



Introduction to Medical Image Analysis

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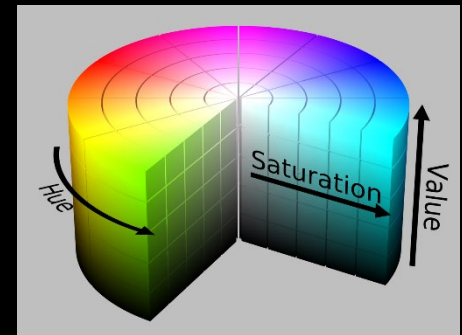
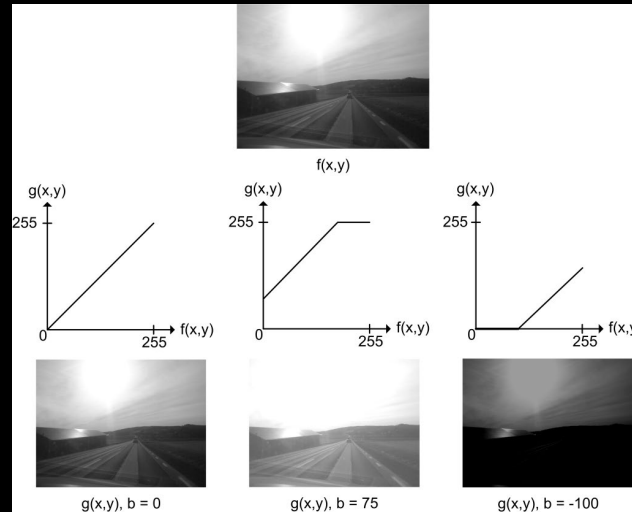
<http://courses.compute.dtu.dk/02512>

Plenty of slides adapted from Thomas Moeslunds lectures



Week 3

Pixelwise operations and colour images





What can you do after today?

- Compute and apply a linear gray transformation
- Describe and compute the image histogram
- Implement and apply histogram stretching
- Implement and apply gamma transformation
- Implement and apply log and exp mappings
- Describe and use thresholding
- Describe and use automatic thresholding
- Perform conversions between bytes and doubles
- Use addition and subtraction of images
- Explain the benefits of bi-modal histograms
- Identify images where global thresholding can be used for object extraction

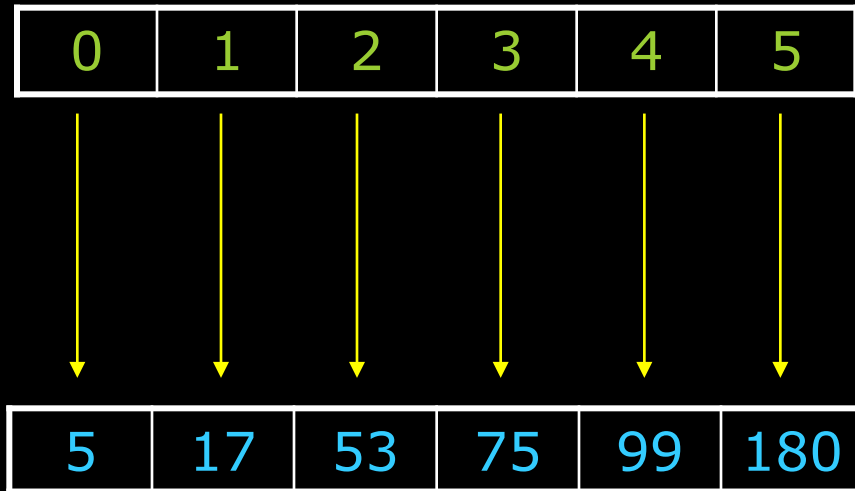


...and you can even more

- Describe the basic human visual system including rods and cones
- Describe subtractive colors
- Describe additive colors
- Describe the RGB color space
- Describe the normalised RGB color representation
- Describe the use of the Bayer pattern in digital cameras
- Describe the HSI color space
- Convert from an RGB to a grey level value
- Convert from an RGB value to an HSI value
- Describe the use of different color spaces
- Implement and use color thresholding in RGB space
- Implement and use color thresholding in HSI space



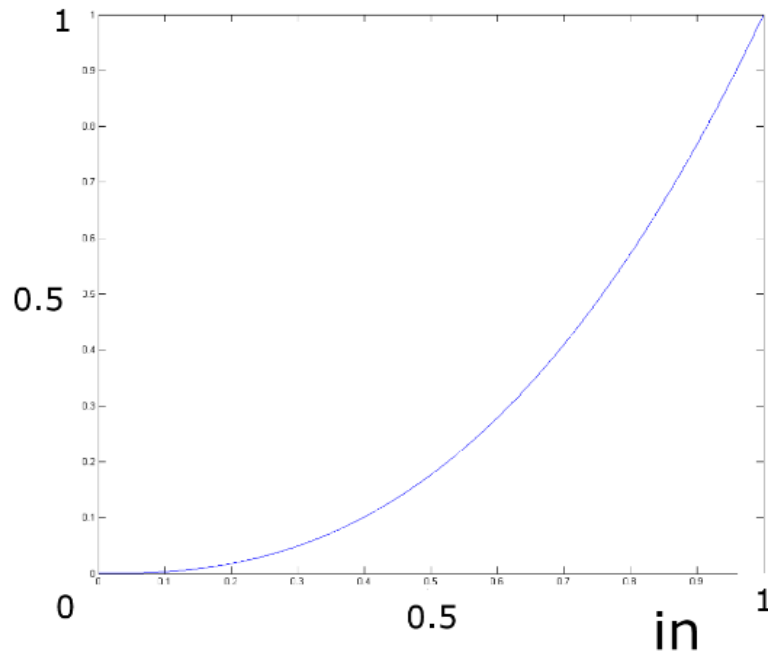
Gray value mappings



- Mapping
 - To make correspondence between two sets of values
- Look-up-table
 - A table of mappings



Mapping Function



$f(0.5)?$

0.1 **A**

0.2 **B**

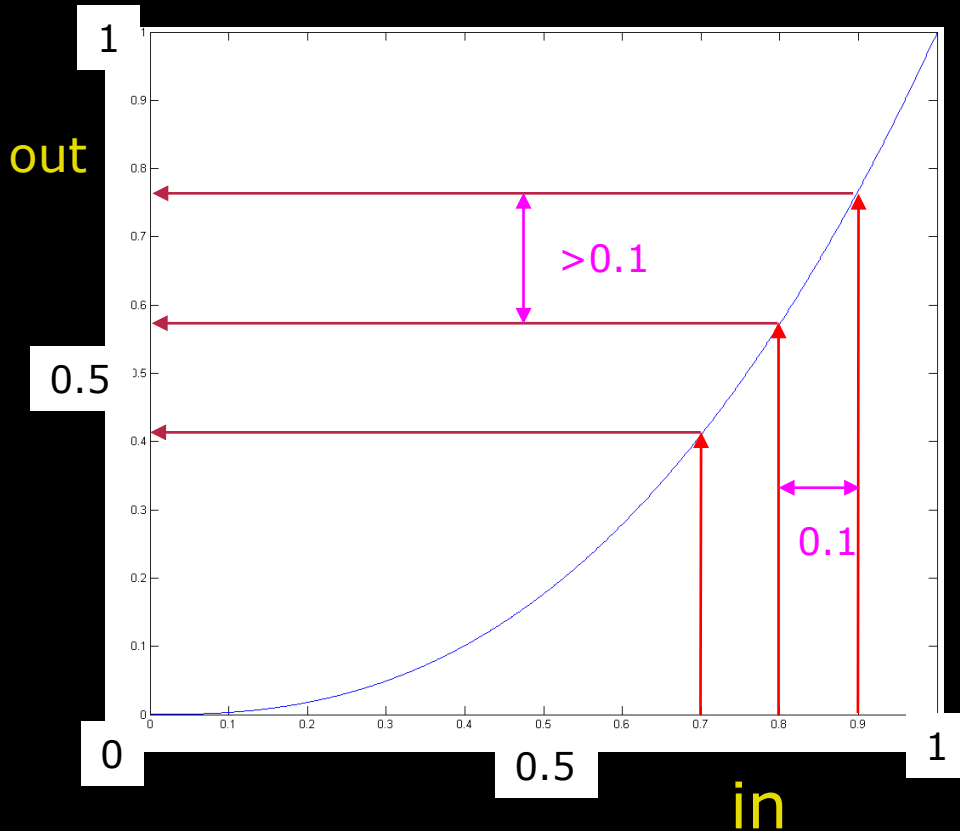
0.3 **C**

0.4 **D**

0.5 **E**



Gray value mappings



- Mapping
 - To make correspondence between two sets of values
- Mapping function
 - $\text{out} = f(\text{in})$
- What happens with the values?
 - Values with difference 0.1
 - Output values "spread out"



Why change gray level values

- When could it be good to change the gray level values?
 - Lack of contrast
 - Make the image lighter
 - Make the image darker



Point processing

Input

1	2	0	1	3	
2	1	4	2	2	
1	0	1	0	1	
1	2	1	0	2	
2	5	3	1	2	

Output

	$\frac{12}{9}$				

- The value of the output pixel is only dependent on the value of one input pixel
- A global operation – changes all pixels



Point processing

■ Grey level enhancement

- Process one pixel at a time independent of all other pixels
- For example used to correct Brightness and Contrast
 - Known from the television remote control

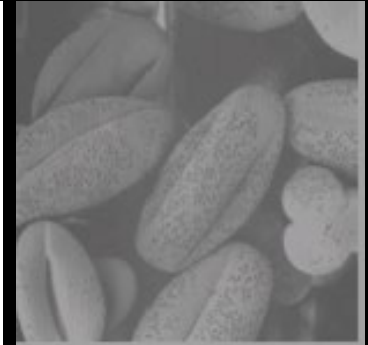
Correct

Too high
brightness

Too low
brightness

Too high
contrast

Too low
contrast





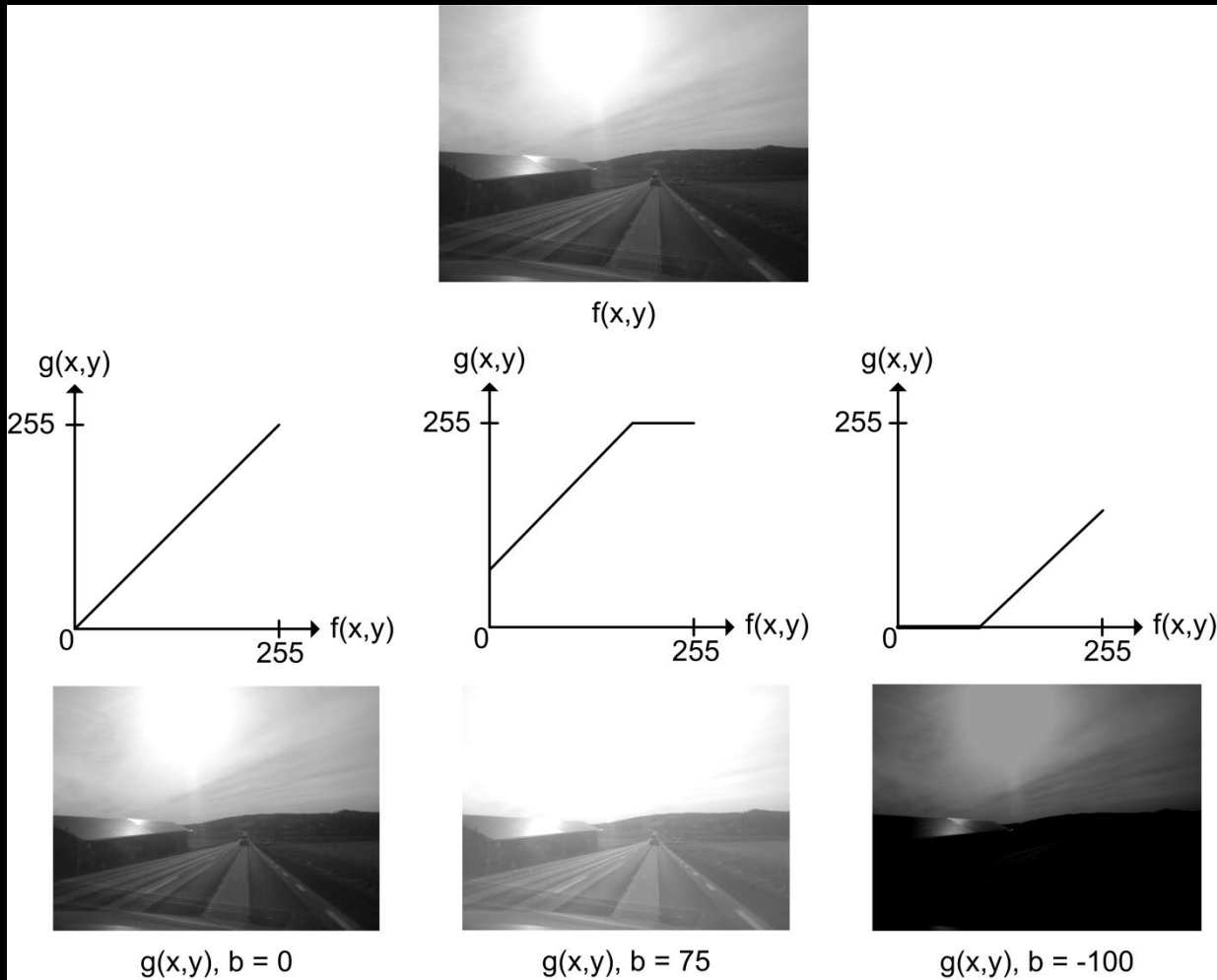
Brightness

- The brightness is the intensity
- Change brightness:
 - To each pixel is added the value b
 - $f(x, y)$ is the input image
 - $g(x, y)$ is the (enhanced) output image
- If $b > 0$: brighter image
- If $b < 0$: less bright image

$$g(x, y) = f(x, y) + b$$



Brightness

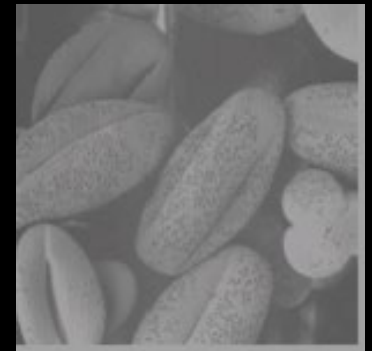




Contrast

- The contrast describes the level of details we can see
- Change contrast
- Each pixel is multiplied by a
 - $f(x, y)$ is the input image
 - $g(x, y)$ is the (enhanced) output image
- If $a > 1 \Rightarrow$ more contrast
- If $a < 1 \Rightarrow$ less contrast

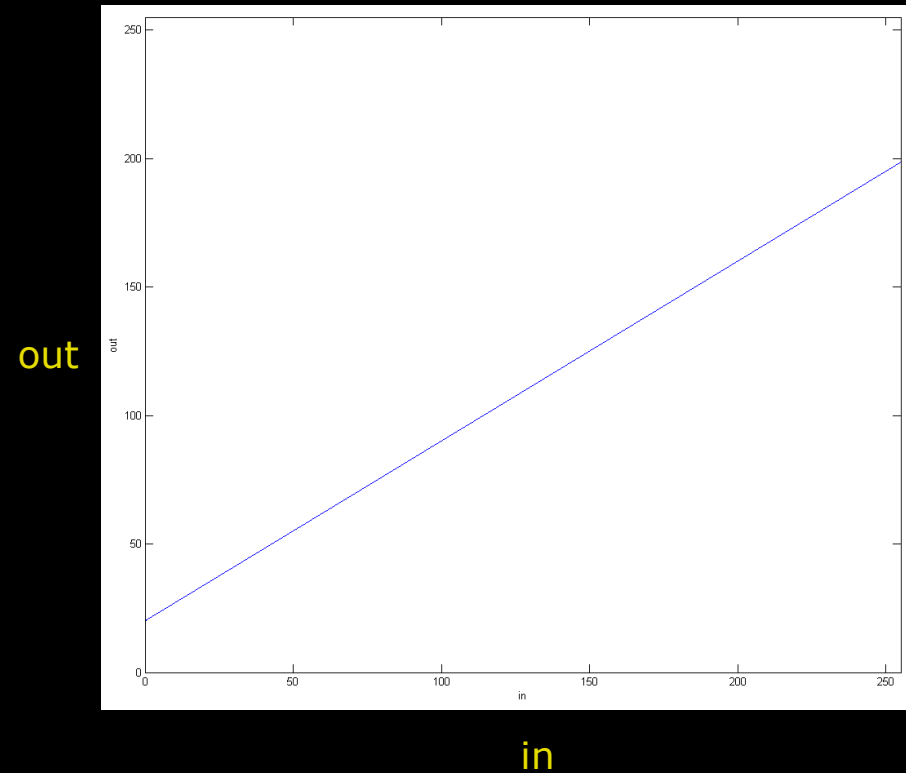
$$g(x, y) = a * f(x, y)$$





Combining brightness and contrast

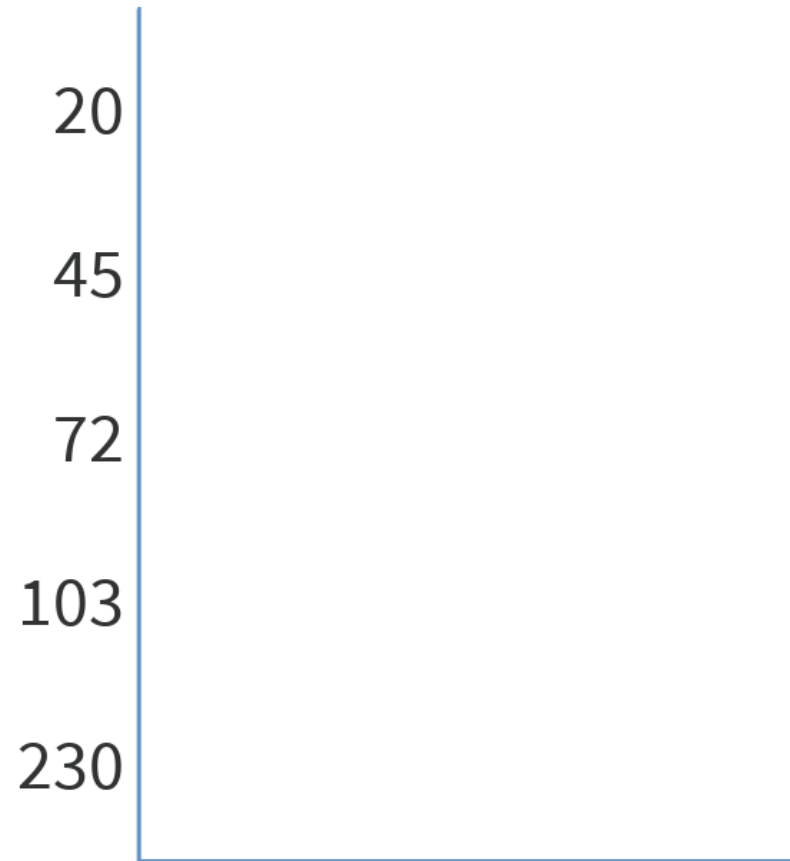
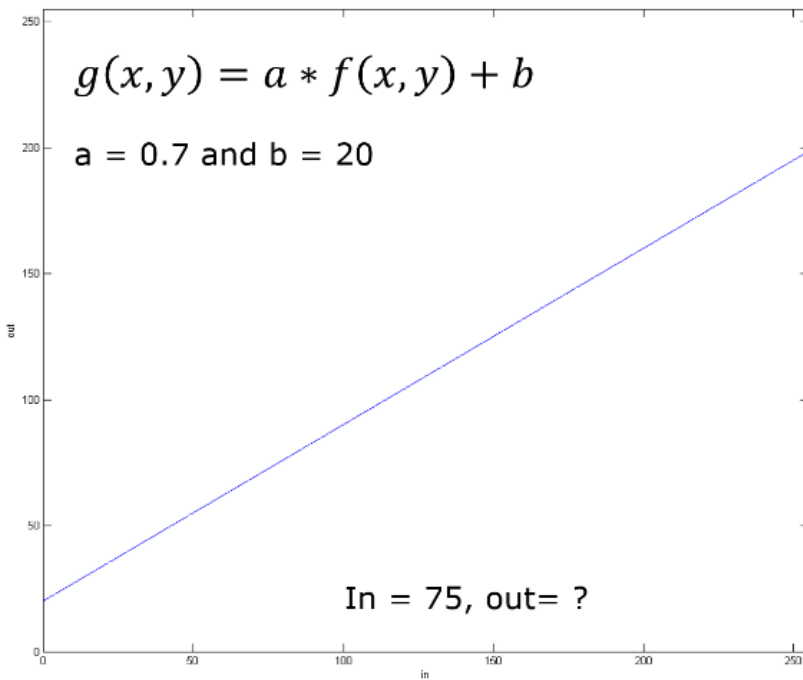
- A straight line
- Called a *linear transformation*
- Here $a = 0.7$ and $b = 20$



