Advanced Microscopy Course 2014

Lecture 5:

Basic Image Processing

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^{*} Dominic Waithe, Lecture 17: Applied Image Analysis and Matlab *

Basic Image Processing

- What is a digital image?
- What makes a good image?

Correct image acquisition

Signal to Noise

Resolution and Sampling

The basics of image processing

Golden rules of image processing

Conceptual Hierarchy of Image Processing

Low Level Processing:

Display (Figures)

Filtering

Mid - Level Processing:

Segmentation

Spectral unmixing

High -Level Processing / Analysis:

Colocalisation

Tracking

Statistics

What is a digital image?



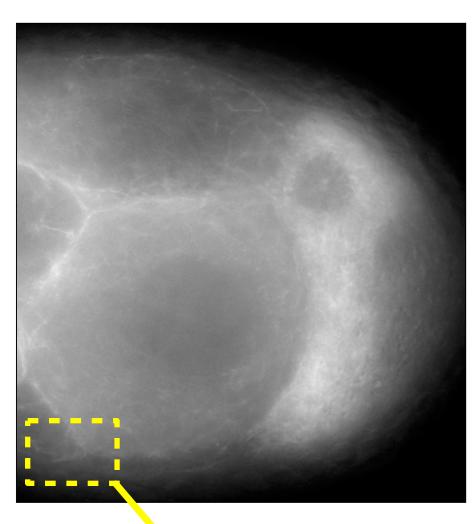
http://en.wikipedia.org/wiki/Money_for_Nothing_(song)

What is a digital image?

An image represents the output of the optics and detector of the imaging system

image ≠ object

image = object ⊗ PSF



622 629

526

613

243

819 1026 1148

- A digital image is a numerical array: elements = pixels or voxels with:
 - defined size (sampling resolution)
 - defined no. of grey levels (bit depth)
- In addition to "useful" signal there is:
 - dark signal from the detector
 - autofluorescence (background)
 - statistical noise of photon detection
- Details are detected within the limitations of:
 - the imaging optics
 - the sampling rate (pixel size)
 - the statistical noise
 - the sample contrast / detector dynamic range

Image Parameters - what to record (= image metadata)

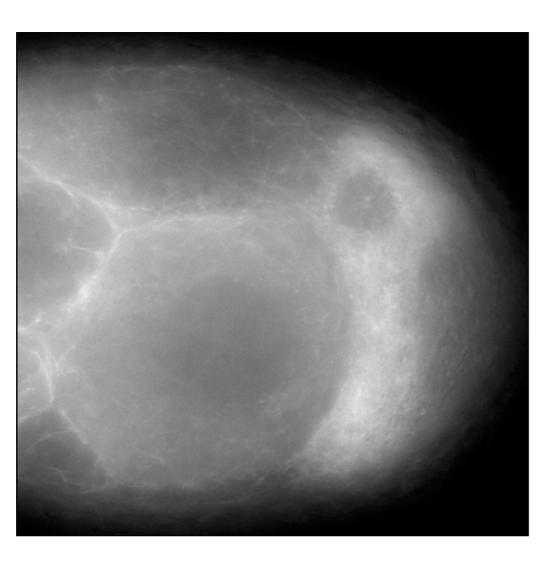
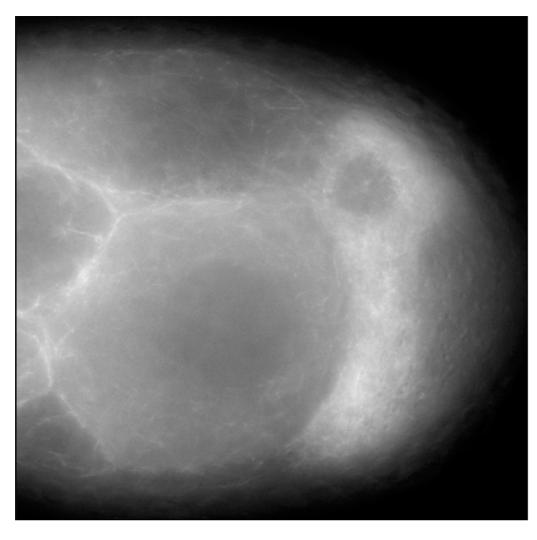


Image Parameters - what to record (= image metadata)



Wide-field fluorescence 490 ex 520 em X60 1.2 WI xy 212 nm; 60 (z step 200 nm) Bin 2x2 250 ms exposure Contrast stretched to fill 8 bit display Tau-GFP Oocyte

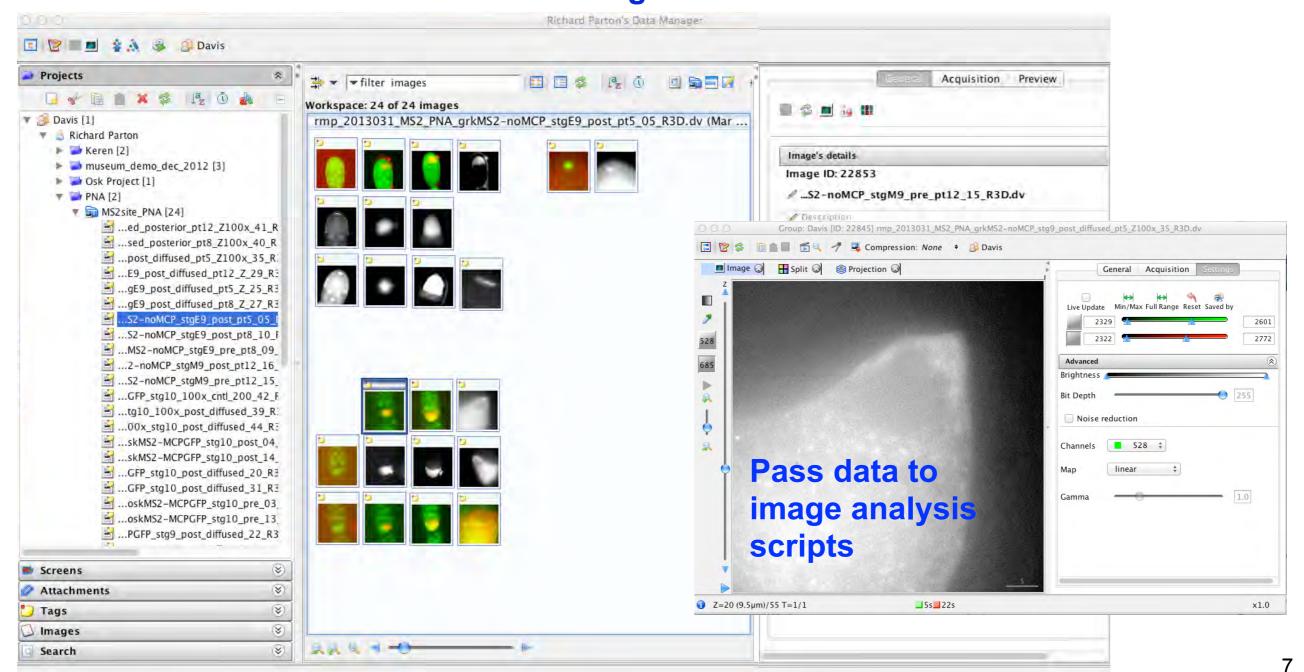
- Type of imaging Wide-field fluorescence
- Excitation and Emission wavelengths 490/520
- Optics used x60 NA 1.2 water immersion
- Image pixel dimensions 212x212 (x200) nm
- Depth or Dynamic range 8 bit; 256 greys
- Any processing performed
 - 12 bit to 8 bit conversion
 - contrast adjustment
- Display parameters range 0-255, grey scale- gamma = 1
- The Biology Drosophila stage 8 egg chamber
 Tau GFP, labelling microtubules

OME - Open Microscopy Environment



* Douglas Russell, Lecture 18: Image Management *

 Purpose: Supporting Metadata Management for Microscopy avoids problems of image formats archiving and retrieval data sharing



What makes a good image?



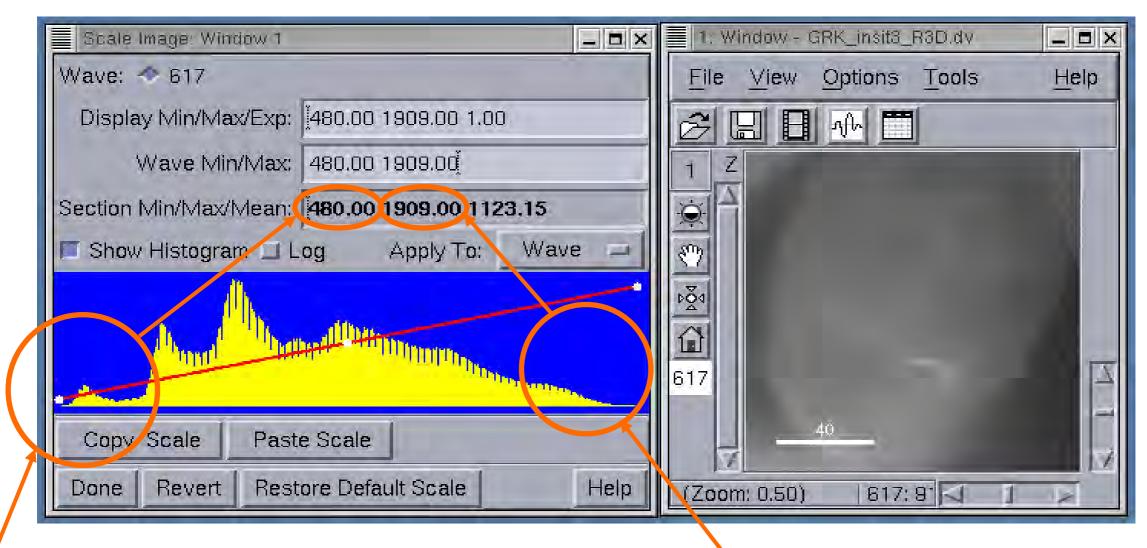
Adapted From - http://www.go4costumes.com/products/Black-High-Afro-Wig

Correct Image Acquisition

- The system must be correctly set up and aligned
 PSF verification (beads) *Practical 2*
- The specimen should not cause undue optical aberration mounting / appropriate optics
- Avoid underflow and overflow but fill the dynamic range
 - use a colour LUT
 - beware of auto-intensity scaling
- Take a dark signal image and/or background
 - Dark subtraction processing
- Be aware of XYZ optical resolution of the system and sample appropriately
 - PSF of the imaging system
- *Nyquist*
- Pixel (voxel in 3D) size in the image
- Take care with signal to noise limitations
 - collect enough light: integrate, average

Correct image acquisition

 Use histogram analysis to ensure you collect enough light in the features of interest



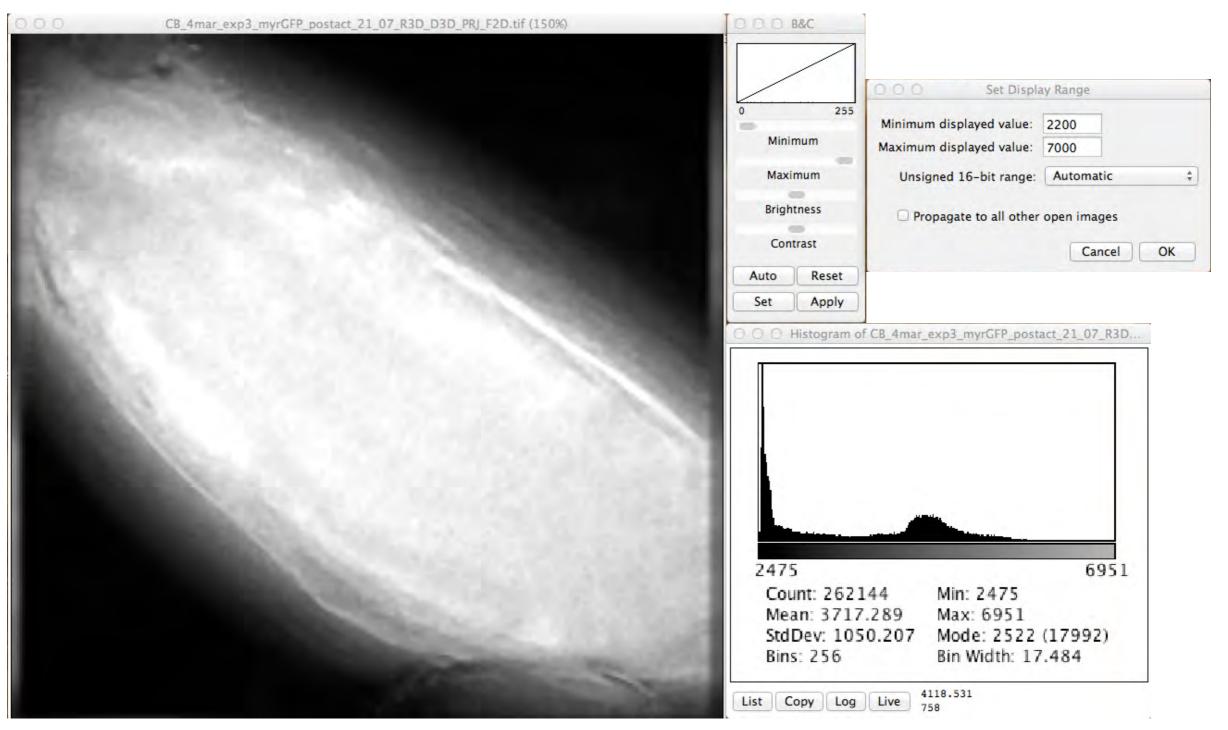
mostly camera offset, dark signal and background max value should be below detector saturation = avoid saturation

