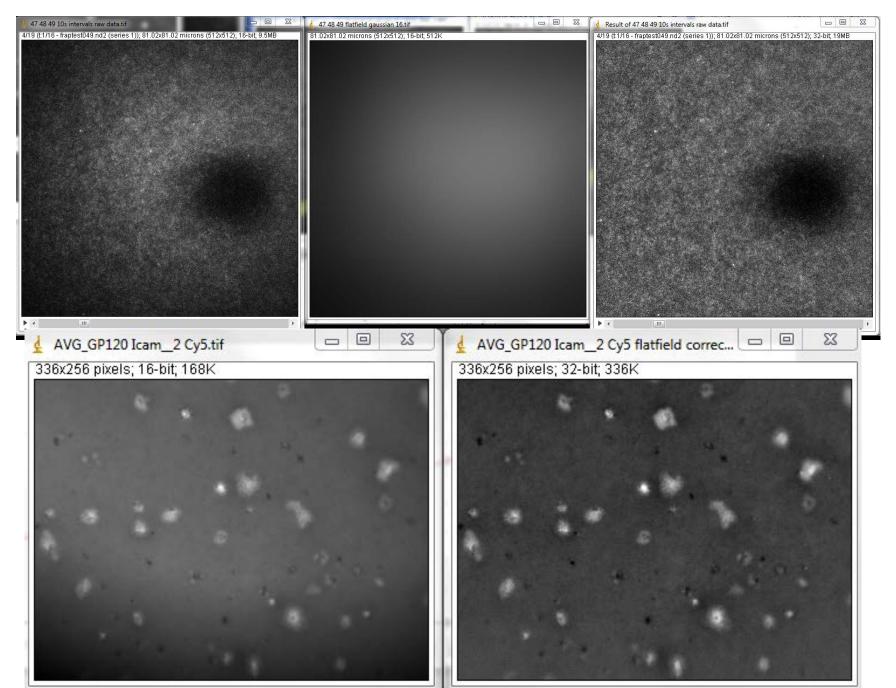
- What is it?
- Why do you want to use it?
  - Remove dirt in optics.
  - Mask periodic noise.
  - Consistent intensity measurements across field.
  - Increase contrast.
- Examples & how to do it
  - To remove dirt, divide by high frequency low noise image.
  - To remove broad uneven lighting, divide by low frequency image.

### To remove fine detail, divide by low noise image.

Example using real collected background image:

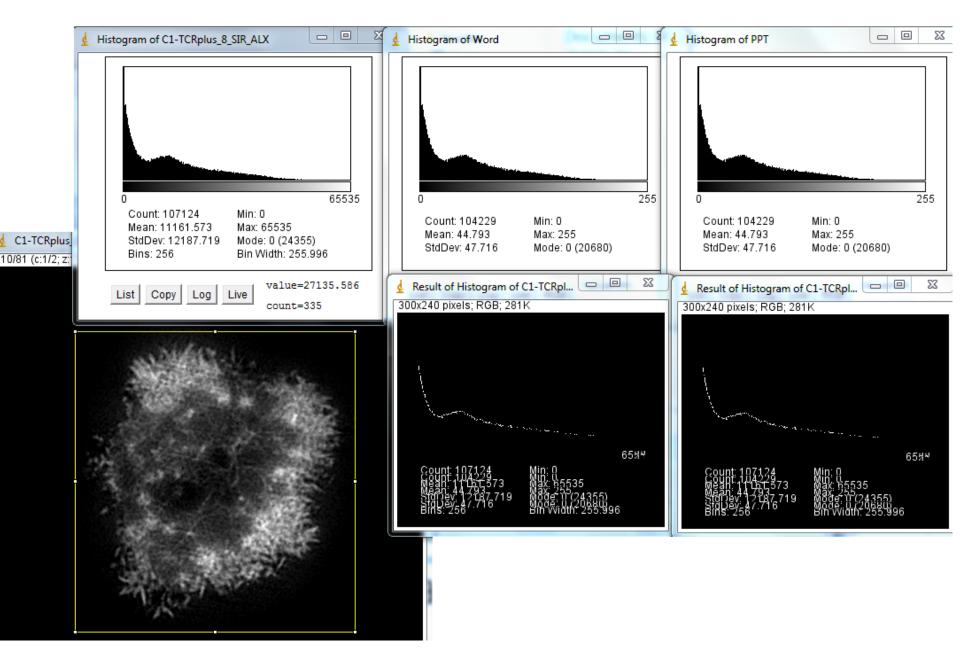


Examples using synthetic background image; divide by Gaussian blurred or median filtered image:



If measuring all the protein in a cell based on intensity, are confocal and high N.A. preferable? Explain.

