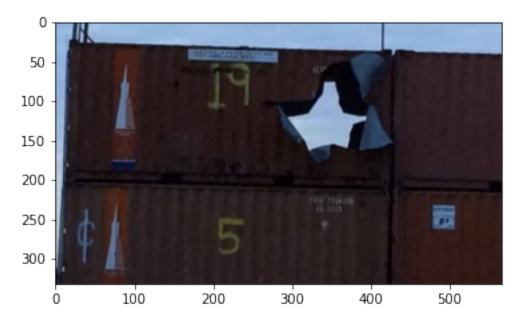
container-2

July 26, 2022

1 Detección de Agujeros en Containers

```
[212]: # Importing necessary libraries
from skimage.io import imread, imshow
import matplotlib.pyplot as plt
import numpy as np

container = imread('images/foto28.jpg')
plt.imshow(container)
plt.show()
```



The next step is convert the image in a gray image.

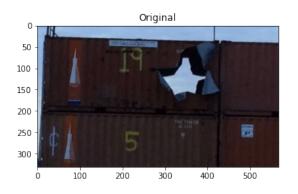
```
[213]: from skimage.color import rgb2gray

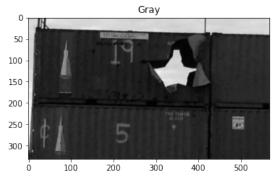
# To gray
```

```
gray_container = rgb2gray(container)

fig, ax = plt.subplots(1, 2, figsize=(12, 4))
ax[0].imshow(container)
ax[0].set_title('Original')
ax[1].imshow(gray_container, cmap='gray')
ax[1].set_title('Gray')
```

[213]: Text(0.5, 1.0, 'Gray')





Next, we need to obtain a binarized image using a threshold value.

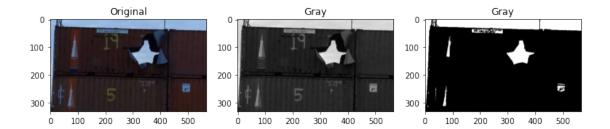
```
# Computing Otsu's thresholding value
threshold = filters.threshold_otsu(gray_container)
print(threshold)

# Binarized image
thresh_gray_container = (gray_container > threshold)*1

fig, ax = plt.subplots(1, 3, figsize=(12, 4))
ax[0].imshow(container)
ax[0].set_title('Original')
ax[1].imshow(gray_container, cmap='gray')
ax[1].set_title('Gray')
ax[2].imshow(thresh_gray_container, cmap='gray')
ax[2].set_title('Gray')
```

0.40399017003676474

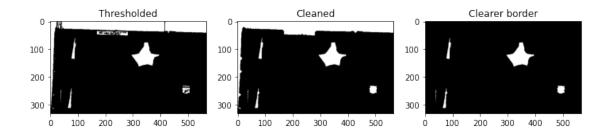
[214]: Text(0.5, 1.0, 'Gray')



Next, now we must clean the data. For that, we use ditations and erosions.

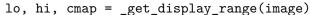
```
[215]: from skimage import morphology
       from skimage.segmentation import clear_border
       def multi_dil(im, num):
           for i in range(num):
               im = morphology.binary_dilation(im)
           return im
       def multi_ero(im, num):
           for i in range(num):
               im = morphology.binary_erosion(im)
           return im
       # We use this chain operation to clean the image:
       container_cleaned = multi_ero(multi_dil(thresh_gray_container, 5), 5)
       # remove artifacts connected to image border
       container_cleared = clear_border(container_cleaned)
       fig, ax = plt.subplots(1, 3, figsize=(12, 4))
       ax[0].imshow(thresh_gray_container, cmap='gray')
       ax[0].set_title('Thresholded')
       ax[1].imshow(container_cleaned, cmap='gray')
       ax[1].set_title('Cleaned')
       ax[2].imshow(container_cleared, cmap='gray')
       ax[2].set_title('Clearer border')
```

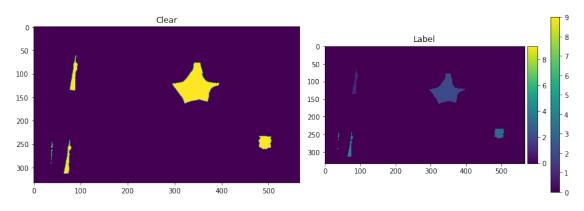
[215]: Text(0.5, 1.0, 'Clearer border')



Next, and now that this is relatively clean, let us get the labels and properties of this last image!