- NOTE the brightness of the display is not the best indicator of image signal
- Make good use of the dynamic range:8-bit = 255, 12-bit = 4095, 16 bit = 65535

Correct image acquisition







Noise / Signal to Noise (S/N)



http://rogewu.comyr.com/brad-pitt-meet-joe-black-wiki.php

Signal to Noise - definitions:

One of the most important limitations to image quality and image processing



- Noise is NOT background, auto-fluorescence or dark signal
- Good image data has a high S:N ratio (>4)
- Fundamental limit = Poisson distributed statistics of photon detection (shot noise)

```
Poisson n
distributed S:N ratio = √n
```

Statistics of photon counting dictate the minimum useful signal

```
Average signal = 9, S:N ratio = 3
Average signal = 100, S:N ratio = 10
Average signal = 10,000, S:N ratio = 100
```

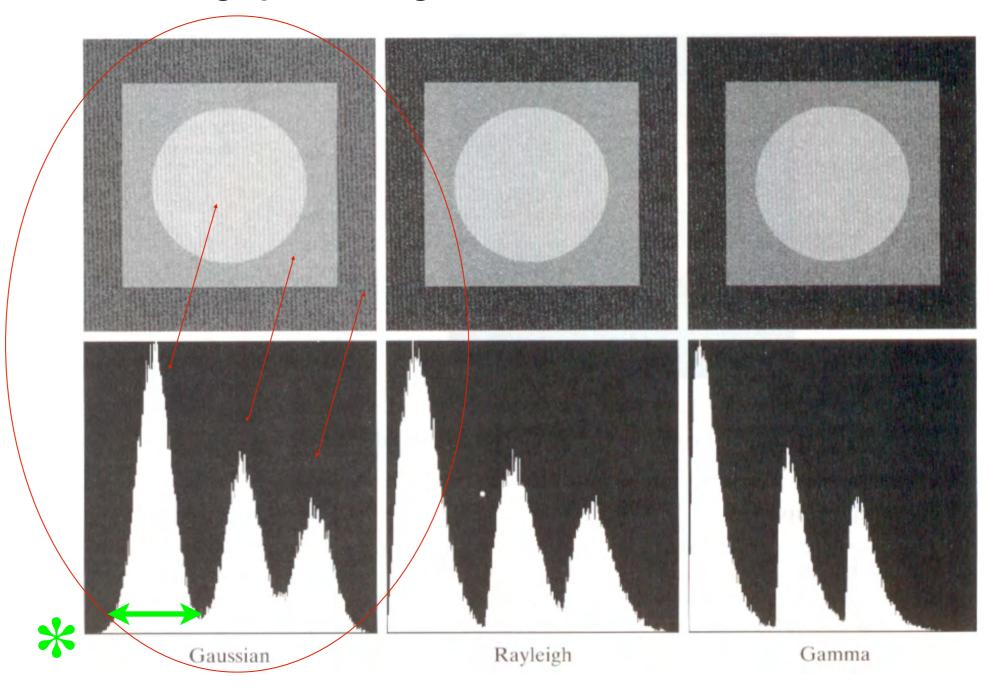
A meaningful difference in intensity needs to be at least three times the noise level

Additional sources of noise from digitisation, detector readout, thermal noise.

Signal to noise

- noise types

Most commonly used noise model for image processing = Gaussian



How to deal with signal to noise

Acquisition

- Use sensitive, high dynamic range, low noise detectors: cooled CCD, EMCCD
- Count as many photons as possible:

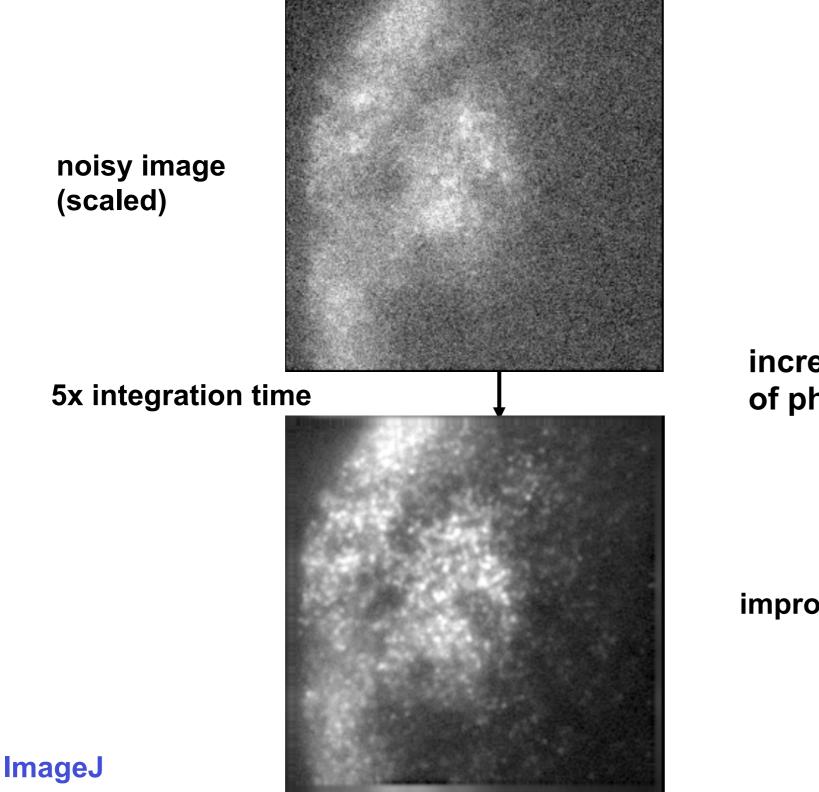
Bright dyes
Good excitation / emission
Integration time (accumulation or averaging)

Post Acquisition

- Image averaging
- Noise reduction filtering using a spatial filtering mask 3x3 median filter
- Noise reduction filtering in the frequency domain Fourier bandpass filter

Improving signal to noise

increased signal increases S/N = improved contrast



increased number of photons counted

improved S/N

Avoid propagating noise

Noise is additive:

SO subtracting one noisy image from another propagates noise

THEREFORE

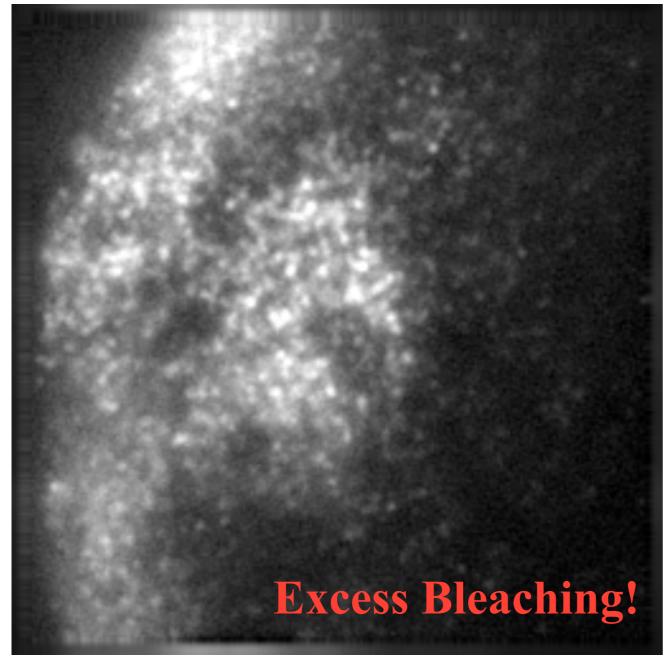
Or

Subtract an AVERAGE signal to avoid noise propagation

Where the signal is non-uniform across the field subtract a 4x AVERAGED image to avoid noise propagation

Detail Preserving Denoising Algorithms

10 ms; 50% power; projected 4Z; 100T



grk MS2(12), MCP-GFP 7Z, 3stacks/s

^{*} Collaboration with Graeme Ball

Detail Preserving Denoising Algorithms

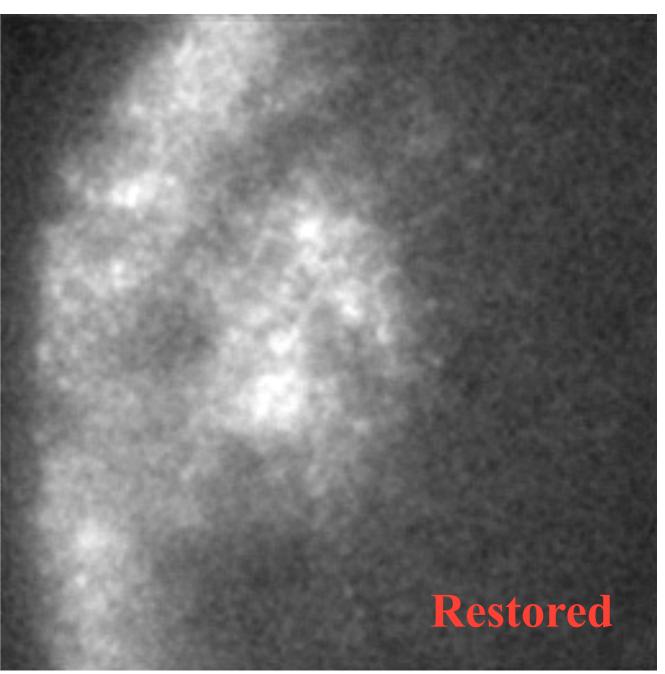
Patch-Based Denoising Kervrann and Boulanger

Boulanger, J., Kervrann, C., Bouthemy, P., Elbau, P., Sibarita, J.-B., & Salamero, J. (2010)

10 ms; 1% power; projected 4Z; 100T

Too Noisy!

dn_Iter2_P3

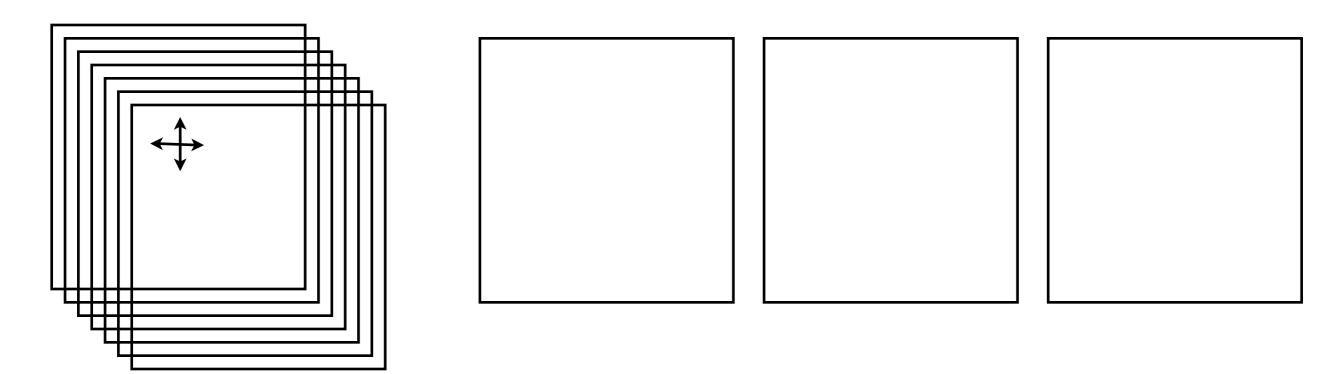


grk MS2(12), MCP-GFP 7Z, 3stacks/s

^{*} Collaboration with Graeme Ball

Detail Preserving Denoising Algorithms

* effectively averaging redundant data across a data set



XY redundancy at least 3 x 3 pixels = 9 (multiplies for each Z frame)

