

## *Exercise 5*

# ***Fundamentals of machine vision algorithms***

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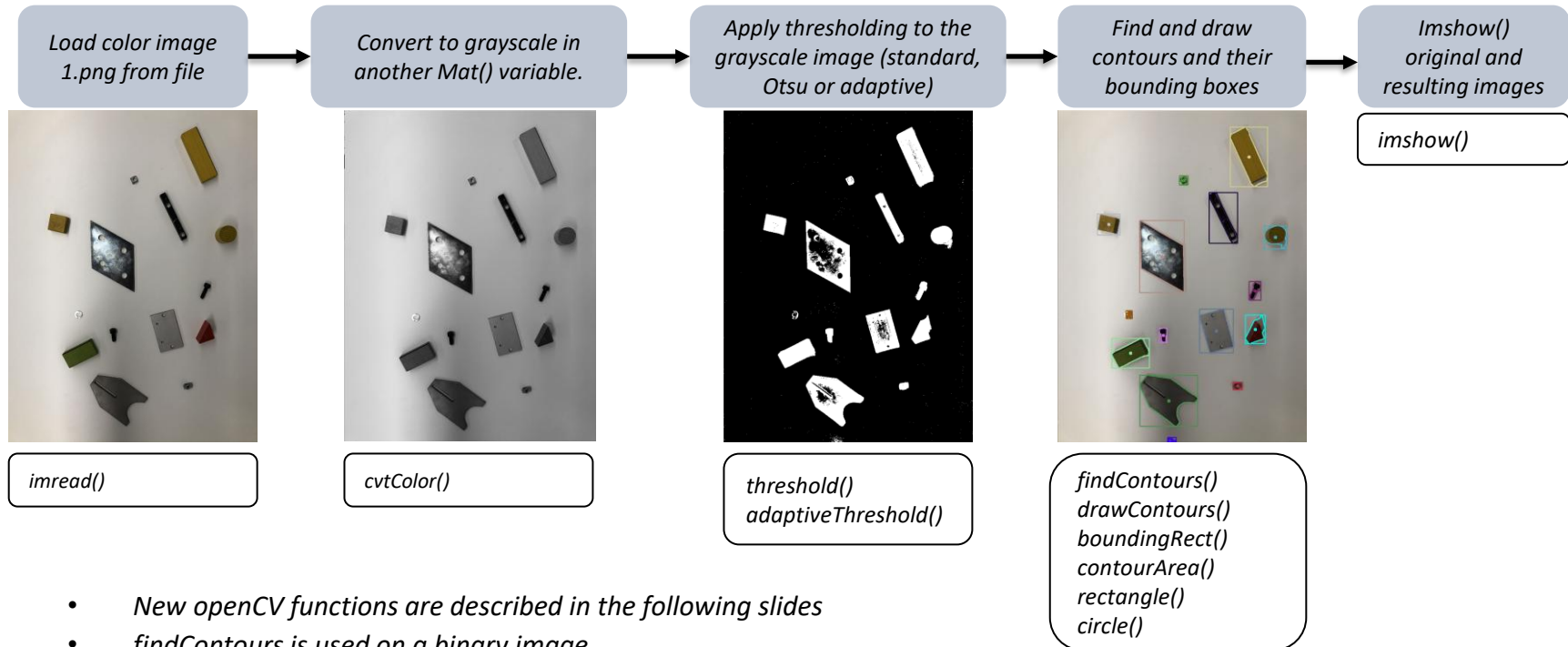
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# Practical task 1

*Object localization. Generate contours around thresholded parts:*



- *New openCV functions are described in the following slides*
- *findContours is used on a binary image*
- *Image processing prior to contour extraction is crucial for good results*
  - *Thresholding method and parameters should be found that give best possible results*
  - *Small particles can be removed with morphological operations (opening) – Practical task 2*
- *Shadows and light reflection can deteriorate localization results (notice reflective surfaces)*
- *Contours can be filtered based on area size – Practical task 3*
- *Objects can be filtered based on color – Practical task 4*

[https://docs.opencv.org/master/d3/dc0/group\\_imgproc\\_shape.html#gadf1ad6a0b82947fa1fe3c3d497f260e0](https://docs.opencv.org/master/d3/dc0/group_imgproc_shape.html#gadf1ad6a0b82947fa1fe3c3d497f260e0)

# Practical task 1

## findContours()

[https://docs.opencv.org/master/d3/dc0/group\\_imgproc\\_shape.html#gadf1ad6a0b82947fa1fe3c3d497f260e0](https://docs.opencv.org/master/d3/dc0/group_imgproc_shape.html#gadf1ad6a0b82947fa1fe3c3d497f260e0)

### ◆ findContours() [1/2]

```
void cv::findContours ( InputArray      image,
                       OutputArrayOfArrays contours,
                       OutputArray      hierarchy,
                       int              mode,
                       int              method,
                       Point            offset = Point() )
```

#### Python:

```
cv.findContours( image, mode, method[, contours[, hierarchy[, offset]]] ) -> contours, hierarchy
```

```
#include <opencv2/imgproc.hpp>
```

Finds contours in a binary image.

The function retrieves contours from the binary image using the algorithm [233]. The contours are a useful tool for shape analysis and object detection and recognition. See squares.cpp in the OpenCV sample directory.

#### Note

Since opencv 3.2 source image is not modified by this function.

#### Parameters

- image** Source, an 8-bit single-channel image. Non-zero pixels are treated as 1's. Zero pixels remain 0's, so the image is treated as binary. You can use `compare`, `inRange`, `threshold`, `adaptiveThreshold`, `Canny`, and others to create a binary image out of a grayscale or color one. If mode equals to `RETR_CCOMP` or `RETR_FLOODFILL`, the input can also be a 32-bit integer image of labels (CV\_32SC1).
- contours** Detected contours. Each contour is stored as a vector of points (e.g. `std::vector<std::vector<cv::Point>>`).
- hierarchy** Optional output vector (e.g. `std::vector<cv::Vec4i>`), containing information about the image topology. It has as many elements as the number of contours. For each *i*-th contour `contours[i]`, the elements `hierarchy[i][0]`, `hierarchy[i][1]`, `hierarchy[i][2]`, and `hierarchy[i][3]` are set to 0-based indices in contours of the next and previous contours at the same hierarchical level, the first child contour and the parent contour, respectively. If for the contour *i* there are no next, previous, parent, or nested contours, the corresponding elements of `hierarchy[i]` will be negative.
- mode** Contour retrieval mode, see [RetrievalModes](#)
- method** Contour approximation method, see [ContourApproximationModes](#)
- offset** Optional offset by which every contour point is shifted. This is useful if the contours are extracted from the image ROI and then they should be analyzed in the whole image context.

