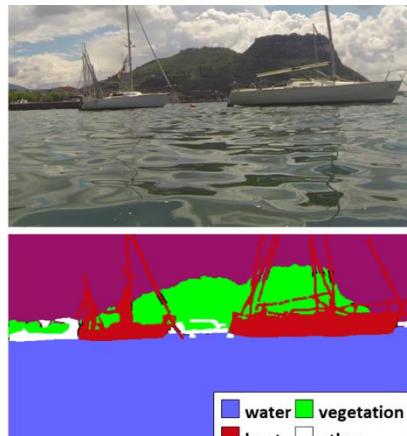
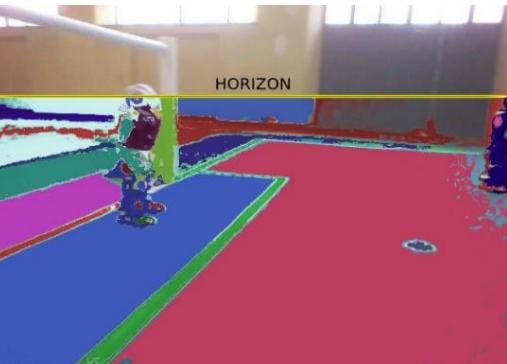




**UNIVERSITÀ DEGLI STUDI
DELLA BASILICATA**

Corso di Visione e Percezione

Trasformazioni



Docente
Domenico D. Bloisi

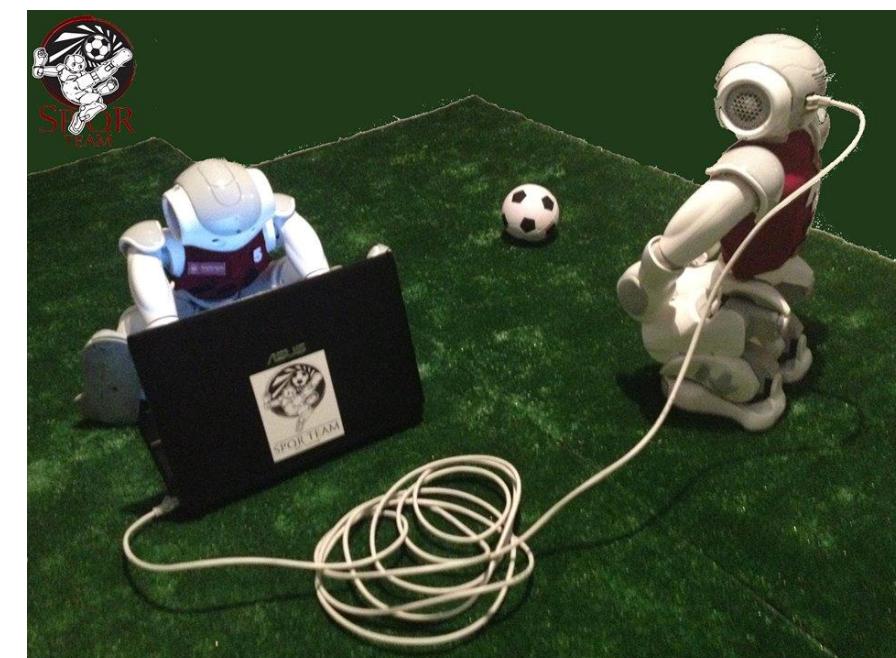
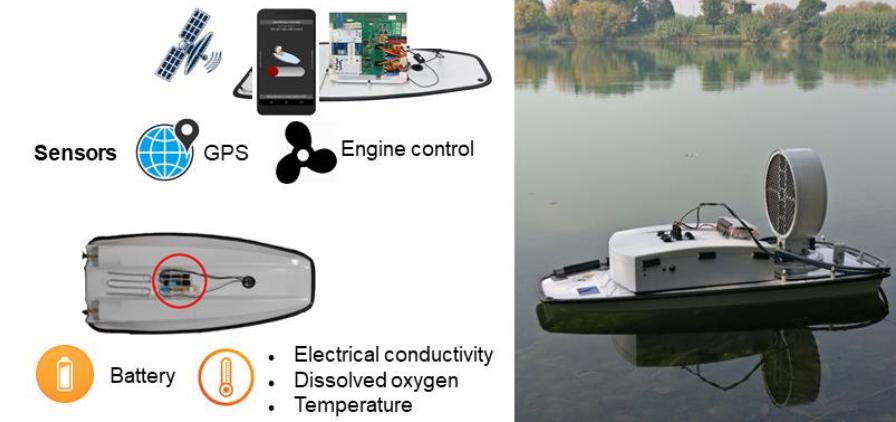
Domenico Daniele Bloisi

- Professore Associato
Dipartimento di Matematica, Informatica
ed Economia
Università degli studi della Basilicata

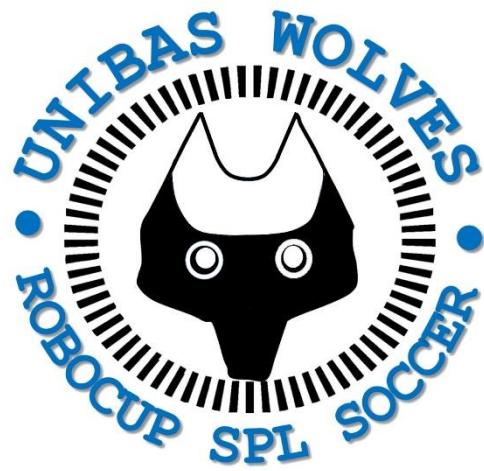
<http://web.unibas.it/bloisi>

- SPQR Robot Soccer Team
Dipartimento di Informatica, Automatica
e Gestionale Università degli studi di
Roma “La Sapienza”

<http://spqr.diag.uniroma1.it>



UNIBAS Wolves <https://sites.google.com/unibas.it/wolves>



- UNIBAS WOLVES is the robot soccer team of the University of Basilicata. Established in 2019, it is focussed on developing software for NAO soccer robots participating in RoboCup competitions.
- UNIBAS WOLVES team is twinned with [SPQR Team](#) at Sapienza University of Rome.



Informazioni sul corso

- Home page del corso:
<https://web.unibas.it/bloisi/corsi/visione-e-percezione.html>
- Docente: Domenico Daniele Bloisi
- Periodo: Il semestre marzo 2022 – giugno 2022
 - Martedì dalle 15:00 alle 17:00 (Aula Copernico)
 - Mercoledì dalle 8:30 alle 10:30 (Aula Copernico)

Ricevimento

- Durante il periodo delle lezioni:

Mercoledì dalle 11:00 alle 12:30 → Edificio 3D, II piano, stanza 15

Si invitano gli studenti a controllare regolarmente la bacheca degli avvisi per eventuali variazioni

- Al di fuori del periodo delle lezioni:

da concordare con il docente tramite email

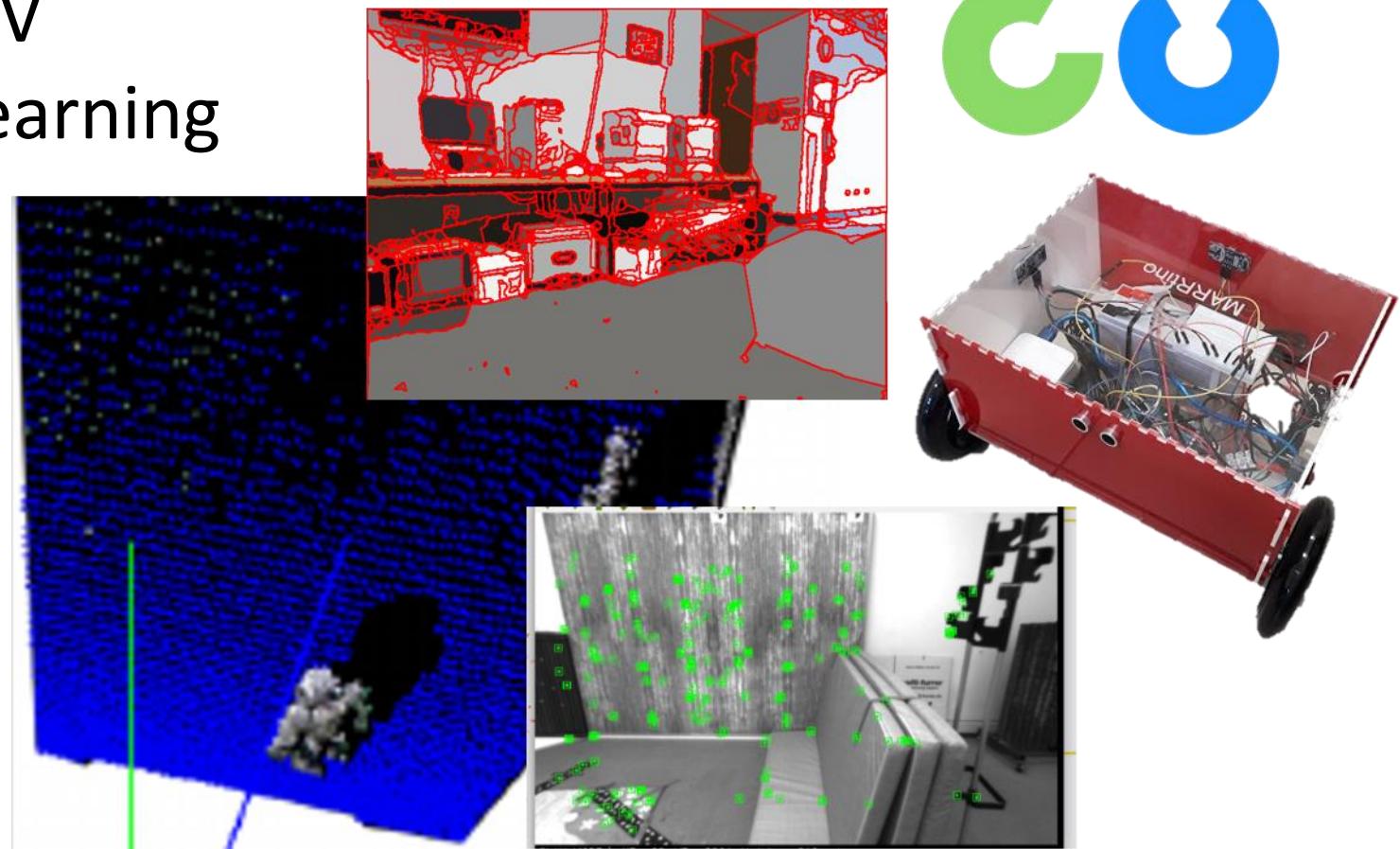
Per prenotare un appuntamento inviare
una email a

domenico.bloisi@unibas.it



Programma – Visione e Percezione

- Introduzione al linguaggio Python
- [Elaborazione delle immagini con Python](#)
- Percezione 2D – OpenCV
- Introduzione al Deep Learning
- ROS
- Il paradigma publisher and subscriber
- Simulatori
- Percezione 3D - PCL



Grayscale image



Immagini a colori

- Simplified object extraction and identification
- Human vision: ~10 million of distinguishable colors
- Digital RGB representation: $256 \times 256 \times 256$ colors per pixel → 16 million possible colors