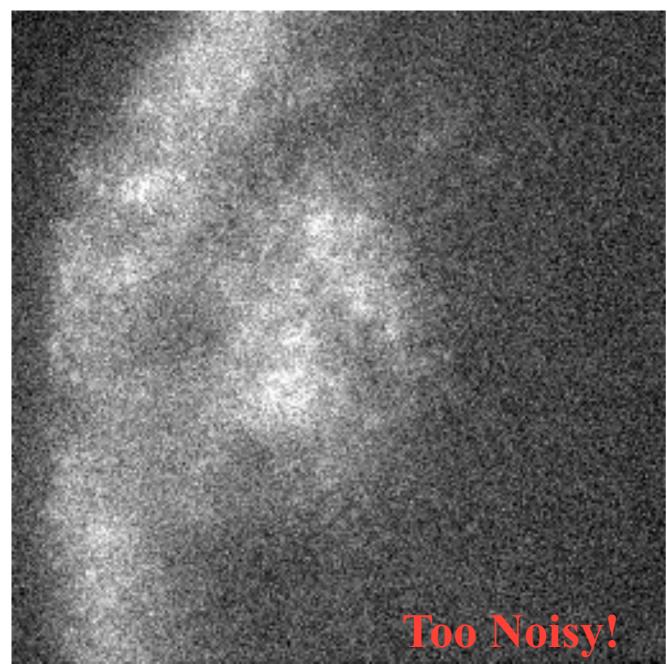
Z redundancy 3 pixels (Z>5 required for deconvolution)

t redundancy 3 to 100 pixels (depends upon events and temporal sampling)

- * root $[(3 \times 3) \times 7z \times 9t] = 24$ times noise reduction
- * root $[(5 \times 5) \times 7z \times 25t] = 66$ times noise reduction

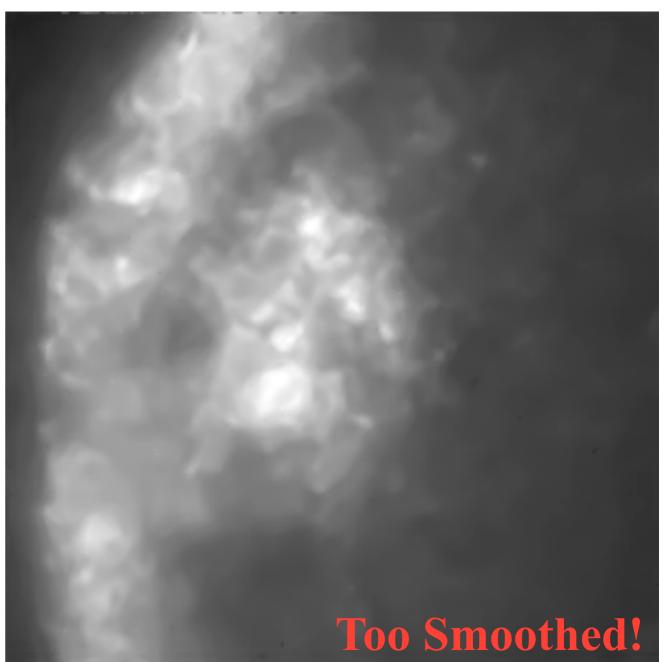
Detail Preserving Denoising Algorithms: oversmoothing

10 ms; 10% power; projected 4Z; 100T

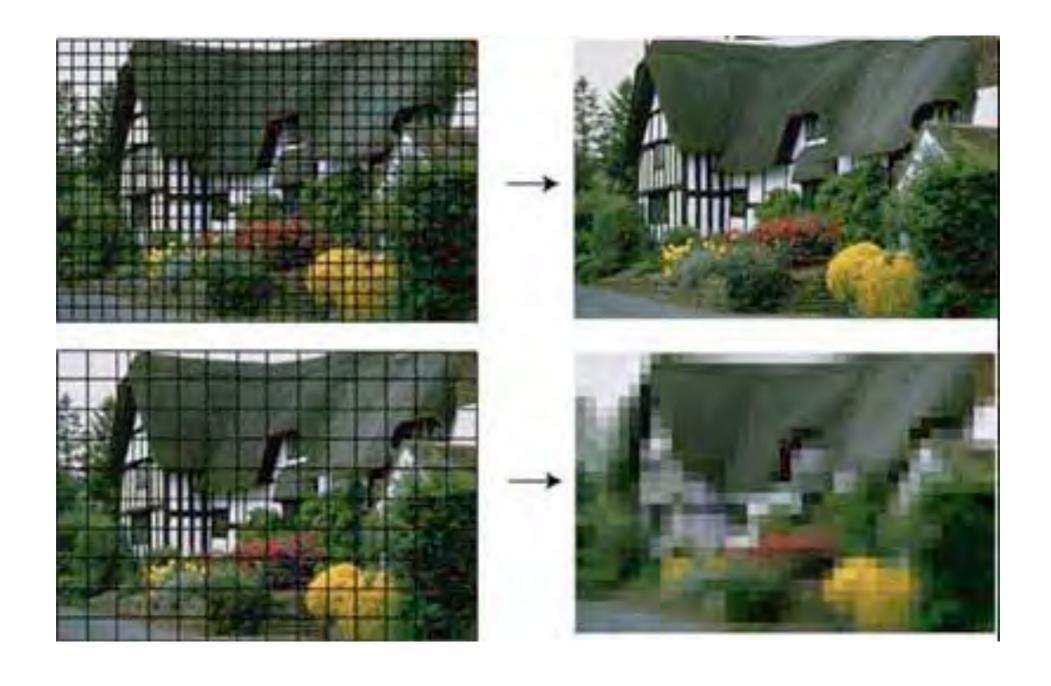


grk MS2(12), MCP-GFP 7Z, 3stacks/s

dn_Iter5_P3



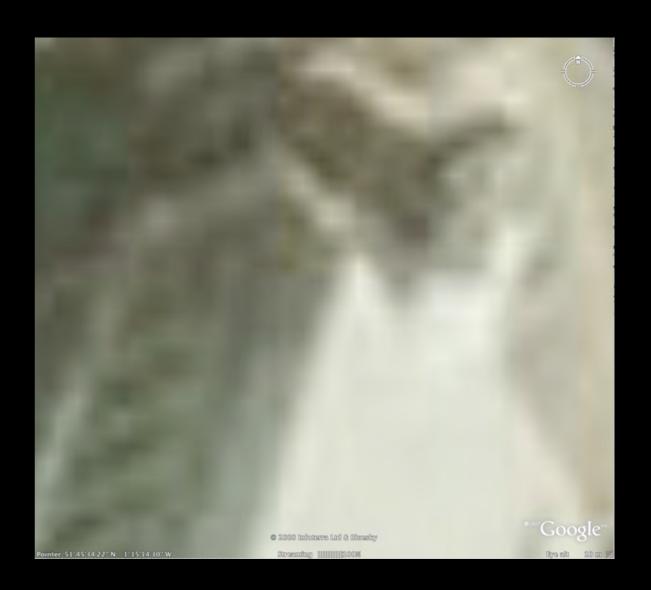
Resolution and Sampling



Resolutionthe ability to see small stuff

Magnifying is not enough:

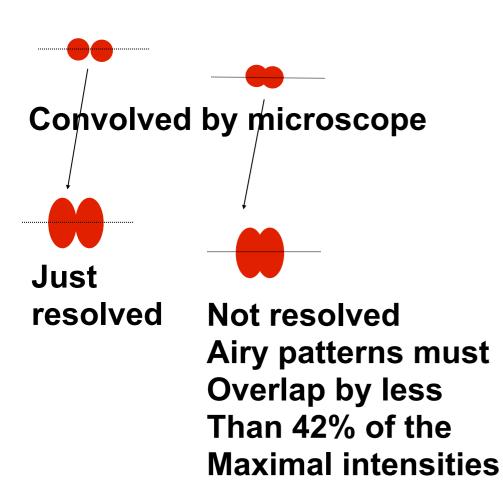




.....resolution is limited

Optical resolution: The Rayleigh Criterion

Two small objects

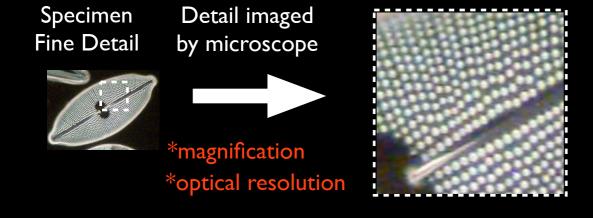


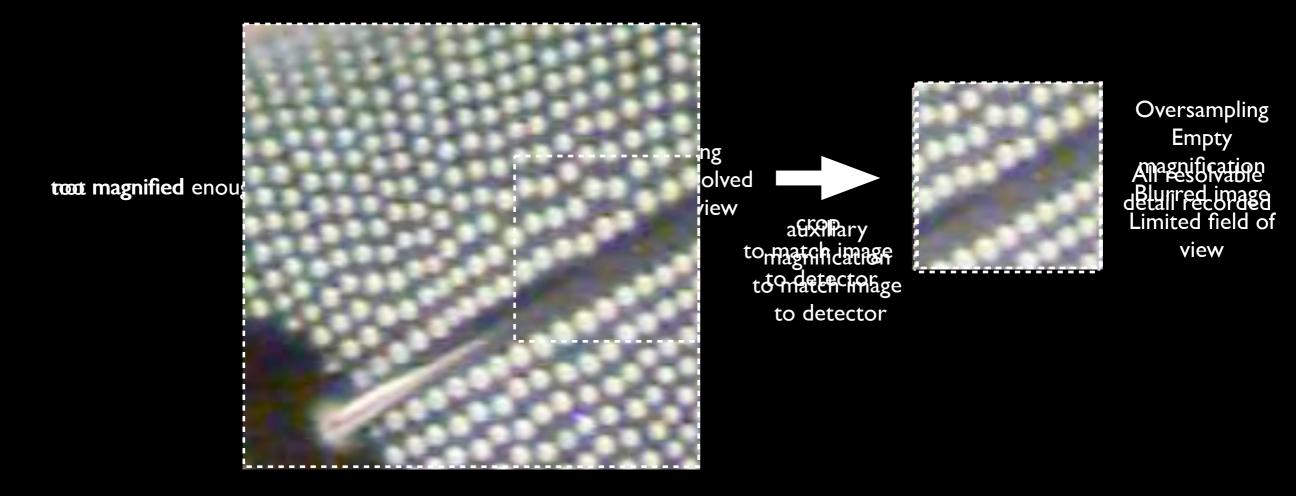
resolution limit $\approx \lambda_{em}/2$

XY resolution $\sim 200 \text{ nm}$ Z resolution $\sim 500 \text{ nm}$

Resolution

......Magnification and Sampling





What is the optimum magnification.....?

For optimal imaging: magnification must match the resolution to the detector

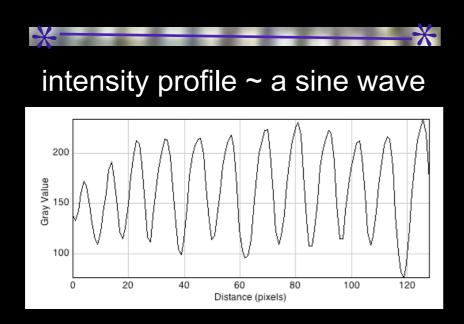
Diatom image: http://www.micromagnus.net

Resolution

......Magnification and Sampling

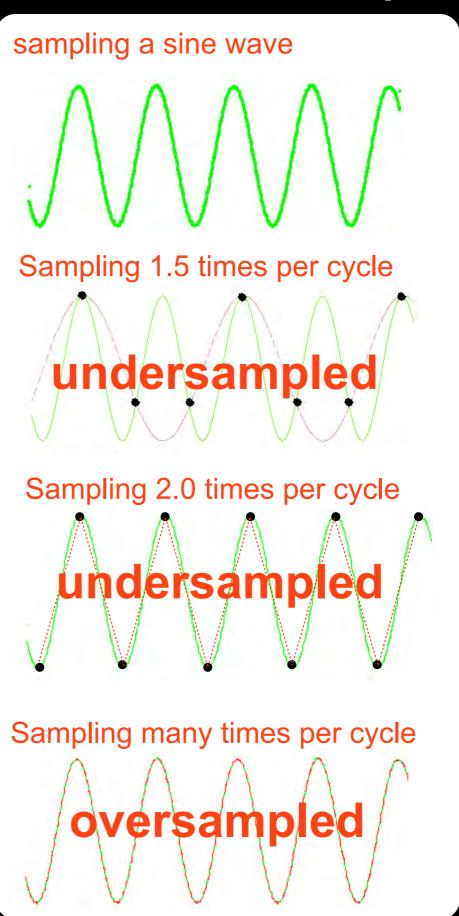
Specimen Detail imaged by microscope

*magnification
*optical resolution



Optimum = 2.3 times per cycle

= Nyquist sampling



Resolution

......Magnification and Sampling

For optimal imaging the magnification must match the resolution to the detector (eye or camera).....

