

# Computer Vision

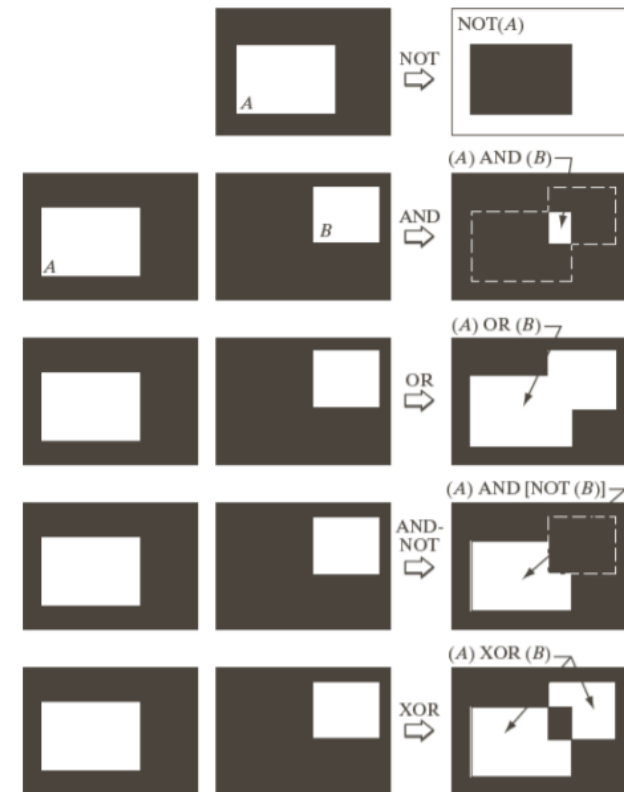
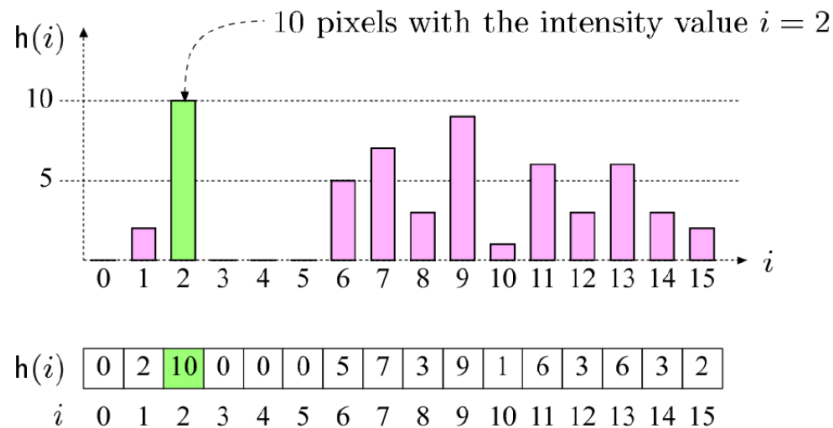
Enhancement  
techniques  
Segmentation