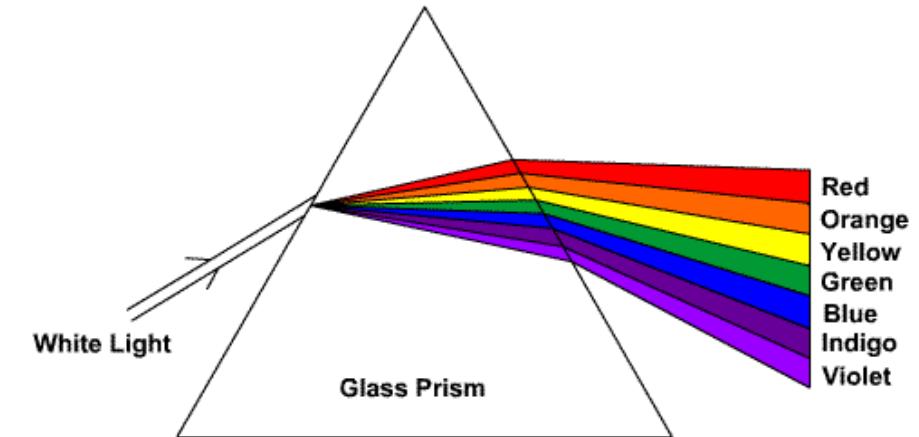
Color image



Color spectrum

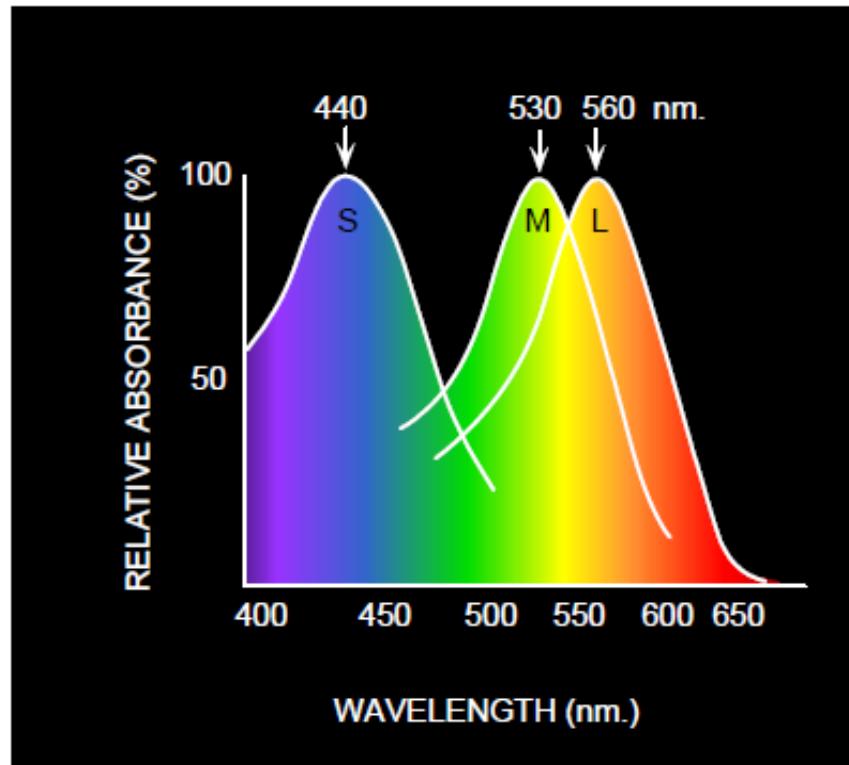


White light with a prism (1666, Newton)



Percezione dei colori

- Approximately 6 million cones in the human eye
- Three different types of cones
- Each cone has a special pigment making it sensitive to specific ranges of wavelengths:
 - Short (S) corresponds to blue
 - Medium (M) corresponds to green
 - Long (L) corresponds to red



Colori primari

- Color representation is based on the theory of T. Young (1802) which states that any color can be produced by mixing three primary colors C_1, C_2, C_3 :
$$C = aC_1 + bC_2 + cC_3$$
- It is therefore possible to characterize a psycho-visual color by specifying the amounts of three primary colors: red, green, and blue, mixed together
- This leads to the standard RGB space used in television, computer monitors, LED screens, etc

Esercizio 1

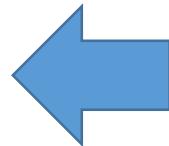
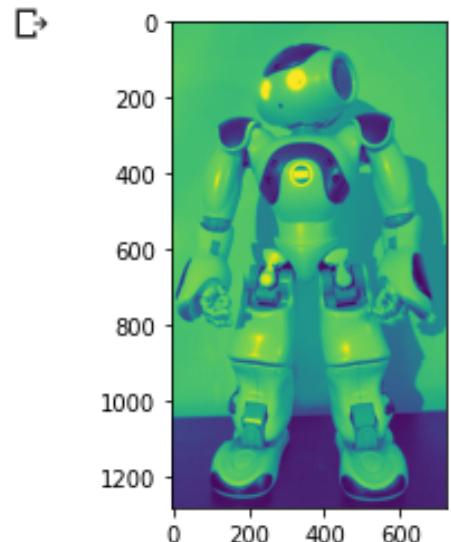
Aprire l'immagine a colori

<https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg>

e trasformarla in grayscale

Esercizio 1 - soluzione

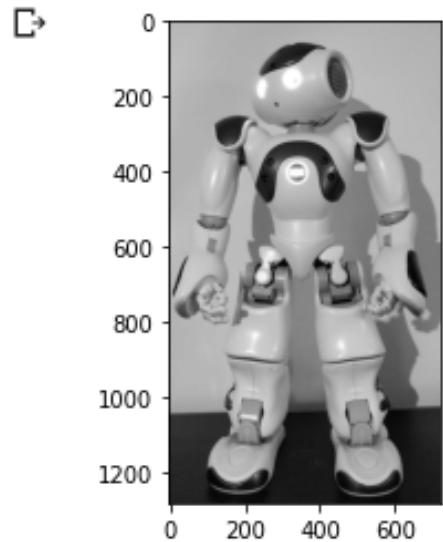
```
from PIL import Image  
from urllib.request import urlopen  
import matplotlib.pyplot as plt  
  
url = "https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg"  
  
img = Image.open(urlopen(url))  
  
gray_img = img.convert("L")  
  
_ = plt.imshow(gray_img)
```



Questa visualizzazione non
sembra corretta!

Esercizio 1 - soluzione

```
▶ from PIL import Image  
from urllib.request import urlopen  
import matplotlib.pyplot as plt  
  
url = "https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg"  
  
img = Image.open(urlopen(url))  
  
gray_img = img.convert("L")  
  
_ = plt.imshow(gray_img, cmap="gray")
```



Questa è la visualizzazione corretta!

Esercizio 2

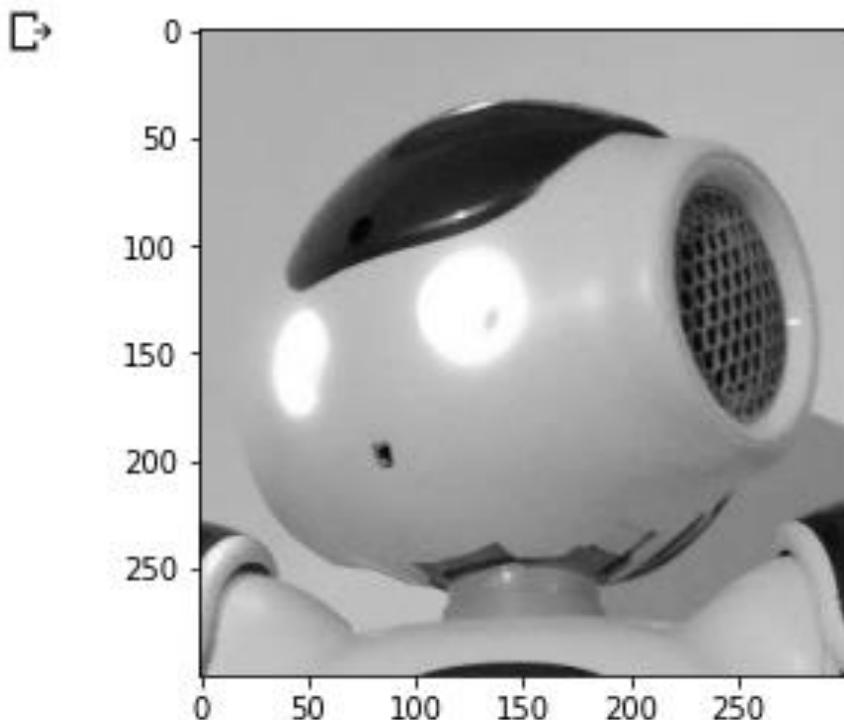
Costruire una immagine 300x300 contenente solo la testa del robot a partire dalla versione grayscale dell'immagine

<https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg>

ottenuta nell'esercizio precedente

Esercizio 2 - soluzione

```
ROI = (200,25,500,325) #left, upper, right, and lower pixel coordinate  
  
face = gray_img.crop(ROI)  
  
_ = plt.imshow(face, cmap="gray")
```



Istogramma di una immagine

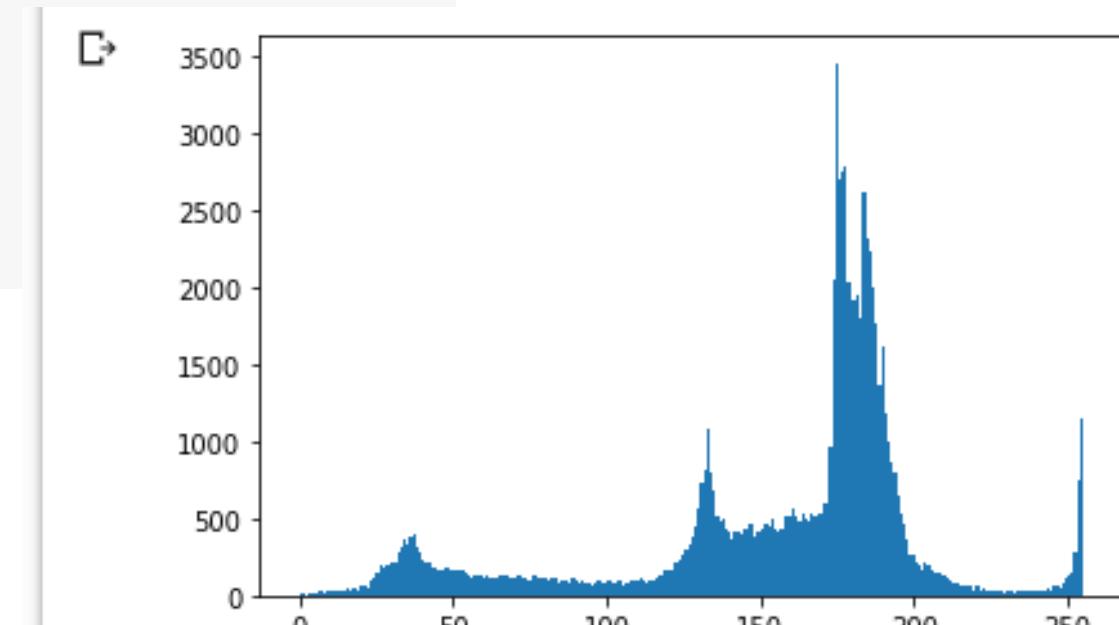


```
from PIL import Image
from urllib.request import urlopen
import matplotlib.pyplot as plt
import numpy as np

url = "https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg"
gray_img = Image.open(urlopen(url)).convert("L")
face = np.array(gray_img.crop((200,25,500,325)))

plt.hist(face.flatten(), 256)
plt.show()
```

ndarray.flatten returns a copy of the array collapsed into one dimension.