[https://docs.opencv.org/master/d3/dc0/group\\_imgproc\\_shape.html#ga819779b9857cc2f8601e6526a3a5bc71](https://docs.opencv.org/master/d3/dc0/group_imgproc_shape.html#ga819779b9857cc2f8601e6526a3a5bc71)

#### Enumerator

RETR_EXTERNAL
Python: cv.RETR_EXTERNAL
RETR_LIST
Python: cv.RETR_LIST
RETR_CCOMP
Python: cv.RETR_CCOMP
RETR_TREE
Python: cv.RETR_TREE
RETR_FLOODFILL
Python: cv.RETR_FLOODFILL

CHAIN\_APPROX\_NONE  
Python: cv.CHAIN\_APPROX\_NONE

CHAIN\_APPROX\_SIMPLE  
Python: cv.CHAIN\_APPROX\_SIMPLE

CHAIN\_APPROX\_TC89\_L1  
Python: cv.CHAIN\_APPROX\_TC89\_L1

CHAIN\_APPROX\_TC89\_KCOS  
Python: cv.CHAIN\_APPROX\_TC89\_KCOS



# Practical task 1

*drawContours(), boundingRect(), rectangle(), circle()*

## ◆ drawContours()

[https://docs.opencv.org/master/d6/d6e/group\\_imgproc\\_draw.html#ga746c0625f1781f1ffc9056259103edbc](https://docs.opencv.org/master/d6/d6e/group_imgproc_draw.html#ga746c0625f1781f1ffc9056259103edbc)

```
void cv::drawContours ( InputOutputArray image,
    InputArrayOfArrays contours,
    int contourIdx,
    const Scalar & color,
    int thickness = 1,
    int lineType = LINE_8,
    InputArray hierarchy = noArray(),
    int maxLevel = INT_MAX,
    Point offset = Point()
)
```

Diagram illustrating the parameters for `drawContours()`:

- `contours` (InputArrayOfArrays) is mapped to `vector<vector<Point>> contours;`
- `contourIdx` (int) is mapped to `Contour number (0....X)`
- `color` (const Scalar &) is mapped to `Scalar (...,...)`

Python:

```
cv.drawContours( image, contours, contourIdx, color[, thickness[, lineType[, hierarchy[, maxLevel[, offset]]]]) -> image
```

## ◆ boundingRect()

`Rect cv::RotatedRect::boundingRect ( ) const`

returns the minimal up-right integer rectangle containing the rotated rectangle → `Rect bounding = boundingRect(contours.at(i)); //creates bounding rectangle around contour "i"`

```
void rectangle(InputOutputArray img, Rect rec, const Scalar &color, int thickness = 1, int lineType = LINE_8, int shift = 0)
rectangle(rectangle1, Rect(20,20,160,160), Scalar(255), -1, 8, 0);

void circle(InputOutputArray img, Point center, int radius, const Scalar &color, int thickness = 1, int lineType = LINE_8, int shift = 0)
circle(circle1, cv::Point(100,100), 80, Scalar(255), -1, 8, 0)
```

**Example code for drawing random color contours and their bounding boxes in the color image:**

```
RNG rng(12345); //range for random number generation
for (size_t i = 0; i < contours.size(); i++) //for each contour
{
    Scalar color = Scalar(rng.uniform(0, 256), rng.uniform(0, 256), rng.uniform(0, 256)); //generate random color
    drawContours(image, contours, (int)i, color, 2, LINE_8); //draw contour "i"
    Rect bounding = boundingRect(contours.at(i)); //creates bounding rectangle around contour "i"
    rectangle(image, bounding, color, 2, LINE_8, 0); //draw bounding rectangle
    circle(image, Point(bounding.x+bounding.width/2, bounding.y+bounding.height/2), 5, color, -1, LINE_8, 0); // draw center of bounding rectange
}
```



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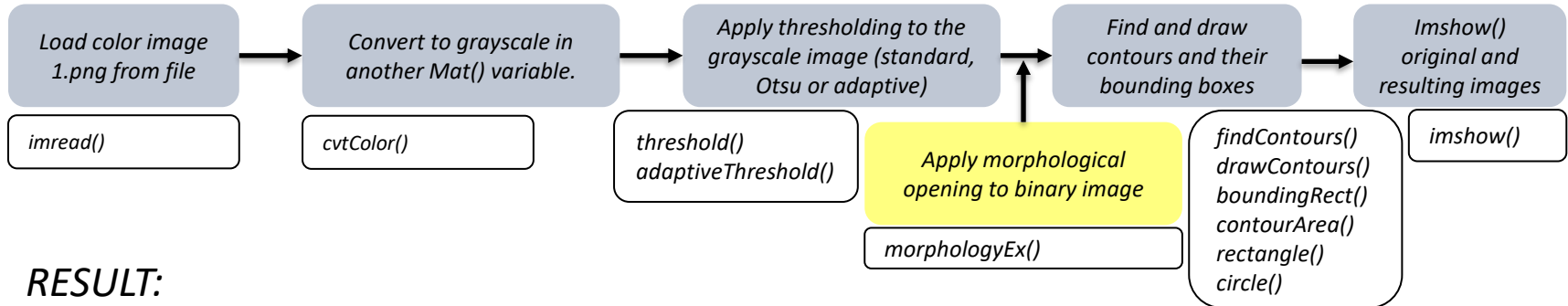


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# Practical task 2

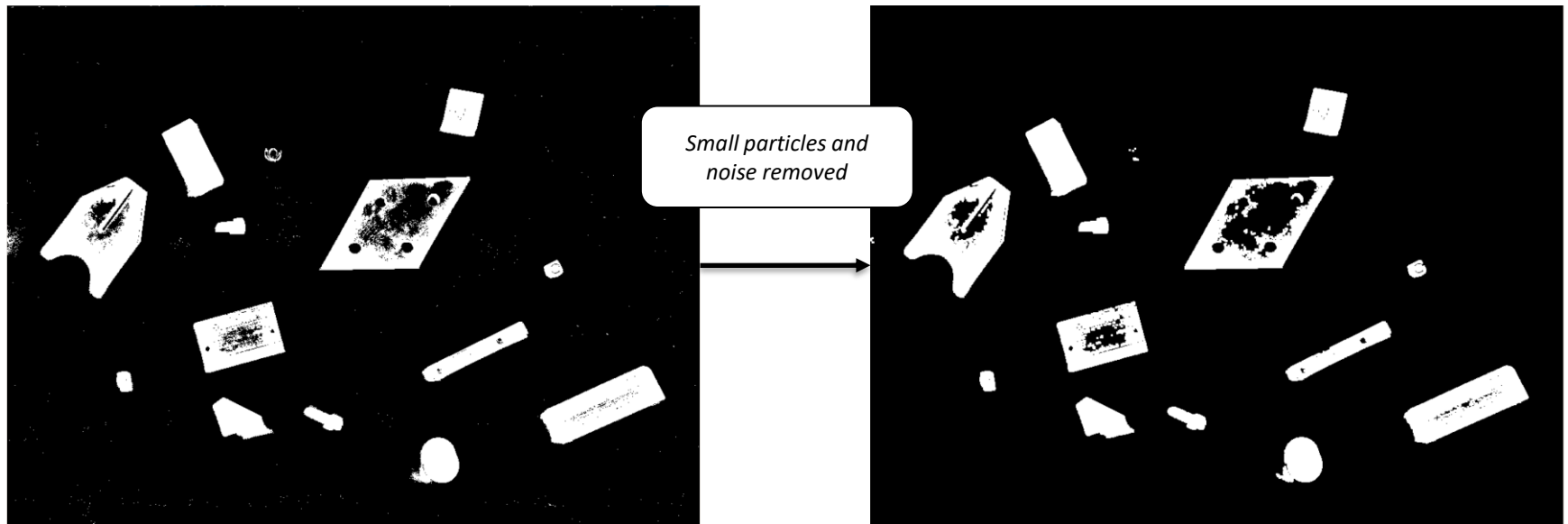
Upgrade Practical task 1 by removing small particles or parts using:

