

# Bioimage Analysis

Basics of Image - anatomy of an image



# What is an image

Image is an array of information.

Let's have a look at Fabio. He'll be with us for for a while as our 'example' image.

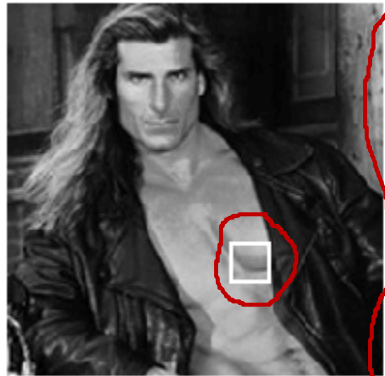


# Fabio is just numbers

Image is an intuitive way to show 2D data.

This data could be for example, intensity of fluorophore at given part of the cell, or it could be Leonardo's Mona Lisa.

Or it could be Fabio.



⇔

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158 169 178 185 183 177 174 177 181 179 184 187 188 189 193 195 195 193 204 186 103
147 165 173 180 181 177 175 176 178 179 183 187 190 192 193 195 195 190 200 191 116
140 155 164 175 178 177 174 173 173 177 179 185 190 192 193 193 194 187 192 187 123
135 138 154 170 176 172 168 169 172 175 177 182 187 189 189 189 188 187 183 129
127 130 144 159 166 168 167 168 169 173 176 179 183 185 185 185 184 184 180 176 131
124 130 136 144 152 158 163 163 162 167 170 173 174 177 180 180 178 173 168 163 124
127 124 131 138 141 144 149 154 157 157 161 164 164 168 173 175 172 163 159 153 117
141 127 129 131 134 137 141 145 148 147 150 154 157 159 163 164 159 151 145 141 111
163 146 130 119 123 132 136 138 142 141 144 148 149 147 151 153 151 146 139 133 103
179 163 141 122 121 128 132 134 135 133 137 142 142 139 143 148 147 138 132 127 94
183 173 163 145 128 122 126 130 130 134 136 141 141 137 138 141 139 129 127 123 85
178 181 179 166 143 125 119 121 123 124 126 130 133 133 135 135 132 128 125 113 71
165 180 181 178 166 147 128 118 116 120 119 122 126 128 130 129 125 119 117 98 59
156 168 174 180 180 170 152 134 122 117 112 109 108 105 104 102 98 93 96 79 49
157 157 168 176 177 175 170 155 139 128 121 117 114 111 112 113 111 118 125 107 75
152 147 156 165 175 181 175 162 154 155 147 146 150 150 149 156 164 170 171 167 122
143 143 153 163 171 178 176 168 162 165 165 171 175 175 175 177 178 177 175 173 132
143 151 162 170 174 177 176 170 166 163 166 171 170 169 174 177 175 170 167 170 138
154 165 176 181 179 177 175 170 168 167 168 168 166 165 172 175 172 170 164 168 143
163 168 179 183 181 179 177 174 174 172 170 172 174 175 178 179 175 174 166 166 145
172 171 178 182 182 181 180 179 181 174 170 170 173 175 178 180 179 174 166 162 143
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# 50 Shades of Gray (and more)

We saw that Fabio's pectorialis major was made from numbers between 80 and 200~ish.

These numbers are actually arbitrary and have to do with the **bit depth** of the image. **Bit depth** explains how many values we have between black and white. We will discuss this later.

The values between black and white are greys, forming a greyscale.



# 50 Shades of Gray

Black and white image (also known as greyscale image) has some white, some black and many kinds of gray in it.

When people began creating first computer screens and printers, there were to things to be considered.

- 1) How many different greys we can show?
- 2) How many different greys humans can see?



# 50 Shades of Gray

Let's take black and add white until we get pure white. Now, let's do these additions in  $x$  steps.

Let  $x$  be 20, 50, 100 and 256.

As you can see, the transitions become smooth between 50 and 100.



# Should we worry about the shades of gray?

Sometimes there is little need to worry. Human eye can only see so much, and if all you want to show is Fabio (or his biological equivalent to him) you can go with whatever the machine does.

On the other hand, if you plan to **quantify** your data, say compare intensities or do colocalization there is all the reason to worry. Because now, the different values are measurements and not just an image.



