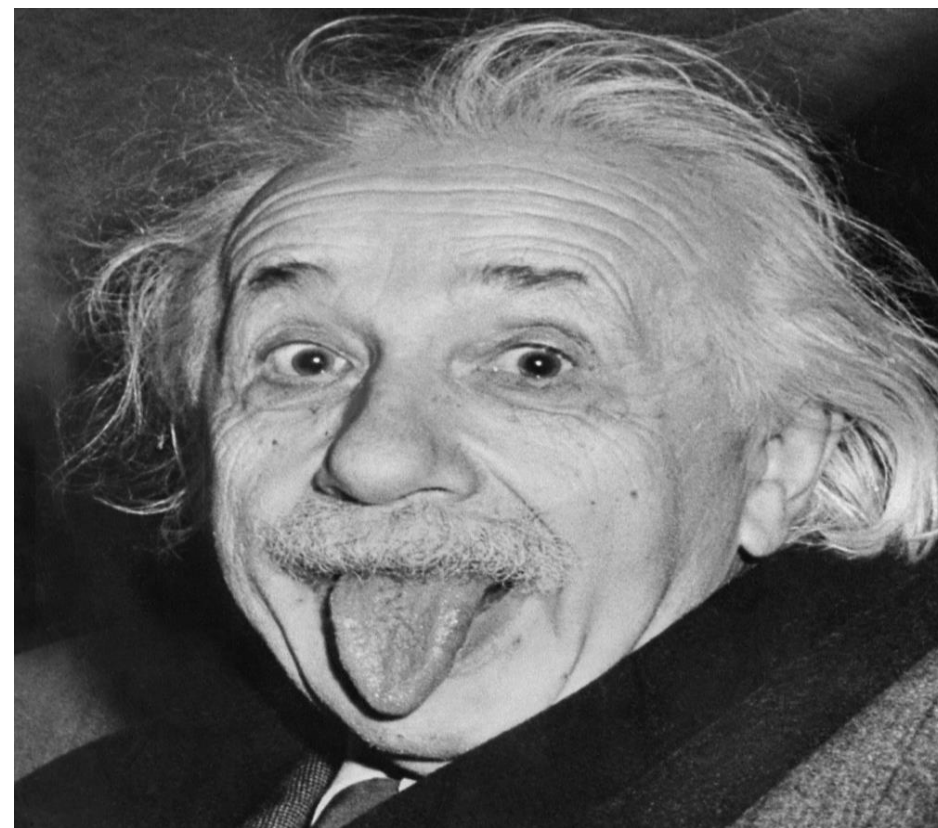


TÉCNICAS PARA UN MEJOR COMPUTER VISION

- Dr. Rubén Alvarez



- *"Everything human beings can imagine; nature has already created..."*
- -Albert Einstein.



WHAT IS COMPUTER VISION?



DataLab
Community

- Humans use our eyes and our brains to see and visually perceive the world around us. Computer Vision is the science that aims to provide a similar, if not better, capability to a machine or computer.



COMPUTER VISION



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1. Optics and 3D Reconstruction
2. Image Processing
3. Machine Learning in images

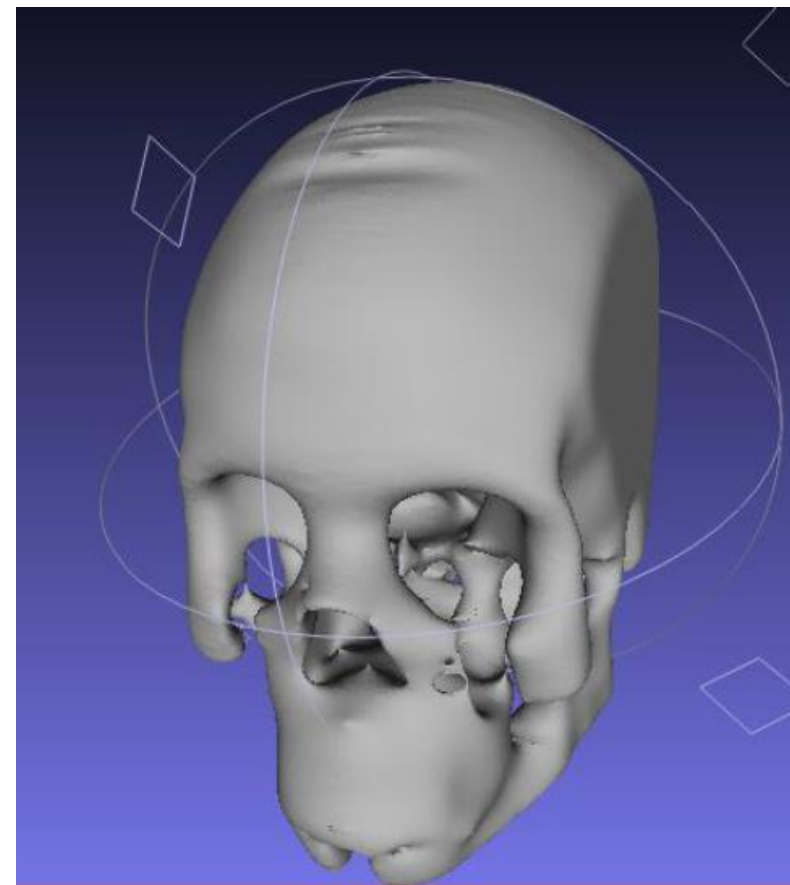
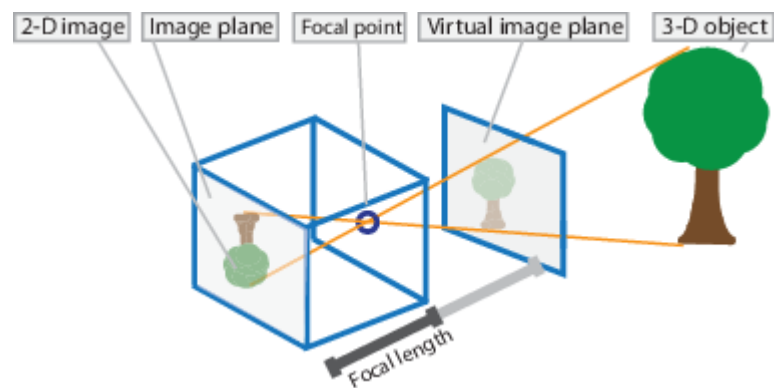