Fátima Leal

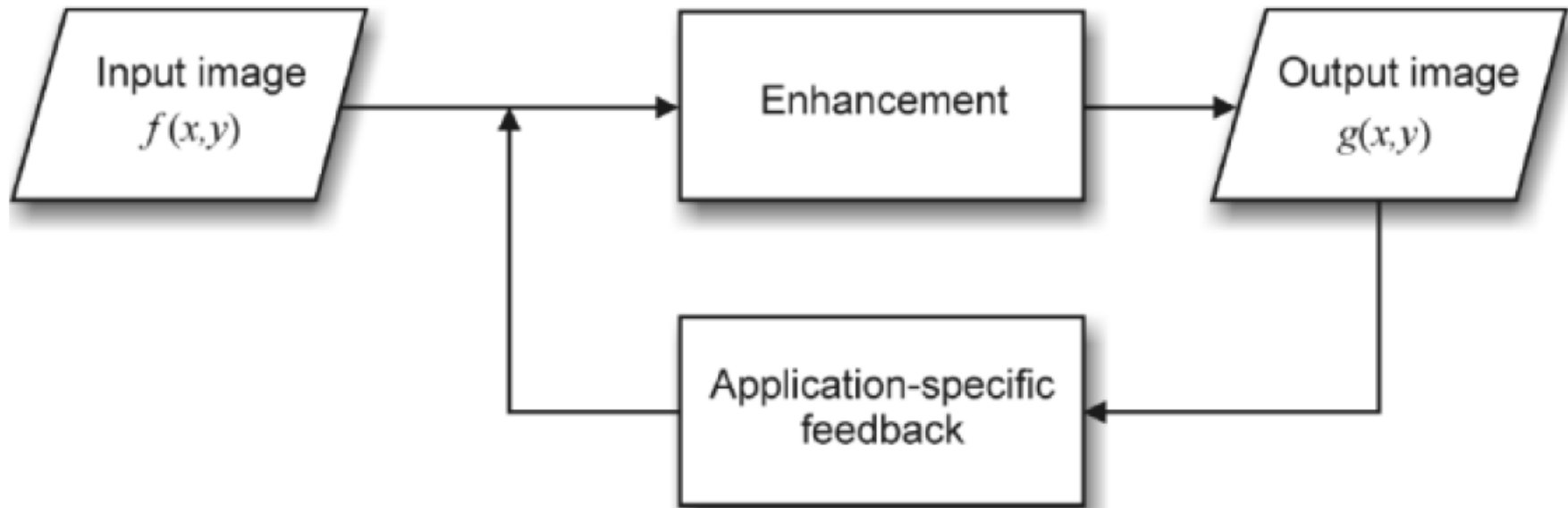
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# What we have learnt

- Arithmetic and Logic operations
- Histograms



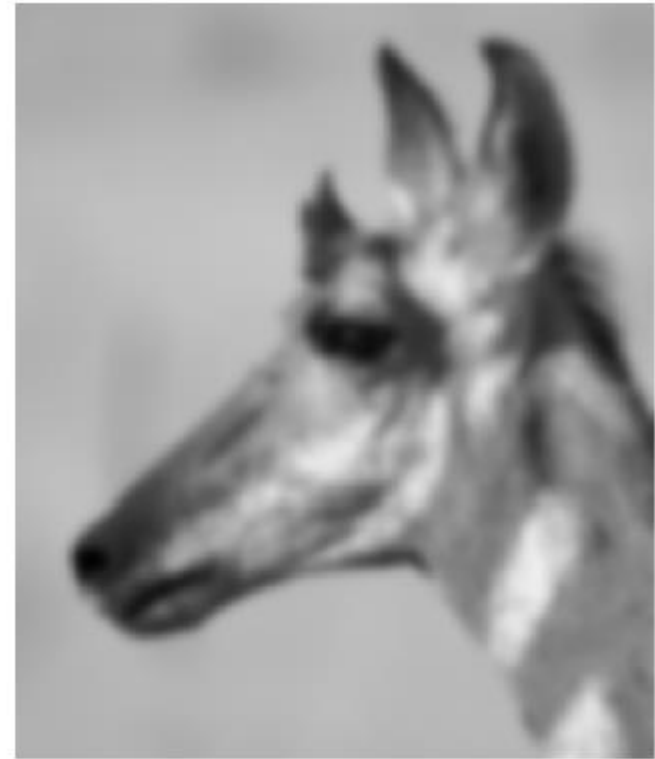
# What we have learnt



# Content

- Enhancement Techniques
- Segmentation
- Low Pass Filters
- High Pass Filters

# Image Enhancement Techniques



# Image Enhancement Techniques

- **A blurring image** happens when the camera takes a picture out of focus.
- Sharper regions in the image lose their detail, normally as a disc/circular shape. Then each pixel in the image is mixed in with its surrounding pixel intensities.
- **Image enhancement** techniques are very important not only to improve the subjective quality of an image for human viewing, but also to modify the image in such a way as to make it more suitable to further analysis and automatic extraction of its contents.