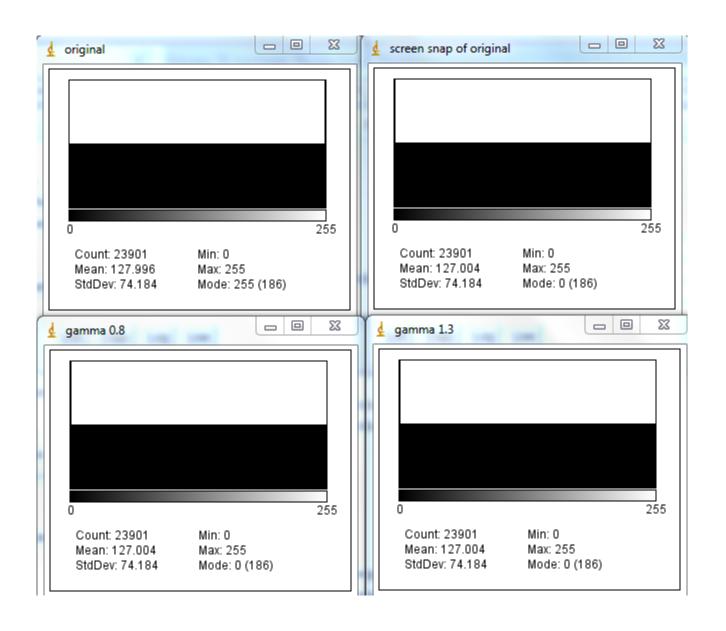
An important but often overlooked concern is to avoid using programs that alter the image without the user being aware of it. For example, carefully correcting the colors in an image using Photoshop and then placing it in Powerpoint for presentation will cause changes even on the same computer screen (as well discarding pixels and reducing resolution if copy-and-paste is used for the transfer). In addition the image may appear different on another computer monitor, or when using a projector. Printing an image will also alter colors [.] ... Placing images into a figure using Illustrator makes it all too easy to adjust the size or position of an image that will result in interpolation and alteration of pixel values (rotation is especially bad in this regard). Placing images into a web page for viewing can cause all of these problems at once. And using any lossy compression method such as jpeg will discard potentially important information that cannot be recovered.



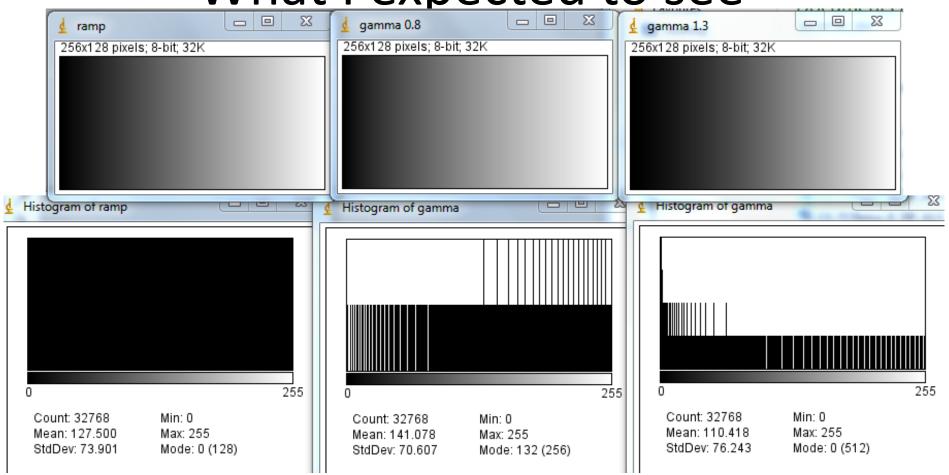
### Every Video Card and Screen Have Their Own Biases

