

```
from google.colab import drive
drive.mount('/content/drive',force_remount=True)
```

Mounted at /content/drive

/content/drive/MyDrive/VR/coins.jpg

```
import cv2
import matplotlib.pyplot as plt
import numpy as np
import os

path = "/content/drive/MyDrive/VR/coins.jpg"

def read_image(path):
    if os.path.exists(path):
        img = cv2.imread(path)
        imgrgb = cv2.cvtColor(img,cv2.COLOR_BGR2RGB)
        plt.imshow(imgrgb)
        plt.title("Original Image of coins")
        plt.axis('off')
        return img
    else:
        print("file not found")
        exit()

img = read_image(path)
```

Original Image of coins



```
gray = cv2.cvtColor(img,cv2.COLOR_BGR2GRAY)
plt.imshow(gray,cmap='gray')
plt.title("Greyscale of original Image")
plt.axis('off')
plt.show()
```

Greyscale of original Image



#Downsample images based on size

```
print(gray.shape)
(3000, 4000)
def downsample_images(image):
    if image.shape[0] > 512 or image.shape[1] > 512:
        while image.shape[0] > 512 or image.shape[1] > 512:
            image = cv2.pyrDown(image)
    return image
gray = downsample_images(gray)
print(gray.shape)
(375, 500)
```

Make image sharper to detect edges better

```
def high_pass_filter(image):
    # Apply Laplacian(2nd order) filter (keep as float64)
    laplacian = cv2.Laplacian(image, cv2.CV_64F)
```

```

# Convert image and Laplacian to same type
image_64f = image.astype(np.float64)
laplacian_64f = laplacian.astype(np.float64)

# Sharpen the image => 1.5*image - 0.5*laplacian
sharpened_image = cv2.addWeighted(image_64f, 1.5, laplacian_64f, -
0.5, 0)

# Convert back to uint8
sharpened_image = np.clip(sharpened_image, 0,
255).astype(np.uint8)
laplacian = np.clip(laplacian, 0, 255).astype(np.uint8) # Convert
Laplacian for display

return laplacian, sharpened_image

laplacian, sharpened_image = high_pass_filter(gray)

fig, ax = plt.subplots(1, 3, figsize=(12, 16))

ax[0].imshow(gray, cmap='gray')
ax[0].set_title('Original Image')
ax[0].axis('off')

ax[1].imshow(laplacian, cmap='grey')
ax[1].set_title('Laplacian Image')
ax[1].axis('off')

ax[2].imshow(sharpened_image, cmap='gray')
ax[2].set_title('Sharpened Image')
ax[2].axis('off')

(-0.5, 499.5, 374.5, -0.5)

```



going with original image since sharpened image removed too much information

blur image to remove noise

```
# image, kernel size, sigma
image_blurred = cv2.GaussianBlur(sharpened_image,(3, 3),0)

fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(12, 6))

ax1.imshow(sharpened_image,cmap='grey')
ax1.set_title('Sharpened Image')
ax1.axis('off')

ax2.imshow(image_blurred, cmap='gray')
ax2.set_title('Blurred Image')
ax2.axis('off')

(-0.5, 499.5, 374.5, -0.5)
```

Sharpened Image



Blurred Image



#Trying out different values of canny

```
def cannyplots(image):
    fig, ax = plt.subplots(2, 6, figsize=(18, 4))
    for i in range(11):
        row = i // 6
        col = (i % 6)

        #image, lower threshold, higher threshold
        edges_img = cv2.Canny(image, i * 30, i * 30 + 60)
        edges_sharpened_img = cv2.Canny(np.uint8(image_blurred), i * 30,
i * 30 + 60)

        ax[row, col].imshow(edges_img, cmap='gray')
        ax[row, col].set_title(f'Canny {i * 30}-{i * 30 + 60}')
        ax[row, col].axis('off')

    ax[1, 5].imshow(image, cmap='gray')
```