Morphological opening (use code from previous excersises)



## RESULT:

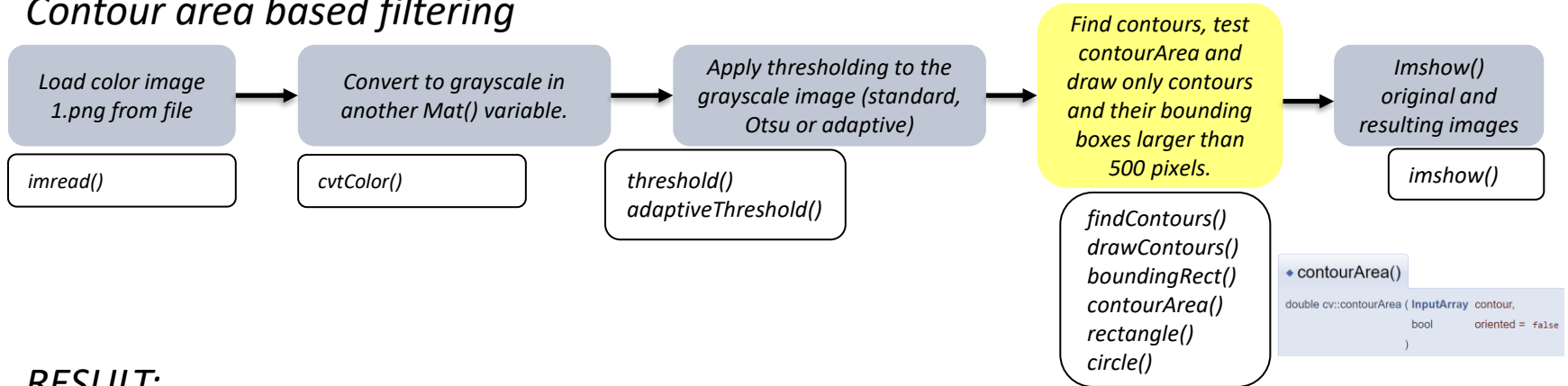
```
adaptiveThreshold(image_gray, image_binary, 255,  
ADAPTIVE_THRESH_GAUSSIAN_C, THRESH_BINARY_INV, 131, 5);
```



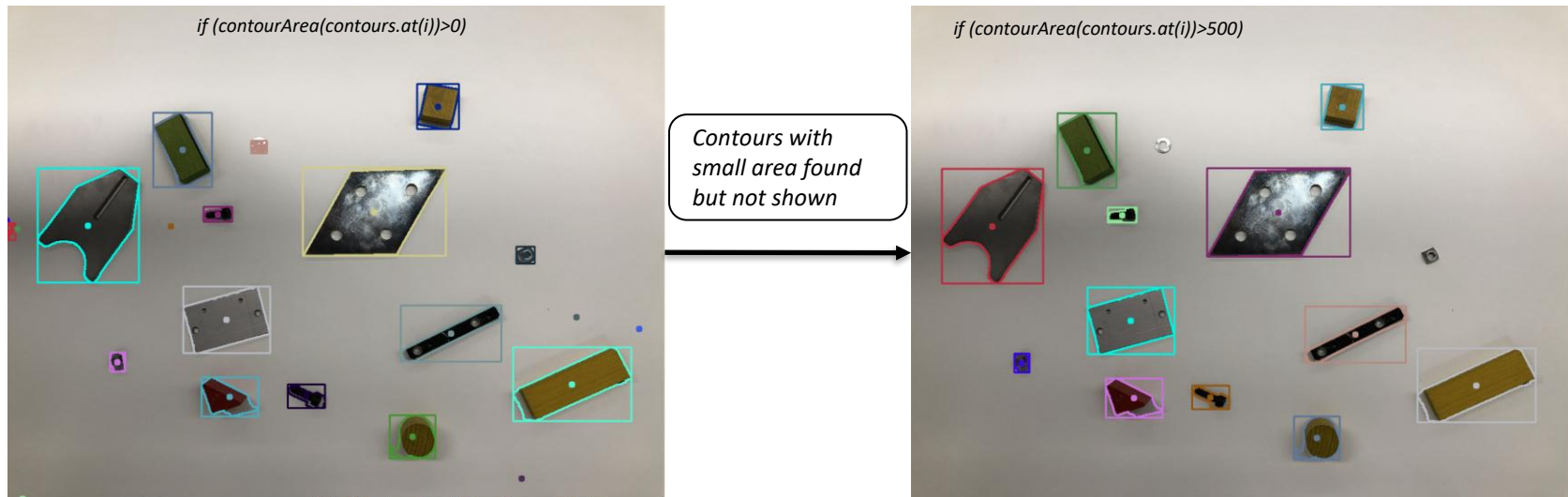
# Practical task 3

Upgrade Practical task 1 by removing small particles or parts using:

## Contour area based filtering



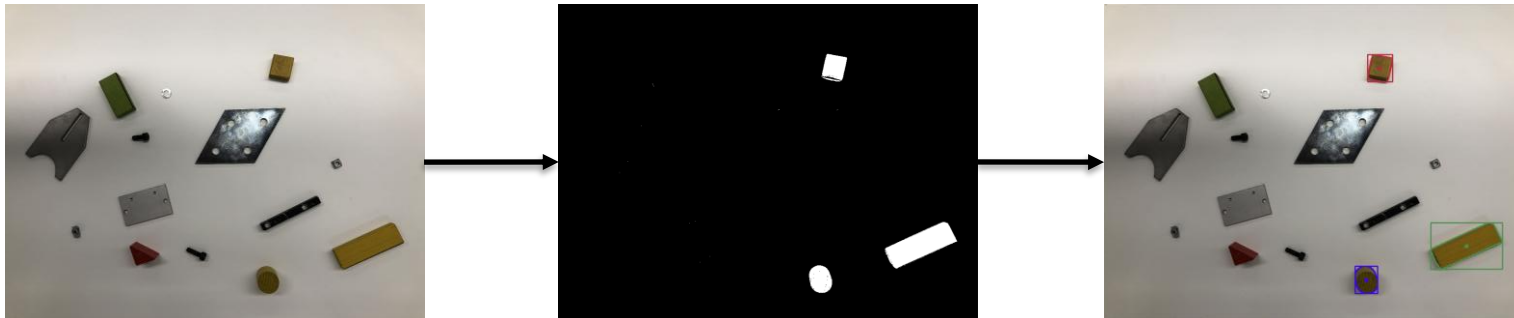
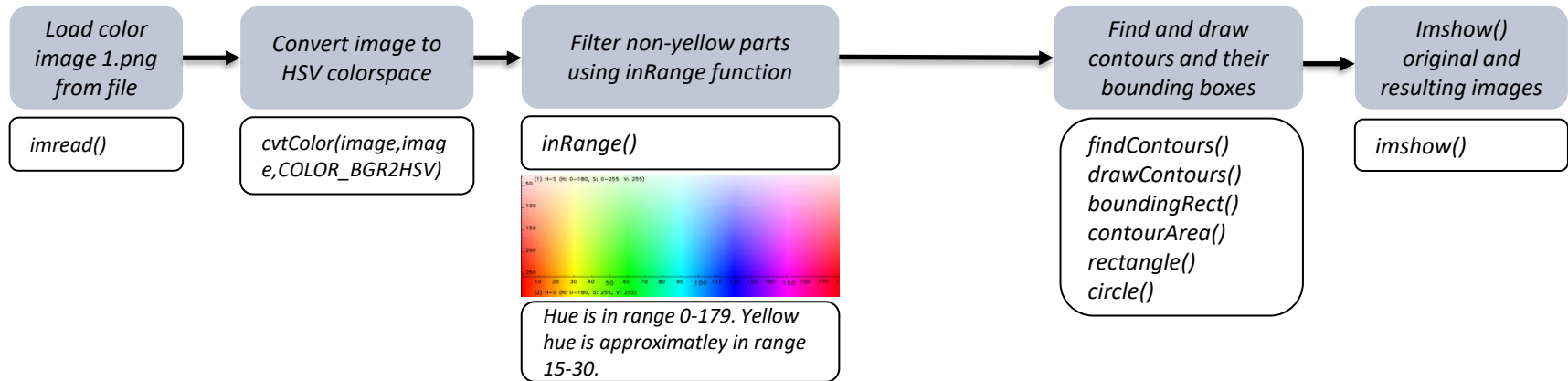
RESULT:





# Practical task 4

Upgrade Practical task 1 by converting the image to HSV color space and creating a binary image based on color of interest:



# Practical task 4

## *inRange()*

### ◆ inRange()

```
void cv::inRange ( InputArray  src,  
                  InputArray  lowerb,  
                  InputArray  upperb,  
                  OutputArray dst  
                  )
```

Scalar (0-179, 0-255, 0-255)

Scalar (0-179, 0-255, 0-255)

#### Python:

```
cv.inRange( src, lowerb, upperb[, dst] ) -> dst
```

```
#include <opencv2/core.hpp>
```

Checks if array elements lie between the elements of two other arrays.

The function checks the range as follows:

- For every element of a single-channel input array:

$$\text{dst}(l) = \text{lowerb}(l)_0 \leq \text{src}(l)_0 \leq \text{upperb}(l)_0$$

- For two-channel arrays:

$$\text{dst}(l) = \text{lowerb}(l)_0 \leq \text{src}(l)_0 \leq \text{upperb}(l)_0 \text{ \& \& } \text{lowerb}(l)_1 \leq \text{src}(l)_1 \leq \text{upperb}(l)_1$$

- and so forth.

That is, dst (l) is set to 255 (all 1 -bits) if src (l) is within the specified 1D, 2D, 3D, ... box and 0 otherwise.

When the lower and/or upper boundary parameters are scalars, the indexes (l) at lowerb and upperb in the above formulas should be omitted.

#### Parameters

**src** first input array.  
**lowerb** inclusive lower boundary array or a scalar.  
**upperb** inclusive upper boundary array or a scalar.  
**dst** output array of the same size as src and CV\_8U type.

#### Example:

```
inRange(image_hsv,Scalar(15,50,50),Scalar(30,255,255),image_binary); //yellow objects only
```

[https://docs.opencv.org/3.4/d2/de8/group\\_core\\_array.html#gq48af0ab51e36436c5d04340e036ce981](https://docs.opencv.org/3.4/d2/de8/group_core_array.html#gq48af0ab51e36436c5d04340e036ce981)



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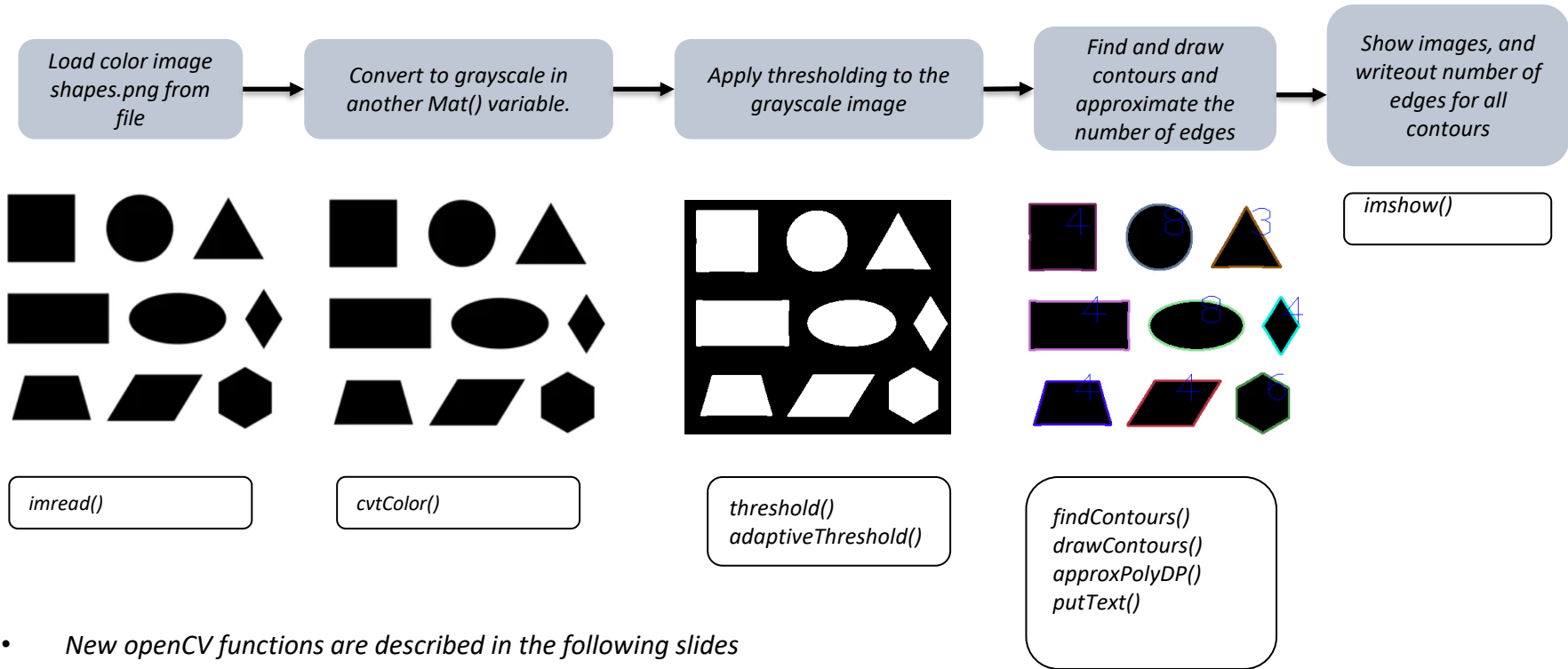


