

# Image Analysis

Rasmus R. Paulsen

Tim B. Dyrby

DTU Compute

<http://compute.dtu.dk/courses/02515>

# Week 1 - today

8:00 – 10:00

Exercises

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10:00 – 12:00

Introduction and practical matters

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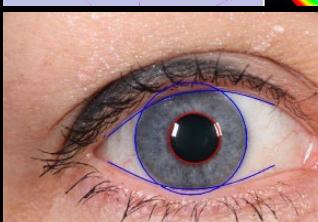
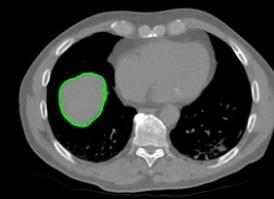
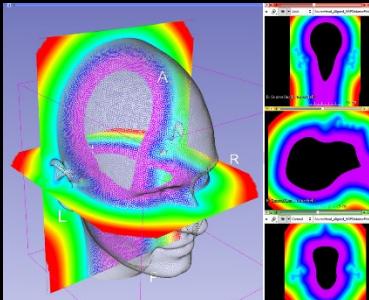
Lecture – An introduction to image analysis

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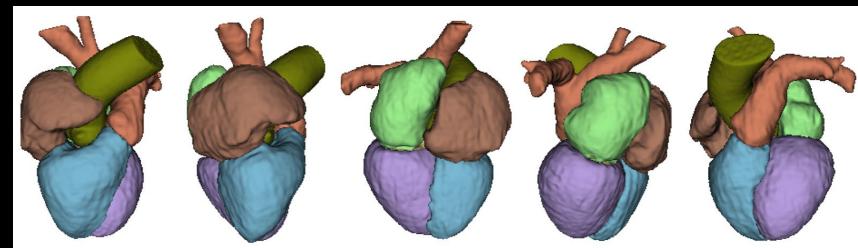
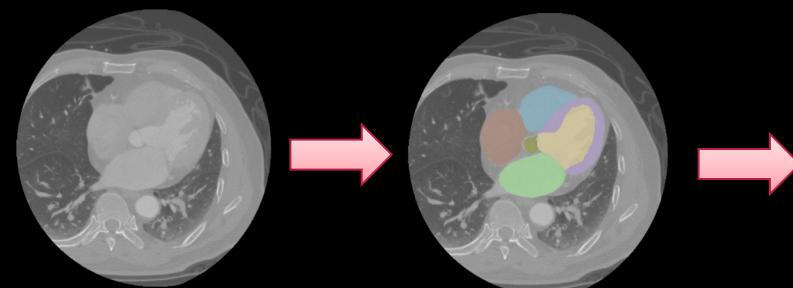
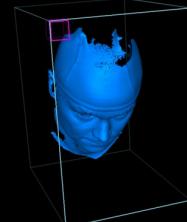
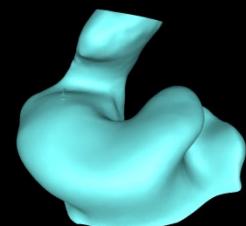
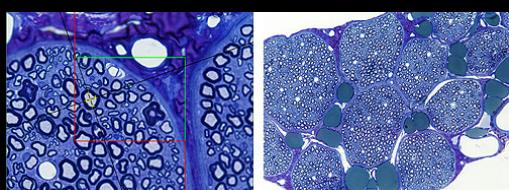
Lecture – A tutorial on Principal Component Analysis (PCA)

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# Rasmus R. Paulsen



- Master of Science (Eng). DTU 1998
- Industrial PhD with Oticon A/S
- Research and development at Oticon A/S
- Professor DTU Compute



# Tim B. Dyrby



■ Professor at DTU Compute and Danish Research Centre for Magnetic Resonance (DRCMR)

# Teaching Assistant

■ Andreas With Aspe

# Practical matters

- 13 days over the DTU 13 week semester
- Flipped class room
  - 8-10 Computer exercises (also on MS Teams)
  - 10-12 Lecture with quizzes
- Lectures are streamed, recorded and made available
  - Links to the stream will be posted on DTU Learn before the lecture
  - Links to video on the homepage (under schedule)
  - [Courses.compute.dtu.dk/02515](http://Courses.compute.dtu.dk/02515)

# The different versions of the image analysis course

- 02502: The previous combined bachelor/master level course
  - Terminated in 2025 since DTU is not allowed to do combine BSc/MSc anymore
- 02503: The bachelor level version of the image analysis course
  - Running from fall 2025
- 02515: The master level version of the image analysis course
  - Running from spring 2026
- The learning objectives and course content is very similar for the three courses. You can only get credits for one of them

# The exercises and the exam

- The exercises are very related to the exam
- Learning objectives stated in all exercises
- You will be examined in these learning objectives
- You will also be examined in the more theoretical learning objectives from the lectures
- We expect that you can run Python during the exam!

**Very Important I: Do the exercises!**

**Very Important II:**  
**We can not help you after the course period!**

# Materials

## ■ Book:

- Rasmus R. Paulsen and Thomas B. Moeslund: *Introduction to Medical Image Analysis* (**MIA**)
- Polyteknisk boghandel
- <http://mediabook.compute.dtu.dk>

## ■ Notes

- Notes will be provided during the course

## ■ At the end of the course a complete reading list will be published

# DTU Learn and the homepage

- Homepage : The main entry to the course
  - <http://courses.compute.dtu.dk/02515>
  - Schedule / Exercises / Data
  - Updates happen!
- Course messages will be given through DTU Learn

#	Date	Topic	Video	Material	Exercise
1	3/2	Introduction to image analysis (Rasmus) Introduction to Principal Component Analysis (PCA) (Rasmus)	Stream	MIA 1, 2, app. A. PCA Note (except Section VI (SVD) and App. A)	1
2	10/2	Cameras, lenses, image compression, image storage and change detection in videos (Rasmus)	Stream	MIA 2, 3 + CDV Note	1 + 1b
3	17/2	Pixelwise operations, Colour images. PCA Analysis on images (Rasmus)	Stream	MIA 4, 8 Eigenfaces article (only sections marked with yellow)	2 + 2b
4	24/2	Neighborhood Processing (Filtering) and Morphology (Tim)	Stream Recordings Fall 2023	MIA 5, 6	3 + 8
5	3/3	Blob analysis and object classification (Rasmus)	Stream	MIA 7	4 + 4b
6	10/3	Pixel classification and advanced classification (Tim)	Stream Recordings Fall 2023	MIA 9 + LDA note on Learn	5
7	17/3	Industry presentations: Radiobotics DaluX JLIVision TrackMan Milestone Videometer IHfood Visiopharm	We cannot stream the Industry presentations	none	Exercise catch-up
8	24/3	Geometric transformations and landmark based registration (Tim)	Stream Recordings Fall 2023	MIA 10, 11	6
9	7/4	Boundary Tracing (Hough Transformation and Dynamic Programming) (Tim)	Stream Recordings Fall 2023	MIA 12	6b
10	14/4	Advanced registration (Tim)	Stream Recordings Fall 2023	Elastix manual (5.2.0) chapter 2.	7
11	21/4	Real time face detection using the Viola Jones method (Rasmus)	Stream	Rapid Object Detection using a Boosted Cascade of Simple Features	9
12	28/4	Statistical models of shape and appearance and active shape models (Rasmus)	Stream	Statistical Models of Appearance for Computer Vision (p. 12 - 20 and p. 29 - 43)	Digital test exam and exercise catch-up
13	5/5	Advanced topics (Claes Nehr Ladefoged)	Stream	none	Digital test exam and exercise catch-up

# Learning Objectives (Læringsmål)

- A list of learning objectives for each lecture and exercise
- A learning objective describes what you can do after the lecture/exercise
- If you fulfil all learning objectives you get 12
- Low-level learning objective
  - Apply the Prewitt edge filter to an image
- High-Level learning objective
  - Evaluate and compare the performance of a selection of image analysis algorithms

# Exam

- Four hours multiple-choice exam
- Please see details here:
  - <http://courses.compute.dtu.dk/02515/exam.html>
- Previous exam sets are also available
  - Most relevant is from Spring 2021 and onwards

# AI assisted tools during the course and the exam

- You are allowed to use AI tools like ChatGPT and Copilot
  - Both during the course and at the exam
- It is your responsibility to
  - install and keep your tools up to date
  - Verify if the output of the tools are correct
- The exam can ALSO be solved without the use of AI assisted tools

# AI related learning objectives

- General 02515 course learning objective:
  - Estimate the correctness of the answer given by an AI assisted tool like ChatGPT and Copilot
- We are gradually adding AI tools related learning objectives to the exercises

# PollEverywhere quizzes

<https://pollev.com/rasmuspaulse538>



## What programming language are you most comfortable with?



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## What is your practical experience with Python programming

I implement and run Python scripts (.py) in an IDE like Jupyter, PyCharm or Visual studio code

81%

I use Notebooks in JupyterLab or similar

15%

I use google Colab or a similar online service

4%

I use another method or tool for my Python programming

0%

I do not have any Python experience

0%

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## What is your experience with Jupyter Notebooks or JupyterLab

I am a complete beginner - I need instructions on software installation and a crash course in Jupyter notebooks.

20%

I have tried it a few times but I need a refresher

4%

I am comfortable with Jupyter notebooks but would not mind a quick refresher

40%

I am all ready - hit me!

28%

I am an expert Jupyter notebook user and have made my own

8%

## What is your experience with image manipulation, image processing and image analysis?

I have never manipulated an image



I have done cropping, rotation and colour enhancements on my phone or in a photo editor



I have used Photoshop or similar to do advanced image manipulation



I have used an image analysis tool in Python, Matlab, C# or similar



I have implemented and tested my own image analysis program



## To what extent are you currently using AI assisted tools like ChatGPT and Copilot

26

They are invaluable tool in my learning and university life

23%

I use them actively and often

54%

I am an occasional user

19%

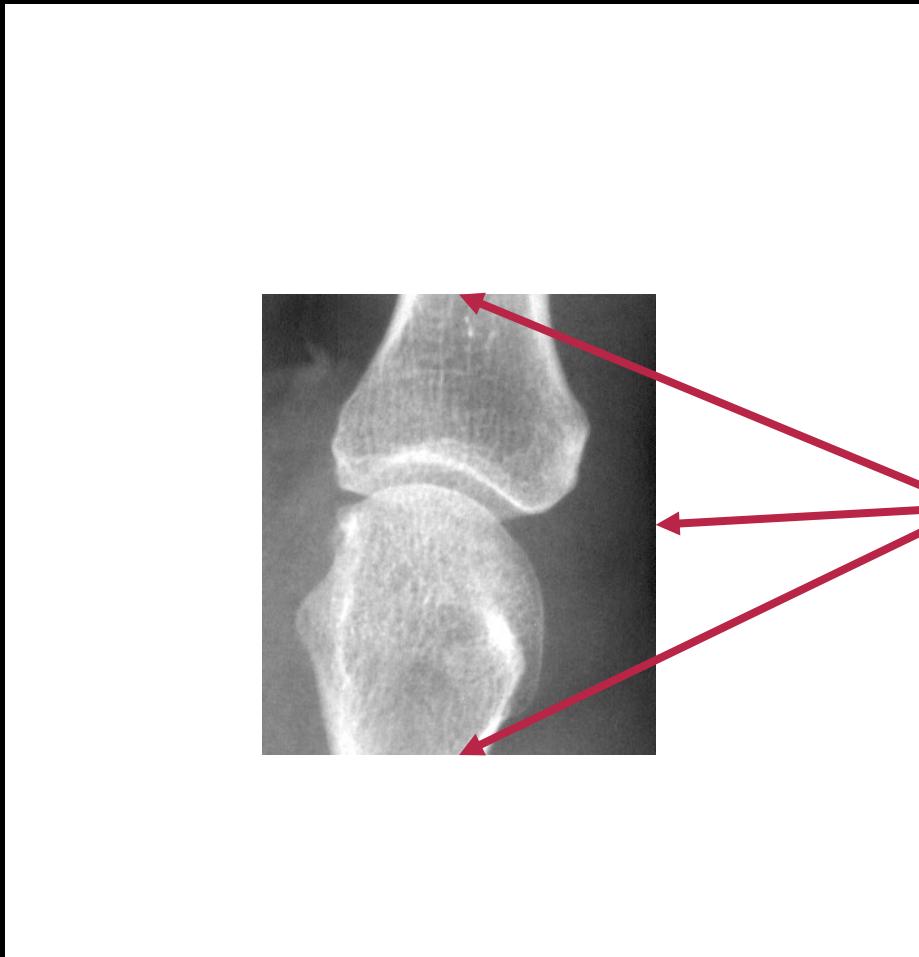
I rarely use the tools

4%

I have never or very rarely used these tools

0%

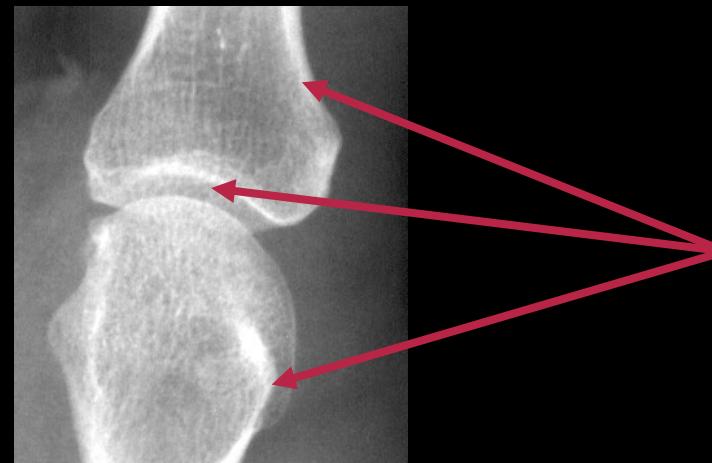
# Why are my slides black?