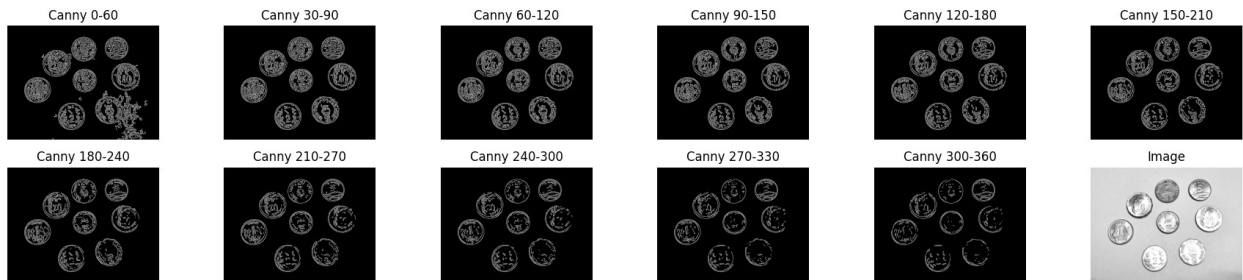
```

ax[1, 5].set_title('Image')
ax[1, 5].axis('off')

plt.tight_layout()
plt.show()

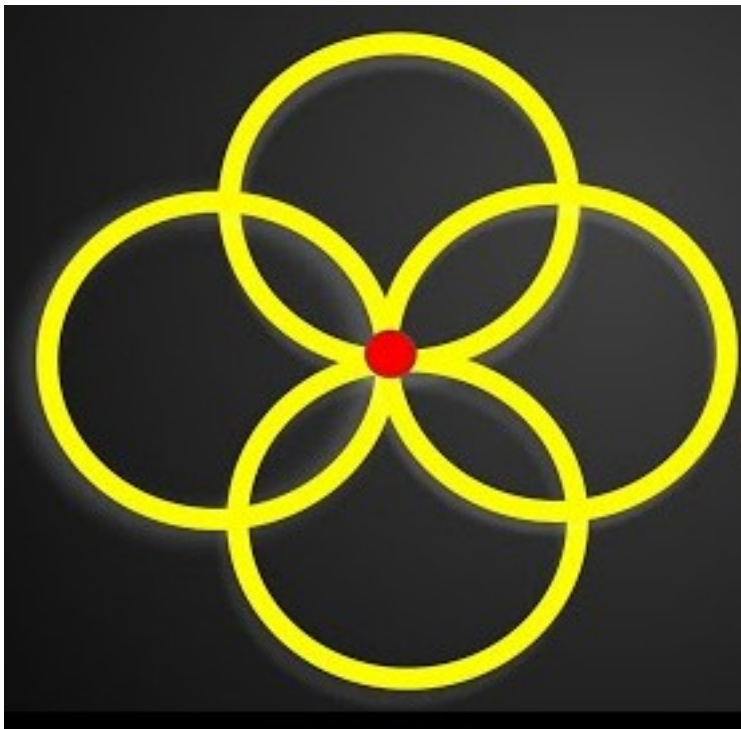
cannyplots(image_blurred)

```



Method 1 : Non overlapping coins : Hough Transform

Works by drawing circles on image where edges are present if the concentric circles zero in some point then this point is deemed center of circle



```

canny_lower = int(input("Enter appropriate value for lower
threshold : "))
canny_higher = int(input("Enter appropriate value for higher
threshold : "))

Enter appropriate value for lower threshold : 60
Enter appropriate value for higher threshold : 120

hough_image_input = image_blurred.copy()

circles = cv2.HoughCircles(
    hough_image_input,                # Image input
    cv2.HOUGH_GRADIENT,              # Detection method
    cv2.HOUGH_STANDARD, cv2.HOUGH_PROBABILISTIC for lines
    cv2.HOUGH_GRADIENT for circles

    ##### ADJUST THESE PARAMETERS FOR BETTER DETECTION
    #####

    dp=1,                            # Inverse ratio of
    resolution                        # Minimum distance
    minDist = gray.shape[0] // 8,     # Minimum distance
    between detected centers
    param1 = canny_higher,            # Canny edge upper
    threshold
    param2 = 30,                      # Hough accumulator
    threshold - minimum 30 overlapping points with circle
    minRadius = (gray.shape[0]//15),  # Minimum coin radius
    maxRadius = (gray.shape[0]//8)    # Maximum coin radius

    #####
    #####
)

circles = circles[0]
print(circles)

[[208.5 297.5 41. ]
 [389.5 168.5 45.4]
 [250.5  79.5 39.2]
 [ 99.5 212.5 41.6]
 [255.5 181.5 37.2]
 [360.5  75.5 33.7]
 [331.5 276.5 42.1]
 [158.5 126.5 41. ]]

if circles is not None:
    circles = np.uint16(np.around(circles))
Convert to integers
#