Scipy library

Scipy library
(<https://www.scipy.org/scipylib>)
è una libreria contenente
l'implementazione di algoritmi e
tool matematici compatibili con
NumPy

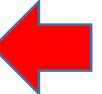


Scipy sub-modules

La Scipy library contiene diversi sub-moduli specializzati per particolari compiti

`scipy.ndimage` è il package per il processamento delle immagini

<https://docs.scipy.org/doc/scipy/reference/ndimage.html>

- Clustering package ([scipy.cluster](#))
- Constants ([scipy.constants](#))
- Discrete Fourier transforms ([scipy.fft](#))
- Legacy discrete Fourier transforms ([scipy.fftpack](#))
- Integration and ODEs ([scipy.integrate](#))
- Interpolation ([scipy.interpolate](#))
- Input and output ([scipy.io](#))
- Linear algebra ([scipy.linalg](#))
- Miscellaneous routines ([scipy.misc](#))
- Multi-dimensional image processing ([scipy.ndimage](#)) 
- Orthogonal distance regression ([scipy.odr](#))
- Optimization and Root Finding ([scipy.optimize](#))
- Signal processing ([scipy.signal](#))
- Sparse matrices ([scipy.sparse](#))
- Sparse linear algebra ([scipy.sparse.linalg](#))
- Compressed Sparse Graph Routines ([scipy.sparse.csgraph](#))
- Spatial algorithms and data structures ([scipy.spatial](#))
- Special functions ([scipy.special](#))
- Statistical functions ([scipy.stats](#))
- Statistical functions for masked arrays ([scipy.stats.mstats](#))
- Low-level callback functions

Trasformazioni geometriche in Scipy

- Shift
- Rotazione
- Zoom
- Flip

Shift

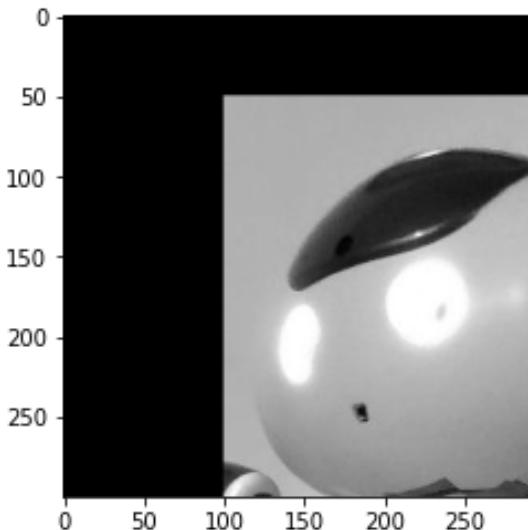
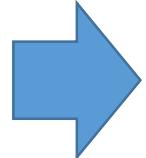
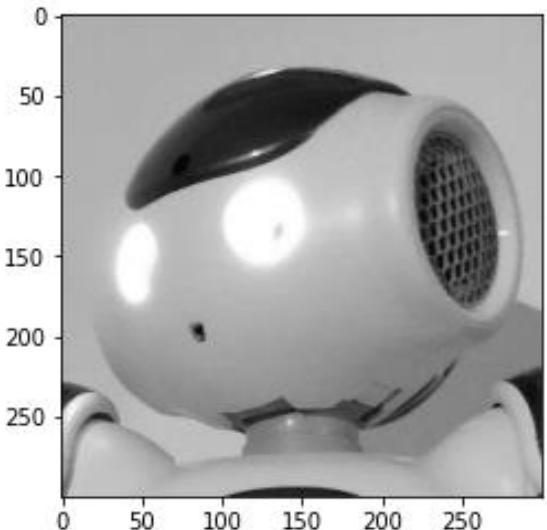


```
from PIL import Image
from urllib.request import urlopen
import matplotlib.pyplot as plt
from scipy import ndimage

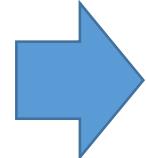
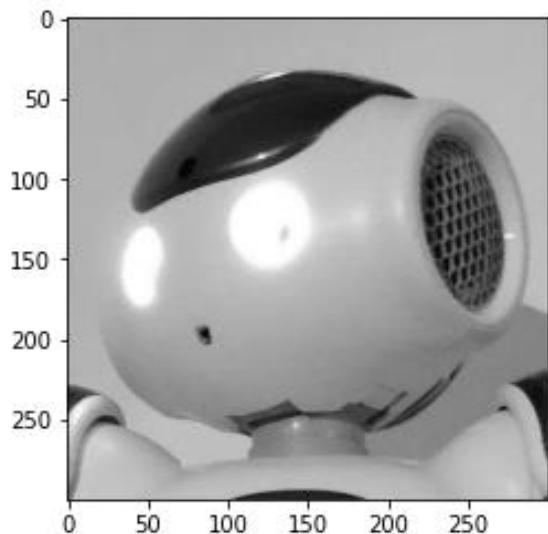
url = "https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg"
gray_img = Image.open(urlopen(url)).convert("L")
face = gray_img.crop((200,25,500,325))

shifted_face = ndimage.shift(face, (50, 100))

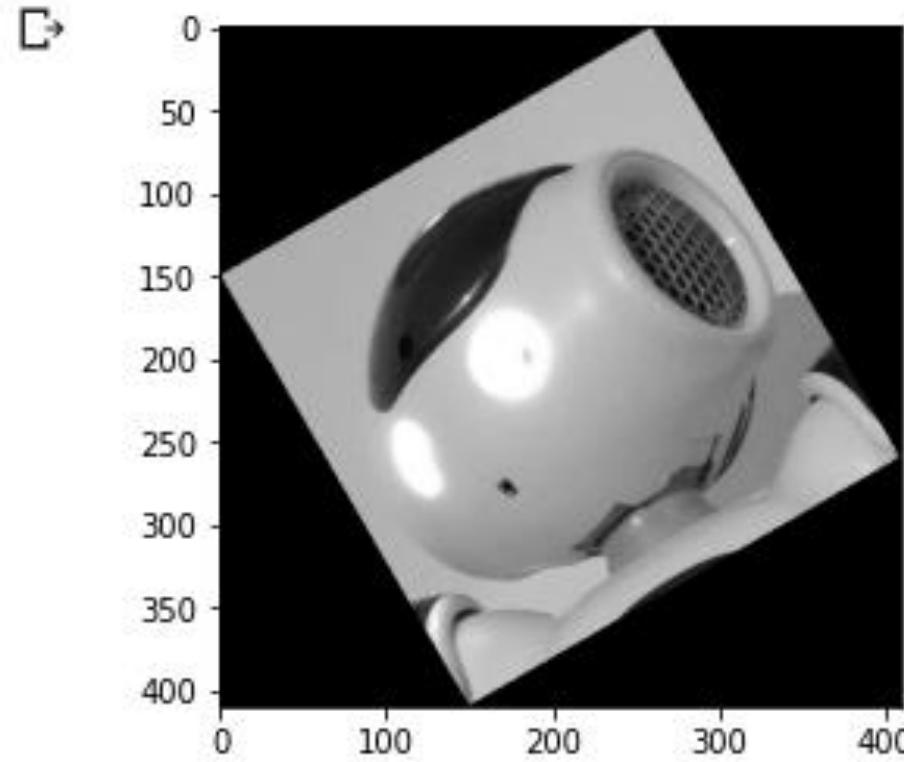
_ = plt.imshow(shifted_face, cmap="gray")
```



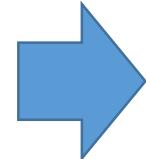
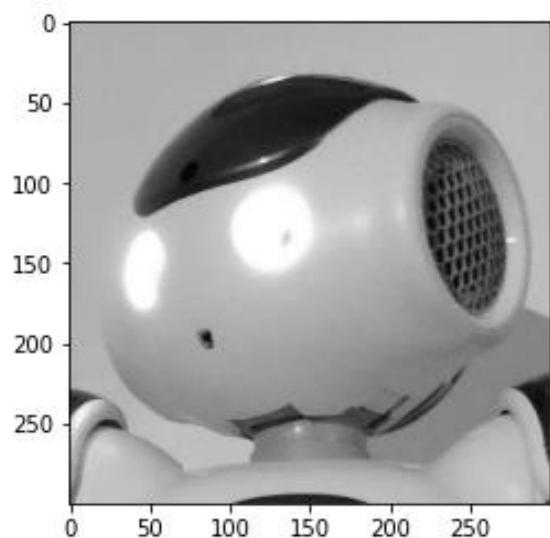
Rotazione



```
rotated_face = ndimage.rotate(face, 30)  
  
_ = plt.imshow(rotated_face, cmap="gray")
```



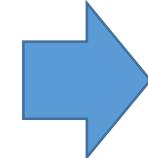
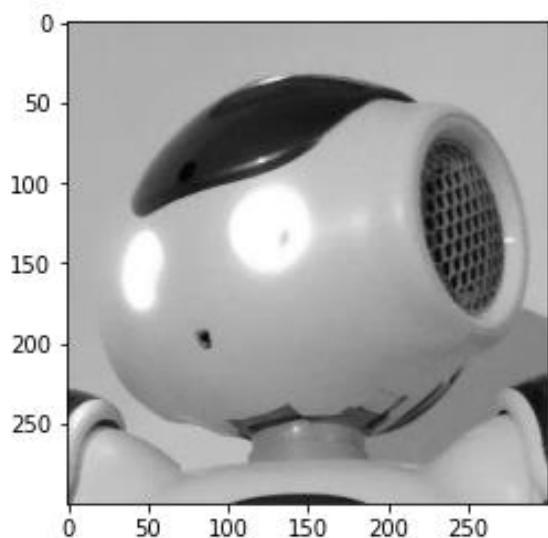
Zoom



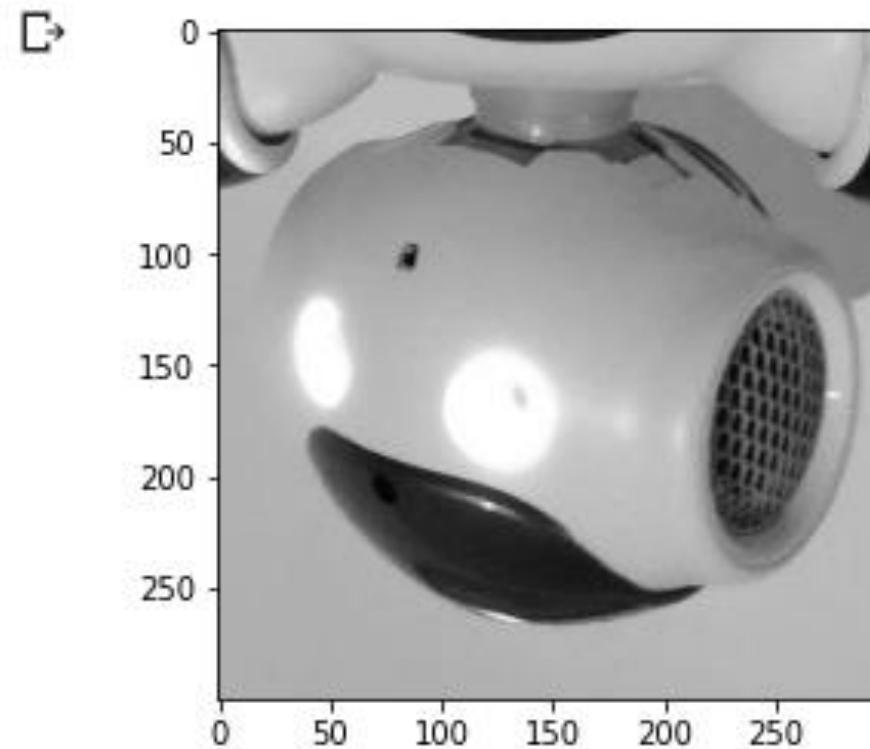
```
zoomed_face = ndimage.zoom(face, 2)  
  
_ = plt.imshow(zoomed_face, cmap="gray")
```

A zoomed-in grayscale image of the same robot head, showing a close-up view of the right side. The image is plotted on a coordinate system with both x and y axes ranging from 0 to 500, with major ticks at 0, 100, 200, 300, 400, and 500. The robot's features are more detailed due to the magnification.

Flip



```
# up <-> down flip  
flip_ud_face = np.flipud(face)  
  
_ = plt.imshow(flip_ud_face, cmap="gray")
```



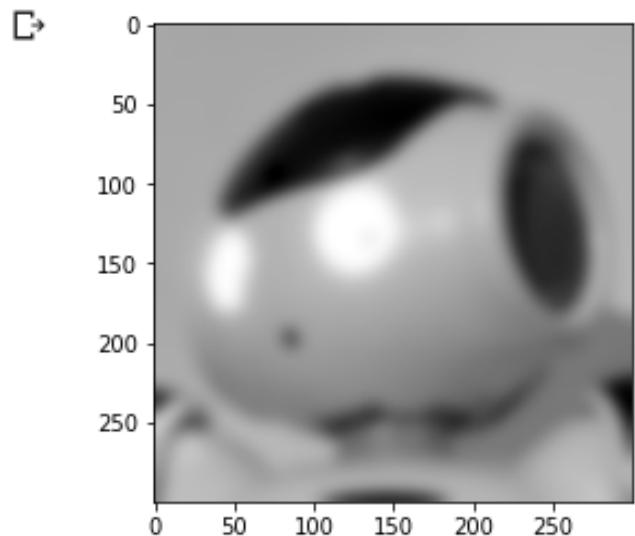
Filtering in Scipy

```
from PIL import Image
from urllib.request import urlopen
import matplotlib.pyplot as plt
from scipy.ndimage import filters

url = "https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg"
gray_img = Image.open(urlopen(url)).convert("L")
face = gray_img.crop((200,25,500,325))

blurred_face = filters.gaussian_filter(face,5)

_ = plt.imshow(blurred_face, cmap="gray")
```



Il secondo parametro di gaussian_filter() è la standard deviation

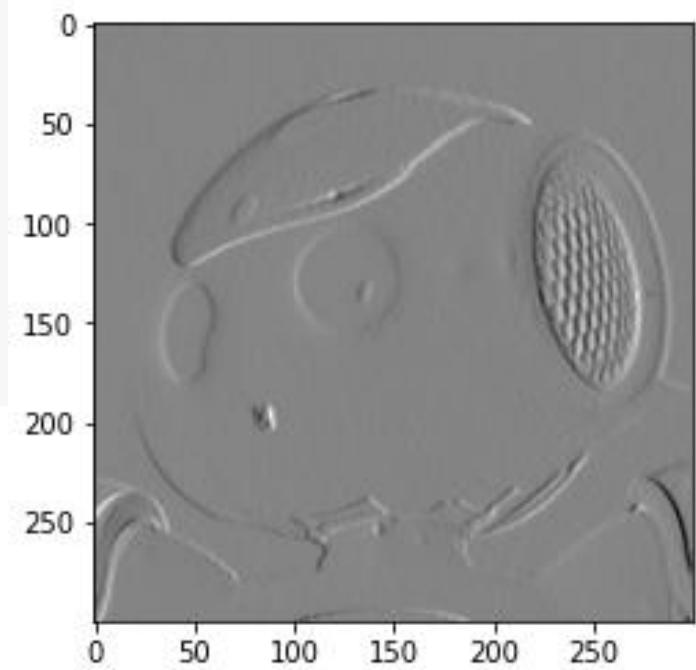
Derivate in Scipy

```
from PIL import Image
from urllib.request import urlopen
import matplotlib.pyplot as plt
from scipy.ndimage import filters
import numpy as np

url = "https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg"
gray_img = Image.open(urlopen(url)).convert("L")
face = np.array(gray_img.crop((200,25,500,325)))
```

Three blue arrows point from the code line `np.array(gray_img.crop((200,25,500,325)))` to the corresponding crop operation in the image on the right.