Norwegian Black Metal

With a white background,  
the strongest visual  
contrast is here

# Why are my slides black?



With a dark background, the strongest visual contrast is here  
(which I find more important)

# What is image analysis

- Automatic extraction of information from images
- A sub-topic within
  - Pattern recognition
  - Machine learning
  - Deep learning

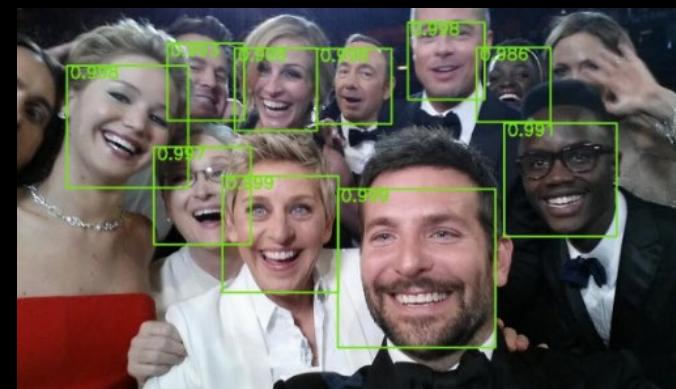
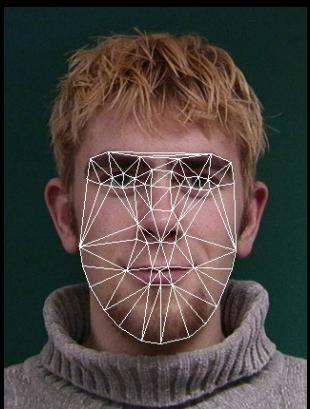
# What is image processing

- Changing the information in images – but not necessarily getting any knowledge
  - Photoshopping
  - Changing the visual appearance of photos
  - Cropping / rotating
  - Filters / effects



# Face tracking – all features including eyes

- For digital cameras / phones
  - Automatic focus on the face + face beautification
- Tracking and manipulation for apps
  - Messenger / WhatsApp / SnapChat ...
- Awareness tracking for car drivers
  - Warning if you fall a sleep



# A 100 million \$ industry



- This image is worth 100 of millions of dollars!
- Well – perhaps not that exact photo.
- The ability to track faces fast and accurate
  - Including estimates of 3D structure
  - App developers pays buckets of money for that
- It all started in 2001 with:  
P. Viola and M. Jones. "Rapid object detection using a boosted cascade of simple features.". CVPR 2001
- Suddenly you could track faces fast and relatively accurate
- Now a lot of focus on deep learning

# Self driving cars

## ■ Modern self driving cars rely on many sensors

- Lidar – radar system
- GPS
- Accelerometers, gyroscopes, magnetometers etc.
- Stereo cameras or multiple cameras
- Lots of advanced image analysis – sensor fusion



# Sports tracking – human body tracking

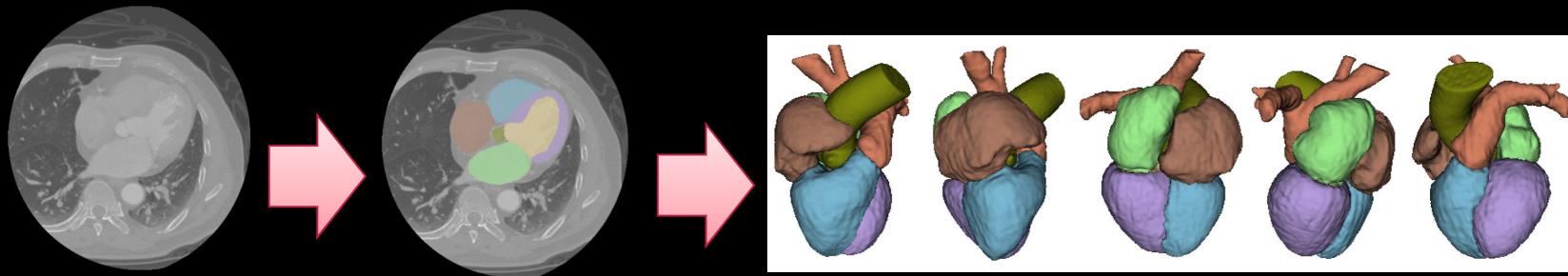


- Huge commercial impact
- Lots of research in human body tracking
- Personal trainers
- TV player tracking and smart overlays



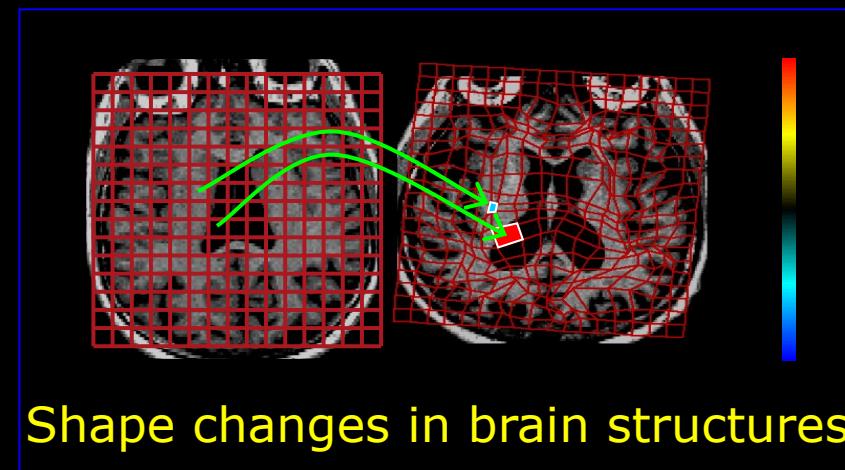
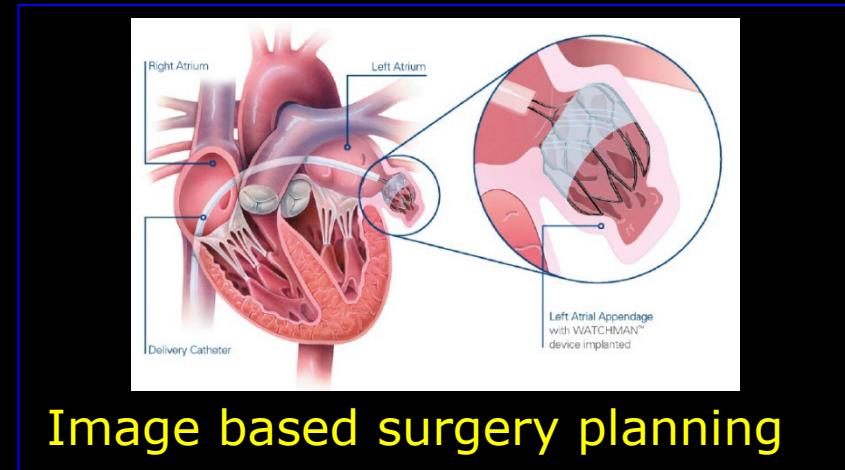
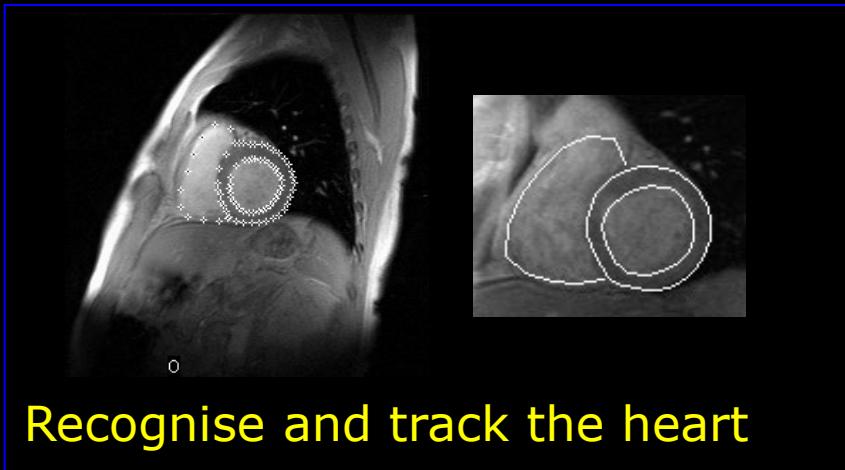
# What is medical image analysis?

- Extraction of information from digital images
- Find unknown connections between diseases and what can be seen in images
- Can enhance the signs of diseases
  - Tumours / heart diseases / brain diseases / bone fractures
- Reproduce expert diagnostics
  - More accurate
  - Variation between doctors opinions removed
- Computer aided diagnostics



Automatic localization of the heart and its major substructures

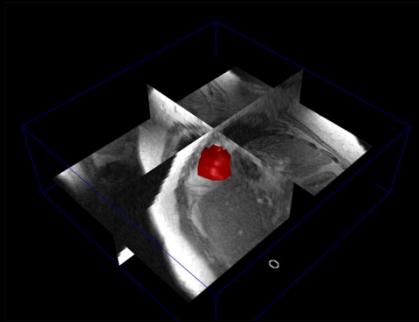
# Medical image analysis examples



# Relevance



1980  
Magnetic  
resonance  
prototype

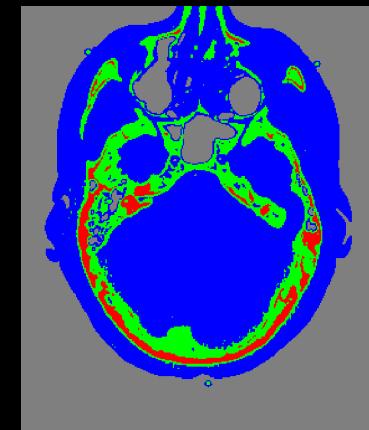
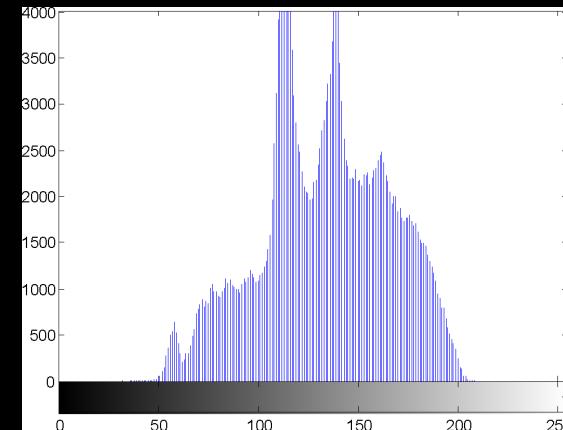


Now – PET/MR

- Images is an important tool in
  - Diagnosis
  - Treatment
  - Follow-up
- Very high-tech!
- New imaging technologies are developed all the time.

# Digital Images – Learning Objectives

- Describe the fundamental properties of a digital image
- Describe and use the commonly used image coordinate systems
- Describe pixel types
- Describe the binary, the color, the label, the multispectral, the floating point, and the 16-bit image



# A digital image

23	216	120	55
4	89	158	130
65	76	189	34
19	234	7	45

- Consists of pixels (picture elements)
- Each pixel has a value between 0 and 255? Why?

# Bits and Bytes!

- A **bit** is a tiny tiny little switch that can be either 0 or 1 – the “memory of a computer” consists of insanely many bits
- One **byte** is 8 bits together. It is the “basic” unit in a computer.
- With 8 bits how many possible values can be made?
  - $(2^8 = 256)$
  
- 00000001 = 1
- 00000010 = 2
- 00000100 = 4
- 00001010 = 10
- 00001111 = 15
- 11111111 = 255

128	64	32	16	8	4	2	1
<input type="checkbox"/>							

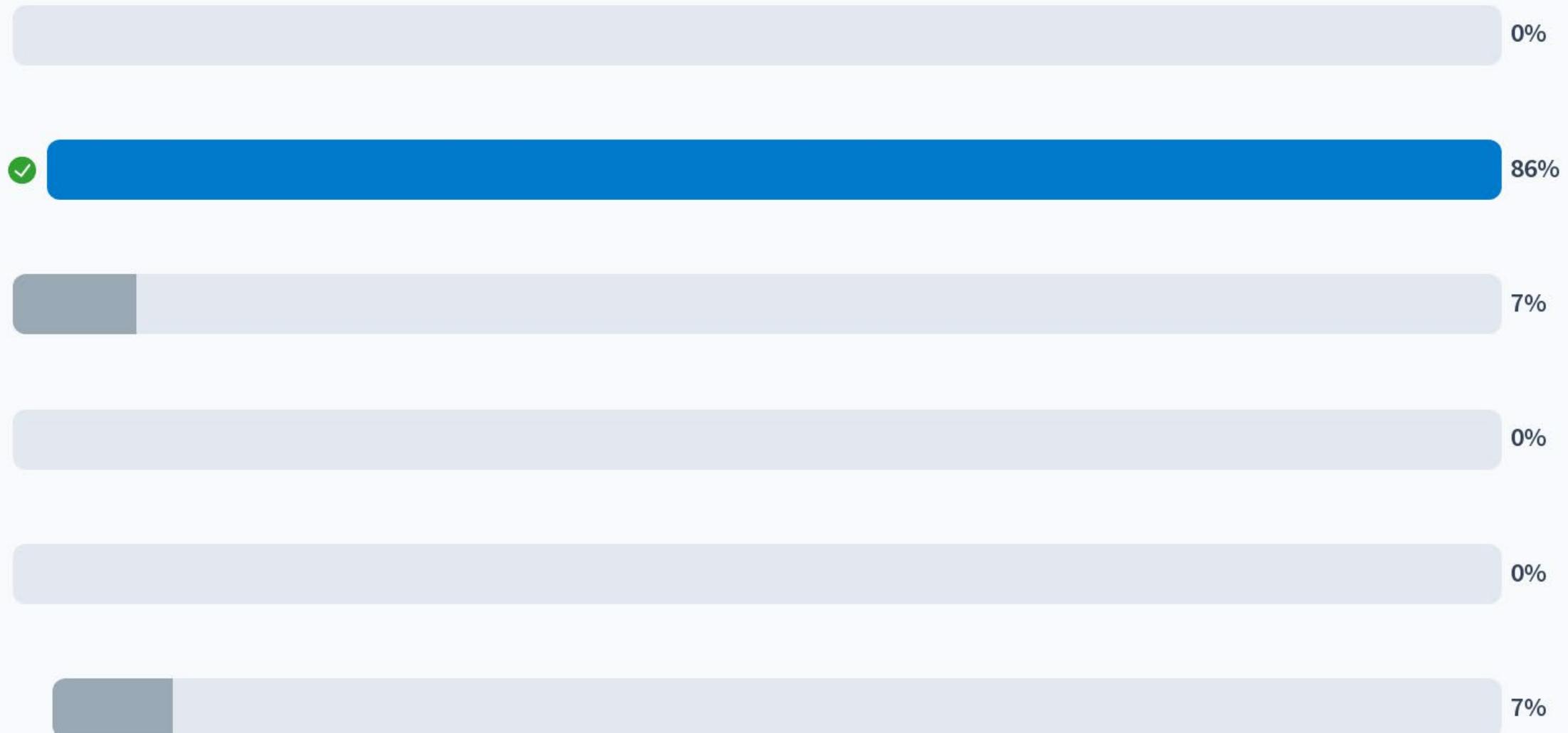
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## What is decimal 67 as a binary number?



