

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

- Circle
- Rectangle → `cv2.rectangle(img, start_point, end_point, color, thickness)`
- 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`
- `cv2.COLOR_BGR2HSV`
- `cv2.COLOR_BGR2LAB` → A perceptually uniform color space
- `cv2.COLOR_BGR2GRAY`

➤ HSV color space

- 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

- Advantages
- Disadvantages
- 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.adaptiveThresholding(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

- Advantages
- Disadvantages
- Average blurring
 - Each pixel in image is the average of its neighbours
 - `cv2.blur(original_img, (ksize, ksize))`
- 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* → 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*
 - *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)`