$$g(x, y) = a * f(x, y) + b$$



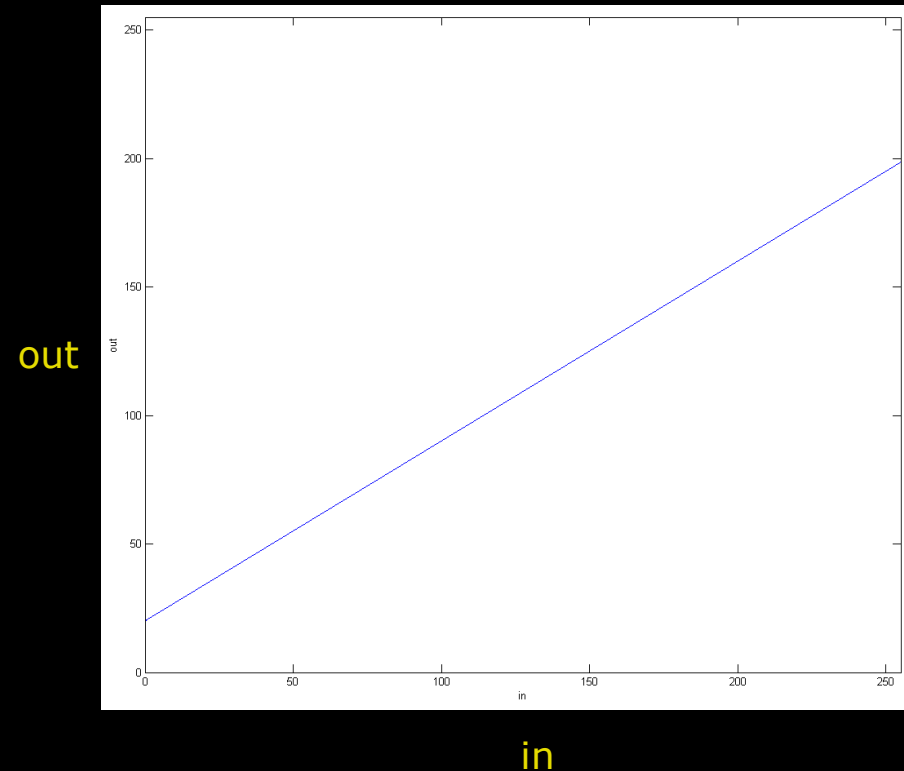
Linear Transformation





Combining brightness and contrast

- A straight line
- Called a *linear transformation*
- Here $a = 0.7$ and $b = 20$
- What will the result be on the output image?
 - More bright ($b > 0$)
 - Less contrast ($a < 1$)



$$g(x, y) = a * f(x, y) + b$$

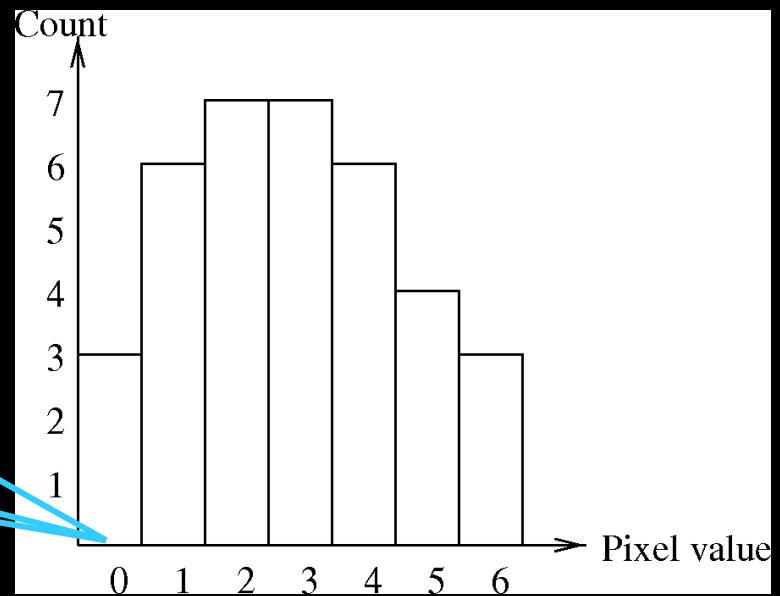


Histogram Reminder

- A histogram normally contains the same number of “bins” as the possible pixel values
- A bin stores the number of pixel with that value

0	2	6	6	3	3
1	4	3	4	4	4
3	2	5	1	5	2
1	4	2	1	3	1
2	5	3	0	2	0
4	2	5	6	3	1

3



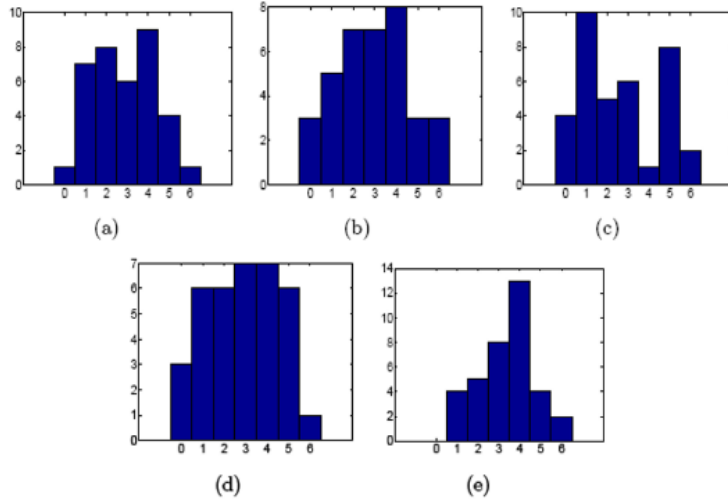


Choose the histogram that represents the image



0	5	3	5	2	1
3	5	5	3	3	1
1	1	1	3	2	3
6	2	2	1	0	0
0	2	1	5	1	5
5	5	1	4	1	6

Figur 6: Grayscale billede.



A

B

C

D

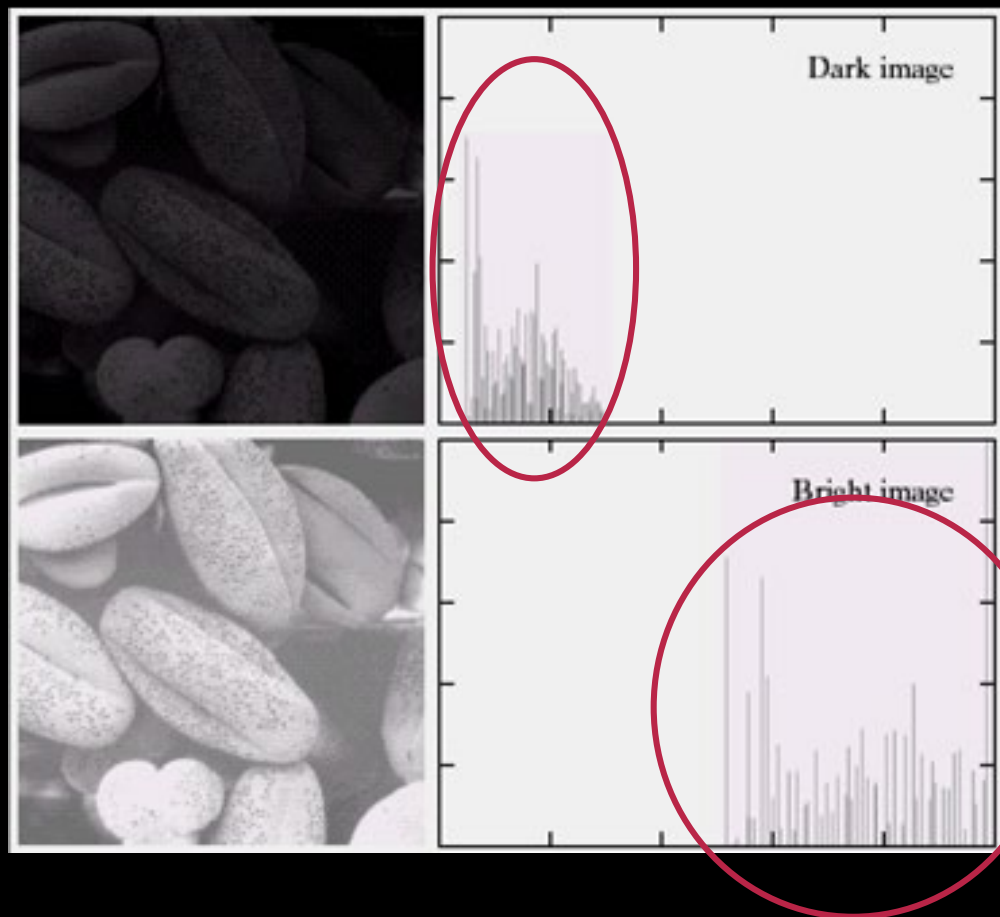
None of the above



Back to the histogram

- The shape of the histogram tells us a lot!

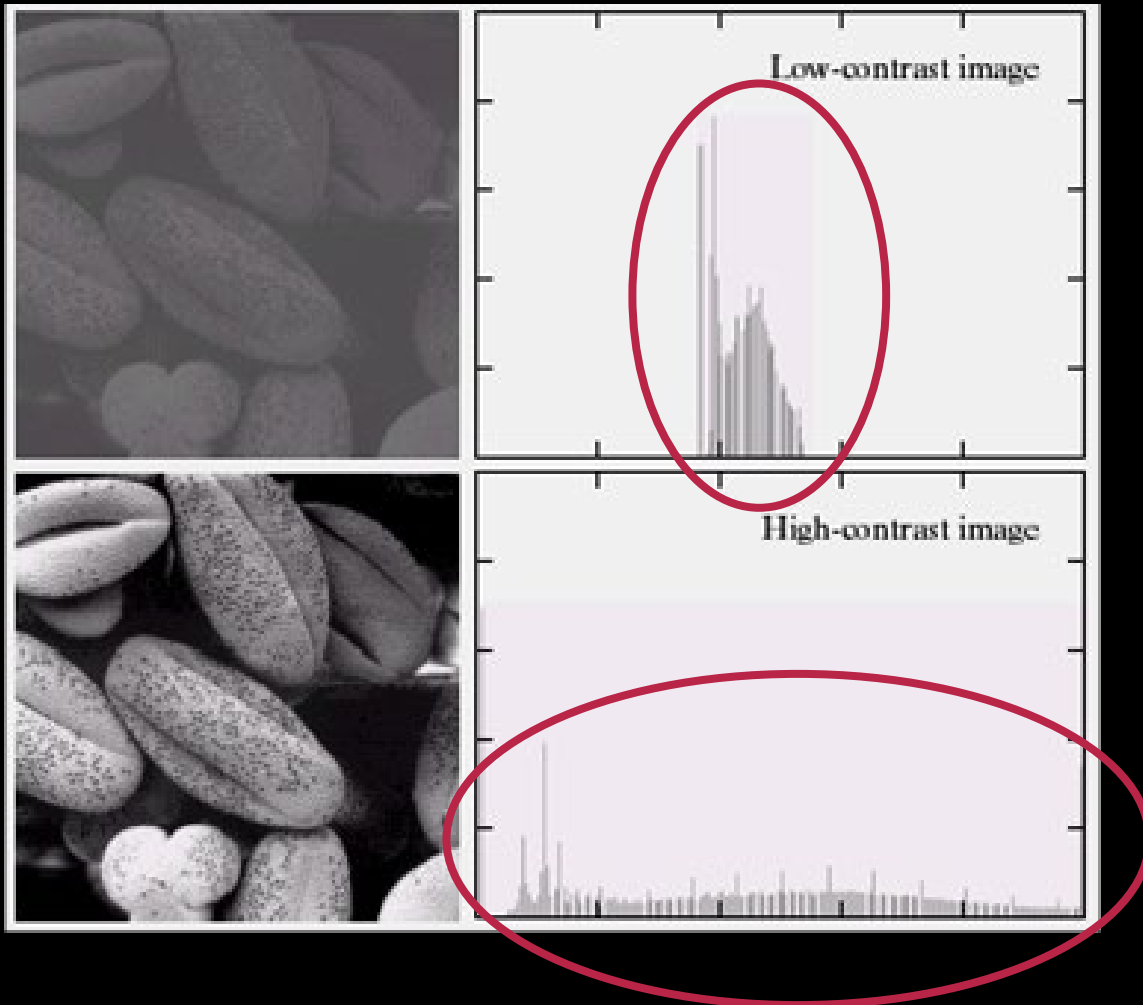
Histogram inspection



Dark image

Bright image

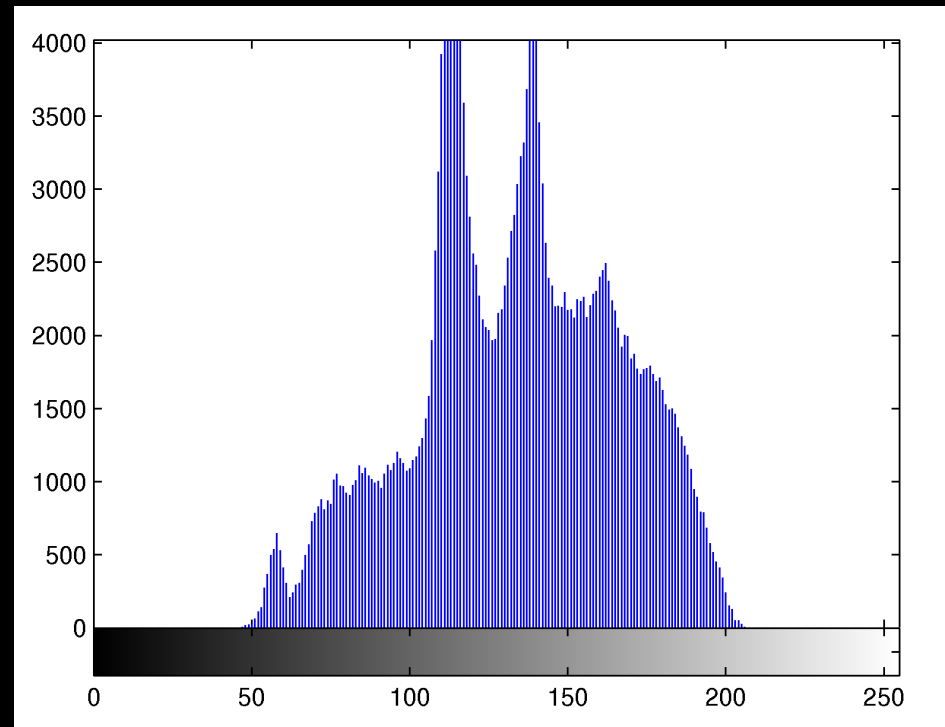
Histogram inspection



Low contrast

High contrast

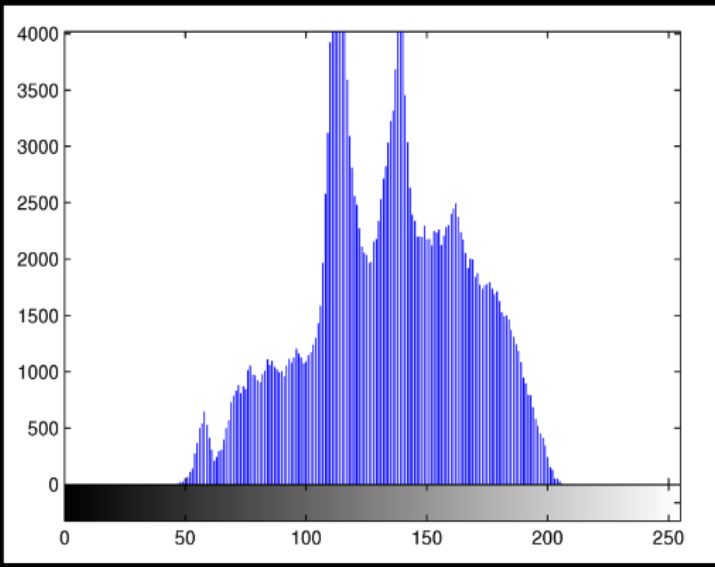
Histogram stretching



- How do we optimise the image using the histogram?
 - Minimum and maximum values?
 - Stretch it so new minimum = 0 and new maximum = 255

Histogram stretching

- We want
 - Min = 0
 - Max = 255
- We have
 - Min = 32
 - Max = 208



Using brightness

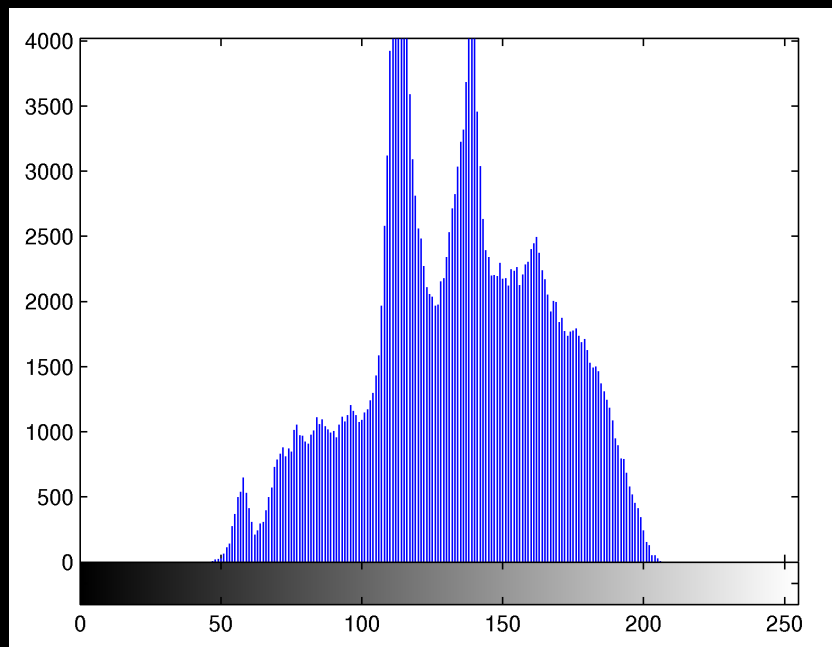
Using contrast

Using brightness
and contrast

None of the
above



Histogram stretching



- We want
 - Min = 0
 - Max = 255
- We have
 - Min = 32
 - Max = 208

$$g(x, y) = \frac{v_{max,d} - v_{min,d}}{v_{max} - v_{min}} (f(x, y) - v_{min}) + v_{min,d}$$