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## What is decimal 67 as a binary number?



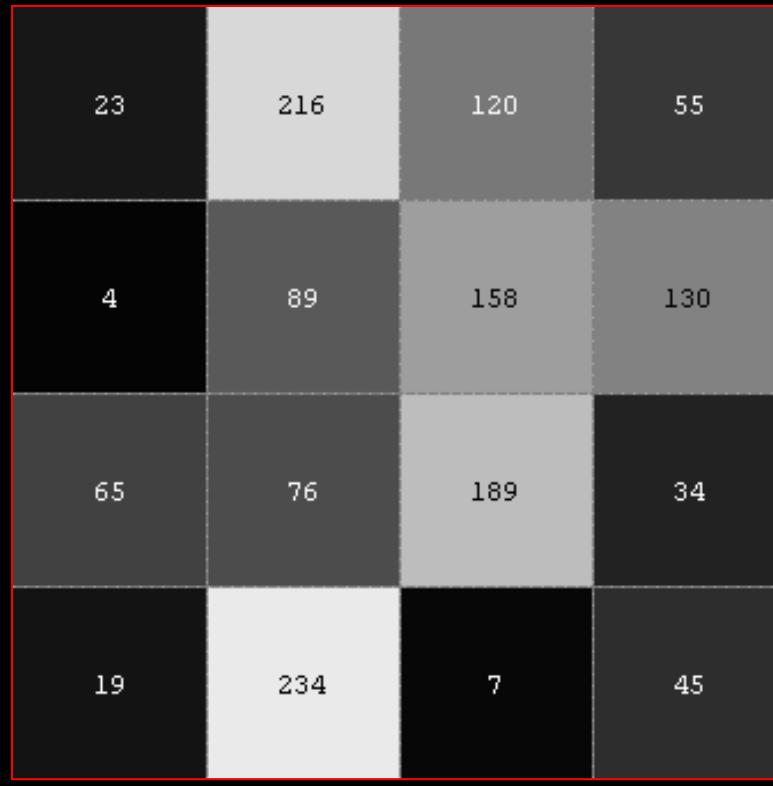
Start the presentation to see live content. For screen share software, share the entire screen. Get help at [pollev.com/app](https://pollev.com/app)

# A digital image

23	216	120	55
4	89	158	130
65	76	189	34
19	234	7	45

- between 0 and 255.
- The pure image data takes up 16 bytes of computer memory

# Grayscale digital images



- 0 is black and 255 is white!
- The values in between are shown as shades of gray



# Typical Grayscale image



- Traditional film X-ray
- Scanned on a flatbed scanner
- Bone is white and air is black
  - The more radiation the darker
- What are they used for?
  - Fractures
  - Arthritis
  - Osteoporosis

# Image Resolution

- Determines how much the image fills in the memory and on the hard disk
- Spatial resolution
- Gray level resolution

# Spatial?

## ■ Spatial

- relating to the position, area and size of things

## ■ Example:

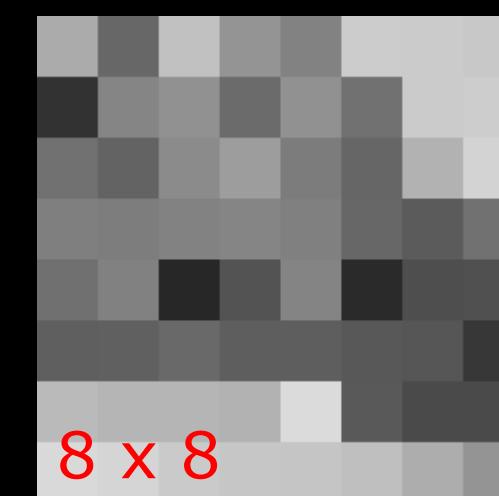
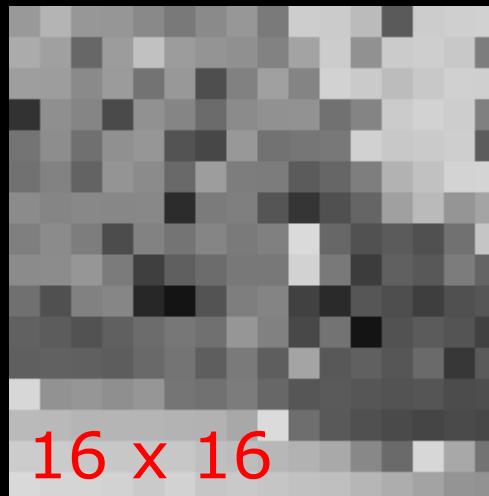
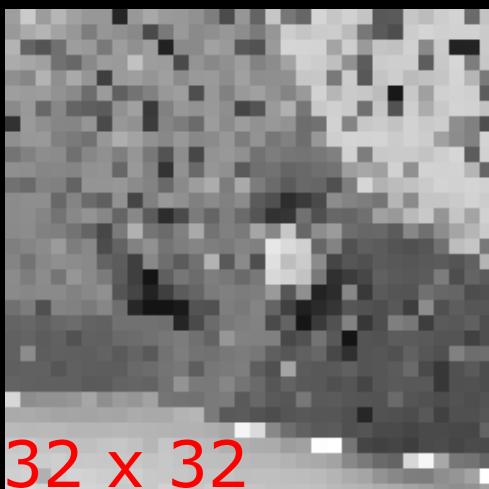
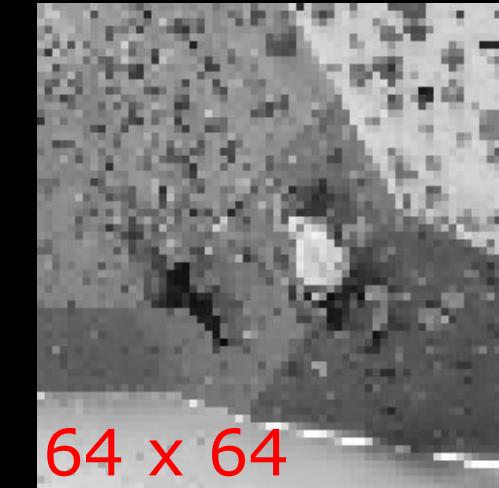
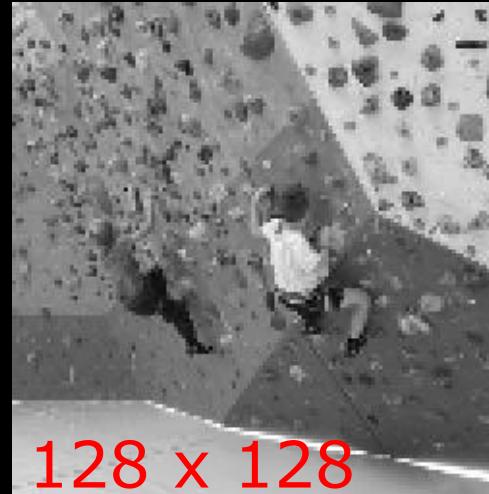
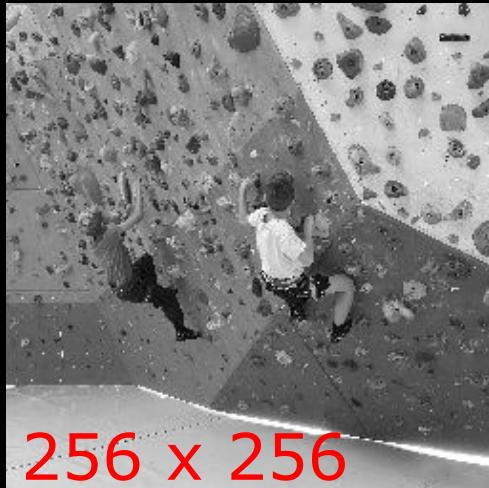
- This task is designed to test the child's *spatial* awareness

## ■ Danish

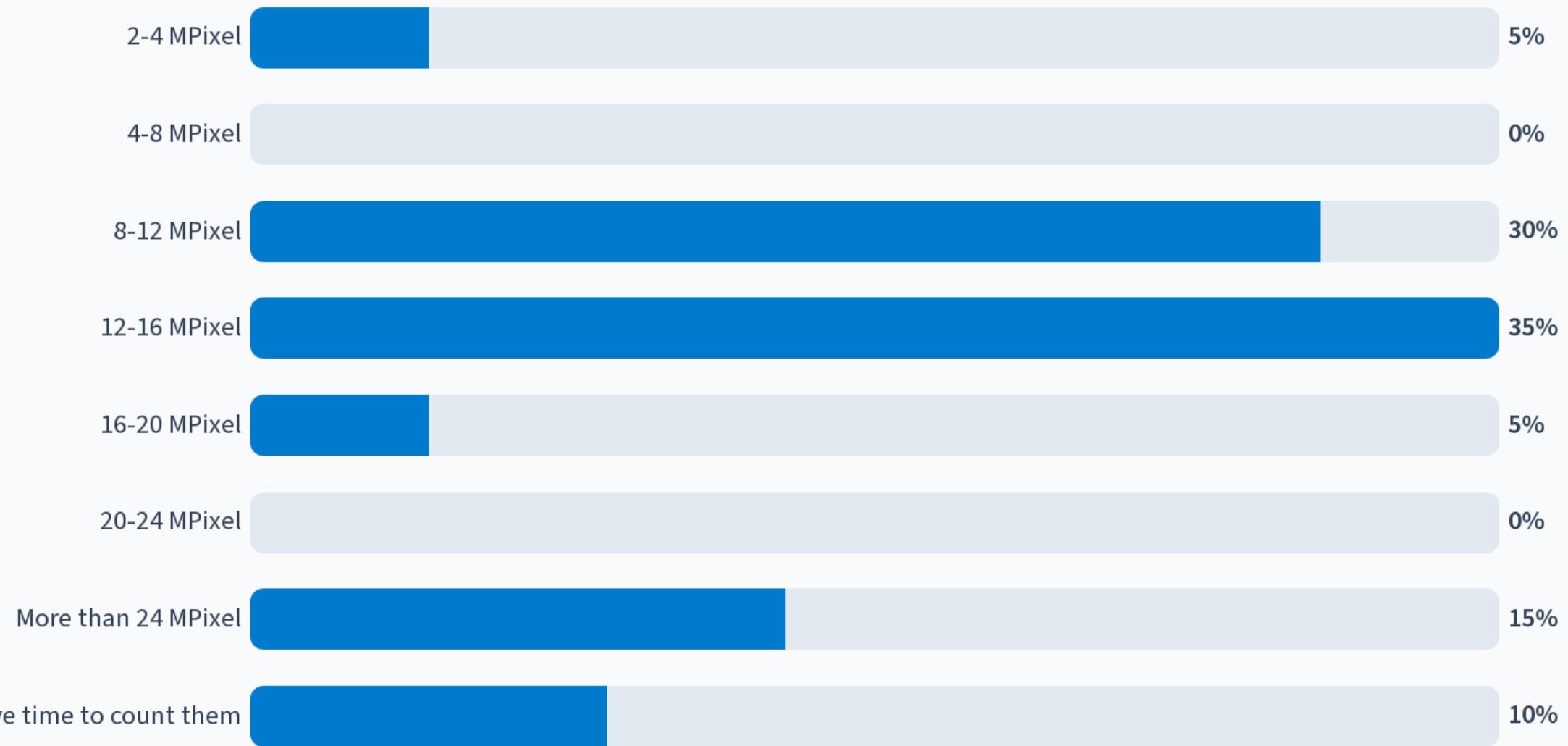
- Rumlig – jeg har en god rumlig forståelse

# Spatial resolution

The number of pixels used to represent the image



## How many megapixels (approximately) do the photos you take with your camera or phone have?



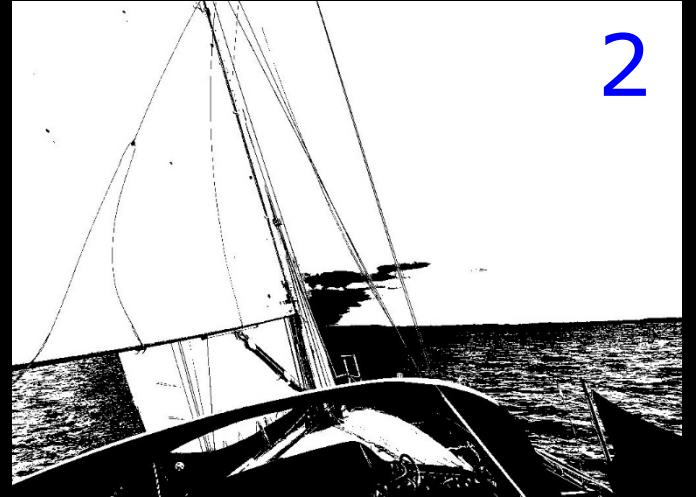
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# How many pixels?