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# Practical task 5

## Shape detection



- New openCV functions are described in the following slides
- `findContours` is used on a binary image
- Contours can be filtered based on area size
- `approxPolyDP()` uses Douglas-Peucker algorithm to approximate a curve or a polygon with another curve/polygon with less vertices
- Writeout the number of edges next to any shape

# Practical task 5

## *approxPolyDP()*

[https://docs.opencv.org/3.4/d3/dc0/group\\_imgproc\\_shape.html#ga0012a5fdae70b8a9970165d98722b4c](https://docs.opencv.org/3.4/d3/dc0/group_imgproc_shape.html#ga0012a5fdae70b8a9970165d98722b4c)

[https://docs.opencv.org/master/d3/dc0/group\\_imgproc\\_shape.html#ga819779b9857cc2f8601e6526a3a5bc71](https://docs.opencv.org/master/d3/dc0/group_imgproc_shape.html#ga819779b9857cc2f8601e6526a3a5bc71)

### ◆ approxPolyDP()

```
void cv::approxPolyDP ( InputArray  curve,  
                        OutputArray approxCurve,  
                        double      epsilon,  
                        bool        closed  
                      )
```

→ A member `contours[i]` in a `vector<vector<Point>> contours`

→ Can be related to contour arc length:  
`arcLength(contours[i], true)*0.02`

→ `vector<Point> approx;`

### Python:

```
cv.approxPolyDP( curve, epsilon, closed[, approxCurve] ) -> approxCurve
```

```
#include <opencv2/imgproc.hpp>
```

Approximates a polygonal curve(s) with the specified precision.

The function `cv::approxPolyDP` approximates a curve or a polygon with another curve/polygon with less vertices so that the distance between them is less or equal to the specified precision. It uses the Douglas-Peucker algorithm [http://en.wikipedia.org/wiki/Ramer-Douglas-Peucker\\_algorithm](http://en.wikipedia.org/wiki/Ramer-Douglas-Peucker_algorithm)

### Parameters

- curve** Input vector of a 2D point stored in `std::vector` or `Mat`
- approxCurve** Result of the approximation. The type should match the type of the input curve.
- epsilon** Parameter specifying the approximation accuracy. This is the maximum distance between the original curve and its approximation.
- closed** If true, the approximated curve is closed (its first and last vertices are connected). Otherwise, it is not closed.

### Examples:

`samples/cpp/contours2.cpp`, `samples/cpp/squares.cpp`, and `samples/tapi/squares.cpp`.



# Practical task 5

## *drawContours(), putText()*

### ◆ drawContours()

[https://docs.opencv.org/master/d6/d6e/group\\_imgproc\\_draw.html#ga746c0625f1781f1ffc9056259103edbc](https://docs.opencv.org/master/d6/d6e/group_imgproc_draw.html#ga746c0625f1781f1ffc9056259103edbc)

```
void cv::drawContours ( InputOutputArray image,
                      InputArrayOfArrays contours,
                      int contourIdx,
                      const Scalar & color,
                      int thickness = 1,
                      int lineType = LINE_8,
                      InputArray hierarchy = noArray(),
                      int maxLevel = INT_MAX,
                      Point offset = Point()
                      )
```

Annotations:

- `contours` → `vector<vector<Point>> contours;`
- `contourIdx` → `Contour number (0....X)`
- `color` → `Scalar (...,...,...)`

#### Python:

```
cv.drawContours( image, contours, contourIdx, color[, thickness[, lineType[, hierarchy[, maxLevel[, offset]]]] ] -> image
```

### ◆ putText()

```
void cv::putText ( InputOutputArray img,
                  const String & text,
                  Point org,
                  int fontFace,
                  double fontScale,
                  Scalar color,
                  int thickness = 1,
                  int lineType = LINE_8,
                  bool bottomLeftOrigin = false
                  )
```

#### Python:

```
cv.putText( img, text, org, fontFace, fontScale, color[, thickness[, lineType[, bottomLeftOrigin]] ] -> img
```

```
#include <opencv2/imgproc.hpp>
```

Draws a text string.

The function `cv::putText` renders the specified text string in the image. Symbols that cannot be rendered using the specified font are replaced by question marks. See [getTextSize](#) for a text rendering code example.

#### Parameters

<code>img</code>	Image.
<code>text</code>	Text string to be drawn.
<code>org</code>	Bottom-left corner of the text string in the image.
<code>fontFace</code>	Font type, see <a href="#">HersheyFonts</a> .
<code>fontScale</code>	Font scale factor that is multiplied by the font-specific base size.
<code>color</code>	Text color.
<code>thickness</code>	Thickness of the lines used to draw a text.
<code>lineType</code>	Line type. See <a href="#">LineTypes</a>
<code>bottomLeftOrigin</code>	When true, the image data origin is at the bottom-left corner. Otherwise, it is at the top-left corner.

