

Lecture 06 - Image Descriptors

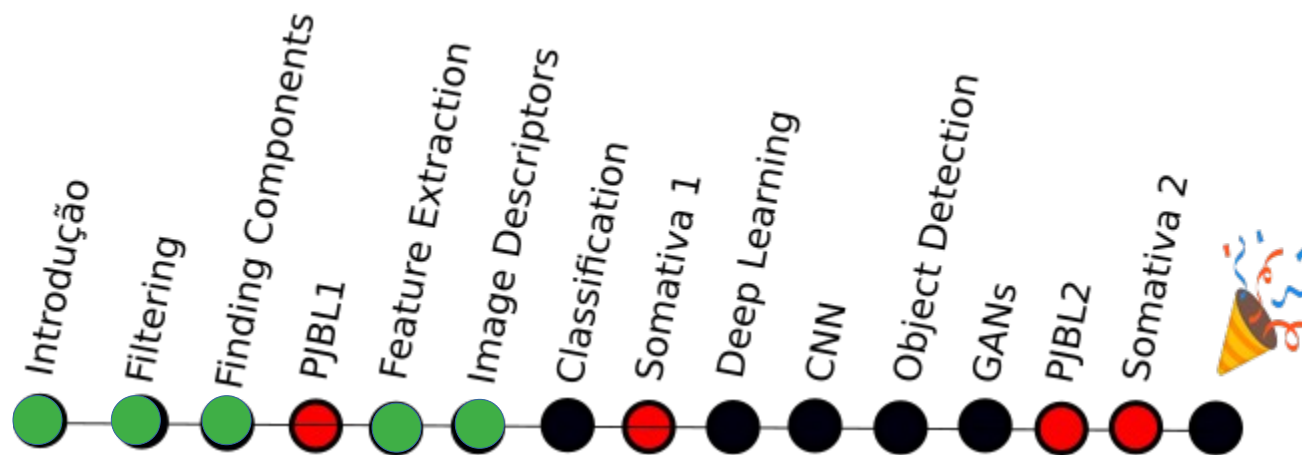
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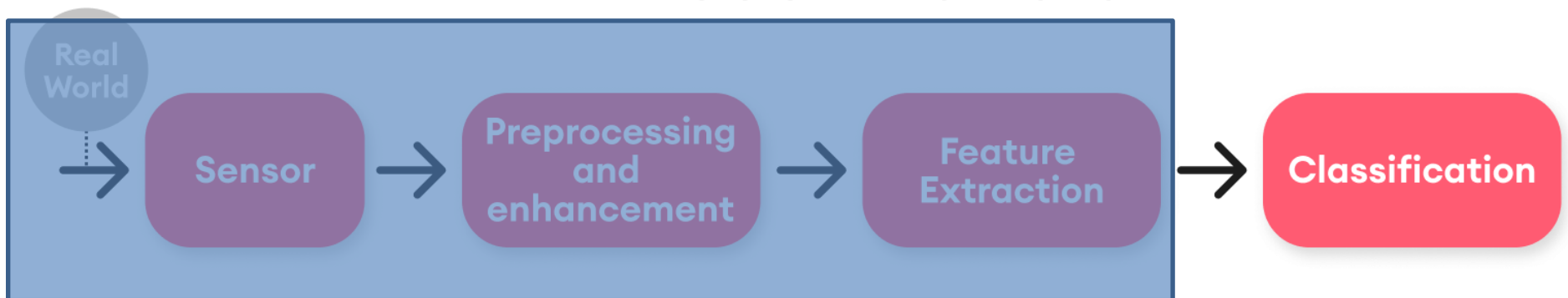
Topics

- Discussion of Lecture #05
 - Feature Vector
 - Horizontal and Vertical Projections
- Image Descriptors
- Practice



Computer Vision & Pattern Recognition Pipeline

PATTERN RECOGNITION SYSTEM



Remembering....