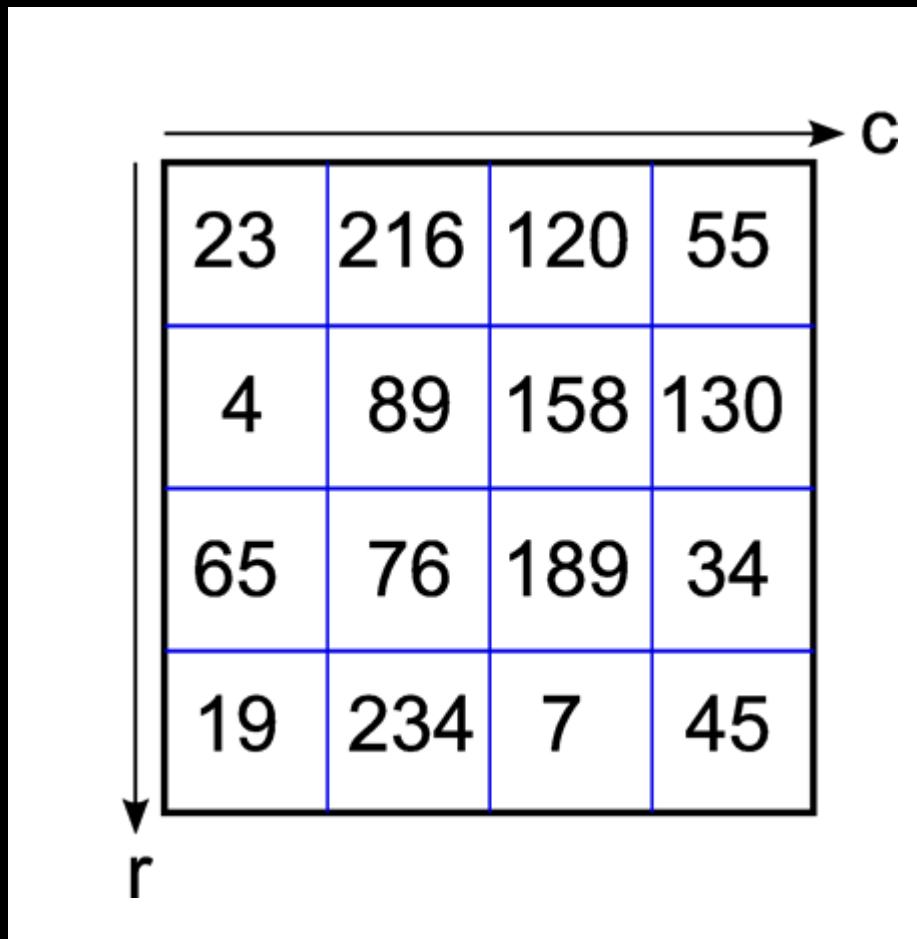
Width	Height	Pixels	Mega-pixels	Camera
320	240	10.000	0.01	Prototype 1975
1600	1200	1.920.000	2	Nikon Coolpix 950
4032	3024	12.192.768	12	Samsung Galaxy S7 edge
6240	4160	26.000.000	26	Canon EOS 6D M2
8984	6732	60.480.288	60.5	Phase One P65+

# Grey level resolution

## The number of grey levels in an image



# An image as a matrix



- An image is stored in the computer memory as a 2 dimensional matrix
- 4 rows and 4 columns
- Can also be seen as a discrete function  $f(r, c)$
- In Python a pixel can be stored as an `uint8`
- `uint8` = Unsigned 8-bit integer = 1 byte

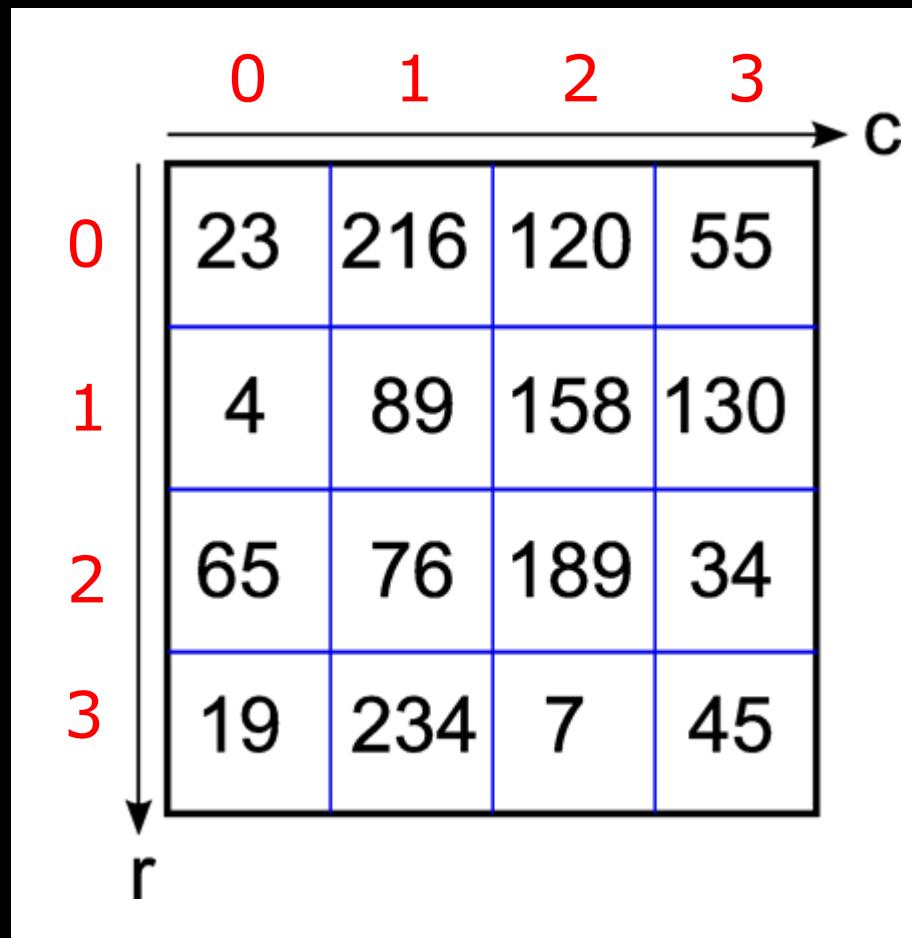
# Pixel types and their ranges

Data type	Range
uint8	0 to 255
uint16	0 to 65535
uint32	0 to $2^{32} - 1$
float	-1 to 1 or 0 to 1
int8	-128 to 127
int16	-32768 to 32767
int32	$-2^{31}$ to $2^{31} - 1$

- A pixel can be processed and stored as different *types*
- The uint8 is the most common type
- For processing a pixel is often transformed to a float
- When processing speed and memory space is an issue you should be careful about the pixel type – more about that later in the course.

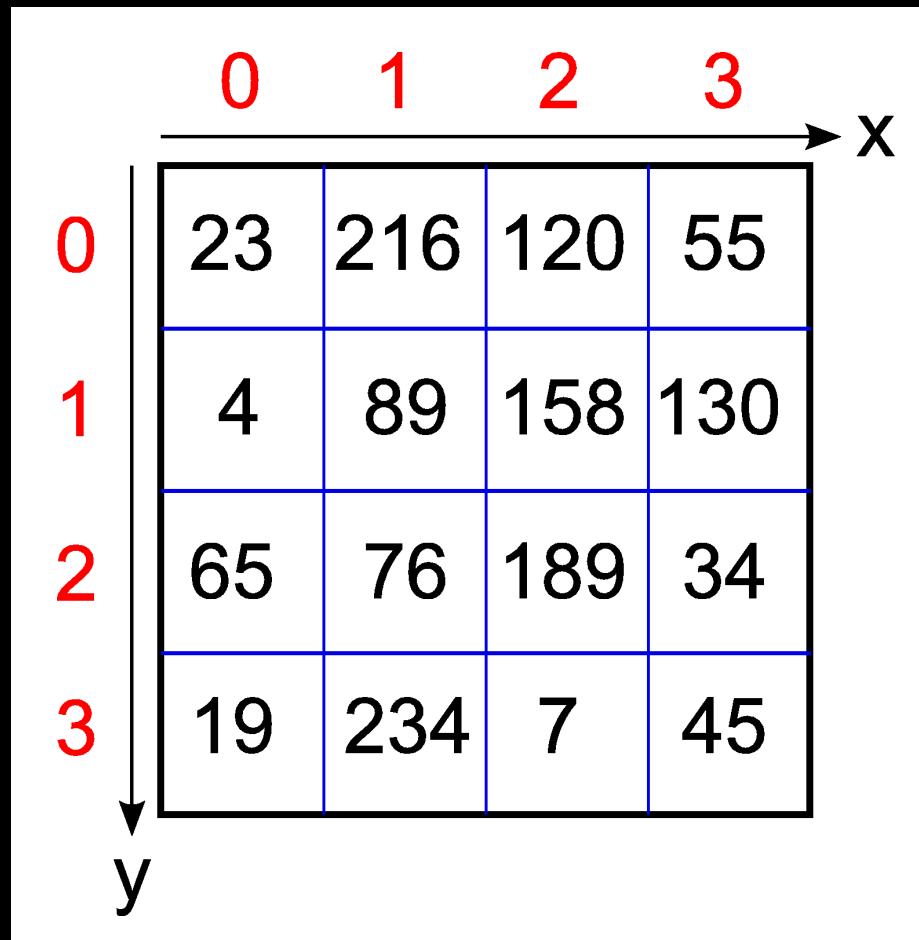
[https://scikit-image.org/docs/stable/user\\_guide/data\\_types.html](https://scikit-image.org/docs/stable/user_guide/data_types.html)

# Pixel coordinates – Python matrix



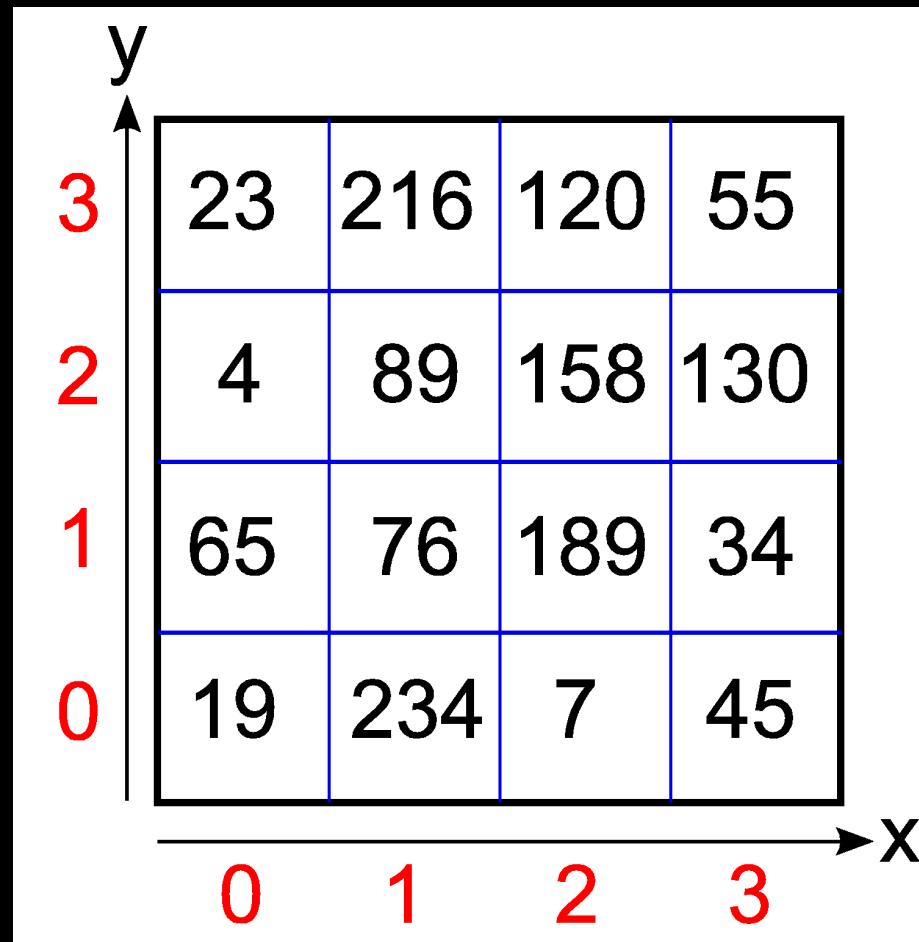
- Origin is in upper left corner
- 0-based
- (row, column) system
  - Vertical axis is the first axis
- M rows and N columns
- Row range  $[0, M-1]$
- Column range  $[0, N-1]$

# Pixel coordinates – Photoshop etc.



- Used in many graphics programs
- Origin in upper left corner
- 0-based
- (X,Y) system
  - Horizontal axis is the first coordinate
- Often width (W) and height (H) are used to denote image dimensions
- X range [0, W-1]
- Y range [0, H-1]