```

```

    for circle in circles:
        x, y, r = circle
        # image, center, radius, color,
        cv2.circle(hough_image_input, (x, y), r, (0, 0, 0), 2) # Draw
        # outer circle
        cv2.circle(hough_image_input, (x, y), 0, (0, 0, 0), 3) # Draw
        # center

plt.imshow(hough_image_input, cmap='gray')
plt.title(f'Detected Coins - number of coins {len(circles)}')
plt.axis('off')
plt.show()

```

Detected Coins - number of coins 8



Method 2 : Overlapping coins : Watershed

```

path = "/content/drive/MyDrive/VR/coins4.jpg"
img = read_image(path)

```


Original Image of coins



```
gray = cv2.cvtColor(img,cv2.COLOR_BGR2GRAY)
plt.imshow(gray,cmap='gray')
plt.title("Greyscale of original Image")
plt.axis('off')
plt.show()
```

Greyscale of original Image



```
gray = downsample_images(gray)
print(gray.shape)

(312, 252)

laplacian, sharpened_image = high_pass_filter(gray)
```

Apply threshold

```
"""
    Lower and upper threshold are dummy as
    OTSU chooses best threshold by itself
    based on background and foreground
"""

# image, lower threshold, upper threshold, apply OTSU's [chooses
# optimal threshold] and invert the values
ret, thresh = cv2.threshold(sharpened_image, 160, 255, cv2.THRESH_OTSU +
cv2.THRESH_BINARY_INV)

fig, ax = plt.subplots(1, 2, figsize=(10, 6))

ax[0].imshow(cv2.cvtColor(sharpened_image, cv2.COLOR_BGR2RGB))
ax[0].set_title('Sharpened Image')
```

```

ax[0].axis('off')

ax[1].imshow(thresh,cmap='gray')
ax[1].set_title(f'Threshold Image with threshold- {int(ret)}')
ax[1].axis('off')

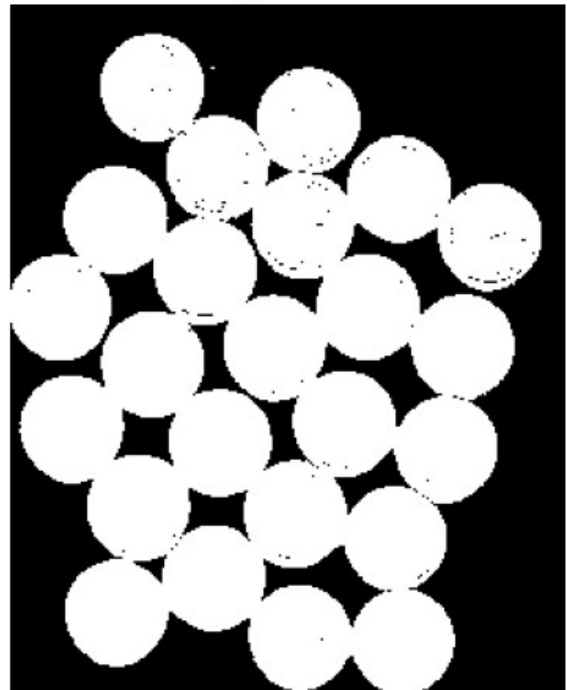
(-0.5, 251.5, 311.5, -0.5)

```

Sharpened Image



Threshold Image with threshold- 192



#Fill empty spaces if region is bounded by filled regions

*# morphologyEx => Closing - Dilation followed by Erosion.
It is useful in closing small holes inside the foreground objects,
or small black points on the object.*

```

kernel = np.ones((5,5), np.uint8)
filled_image = cv2.morphologyEx(thresh, cv2.MORPH_CLOSE, kernel,
iterations=1)

```

```

fig, ax = plt.subplots(1, 2, figsize=(10,6))

```

```

ax[0].imshow(cv2.cvtColor(img,cv2.COLOR_BGR2RGB))
ax[0].set_title('Original Image')
ax[0].axis('off')

