

# IMAGE PROCESSING PART I

- Professor: Rubén Alvarez

# CONTENT

- Digital filters
- Filter analysis
- Filter design
- Filters comparison
- Model colors

# REMEMBER

- Convolution theorem

$$\mathcal{F}(f * g) = F(\omega)G(\omega)$$

# 2D CONVOLUTION

$$y_{k,l} = \sum_{m=-1}^1 \sum_{n=-1}^1 x_{k-m,l-n} g_{m,n}$$

## DIGITAL FILTER

# WHAT IS A DIGITAL FILTER AND WHAT IS IT FOR IMAGES?

# DIGITAL FILTER

- Objectives:
  - Image smoothing
  - Noise removal
  - Improve the image
  - Edge detection?

# WHAT IS NOISE?

- In images is the unwanted information that contaminates the image.
- Defined as

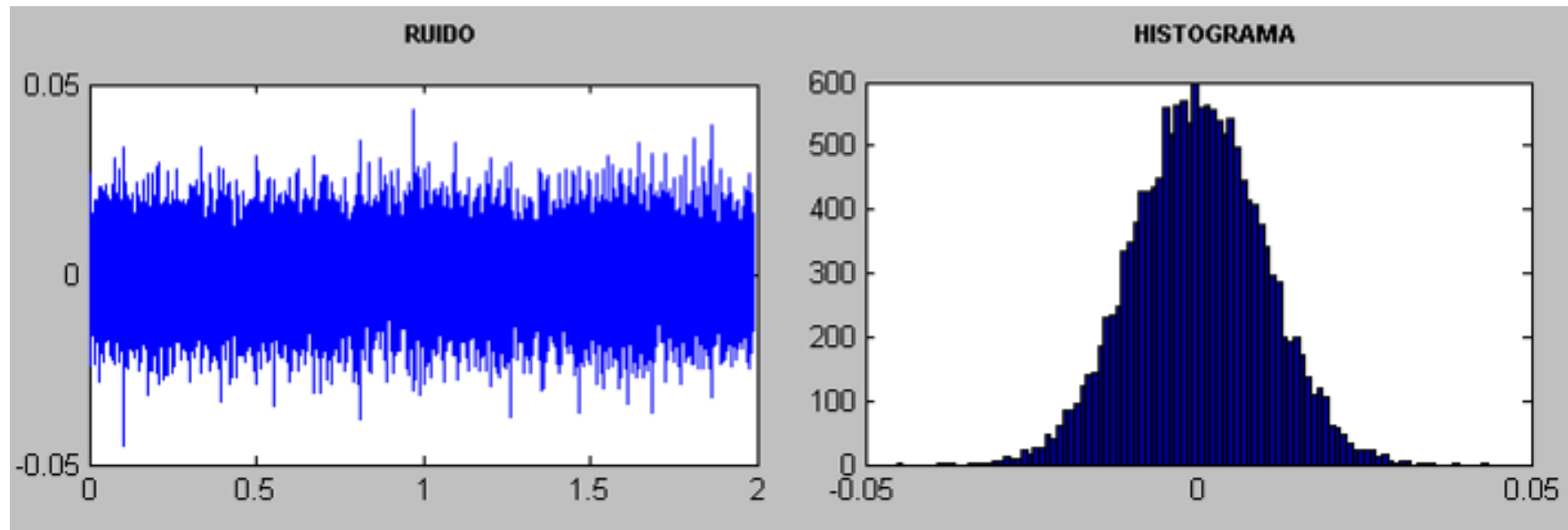
$$g(x, y) = f(x, y) + r(x, y)$$

# TYPES OF NOISE

- Gaussian Noise
- Uniform noise ... (Bias?)
- Impulsive noise (Salt and pepper)
- Etc.



# GAUSSIAN NOISE



# SALT AND PEPPER NOISE



KERNEL?

KERNEL OR CONVOLUTION MASK?

# AVERAGE FILTER

$$f'(x, y) = \frac{1}{MN} \sum_{i=-\frac{M}{2}}^{\frac{M}{2}} \sum_{j=-\frac{N}{2}}^{\frac{N}{2}} f(x + i, y + j)$$

- Where M and N sun the size of the kernel or filter convolution mask.