# Plot coordinates

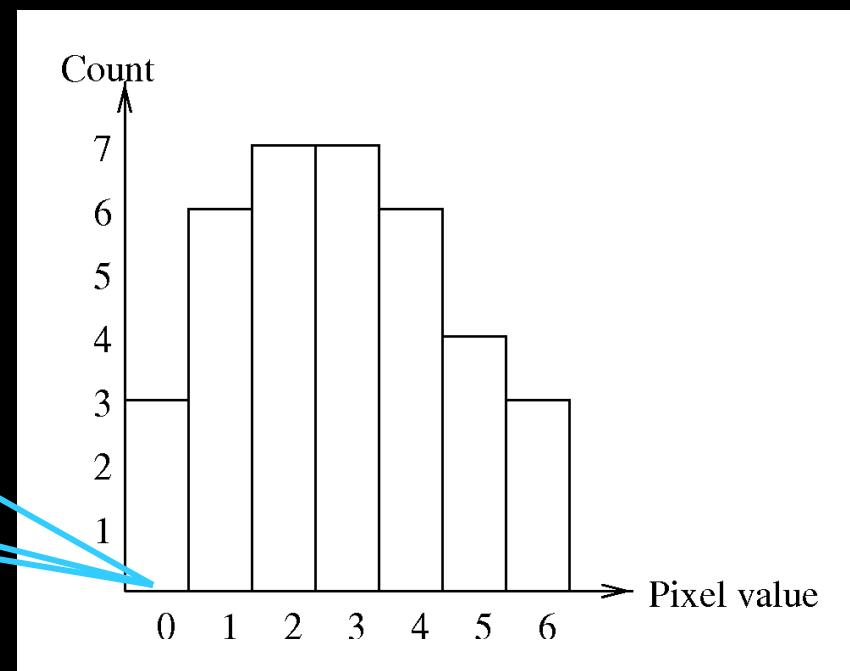


- Used when plotting – known from mathematics
- Origin in lower left corner
- 0-based
- (X,Y) system
  - Horizontal axis is the first axis

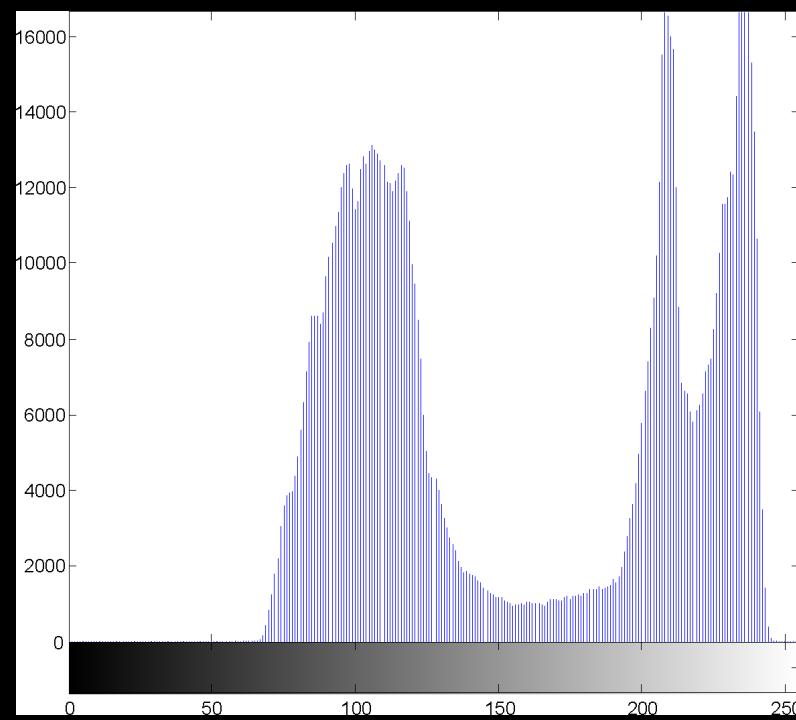
# The Image Histogram

- 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



# A real grayscale image histogram



- 256 gray levels in the image = 256 bins in the histogram
- The shape of the histogram tells us something about the image

# Where are the flower leaves in the histogram?

Range 1

Range 2

Range 3

Range 4

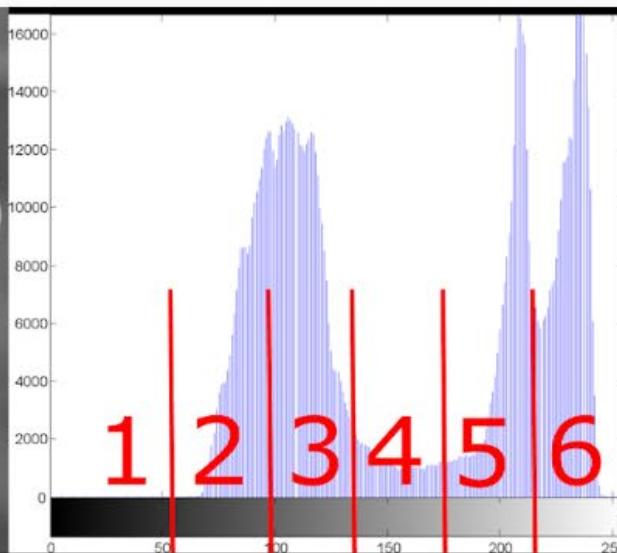
Range 5

Range 6

Oh bugger - my flower allergy has been triggered

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## Where are the flower leaves in the histogram?



Range 1 0%

Range 2 8%

Range 3 0%

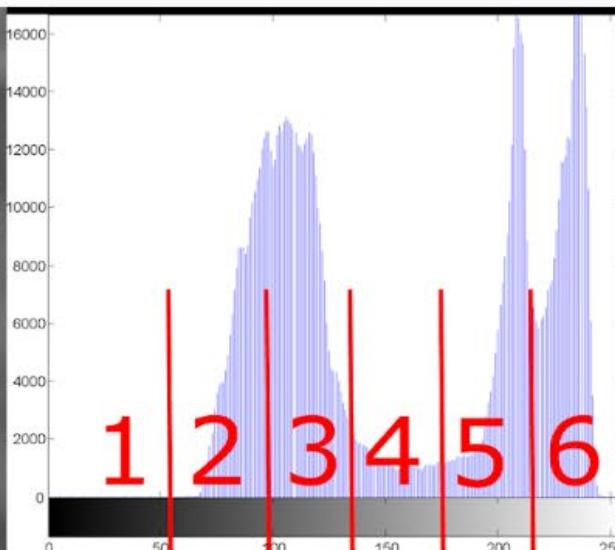
Range 4 0%

Range 5  8%

Range 6  83%

Oh bugger - my flower allergy has  
been triggered 0%

## Where are the flower leaves in the histogram?



Range 1 0%

Range 2 8%

Range 3 0%

Range 4 0%

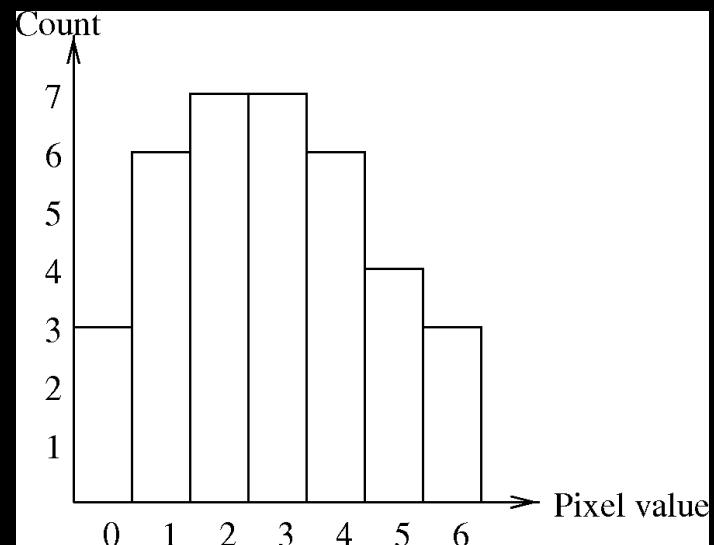
Range 5  8%

Range 6  83%

Oh bugger - my flower allergy has  
been triggered 0%

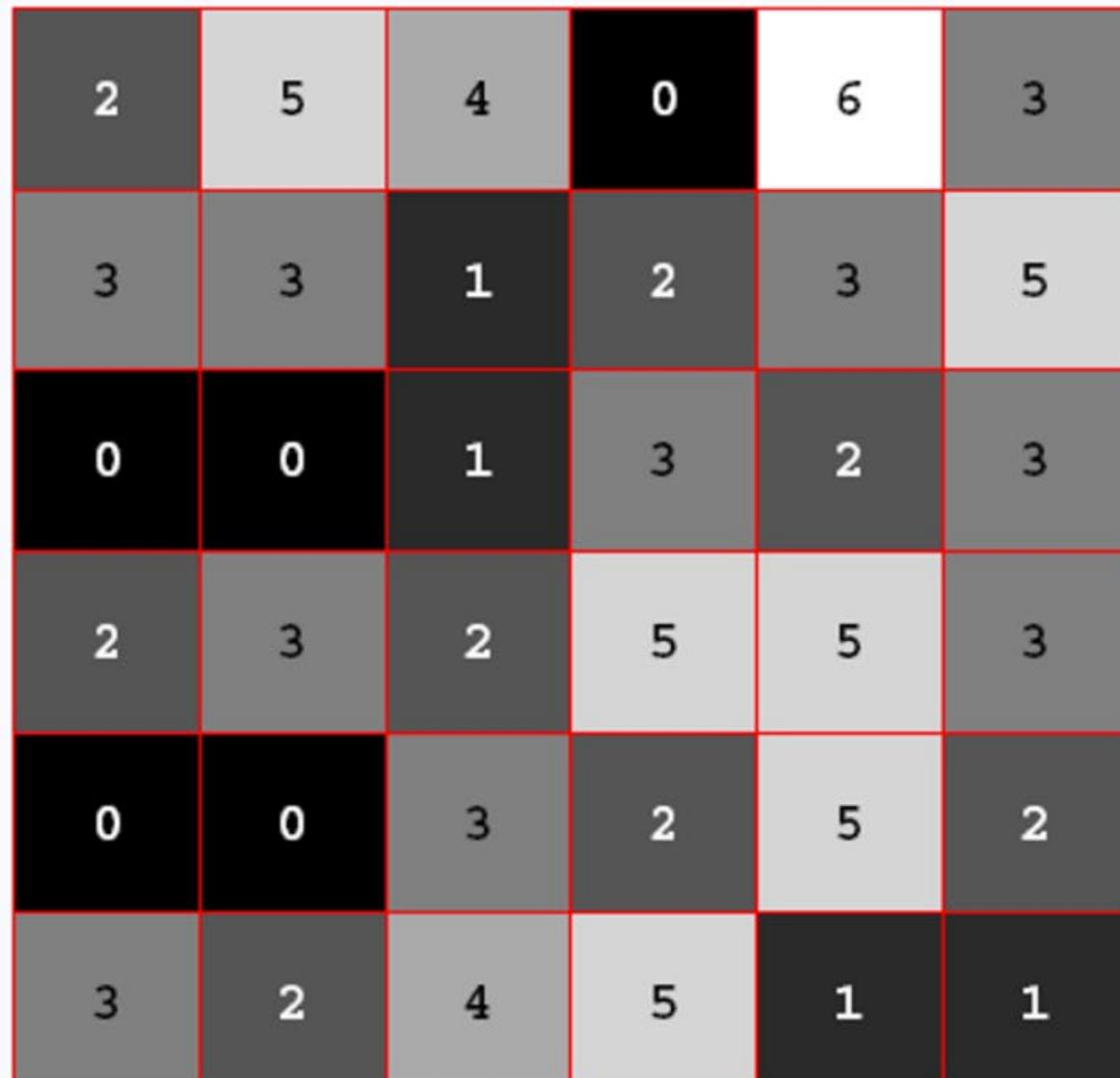
# Pixel value statistics

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



- Pick a random pixel in the image
- What is the probability of it having value 3?  $P(v=3)$
- $h(3) = 7$
- $N_p = 36$
- $P(v=3) = 7/36 * 100\%$
- The histogram divided by the total pixel count can be seen as a probability density function

A random pixel is chosen in the image. What is the probability that the value of the pixel is 3?