/home/juan/.virtualenvs/dl4cv/lib/python3.8/site-packages/skimage/io/_plugins/matplotlib_plugin.py:150: UserWarning: Low image data range; displaying image with stretched contrast.





There are a lot of features that can be extracted from this labelled image. To mention a few we have:

Now, let's use regionprops and look at the following properties:

```
1. area
```

- 2. perimeter
- 3. bbox bounding box dimensions
- 4. bbox area area of bounding box
- 5. centroid coordinate of centroid
- 6. convex image convex hull of the blob
- 7. convex area area of the convex hull
- 8. eccentricity measure how it fits into an ellipse (0) for circles (how elongated is your object)
- 9. major axis length length of the major moment of the ellipse fitted
- 10. minor axis length length of the minor moment of the ellipse fitted

Let us try to take the area of the bigger area object:

```
[217]: from skimage.measure import regionprops

props=regionprops(label_container)

print(len(props))

for i in range(len(props)):
    print(i)
    print(props[i].area) #area (zero) Oth object in the image
    print(props[i].bbox)
```

```
9
0
536
(62, 77, 136, 94)
3978
(77, 296, 164, 396)
2
643
(233, 480, 261, 507)
3
421
(241, 64, 313, 81)
4
(246, 37, 258, 41)
5
21
(262, 38, 279, 40)
```

```
3 (281, 37, 283, 39) 7 1 (286, 38, 287, 39) 8 3 (289, 37, 292, 38)
```

Now, I can calculate the bigger area in the image. But I want to see some statistics from objects earlier.

```
centroid-0 centroid-1 area
[218]:
                                       bbox-0
                                                bbox-1
                                                        bbox-2 bbox-3
       0 105.608209
                      85.641791
                                   536
                                            62
                                                    77
                                                           136
                                                                    94
       1 126.394419 343.692810 3978
                                            77
                                                   296
                                                           164
                                                                   396
                      493.063764
       2 245.832037
                                   643
                                           233
                                                   480
                                                           261
                                                                   507
       3 284.912114
                      72.422803
                                   421
                                           241
                                                    64
                                                           313
                                                                    81
       4 252.631579
                      38.789474
                                                    37
                                    19
                                           246
                                                           258
                                                                    41
       5 269.095238
                      38.190476
                                    21
                                           262
                                                    38
                                                           279
                                                                    40
       6 281.333333
                       37.666667
                                     3
                                           281
                                                    37
                                                           283
                                                                     39
       7 286.000000
                       38.000000
                                           286
                                                    38
                                     1
                                                           287
                                                                     39
       8 290.000000
                       37.000000
                                     3
                                           289
                                                    37
                                                           292
                                                                     38
```

First af all, i want to have all data about the main feature which is very significative for that object. In this case area.

```
[219]: onlyArea = [props[i].area for i in range (len(props))]

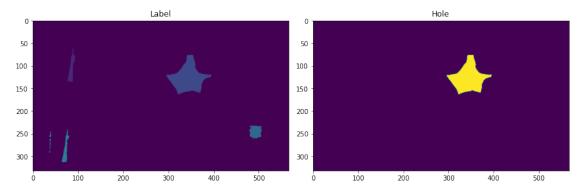
# With index_ob we can obtain any feature of the object (our hole) using
# props[index_ob].feature
index_ob = onlyArea.index(max(onlyArea))

# labels is 1 from N but index is from 0 to N-1.
hole_container = label_container == index_ob+1

fig, ax = plt.subplots(1, 2, figsize=(12,4))
ax[0].imshow(label_container)
ax[0].set_title('Label')
```

```
ax[1].imshow(hole_container)
ax[1].set_title('Hole')

#fig.colorbar(imshow(label_container), ax=ax[1])
fig.tight_layout()
```



Finally, I only have that draw the segmented object on the original figure (in gray for contrast)

[220]: <matplotlib.patches.Rectangle at 0x7f2743dc9880>