# AVERAGE FILTER

- Then for a neighborhood of 3...

$$\frac{1}{9} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}$$

# AND IN PICTURES...

$F[x, y]$

0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0	0
0	0	0	90	90	90	90	90	0	0	0
0	0	0	90	90	90	90	90	0	0	0
0	0	0	90	0	90	90	90	0	0	0
0	0	0	90	90	90	90	90	0	0	0
0	0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0

$G[x, y]$

	0									

# AND IN PICTURES...

$$F[x, y]$$

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

$$G[x, y]$$

0	10								



# AND IN PICTURES...

$$F[x, y]$$

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

$$G[x, y]$$

	0	10	20						

# AND IN PICTURES...

$$F[x, y]$$

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

$$G[x, y]$$

	0	10	20	30					

# AND IN PICTURES...

$$F[x, y]$$

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

$$G[x, y]$$

	0	10	20	30					

# AND IN PICTURES...

$$F[x, y]$$

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

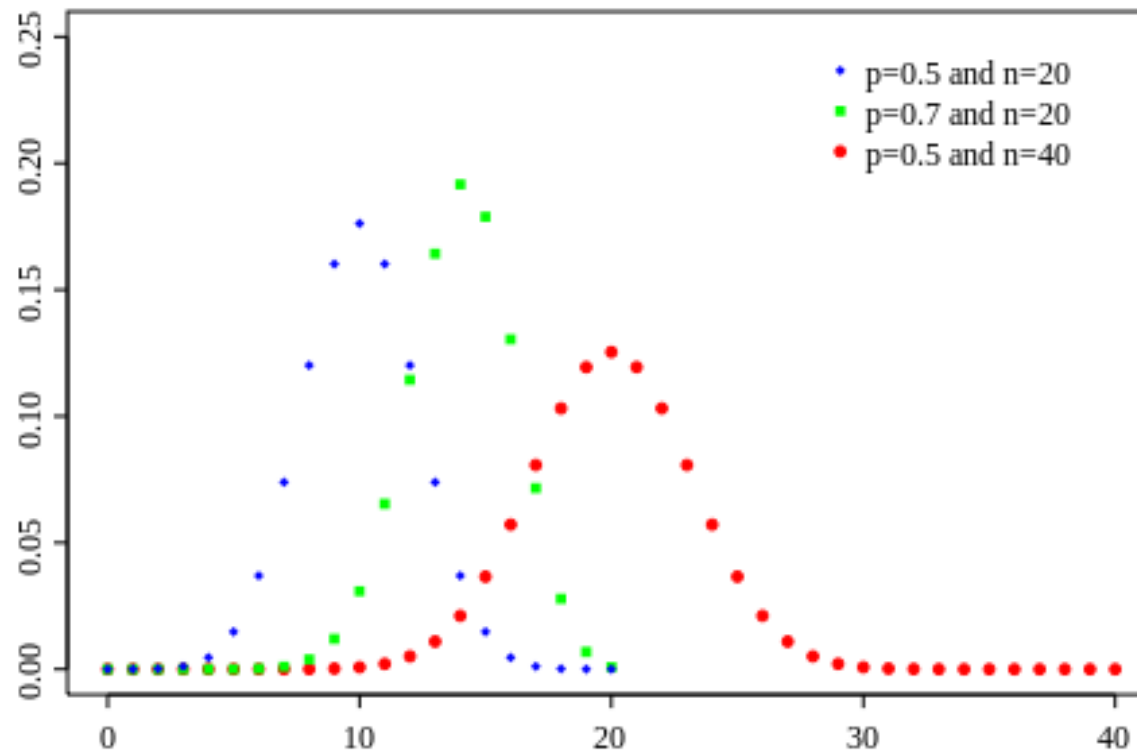
$$G[x, y]$$

	0	10	20	30	30	30	20	10	
	0	20	40	60	60	60	40	20	
	0	30	60	90	90	90	60	30	
	0	30	50	80	80	90	60	30	
	0	30	50	80	80	90	60	30	
	0	20	30	50	50	60	40	20	
	10	20	30	30	30	30	20	10	
	10	10	10	0	0	0	0	0	

## BINOMIAL FILTER

- In statistics, binomial distribution is a discrete probability distribution that counts the number of successes in a sequence of Bernoulli independent  $n$  trials independent of each other, with a fixed probability  $p$  of success occurrence between trials.