optimal total mag X resolvable distance = detector element size

Considering Fluorescence imaging x100 objective; I.4 Na; 520 nm emission,....

total mag

resolvable distance

Detector Element

x100 objective X x1.0 Aux mag

 $1.22 \times \lambda 520/2Na$

Camera pixel element = 6.6 um (/~3 taking into account Nyquist)

Rearranging to find the optimum magnification

optimal total mag = Na \times 2 x Detector Element /1.22 x 520 nm

≈ ideal pixel size ~ 80 nm

 $\approx 87 \text{ x mag}$

x100 obj = GOOD SAMPLING

Resolution: sampling

Theoretical Axial Resolution (em 525 nm)	Appropriate Sampling According to Nyquist theorem - at least half the size	Pixel size on Delta Vision
1.4 oil = 229 nm	Nyquist ~ sample at 0.100 um/pixel	x100 = 0.063 um/pixel
1.35 oil = 237 nm	Nyquist – sample at 0.103 um/pixel	x60 = 0.106 um/pixel
1.2 water = 267 nm	Nyquist – sample at 0.116 um/pixel	x40 = 0.158 um/pixel
0.75 air = 427 nm	Nyquist – sample at 0.186 um/pixel	x20 = 0.317 um/pixel

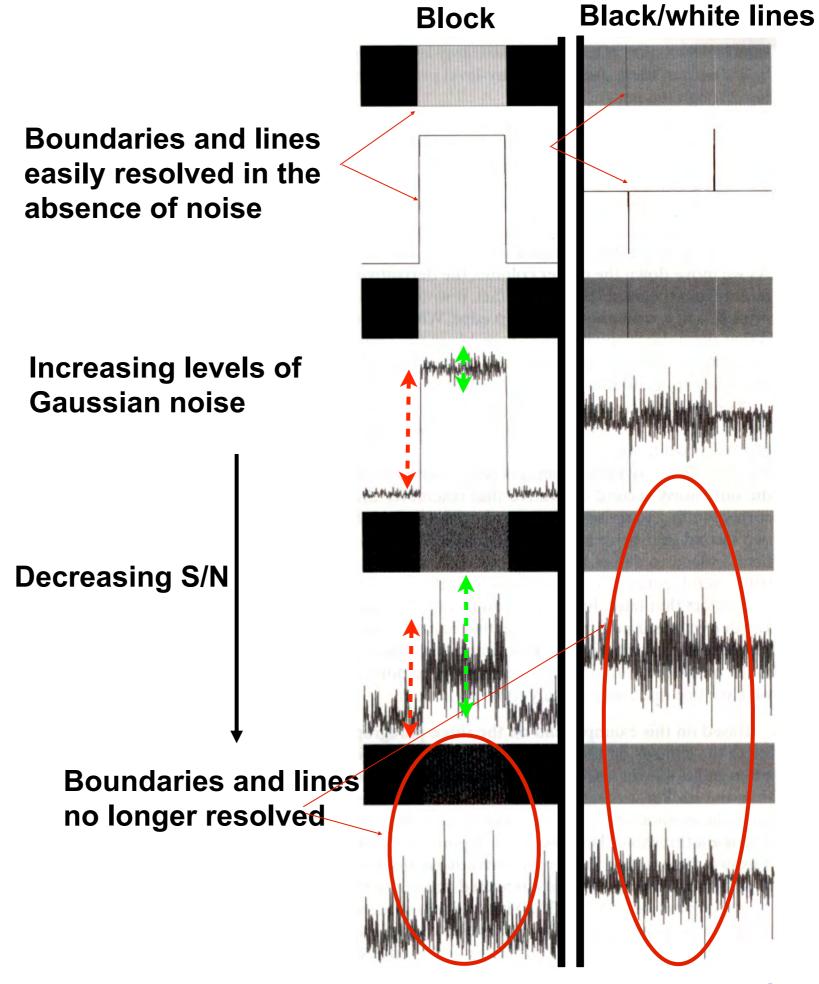
Undersampling limits the data available

Resolution, contrast, noise

- Noise limits the contrast which limits the details that can be resolved
 - = Noise limits resolution

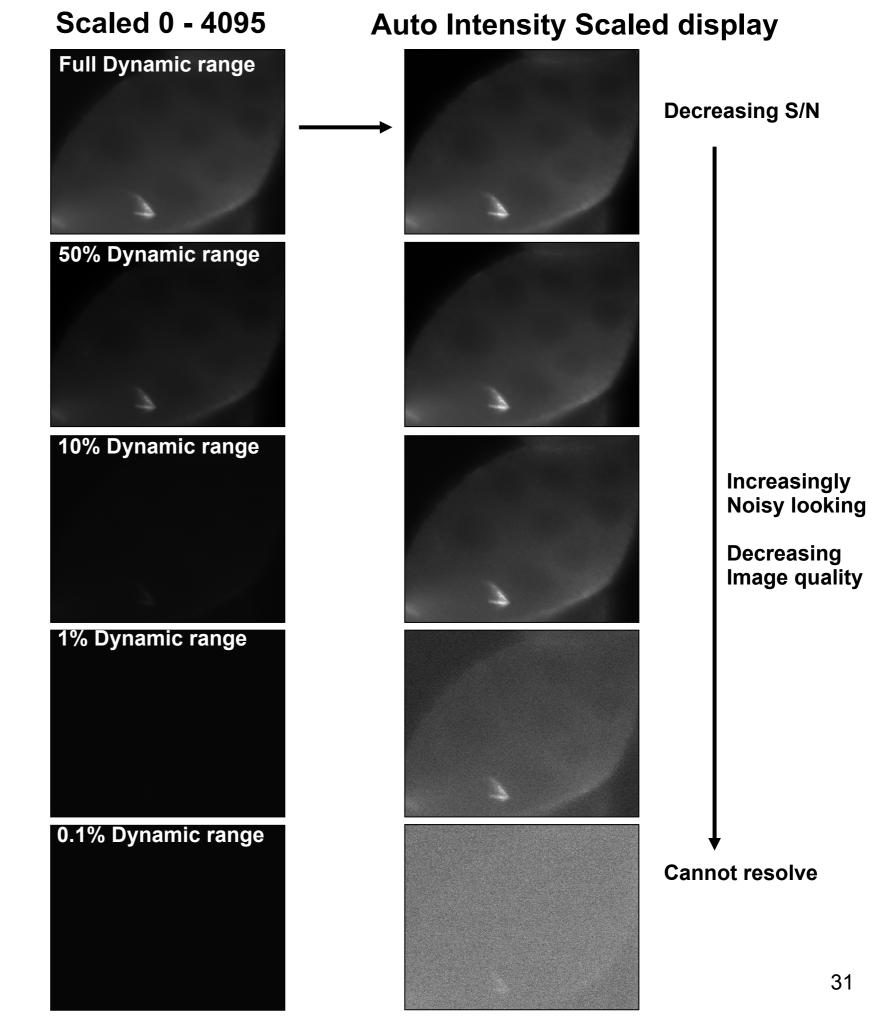
Resolution, contrast, noise

The difference between signal and background must be at least 3X the noise to be detectable



Resolution Contrast noise

Image series collected by decreasing the excitation lamp intensity from 100% to 50%, 10%, 1%, and 0.1%



The basics of image processing



"I want you to make me the fairest of them all."

Remember what makes a good image

- Good image data has a high S:N ratio (count more photons)
- Correctly sampled to reproduce the optical resolution (pixel = optical resolution/2)
- Avoid aberrations (sample prep / choice of objective / technique)
 - spherical aberration (SA)
 - motion blur
 - bad system alignment
- Correctly annotated (Metadata retained)

Image Processing is NOT a substitute for a good image

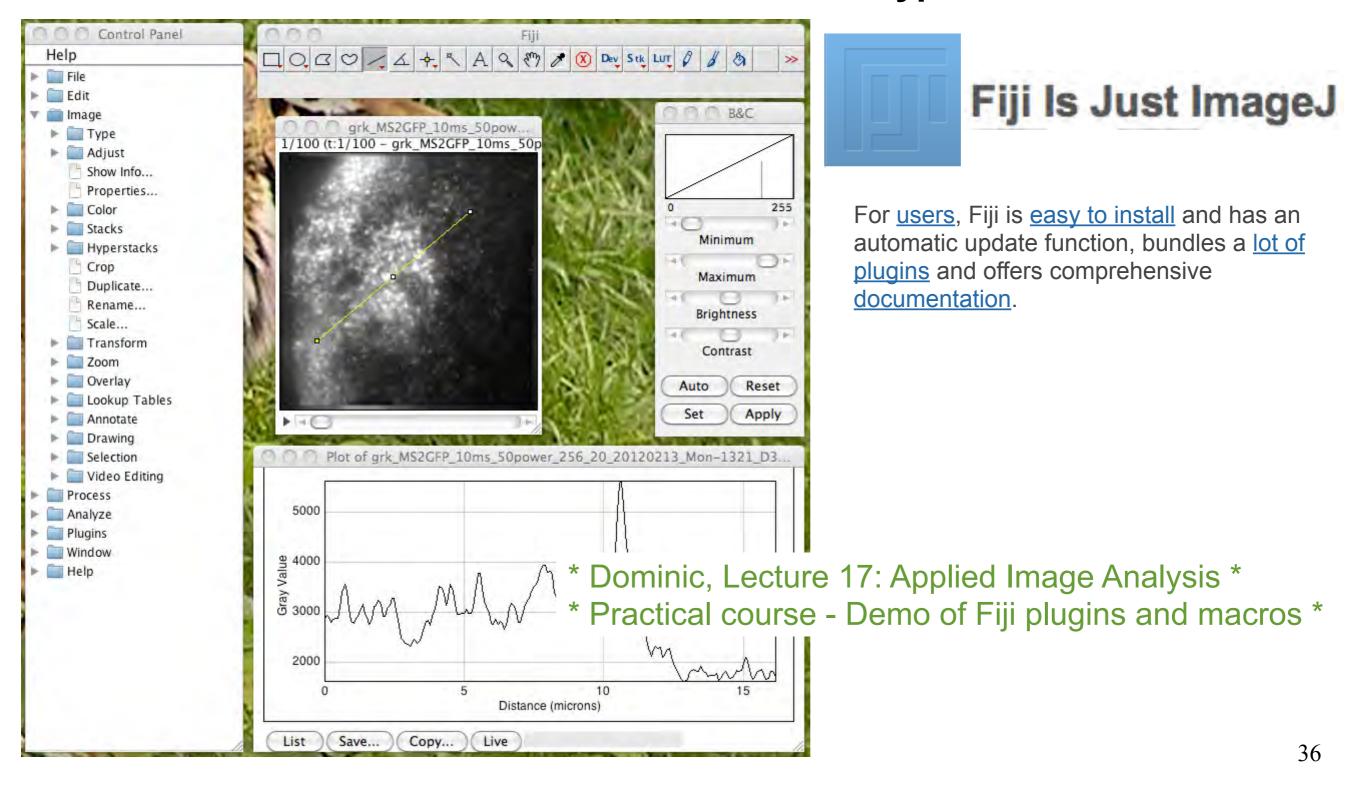


Golden Rules of Image Processing

- Always retain the original data
- Do not corrupt the integrity of the original data through processing:
 Processing should NOT generate data not present in the original image
- Images are arrays of numerical data and should be given appropriate consideration.
- Always record and report all processing steps.