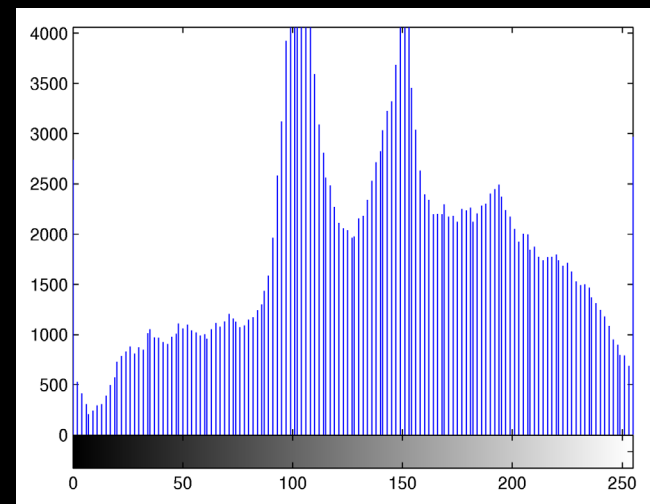
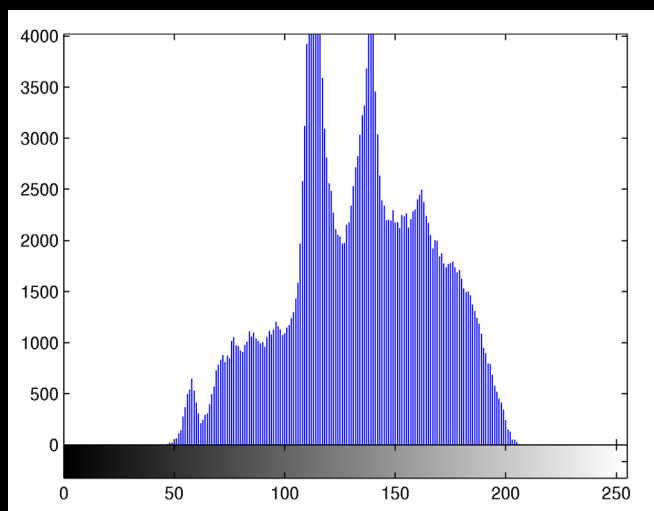


Histogram stretching formula

$$g(x, y) = \frac{v_{max,d} - v_{min,d}}{v_{max} - v_{min}} (f(x, y) - v_{min}) + v_{min,d}$$

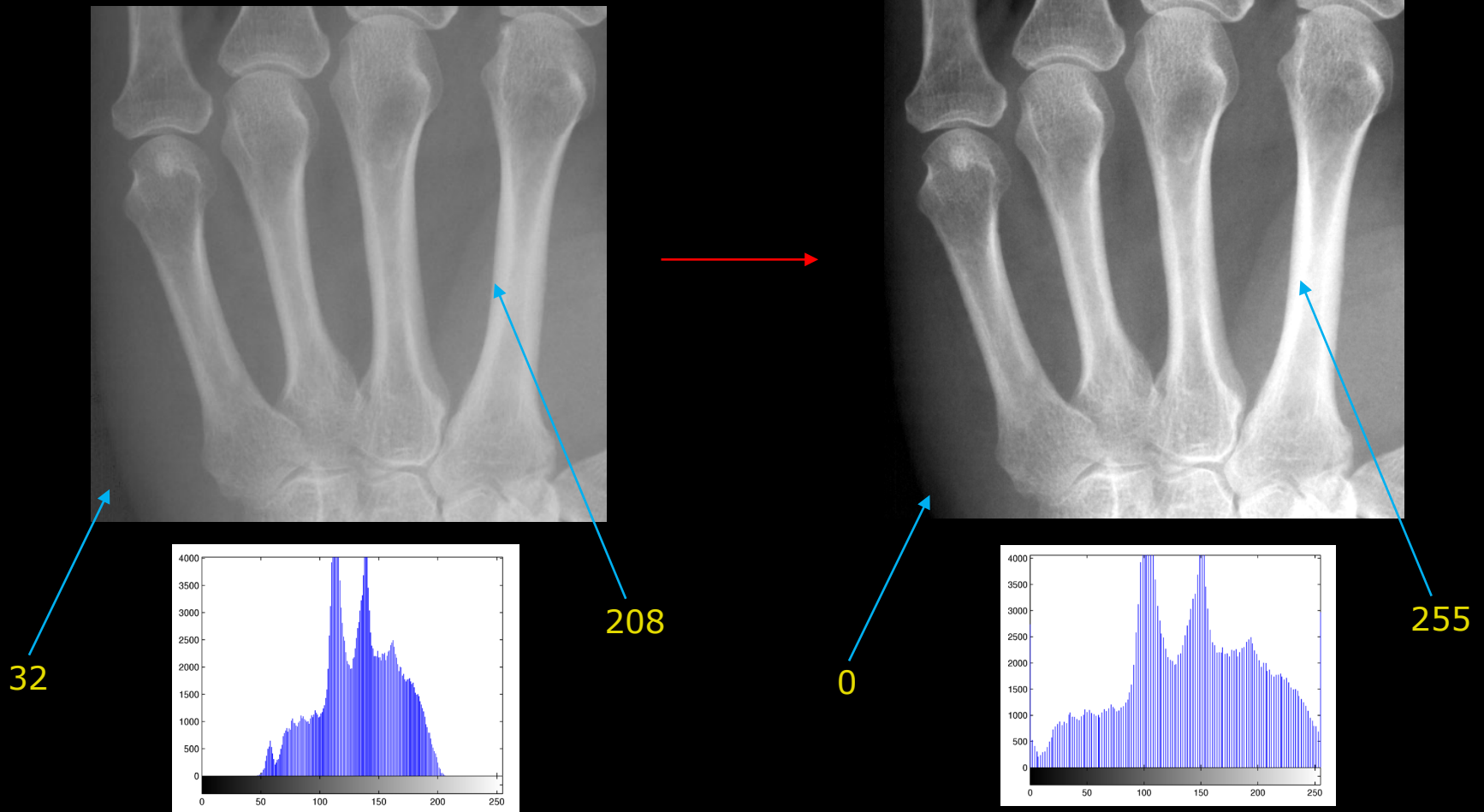
- Desired min value $v_{min,d} = 0$
- Desired max value $v_{max,d} = 255$
- Current min value $v_{min} = 32$
- Current max value $v_{max} = 208$

Histogram stretching



$$g(x, y) = \frac{255}{176} (f(x, y) - 32)$$

Effect of histogram stretching





Histogram stretching – weaknesses

- A single pixel value of 0 or 255 ruins it
- Sometimes you want
 - To stretch only the high pixel values
 - While “compressing” the low pixel values
 - Non-linear mapping

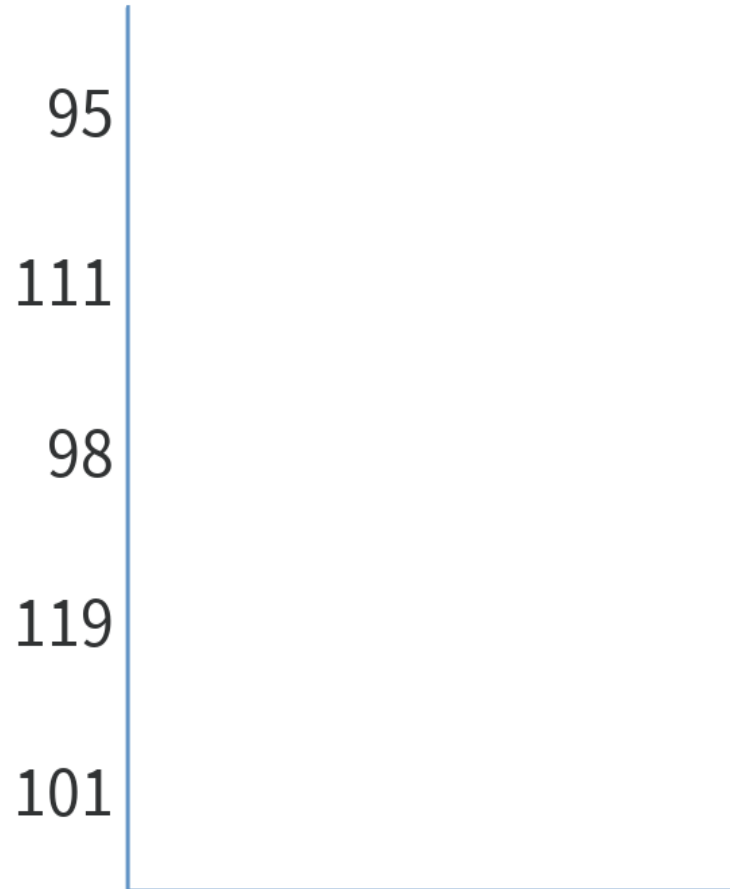
Linear mapping on an image

Der udføres en *linear mapping* på billedet i Figur 16 hvor resultatet er et ny grayscale billede med maksimum værdi 255 og minimum værdi 0. Hvad er den nye værdi af den pixel, der har værdi 108?

1. 95
2. 111
3. 98
4. 119
5. 101
6. Ved ikke

208	25	40	36	167
231	71	23	108	18
32	139	244	234	217
233	244	124	202	238
161	245	204	245	173

Figur 16: Grayscale billede.





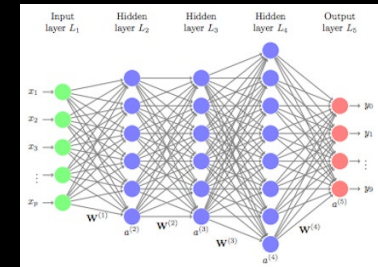
Deep learning and color/gray scale transformations

■ Deep learning needs training data

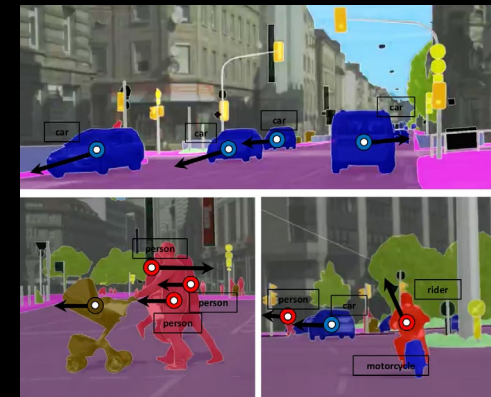
- Input image
- Ground truth labels or classes

■ When you lack data you can *augment* your data

- Create artificial versions
- Adding variation
- Changing gray / color levels in the image
- Point wise operations



http://uc-r.github.io/feedforward_DNN



Luc, Pauline, et al. "Predicting deeper into the future of semantic segmentation." IEEE International Conference on Computer Vision (ICCV). Vol. 1. 2017.



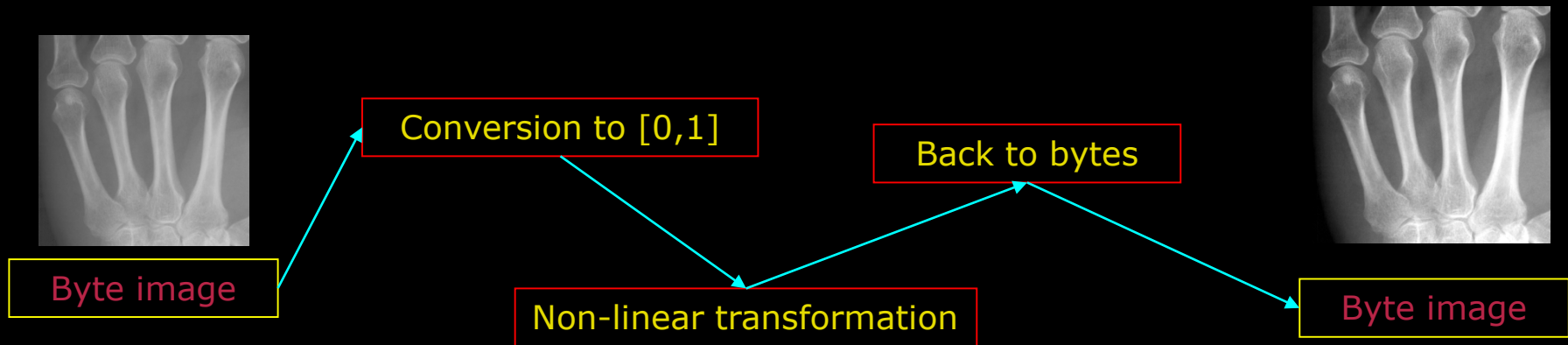
<https://www.quora.com/What-does-the-term-semantic-segmentation-mean-in-the-context-of-Deep-Learning>





Other mappings

- Non-linear mappings
- Not always nice to work with byte images
 - Better to work with image with values in $[0,1]$





Working with bytes and doubles

- A byte contains integer values [0,255]
 - A byte can not store 127.4232
- A value of type *double* can contain “all numbers”
- Why not use doubles always?
 - One double = 8 bytes in the memory
 - Images become very large!
 - Many things can be done with bytes



Map pixels to [0,1]

- In Matlab it is easiest to create a new image of type double
 - `Itemp = double(I);`
 - (temp means temporary and is used by many programmers for variables that quickly are thrown away)
- Conversion to [0,1]

$$g(x, y) = \frac{1}{255} f(x, y)$$



Pixels back to bytes

- Input pixels are $[0,1]$
- We want them to be $[0,255]$
- Simple linear transformation equal to
 - Contrast?
 - Brightness?

$$g(x, y) = 255 * f(x, y)$$

- Back to bytes
 - `Ifinal = uint8(Itemp);`

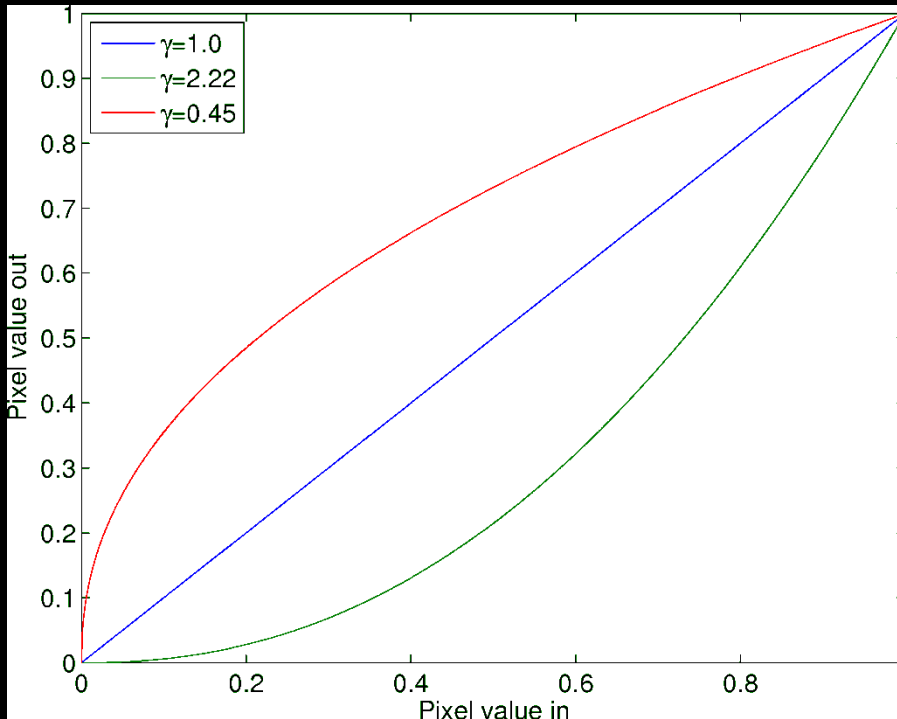


Gamma mapping

- Gamma mapping is used in televisions and flat panels
- Can increase the contrast (dynamics) in more selected part of the histogram
- Many games have a possibility for a gamma correction



Gamma curves

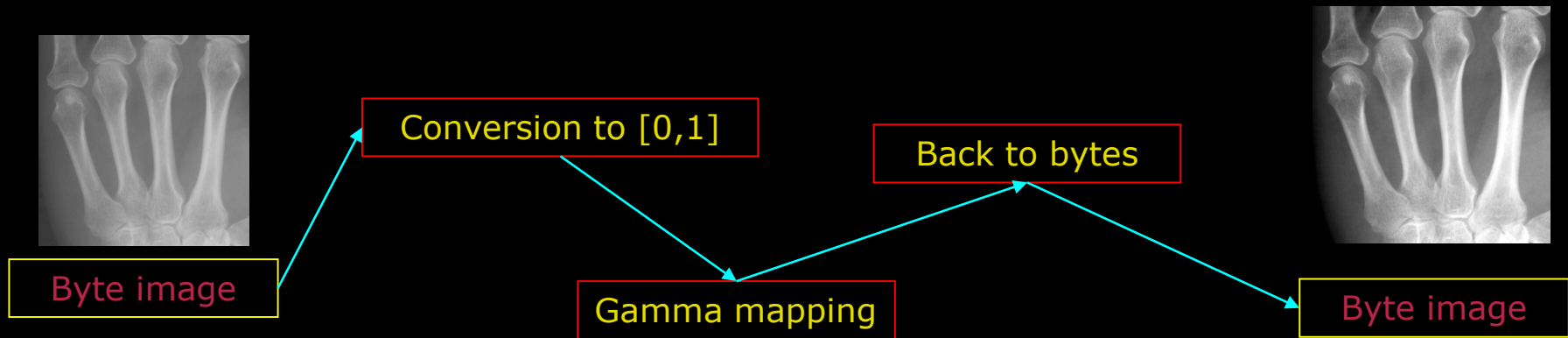


- Named after the Greek letter gamma
- What happens to the dark areas
 - With 0.45?
 - With 2.22?

$$g(x, y) = f(x, y)^\gamma$$

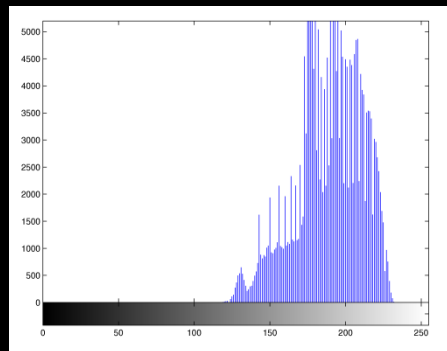


Perform the gamma mapping

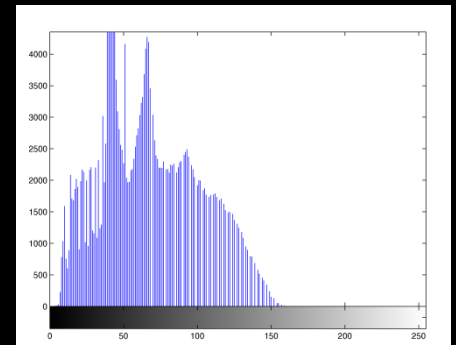
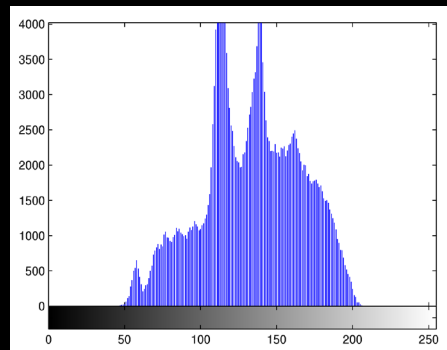


Results of gamma mapping

0.45



2.22



Gamma mapping on an image

Der udføres en *gamma mapping* med $\gamma = 1.3$ på billedet i Figur 3. Resultatet er et ny grayscale billede. Hvad er den mindste og største pixelværdi i det nye billede?