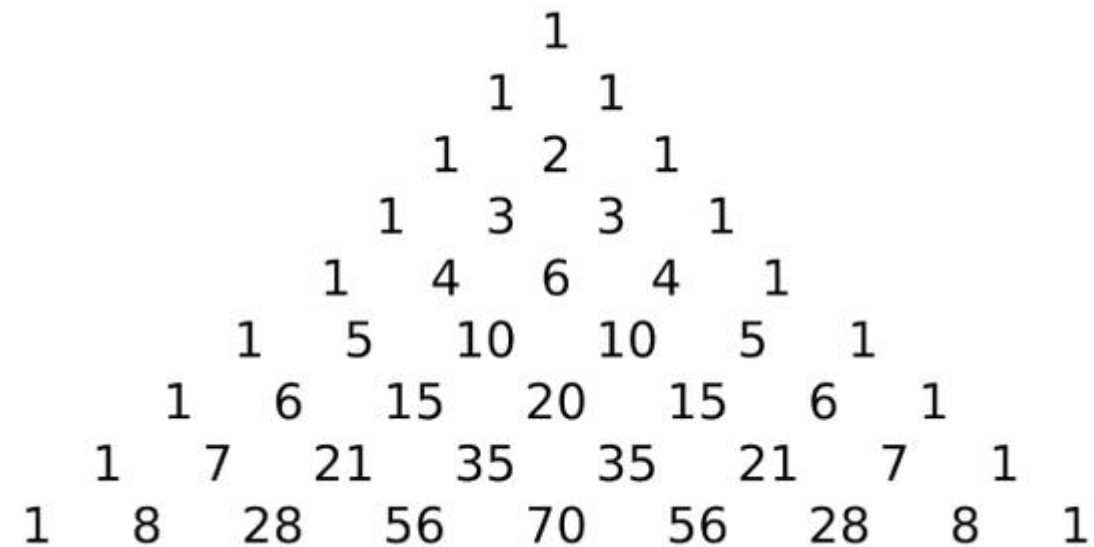
# BINOMIAL FILTER



# BINOMIAL FILTER

- How do I build a binomial filter for images?

# PASCAL TRIANGLE



				1					
			1		1				
		1		2		1			
	1		3		3		1		
	1	4		6		4		1	
	1	5	10		10	5		1	
	1	6	15	20		15	6		1
	1	7	21	35	35	21	7		1
1	8	28	56	70	56	28	8		1



# WE USE CONVOLUTION

$$h_{2D}^n = h_x^n * h_y^n$$

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$$h_{2D}^n = h_x^n * h_y^n$$

$$h_x^2 * h_y^2 = \frac{1}{4} \begin{pmatrix} 1 & 2 & 1 \end{pmatrix} * \frac{1}{4} \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$$

# WE USE CONVOLUTION

$$\begin{aligned} h_{2D}^n &= h_x^n * h_y^n \\ h_x^2 * h_y^2 &= \frac{1}{4} \begin{pmatrix} 1 & 2 & 1 \end{pmatrix} * \frac{1}{4} \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix} \\ &= \frac{1}{16} \begin{pmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{pmatrix} \end{aligned}$$

**BUT HOW CAN WE KNOW THE BEHAVIOR OF THE  
FILTER BEFORE APPLYING IT TO THE IMAGE?**

**BUT HOW CAN WE KNOW THE BEHAVIOR OF THE FILTER  
BEFORE APPLYING IT TO THE IMAGE?**

OR IS THE FILTER DESIGN EMPIRICAL?

## WITH FOURIER'S TRANSFORM IN FREQUENCY!

$$G_{k,l} = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} g(m,n) e^{-j\frac{2\pi k}{K}m} e^{-j\frac{2\pi l}{L}n}$$

- With  $k = 0, 1, 2, \dots, K-1$  &  $l = 0, 1, 2, \dots, L-1$  for 2 dimensions.