Fiji = mage J http://fiji.sc/wiki/index.php/Fiji (http://rsb.info.nih.gov/ij/)

- Fiji is FREE and works on MAC, PC and linux
- Consists of a core program and plugins
- Uses Loci Bioformats to convert between file types



Do not corrupt the integrity of the original data

- Retain your original data in its original file format and original metadata associations
- Consider OMERO for data archiving (lecture 18 OME, bioformats)
 http://loci.wisc.edu/software/bio-formats
- Ideally use Uncompressed TIF (tagged image file format) for processed data
- AVOID compressed file formats when processing: JPEG, PSD, PDF, compressed
 TIF...This will cause data corruption and loss
- Most data is collected as single channel grey scale image.
 Avoid saving primary image data in colour formats (RGE)
- Avoid repeated inter-conversions of file formats
- Prepare figures for publication in Adobe Photoshop in T and annotate in Adobe Illustrator

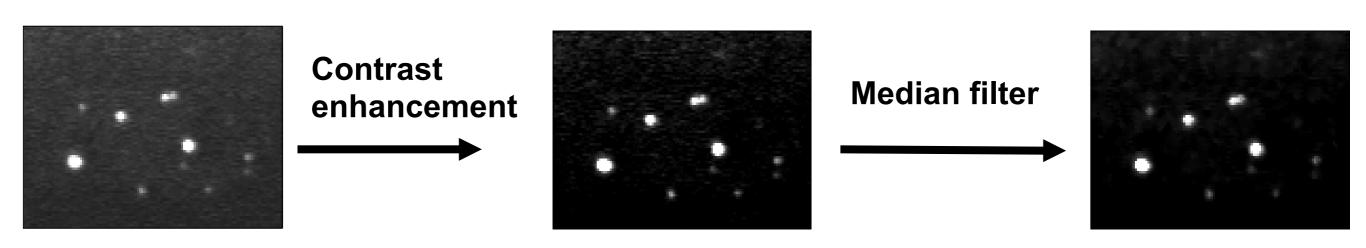
"The Scream" Edvard Munch

Conceptual Hierarchy of image processing

- Low level processing = Image enhancement (most common)
- Mid level processing = Features and attributes extracted
- High level processing = Analysis = Interpretation of images

LOW LEVEL Processing

- Visual enhancement
- Subjective = looks better
- Input is an image, output is an image
- Enhancement Filters
- Adjustments for image **Display** / Making figures



LOW LEVEL Processing - Image Restoration

- "Restoration" = attempting to reverse distortions of the object which arise during image capture
- * Flat-field correction correcting uneven illumination
- * Deblurring = Deconvolution, with or without PSF, unsharp mask
- * Simple filtering post acquisition increase in S/N e.g. by averaging
 - * Dominic, Lecture 17: Applied Image Analysis *
- * Normalization intensity of each time-point scaled to correct bleaching (or flicker)
- * Denoising post acquisition increase in S/N without loss of resolution
- * Image registration alignment of multiple channels for co-localisation analysis

LOW LEVEL Processing - "Background" Correction

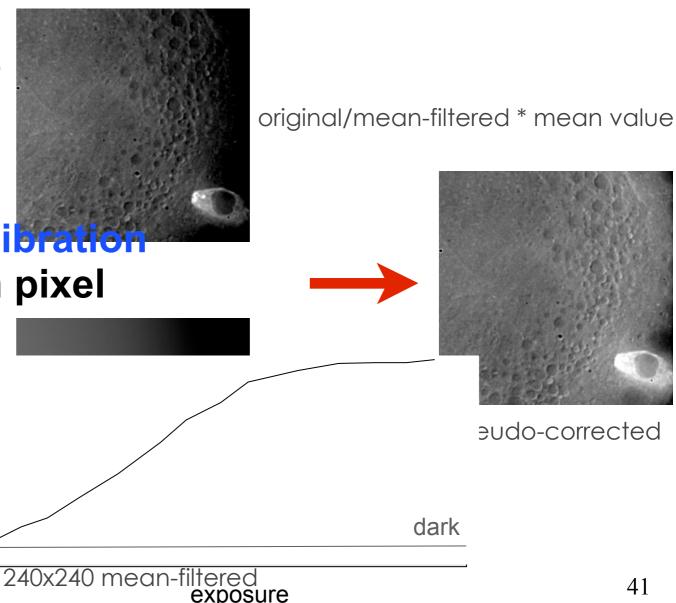
Requires dark image, flat field image and detector calibration

grey level

original image

- Detector offset / background subtract averaged dark value or image
- Uneven illumination correct using a flat field image

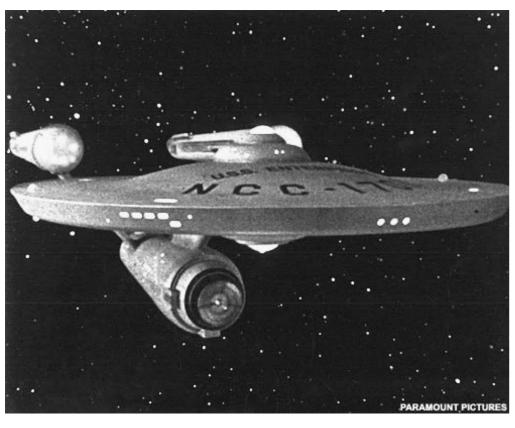
 Uneven detector response normalise using a detector calibration mapping the response of each pixel



pseudo-flat field

LOW LEVEL Processing - Simple filtering

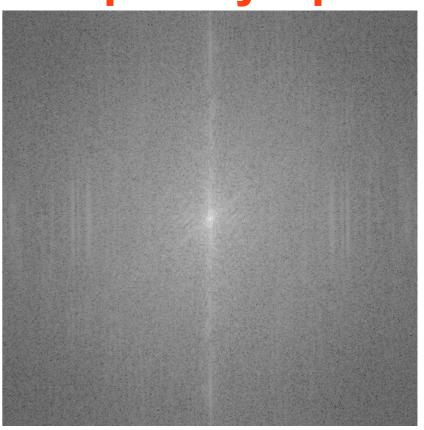
Processing in Real Space



or

Fourier Transform

Frequency Space



Example:

- Noise reduction filtering using a spatial filtering mask 3x3 median filter
- Noise reduction filtering in the frequency domain Fourier bandpass filter

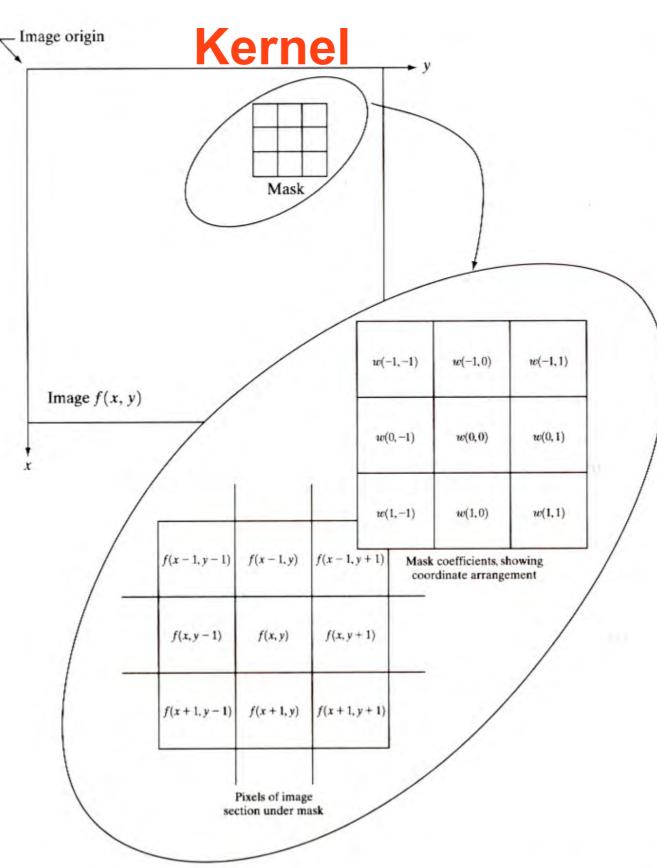
LOW LEVEL Processing - Simple filter, spatial domain

Real space - pixel by pixel

e.g. Noise reduction filtering using a 3x3 median filter

MEDIAN 3x3:

Replaces value of a pixel by the median grey scale value of the ranked values of the 9 neighbourhood pixels.

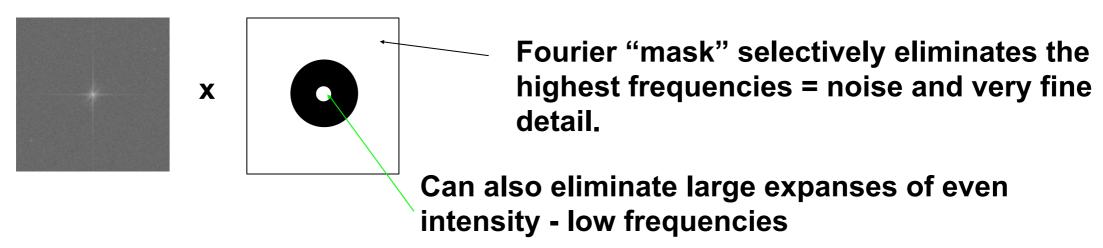


rigure 3.32 The mechanics of spatial filtering. The magnified drawing shows a 3 × 3 mask and the image section directly under it; the image section is shown displaced out from under the mask for ease of readability.

Gonzales & Woods, 2002. Digital Image Processing 2nd Ed. Prentice-Hall Inc, USA.

LOW LEVEL Processing - Simple filter, Fourier domain

 Frequency domain - images converted to Fourier space e.g. Noise reduction using a low pass or band pass filter



F(u, v)

Filter

function

H(u, v)

H(u, v)F(u, v)

Inverse

Fourier

transform

Works on the whole image at once

processing Gonzales & Woods, f(x, y)2002. Digital Image Input Processing 2nd Ed. image Prentice-Hall Inc.

Pre-

Fourier

transform

USA.

FIGURE 4.5 Basic steps for filtering in the frequency domain.

Post-

processing

g(x, y)

Enhanced

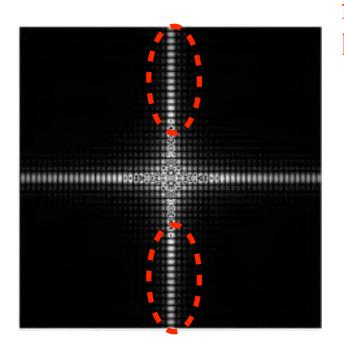
image

Fourier Space = Frequency space = Reciprocal space = K space

Data is broken down into its "frequency" components real space reciprocal space

fine detail = high frequencies 1D case Boxcar

2D case



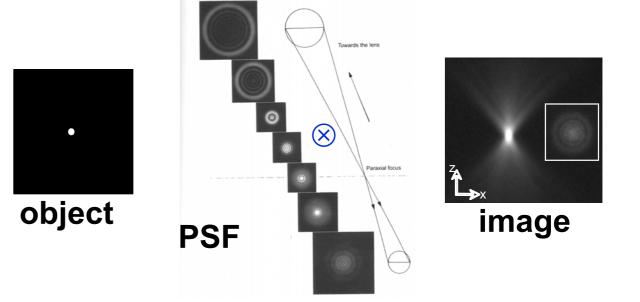
fine detail = high frequencies

LOW LEVEL Processing - Deconvolution

Blur = out of focus information

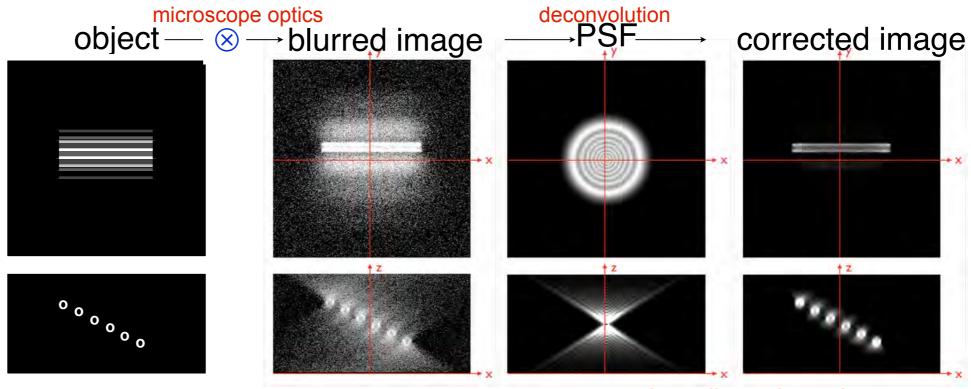
An image represents the output of the optics and detector of the imaging system

image ≠ object ⊗ PSF



Deblurring ≈ Deconvolution = out of focus information removed or

reassigned



Different classes of deconvolution

 Deblurring: nearest neighbours / no neighbours / unsharp mask Not true deconvolution, subtractive (throws away light)
 Quick and easy

Image restoration: (inverse filter) constrained iterative algorithms
 True deconvolution, light is re-assigned to its point of origin.
 Can use measured (empirical), theoretical or derived (blind) PSF