6%

28%

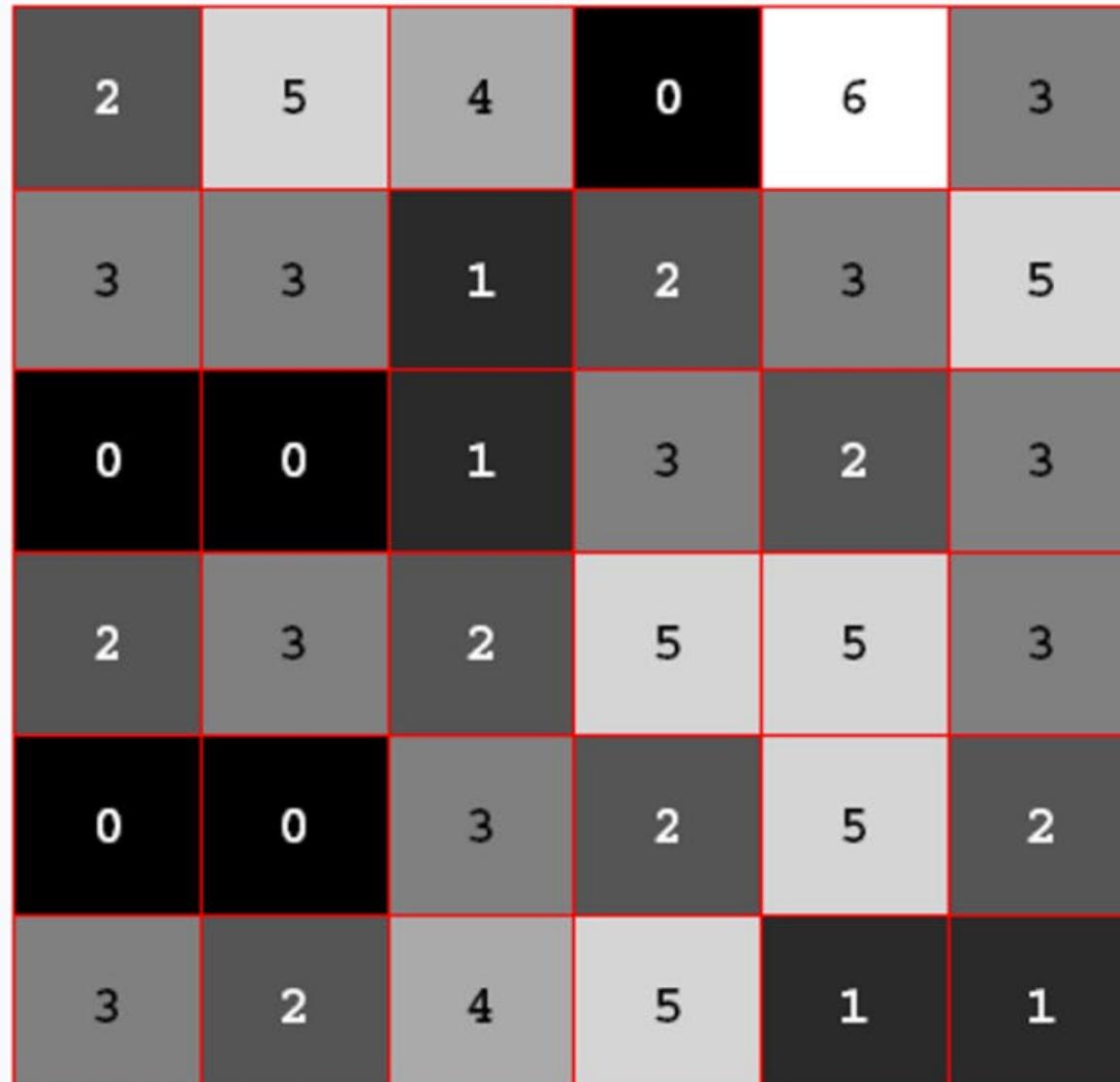
39%

51%

72%

I do not know

A random pixel is chosen in the image. What is the probability that the value of the pixel is 3?



6%

0%

28%

79%

39%

21%

51%

0%

72%

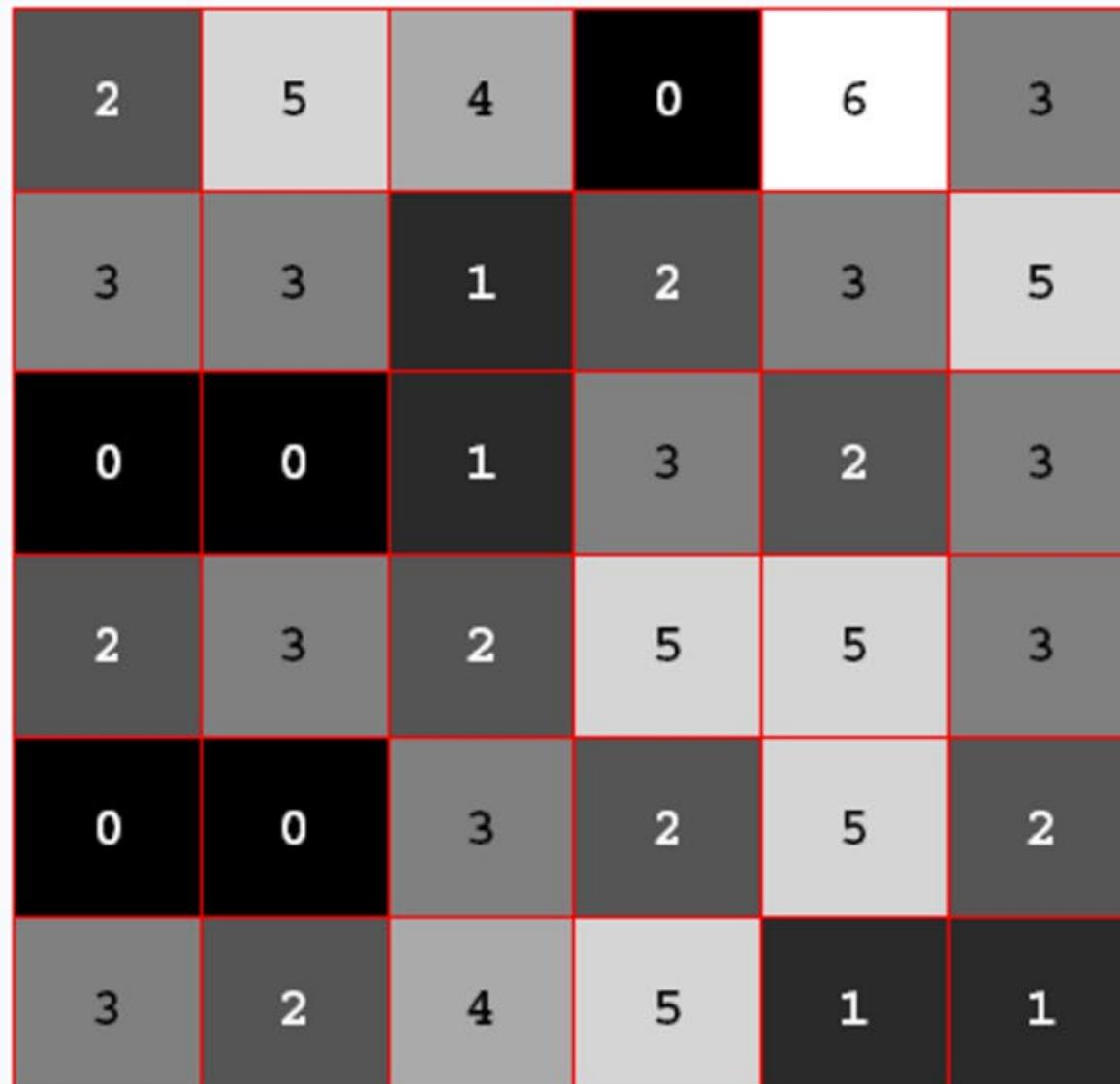
0%

I do not know

0%

Start the presentation to see live content. For screen share software, share the entire screen. Get help at [pollev.com/app](https://pollev.com/app)

A random pixel is chosen in the image. What is the probability that the value of the pixel is 3?



6%

0%

28%

79%

39%

21%

51%

0%

72%

0%

I do not know

0%

Start the presentation to see live content. For screen share software, share the entire screen. Get help at [pollev.com/app](https://pollev.com/app)

# Other Image Types

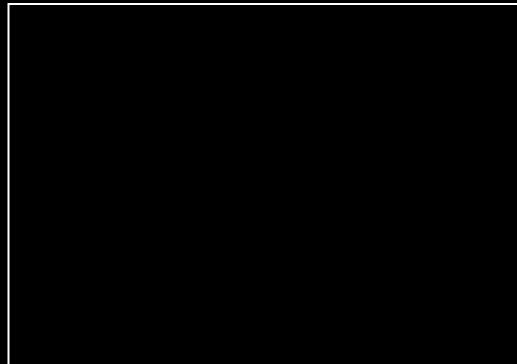
- Colour images
- Binary Images
- Label Images
- 16-bit images
- Floating point images

# Colour images



- RGB = Red, Green, and Blue
- Television, computers, digital cameras use the “RGB color space”
- Additive colours: Final colour is made by mixing red, green, and blue
- Typically the values of R, G, and B lie between 0 and 255 (total 3 bytes)!

# RGB Colours

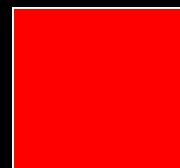


RGB = (0,0,0)

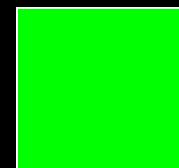


RGB = (255,255,255)

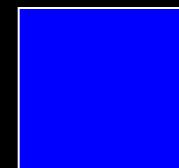
- When alle three "Lamps" are turned of we get black
- When all three "lamps" are on what do we get?



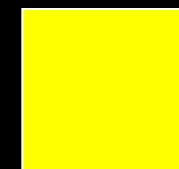
(255,0,0)



(0,255,0)



(0,0,255)



(255,255,0)

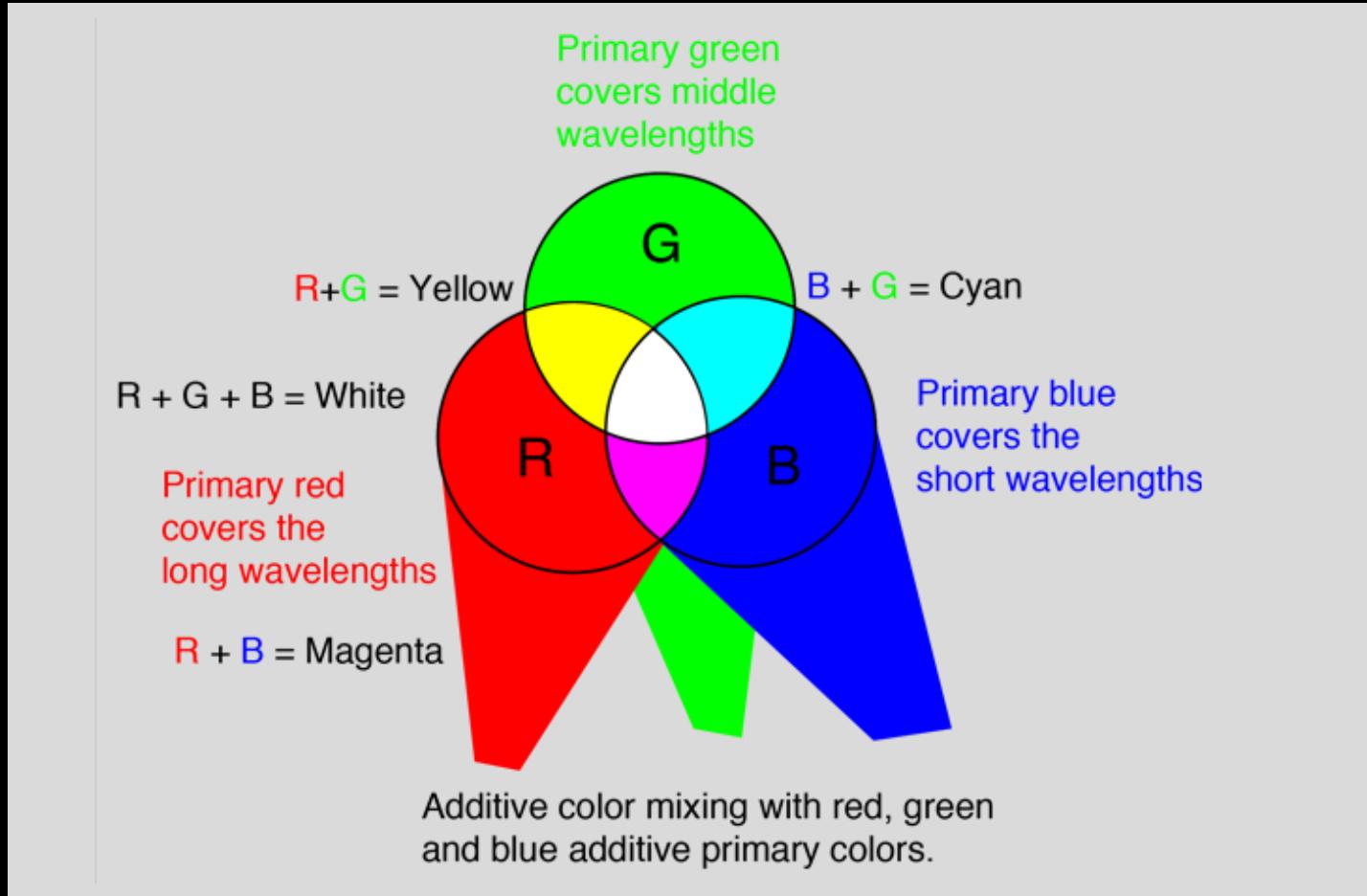


(0,255,255)



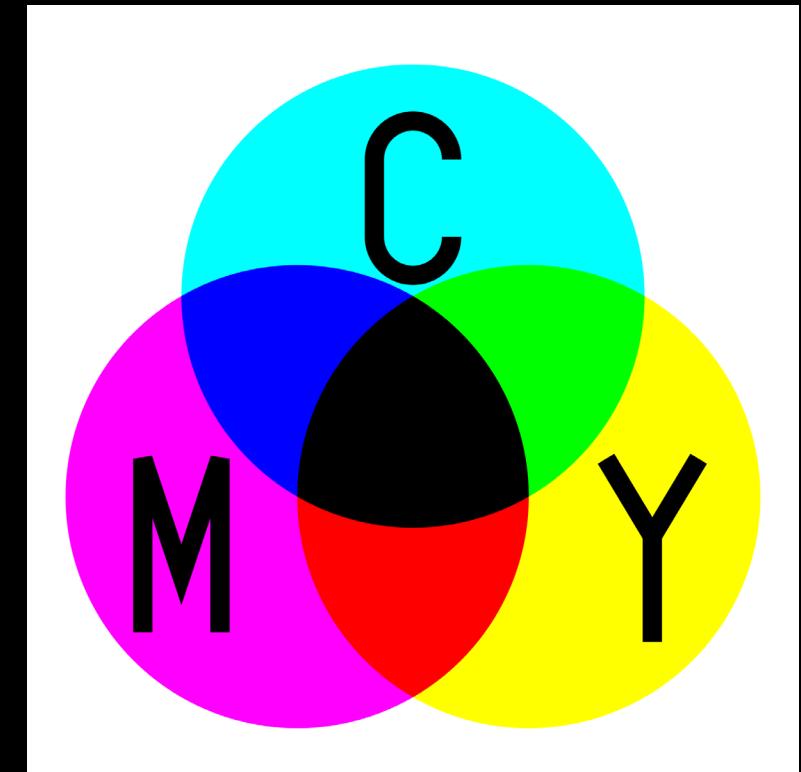
(255,0,255)