# Whiter shade of pale

An image on a microscope can be acquired with various **bit depths**. This is to say, we get to decide how many numbers there are between black and white.

Bit depth represents the accuracy of division between white and black. The more bits we have, the more shades of grey we have and the more miniscule differences we can show.

$$8 \text{ bits} \rightarrow 2^8 = \underline{256 \text{ grayscales}}$$

$$12 \text{ bits} \rightarrow 2^{12} = \underline{4096 \text{ grayscales}}$$

$$16 \text{ bits} \rightarrow 2^{16} = \underline{65536 \text{ grayscales}}$$

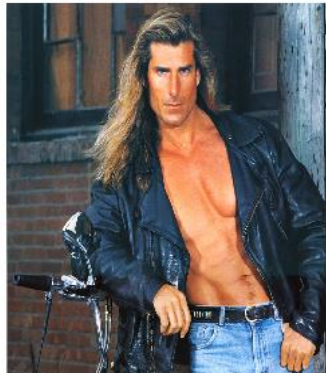
You might recall that already 8 bits will give you a 'smooth' image. The additional bits only matter if we want to know how much signal we are getting out.





# Color images!

RGB stands for red/green/blue and RGB images are basically three greyscale images layered on each other, so that each pixel has three values. This is why we tend to talk about 'green' or 'blue' channel. Let Fabio demonstrate.



Red



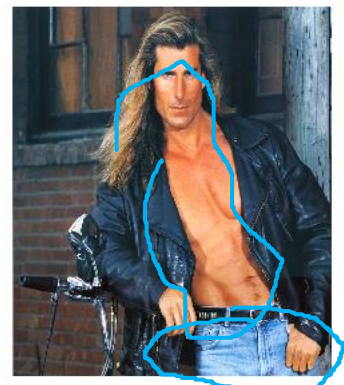
Green



Blue



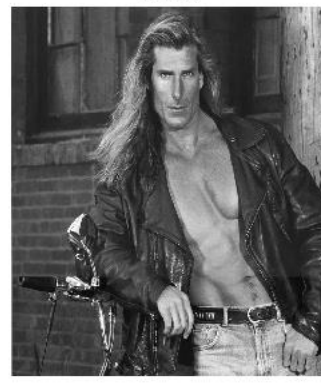
You'll notice that it is hard to see any detail on the blue image.