- A feature descriptor translates high-dimensional data to a low dimension feature space
- A feature vector represents the input data produced by the feature descriptor
- Later, a machine learning model will learn the representations

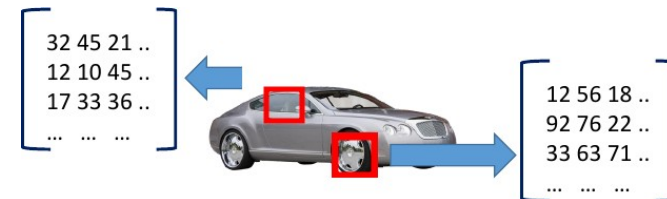
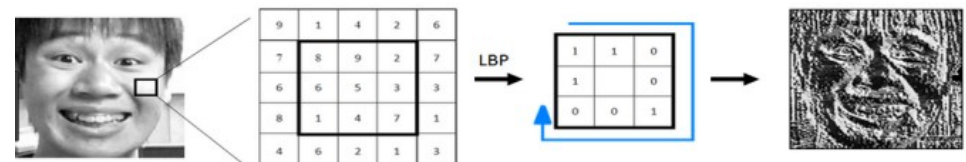
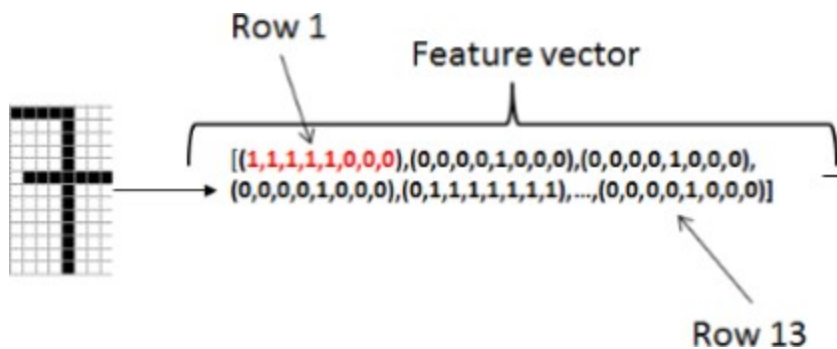
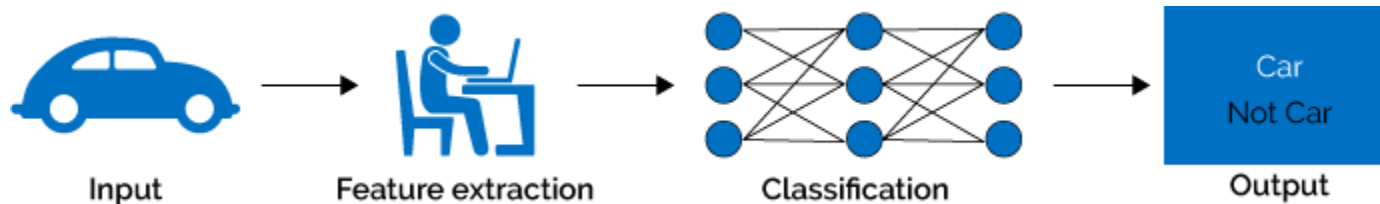
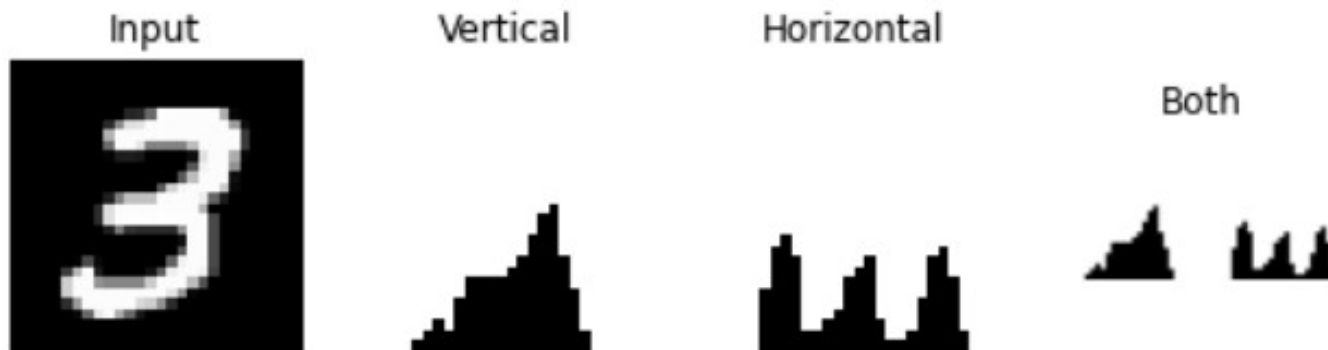