# Image Enhancement Techniques

- We have different types of enhancement techniques:
  - Low pass Filters
    - Average Filtering
    - Gaussian filter
    - Nonlinear filtering
  - High Pass filters
    - Laplacian filter

# LOW PASS FILTER



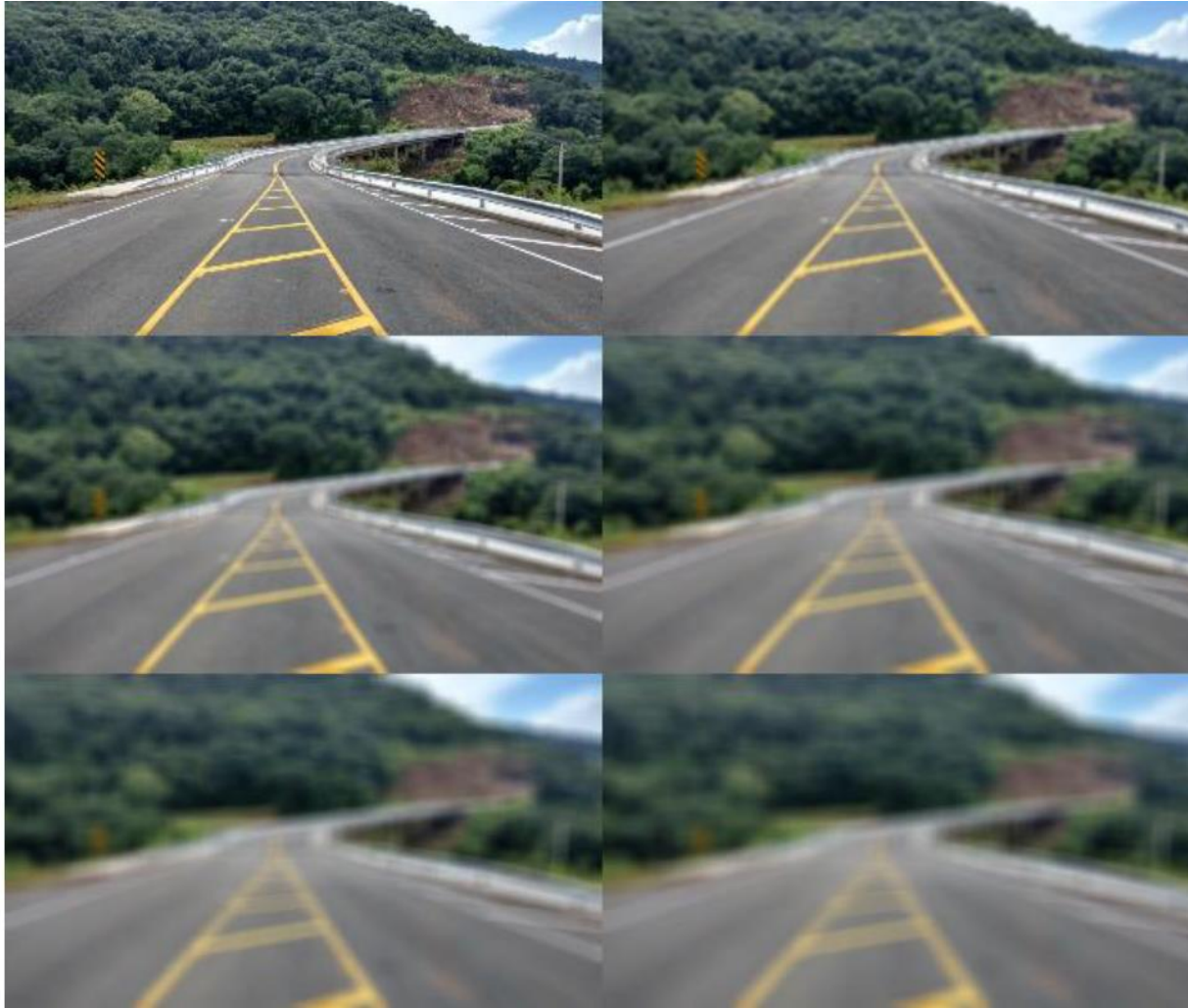
# Average Filtering

- In this technique it is calculates the average of the neighbourhood pixels
- This box is named as kernel box.

30	100	130
130	Pixel	160
50	100	210

- No OpenCV it is used the blur method for this technique.

# Average Filtering



# Low Pass Filter

## Average Filter



(a)



(b)



(c)



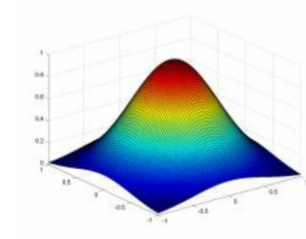
(d)

# Gaussian Filter

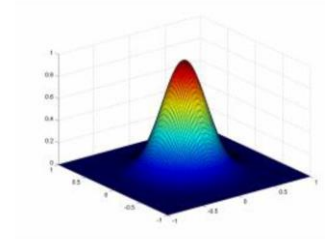
- It is used a kernel gaussian
- This is calculated using the `cv2.GaussianBlur`.
- The function requires the specification of a width and height with odd numbers
- Optionally, it is possible to specify the number of standard deviations on the X and Y axis (horizontal and vertical).