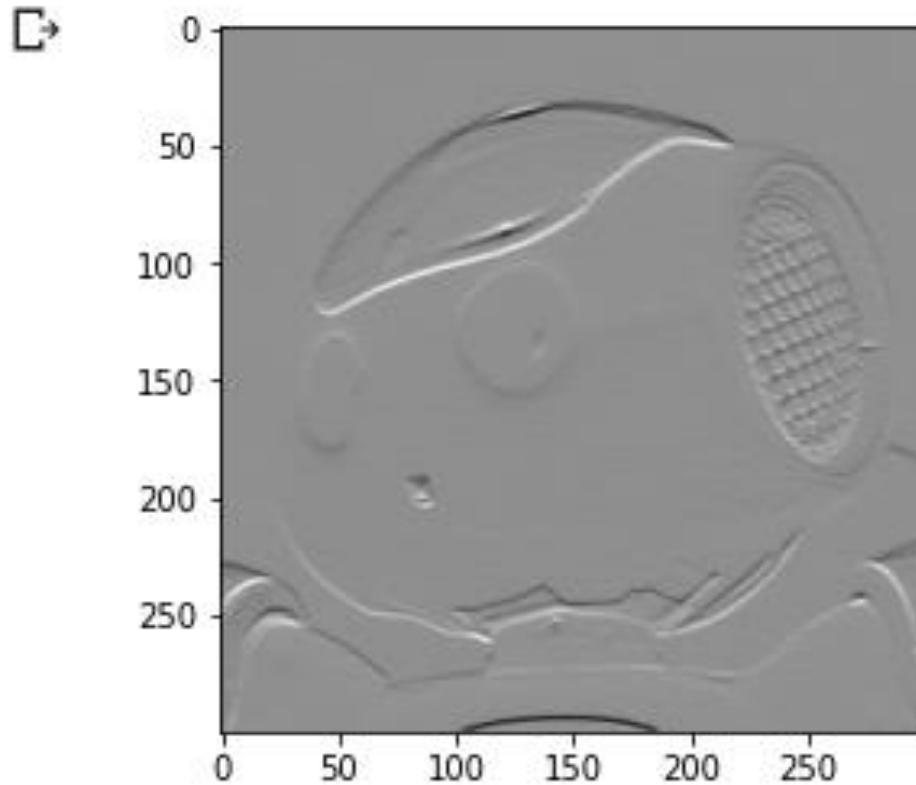
```
dx = np.zeros(face.shape)
filters.sobel(face,1,dx)
_ = plt.imshow(dx, cmap="gray")
```



Derivate in Scipy



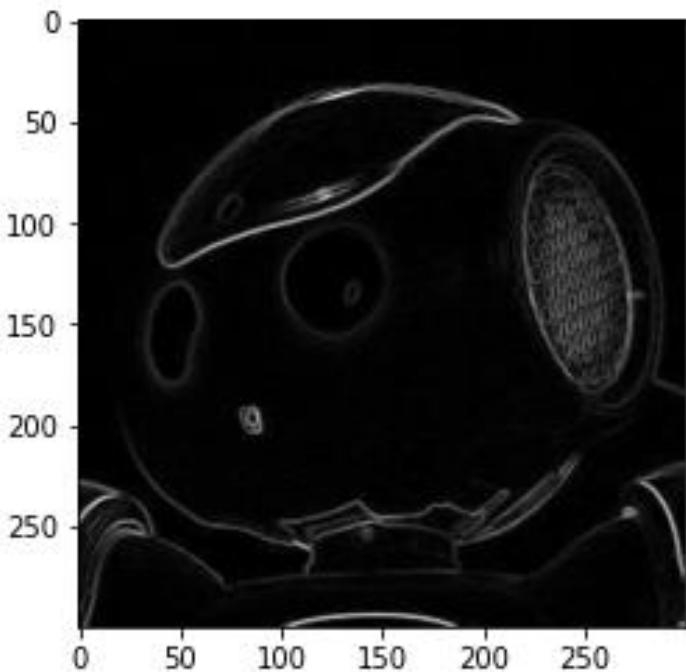
```
dy = np.zeros(face.shape)
filters.sobel(face, 0, dy)
_ = plt.imshow(dy, cmap="gray")
```



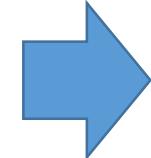
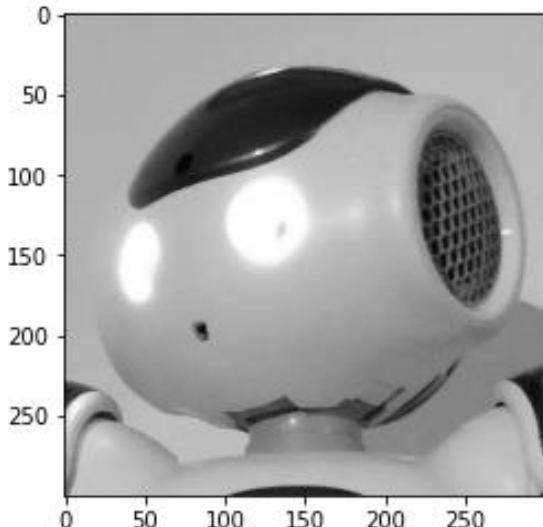
Gradient magnitude



```
from numpy import sqrt  
  
magnitude = sqrt(dx**2+dy**2)  
  
_ = plt.imshow(magnitude, cmap="gray")
```



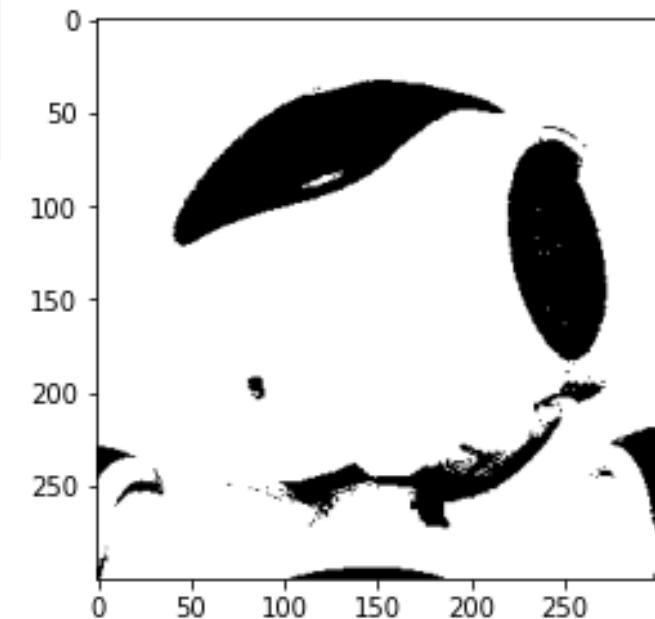
Thresholding



```
from PIL import Image
from urllib.request import urlopen
import matplotlib.pyplot as plt
import numpy as np

url = "https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg"
gray_img = Image.open(urlopen(url)).convert("L")
face = np.array(gray_img.crop((200,25,500,325)))

t = 120
mask = face > t
_ = plt.imshow(mask, cmap="gray")
```



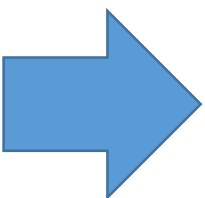
Otsu Thresholding

```
import matplotlib.pyplot as plt
from skimage import data
from skimage import filters
from skimage import exposure
from PIL import Image
from urllib.request import urlopen
import matplotlib.pyplot as plt
import numpy as np

url = "https://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg"
gray_img = Image.open(urlopen(url)).convert("L")
face = np.array(gray_img.crop((200,25,500,325)))

val = filters.threshold_otsu(face)
print("val: %d" % val)
```

Assumption:
the image
histogram is
bimodal

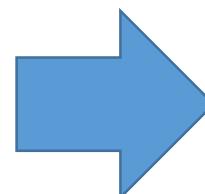


Otsu Thresholding

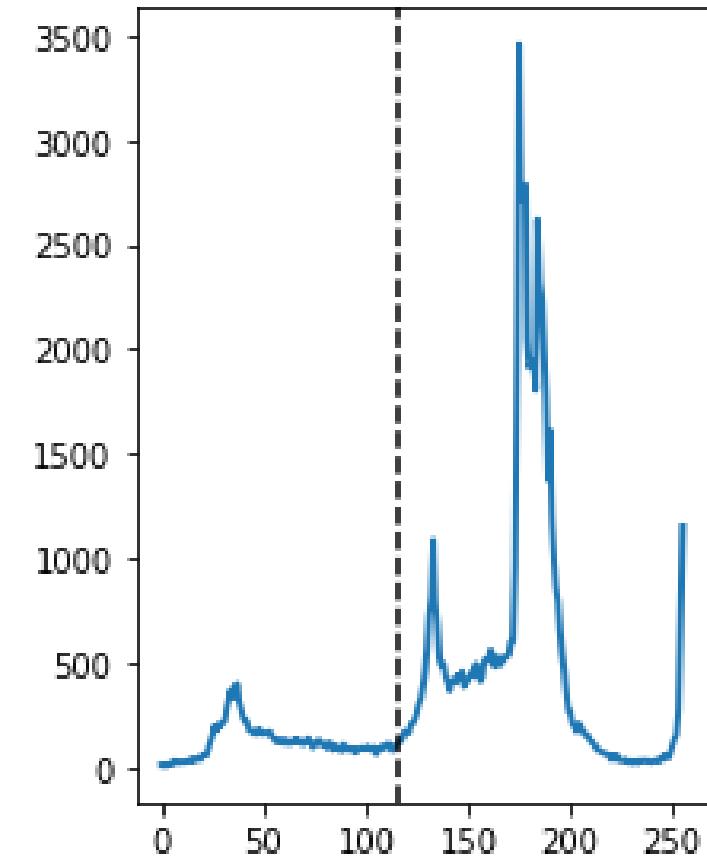
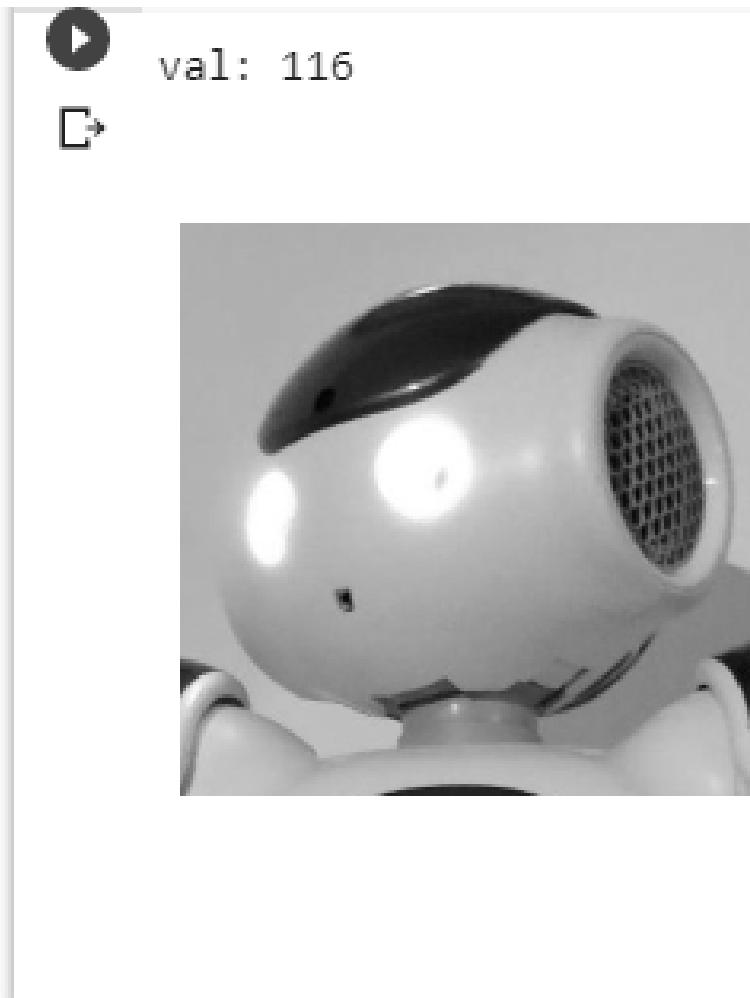
```
hist, bins_center = exposure.histogram(face)

plt.figure(figsize=(9, 4))
plt.subplot(131)
plt.imshow(face, cmap='gray')
plt.axis('off')
plt.subplot(132)
plt.imshow(face < val, cmap='gray') ←
plt.axis('off')
plt.subplot(133)
plt.plot(bins_center, hist, lw=2)
plt.axvline(val, color='k', ls='--')

plt.tight_layout()
plt.show()
```