Image Descriptors - Shape / Edges

- Gradient Based

- Projections



- Convolutional (Filters)

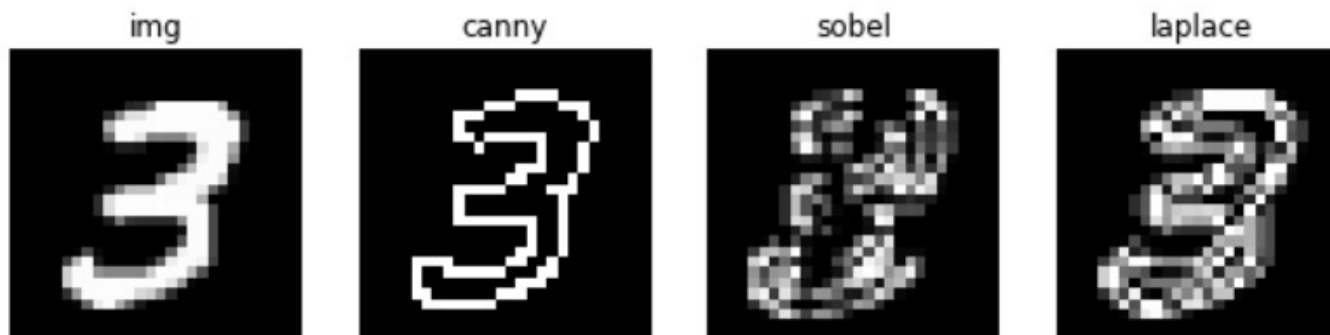


Image Descriptors - Shape / Edges

- Histogram Projection

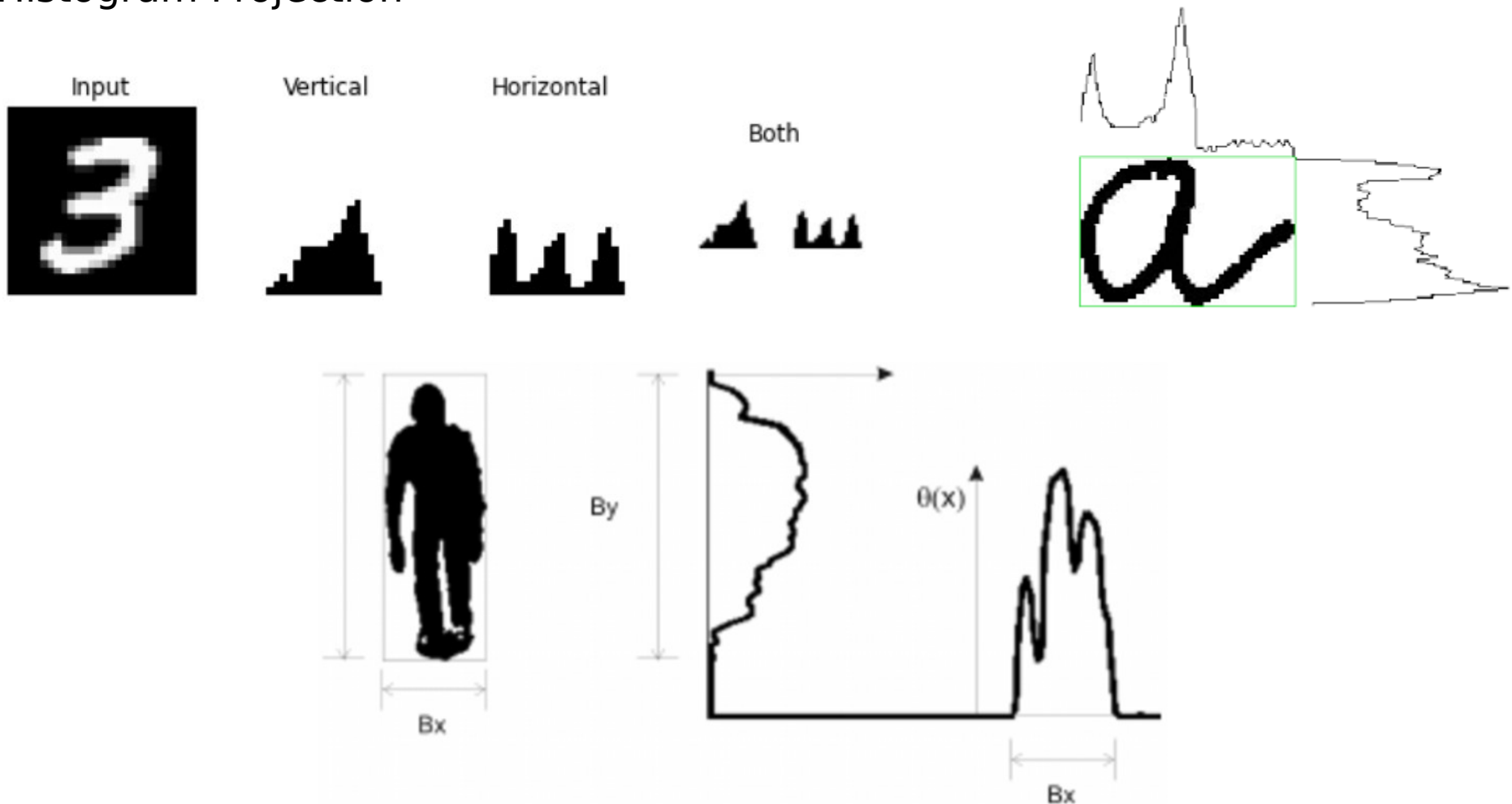


Image Descriptors - Shape / Edges

- Sobel Filter

-1	0	+1
-2	0	+2
-1	0	+1
Gx		

+1	+2	+1
0	0	0