#### Example:

```
putText(image,"text",Point2i(100,100),1,4,Scalar(255,0,0),1,LINE_8);
```



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# Practical task 5

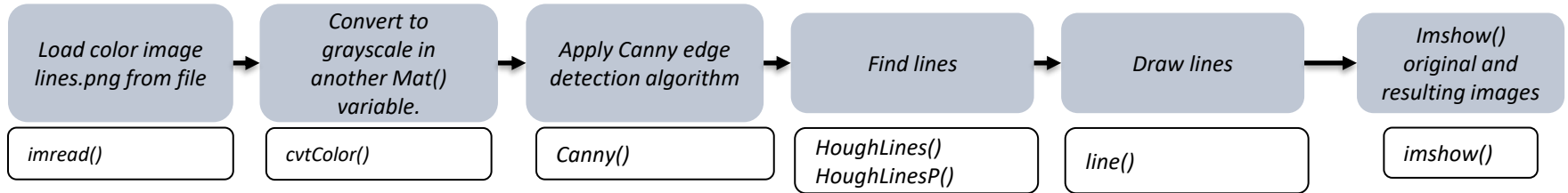
Use basic code from prior exercise to get a faster start:

```
int main(){
    Mat image=imread("1.png",1);
    Mat image_gray=imread("1.png",0);
    Mat image_binary,image_hsv;
    //threshold(image_gray,image_binary,100,255,THRESH_BINARY_INV+THRESH_OTSU);
    //threshold(image_gray,image_binary,100,255,THRESH_BINARY_INV);
    adaptiveThreshold(image_gray, image_binary, 255, ADAPTIVE_THRESH_GAUSSIAN_C, THRESH_BINARY_INV, 131, 5);
    vector<vector<Point> > contours;
    findContours(image_binary,contours,RETR_EXTERNAL,1);
    RNG rng(12345); //range for random number generation
    for (size_t i = 0; i < contours.size(); i++)//for each contour
    {
        if (contourArea(contours.at(i))>100) //filter for removing small contours
        {
            Scalar color = Scalar(rng.uniform(0, 256), rng.uniform(0, 256), rng.uniform(0, 256)); //generate random color
            drawContours(image, contours, (int)i, color, 2, LINE_8);//draw contour "i"
            Rect bounding = boundingRect(contours.at(i)); //creates bounding rectangle around contour "i"

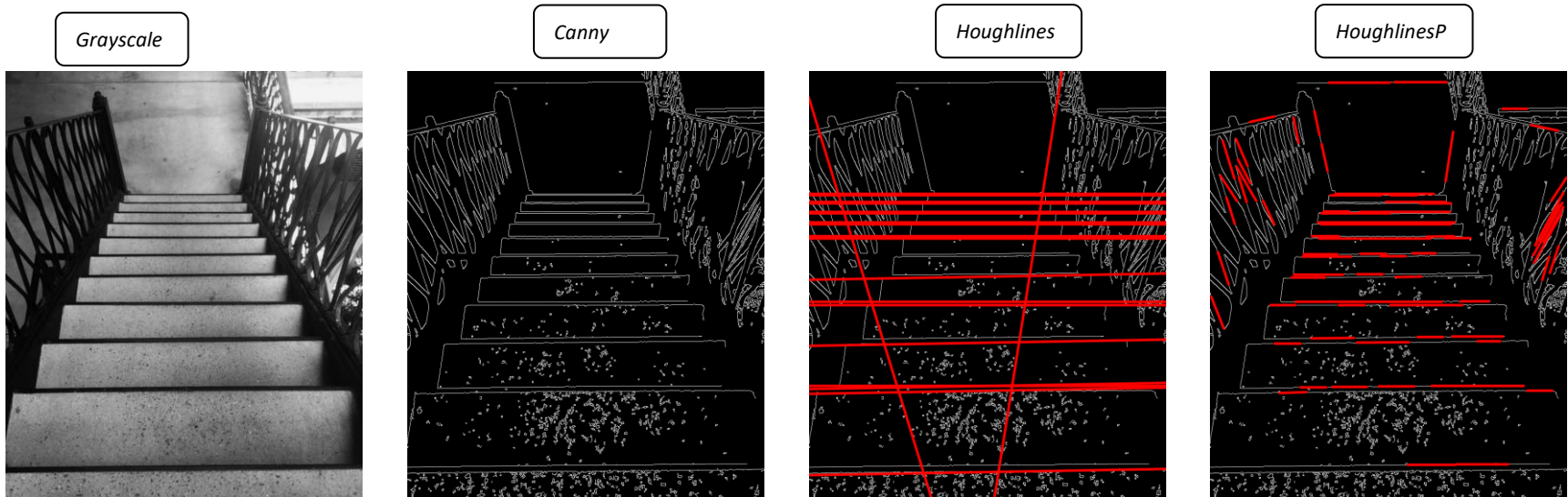
            // approximate contour with accuracy proportional to the contour perimeter
        }
    }
    imshow("Color", image);
    imshow("Gray", image_gray);
    imshow("Binary", image_binary);
    waitKey(0);
    return 1;
}
```

# Practical task 6

## Detect lines



## RESULT:



# Practical task 6

## HoughLines(), HoughLinesP()

[https://docs.opencv.org/3.4/dd/d1a/group\\_imgproc\\_feature.html#ga46b4e588934f6c8dfd509cc6e0e4545a](https://docs.opencv.org/3.4/dd/d1a/group_imgproc_feature.html#ga46b4e588934f6c8dfd509cc6e0e4545a)

### ◆ HoughLines()

```
void cv::HoughLines ( InputArray  image,
                      OutputArray lines,
                      double      rho,
                      double      theta,
                      int          threshold,
                      double      srn = 0 ,
                      double      stn = 0 ,
                      double      min_theta = 0 ,
                      double      max_theta = CV_PI
                    )
```

#### Python:

```
cv.HoughLines( image, rho, theta, threshold[, lines[, srn[, stn[, min_theta[, max_theta]]]] ] -> lines
```

```
#include <opencv2/imgproc.hpp>
```

Finds lines in a binary image using the standard Hough transform.

The function implements the standard or standard multi-scale Hough transform algorithm for line detection. See <http://homepages.inf.ed.ac.uk/rbf/HIPR2/hough.htm> for a good explanation of Hough transform.

#### Parameters

**image** 8-bit, single-channel binary source image. The image may be modified by the function.

**lines** Output vector of lines. Each line is represented by a 2 or 3 element vector  $(\rho, \theta)$  or  $(\rho, \theta, \text{votes})$ .  $\rho$  is the distance from the coordinate origin  $(0, 0)$  (top-left corner of the image).  $\theta$  is the line rotation angle in radians ( $0 \sim \text{vertical line}, \pi/2 \sim \text{horizontal line}$ ). votes is the value of accumulator.

**rho** Distance resolution of the accumulator in pixels.

**theta** Angle resolution of the accumulator in radians.

**threshold** Accumulator threshold parameter. Only those lines are returned that get enough votes ( $> \text{threshold}$ ).

**srn** For the multi-scale Hough transform, it is a divisor for the distance resolution rho. The coarse accumulator distance resolution is rho and the accurate accumulator resolution is rho/srn. If both srn=0 and stn=0, the classical Hough transform is used. Otherwise, both these parameters should be positive.

**stn** For the multi-scale Hough transform, it is a divisor for the distance resolution theta.

**min\_theta** For standard and multi-scale Hough transform, minimum angle to check for lines. Must fall between 0 and max\_theta.

**max\_theta** For standard and multi-scale Hough transform, maximum angle to check for lines. Must fall between min\_theta and CV\_PI.

### ◆ HoughLinesP()

```
void cv::HoughLinesP ( InputArray  image,
                      OutputArray lines,
                      double      rho,
                      double      theta,
                      int          threshold,
                      double      minLineLength = 0 ,
                      double      maxLineGap = 0
                    )
```

#### Python:

```
cv.HoughLinesP( image, rho, theta, threshold[, lines[, minLineLength[, maxLineGap]] ] -> lines
```

```
#include <opencv2/imgproc.hpp>
```

Finds line segments in a binary image using the probabilistic Hough transform.