# Additive color mixing



<http://hyperphysics.phy-astr.gsu.edu/hbase/vision/addcol.html>

# Subtractive color mixing



Wikipedia

# Processing RGB images

- Each pixel in a colour image contains 3 values
- Equal to a “vector function” in mathematics
- More complicated to analyse
- Medical images are typically grayscale
  - Why?
- Often images are converted from colours to grayscale before the analysis

# Converting colour to grayscale

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



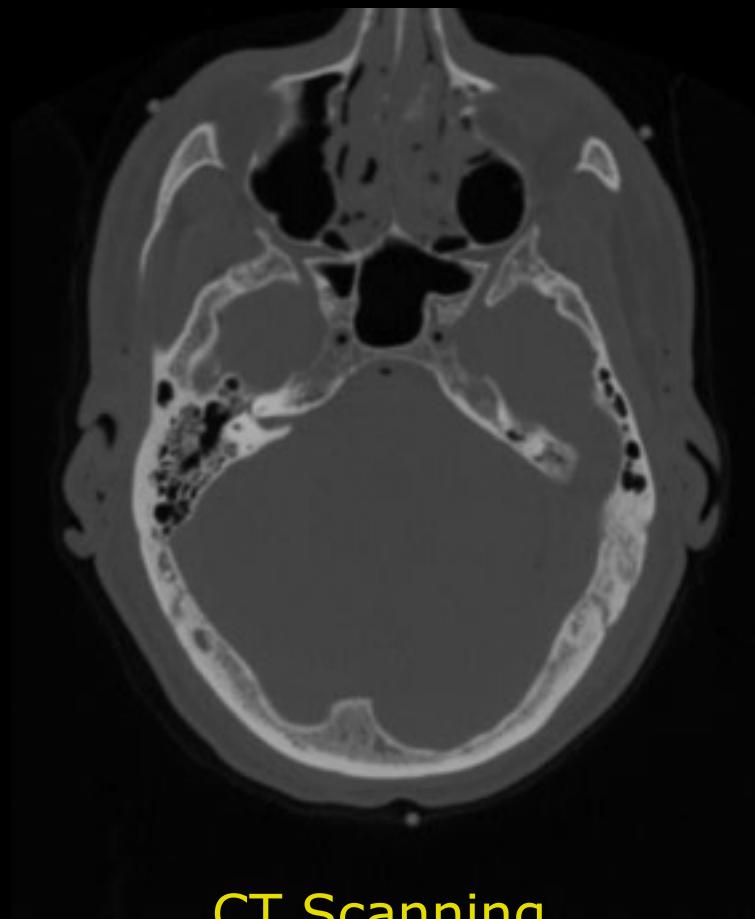
Is it possible to convert a grayscale image back to a color image?

# Binary images



- Binary – means on or off
- Binary image – only two colors
- Background (0 = black)
- Foreground (1 = white)
  
- Simple representation of CT scanning of the head

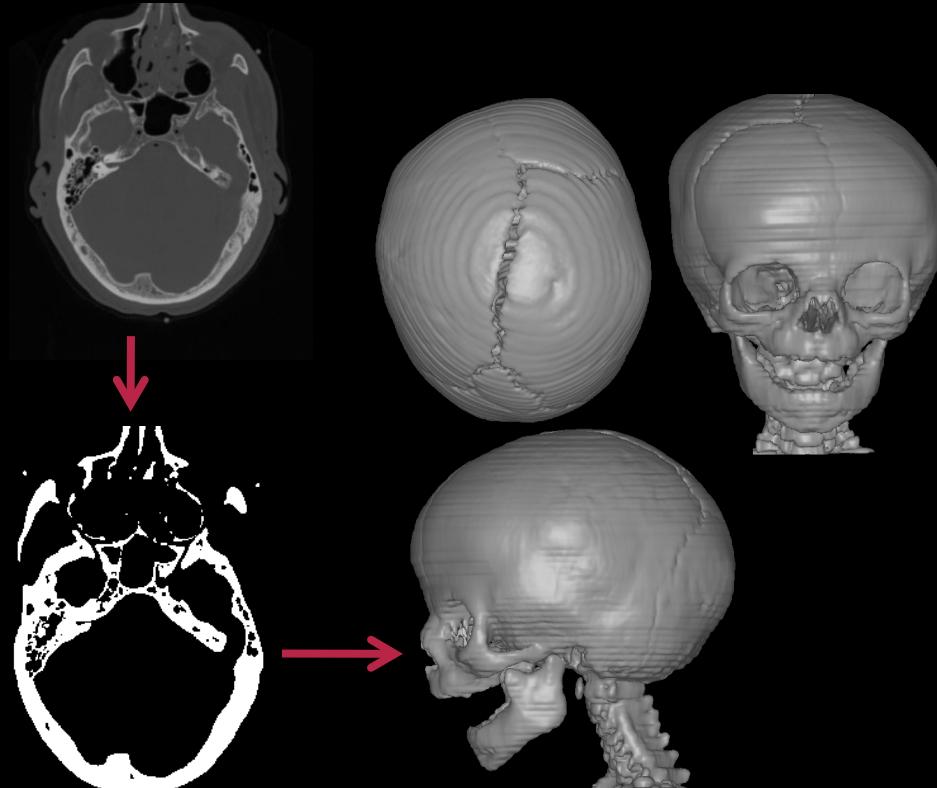
# Gray scale to Binary Image



Threshold



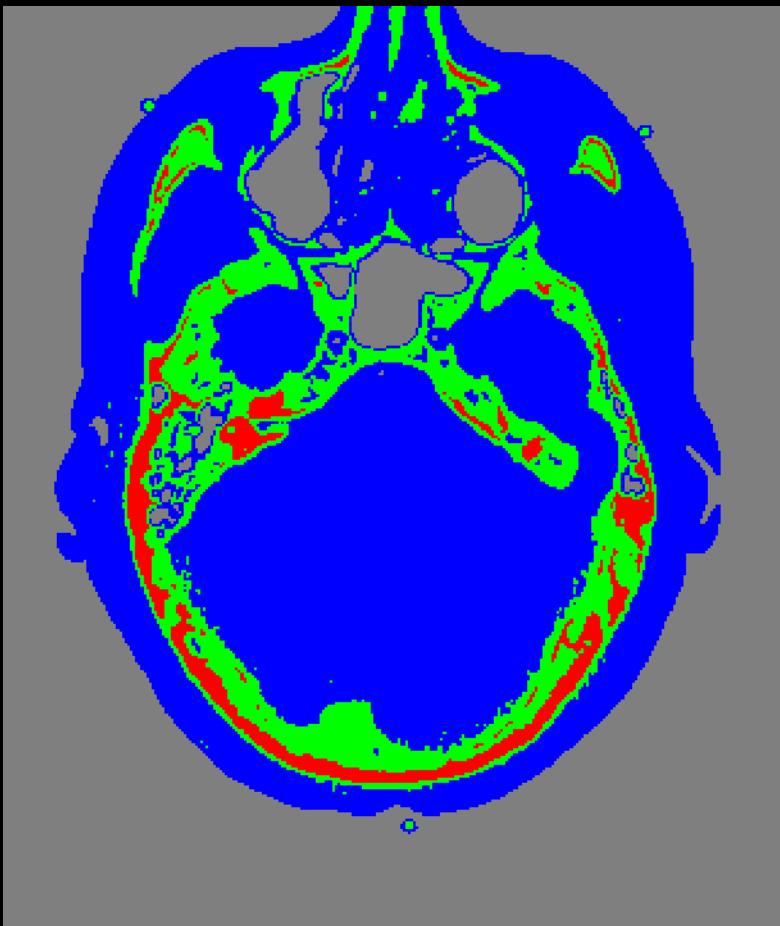
# Binary image – why?



- Separating objects from background
- Count the number of the objects
- Measure the size and shape of objects
- Advanced 3D visualisations

Image from 3D laboratory

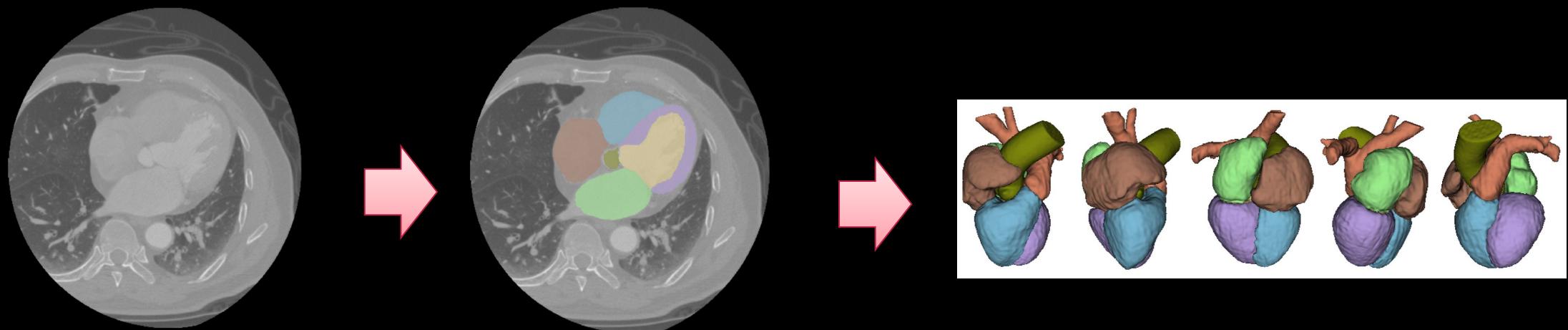
# Label Image



- The pixel value tells the *type* of the pixel
  - (0) Gray – background
  - (1) Blue – soft tissue
  - (2) Green – hard bone
  - (3) Red – spongy bone
- Only 4 different pixel values
- Colours made using a *look-up-table*

# Label Image – why?

- Segment images into regions
- **Example:** Recognize the major structures of the human heart as seen in a computed tomography image. Construct a 3D model of a given patient heart. Use the 3D model for diagnostics and surgery planning.



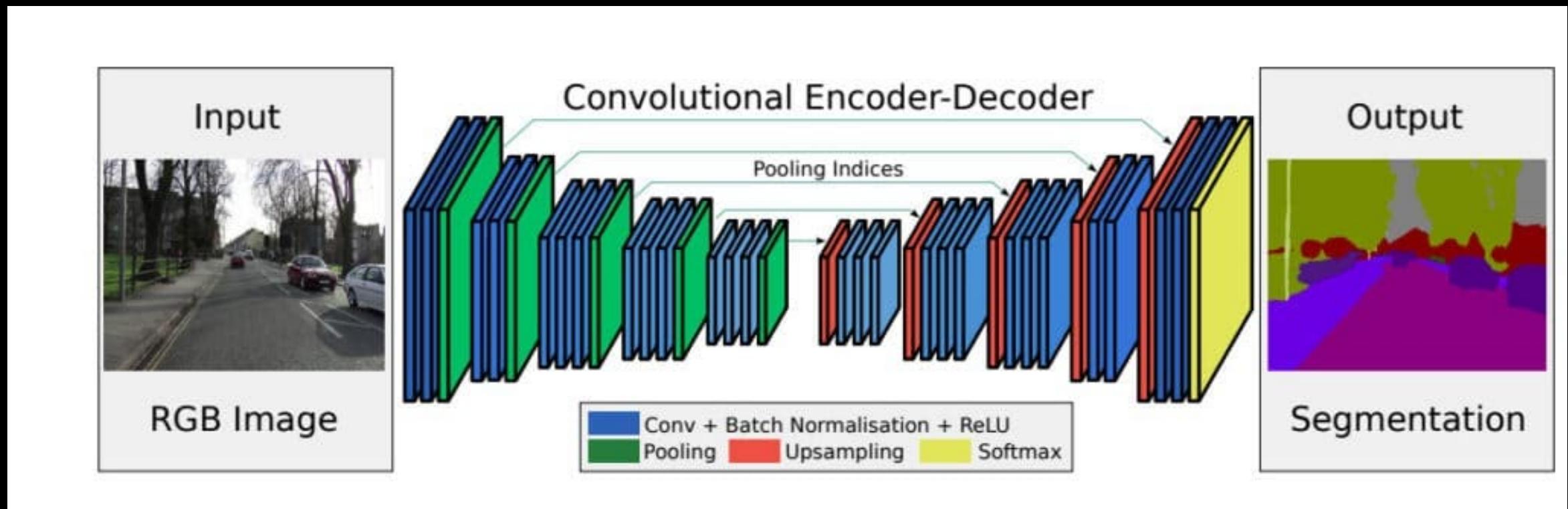
# Label image from semantic segmentation