Red



Green

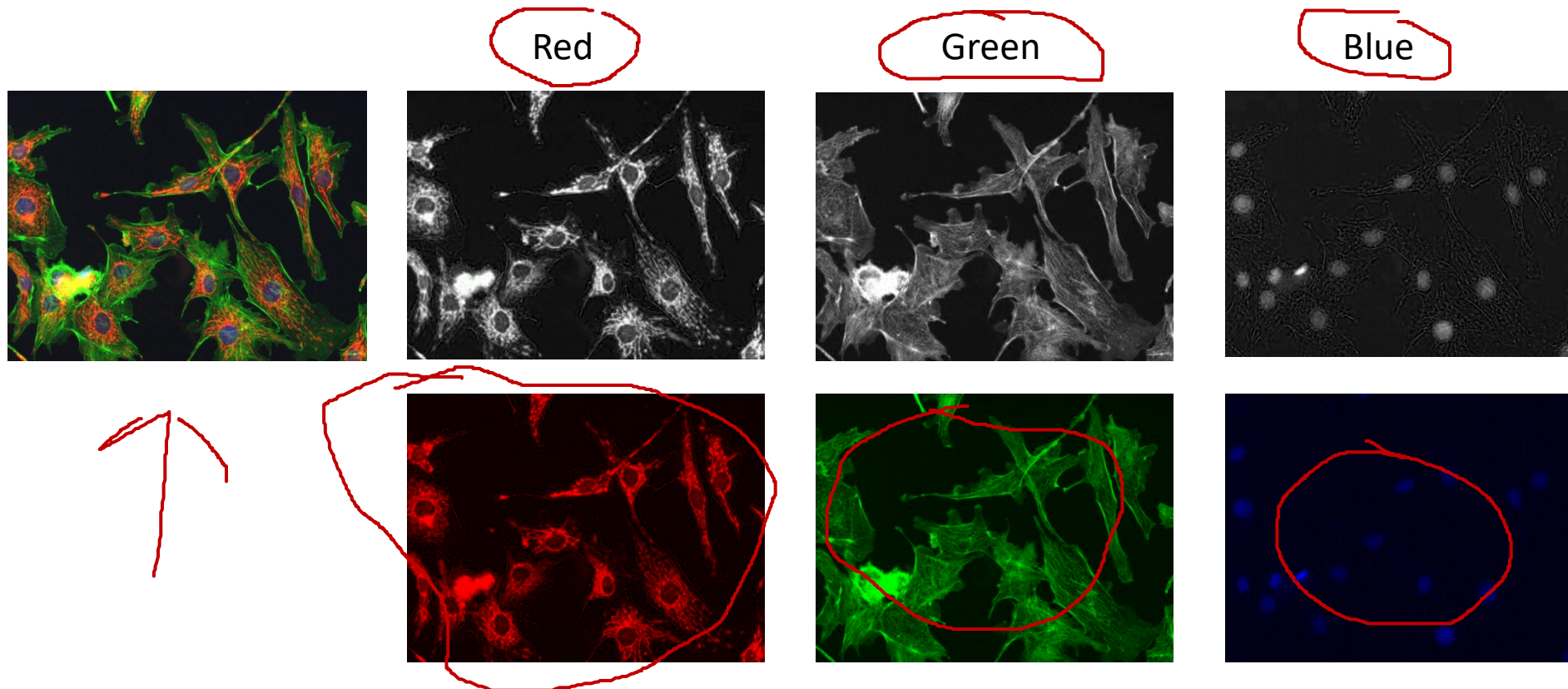


Blue



# Microscopy images are not color images

You might think they are, but they really are not.



Each channel is greyscale image. In the final product these images tinted to a chosen colour and overlaid.

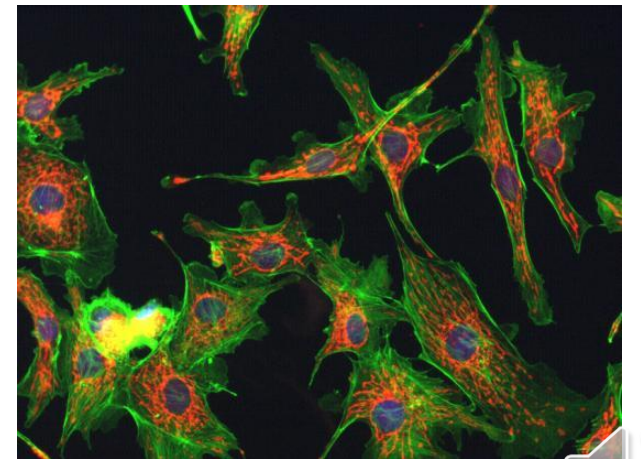
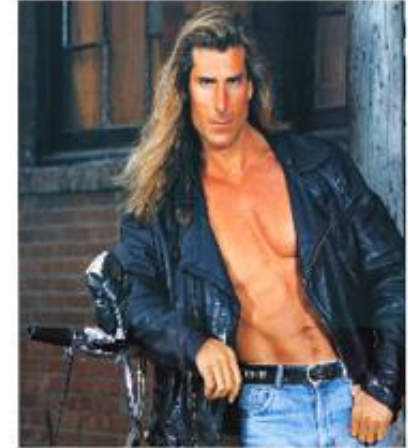


# Picture v.s data

You might be inclined to think that these are both images and in everyday language this is the case.

However the image below is actually data while the image above is a picture.

While you can take pictures haphazardly, a great care should be taken when acquiring data.



# Picture v.s data

The same rules do not apply to pictures and data.

Picture where you cannot see anything is pretty much useless.

Measurement where the computer can see things, even if you cannot see them can still be useful.

Don't judge your data on the criteria you might apply to pictures.