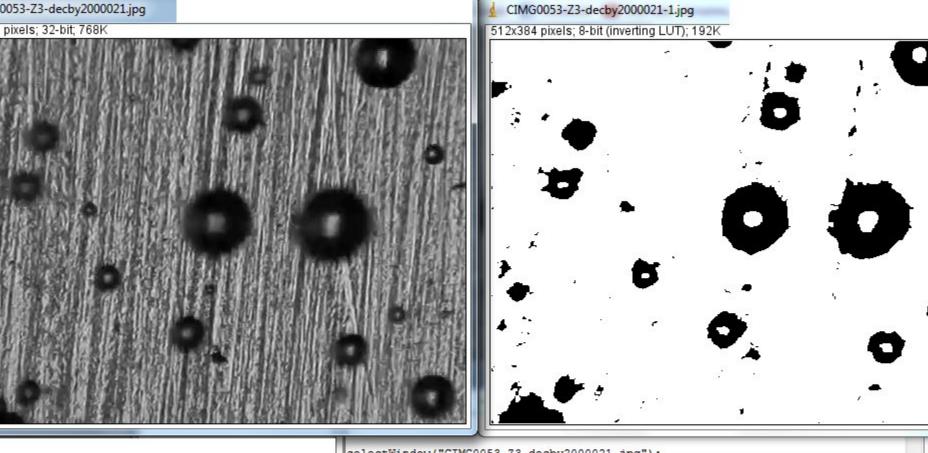


#### Screen snaps actually record raw data, not displayed LUT



What I expected to see

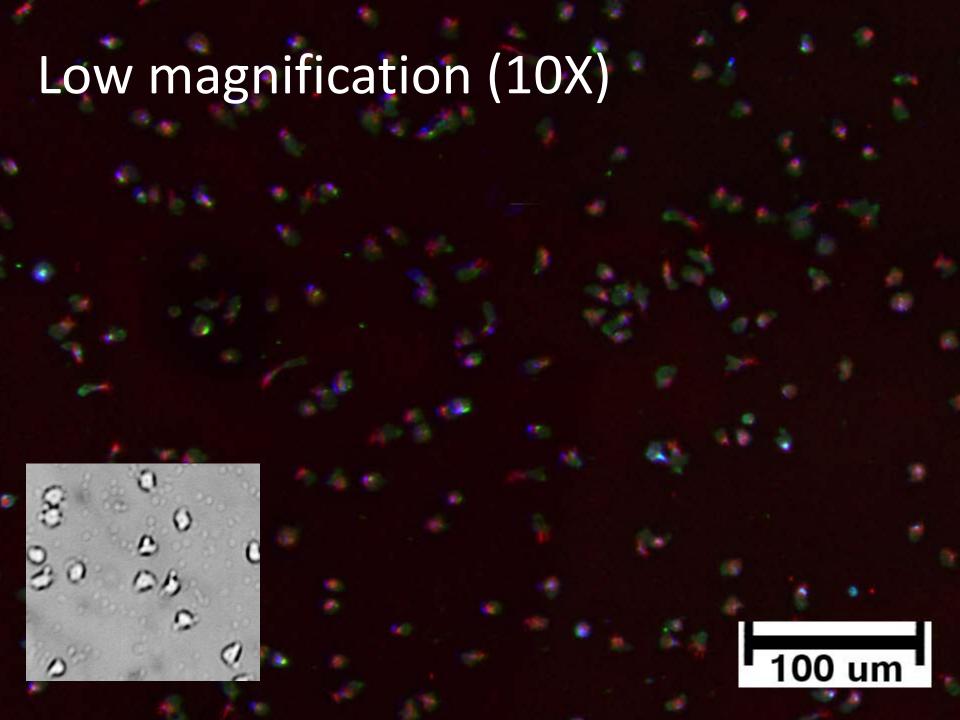


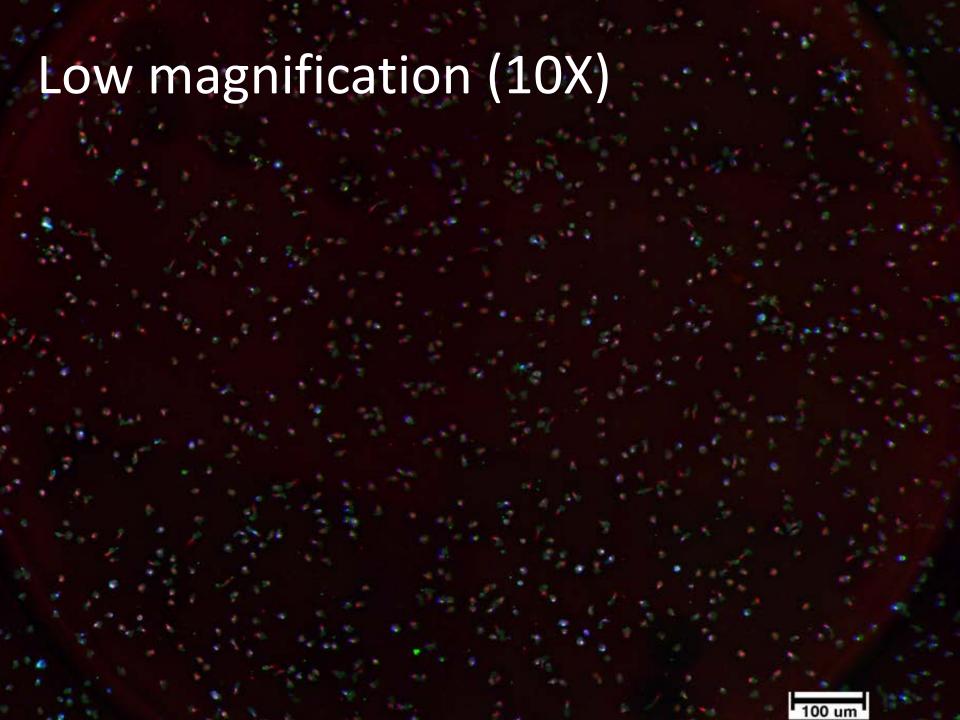


```
selectWindow("CIMG0053-Z3-decby2000021.jpg");
run("Duplicate...", "title=CIMG0053-Z3-decby2000021-1.jpg");
run("32-bit");
run("Median...", "radius=6");
run("South");
//run("Threshold...");
setAutoThreshold("Default");
setAutoThreshold("Default");
run("Convert to Mask");
run("Erode");
```

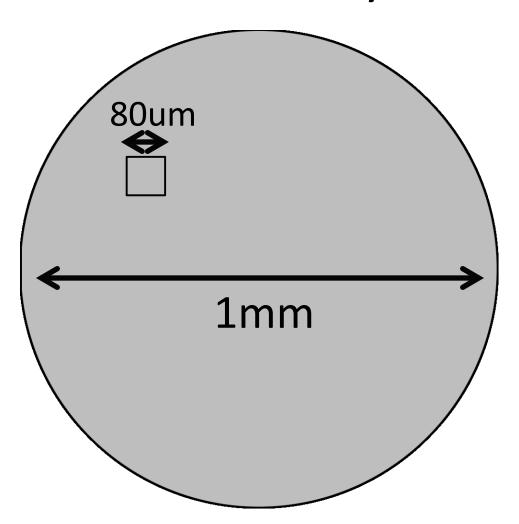
A few examples of ethical choices.