-1	-2	-1
Gy		

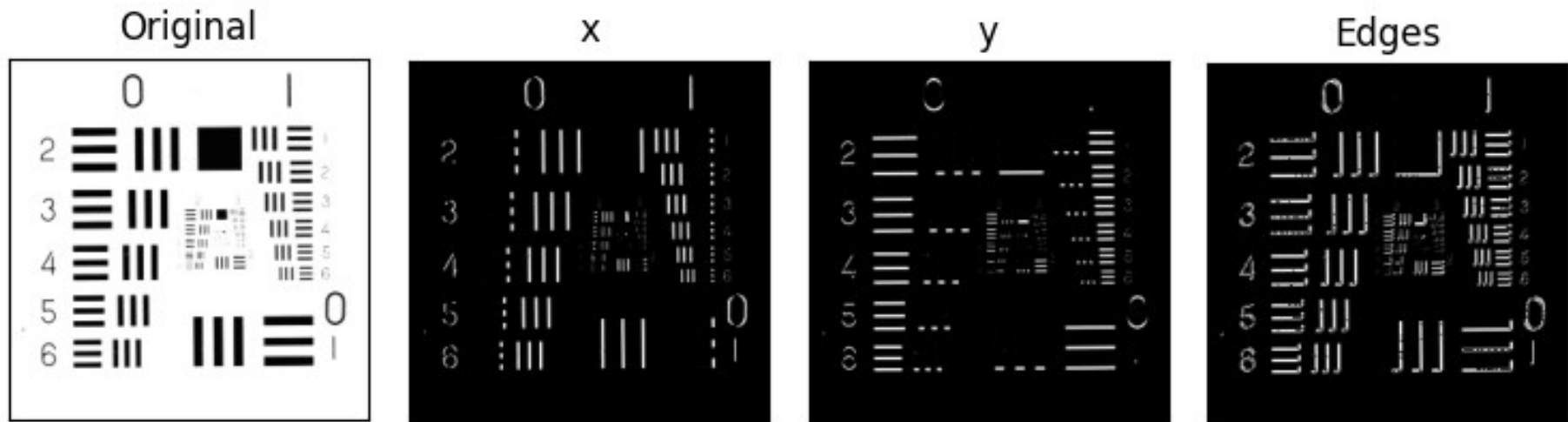


Image Descriptors - Shape / Edges

- Sobel Filter

-1	0	+1		+1	+2	+1
-2	0	+2		0	0	0
-1	0	+1		-1	-2	-1
Gx				Gy		

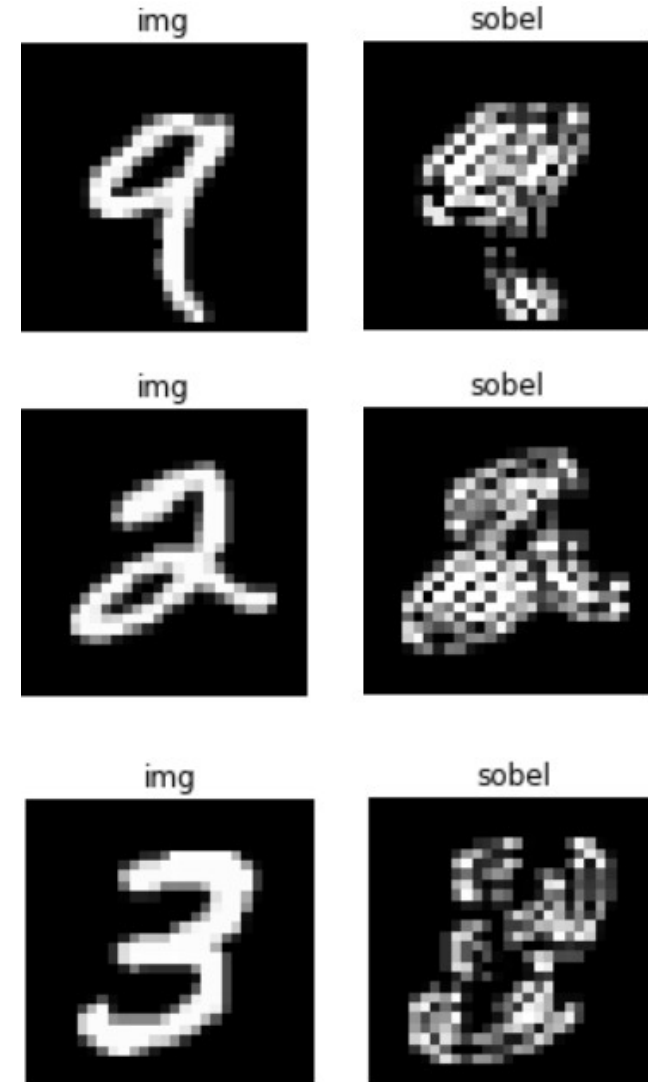


Image Descriptors - Shape / Edges

- Laplace

$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