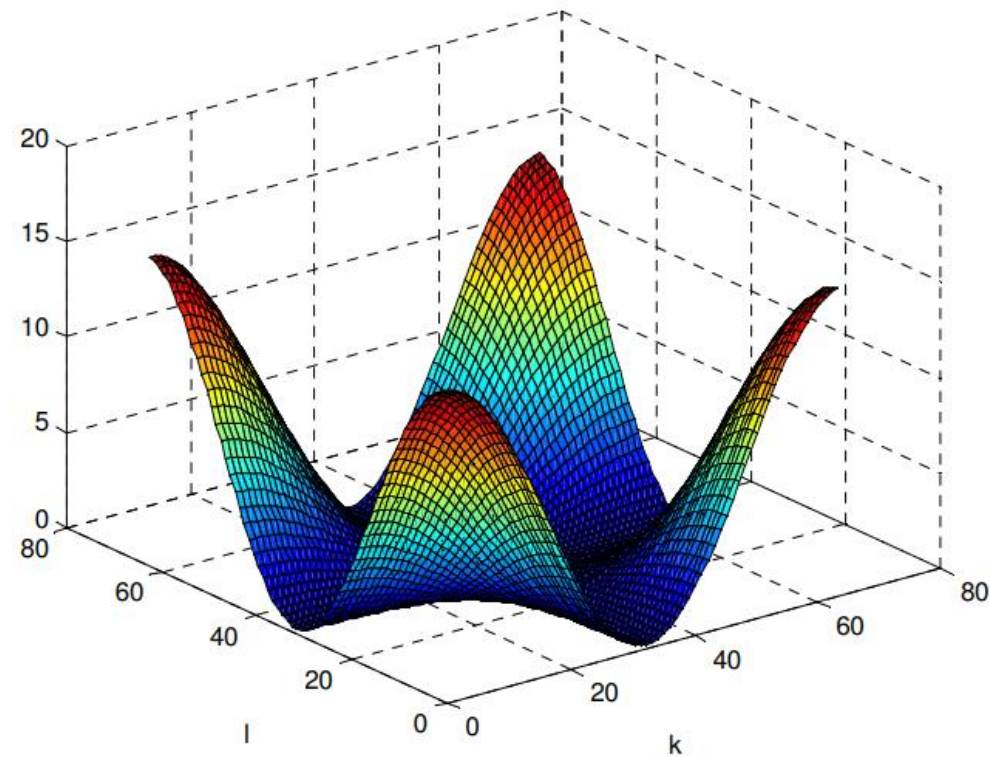
# EXERCISE

- We have:

$$\{g_n\} = \frac{1}{4}\{-1, \quad 0, \quad 2, \quad 0, \quad -1\}$$



# VIEWING IN FREQUENCY EXAMPLE



# GAUSSIAN FILTER



# QUALITY VS COMPLEXITY?

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IS THERE A BETTER FILTER FOR SALT AND PEPPER NOISE?

# GAUSSIAN FILTER

- By definition

$$h_{\sigma^2}(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{x^2}{2\sigma^2}}$$

# HOW IS THE GAUSSIAN FILTER BUILT?

- An approximation is used to determine the size of the kernel in 2D

$$w \geq 3c$$

- We look for radial symmetry
- Where

$$c = 2\sqrt{2\sigma^2}$$

## EXAMPLE

- If we want a Gaussian filter with variance of  $\frac{1}{4}$

$$c = 2 \sqrt{\frac{2}{4}} \text{ then } w \approx 3$$

Then...

- If we want a Gaussian filter with variance of  $1/4$

$$c = 2\sqrt{\frac{2}{4}} \text{ then } w \approx 3$$

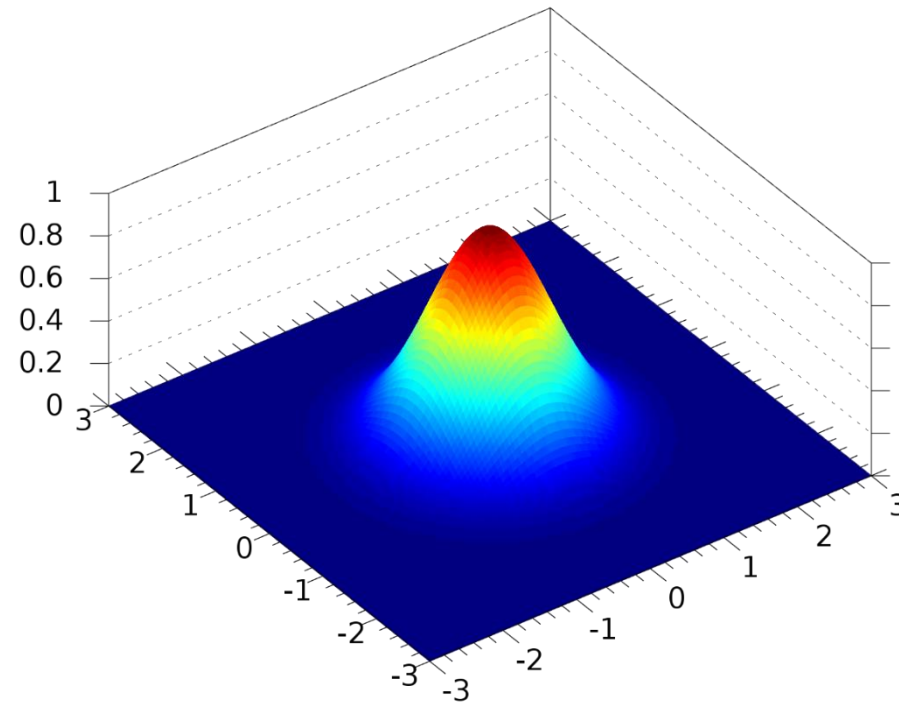
Then...