# How to choose fields of cells fairly for quantification

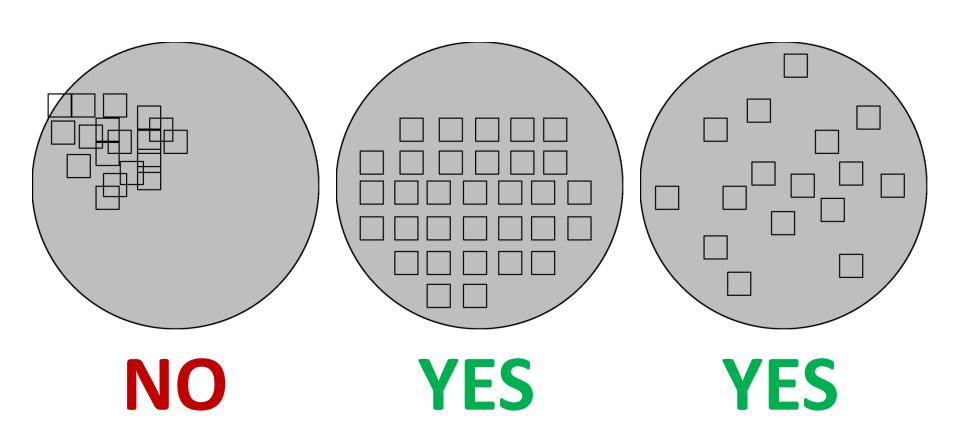




# Imaging "representative" or random cells on a bilayer



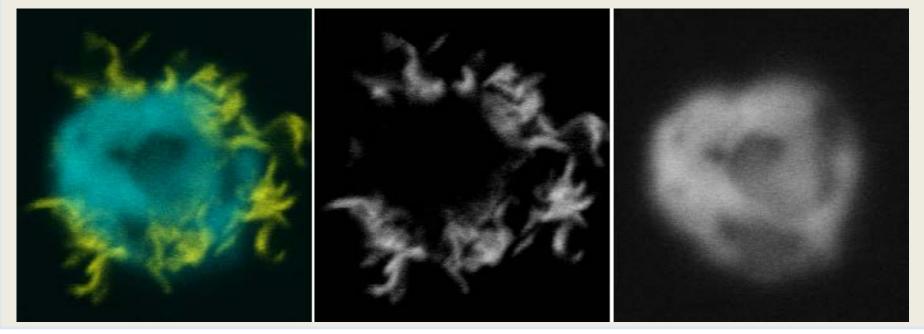
### Distribution of fields imaged

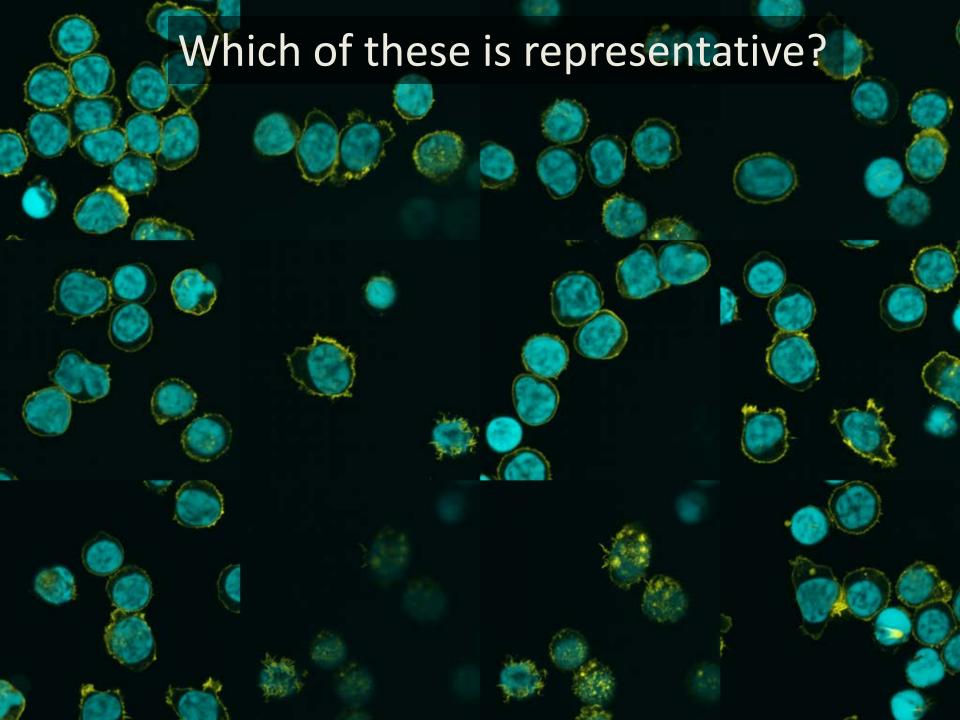


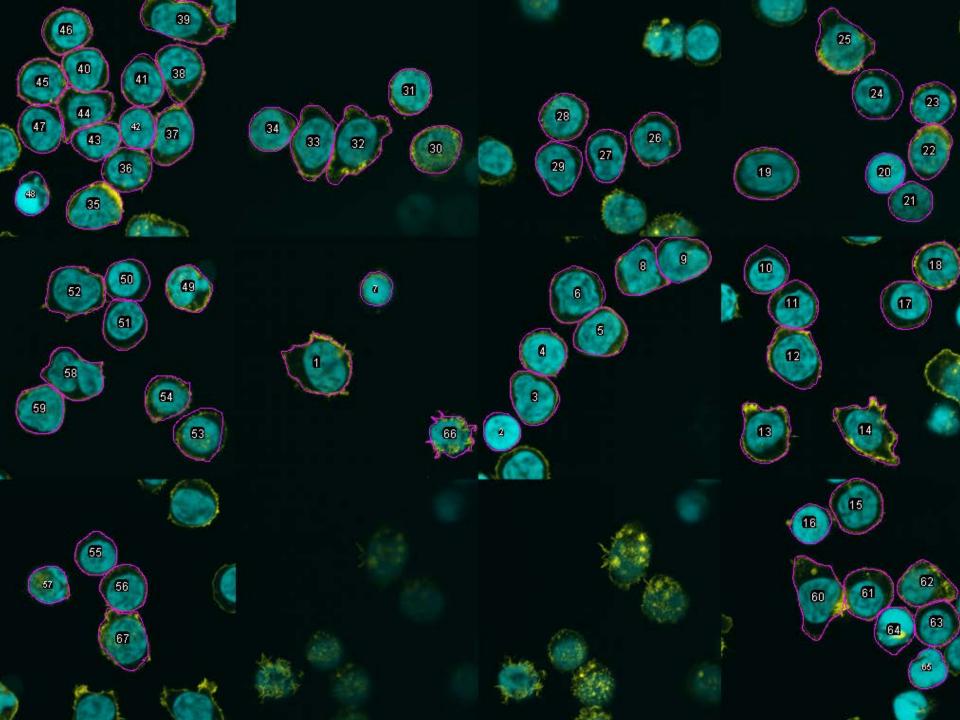
### Showing the Typical Cell

You have hundreds of pictures of cells but for the figure in your paper you are going to show one. How do you choose it?