# Resolution and sampling

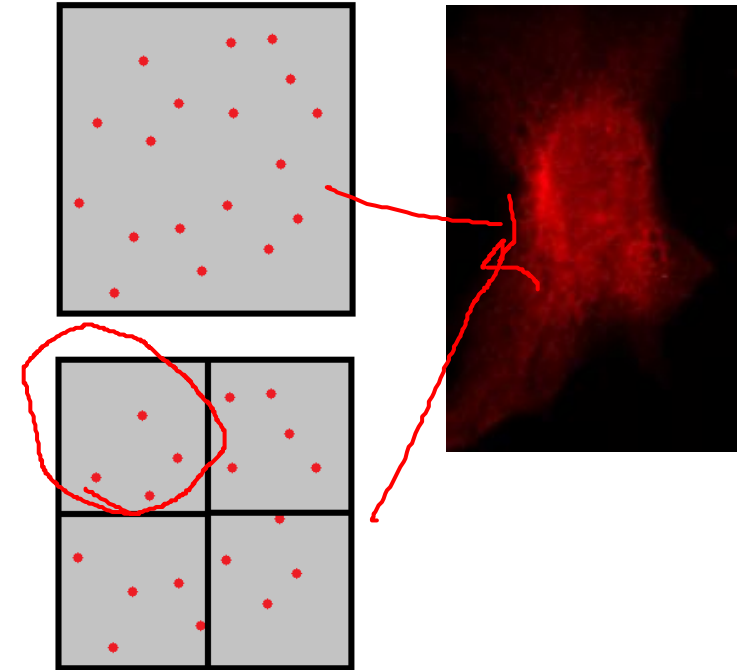


# Resolution

The physical size of your pixels affects the quality of your signal.

The bigger the pixels, the more signal you get  
However, bigger pixels mean also having worse resolution.

This brings us to the matters of **sampling** and **resolution**.



# Sampling

**Sampling** or in layman's terms 'how big pixels you take'.

Optimal sampling = using the optimal pixel size.

Oversampling = using smaller pixels than physically possible.

Undersampling = using larger pixels than the optimal.



Undersampling might sound bad, but there are a lot of cases where it is a valid strategy!



# Sampling

In simplest terms it means 'how many pixels we are using to portray information'. Higher the sampling, higher the resolution.

High resolution Fabio

Low resolution Fabio



As you can see, here the low resolution Fabio is just as fabulous as his high resolution counterpart.