- $h_{\sigma^2}$

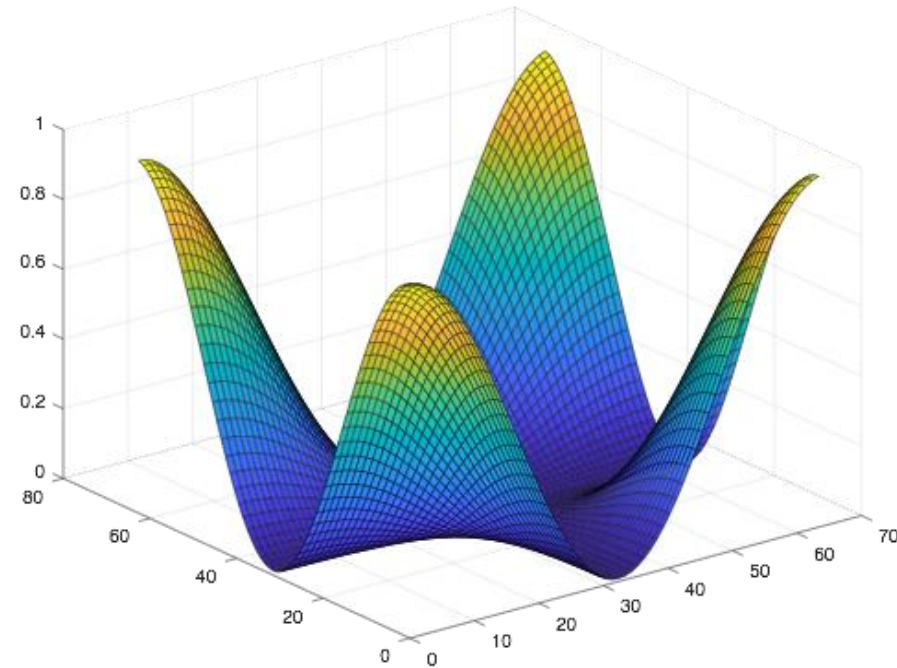
$$= \frac{\pi}{2} \begin{pmatrix} e^{-4} & e^{-2} & e^{-4} \\ e^{-2} & 1 & e^{-2} \\ e^{-4} & e^{-2} & e^{-4} \end{pmatrix}$$