The PSF and OTF

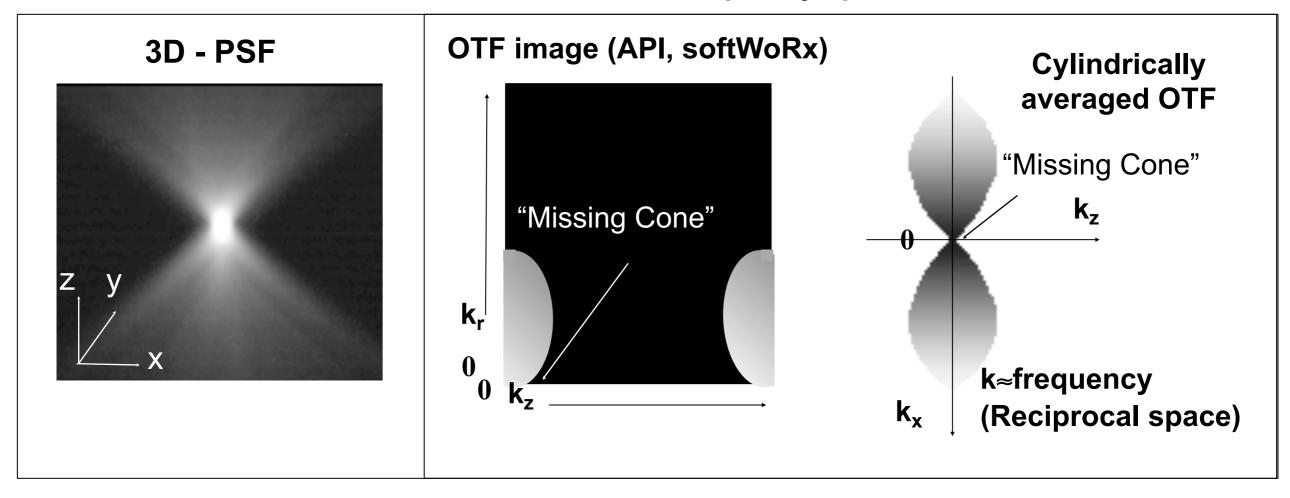
But in Fourier space (or frequency space)

- Calculations for deconvolution are done in Fourier space (simpler / faster)
- The PSF is Fourier transformed to the OTF (optical transfer function)
- The inverse calculation to obtain a "true" image = a linear inverse filter

The PSF and OTF

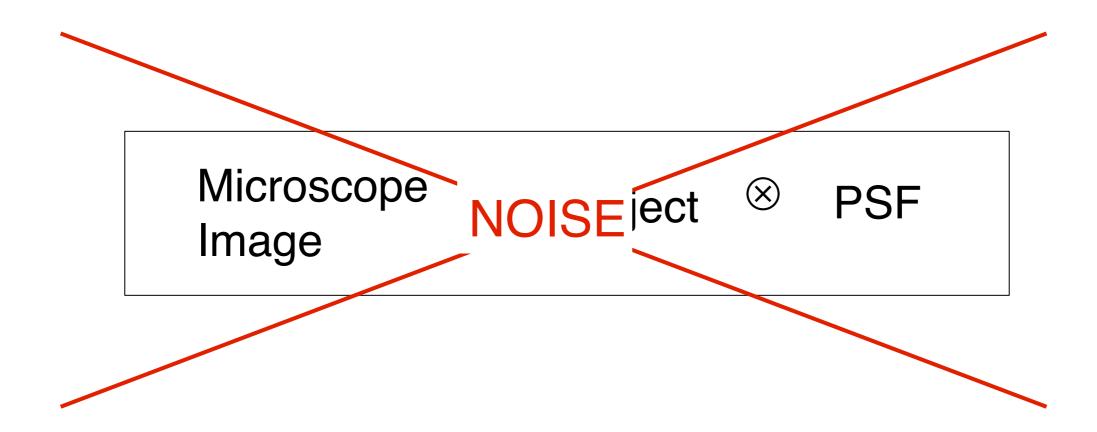
Real space

Frequency space



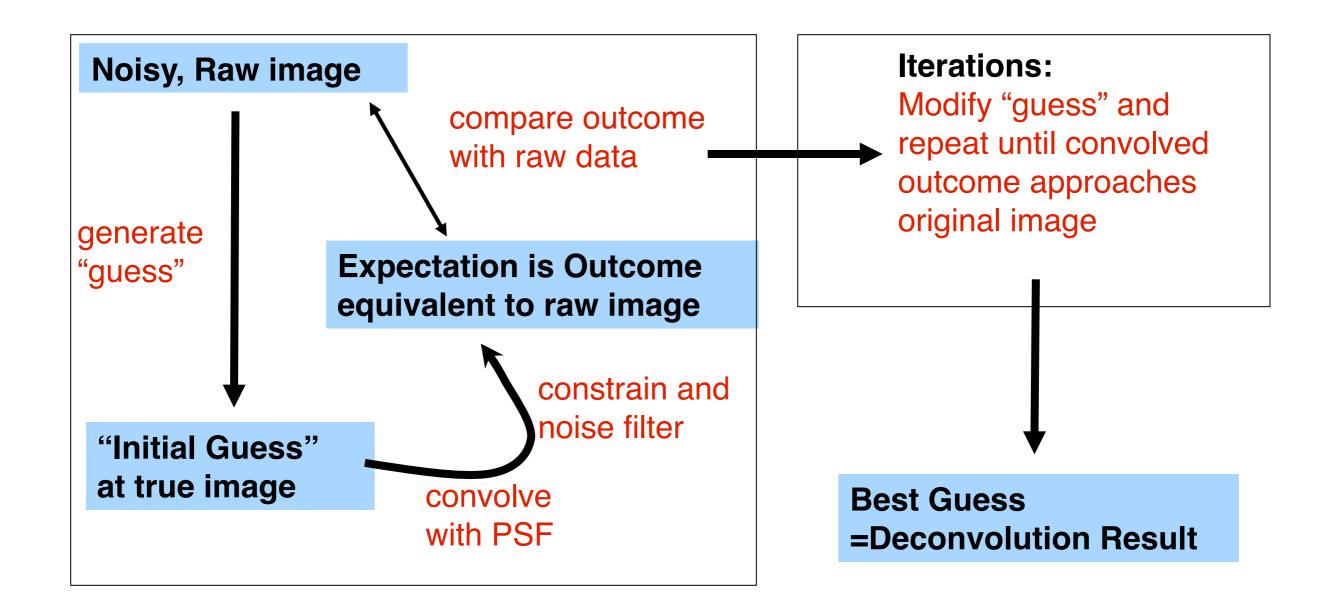
- To simplify deconvolution calculations the OTF is often simplified by assuming radial symmetry for the PSF
- The "missing cone" represents missing frequency information not collected by an objective with a limited NA (not all light can be gathered)

Deconvolution and Noise



Much of the complexity of deconvolution is a direct consequence of having to deal with "noise" inherent in image data

Constrained iterative deconvolution



Data requirements for deconvolution

- Ensure that the imaging system is correctly set up, aligned and calibrated (pixel sizes).
- Reduce aberrations if possible:
 Sample preparation and correction of optics
 Collect more light / average images
 Aim for a high S/N in image data.
- Make sure that images are collected according to the Nyquist sampling criteria (pixel size in XY and Z step).
- Collect sufficient image planes in Z.
 (2D data can be deconvolved but lacks Z information so restoration is limited.)
- Minimise lamp flicker between Z sections. (corrected for on the DV system)
- Avoid motion blur from live specimens. (short exposure times)

Has deconvolution worked?

- Should look sharper and more contrasted and not excessively noisy
- Should NOT "invent" features not visible in the original data
- Be wary of very small punctate features
- Be wary of "ringing" artefacts dark circles round structures

LOW LEVEL Processing - Making Figures



LOW LEVEL Processing - Figure Making Guidelines

 Carry out all processing and analysis of images before making figures by using pixel based (raster) programs:

Handling of images (tif files) for figures should use pixel based (raster) programs
 Arranging multi-panel figures

Before using vector graphics based programs





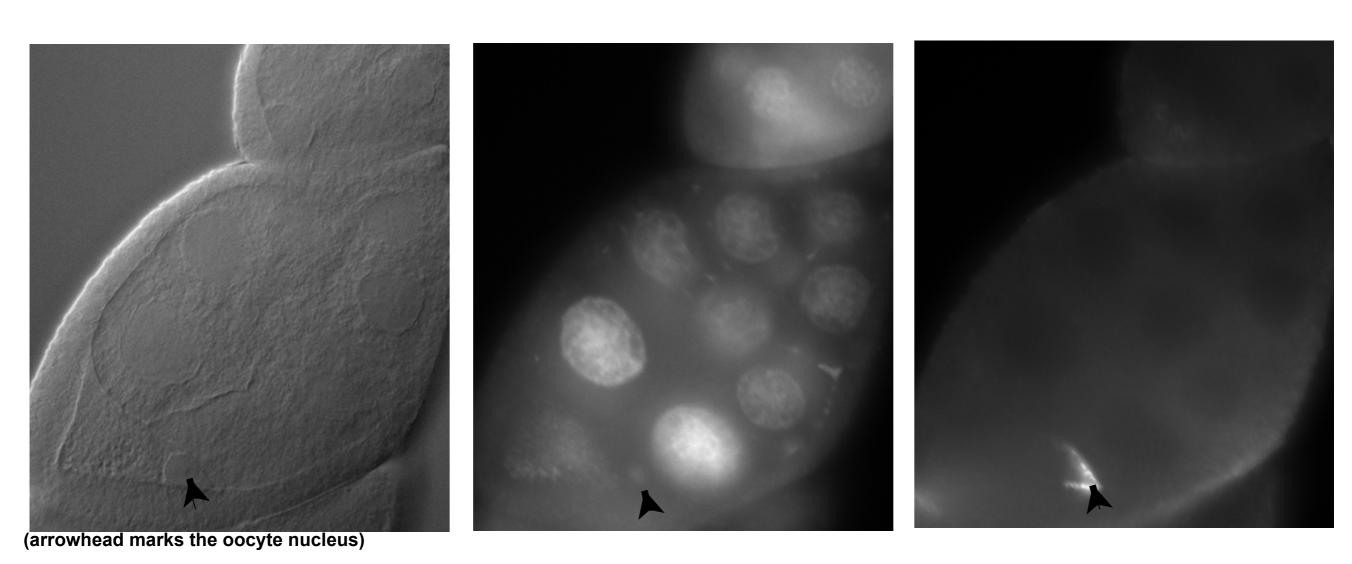
Adding lettering, arrows, charts diagrams....

Cropping, resizing, rotating

- Do not prepare figures in powerpoint or keynote or by screen capture
- Understand what happens when you resize an image
- Be consistent with processing steps, especially contrasting

LOW LEVEL Processing - Display, Grey Scale

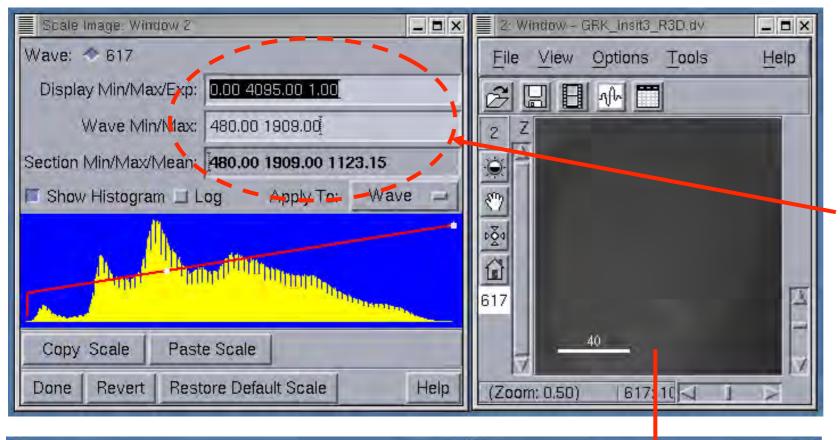
DIC/DAPI/Grk in situ - grey scale images



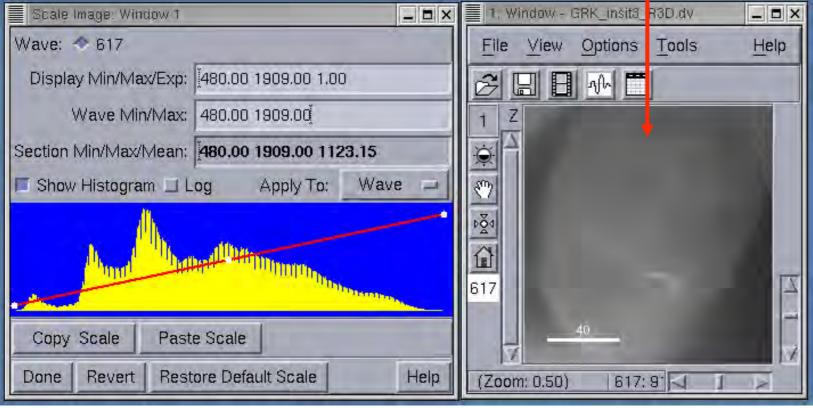
• For viewing and display use grey scale scale images to see fine detail.

LOW LEVEL Processing - Display, Brightness/Contrast

 Brightness and contrast - Enhancing details which are too close in grey level to be easily discernible.