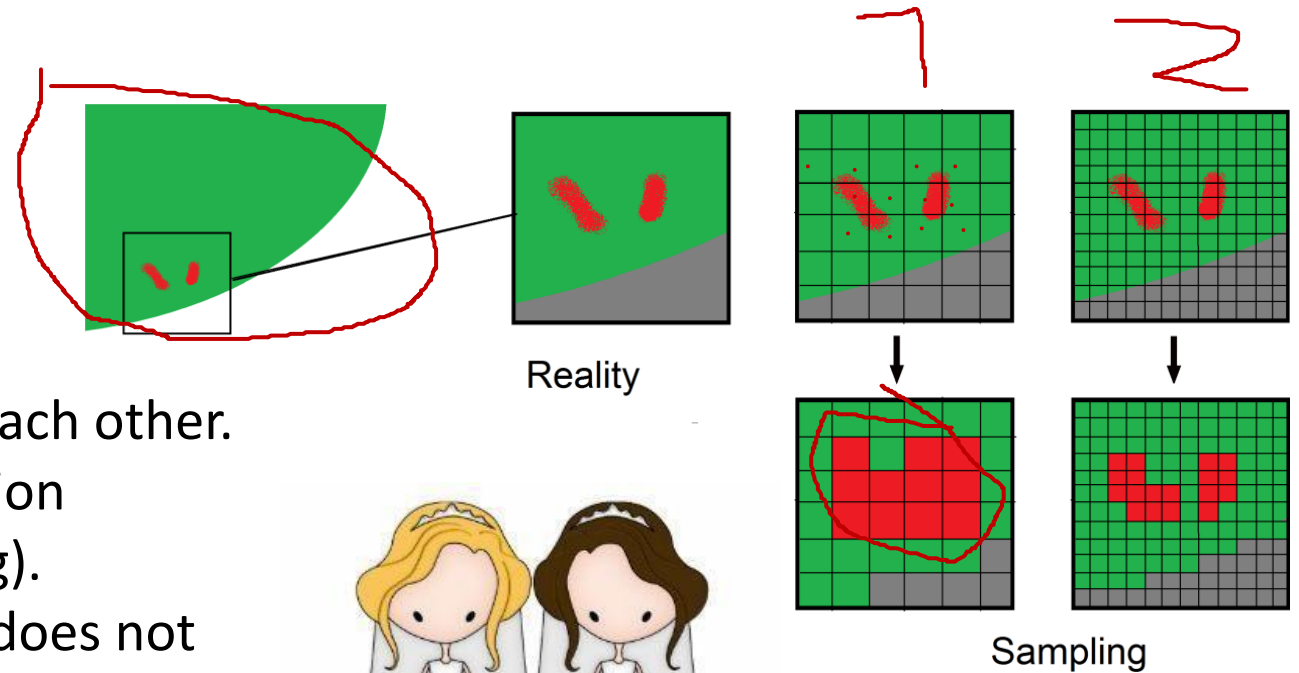
# Resolution

Image size (e.g. 512x512)  $\neq$  resolution

Zoom  $\neq$  resolution

Pixel size and resolution are married to each other.  
You cannot achieve best possible resolution  
without best possible pixel size (sampling).  
However smaller than optimal pixel size does not  
improve resolution.

Ability to differentiate objects from others, that  
is ability to resolve structures = **resolution**



Mrs. Pixelsize & Mrs. Resolution



Resolution = ability to  
resolve structures



# Resolution

Your pixel size might be for example 70 nm. But that doesn't mean your ability to resolve things is 70 nm. The rough equation\* for **lateral** (xy) resolution reads:

$$r = \frac{\overset{\text{---}}{\underset{\text{---}}{\text{---}}}{\text{---}} \lambda}{\text{---} \text{NA} \text{---}}$$

where

**r** = shortest distance between two points on a specimen that can still be distinguished by the observer

**λ** = excitation wavelength for confocals and emission wavelength for widefield systems

**NA** = numerical aperture of your objective

 = multiplication factor that depends on the modality of the microscope

\* This is the rough equation. The exact equation we leave for the physicists.




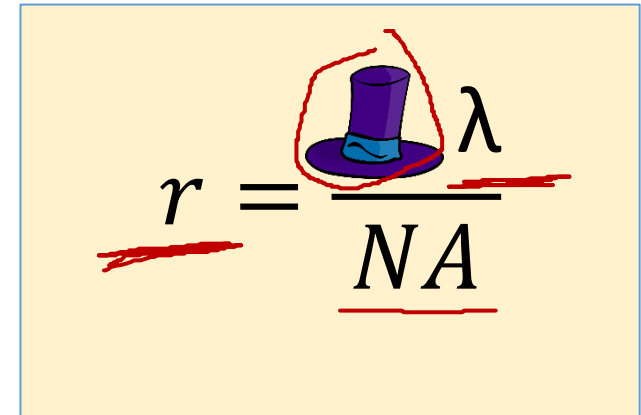
# Resolution

The equation on the previous page has few immediate consequences.

Firstly, bigger the NA, the better the resolution.

Secondly, smaller the hat, the better the resolution.


Usually given values for the  are 0.4, 0.5 and 0.61 depending on the modality.




The equation  $r = \frac{\lambda}{NA}$  is displayed on a yellow background. A red top hat icon is placed above the fraction line. Red underlines are drawn under the  $\lambda$  and the  $NA$  in the denominator. A red circle is drawn around the top hat icon.