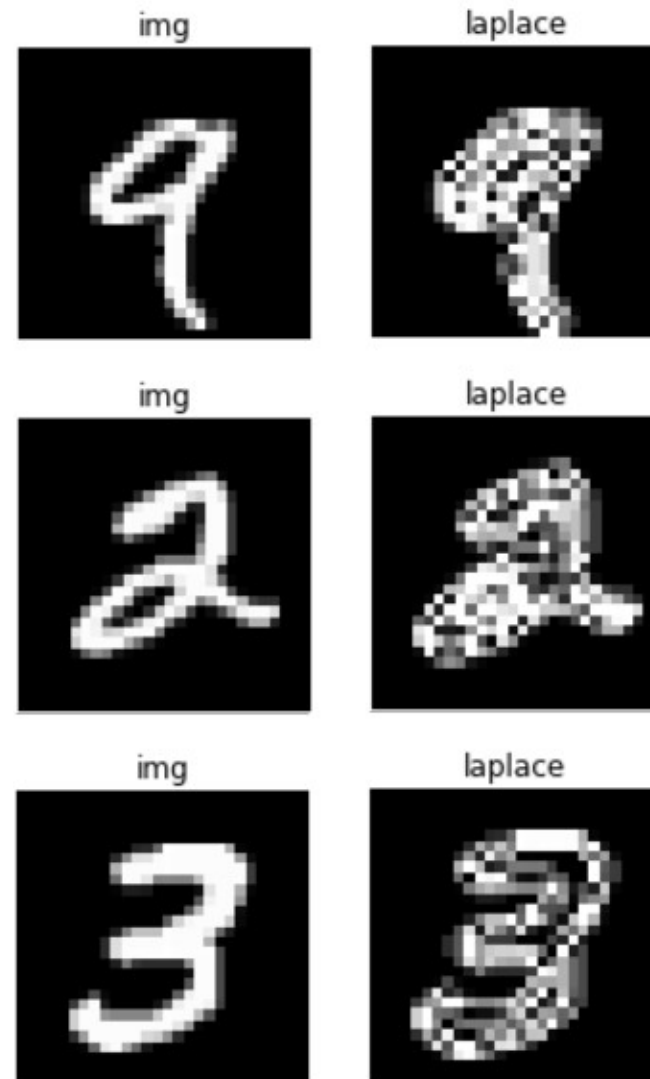


Image Descriptors - Shape / Edges

- Canny
 - Gaussian Gradient Based Filter (John F. Canny)
 - Gaussian Blur
 - Gradient Detection

$$\mathbf{B} = \frac{1}{159} \begin{bmatrix} 2 & 4 & 5 & 4 & 2 \\ 4 & 9 & 12 & 9 & 4 \\ 5 & 12 & 15 & 12 & 5 \\ 4 & 9 & 12 & 9 & 4 \\ 2 & 4 & 5 & 4 & 2 \end{bmatrix} * \mathbf{A}.$$