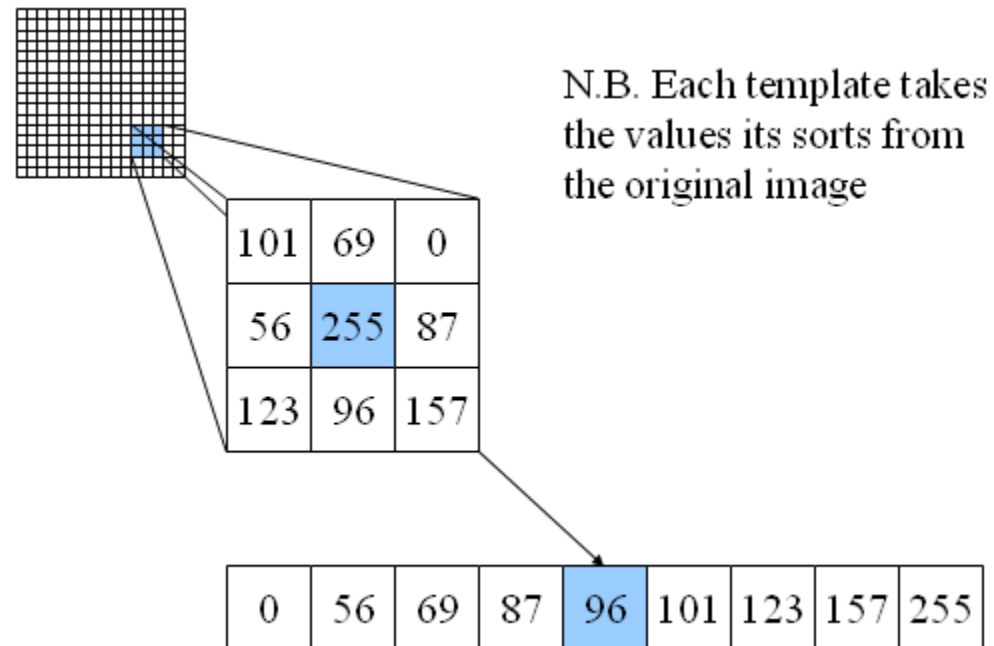
# GAUSSIAN FILTER IN TIME



# GAUSSIAN FILTER IN FREQUENCY



# MEDIAN FILTER



# MEDIAN FILTER

Noisy image



Median  $r=1$



Median  $r=5$



Median  $r=20$



# MEDIAN FILTER, COMPLEXITY?

Noisy image



Median  $r=1$



Median  $r=5$



Median  $r=20$



# REMEMBERING GAUSSIAN FILTER



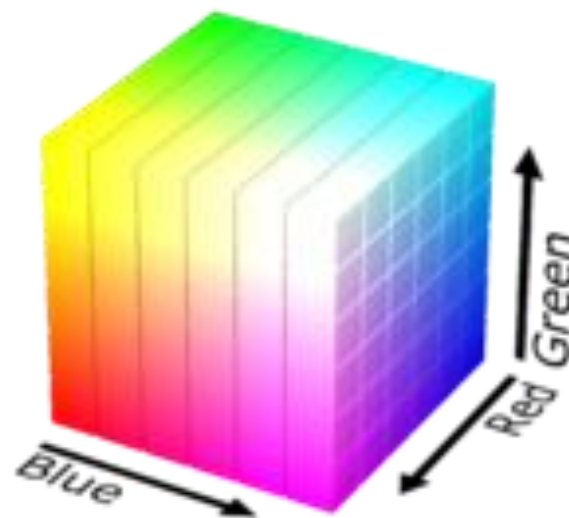
# QUALITY VS COMPLEXITY?