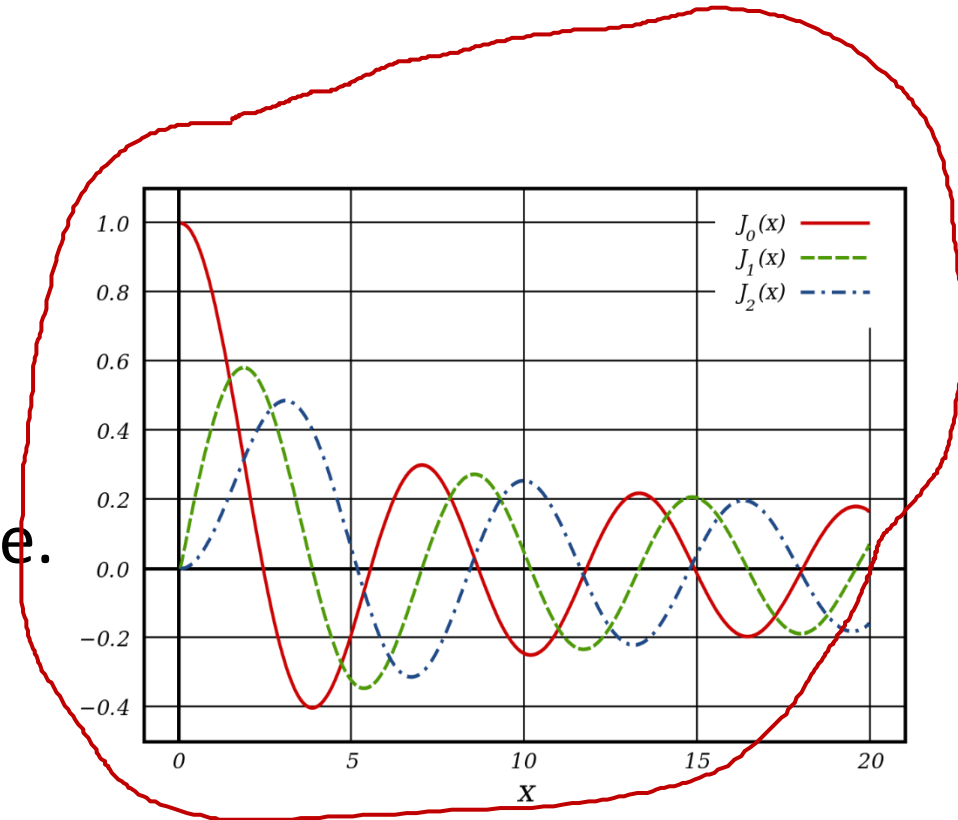


# Resolution

Those different values for  come from theoretical models based on Bessel functions used to model the propagation of light and which are way beyond the scope of this course.

It's not magic, it's physics.

For confocal the  is about 0.4 while for the widefield it's about 0.61



$$r = \frac{\text{top hat icon} \lambda}{NA}$$



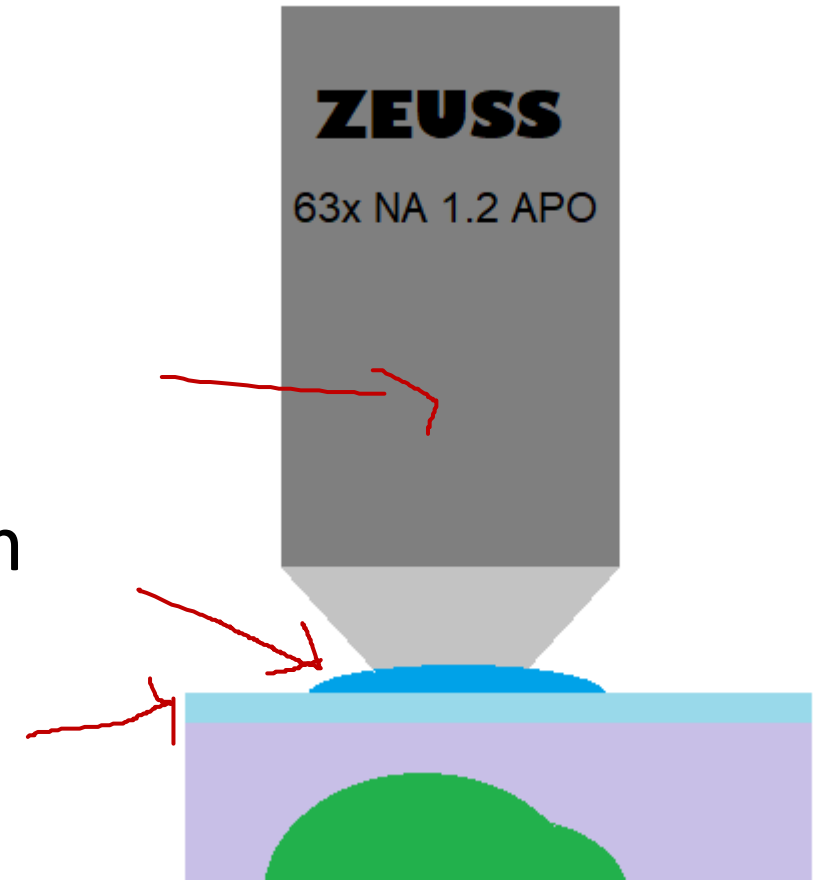
# Resolution

For the z-resolution the equation reads

$$r = \frac{\hat{\lambda} n}{NA^2},$$

where  $n$  Refractive Index of the medium between the lens and the specimen.