- Scene understanding for self navigating vehicles



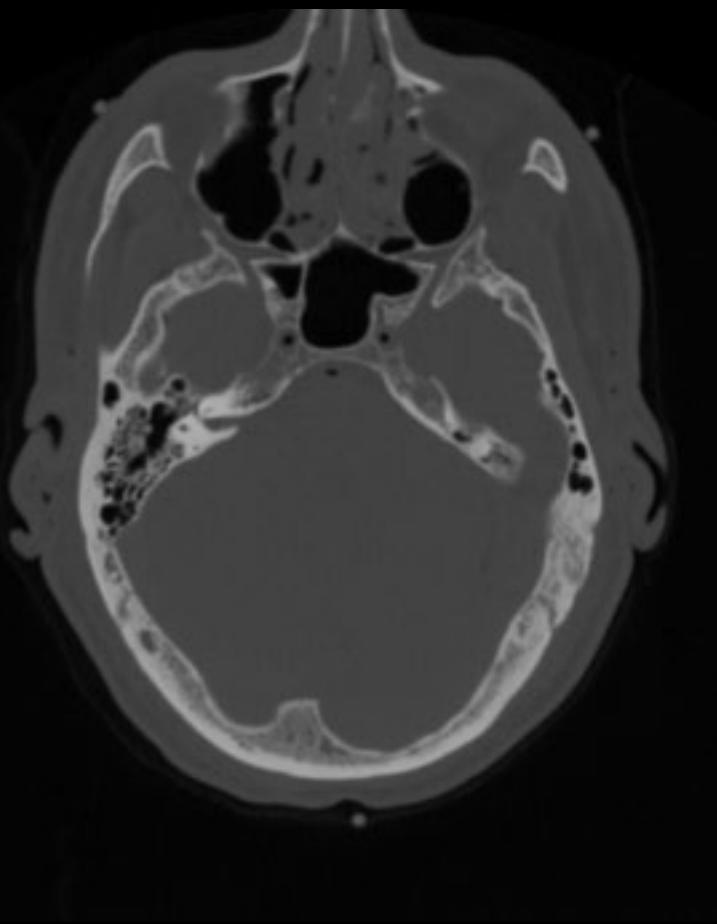
<https://towardsdatascience.com/semantic-segmentation-of-150-classes-of-objects-with-5-lines-of-code-7f244fa96b6c>

# Deep learning for semantic segmentation



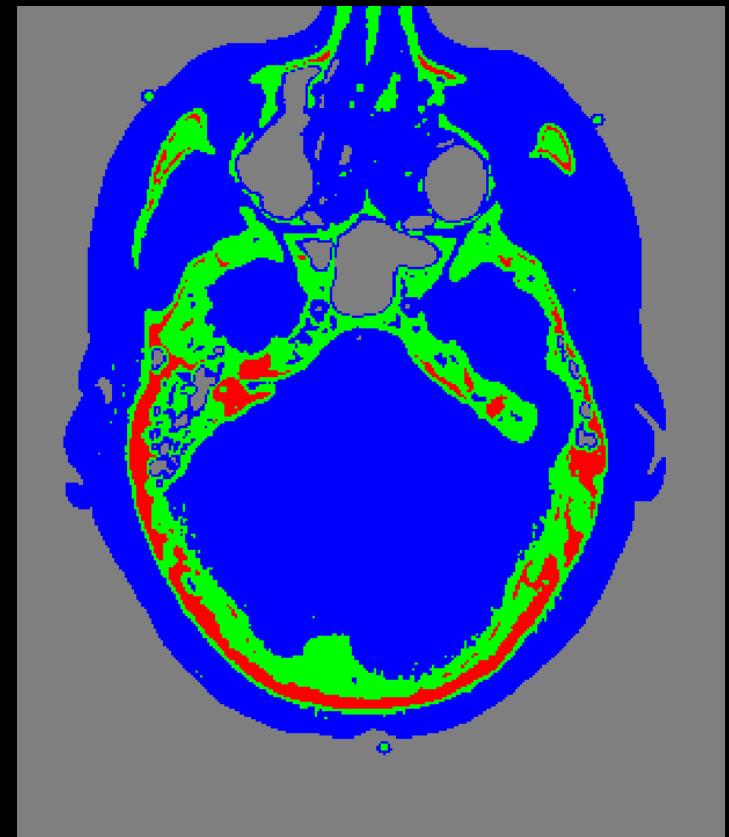
Badrinarayanan, Vijay, Alex Kendall, and Roberto Cipolla. "Segnet: A deep convolutional encoder-decoder architecture for image segmentation." *IEEE transactions on pattern analysis and machine intelligence* 39.12 (2017): 2481-2495.

# Label images in this course



Pixel Classification

BLOB analysis and classification



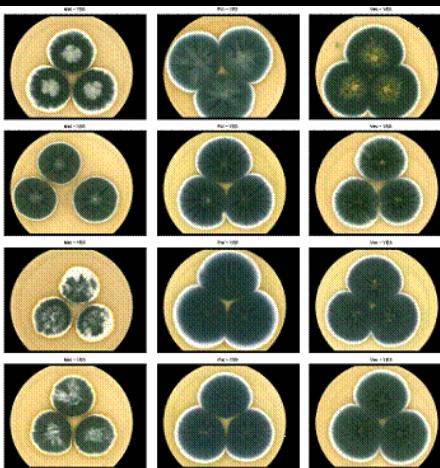
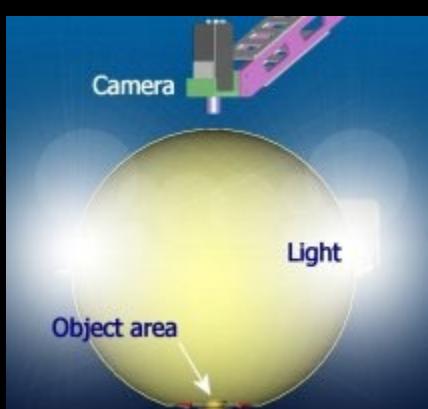
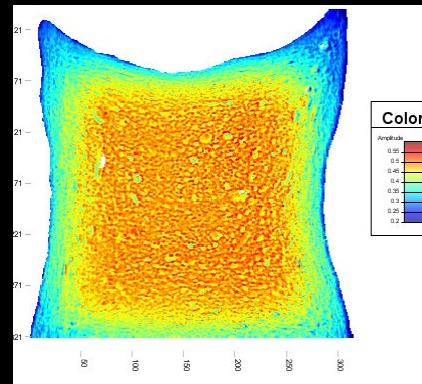
# Multispectral images



Infrared

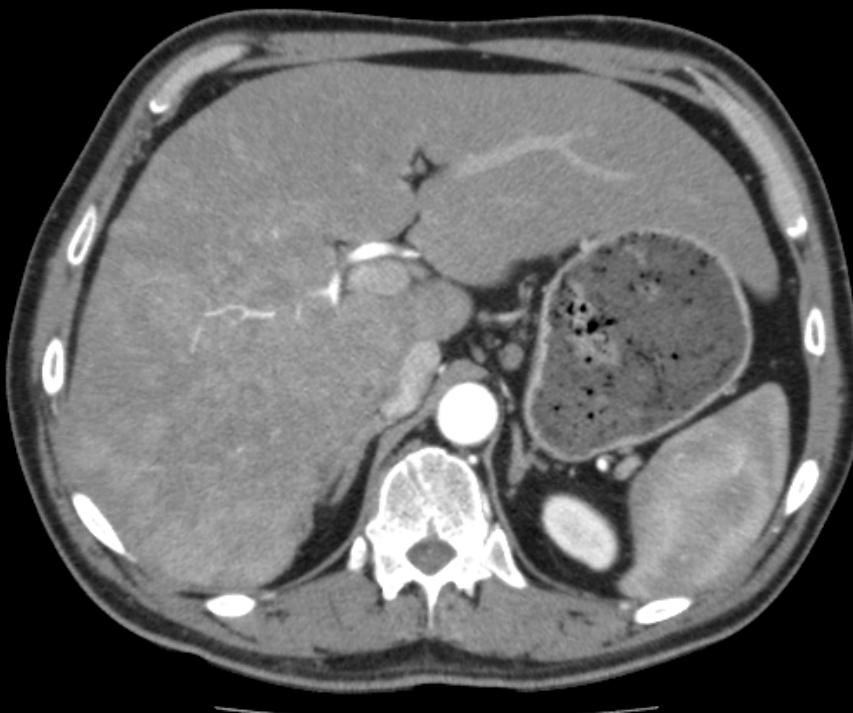
- There are more visual information than what can be seen with the human eye
- Standard cameras captures the red, green, blue colours
- Capture systems that capture more bands and other frequencies exist
- Creates multispectral images
  - Each pixel contains perhaps 20 values from different spectral bands

# Multispectral System - VideometerLab



- Integrating sphere
- Light emitting diodes with different wavelengths
  - From near infrared to ultraviolet
- High resolution camera
- Water in bread
- Classification of fungi
- Skin diseases

# 16-bit images



- 256 values fine for the human eye
- Pixel values not only for display
  - Physical meaning
- Computed Tomography
  - X-ray attenuation
- Hounsfield units
  - 0 water
  - -1000 air
  - -120 fat
  - 400+ bone

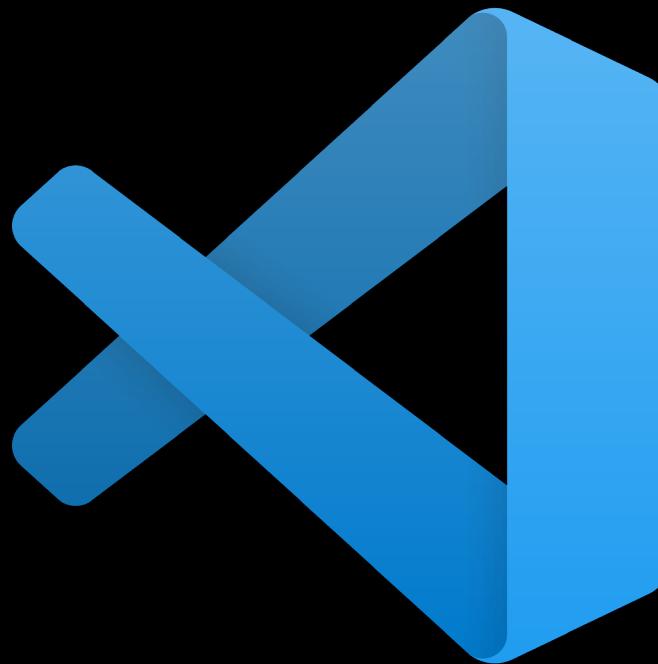
# Floating point images

- The pixel type is often changed when applying image processing functions
- For example when scaling an image, the output will be a floating point image:

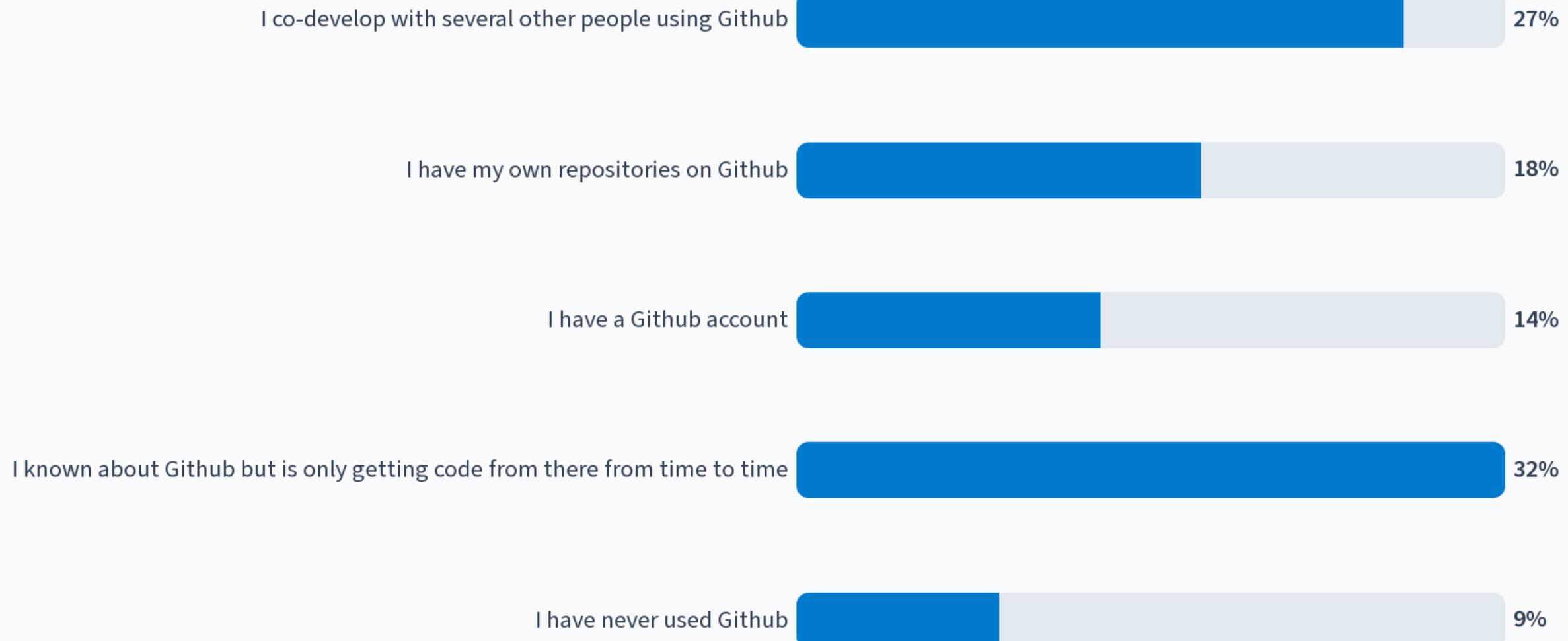
```
image_rescaled = rescale(im_org, 0.25, anti_aliasing=True, channel_axis=2)
print(image_rescaled.shape)
print(image_rescaled.dtype)
```

# Python scripts vs. Jupyter Notebooks

- In this course, you can do the exercises and the exam in both Jupyter Notebooks or as Python scripts
- Strengths and benefits of both approaches



## What is your experience with Github?



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# PCA Analysis

# Next week:

## Image acquisition, digital cameras, compression and storage and real-time image analysis