# COLOUR MODELS

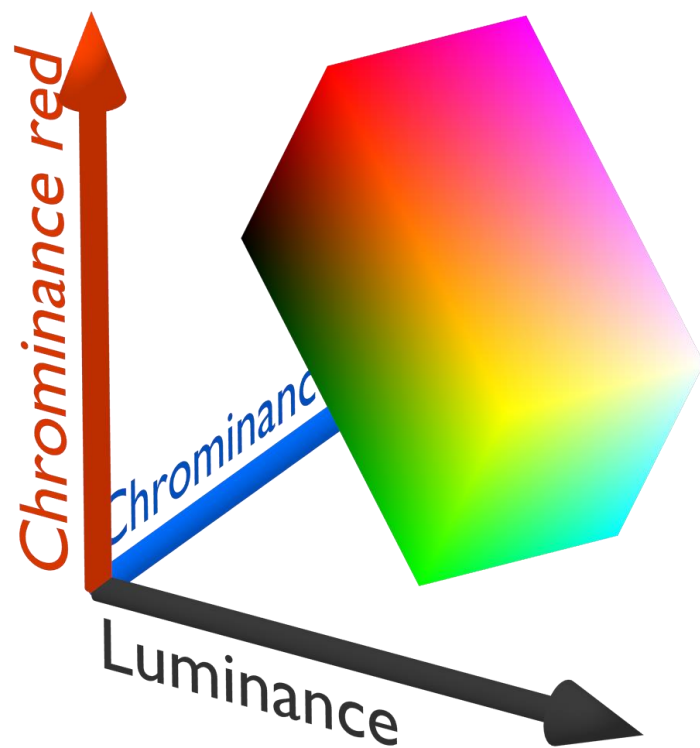
- RGB
- YCbCr
- HSV
- HSL



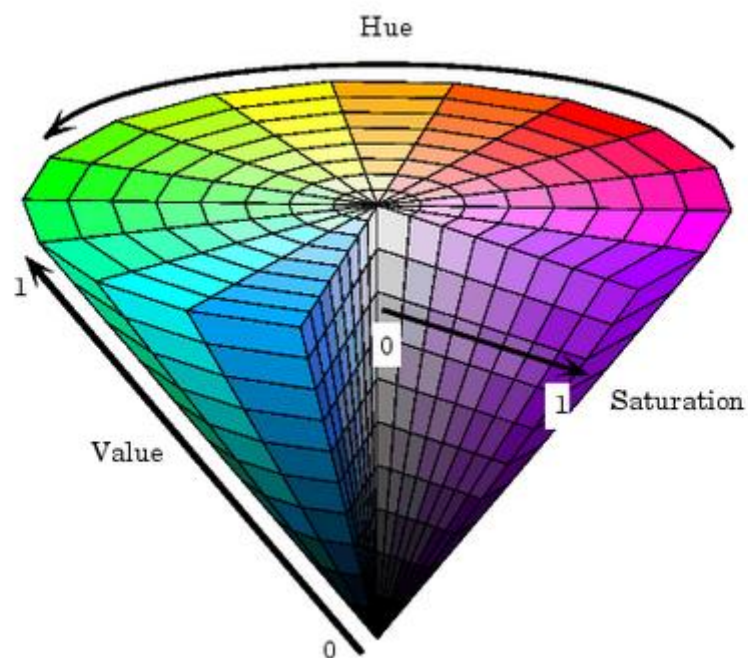
# RGB



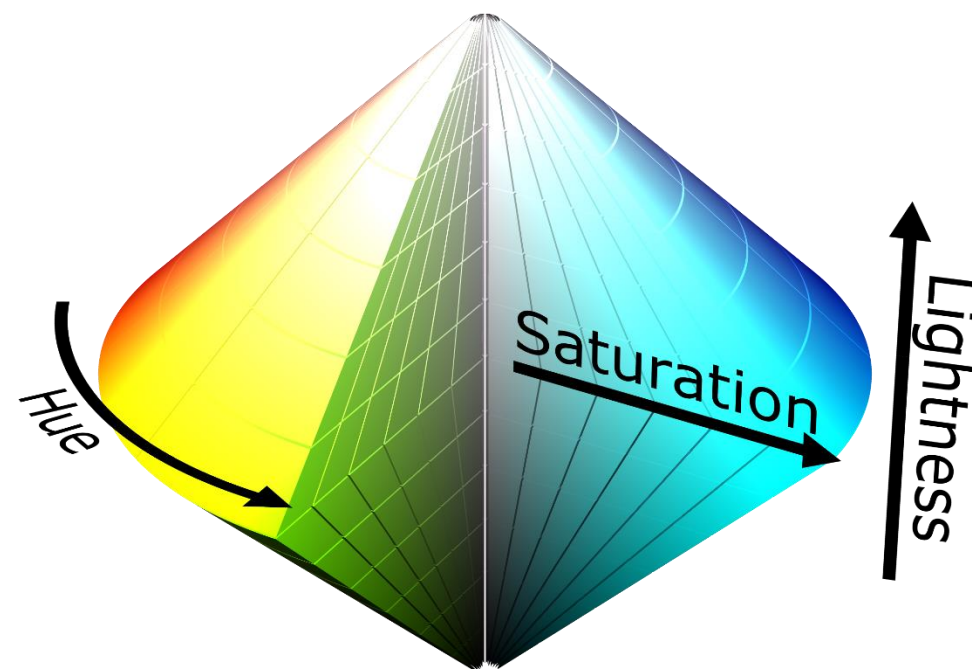
# YCBCR



# HSV



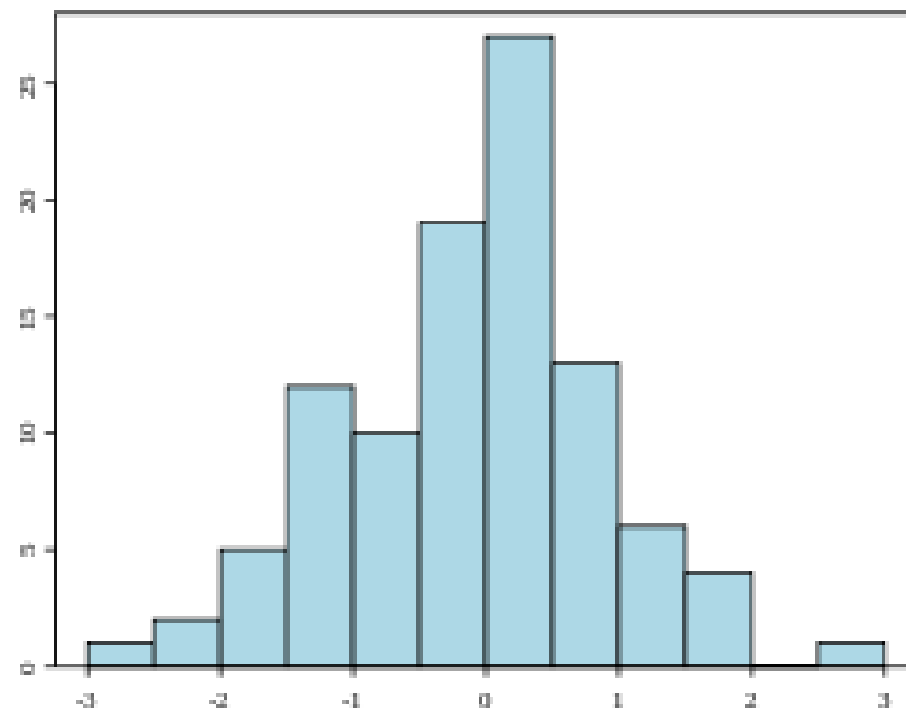
# HSL



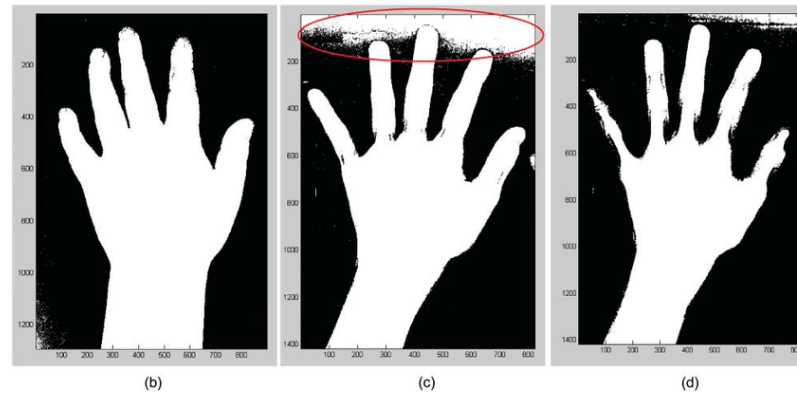
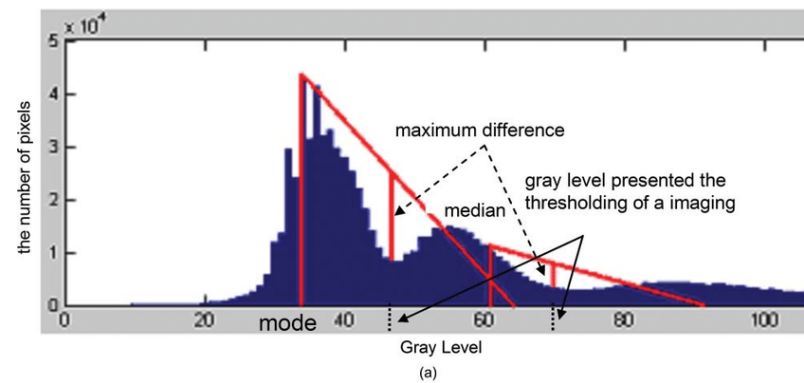
# HISTOGRAM

- In statistics, a histogram is a graphical representation of a variable in the form of bars, where the surface of each bar is proportional to the frequency of the represented values. They serve to obtain a general "first view", or panorama, of the distribution of the population, or of the sample, with respect to a characteristic, quantitative and continuous (such as length or weight).