Fig. 1 Showing this cell supports hypothesis that cells are small & have a lot of ruffles after treatment.

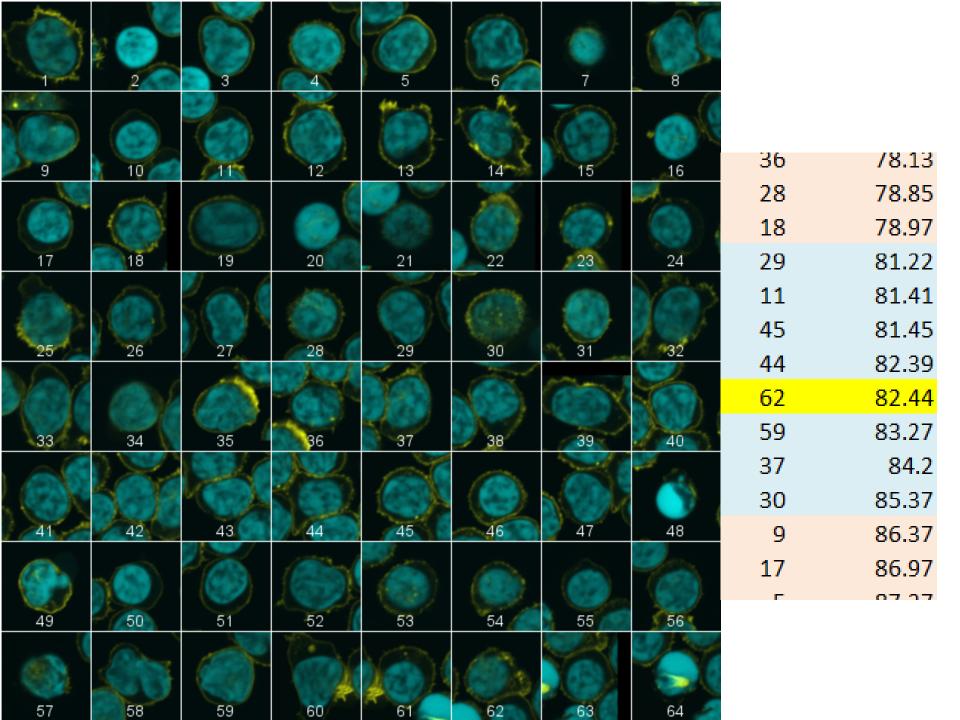




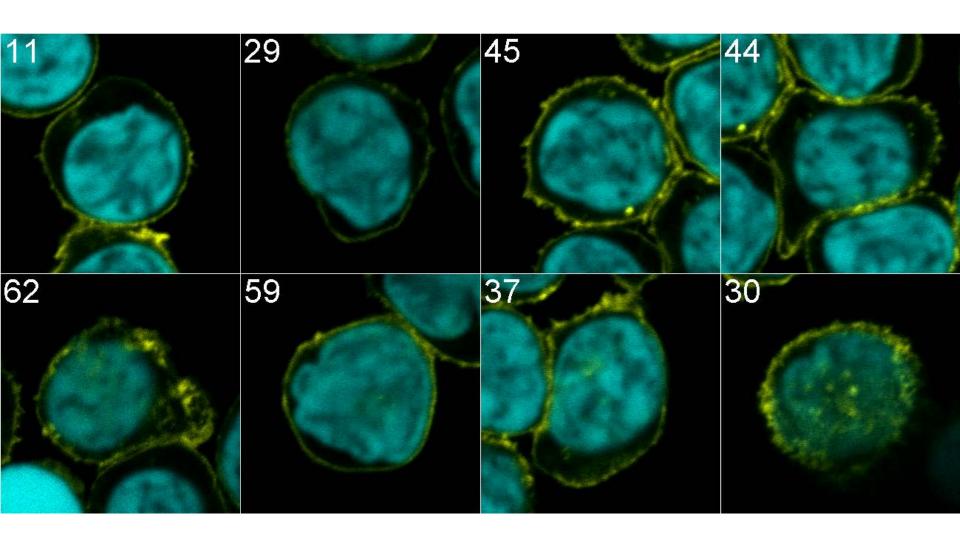


	Α	В	С	D	Е	F	G	Н	1
1		Area (um^2)	Circ.	AR	Round			Average	
2	1	129.04	0.83	1.21	0.83		Area um^2	82.52	
3	2	49.97	0.97	1.06	0.94		stdev	20.50	
4	3	90.84	0.94	1.22	0.82		n=	67	
5	4	76.69	0.96	1.07	0.93		SEM	2.50	
6	5	87.37	0.95	1.16	0.86				
7	6	106.81	0.97	1.1	0.91				
8	7	38.17	0.98	1.08	0.92				
9	8	88.22	0.83	1.1	0.91				
10	9	86.37	0.95	1.23	0.81				
11	10	71.39	0.96	1.09	0.92				
12	11	81.41	0.97	1.04	0.96			-:+-	
13	12	110.95	0.91	1.34	0.75			rite	rion Area
14	13	106.63	0.82	1.18	0.85		_		
15	14	116.23	0.63	1.56	0.64				
16	15	99.05	0.96	1.06	0.94				
17	16	53.68	0.97	1.08	0.93				
18	17	86.97	0.98	1.06	0.95				
19	18	78.97	0.98	1.03	0.97				
20	19	118.18	0.97	1.24	0.81				
21	20	54.34	0.98	1.11	0.9				
22	21	59.79	0.97	1.14	0.88				
23	22	77.95	0.93	1.38	0.72				
24	23	67.82	0.97	1.1	0.91				
25	24	88.43	0.98	1.06	0.94				
26	25	113.08	0.85	1.24	0.81				
27	26	87.96	0.95	1.07	0.94				
28	27	74.9	0.93	1.34	0.75				
29	28	78.85	0.98	1.07	0.94				

	Α	В	С	D	Е			
	20	54.34	0.98	1.11	0.9			
	21	59.79	0.97	1.14	0.88			
	43	60.06	J.91	1.4	0.71			ı
	55	62.31	7 95	1.03	0.98			
	66	63.75	0.45	1.09	0.92			
	64	64.85	0.92	1.17	0.86			
	23	67.82	0.97	1.1	0.91		_	