Better to set your contrast by numbers to make the operation more systematic 0-4095 is full range for 12 bit



Auto-intensity scaling by softWoRx scales the display to fill the dynamic range

LOW LEVEL Processing - Display, Bit Depth (levels)

- "grey levels" = the number of discrete values in an image
- **Dynamic range = the number of possible grey levels**
- Imaging detectors:

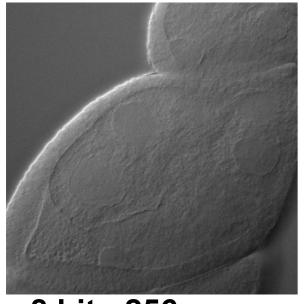
8 bit = 2^8 = 256 grey levels

12 bit = 2^{12} = 4096 grey levels

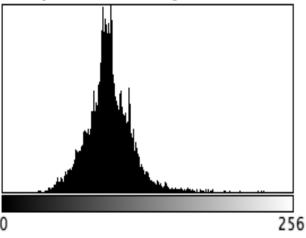
16 bit = 2^{16} = 65536 grey levels

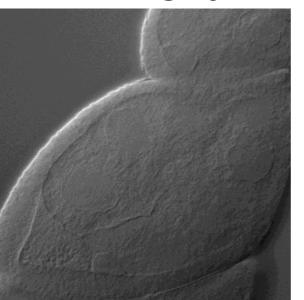
Eye has limited ability to distinguish grey levels/colours Above 32 grey levels images look smooth - 16 and below grey levels eye perceives objectionable banding = false contours.

False contouring due to insufficient grey levels

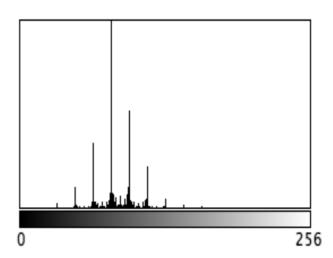


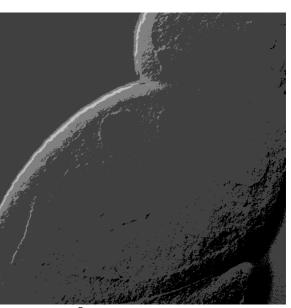
8 bit - 256 greys dynamic range filled



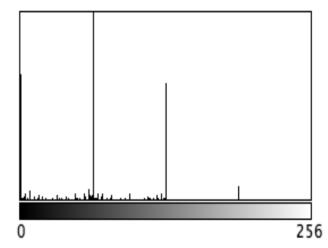


16 greys

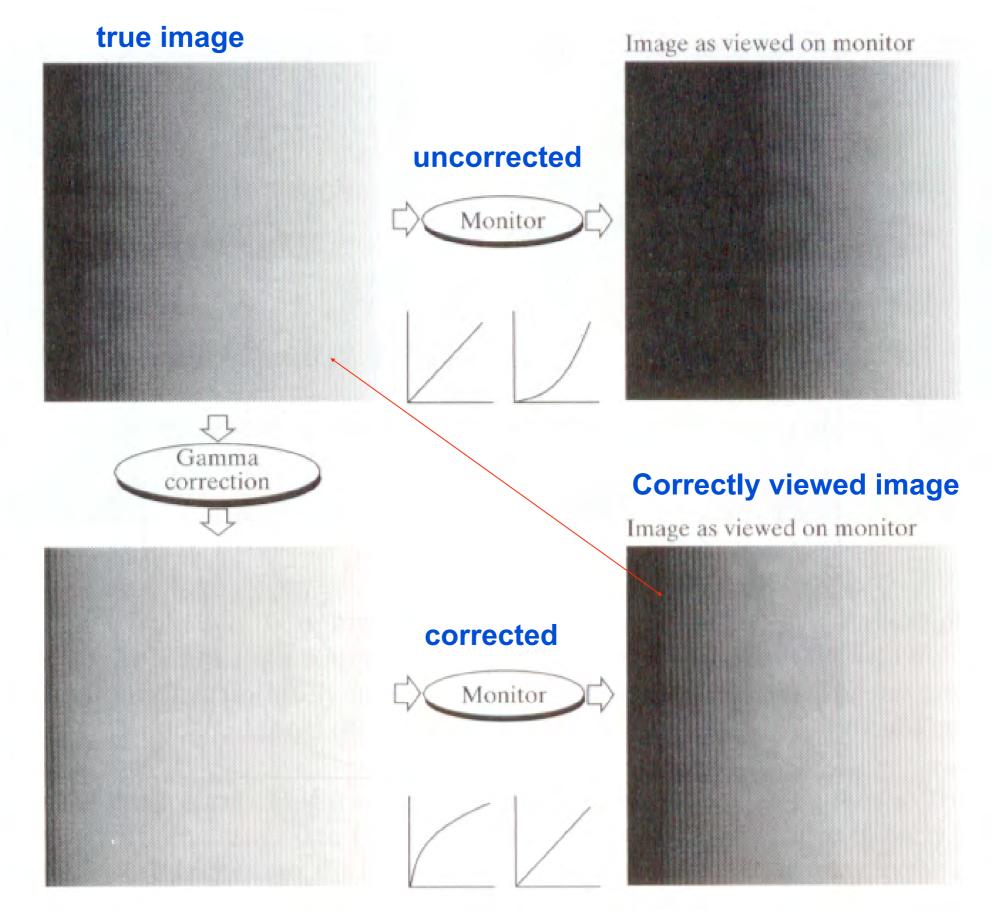




4 greys



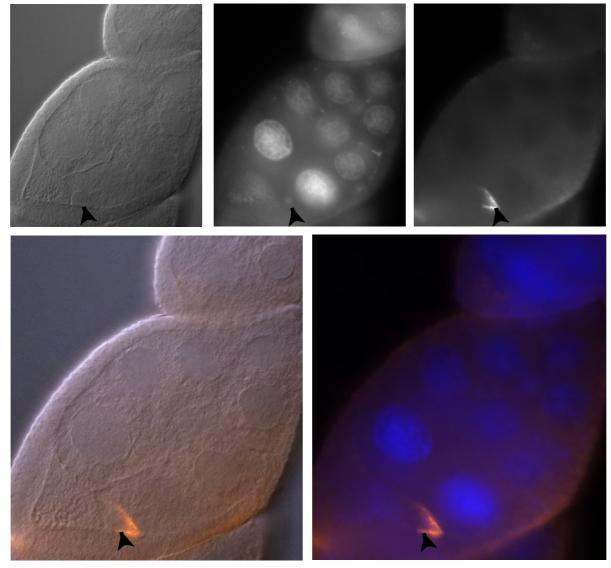
LOW LEVEL Processing - Display, Using Gamma



LOW LEVEL Processing - Display, Colour

DIC/DAPI/Grk in situ - grey scale images, colour blended, additive overlays

Look-Up Tables (LUTs) map intensity to color

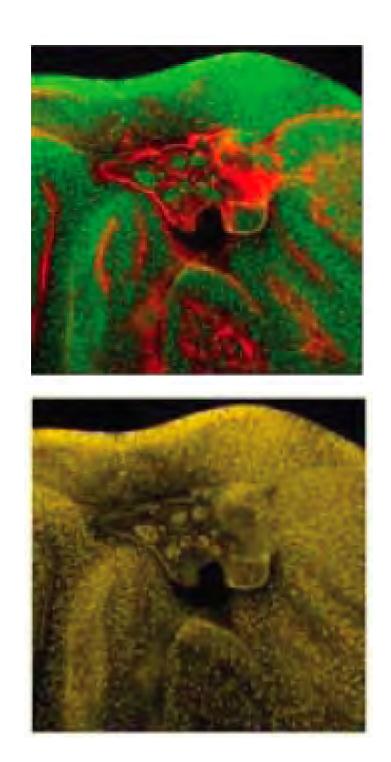


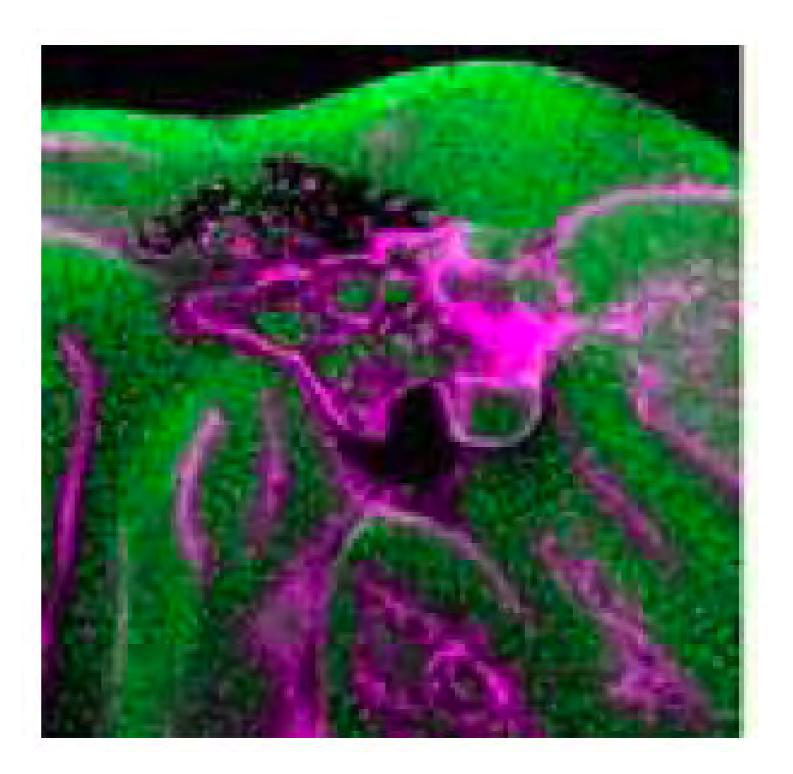
(arrowhead marks the oocyte nucleus)

- Colour should be used for highlighting particular intensity differences / co-localisation.
- For publication show greyscale images alongside colour overlays.

LOW LEVEL Processing - Display, Colour

Consider colour blind friendly colours: magenta, green





LOW LEVEL Processing - Display, Making Movies

- Movie formats: .avi; .mov; (.mpeg)
- Considerations:

image quality vs movie size (use of compression).

speed of play (frames / second).

speed of play on the computer may be slowed by large movies

ImageJ

Will open many file formats and export straight to .avi or .mov., also can open tif image series (image001.tif; image002.tif; etc) restack and export to .avi

Quicktime Pro 7

Can open tif image series and export to .avi or .mov Can interconvert movie formats.

Has a range of compression options for .mov

Bad Imaging Practices

Do not corrupt the integrity of the original data

Examples of Bad Imaging Practices

Rossner & Yamada (2004). What's in a picture? The temptation of image manipulation. J. Cell Biology 166: 11–15.

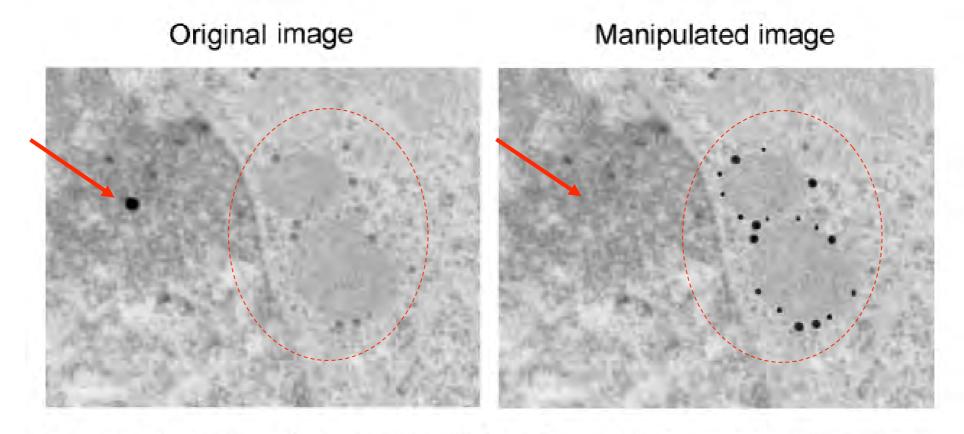
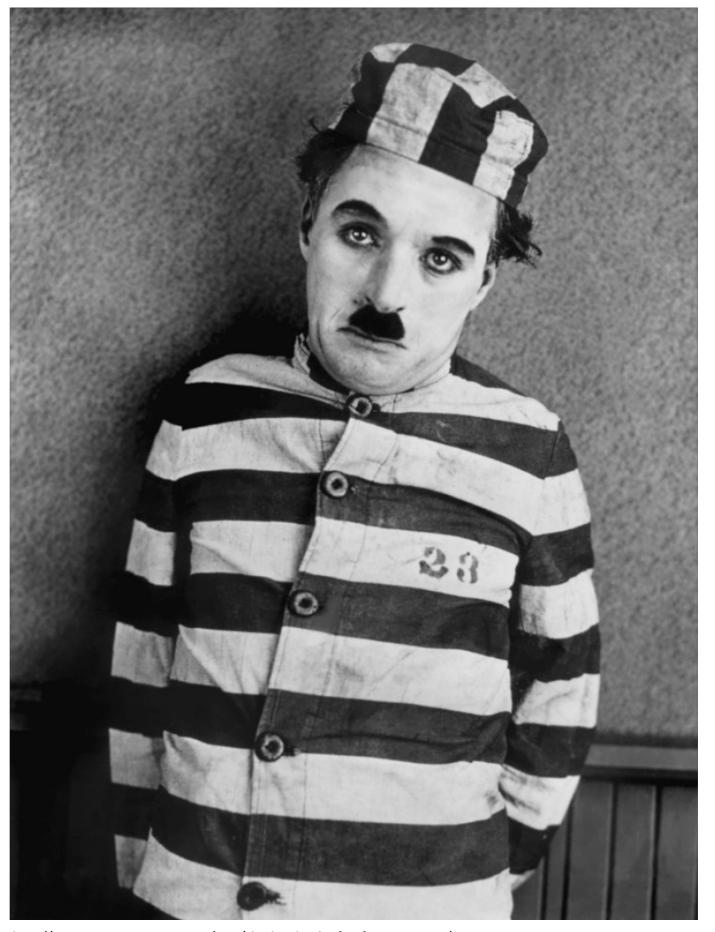


Figure 5. **Misrepresentation of immunogold data.** The gold particles, which were actually present in the original (left), have been enhanced in the manipulated image (right). Note also that the background dot in the original data has been removed in the manipulated image.

