

# Lab-7

# BLOB Detection

- Blob detection is a basic method in computer vision used to locate **areas of interest** in a picture.
- Blob detection is the process of **finding related areas** in an image that share features.
- It assembles linked **pixels into clusters**.
- **Applications** for blob detection include image segmentation, feature extraction, object tracking, etc.
- **Note:-** you can apply various **morphological, smoothing** operations in combination before apply blob detector function for better results.
  - To improve blob detection performance, image preprocessing, including **noise reduction, contrast enhancement**, and **color space conversions**.

# Techniques

- **Laplacian of Gaussian (LoG)**: Gaussian-smoothed image's Laplacian is the method used here.
  - **Most accurate** but slowest
- **Difference of Gaussian (DoG)**: DoG algorithm computes the difference between two Gaussian-smoothed pictures.
  - Faster approximation of LoG
- **Determinant of Hessian (DoH)**: Uses the determinant of the Hessian matrix.
  - Here, local curvature of an image is represented by the Hessian matrix.
  - **Fastest Method**

# Steps and Functions Used

- Load libraries and images.
- Set Up the Simple Blob Detector
  - `cv2.SimpleBlobDetector()` (**old version**)
  - `cv2.SimpleBlobDetector_Create ()`
  - `cv2.SimpleBlobDetector_Params()`
- Create keypoints detector using `.detect(img)` function
- Obtain Key Points on the Image `cv2.drawKeypoints( )`
- Display using `cv2.imshow()` function
- **DoG, LoG and DoH**
  - `from skimage.feature import blob_dog, blob_log, blob_doh`

## # Simple blob detector

```
import cv2
```

```
import numpy as np;
```

```
# Read image
```

```
im = cv2.imread("blob.jpg", cv2.IMREAD_GRAYSCALE)
```

```
# Set up the detector with default parameters.
```

```
detector = cv2.SimpleBlobDetector()
```

```
# Detect blobs.
```

```
keypoints = detector.detect(im)
```

```
# Draw detected blobs as red circles.
```

```
# cv2.DRAW_MATCHES_FLAGS_DRAW_RICH_KEYPOINTS ensures the size of the circle corresponds to  
the size of blob
```

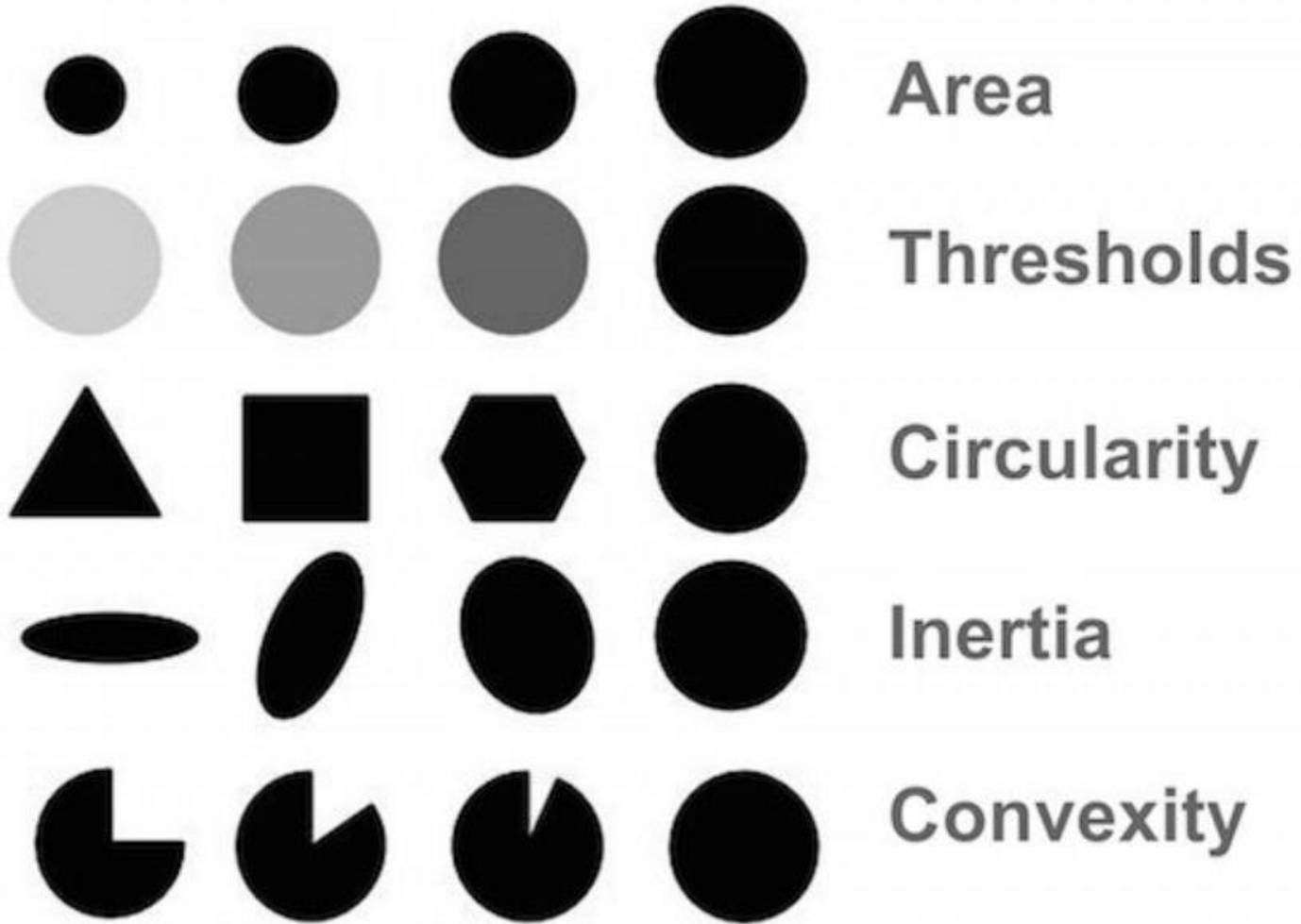
```
im_with_keypoints = cv2.drawKeypoints(im, keypoints, np.array([]), (0,0,255),  
cv2.DRAW_MATCHES_FLAGS_DRAW_RICH_KEYPOINTS)
```

```
# Show keypoints
```

```
cv2.imshow("Keypoints", im_with_keypoints)
```

```
cv2.waitKey(0)
```