1. **Optics and 3D Reconstruction**
2. Image Processing
3. Machine Learning in images

OPTICS AND 3D RECONSTRUCTION



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COMPUTER VISION

3D CLOUD SAMPLES



1. Optics and 3D Reconstruction
- 2. Image Processing**
3. Machine Learning in images

IMAGE PROCESSING



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COMPUTER VISION



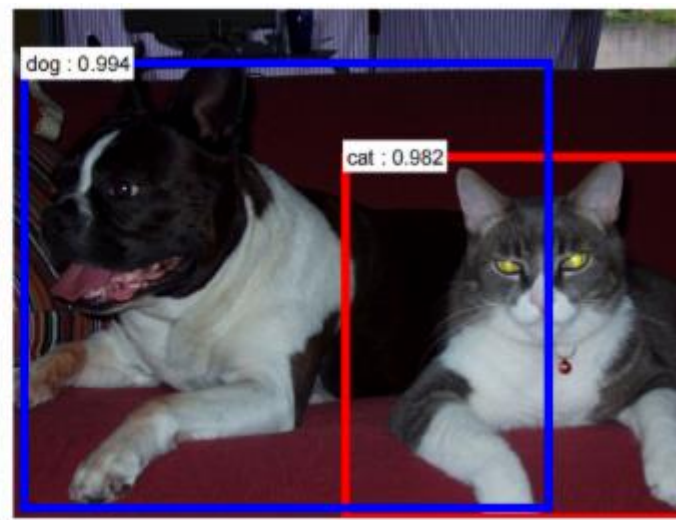
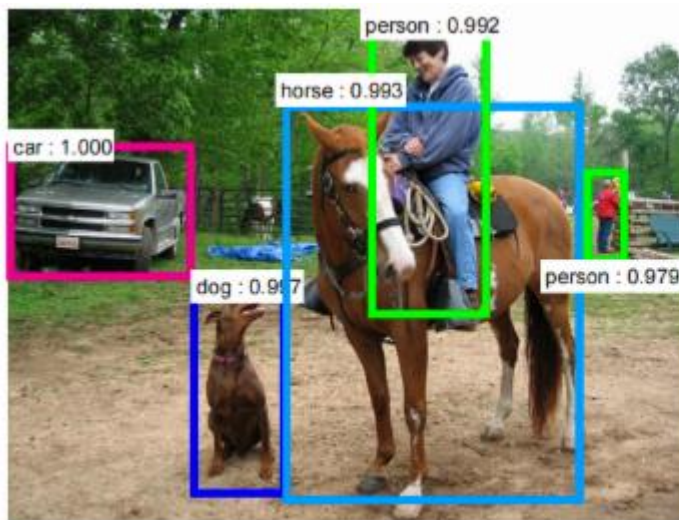
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1. Optics and 3D Reconstruction
2. Image Processing
3. **Machine Learning in images**