	50	69.17	0.96	1.09	0.92	Critoria	$\sim$ $\sim$ $\sim$	
	54	70.75	0.91	1.13	0.88	. (,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	on Area	
	31	71.02	0.98	1.08	0.92			
	10	71.39	0.96	1.09	0.92			
	34	71.73	0.96	1.16	0.86			
	63	72.29	0.96	1.07	0.93			
	49	72.46	0.96	1.11	0.9			
	40	72.96	0.96	1.14	0.87			
	27	74.9	0.93	1.34	0.75			
	56	75.46	0.97	1.03	0.97			
	41	76.64	0.96	1.25	0.8			
_	4	76.69	0.96	1.07	0.93			
_	46	77.09	0.93	1.16	0.86			
_	47	77.16	0.96	1.12	0.9			
_	51	77.52	0.97	1.17	0.85			
_	22	77.95	0.93	1.38	0.72			
_	36	78.13	0.98	1.13	0.88			
4	28	78.85	0.98	1.07	0.94			
_	18	78.97	2.98	1.03	0.97			
4	29	81.22	11,92	1.18	0.84	_		
4	11	81.41	0.97	1.04	0.96		CEN 4	STDEV
4	45	81.45	0.98	1.08	0.92		SEM	SIDLV
4	44	82.39 82.44	0.7	1.34	0.75			Range
-	62 59	83.27	b.96		1	THE AVERAGE CELL	Range	Marige
$\dashv$	37	84.2	0.36	1.58	0.63			
$\dashv$	30	85.37	1196	1.30	0.63			
$\dashv$	9	86.37	95	1.23	0.81			
$\exists$	17	86.97	0.98	1.06	0.95			
$\exists$	5	87.37	0.95	1.16	0.86			
$\exists$	26	87.96	0.95	1.07	0.94			
$\exists$	8	88.22	0.83	1.1	0.91			
١	38	88.38	0.89	1.32	0.76			
+	24	88.43	0.98	1.06	0.94			
۲	53	89.12	0.93	1.12	0.89			
1	61	89.4	0.96	1.05	0.96			
1	3	90.84	0.94	1.22	0.82			
	35	92.29	0.94	1.25	0.8			
	67	97.85	0.82	1.36	0.73			
	15	99.05	<u>2</u> 96	1.06	0.94			
	58	99.33	City.		0.70			
	52	105.22	0.9	1.22	0.82			
	13	106.63	0.82	1.18	0.85			
	6	106.81	0.97	1.1	0.91			