```

```

ax[1].imshow(filled_image,cmap='gray')
ax[1].set_title('Filled Image')
ax[1].axis('off')

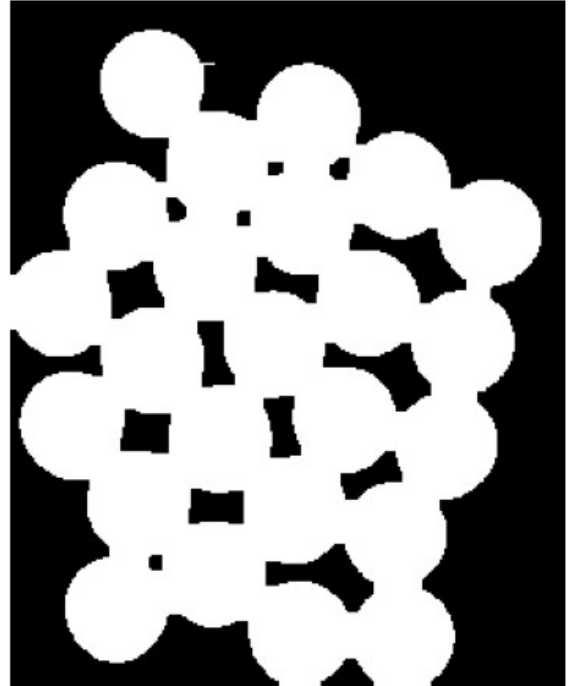
```

```
plt.show()
```

Original Image



Filled Image



#Use this block of code only when there is too much overlapping

```
# sure_bg = cv2.erode(filled_image, kernel, iterations=1)
# fig, ax = plt.subplots(1, 2, figsize=(10,6))
# ax[0].imshow(cv2.cvtColor(img, cv2.COLOR_BGR2RGB))
# ax[0].set_title('Original Image')
# ax[0].axis('off')
# ax[1].imshow(sure_bg, cmap='gray')
# ax[1].set_title('Sure Background Image')
# ax[1].axis('off')
# plt.show()
```

#Extracting center of coins

```
"""
    dist_transform = for every image find out it's closes pixel
    nearest to 1
    so points in centre of circle will be light and points outside
    circle will be darker

```

```

    cv2.threshold = extracts regions that are closer to centre of the
    circle
    """

dist_transform = cv2.distanceTransform(filled_image,cv2.DIST_L2,3)
ret, sure_fg =
cv2.threshold(dist_transform,0.8*dist_transform.max(),255,cv2.THRESH_B
INARY)

sure_fg = np.uint8(sure_fg)

if 'sure_bg' not in globals():
    sure_bg = filled_image

difference_image = cv2.subtract(filled_image,sure_fg)

fig, ax = plt.subplots(1, 5, figsize=(18,6))

ax[0].imshow(cv2.cvtColor(img,cv2.COLOR_BGR2RGB))
ax[0].set_title('Original Image')
ax[0].axis('off')

ax[1].imshow(dist_transform,cmap='gray')
ax[1].set_title('Distance Transform Image')
ax[1].axis('off')

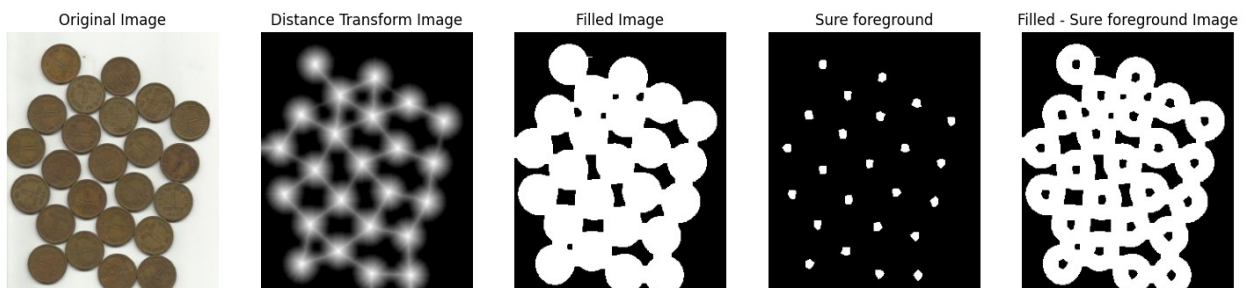
ax[2].imshow(filled_image,cmap='gray')
ax[2].set_title('Filled Image')
ax[2].axis('off')

ax[3].imshow(sure_fg,cmap='gray')
ax[3].set_title('Sure foreground')
ax[3].axis('off')

ax[4].imshow(difference_image,cmap='gray')
ax[4].set_title('Filled - Sure foreground Image')
ax[4].axis('off')

plt.show()