OBJECT RECOGNITION



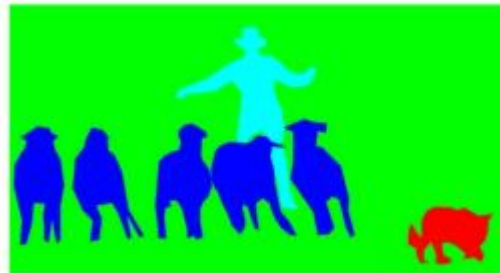
DataLab
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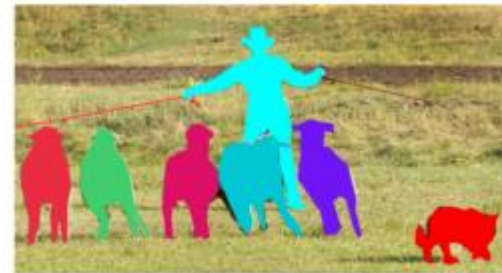
SEGMENTATION



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(c) Semantic segmentation



(d) Instance segmentation

STYLE TRANSFER



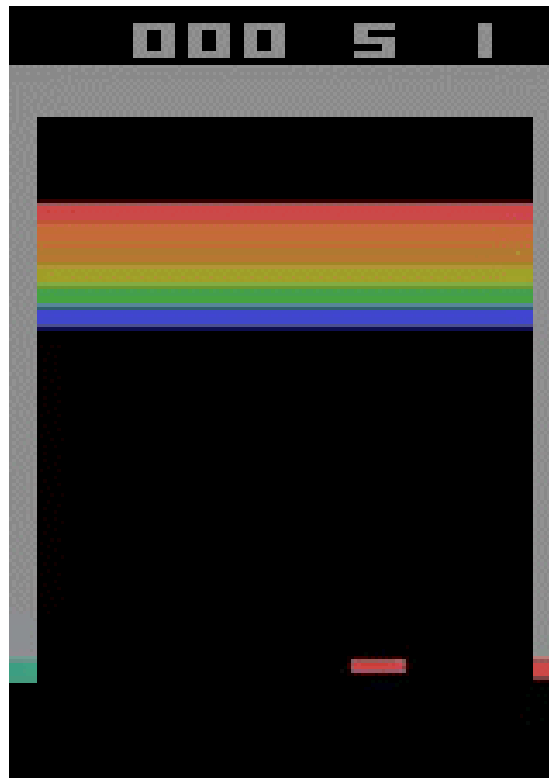
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VIDEO GAMES