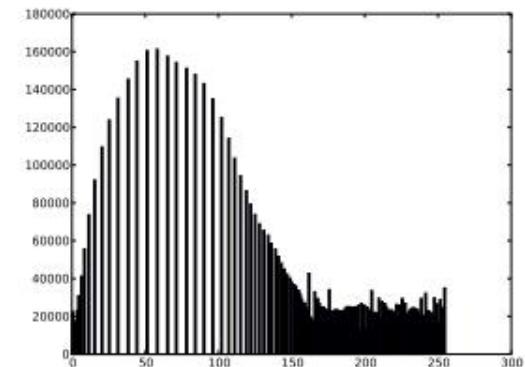
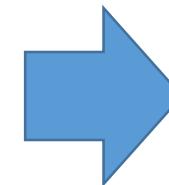
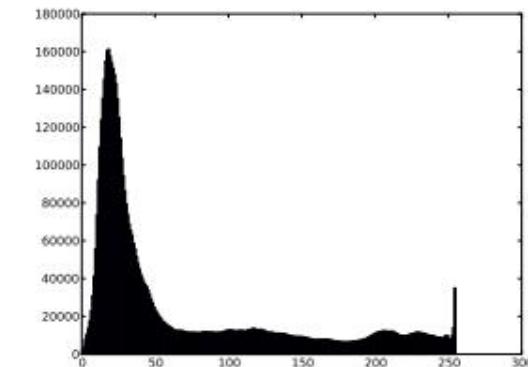
Otsu Thresholding



Histogram Equalization

Histogram equalization flattens the graylevel histogram of an image so that all intensities are as equally common as possible.

This is often a good way to normalize image intensity before further processing and also a way to increase image contrast.



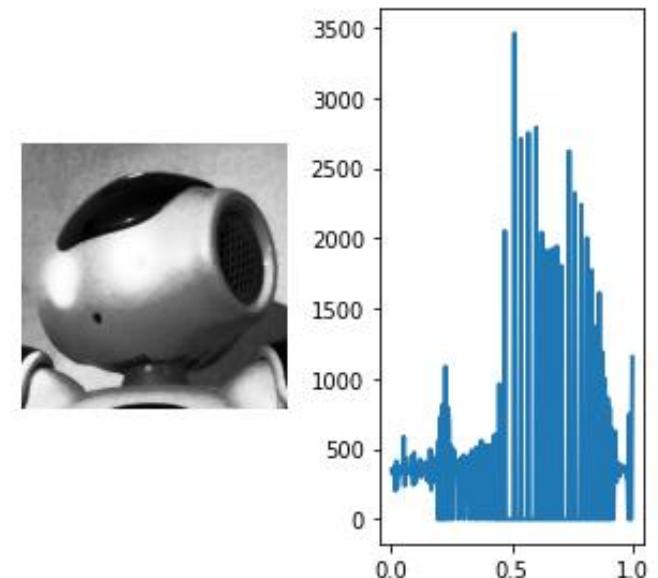
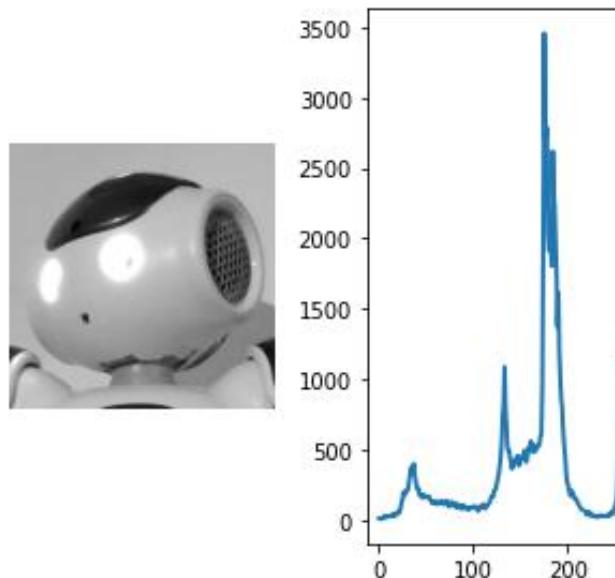
Histogram Equalization

```
equalized_face = exposure.equalize_hist(face)

hist_eq, bins_center_eq = exposure.histogram(equalized_face)

plt.figure(figsize=(9, 4))
plt.subplot(141)
plt.imshow(face, cmap='gray')
plt.axis('off')
plt.subplot(142)
plt.plot(bins_center, hist, lw=2)
plt.subplot(143)
plt.imshow(equalized_face, cmap='gray')
plt.axis('off')
plt.subplot(144)
plt.plot(bins_center_eq, hist_eq, lw=2)

plt.tight_layout()
plt.show()
```



Mathematical Morphology

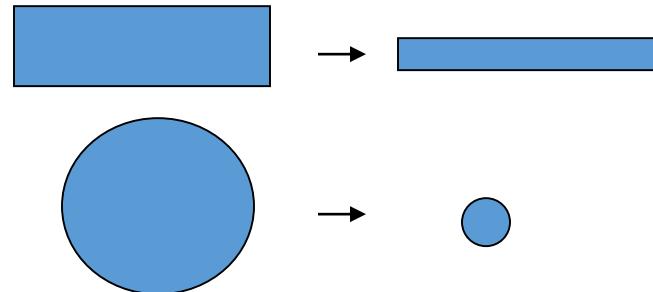
- Erosion
- Dilation
- Closing
- Opening

Erosion

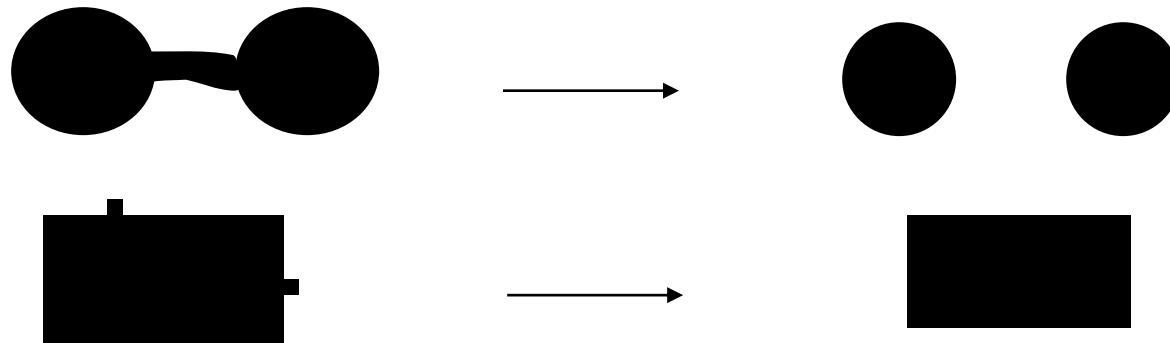
Erosion **shrinks** the connected sets of 1s of a binary image.

It can be used for

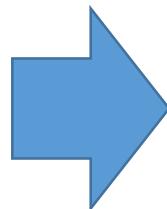
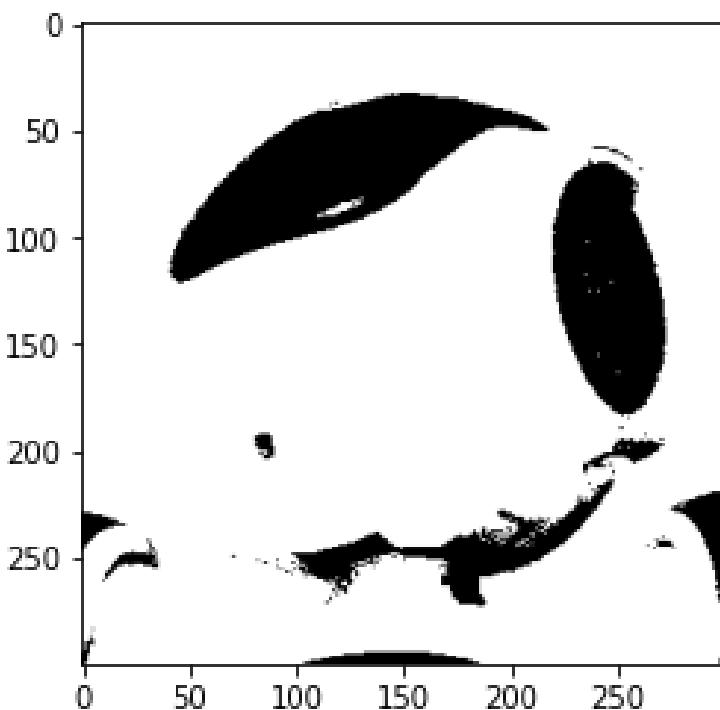
1. shrinking features



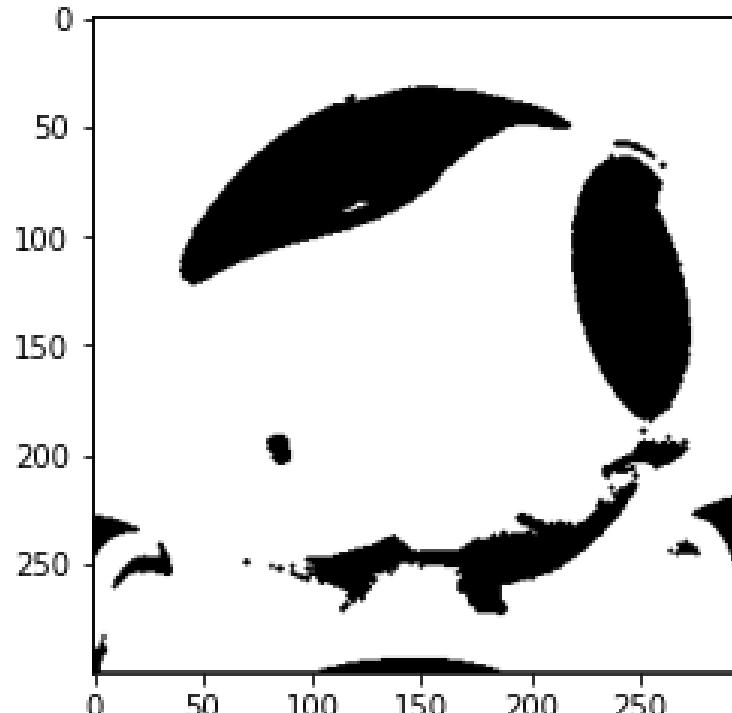
2. Removing bridges, branches and small protrusions