For confocals  $\hat{\lambda}$  is about 1.4.



# Resolution

Recalling our equations

$$\underline{r_z = \frac{\hat{\text{red hat}} \lambda n}{NA^2}}, \quad \underline{r_{xy} = \frac{\hat{\text{purple top hat}} \lambda}{NA}}$$

Assuming green 488 nm excitation laser, we can tabulate the following.

	XY resolution	z resolution
<u>Oil 63x 1.4 NA</u>	<u>140 nm</u>	530 nm
<u>Glycerol 63x 1.3 NA</u>	<u>150 nm</u>	600 nm
<u>Glycerol 20x 0.75 NA</u>	<u>260 nm</u>	1800 nm

While oil objectives give nominally better resolution than glycerol objectives, aberrations caused by refractive index mismatch will likely undo any potential gains as long as you are using conventional mounting media.

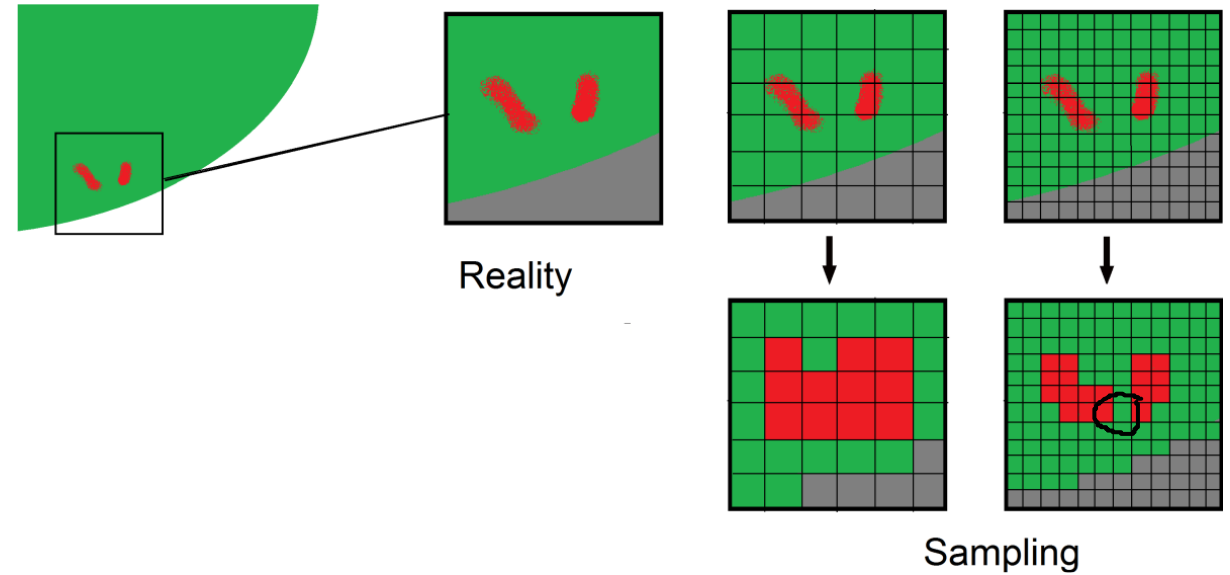
Prolong glass with refractive index 1.52 is notable exception.



# Optimal sampling

However as noted earlier, resolution and sampling are not the same.

You often find a recommendation that your pixel size should be 2.3 times smaller than the optimal resolution if you want to resolve your image as accurately as possible.



	XY sampling	z sampling
Oil 63x 1.4 NA	61 nm	230 nm
Glycerol 63x 1.3 NA	65 nm	243 nm
Glycerol 20x 0.75 NA	113 nm	780 nm





# Questions?