1. 0 og 255
2. 25 og 130
3. 8 og 242
4. 15 og 230
5. 37 og 219
6. Ved ikke

208	25	40	36	167
231	71	23	108	18
32	139	244	234	217
233	244	124	202	238
161	245	204	245	173

Figur 3: Grayscale billede.

0, 255

25, 130

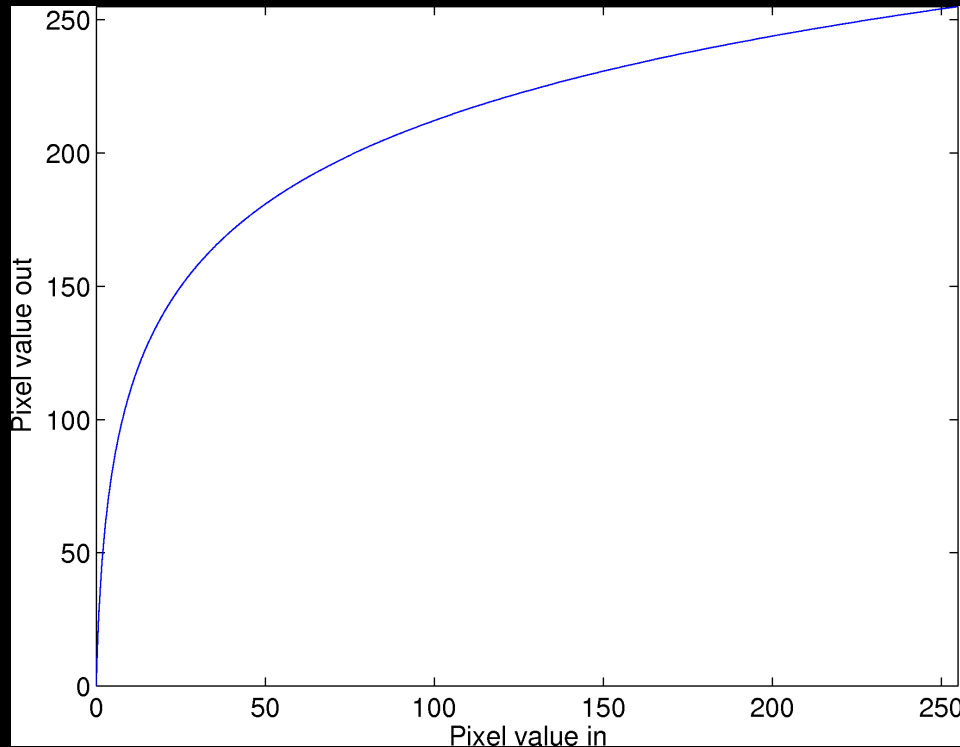
8, 242

15, 230

37, 219



Logarithmic mapping



Maps from [0,255] to [0,255]

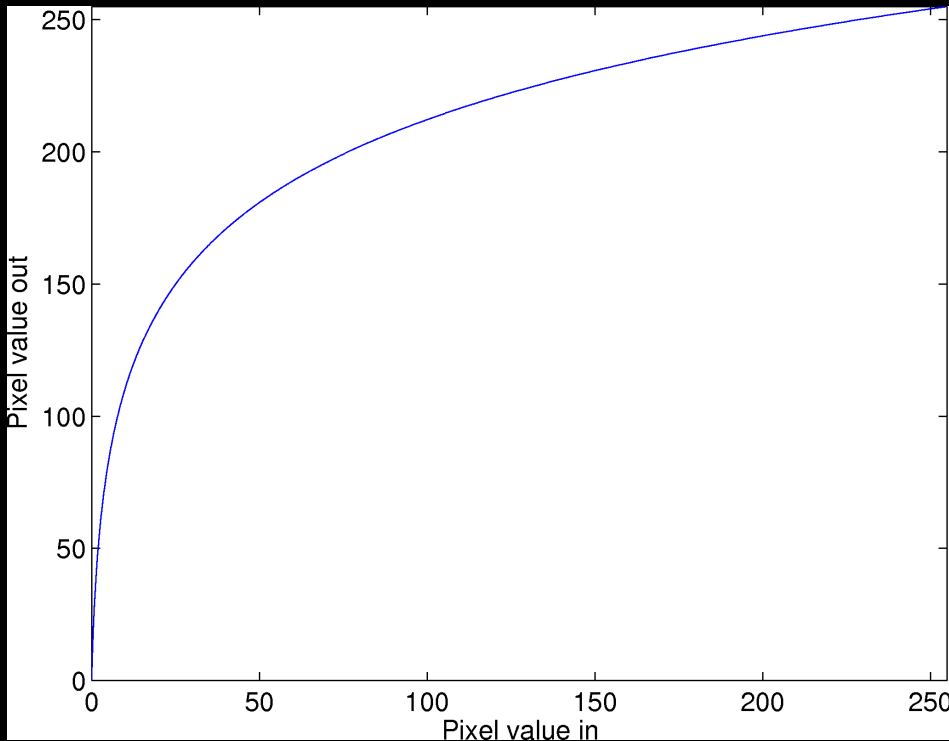
Why?

$$g(x, y) = c \log(1 + f(x, y))$$

$$c = \frac{255}{\log(1 + v_{max})}$$

Logarithmic mapping – when?

- For images with very bright spots
- Low intensity pixel values are enhanced



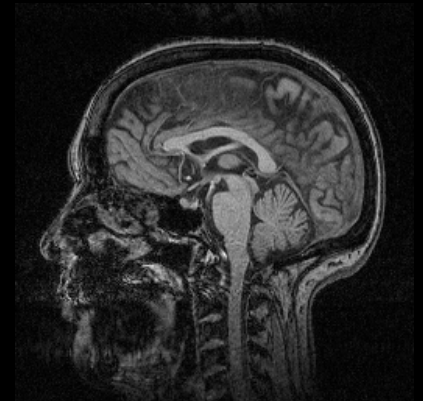
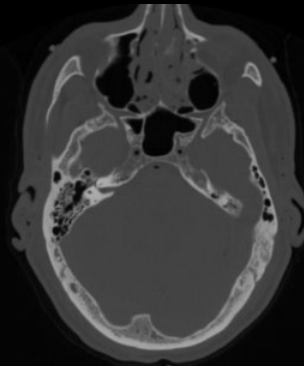


What do we get out of pixel mappings

- Spreading out or compressing pixel values
 - Better for humans to see
 - New information – no!

Now for something different

- Until now image processing
 - Input image transformed to output image
- Now for something more like image analysis
- Segmentation
 - Segment the image into regions
 - Background and objects for example





Thresholding

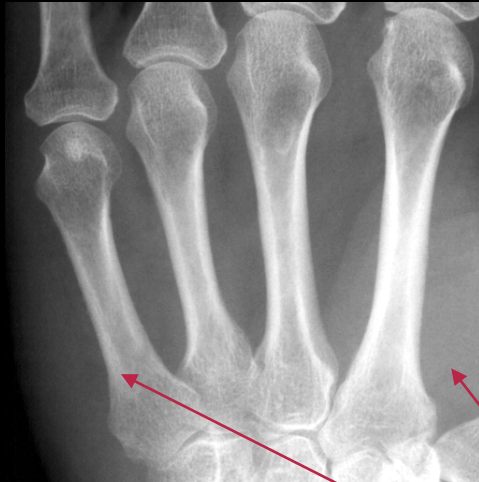
- A threshold T is a value
 - Pixels below that value is set to 0 (background)
 - Pixels equal or above is set to 1 (object)
- One threshold value for the entire image
 - Difficult to choose!

if $f(x, y) \leq T$ then $g(x, y) = 0$

if $f(x, y) > T$ then $g(x, y) = 255$



Thresholding



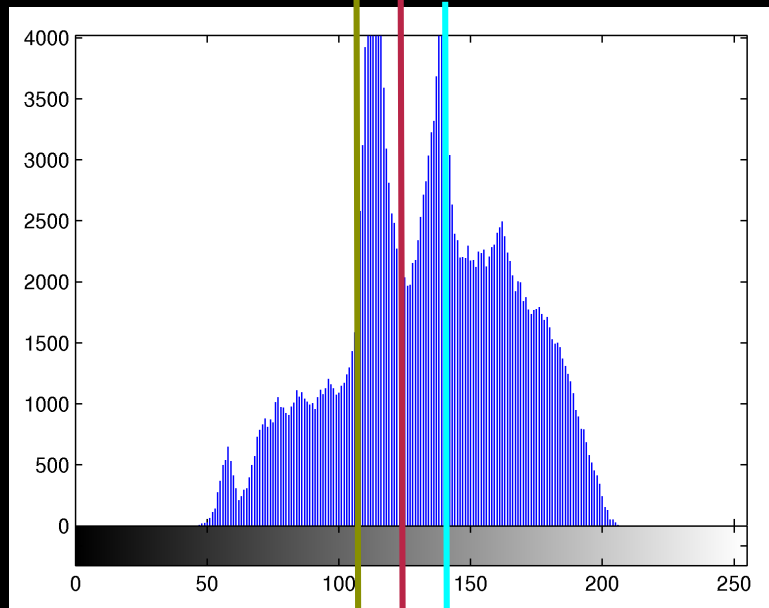
Background and bone have same value!



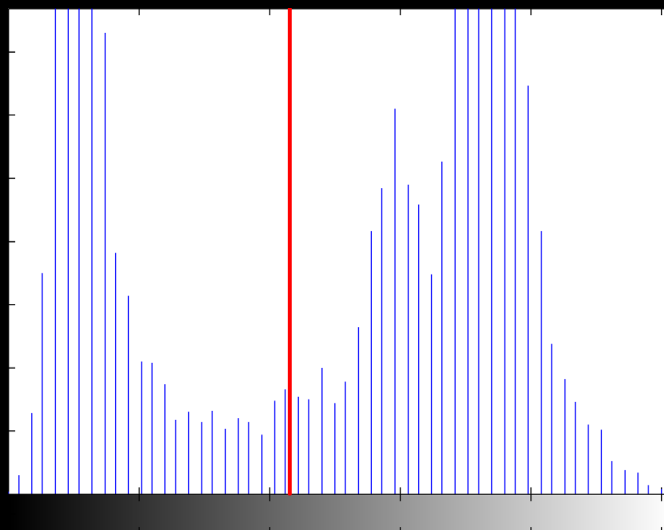
Thresholding based on the histogram



The bones are visible in the histogram!
But mixed with soft-tissue



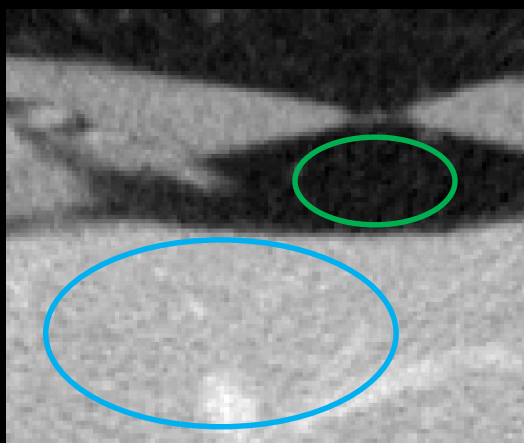
Automatic Tresholding



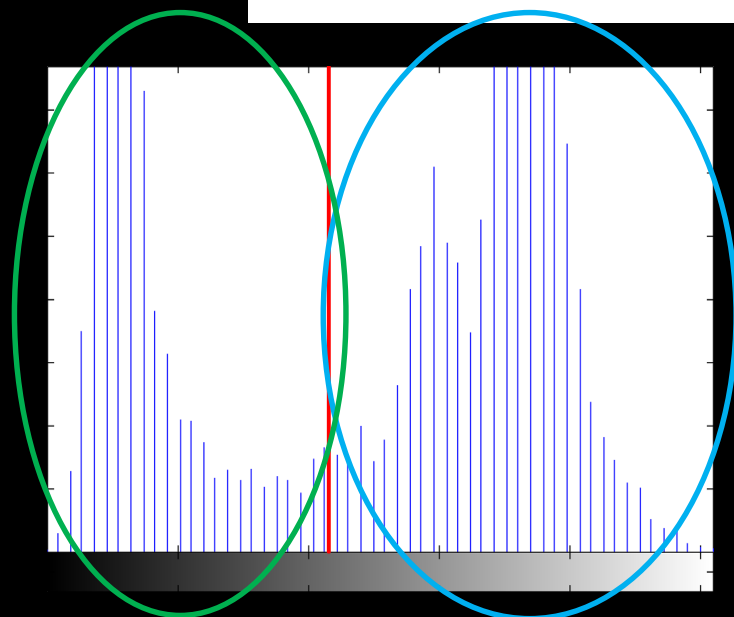


Automatic Tresholding

Otsu's method

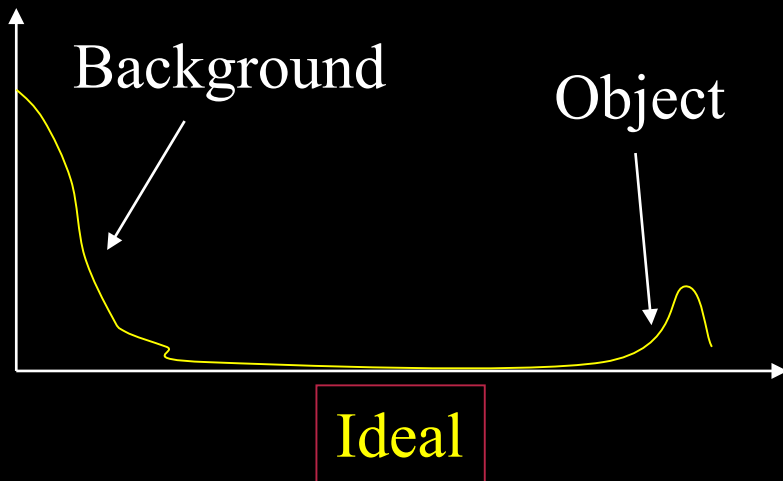
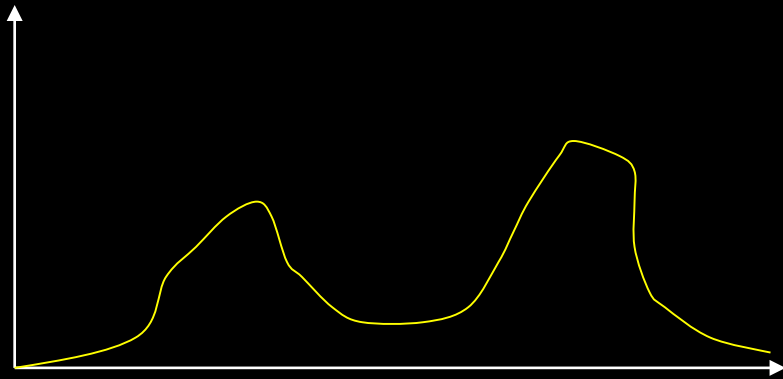


- Two classes: **background** and **object**
- T divides pixels into object and background
- Compute pixel value variance in each class
- Find T that minimises combined variance



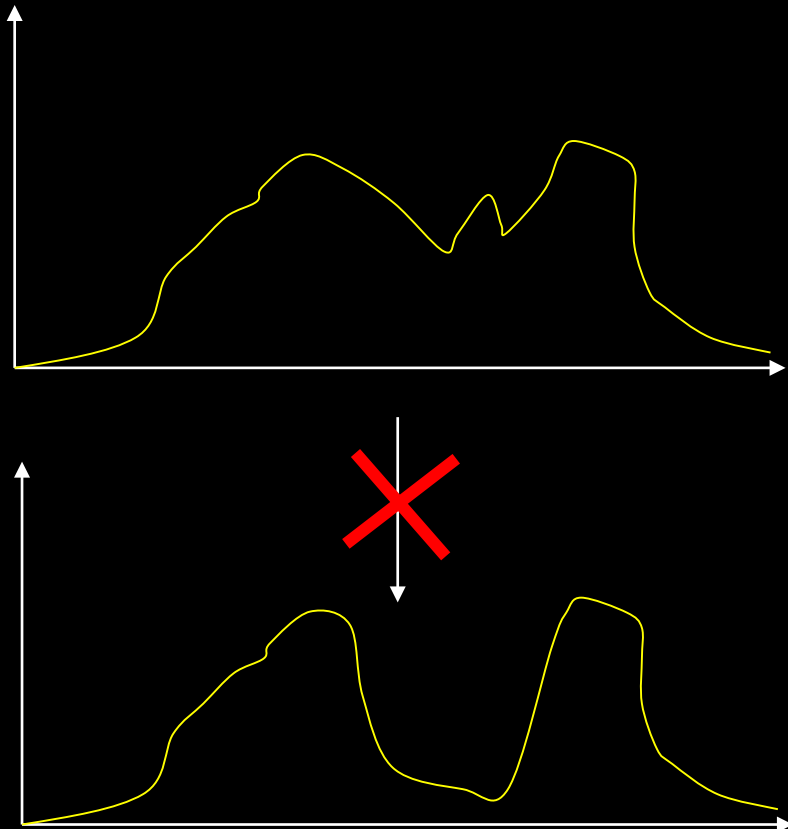


Segmentation – histogram shaping



- With a threshold you want a histogram with two peaks
 - *Bimodal*
- An ideal histogram has well separated peaks
- Obtaining a bi-modal histogram is very important in the image acquisition

Histogram shaping



- It is not possible to “unmix” using gray level transformations



Should be higher

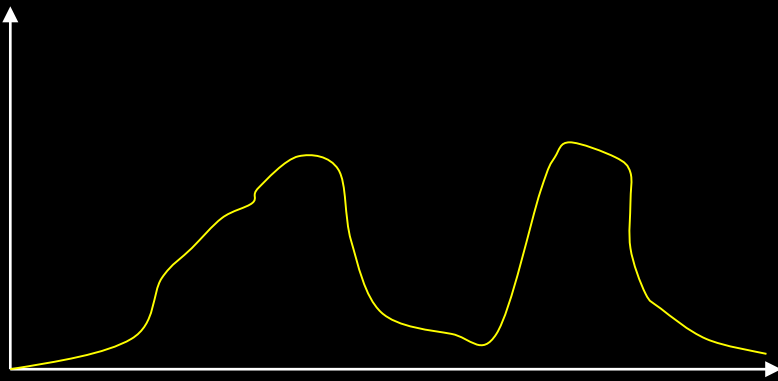
Should be lower



How to obtain good histograms

■ With cameras

- Light
- Setup
- Camera
- Lens
- Backlight?