A 2D Gaussian is just the product of 1D Gaussians:

$$g(x, y; \sigma) = \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}} = \frac{1}{\sqrt{2\pi}\sigma} \exp^{-\frac{x^2}{2\sigma^2}} \cdot \frac{1}{\sqrt{2\pi}\sigma} \exp^{-\frac{y^2}{2\sigma^2}} = g_\sigma(x) \cdot g_\sigma(y)$$

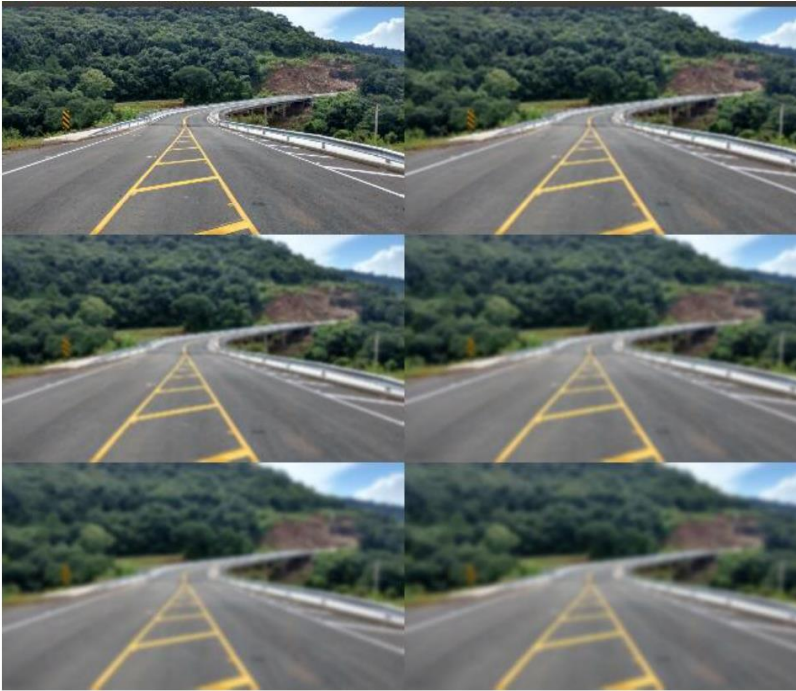


(a)  $\sigma = 0.5$



(b)  $\sigma = 0.25$

# Gaussian Filter vs Neighbourhood



Gaussian



Average

# Nonlinear Filtering

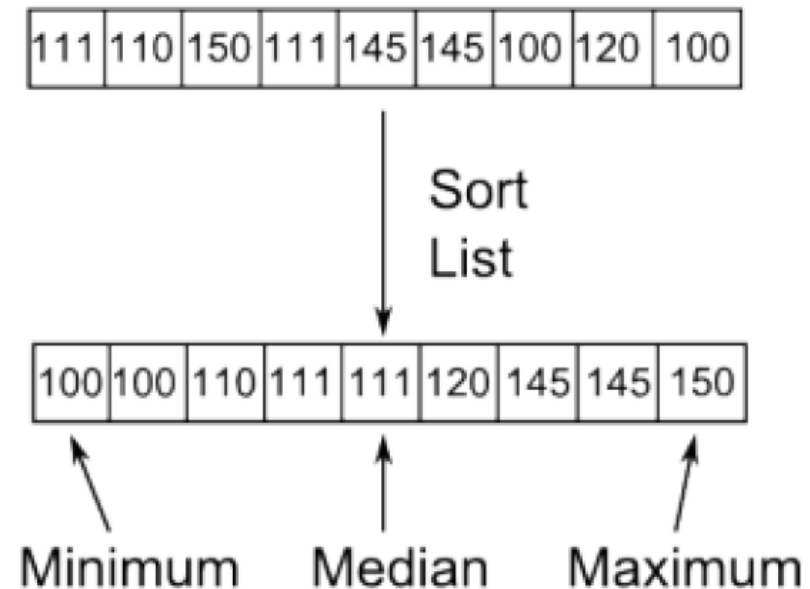
- In linear filtering each output pixel is a weighted summation of some number of input pixels;
- Easier to compose and are amenable to frequency response analysis. In nonlinear filtering a combination of neighbouring pixels is performed.
- They are also known as **rank-order filters**.
- Median Filter consists in finding the median value on a neighbourhood with a given size. It is more effective in not blurring edges.



