OpenCV

Computer vision → A process in which we can understand the image and videos and can be manipulated

> Basic

- Read
 - Image file → imread(path, flag)
 - Video file
 - Camera image
 - Camera video
- Show → imshow(name, image)
- Save → imwrite(path, image)

> Noise

Random variations in brightness or color can be present in an image.

Gaussian Noise

Random distribution of pixel values, followed by Gaussian normal distribution

Salt-and-Pepper Noise

Randomly occurring bright and dark pixels

Speckle Noise

Caused by variations in brightness or color due to random fluctuations in the image.

Quantization Noise

The image is represented with a limited number of intensity levels

- > Geometrical shapes
 - o Circle
 - Rectangle → cv2.rectangle(img, start_point, end_point, color,thickness)
 - o line
- > Channels
 - Split → cv2.split(img)
 - Merge
 - Resize → cv2.resize(img, (new_width, new_height))
- ➤ Color spaces

cv2.cvtColor(image, color code)

- cv2.COLOR_BGR2RGB
- o cv2.COLOR_BGR2HSV
- cv2.COLOR_BGR2LAB → A perceptually uniform color space
- cv2.COLOR_BGR2GRAY
- > HSV color space

- o Hue, Saturation, Value
- Hue → type of color, range (0, 179)
- Saturation → intensity of color, range (0 to 255)
- Value → Brightness of color , range(0 to 255)
 Use cases
- Color filtering → Color-based object detection
- Image thresholding → segmenting images

> Bitwise operators

- AND → pixel-wise and operation of 2 images, useful in masking. The result is non-zero only both values are non-zero
- OR → useful in combining images. The result is nonzero if at least one value is non-zero
- NOT → perform on a single image. Invert the pixel values
- XOR → useful to highlight differences between 2 images. Result non-zero only if both values are different

> Simple thresholding

- o Advantages
- Disadvantages
- o Thresholding is the way to convert an image to a binary image.
- _, binary_image = cv2.threshold(img, threshold value, max, type: cv2.thresh_binary)

> Adaptive image thresholding

- The image will be divided into various regions and the thresholding
- Threshold = cv2.adaptiveThresholding(img, max_val, adaptive method, binarization, size of neighbour, constant subtract)
- Thresh = cv2.adaptiveThreshoding(img, 255, cv2.adaptive_thresh_mean_c, cv2.thresh_binary, 11, 2)

> Bind trackbars

 GUI component that allow interactively change a parameter code

```
def update_val(x):
```

Blah blah

cv2.createTrackbar("threshod", "img window", 0, 255, update_val) cv2.setTrackbarPos("threshold", "img window", 128)

> Smoothing

- o Advantages
- Disadvantages
- Average blurring
 - Each pixel in image is the average of its neighbours
 - cv2.blur(original_img, (ksize, ksize))
- o Gaussian blur
 - Reduce the high frequency noises
 - cv2.GaussianBlur(img, (ksize, ksize), sigma)

- Median Blurring
 - Replace each cell with the median value of its neighbours
 - cv2.medianBlur(original_size, ksize)
- Bilateral filtering
 - Preserves edge while smoothing the image
- Smoothing in coloured images → split, blur, merge
- > Morphological transformations

Operations based on the shape of the image, are only applicable in the binary image

- Use cases
 - Noise Reduction
 - Object detection and segmentation
 - Shape and size adjustment
 - Boundary extraction
- Erosion
 - Expands the white space kernel = np.ones((5, 5), np.uint8) erosion = cv2.erode(image, kernel, iterations=1)
- Dilation
 - Expand the dark part in an image dilation = cv2.dilate(image, kernel, iterations=1)
- Opening
 - Erosion followed by dilation cv2.morphologyEx(image, cv2.MORPH_OPEN, kernel)
- Closing
 - Dilation followed by erosion cv2.morphologyEx(img, cv2.MORPH_CLOSE, kernel)
- > Image gradient
 - Rate of change of intensity in different directions
 - Types
 - Sobel operator
 - Computes the gradient using convolution with sobel kernels in both horizontal and vertical direction
 - cv2.Sobel(image, ddepth, dx, dy, ksize)
- > Edge detection
 - Identifying the boundaries in an image
 - Use cases
 - Object detection
 - Image segmentation
 - Feature extraction
- > Canny edge detection
 - Multistage algorithm that combines gradient calculation, non-maximum suppression, and edge tracking by hysteresis
 - cv2.Canny(img, lower_thresh, upper_thresh)

- Algorithm
 - Image Smoothing → grayscale, gaussian filter
 - Finding Gradients \rightarrow Sobel x and y combining
 - Non-Max Suppression → in each kernel, if the pixel is local maximum, keep it otherwise remove it.
 - Double Thresholding → lower val, upper val
 - Edge tracking by hysteresis
- > Mouse events
 - Listen to the events in the mouse
 - cv2.setMouseCallback("image", mouse_callback)
 def mouse_callback(event, x, y, flags, param)
- > Image pyramids
 - Gaussian image pyramid
 - Laplacian pyramid
 - Application
 - Image compression

Reduce the resolution

Multi-scale processing

Perform operations on different scales

Image blending

Combine 2 images seamlessly by blending them at different resolutions

> Contours

- o It is the boundaries of an object in an image
- Finding the contours
 - contours, _ = cv2.findContours(binary_img, cv2.RETR_EXTERNAL, cv2.CHAIN_APPROX_SIMPLE)
- Draw contours
 - cv2.drawContours(img, contours, -1, (0, 255, 0), 2)
- Contour properties

for contour in contours:

Area = cv2.contourArea(contour)

Perimeter = cv2.arcLength(contour, True)

- > Template matching
 - Used to find a sub-image a template in an image
 - res = cv2.matchTemplate(img, template_img, method=cv2.CCOEFF_NORMED)
 min_val, max_val, min_loc, max_loc = cv2.minMaxLoc(res)