#### Cells that are within SEM of AVG area



- ImageJ Interest Group [IMAGEJ@LIST.NIH.GOV] on behalf of Rasband, Wayne (NIH/NIMH) [E]
  [rasbandw@MAIL.NIH.GOV]
- **Subject:** Count negative pixels in a stack
- To: IMAGEJ@LIST.NIH.GOV

•

- Saturday, August 17, 2013 12:00 AM
- On Aug 16, 2013, at 10:48 AM, sreeyuth wrote:

```
> Hi,
```

-wayne

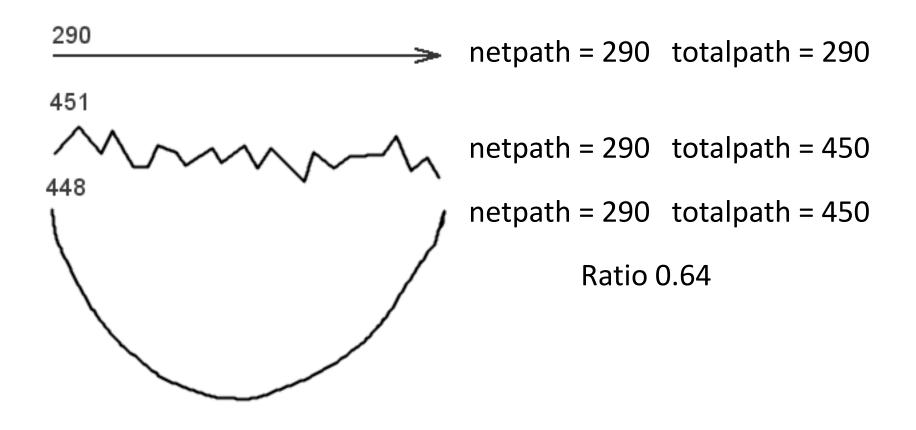
- > I have around 440 images in my stack. From each image, I want to get the
- > total number of pixels having negative values. So in the end I expect a list
- > which has just 2 columns, Slice Number and No. of pixels having negative
- > values in that slice.

Here is a macro that does this:

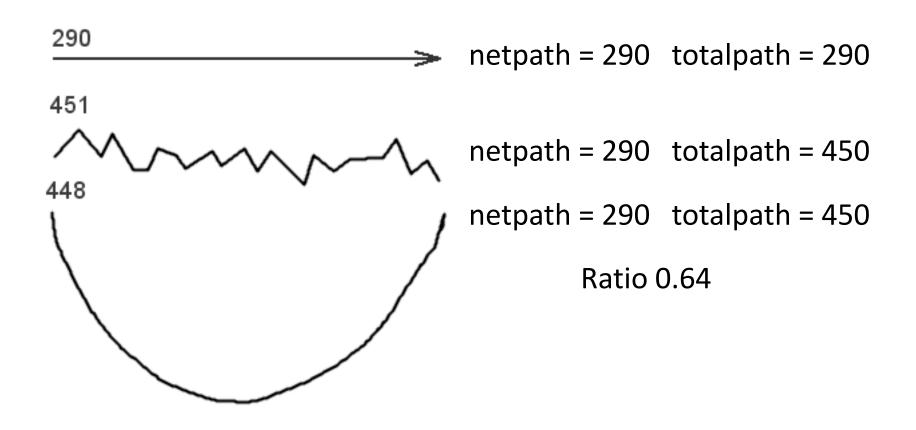
```
saveSettings;
run("Clear Results");
run("Set Scale...", "distance=0 known=0");
setThreshold(-99999, -0.00001);
run("Set Measurements...", "area limit");
for (n=1; n<=nSlices; n++) {
   setSlice(n);
   run("Measure");
}
restoreSettings;</pre>
```

Why to ask the **ImageJ** listserv for help

## Netpath/totalpath may not tell characteristics of movement



## Netpath/totalpath may not tell characteristics of movement



Also, irregular velocity vs. constant speed...

#### The Size Of The Ruler Problem

Message: The spatial and temporal resolution changes the answer. Choose magnification, time intervals, and duration of experiment appropriate for your questions. Be careful comparing to others' results.

For example:

To see ruffling, high mag, high speed, short duration.

To see synapse formation and release, low mag, long intervals, long duration.