# Nonlinear Filtering

## Nonlinear Filter

	111	110	150			
	111	145	145			
	100	120	100			



# Nonlinear Filtering

- When applying a median blur, we first define our **kernel size  $k$** .
- Then, as in the averaging blurring method, we consider all pixels in the neighbourhood of size  $k \times k$ .
- Instead of replacing the central pixel with the average of the neighbourhood, we **replace the central pixel with the median** of the neighbourhood.
- Median blurring is more effective
- Averaging and Gaussian methods can compute means or weighted means for the neighbourhood
- By replacing our central pixel with a median rather than an average, we can substantially reduce noise.



# Nonlinear Filtering



# High PASS FILTER

# Laplacian Filter

- Laplacian filter is a **High-Pass Filter** (HPF) whose effect on the output image is equivalent to preserving or emphasizing its **high-frequency components** (e.g., fine details, points, lines, and edges), i.e., to highlight transitions in intensity within the image.
- Linear HPFs can be implemented using **2D convolution** masks with positive and negative coefficients, which correspond to a digital approximation of the Laplacian, a simple, isotropic (i.e., rotation invariant) second-order derivative that can respond to intensity transitions in any direction.

# Laplacian Filter



# Image Segmentation

- Extract symbolic information as regions of interest
- Finding groups of pixels
- Image segmentation techniques can vary widely according to:
  - **type of image** (e.g., binary, gray, color).
  - **choice of mathematical framework** (e.g., morphology, image statistics, graph theory).
  - **type of features** (e.g., intensity, color, texture, motion) and
  - **approach** (e.g., top-down, bottom-up, graph-based).
- Three categories:
  - **Intensity-based** (non-contextual) methods: work based on pixel distributions (i.e., histograms).
  - **Region-based** (contextual) methods: rely on adjacency and connectivity criteria between a pixel and its neighbours.
  - **Other methods**: segmentation based on texture, edges, and motion

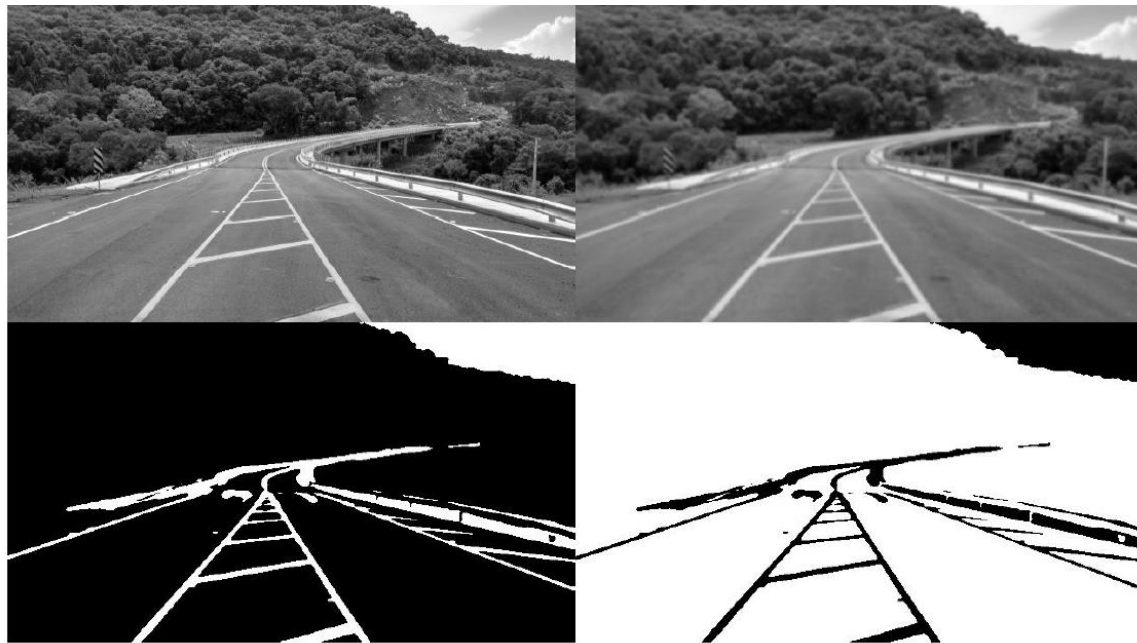
# Image Segmentation: Simple Thresholding