# Parameters



## # Setup SimpleBlobDetector parameters.

```
params =  
cv2.SimpleBlobDetector_Params()      # Create a detector with the parameters  
  
# Change thresholds  
ver = (cv2.__version__).split('.')  
if int(ver[0]) < 3 :  
    detector = cv2.SimpleBlobDetector(params)  
else :  
    detector =  
cv2.SimpleBlobDetector_create(params)  
  
params.minThreshold = 10;  
params.maxThreshold = 200;  
  
# Filter by Area.  
params.filterByArea = True  
params.minArea = 1500  
  
# Filter by Convexity  
params.filterByConvexity = True  
params.minConvexity = 0.87  
  
# Filter by Inertia  
params.filterByInertia = True  
params.minInertiaRatio = 0.01
```

# Comparison

Method	Library	Best For	Speed	Notes
<code>cv2.SimpleBlobDetector()</code>	OpenCV	Simple round blobs	Fast	Good for basic blob detection
<code>blob_dog</code>	scikit-image	Multi-scale bright blobs	Medium	Uses Difference of Gaussians
<code>blob_log</code>	scikit-image	Round blobs	Slow	More accurate but computationally heavy
<code>blob_doh</code>	scikit-image	Elongated blobs	Medium	Best for corner-like or filament structures



# Refs

- <https://towardsdatascience.com/image-processing-blob-detection-204dc6428dd/>
- <https://github.com/AshishPandey88/Blob-Detection?tab=readme-ov-file>
- <https://cyrillugod.medium.com/blob-detection-in-action-7425638f5347>
- <https://mathematica.stackexchange.com/questions/310557/performant-blob-detection-in-images>
- <https://www.geeksforgeeks.org/find-circles-and-ellipses-in-an-image-using-opencv-python/>
- <https://www.geeksforgeeks.org/blob-detection-using-opencv/>

# Image quality enhancement

- Adjusting brightness and contrast
- Sharpening images
- Removing noise from images
- Enhancing color in images
- Image resizing and scaling
- Inverse Transform
- Equalizing histograms
- Super-resolution

# Adjusting brightness and contrast

- `image2 = cv2.addWeighted(image, contrast, np.zeros(image.shape, image.dtype), 0, brightness)`  
# Adjusts the brightness by adding x to each pixel value
  - `brightness = x`
- # Adjusts the contrast by scaling the pixel values
  - `contrast = y`
- **`cv2.convertScaleAbs()`** function adjust the brightness and contrast using a combination of **scaling** and **shifting** the pixel values.

- **Sharpening images**
  - `cv2.filter2D()` function **or**
  - `cv2.Laplacian()` calculates the Laplacian of an image.
  - **The Laplacian is a second-order derivative, meaning it measures the rate of change of the rate of change of pixel intensities.**
- **Removing noise from images**
  - `cv2.medianBlur()` function
  - `cv2.GaussianBlur()`
- **Enhancing color in images**
  - `cv2.cvtColor()` function after converting RGB to HSV and adjusts the hue, saturation, and value (brightness).
- **Image resizing and scaling**
  - `cv2.resize()` function with `cv2.INTER_NEAREST`, `cv2.INTER_LINEAR`, `cv2.INTER_CUBIC`, and others.

- **Inverse Transform**
  - inverse the color by simply subtracting each value from 255.
- **Equalizing histograms**
  - cv2.equalizeHist() function and reproduce the image back.
- **Super-resolution**
  - Use the pyrUp() and pyrDown() functions
- **Color correction**
  - cvtColor() and inRange() function. **To perform task like** color balance, color grading, and white balancing.