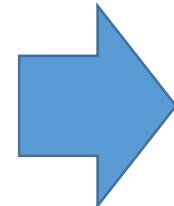
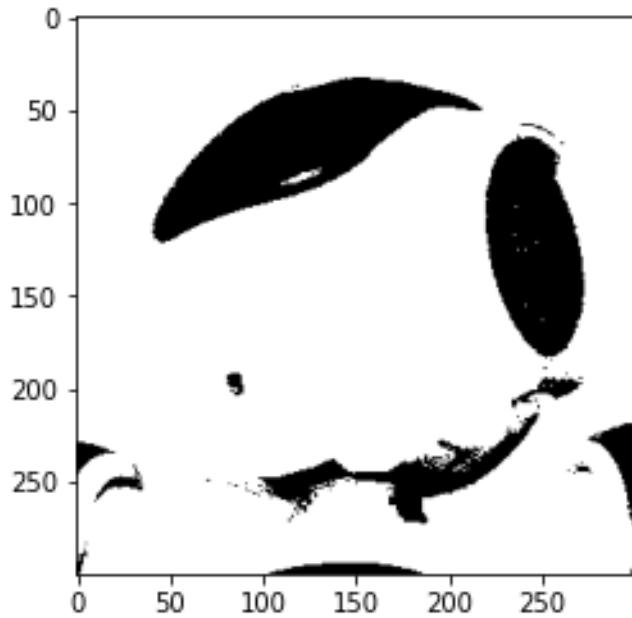
Erosion



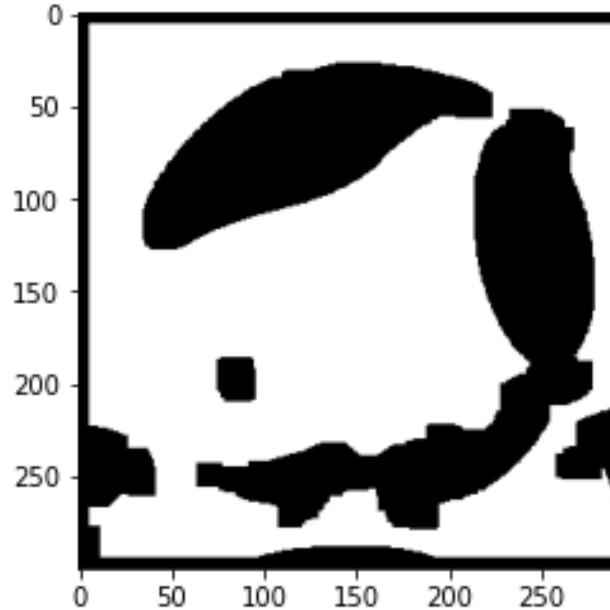
```
from scipy.ndimage import morphology  
  
e = ndimage.binary_erosion(mask)  
  
_ = plt.imshow(e, cmap="gray")
```



Erosion



```
e2 = ndimage.binary_erosion(mask,structure=np.ones((5,5)),iterations=3)  
_ = plt.imshow(e2, cmap="gray")
```

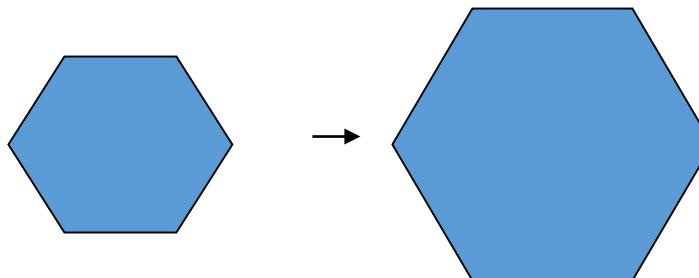


Dilation

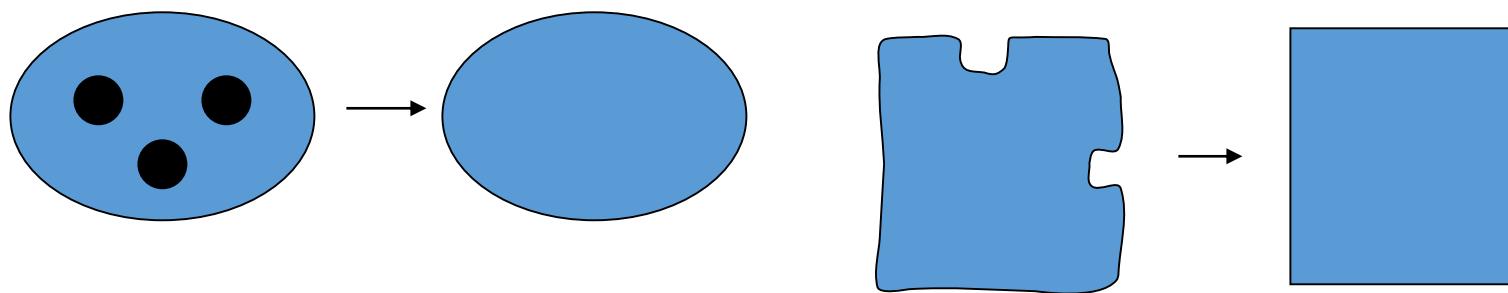
Dilation **expands** the connected sets of 1s of a binary image.

It can be used for

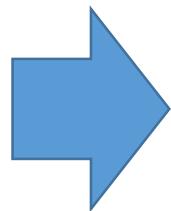
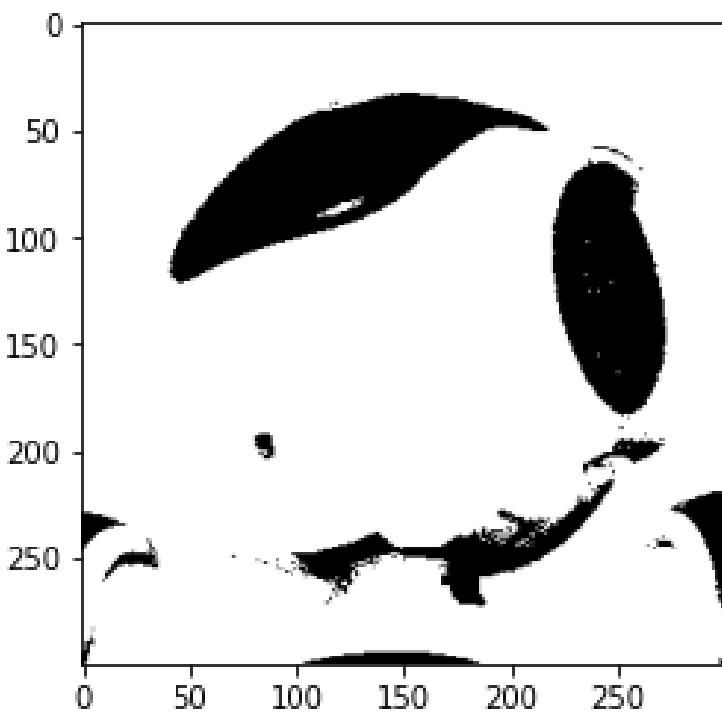
1. growing features



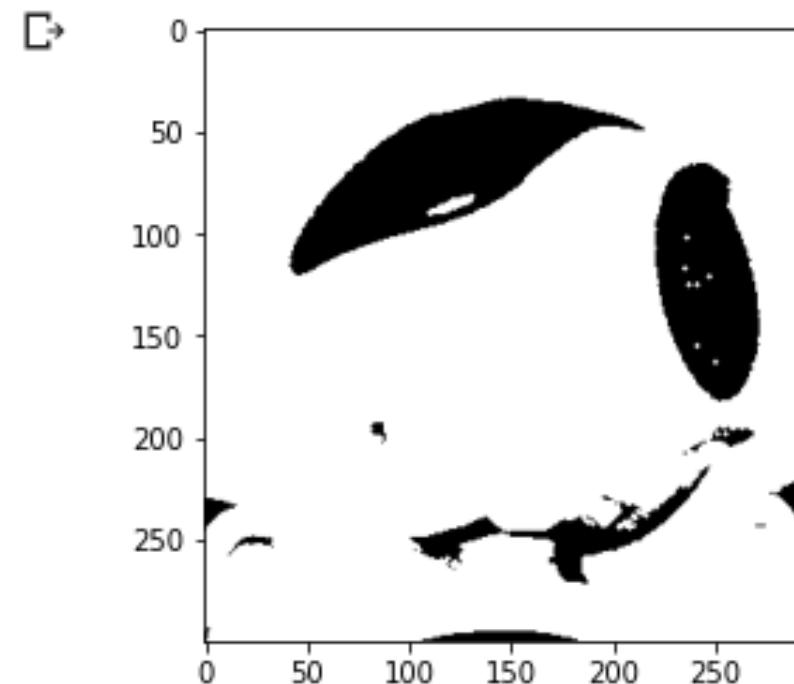
2. filling holes and gaps



Dilation



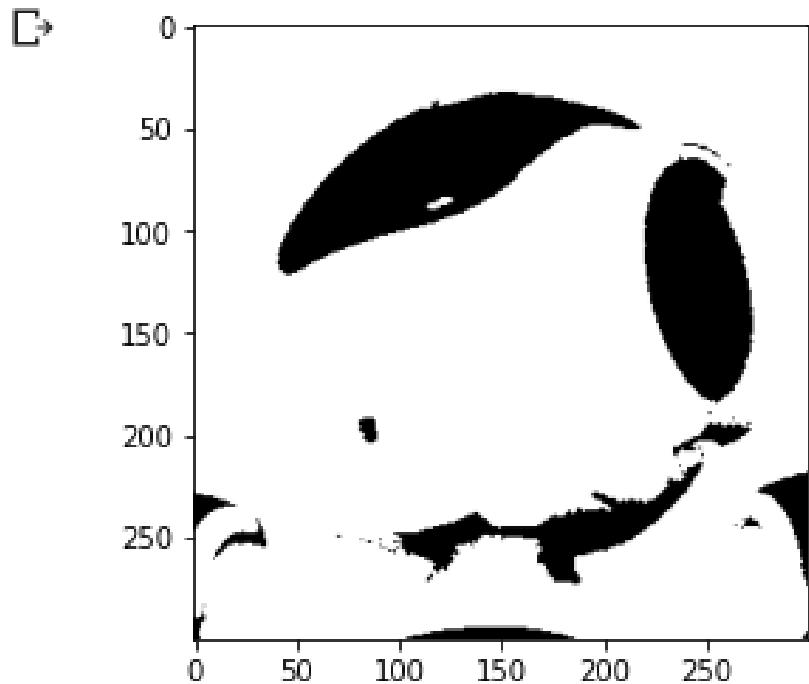
```
▶ from scipy.ndimage import morphology  
  
d = ndimage.binary_dilation(mask)  
  
_ = plt.imshow(d, cmap="gray")
```



Opening



```
from scipy.ndimage import morphology  
  
o = ndimage.binary_opening(mask)  
  
_ = plt.imshow(o, cmap="gray")
```



Opening is the compound operation of erosion followed by dilation

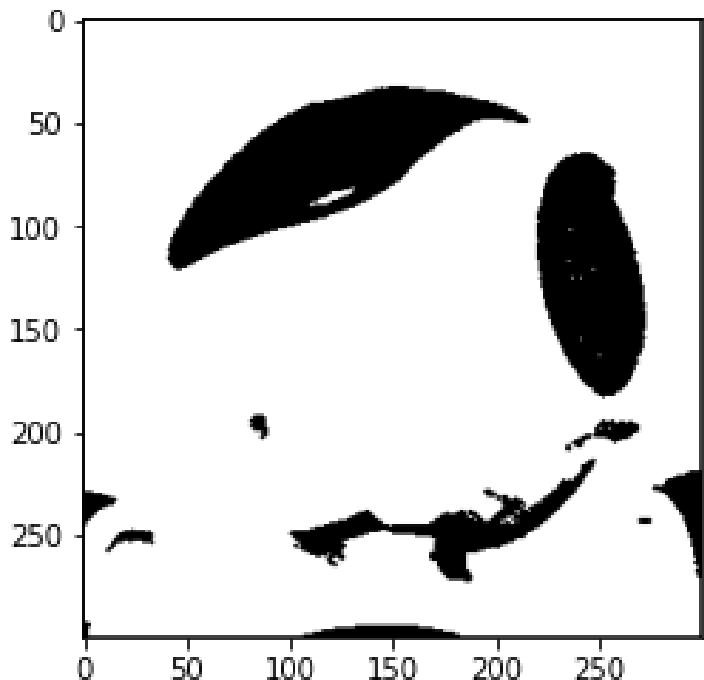
Opening is so called because it can open up a gap between objects connected by a thin bridge of pixels. Any regions that have survived the erosion are restored to their original size by the dilation.