- A simple thresholding example would be selecting a pixel value  $T$ , and then setting all pixel intensities less than  $T$  to zero, and all pixel values greater than  $T$  to 255. In this way, we can create a binary representation of the image.

$$g(x,y) = \begin{cases} 255 & \text{if } f(x,y) \geq T \\ 0 & \text{if } f(x,y) < T \end{cases}$$

# Image Segmentation: Simple Thresholding

- For autonomous vehicle this is one of the techniques used to identify the road
- The same technique is used to identify objects

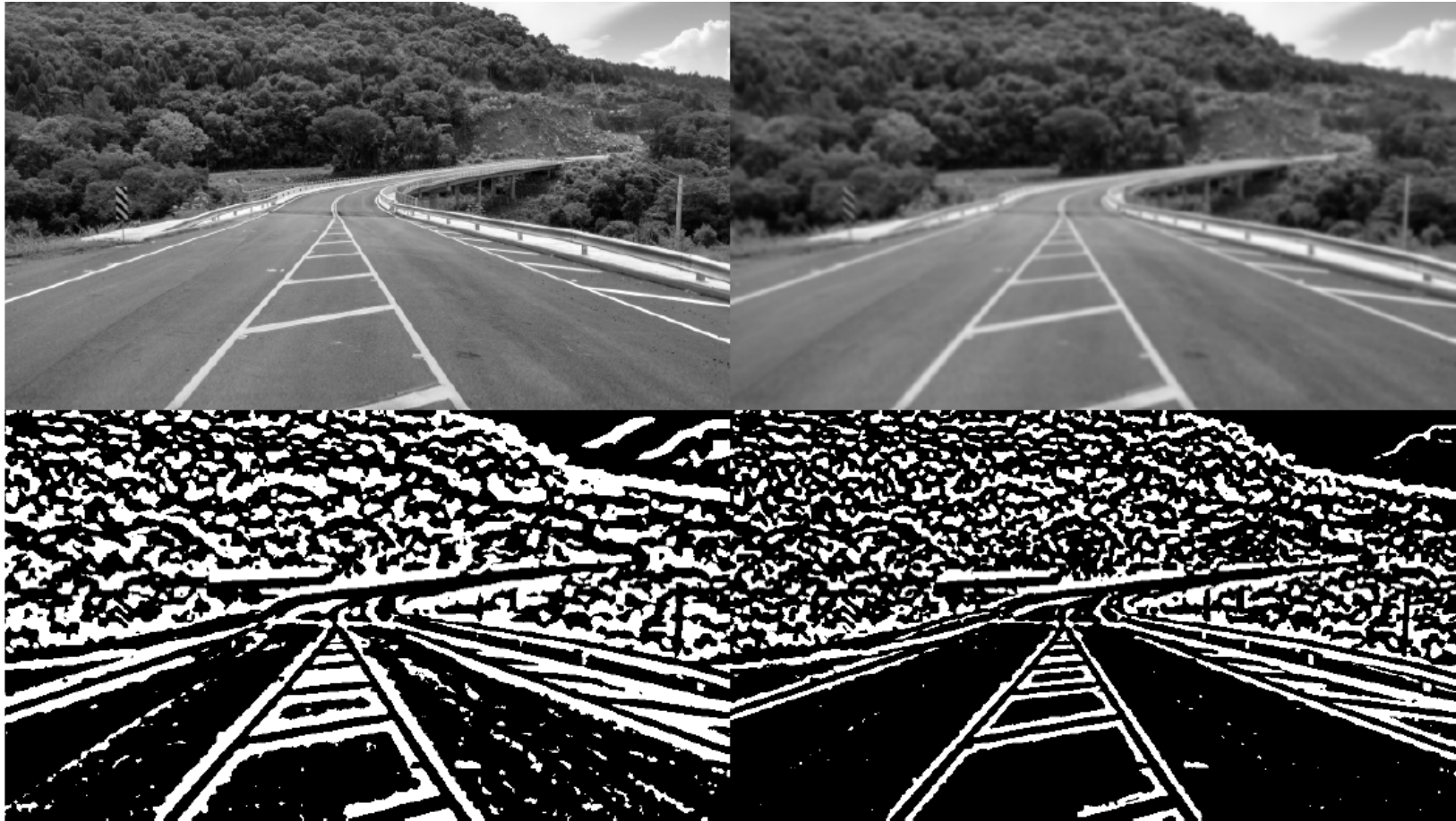


# Image Segmentation: Adaptive Thresholding

- T value used in the previous example was arbitrary. However, there are math techniques to optimise. This is proposed by adaptive thresholding