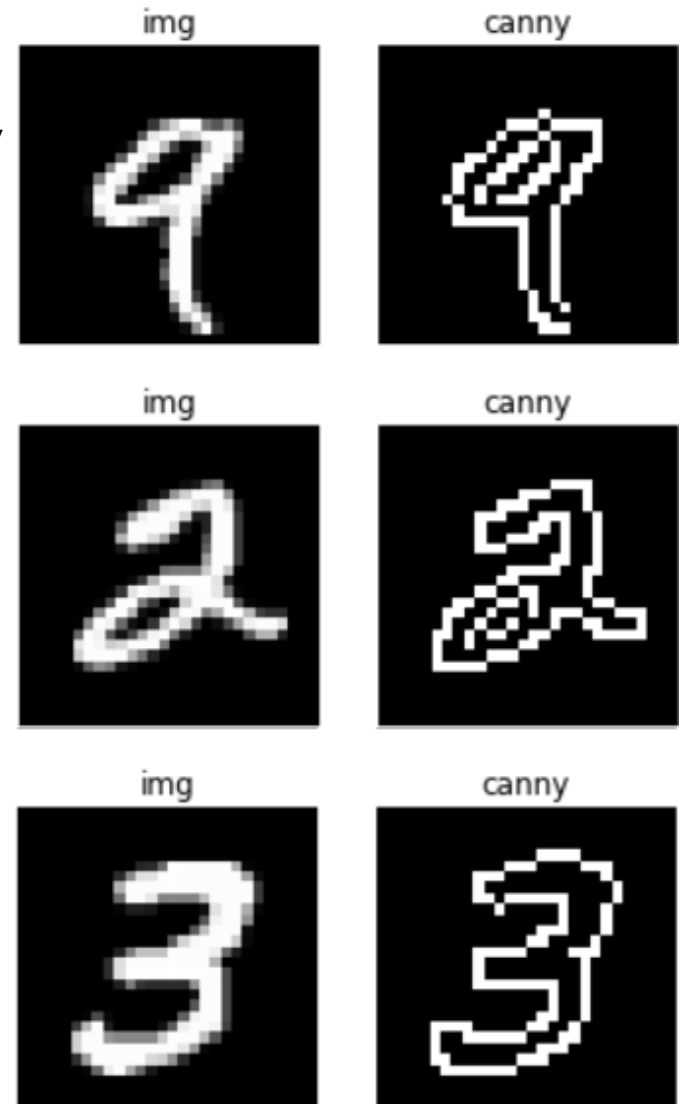


Image Descriptors - Shape / Edges

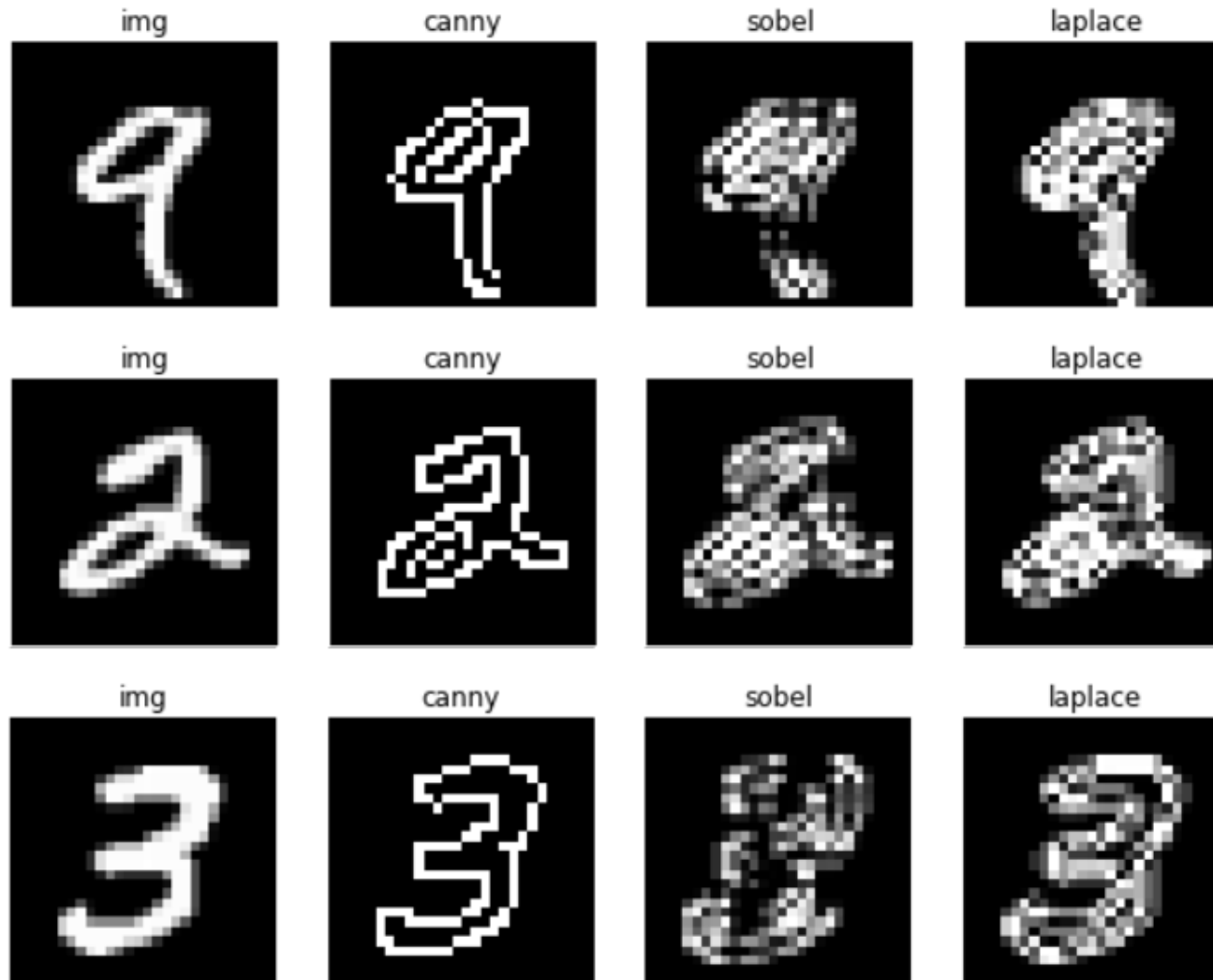


Image Descriptors - Shape / Edges

Original Image



Canny



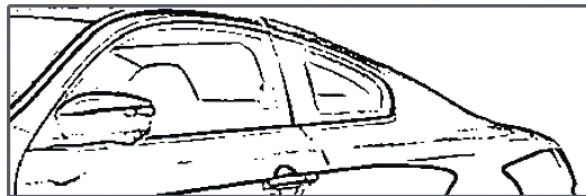
Sobel



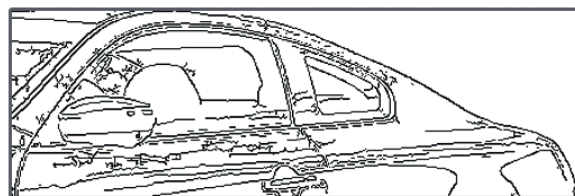
Laplacian



Laplace



Canny



Sobel



Let's Code

- [LINK](#)