Gamma mapping and threshold on an image

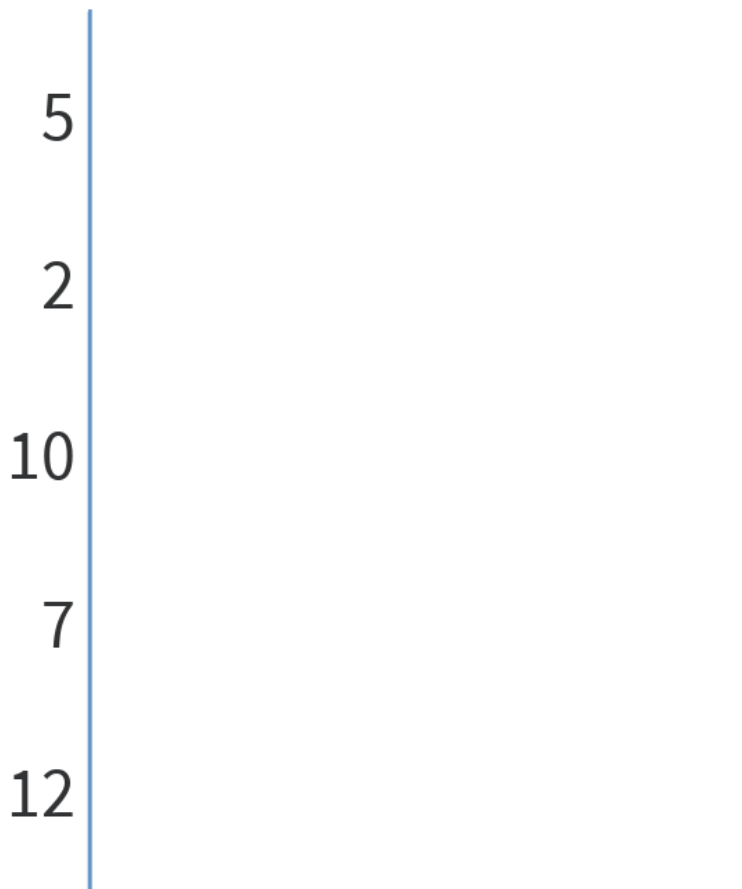


Der udføres en *gamma mapping* med $\gamma = 1.1$ på billedet i Figur 12 og herefter sættes der et *threshold* på 120. Pixels over *threshold* sættes til forgrund og resten til baggrund. Hvor mange forgrundspixler er der i resultatbilledet?

- 1. 5
- 2. 2
- 3. 10
- 4. 7
- 5. 12
- 6. Ved ikke

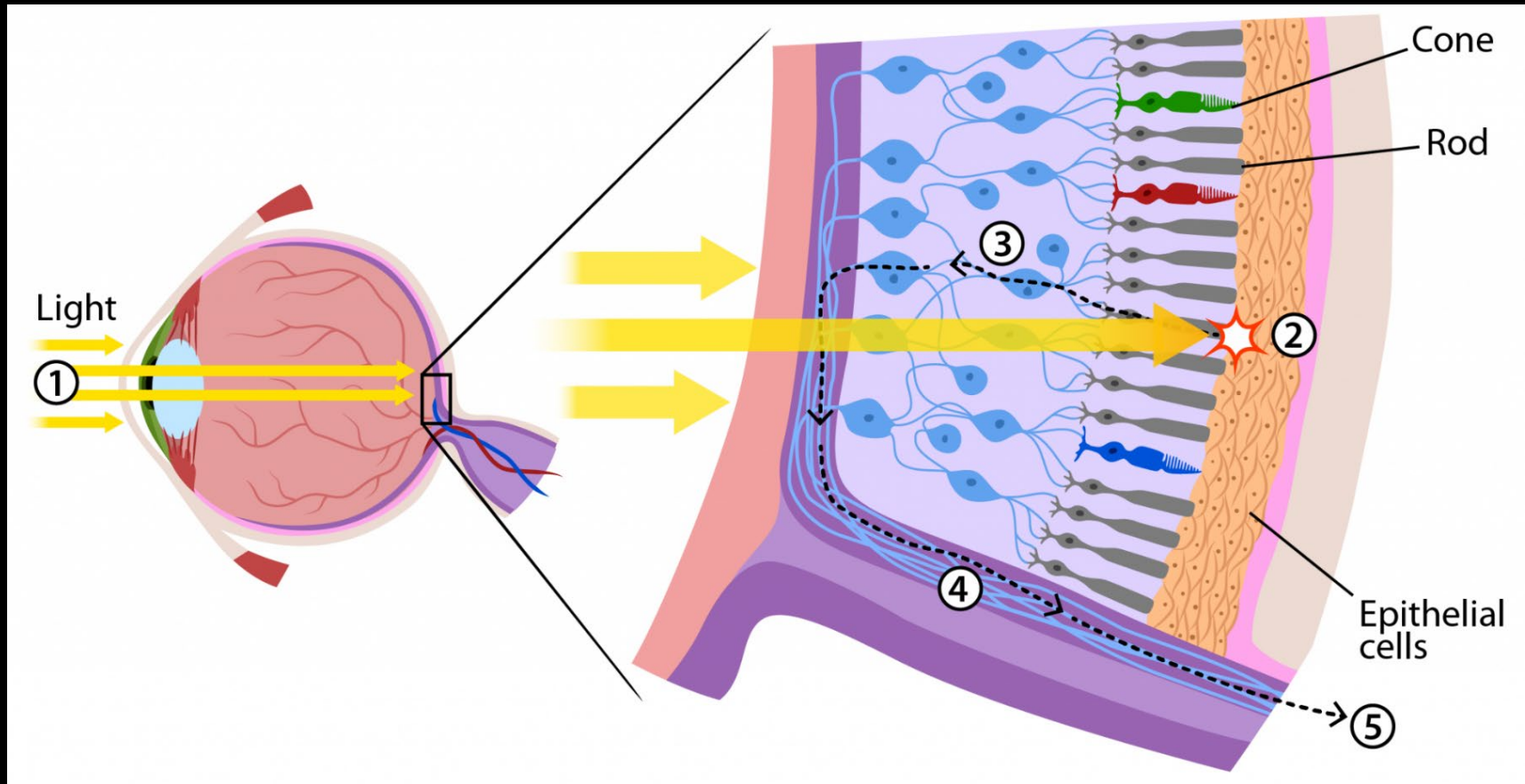
108	245	217
234	167	238
202	9	173

Figur 12: Grayscale billede.



Colour images and colour perception

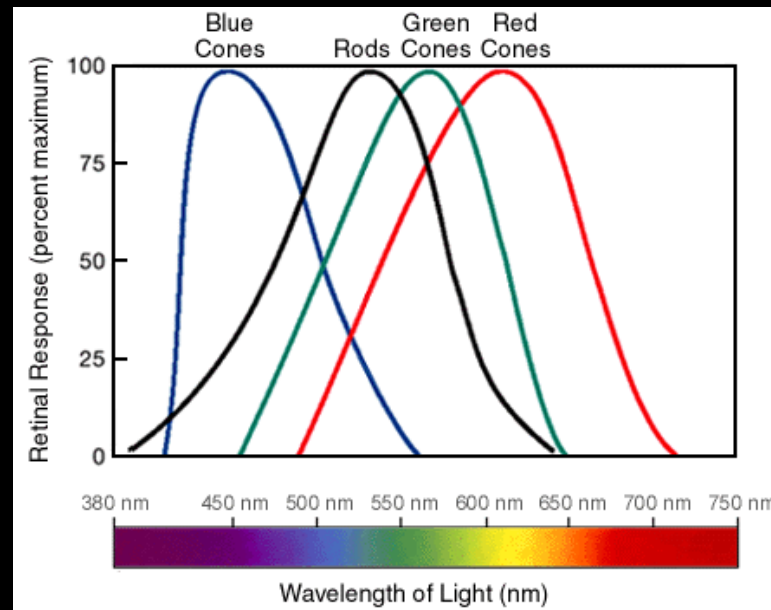
The Human Eye



<https://askbiologist.asu.edu/rods-and-cones>

Color sensitivity

Photoreceptor cell	Wavelength in nanometers (nm)	Peak response in nanometer (nm)	Interpretation by the human brain
Cones (type L)	[400-680]	564	Red
Cones (type M)	[400-650]	534	Green
Cones (type S)	[370-530]	420	Blue
Rods	[400-600]	498	Shade of gray

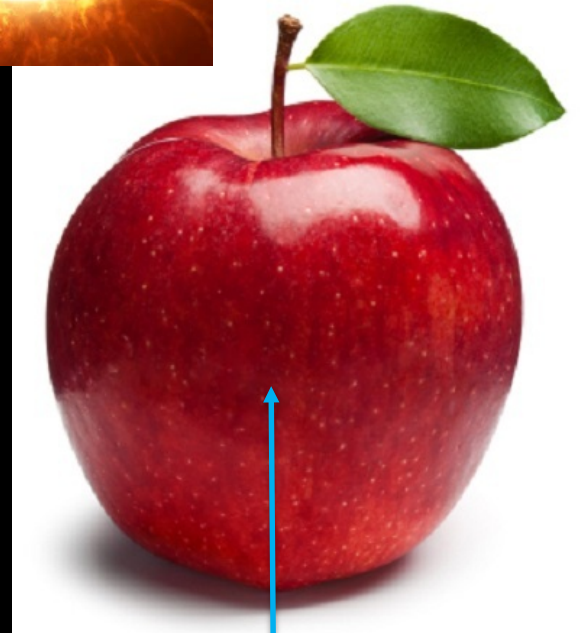
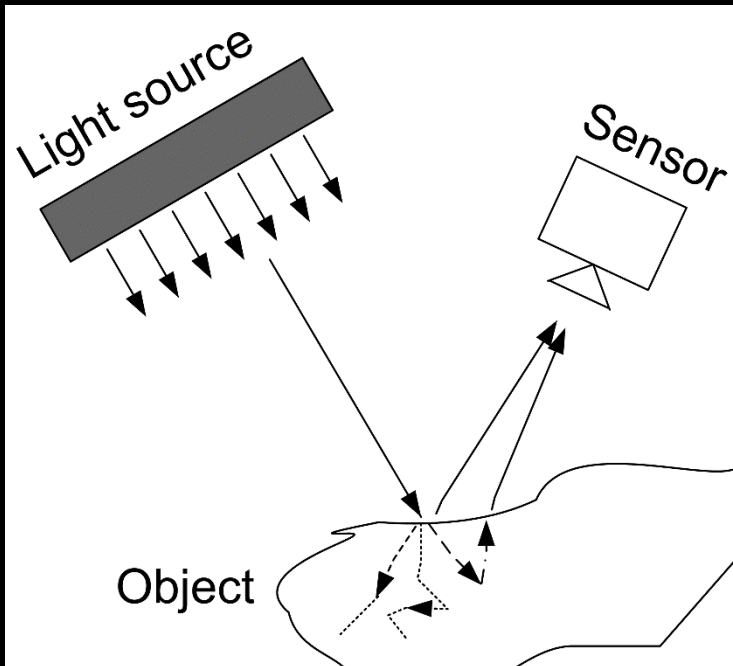
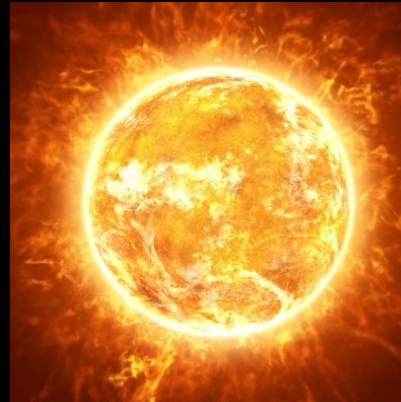


<https://askabiologist.asu.edu/rods-and-cones>



Object colors

Subtractive colors



All other colors than red absorbed





Object colors

Additive colors



- Additive colours: Final colour is made by mixing red, green, and blue
- RGB = Red, Green, and Blue
- Television, computers, digital cameras use the "RGB color space"
- Typically the values of R, G, and B lie between 0 and 255



RGB Colours



RGB = (0,0,0)

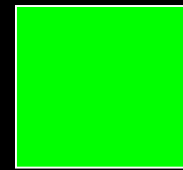


RGB = (255,255,255)

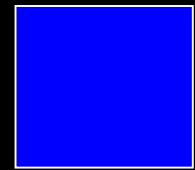
- When all three "Lamps" are turned of we get black
- When all three "lamps" are on what do we get?



(255,0,0)



(0,255,0)



(0,0,255)



(255,255,0)

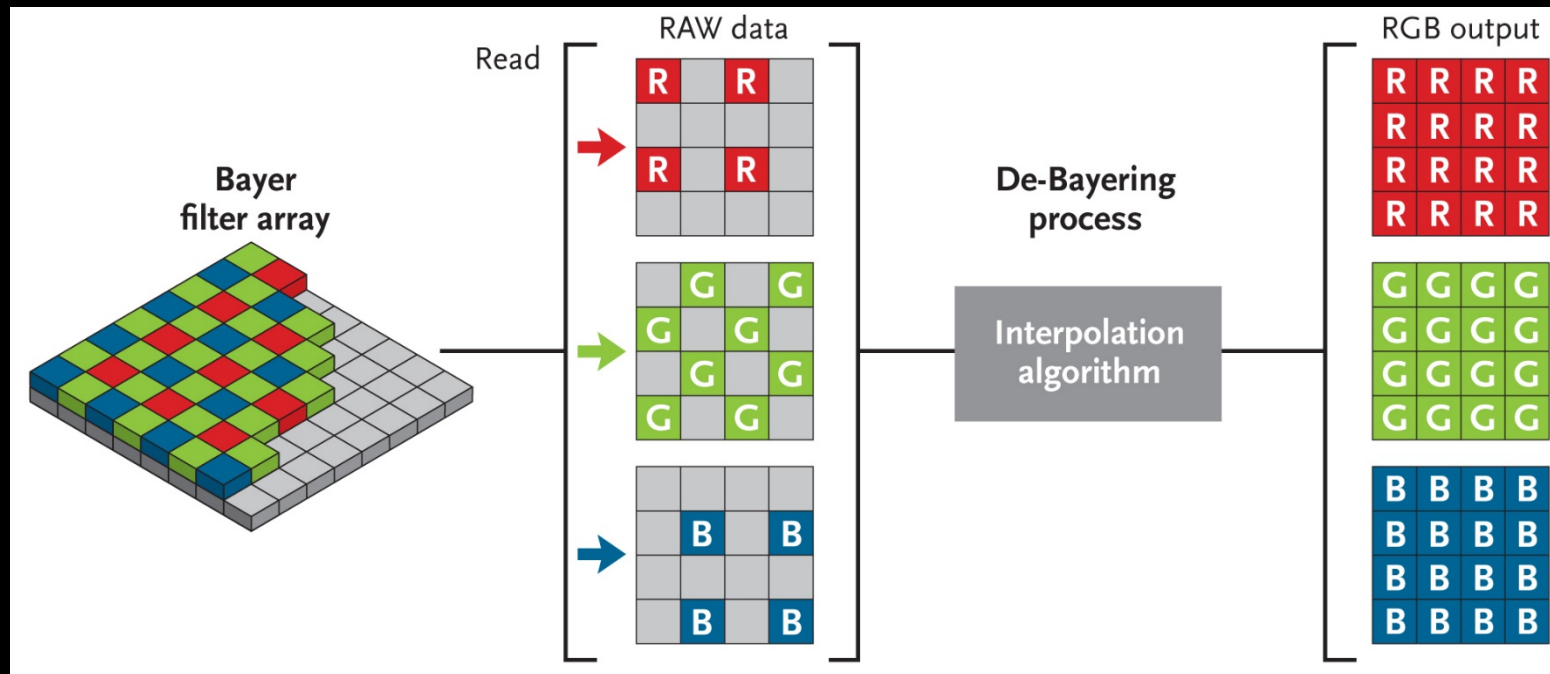


(0,255,255)



(255,0,255)

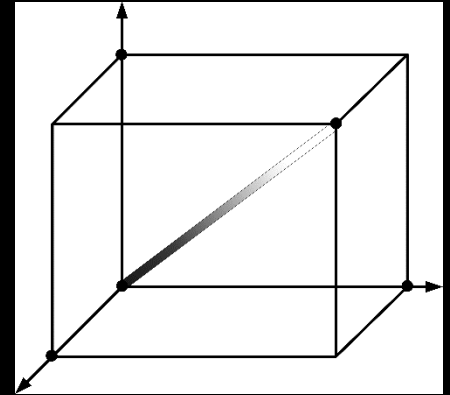
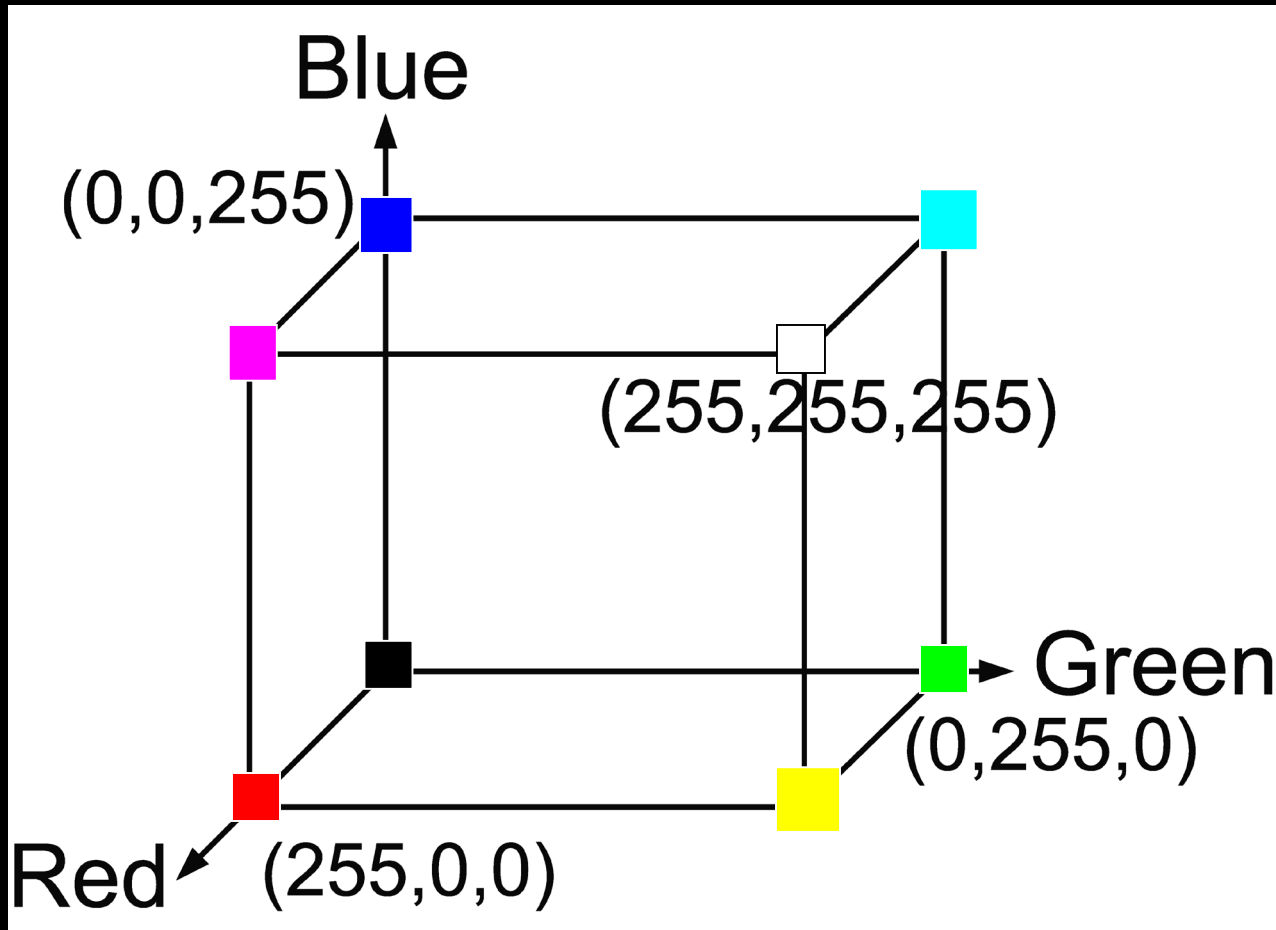
Color camera with one sensor