- We need:
  - A window value where the threshold is calculated using pixels next to the image.
  - A parameter which is subtracted to the average calculated and generate the final threshold.

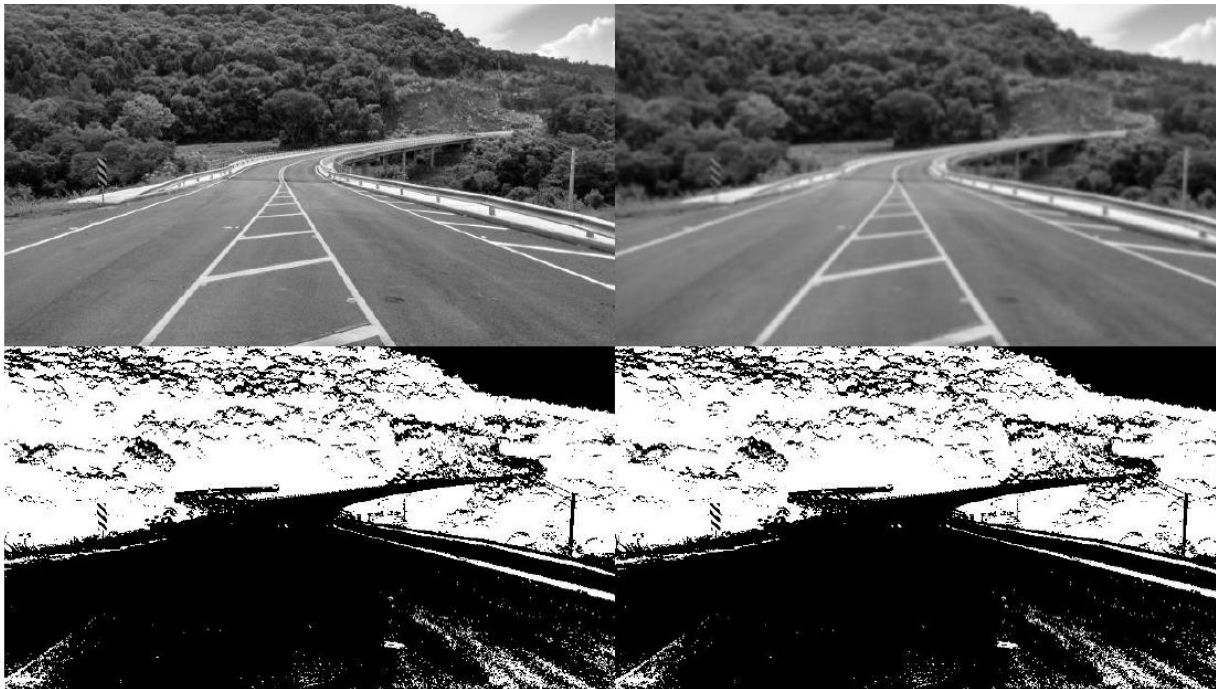


# Image Segmentation: Adaptive Thresholding



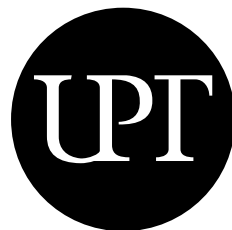
# Otsu

- Otsu's method assumes there are two peaks in the grayscale histogram of the image and then tries to find an optimal value to separate those
- Otsu is common seen as a global thresholding method.



# Let's play with images!





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