<https://www.cs.auckland.ac.nz/courses/compsci773s1c/lectures/ImageProcessing-html/topic4.htm>

Closing



```
from scipy.ndimage import morphology  
  
c = ndimage.binary_closing(mask)  
  
_ = plt.imshow(c, cmap="gray")
```



Closing is the compound operation of dilation followed by erosion

Closing is so called because it can fill holes in the regions while keeping the initial region sizes.

<https://www.cs.auckland.ac.nz/courses/compsci773s1c/lectures/ImageProcessing-html/topic4.htm>

Esercizio 3

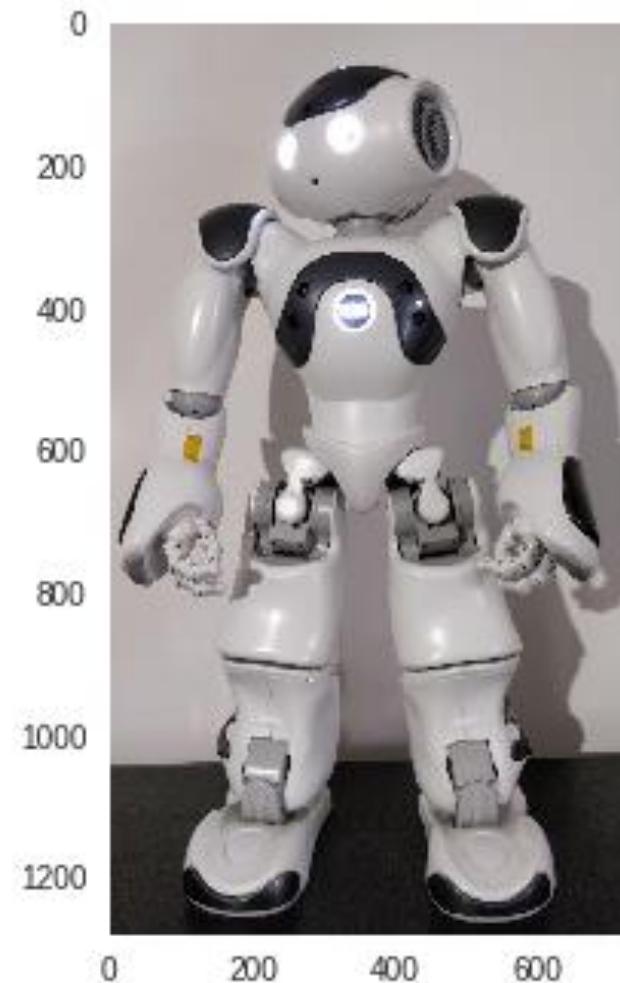
Aprire l'immagine JPEG

<http://web.unibas.it/bloisi/corsi/images/nao-v6-spqr.jpg>

e trasformarla in PNG

Esercizio 3 - soluzione

```
▶ from PIL import Image  
  
import matplotlib.pyplot as plt  
import urllib.request  
  
url = "https://dbloisi.github.io/corsi/images/nao-v6-spqr.jpg"  
  
img = Image.open(urllib.request.urlopen(url))  
  
img.save("nao.png")  
  
!ls  
  
img_png = Image.open("nao.png")  
  
plt.grid(b=False)  
plt.imshow(img_png)
```



Esercizio 4

1. Aprire l'immagine a colori

<https://dbloisi.github.io/corsi/images/nao-v6-spqr.jpg>

2. Estrarre la ROI (300,150,500,200)

3. Incollare la ROI al centro dell'immagine

Esercizio 4 - soluzione

```
from PIL import Image

import matplotlib.pyplot as plt
import urllib.request

url = "https://dbloisi.github.io/corsi/images/nao-v6-spqr.jpg"

img = Image.open(urllib.request.urlopen(url))
print(img.size)

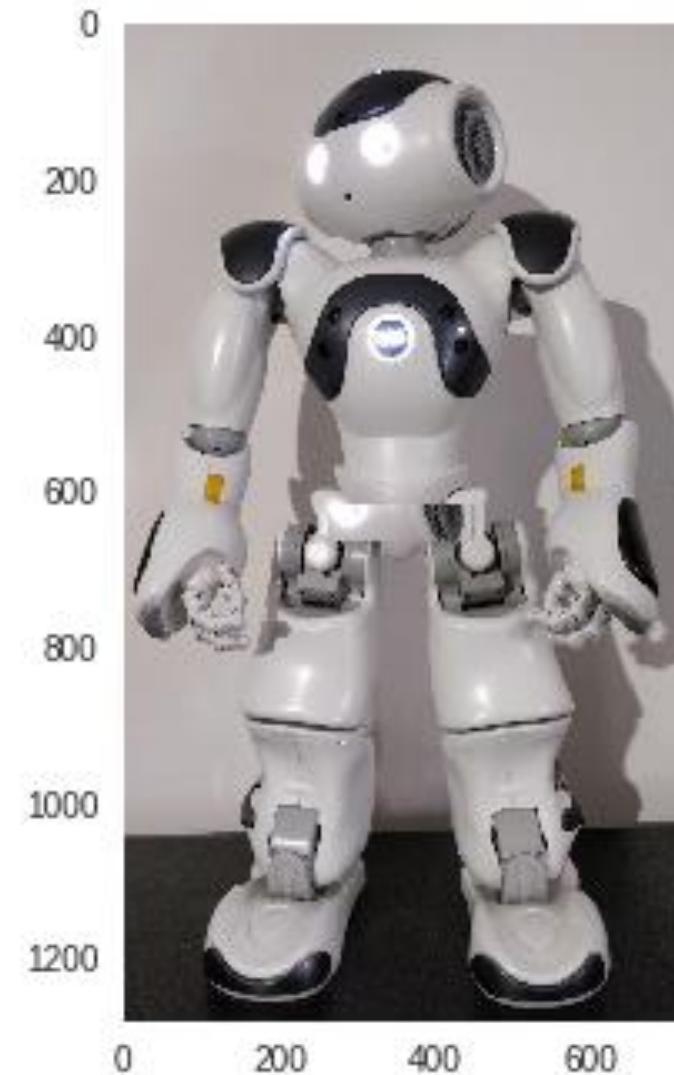
roi = img.crop((300,150,500,200))
print(roi.size)

x = (img.size[0] - roi.size[0]) // 2
y = (img.size[1] - roi.size[1]) // 2

position = (x, y)

img_copy = img.copy()
img_copy.paste(roi, position)

plt.grid(b=False)
plt.imshow(img_copy)
```



Esercizio 4 - soluzione

```
from PIL import Image

import matplotlib.pyplot as plt
import urllib.request

url = "https://dbloisi.github.io/corsi/images/nao-v6-spqr.jpg"

img = Image.open(urllib.request.urlopen(url))
print(img.size)

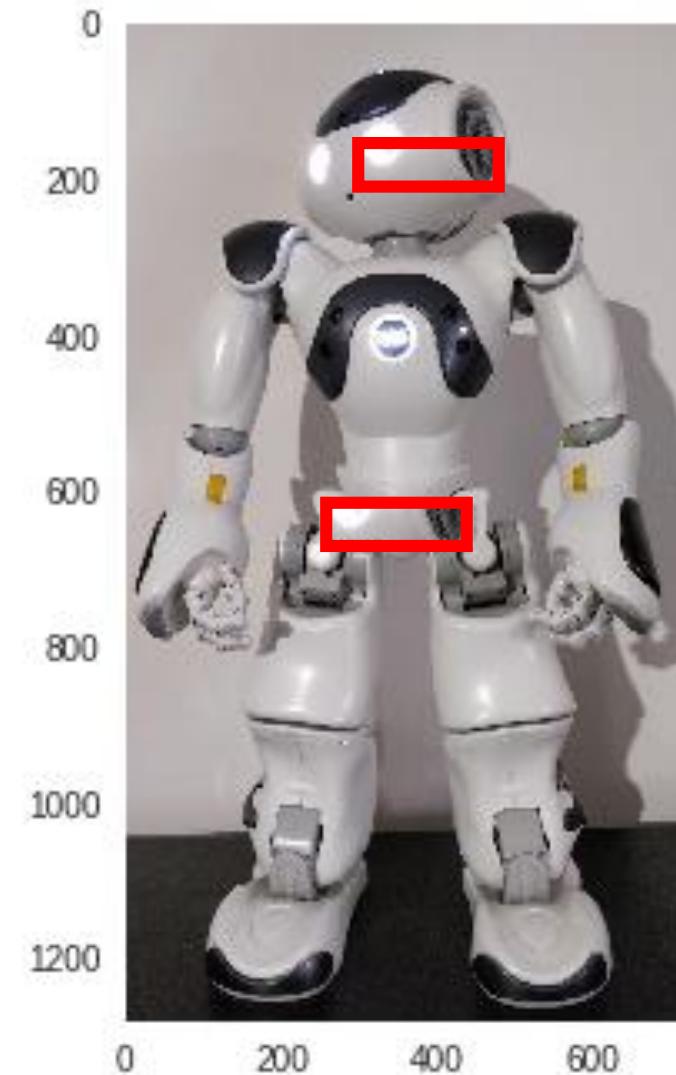
roi = img.crop((300,150,500,200))
print(roi.size)

x = (img.size[0] - roi.size[0]) // 2
y = (img.size[1] - roi.size[1]) // 2

position = (x, y)

img_copy = img.copy()
img_copy.paste(roi, position)

plt.grid(b=False)
plt.imshow(img_copy)
```



Esercizio 5

1. Aprire l'immagine a colori

<https://dbloisi.github.io/corsi/images/nao-v6-spqr.jpg>

2. Salvare una nuova immagine che abbia dimensioni pari a $\frac{1}{4}$ dell'originale

Esercizio 5 - soluzione

```
from PIL import Image

import matplotlib.pyplot as plt
import urllib.request

url = "https://dbloisi.github.io/corsi/images/nao-v6-spqr.jpg"

img = Image.open(urllib.request.urlopen(url))

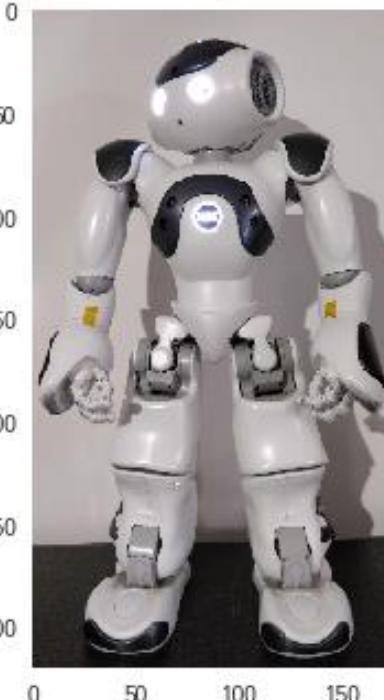
plt.grid(b=False)
plt.imshow(img_copy)

resized_img = img.resize((img.size[0] // 4, img.size[1] // 4))
resized_img.save('resized.jpg')

print(img.size)
print(resized_img.size)

plt.grid(b=False)
plt.imshow(resized_img)
```

(720, 1280)
(180, 320)
<matplotlib.image.AxesImage at 0x7fc05bf212e8>



Esercizio 6

1. Aprire l'immagine a colori

<https://dbloisi.github.io/corsi/images/nao-v6-spqr.jpg>

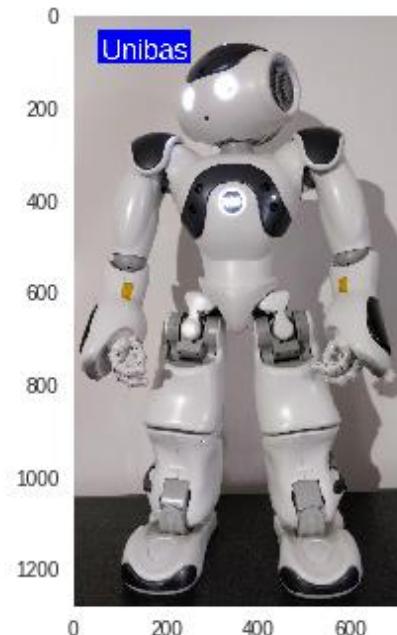
2. Inserire nell'angolo in alto a sinistra dell'immagine la stringa 'Unibas' così come mostrata sotto