<http://www.skyandtelescope.com/astronomy-resources/astrophotography-tips/redeeming-color-planetary-cameras/>



RGB color space



Converting colour to grayscale

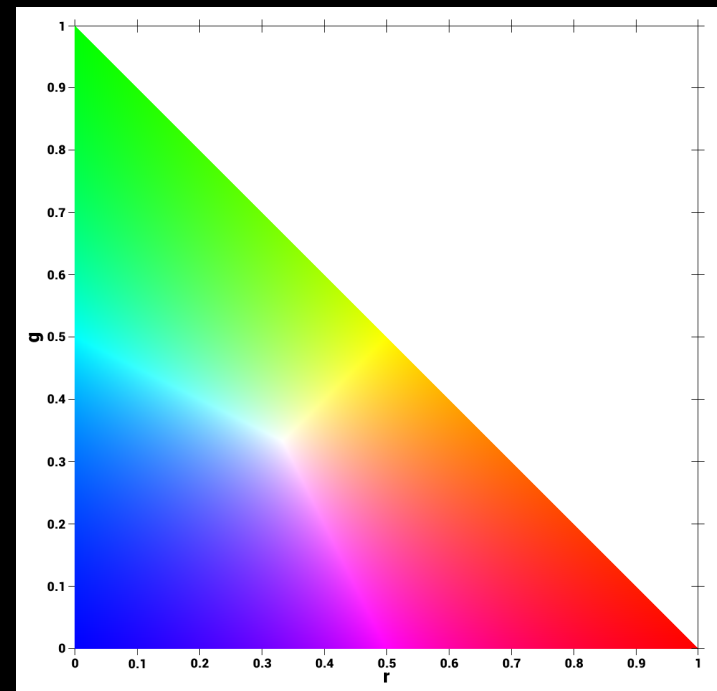
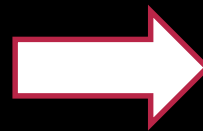
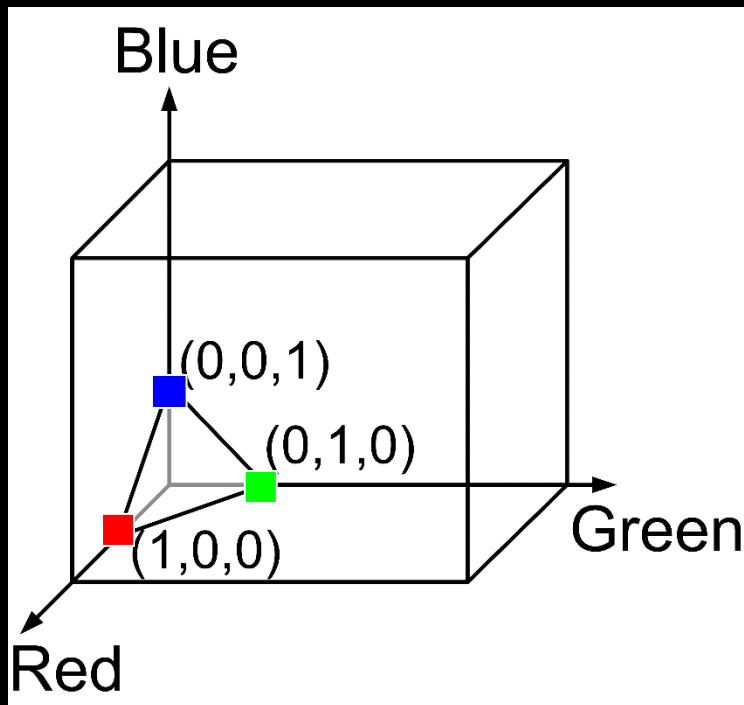
$$v = 0.2989 * R + 0.5870 * G + 0.1140 * B$$





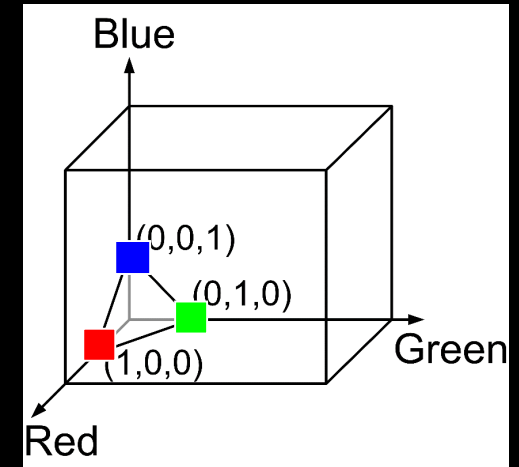
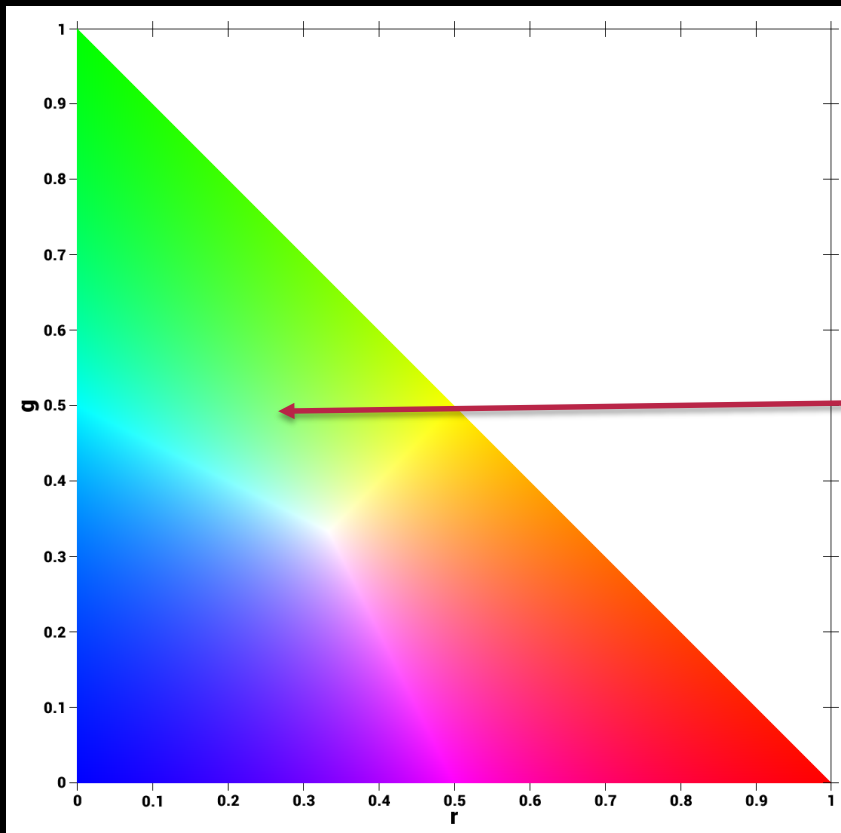
Normalised RGB colors

$$(r, g, b) = \left(\frac{R}{R+G+B}, \frac{G}{R+G+B}, \frac{B}{R+G+B} \right)$$





Another RGB representation

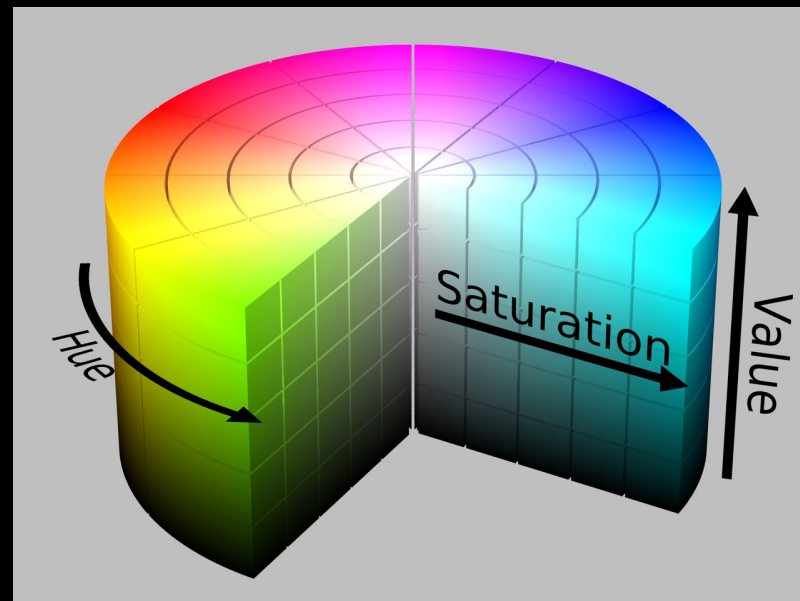


(r, g, I)

$$I = \frac{R+G+B}{3}$$

HSI Color Representation

- **Hue** – the dominant wave length in the perceived light (the pure color)
- **Saturation** – the purity of the color
- **Intensity** – the brightness of the color (sometimes called the value)





Converting between RGB and HSI

- You have an RGB value
- You want the corresponding HSI value

$$H = \begin{cases} \cos^{-1} \left(1/2 \cdot \frac{(R-G)+(R-B)}{\sqrt{(R-G)(R-G)+(R-B)(G-B)}} \right), & \text{if } G \geq B; \\ 360^\circ - \cos^{-1} \left(1/2 \cdot \frac{(R-G)+(R-B)}{\sqrt{(R-G)(R-G)+(R-B)(G-B)}} \right), & \text{Otherwise.} \end{cases} \quad (8.8)$$

$$H \in [0, 360[$$

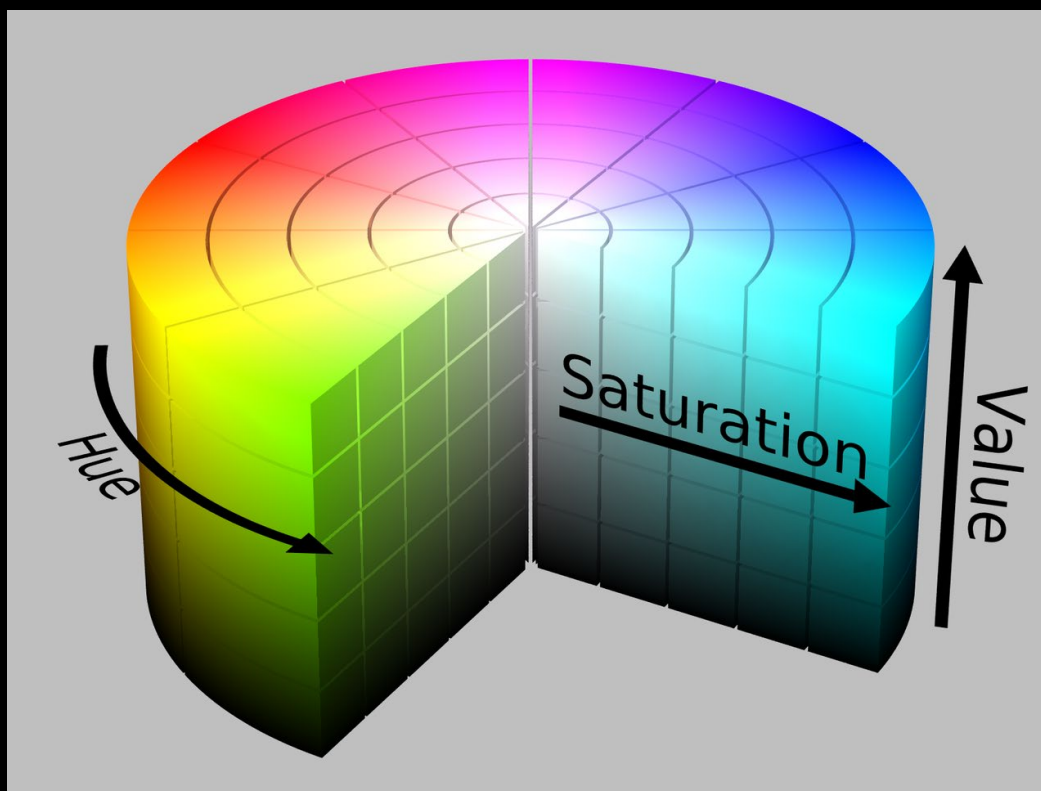
$$S = 1 - 3 \cdot \frac{\min\{R, G, B\}}{R + G + B} \quad S \in [0, 1] \quad (8.9)$$

$$I = \frac{R + G + B}{3} \quad I \in [0, 255] , \quad (8.10)$$



Why other colorspaces

- Why should we use for example HSI ?

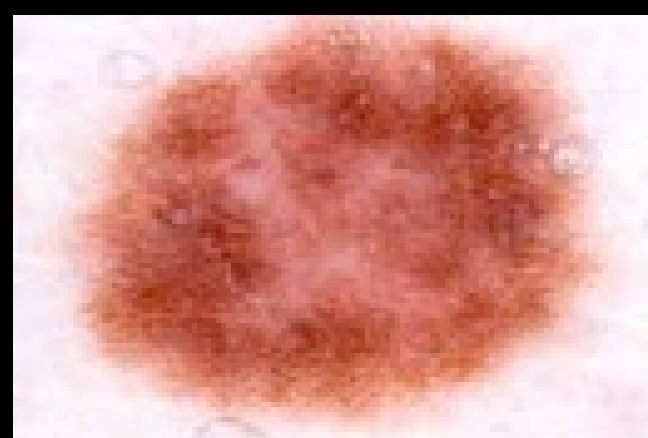
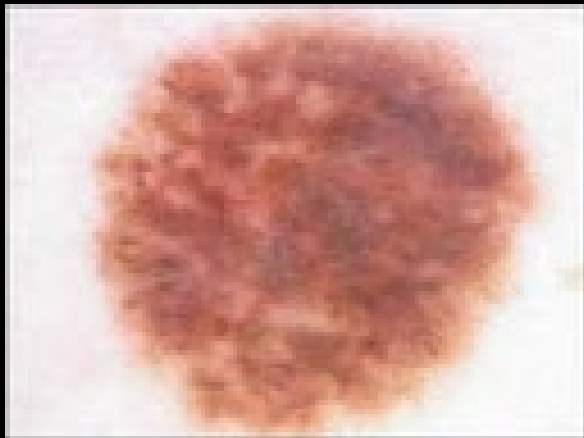
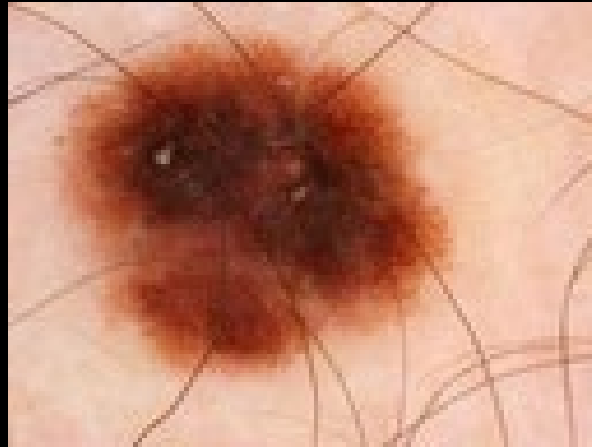


Melanoma segmentation



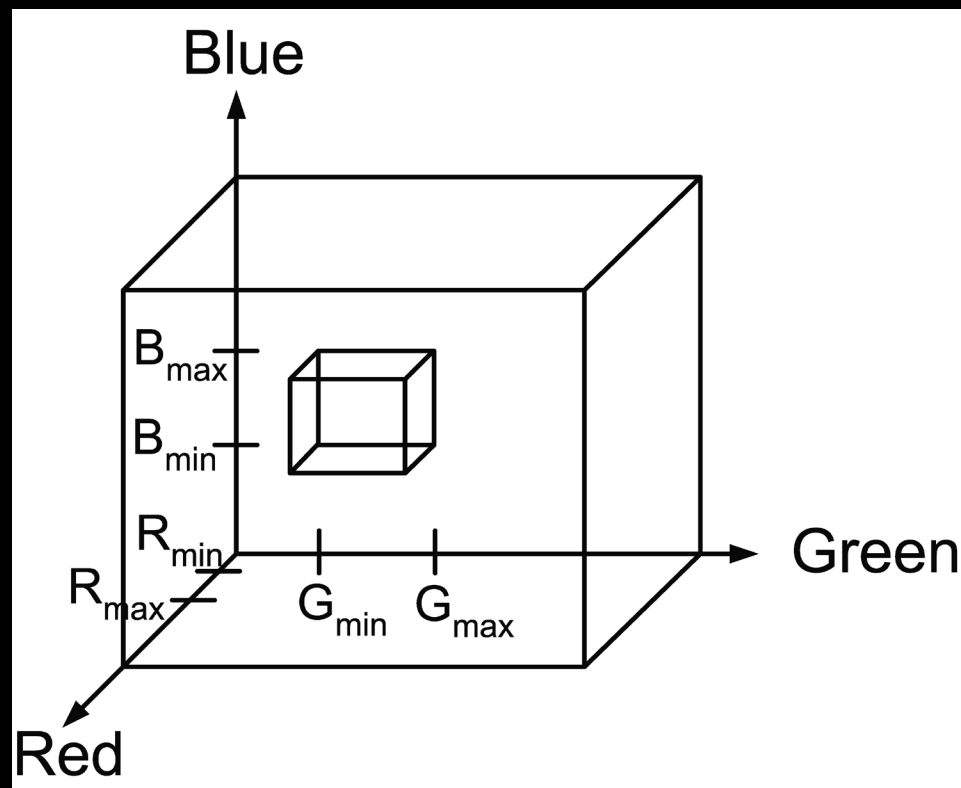
- An algorithm that can do pixelwise classification
 - Background / skin
 - Melanoma
- Use the colors