The function implements the probabilistic Hough transform algorithm for line detection, described in [147]

#### Parameters

**image** 8-bit, single-channel binary source image. The image may be modified by the function.

**lines** Output vector of lines. Each line is represented by a 4-element vector  $(x_1, y_1, x_2, y_2)$ , where  $(x_1, y_1)$  and  $(x_2, y_2)$  are the ending points of each detected line segment.

**rho** Distance resolution of the accumulator in pixels.

**theta** Angle resolution of the accumulator in radians.

**threshold** Accumulator threshold parameter. Only those lines are returned that get enough votes ( $> \text{threshold}$ ).

**minLineLength** Minimum line length. Line segments shorter than that are rejected.

**maxLineGap** Maximum allowed gap between points on the same line to link them.

Example:

`HoughLines(dst, lines, 1, CV_PI/180, 180, 0, 0);`

Example:

`HoughLinesP(dst, linesP, 1, CV_PI/180, 50, 50, 1);`



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# Practical task 6

## Draw lines based on HoughLines output:

```
vector<Vec2f> lines; // will hold the results of the detection
HoughLines(dst, lines, 1, CV_PI/180, 180, 0, 0 ); // runs the actual detection
// Draw the lines
for( size_t i = 0; i < lines.size(); i++ )
{
    float rho = lines[i][0], theta = lines[i][1];
    Point pt1, pt2;
    double a = cos(theta), b = sin(theta);
    double x0 = a*rho, y0 = b*rho;
    pt1.x = cvRound(x0 + 1000*(-b));
    pt1.y = cvRound(y0 + 1000*(a));
    pt2.x = cvRound(x0 - 1000*(-b));
    pt2.y = cvRound(y0 - 1000*(a));
    line( dst, pt1, pt2, Scalar(0,0,255), 3, LINE_AA);
}
```

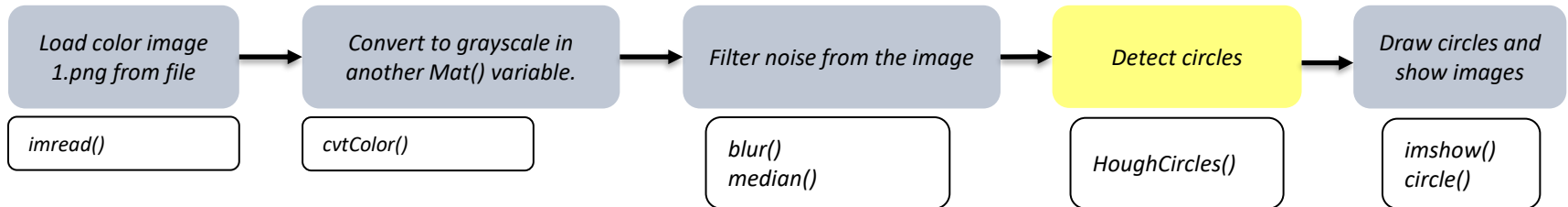
## Draw lines based on HoughLinesP output:

```
// Probabilistic Line Transform
vector<Vec4i> linesP; // will hold the results of the detection
HoughLinesP(dst, linesP, 1, CV_PI/180, 50, 50 , 1 ); // runs the actual detection
// Draw the lines
for( size_t i = 0; i < linesP.size(); i++ )
{
    Vec4i l = linesP[i];
    line( dst, Point(l[0], l[1]), Point(l[2], l[3]), Scalar(0,0,255), 3, LINE_AA);
}
```

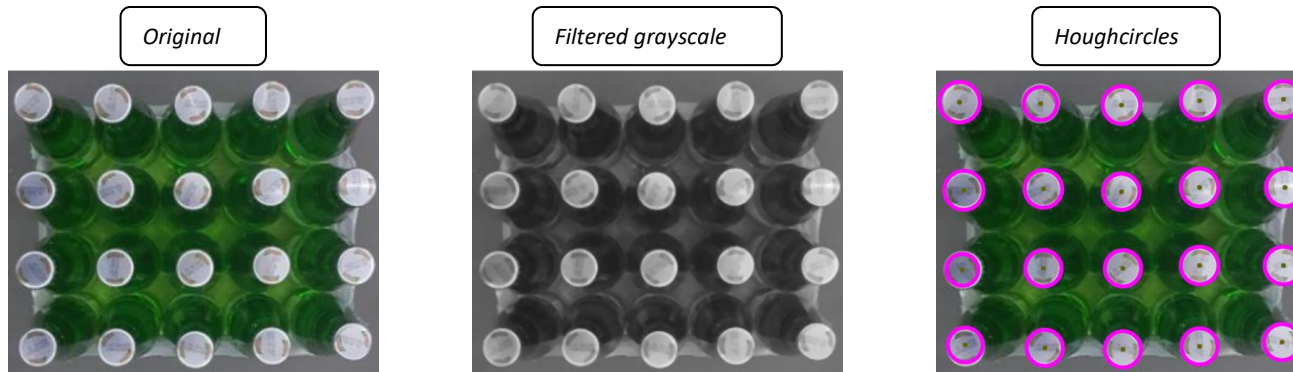


# Practical task 7

*Detect all the bottle caps as circles*



**RESULT:**



# Practical task 7

## HoughCircles()

### ◆ HoughCircles()

```
void cv::HoughCircles ( InputArray  image,
                        OutputArray circles,
                        int          method,
                        double       dp,
                        double       minDist,
                        double       param1 = 100 ,
                        double       param2 = 100 ,
                        int          minRadius = 0 ,
                        int          maxRadius = 0
                      )
```

#### Python:

```
cv.HoughCircles( image, method, dp, minDist[, circles[, param1[, param2[, minRadius[, maxRadius]]]]) -> circles
```

```
#include <opencv2/imgproc.hpp>
```

Finds circles in a grayscale image using the Hough transform.

The function finds circles in a grayscale image using a modification of the Hough transform.

#### Note

Usually the function detects the centers of circles well. However, it may fail to find correct radii. You can assist to the function by specifying the radius range ( minRadius and maxRadius ) if you know it. Or, you may set maxRadius to a negative number to return centers only without radius search, and find the correct radius using an additional procedure.

#### Parameters

- image** 8-bit, single-channel, grayscale input image.
- circles** Output vector of found circles. Each vector is encoded as 3 or 4 element floating-point vector  $(x, y, radius)$  or  $(x, y, radius, votes)$ .
- method** Detection method, see `HoughModes`. Currently, the only implemented method is `HOUGH_GRADIENT`
- dp** Inverse ratio of the accumulator resolution to the image resolution. For example, if  $dp=1$ , the accumulator has the same resolution as the input image. If  $dp=2$ , the accumulator has half as big width and height.
- minDist** Minimum distance between the centers of the detected circles. If the parameter is too small, multiple neighbor circles may be falsely detected in addition to a true one. If it is too large, some circles may be missed.
- param1** First method-specific parameter. In case of `HOUGH_GRADIENT`, it is the higher threshold of the two passed to the Canny edge detector (the lower one is twice smaller).
- param2** Second method-specific parameter. In case of `HOUGH_GRADIENT`, it is the accumulator threshold for the circle centers at the detection stage. The smaller it is, the more false circles may be detected. Circles, corresponding to the larger accumulator values, will be returned first.
- minRadius** Minimum circle radius.
- maxRadius** Maximum circle radius. If  $\leq 0$ , uses the maximum image dimension. If  $< 0$ , returns centers without finding the radius.