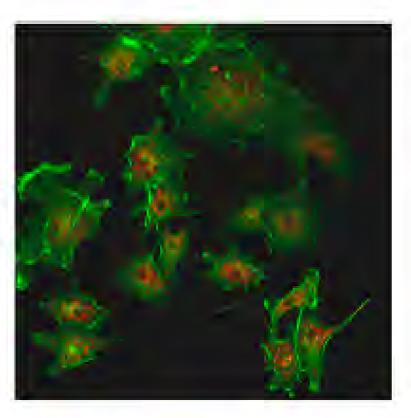
- BAD: manipulated but does not alter interpretation
- VERY BAD: Changes interpretation with intention to defraud
- Adjustments necessary to reveal a feature ALREADY PRESENT in the original data are acceptable if they can be justified

THEY HAVE **WAYS** OF **FINDING** OUT **WHAT** YOU DID!

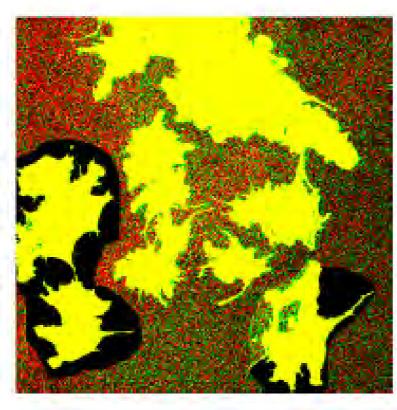


http://www.rottentomatoes.com/quiz/charlie-chaplin-fun-facts-1177891/

Manipulated image

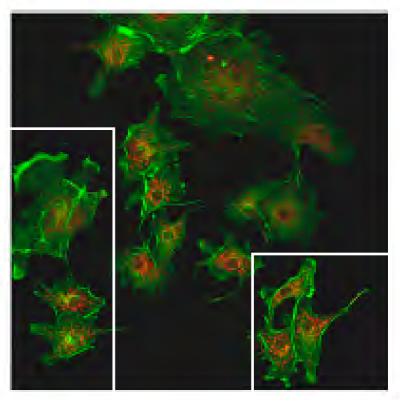


Manipulation revealed by contrast adjustment



- Miss-representation of cell population within an oberved field
- VERY BAD: Changes interpretation with intention to defraud

Correct:

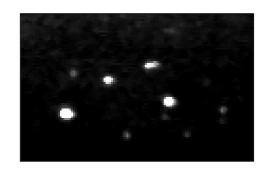


66

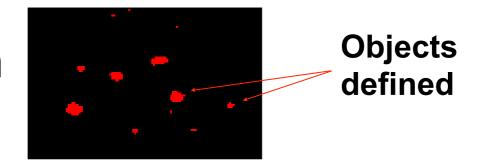
Rossner & Yamada (2004). What's in a picture? The temptation of image manipulation. J. Cell Biology 166: 11–15.

MID LEVEL Processing

 Input is an image, output is an attribute extracted from the image

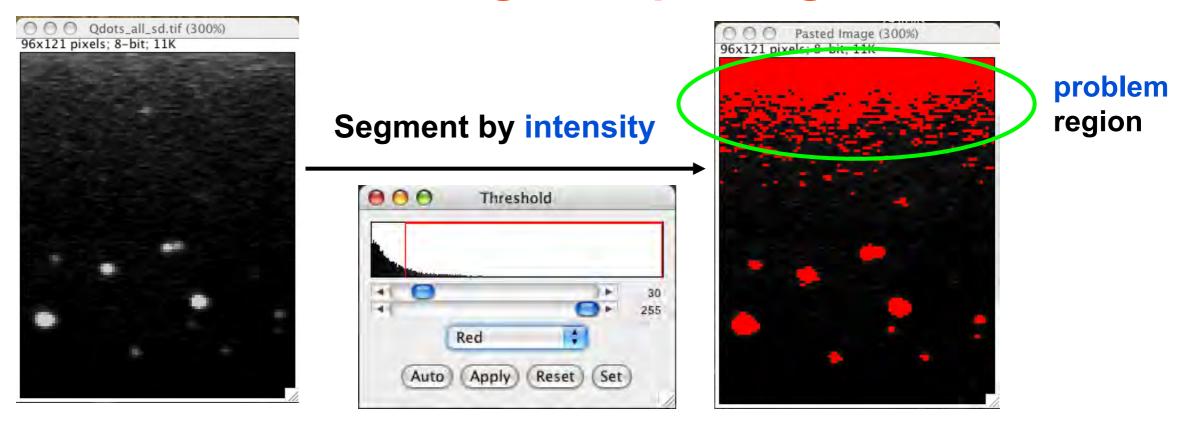


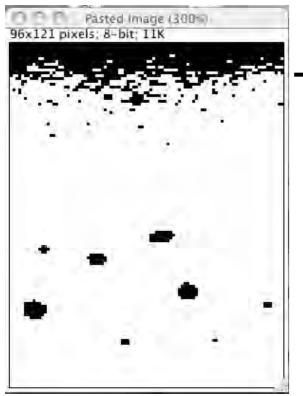
Segmentation



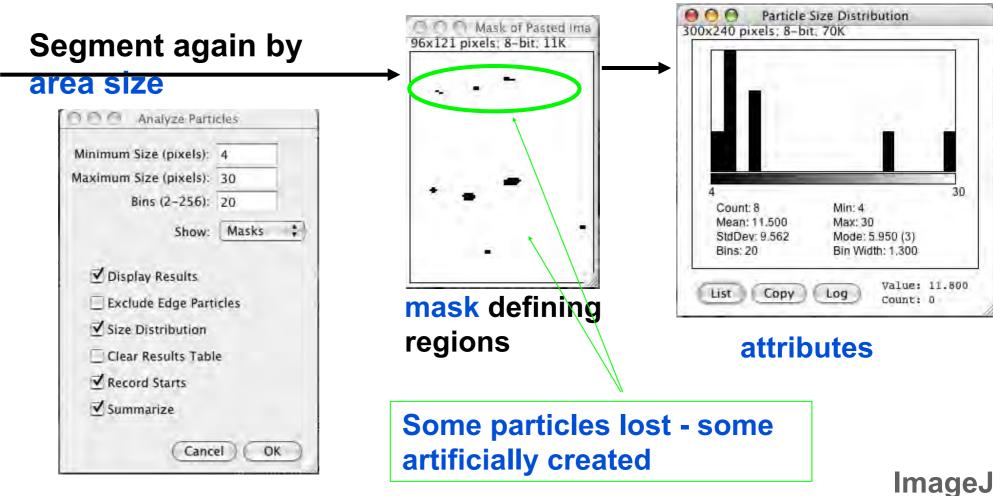
Number of objects Positions of objects Size of objects

MID LEVEL Processing - simple segmentation





Intensity segmented binary image



The human brain is still one of the best segmentation tools

But

Subjective not objective!



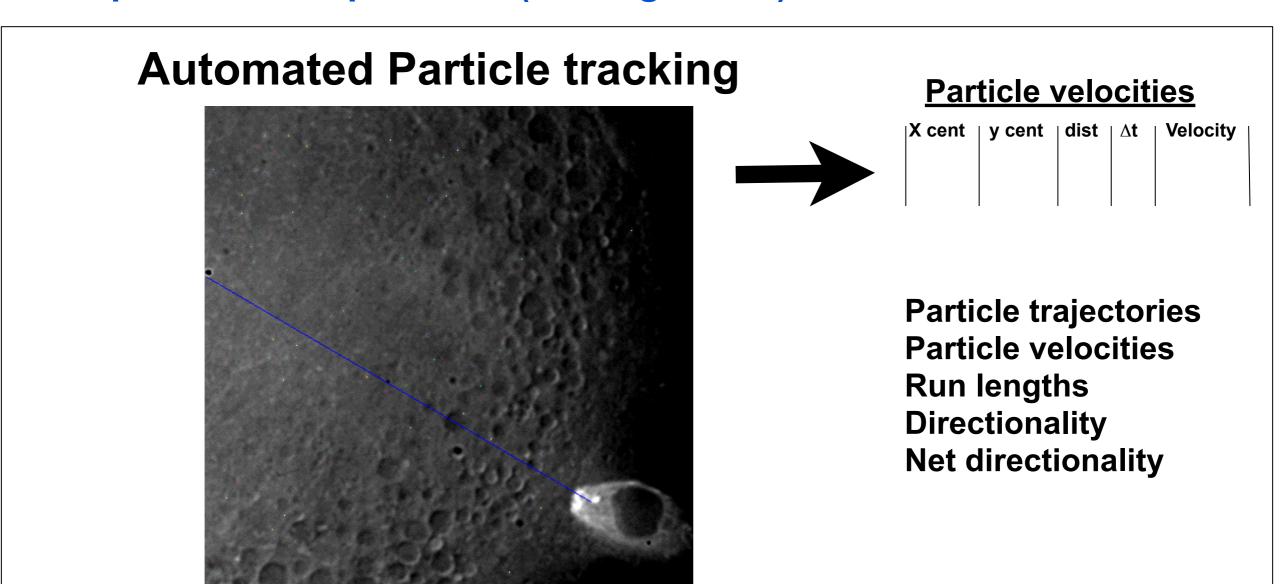
http://www.websitepulse.com/blog/uploads/man_with_magnifying_glass.ipg

Automated segmentation desirable for data quality and sanity

- * Dominic, Lecture 17: Applied Image Analysis *
- * Practical course Demo of Fiji plugins and macros *

HIGH LEVEL Processing = Image Analysis

Outputs are interpretation (making sense)



HIGH LEVEL Processing = Image Analysis

- * MACHINE LEARNING Dominic, Lecture 17: AppliedImage Analysis *
- * Practical course Scripting and Automation: Demo of Fiji plugins and macros *



END



© Warren Photographic http://www.warrenphotographic.co.uk/03360-silver-spotted-cat-looking-back

Reference Material

Fundamentals of light microscopy and electronic imaging Douglas B. Murphy. Wiley-Liss 2001 ISBN 0-471-25391-X

http://www.biology.uoc.gr/courses/BIOL493/documents/book.pdf

* Very good general book for understanding the principles of microscopy.

Gonzales & Woods, 2002. Digital Image Processing 2nd Ed. Prentice-Hall Inc, USA.

* Excellent book to get into image processing, there is also a version of the book which takes you through the use of Matlab for image processing

References for denoising:

Boulanger, J., Kervrann, C., Bouthemy, P., Elbau, P., Sibarita, J.-B., & Salamero, J. (2010). Patch-based nonlocal functional for denoising fluorescence microscopy image sequences. IEEE Transactions on Medical Imaging, 29(2), 442–454. doi:10.1109/TMI. 2009.2033991

Carlton, P. M., Boulanger, J., Kervrann, C., Sibarita, J.-B., Salamero, J., Gordon-Messer, S., et al. (2010). Fast live simultaneous multiwavelength four-dimensional optical microscopy. Proceedings of the National Academy of Sciences of the United States of America, 107(37), 16016–16022. doi:10.1073/pnas.1004037107