Melanoma segmentation – color variation





Color thresholding



If

$R > R_{min}$ and $R < R_{max}$ and

$G > G_{min}$ and $G < G_{max}$ and

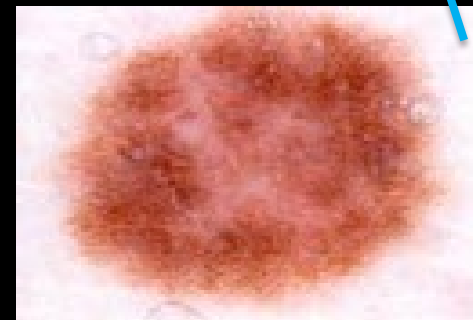
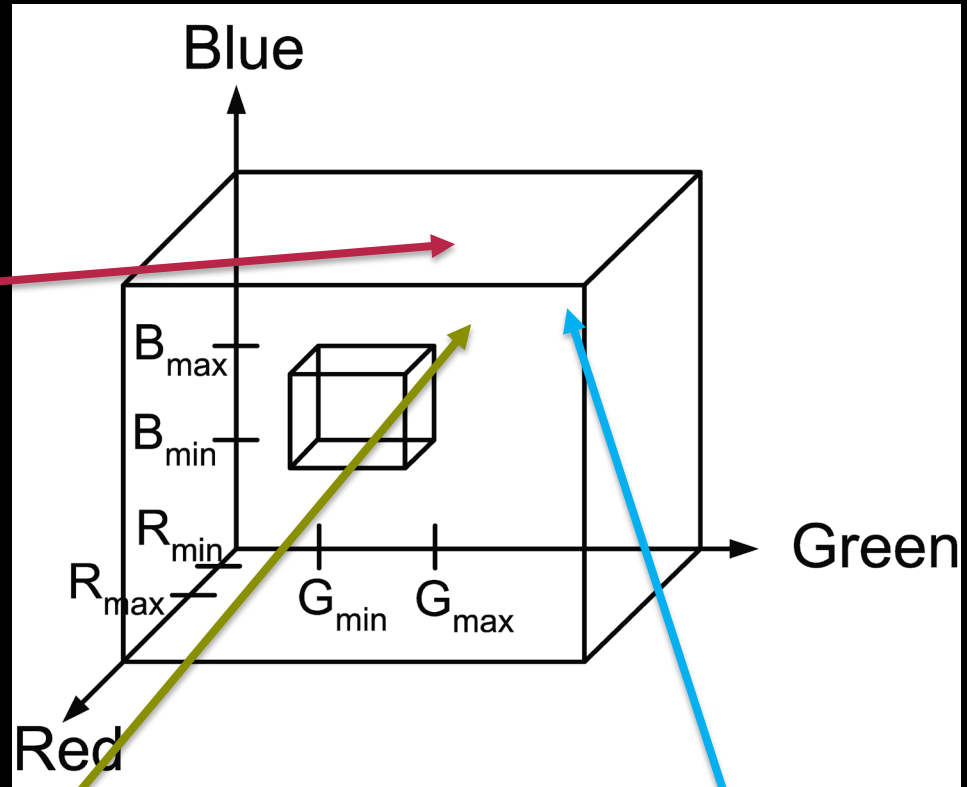
$B > B_{min}$ and $B < B_{max}$

Then $g(x, y) = 255$

Else $g(x, y) = 0$

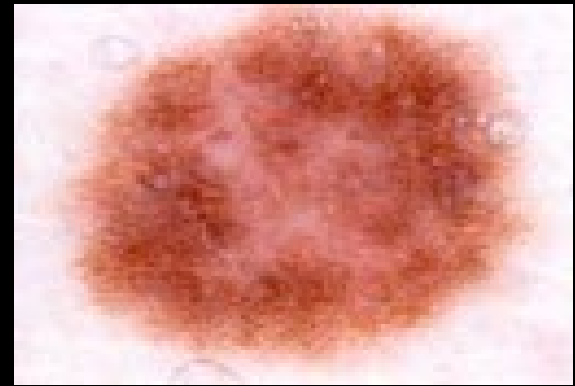
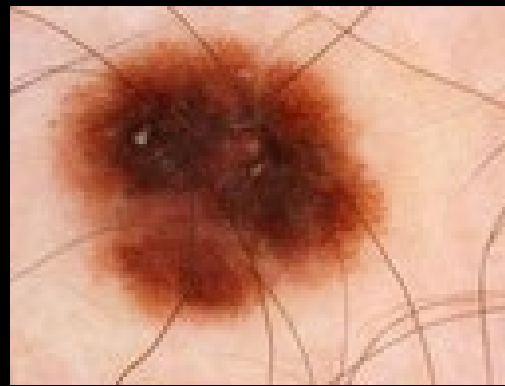
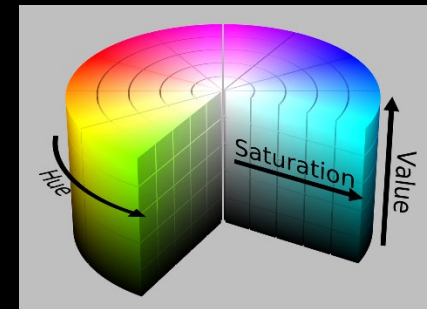


Color thresholding



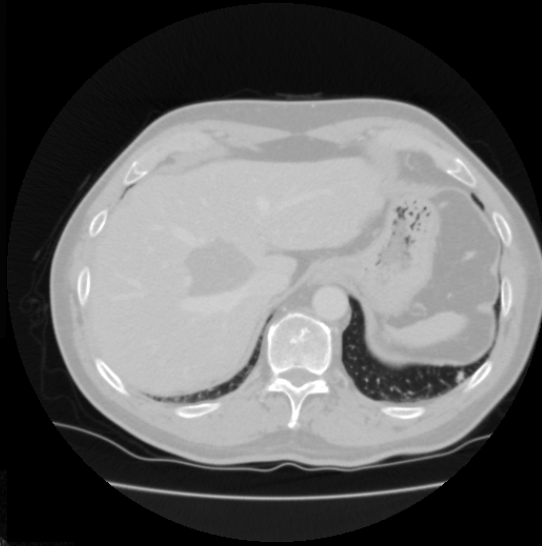
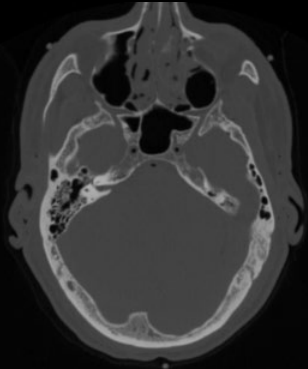
Color variation

- The major variation is in the brightness
 - This will spread out the values in RGB space
- The Hue is rather constant
- HSI Space
 - Hue and saturation rather stable
 - Only variation in intensity / value





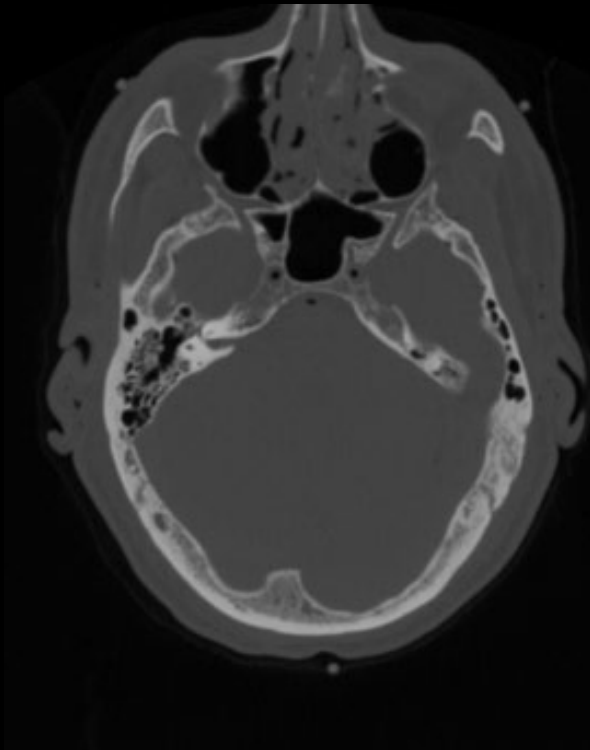
Contrast in medical images



- How do we optimise image acquisition when we want to look at
 - Bones
 - Brain structures
 - Cancer

Image acquisition - bone

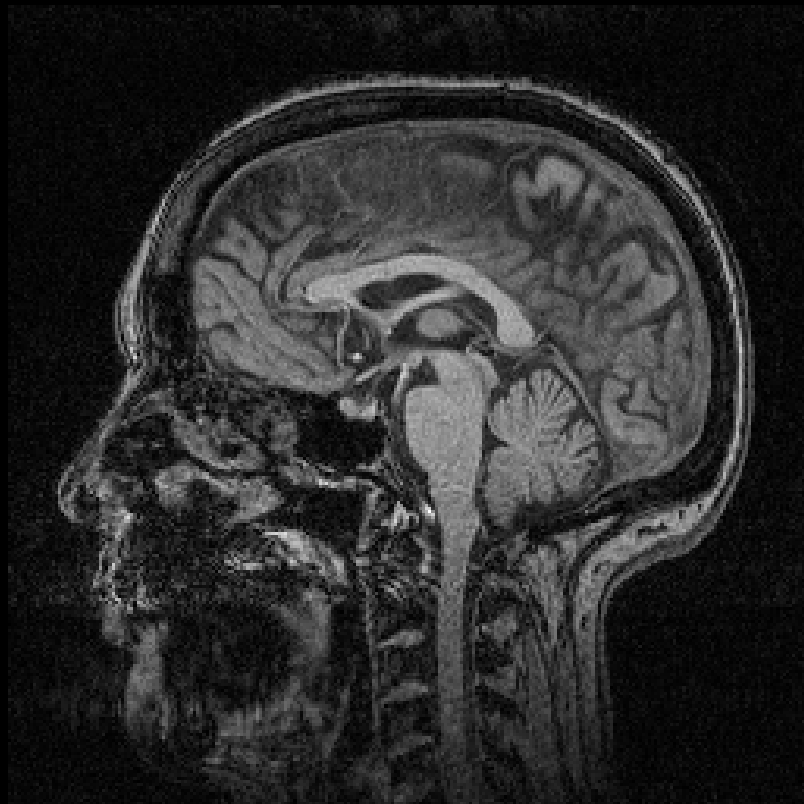
- X-rays
 - goes through soft tissue with little loss
 - are attenuated in bone
- CT scanners use X-rays
 - Good for imaging bones
- A simple threshold can often extract the bones
- Areas with only bone and soft-tissue will have a bimodal histogram



Attenuation - the gradual loss in intensity



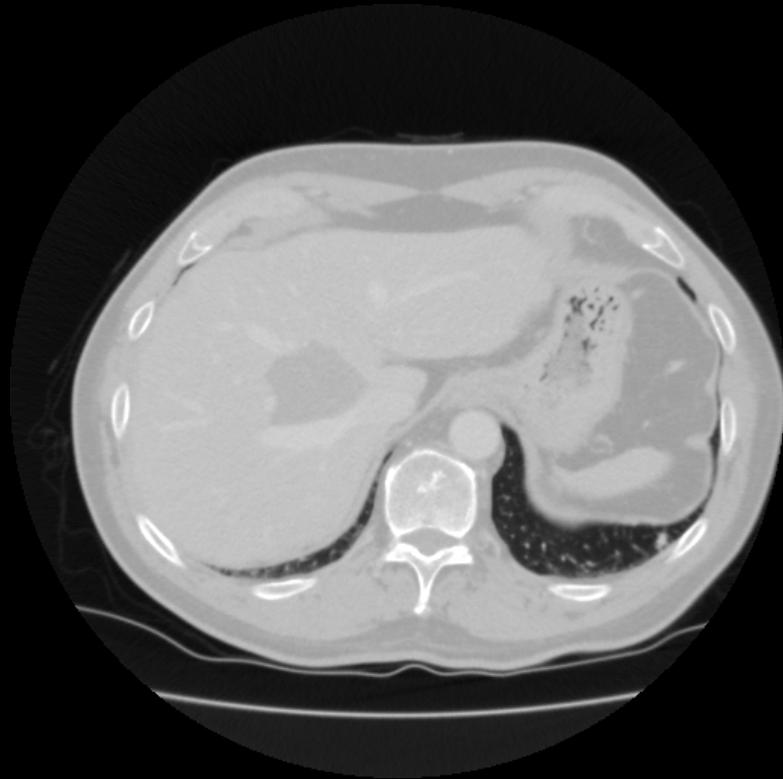
Image acquisition – brain structures



- Magnetic Resonance Imaging (MRI) is often used
- Much more difficult to explain!
 - Based on very powerful magnetic fields and radio waves
- Needs water molecules!
- Bone is black!



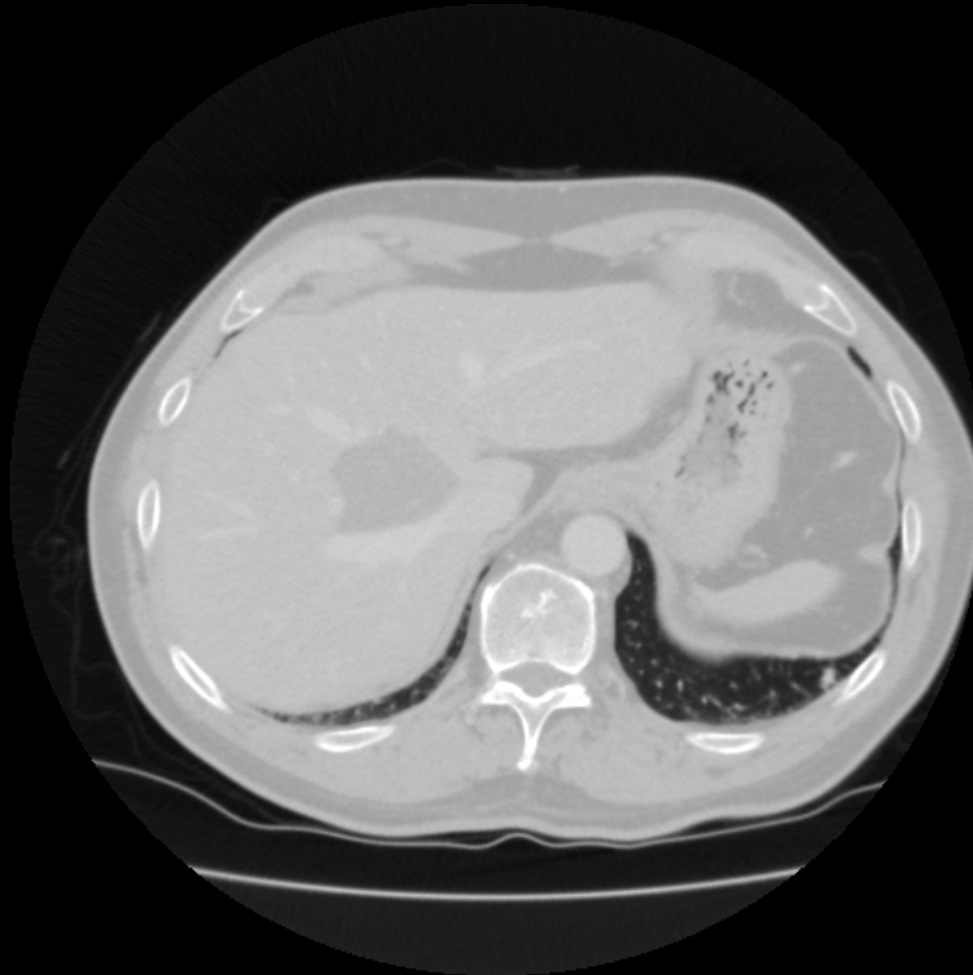
Image acquisition - cancer



- CT scan
- Liver cancer
 - Very difficult to see

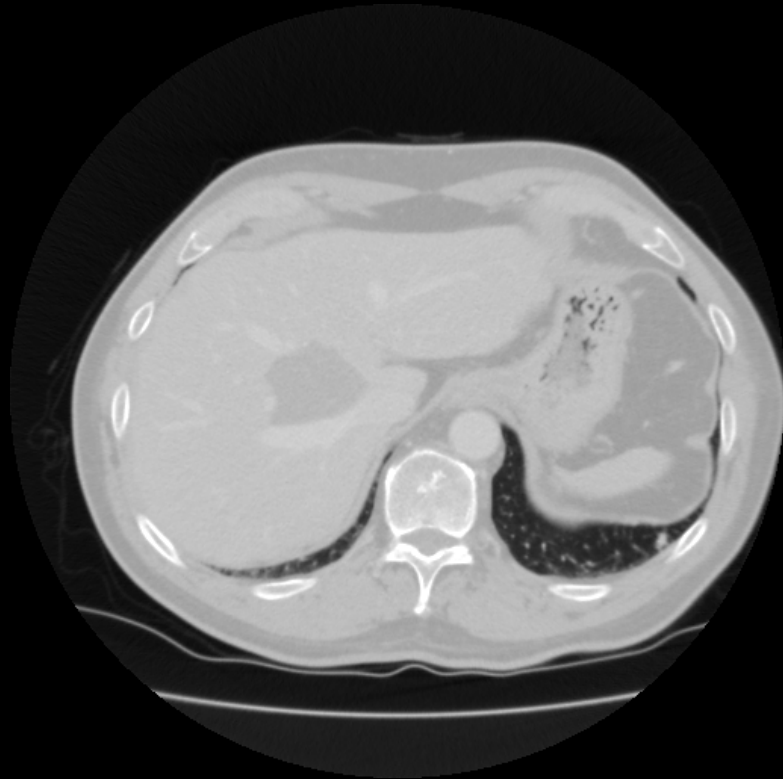


What makes cancer cells special?