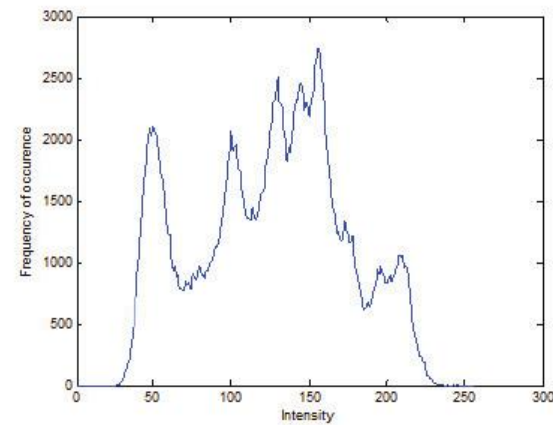
# HISTOGRAM



# HISTOGRAM



# BINARIZATION BY THRESHOLD



a	b
c	d



## PRACTICE 1.3

- Deliver the results of the following to a .pdf file:
  - Calculate the convolution mask for a Gaussian Variance Filter "I".
  - Display the Gaussian filter convolution mask in the frequency spectrum.
  - Make a script that receives images, video, or streaming from a camera and do the following:
    - Have an option to display the histogram (grayscale and/or "n" layers in a color model).
    - Get the binary representation of the input data.
      - The binarization can be grayscale or some color model.
      - You can perform binarization by threshold.
      - Otsu model.
      - etc.

# Questions?