```



Obtaining boundaries of coins

```

"""
sure_fg (Binary Image)      connectedComponents()      markers + 1
difference_image      Final markers
-----
-----
1  1  0  0  0      1  1  0  0  0      2  2  1  1
1      0  0  255 255  0      2  2  0  0  1
1  1  0  0  0      1  1  0  0  0      2  2  1  1
1      0  0  255 255  0      2  2  0  0  1
0  0  1  1  1      0  0  2  2  2      1  1  3  3
3      255 255  0  0  0      0  0  3  3  3
0  0  1  1  1      0  0  2  2  2      1  1  3  3
3      255 255  0  0  0      0  0  3  3  3
0  0  1  1  1      0  0  2  2  2      1  1  3  3
3      0  0  0  0  0      1  1  3  3  3
"""

total_coins, markers = cv2.connectedComponents(sure_fg)
markers = markers+1

#removes regions where you can see white regions in last step above
markers[difference_image==255] = 0

fig, ax = plt.subplots(1, 2, figsize=(10,6))

ax[0].imshow(cv2.cvtColor(img,cv2.COLOR_BGR2RGB))
ax[0].set_title('Original Image')
ax[0].axis('off')

ax[1].imshow(markers,cmap='gray')
ax[1].set_title('Marker Center Image')
ax[1].axis('off')

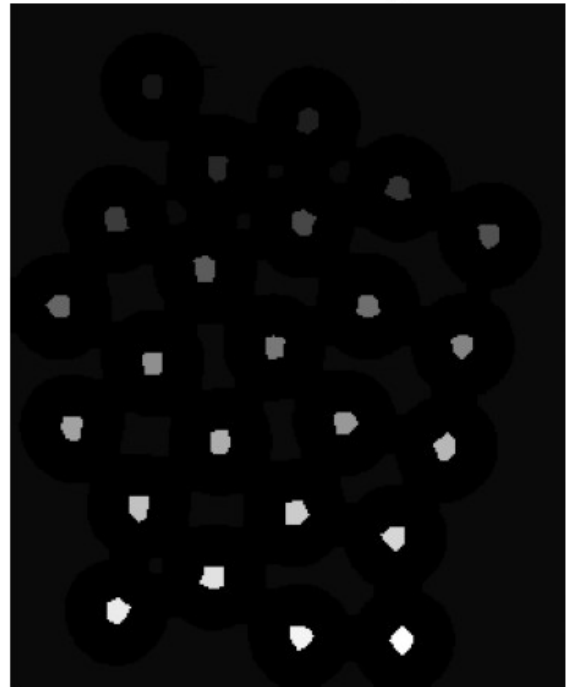
plt.show()

```

Original Image



Marker Center Image



Imagine a geographical map where high areas are ridges (boundaries) and low areas are valleys (objects). The watershed algorithm works like filling a terrain with water, where each region gets flooded from its lowest points until water from different sources meets. The meeting points become the boundaries.

It floods from the markers and expands outward. When two regions meet, a watershed boundary (edge) is formed and labeled as -1.

```
watershed_img = img.copy()

markers = cv2.watershed(watershed_img, markers)
img[markers == -1] = [255, 0, 0]

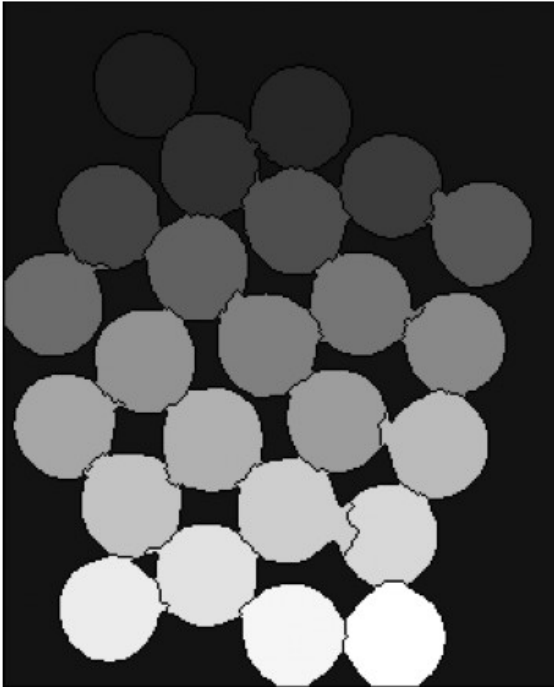
fig, ax = plt.subplots(1, 2, figsize=(10, 6))

ax[0].imshow(markers, cmap='gray')
ax[0].set_title('Marker Image')
ax[0].axis('off')

ax[1].imshow(cv2.cvtColor(watershed_img, cv2.COLOR_BGR2RGB))
ax[1].set_title(f'Original Image with watershed - coins {total_coins-1}')
ax[1].axis('off')

plt.show()
```


Marker Image



Original Image with watershed - coins 24