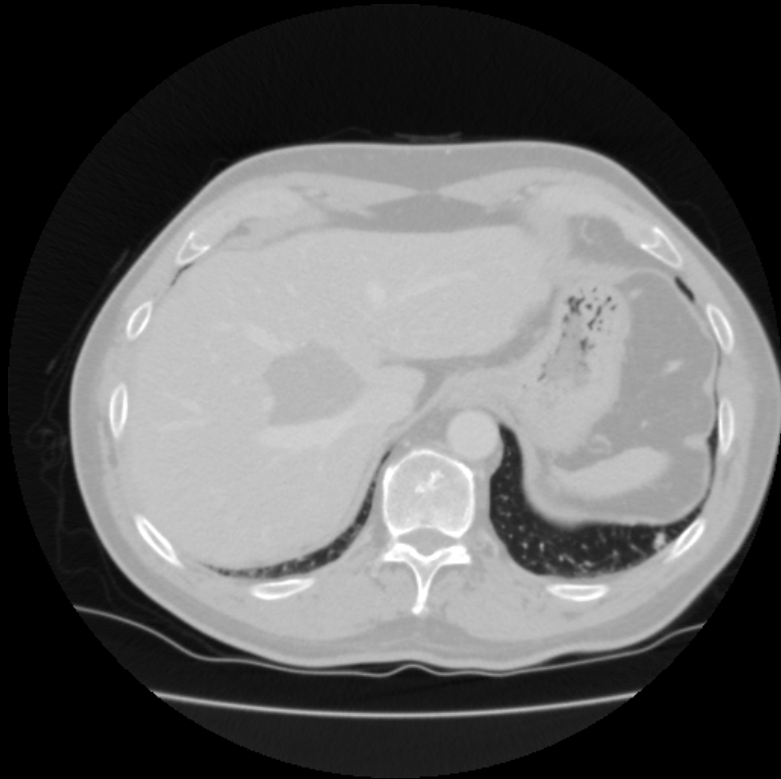
Cancer metabolism



- Cancer cells typically have a high metabolism
 - They eat more!
- Some substances are easier to see on different scanners
 - Bone on CT

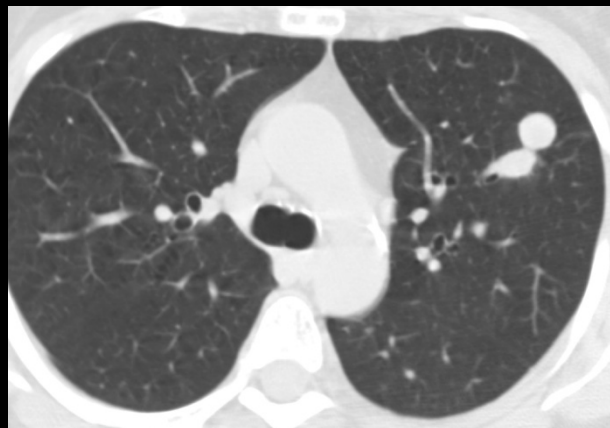


Using the cancer metabolism

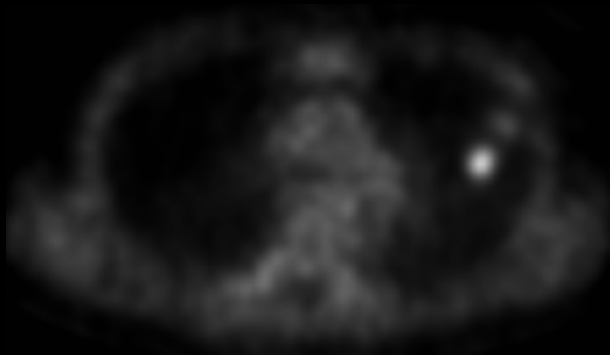


- Something that is to see
+
- Something that is being eaten by the cancer
- A tracer

Contrast using tracers



CT image

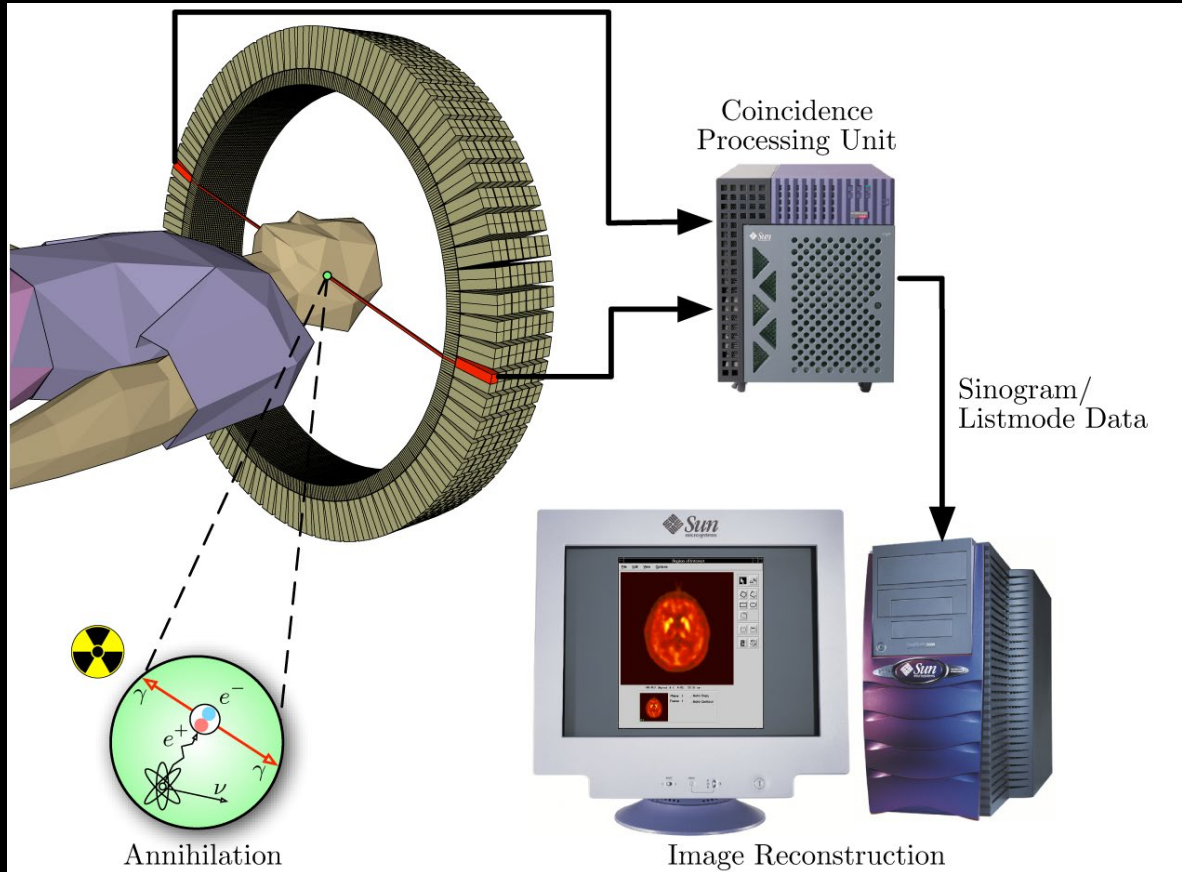


PET-FDG image

- A commonly used tracer is
 - ^{18}F -FDG = ^{18}F -fluorodeoxyglucose
- Used in *oncological PET*
 - *Oncology* : Cancer
 - *PET*: positron emission tomography
- Positron-emitting radioactive isotope fluorine-18
- Glucose is a “sugar”



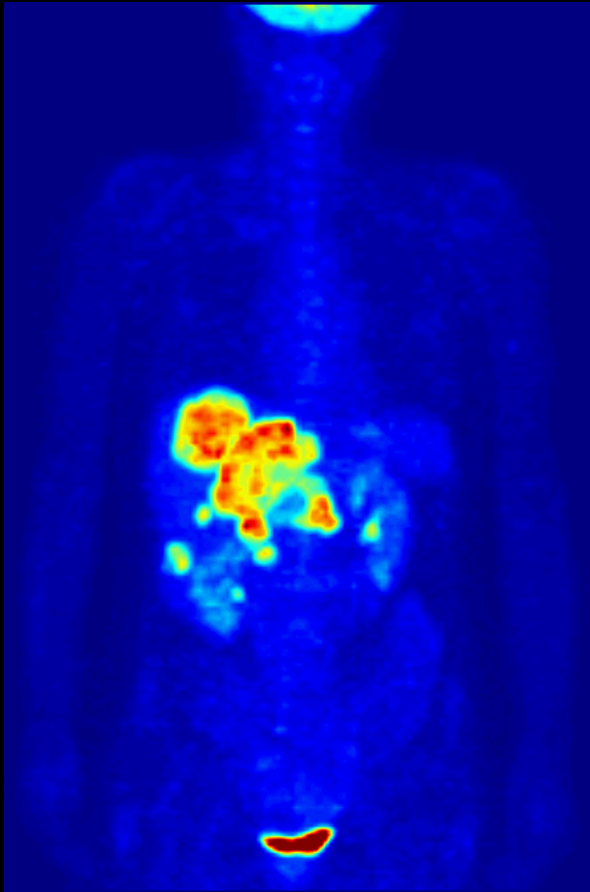
PET



Wikipedia



PET Image

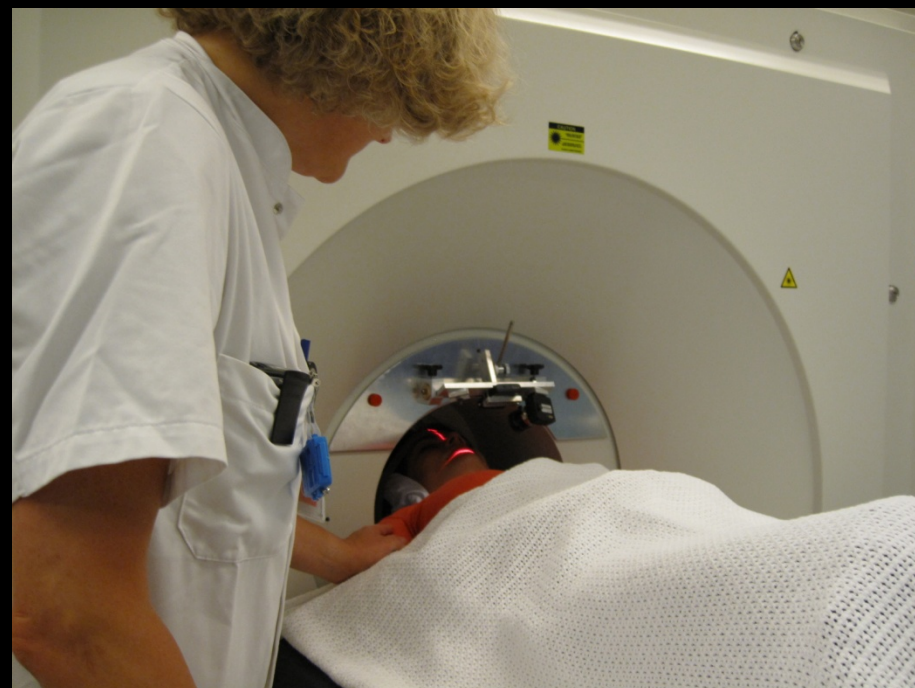
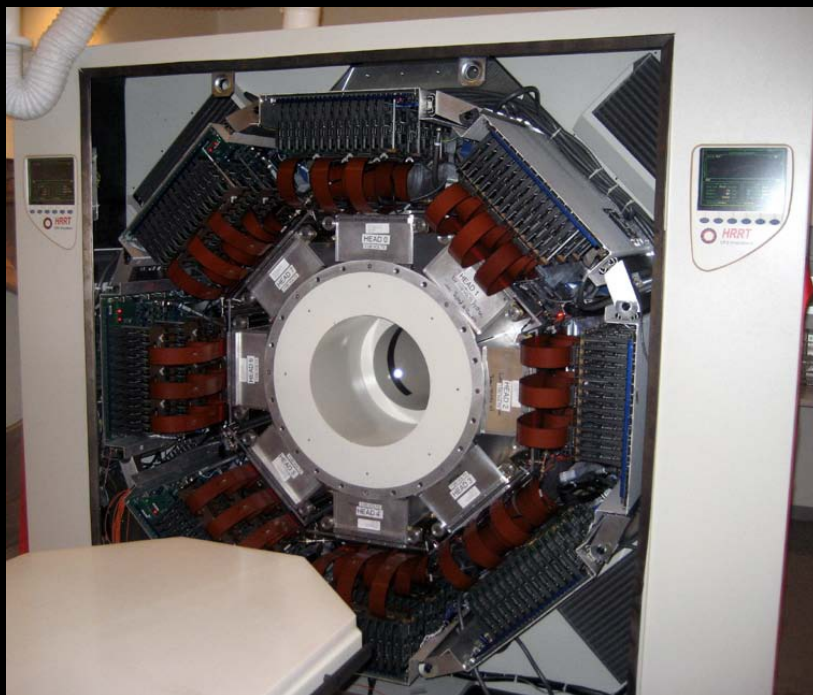


Wikipedia

- Areas with high glucose intake will be brighter
 - Higher intake of radioactive molecules
- Bimodal histograms in areas with cancer cells
- Big research topic



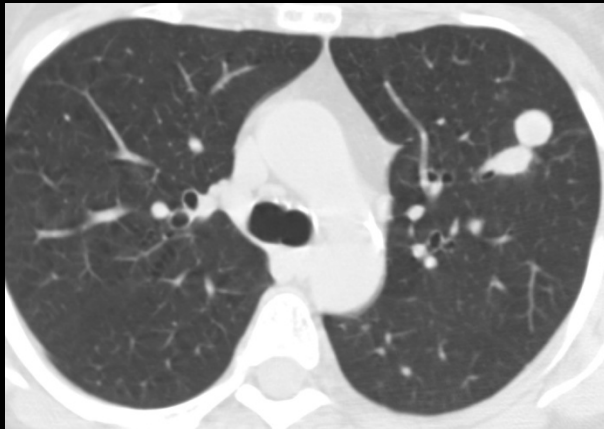
High-Resolution PET scanner at Rigshospitalet



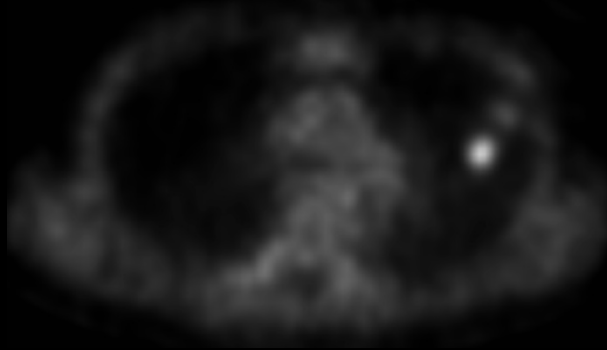
PhD project by Oline Vinter Olesen –
now startup



Combining Images

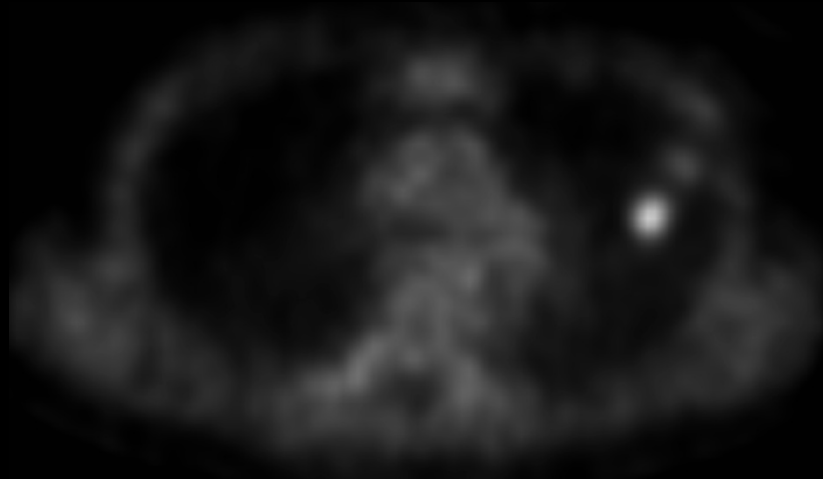


- CT is good for bone
- PET is good for cancer
- What if I want to see both?
- PET/CT scanner
 - Patient scanned in both a CT and a PET scanner
- Image registration
 - Take two or more separate images
 - Combine them using image registration
 - More about that later





Thresholds visited



- The tumour became much more separated from the background
- Perhaps a simple threshold is enough now?
- The best solution
 - Clever imaging techniques and
 - Intelligent image analysis



Level of the lectures

Far too easy

Too easy

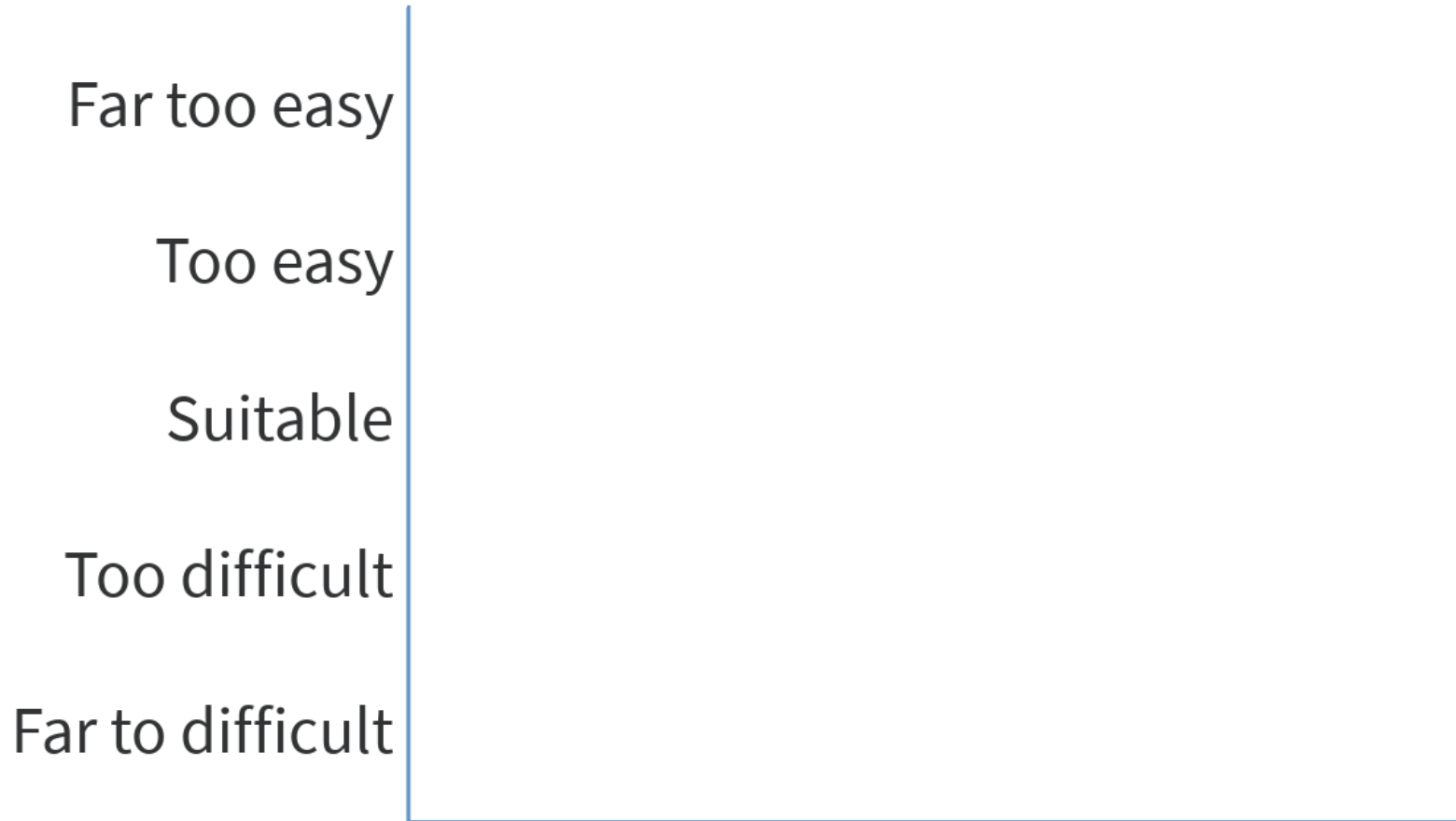
Suitable

Difficult

Too difficult



Level of the exercises





Use of PollAnywhere

I have enough -
please stop!

A little bit less - thank
you

Fine fine

I would like a little
bit more

I just can't get enough

Next week

- Neighbourhood processing
 - Filtering

