

Work 1.1 – Reading and showing an image

OpenCV methods: imread(), namedWindow(), imshow(), waitKey()

Work 1.2 – Reading and showing a video or images from a camera;

OpenCV methods: VideoCapture(); VideoCapture .get(); VideoCapture .read();

Work 1.3 – Image resize with a finger print image OpenCV method: *resize()*.

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Work 2.1 – *Cromakey* (blue screening)

 $ImageOut = objectImage \times Mask + background \times !Mask$







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OpenCV methods: add(); multiply();

Work 2.2 – Image smoothing and median filtering

OpenCV methods: blur(); medianBlur();

Work 2.3 – Edge detector

OpenCV methods: cvtColor(); Sobel(); Canny(); Laplacian();



Work 3.1 – Image Histogram

OpenCV method: calcHist();

Matplotlib (python 2D plotting library) method: bar(); show();

Work 3.2 – Image Binarization

OpenCV method: threshold();

Work 3.3 – Morphological Operators

OpenCV method: getStructuringElement(); morphologyEx(); dilate(); erode().



Work 4.1 – Labeling

Local toolbox: bwLabel and psColor

Used methods: bwLabel.labeling(); psColor. CreateColorMap(); psColor. Gray2PseudoColor();

OpenCV 2.X methods: findContours(); drawContours();

OpenCV 3.X method: connectedComponents()

Work 4.2 – Feature Extraction;

OpenCV 2.X methods: contourArea(); moments(); arcLength(); boundingRect()

OpenCV 3.X method: connectedComponentsWithStats()

Work 4.3 – Classification;



Work 5.1 – Image Gradient

5.1.1 – Compute the modulus and phase of the image gradient based on a differential operator, for example, *Sobel* (work 2.3);

5.1.2 – Based on the modulus of the gradient, compute a contour image (thresholding, work 3.2);

Compare the results with the Canny algorithm.

Work 5.2 – Color Feature Extraction

5.2.1 – Convert an image in RGB format to the HSI color space and visualize each component;

5.2.2 – Compute a color histogram.

Compare these features between several images.

Workout 5 (cont.)



Work 5.3 – Texture Feature Extraction

5.3.1 – Compute the edge density (*edgeness*);

5.3.2 – Compute the normalized histogram of magnitudes and orientations of contours.

Compare these features between several images.

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Work 6 – Motion Detection

Input: Two monochrome images $I_n(r,c)$ and $I_{n-k}(r,c)$ or $I_n(r,c)$ and $B_n(r,c)$ and a threshold τ ;

Output: binary image, I_{out} and a set of bounding boxes, B, with the location of detected objects;

5 - Step Algorithm:

1. Compute the binary image (active pixels)

$$I_{out}(r,c) = \begin{cases} 1 & \text{if } |I_n(r,c) - I_{n-k}(r,c)| > \tau \\ 0 & \text{otherwise} \end{cases}$$

- 2. Perform a morphological closing of I_{out} using a small disc;
- 3. Perform connect components extraction;
- 4. Remove regions with small area (noise);
- 5. For each remaining active region, compute the bounding box;



Work 7 – Sparse Motion Field Detection

Input: Two monochrome images $I_n(r,c)$ and $I_{n-k}(r,c)$;

Output: Set of interesting points and their corresponding motion vectors;

Algorithm:

1. Compute a set of interesting points in the previous image, $I_{n-k}(r,c)$, for example, *corners*;

OpenCV method: goodFeaturesToTrack();

2. Determine the correspondence of these points in the current image, $I_n(r,c)$;

OpenCV method: calcOpticalFlowPyrLK();

3. Determine the motion vectors and represent them graphically; Matplotlib method: *quiver()*.



Work 8.1 – K-means Color Segmentation

Use the OpenCV function *kmeans()* to perform color segmentation, where each pixel is represented by its *RGB* components (use also other color space, for example, HSI).

Work 8.2 – Circle Detection with Hough Transform

Use the OpenCV function *HoughCircles()* to detect circular objects, for example, *coins*.

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