### Example:

```
HoughCircles(gray, circles, HOUGH_GRADIENT, 1,
            gray.rows/16, // change this value to detect circles with different distances to each other
            100, 30, 1, 30 // change the last two parameters
            // (min_radius & max_radius) to detect larger circles
);
```

[https://docs.opencv.org/3.4/dd/d1a/group\\_imgproc\\_feature.html#qa47849c3be0d0406ad3ca45db65a25d2d](https://docs.opencv.org/3.4/dd/d1a/group_imgproc_feature.html#qa47849c3be0d0406ad3ca45db65a25d2d)



# Practical task 7

*Draw circles based on the output from the HoughCircles:*

**Example:**

```
for( size_t i = 0; i < circles.size(); i++ )
{
    Vec3i c = circles[i];
    Point center = Point(c[0], c[1]); // circle center
    circle( src, center, 1, Scalar(0,100,100), 3, LINE_AA); // circle outline
    int radius = c[2];
    circle( src, center, radius, Scalar(255,0,255), 3, LINE_AA);
}
```

[https://docs.opencv.org/3.4/dd/d1a/group\\_imgproc\\_feature.html#qa47849c3be0d0406ad3ca45db65a25d2d](https://docs.opencv.org/3.4/dd/d1a/group_imgproc_feature.html#qa47849c3be0d0406ad3ca45db65a25d2d)



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# *Student assignment (seminar)*



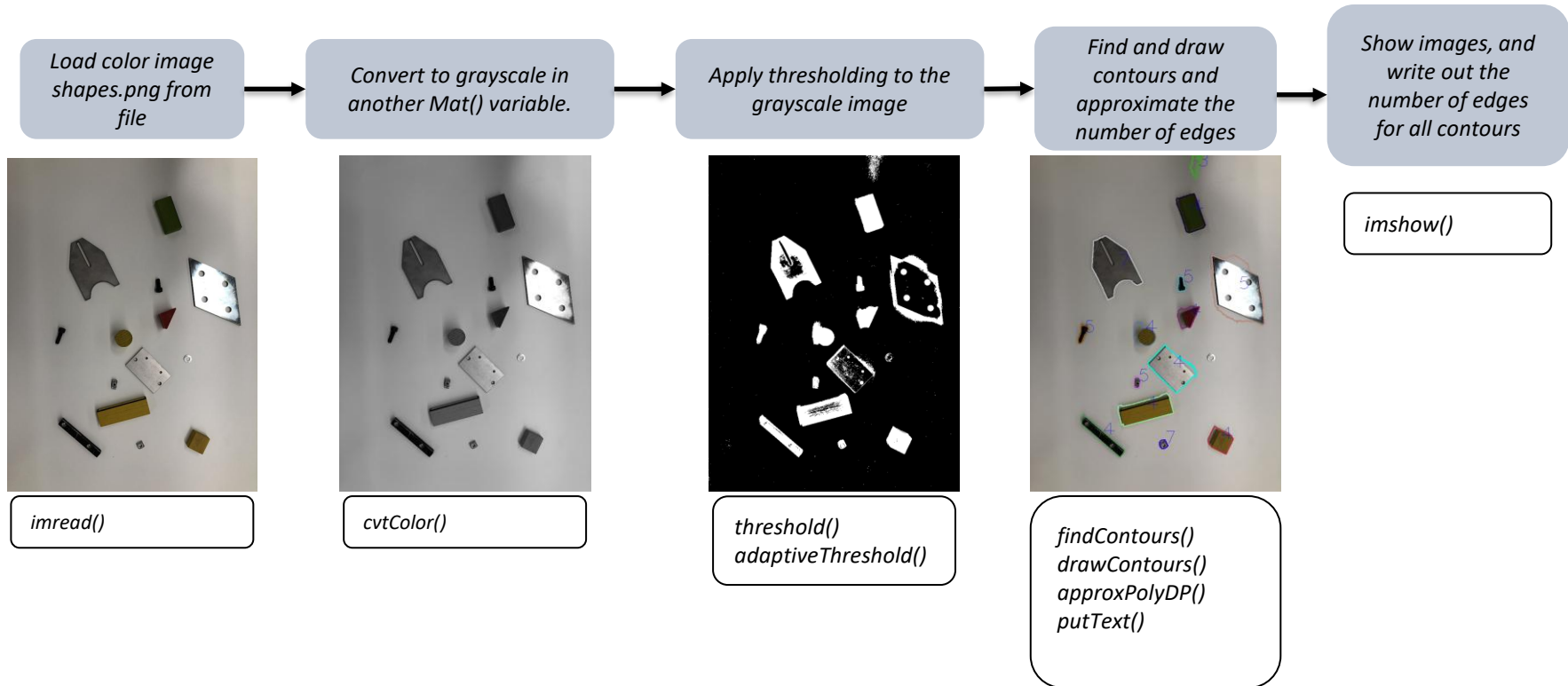
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# Student assignment - seminar

## Shape detection – test the algorithm in a realistic example

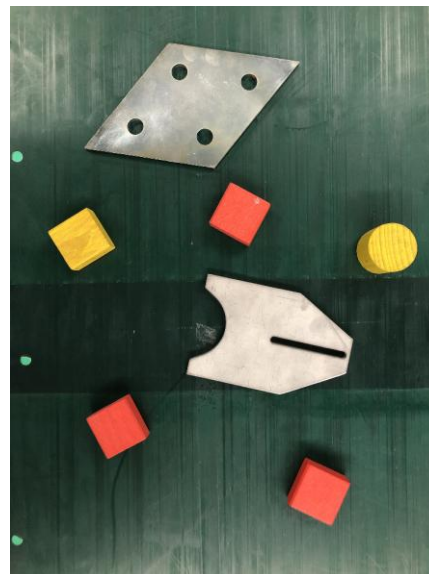


- `approxPolyDP()` uses Douglas-Peucker algorithm to approximate a curve or a polygon with another curve/polygon with less vertices
- Write out the number of edges next to any shape
- NOTICE → image noise and approximation epsilon can give wrong results, test different settings to get best results

# Student assignment - seminar

**Add trackbar and change the threshold value and inRange lower and upper limits, for Practical tasks 1, 2, 3 and 4.**

Test out the object localization on other images („2.png” and „3.png”):



For trackbar use example :

[https://docs.opencv.org/master/da/d6a/tutorial\\_trackbar.html](https://docs.opencv.org/master/da/d6a/tutorial_trackbar.html)

# ***Student assignment - seminar***

*Detect lines – test the algorithm on road.jpg example and change the parameters (canny edge, hough lines,...) to get better results*