Esercizio 6 - soluzione

```
from PIL import Image, ImageDraw, ImageFont  
  
import matplotlib.pyplot as plt  
import urllib.request  
  
url = "https://dbloisi.github.io/corsi/images/nao-v6-spqr.jpg"  
  
img = Image.open(urllib.request.urlopen(url))  
  
img_draw = ImageDraw.Draw(img)  
  
img_draw.rectangle((50, 30, 250, 100), fill='blue')  
  
!ls '/usr/share/fonts/truetype/liberation'  
  
font = ImageFont.truetype(font="LiberationSans-Regular.ttf", size=60)  
  
img_draw.text((60, 40), 'Unibas', fill='white', font=font)  
  
plt.grid(b=False)  
plt.imshow(img)
```

```
LiberationMono-BoldItalic.ttf    LiberationSansNarrow-Bold.ttf  
LiberationMono-Bold.ttf         LiberationSansNarrow-Italic.ttf  
LiberationMono-Italic.ttf     LiberationSansNarrow-Regular.ttf  
LiberationMono-Regular.ttf    LiberationSans-Regular.ttf  
LiberationSans-BoldItalic.ttf  LiberationSerif-BoldItalic.ttf  
LiberationSans-Bold.ttf        LiberationSerif-Bold.ttf  
LiberationSans-Italic.ttf      LiberationSerif-Italic.ttf  
LiberationSansNarrow-BoldItalic.ttf  LiberationSerif-Regular.ttf  
<matplotlib.image.AxesImage at 0x7fc5aa4a9b38>
```



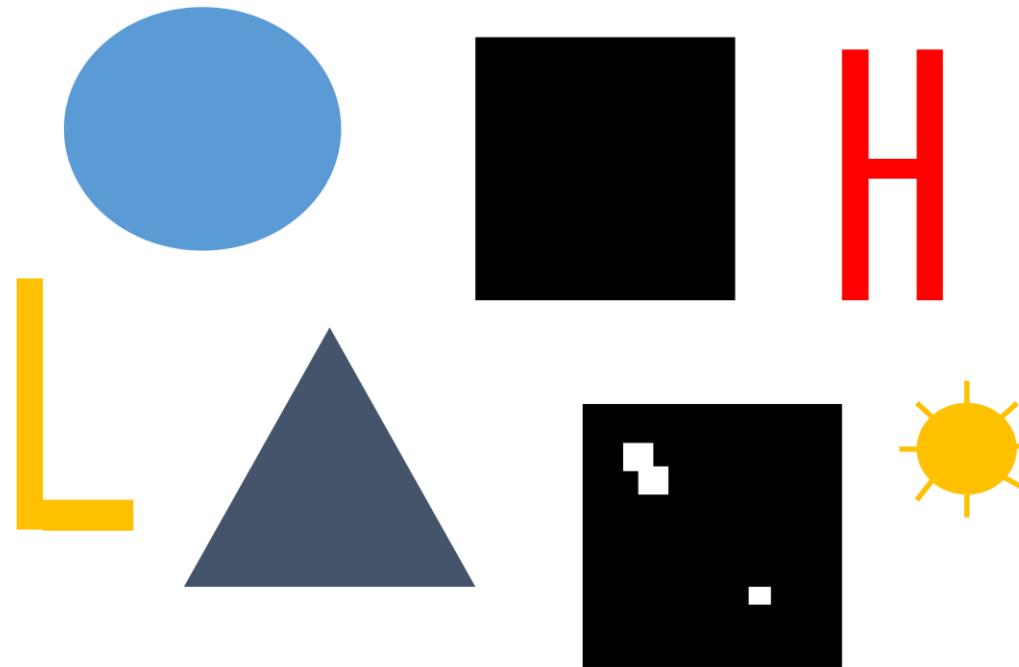
Esercizio 7

Applicare all'immagine

<https://web.unibas.it/bloisi/corsi/images/forme.png>

le operazioni di

- erosion
- dilation
- aperture
- closing

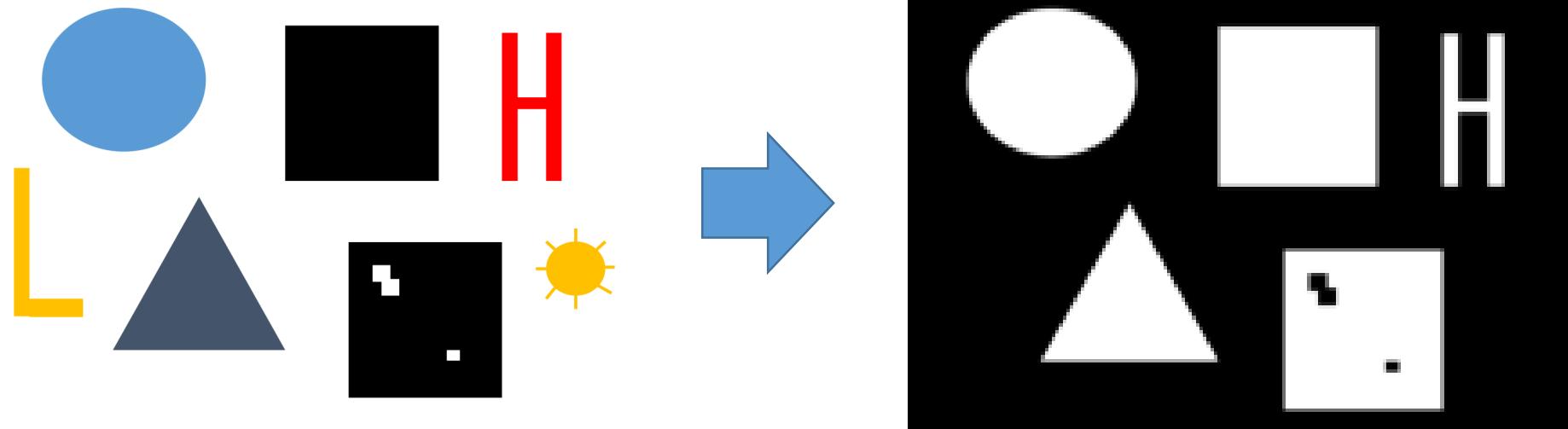


Esercizio 8

Applicare all'immagine

<https://web.unibas.it/bloisi/corsi/images/forme.png>

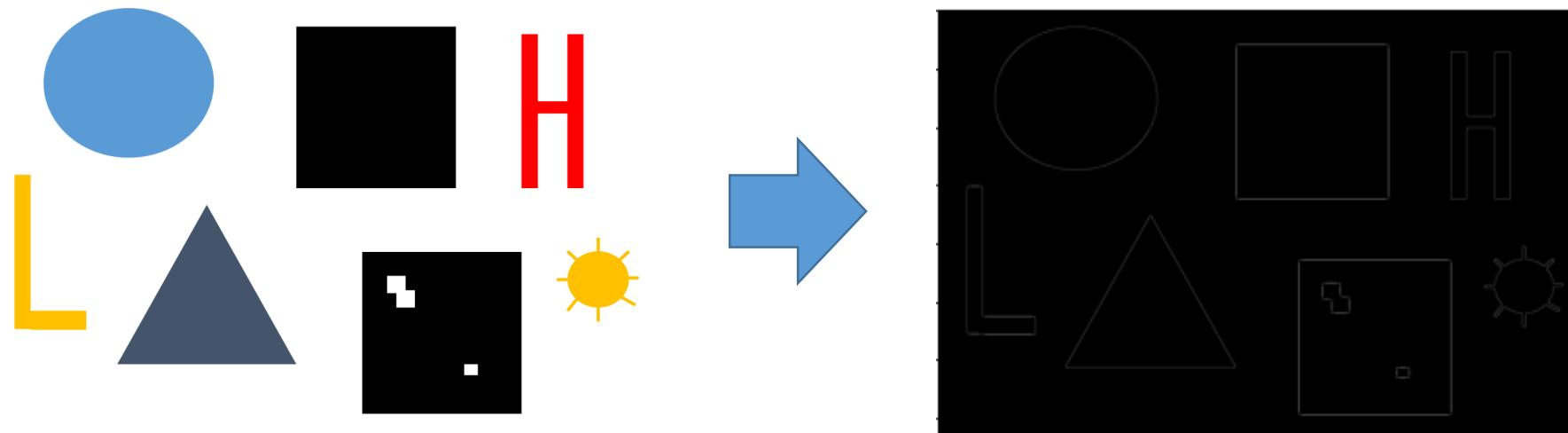
il metodo di thresholding di Otsu



Esercizio 9

Estrarre i contorni dall'immagine

<https://web.unibas.it/bloisi/corsi/images/forme.png>

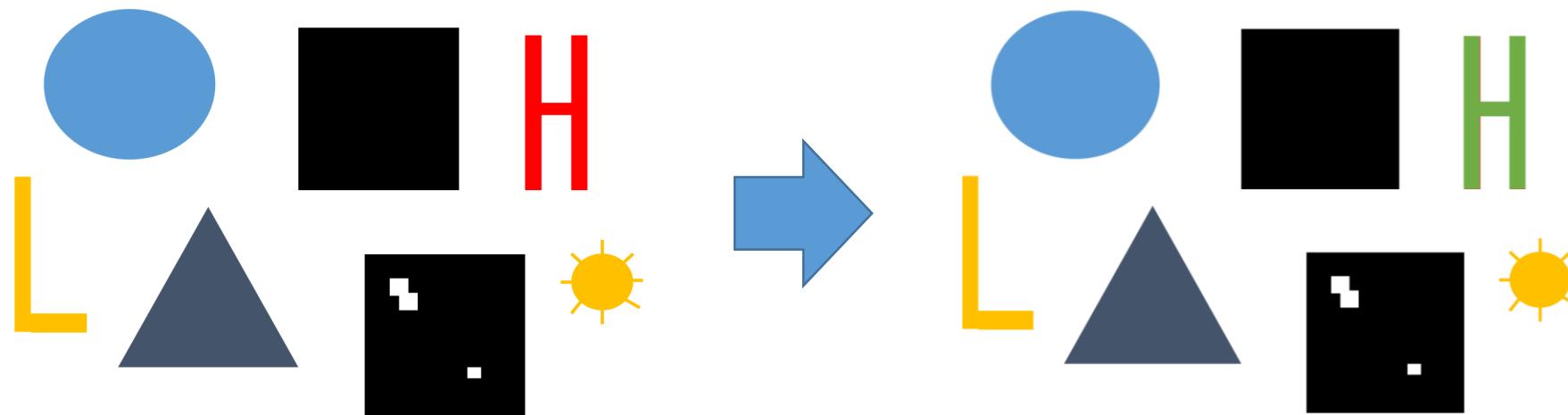


Esercizio 10

Ricolorare la figura in rosso nella immagine

<https://web.unibas.it/bloisi/corsi/images/forme.png>

con il colore verde

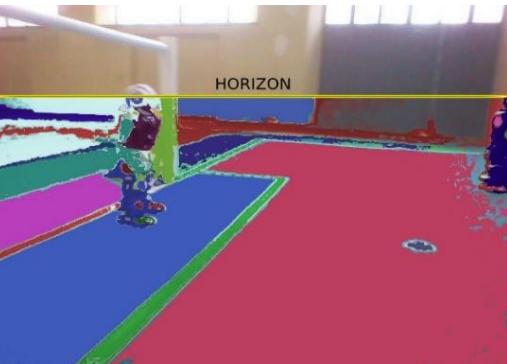




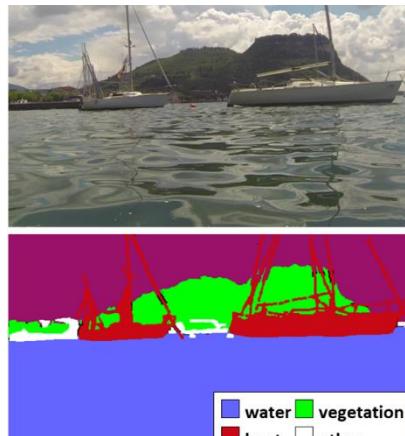
**UNIVERSITÀ DEGLI STUDI
DELLA BASILICATA**

Corso di Visione e Percezione

Trasformazioni



Docente
Domenico D. Bloisi



■ water ■ vegetation
■ boat ■ other