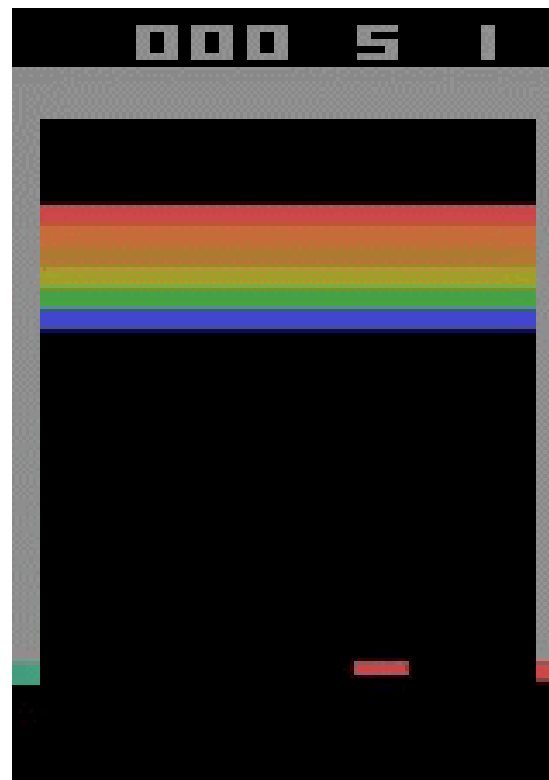
DataLab
Community



VIDEO GAMES



DataLab
Community



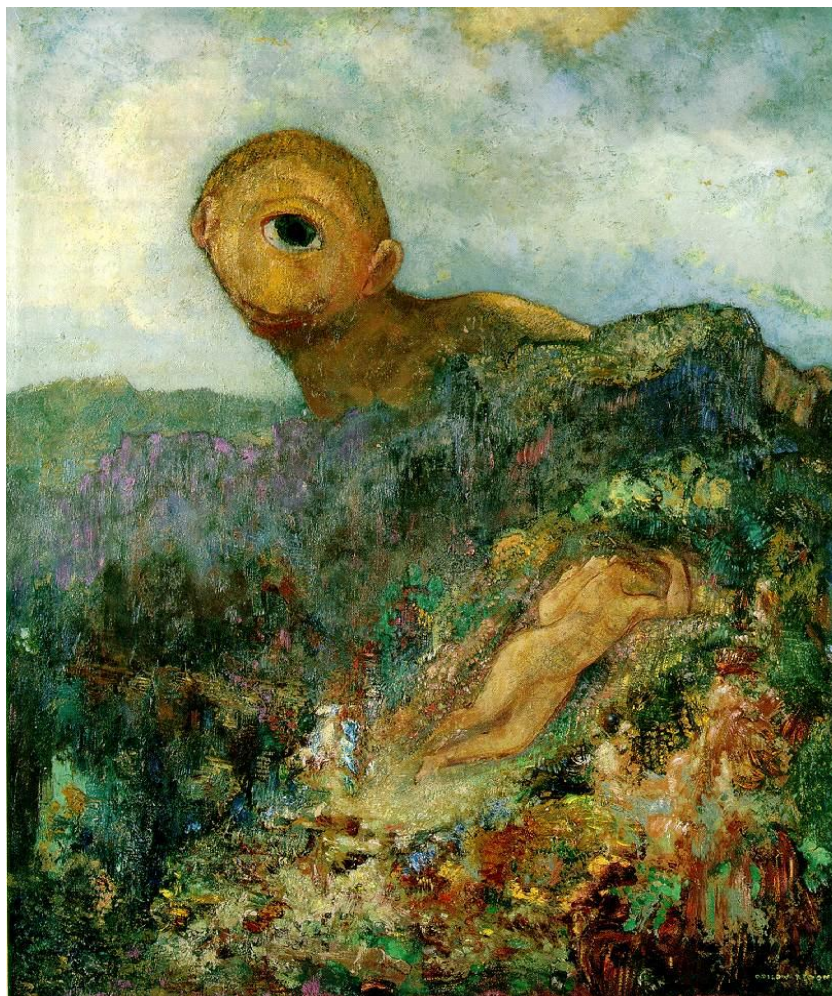
COMPUTER VISION

PERCEPCIÓN 3D

SINGLE-VIEW GEOMETRY



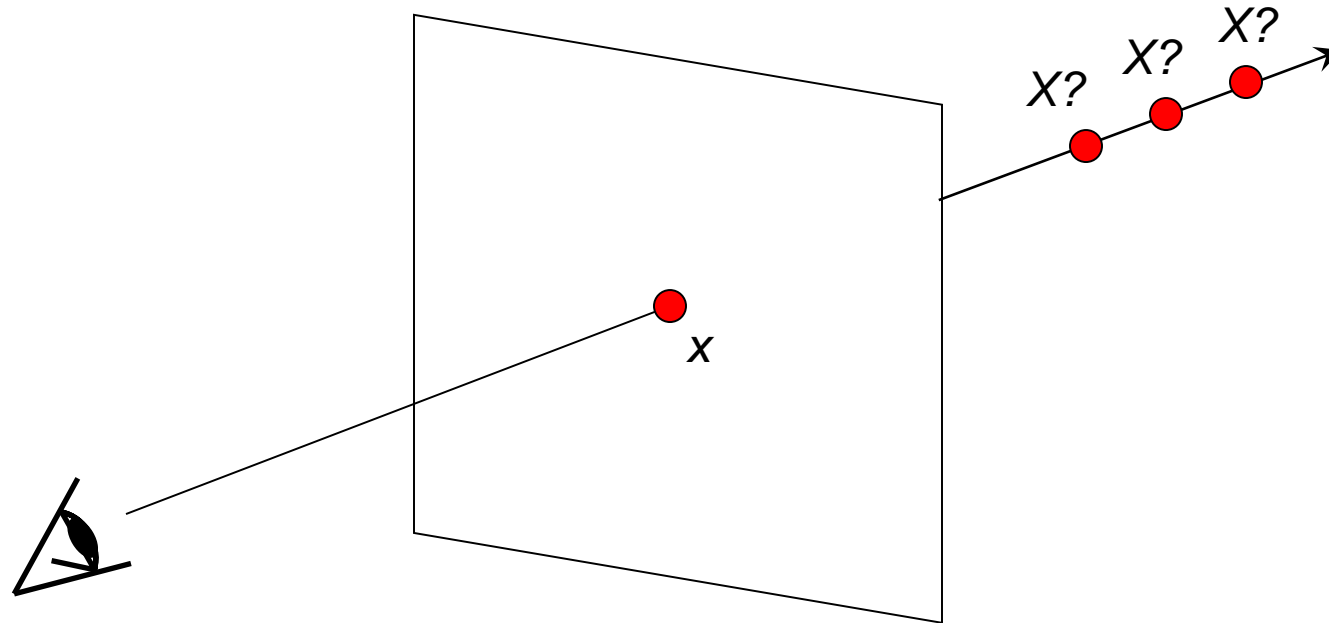
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Community



Odilon Redon, *Cyclops*, 1914

OUR GOAL: RECOVERY OF 3D STRUCTURE

- Is recovering the structure of an image unambiguous?



OUR GOAL: RECOVERY OF 3D STRUCTURE



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Community

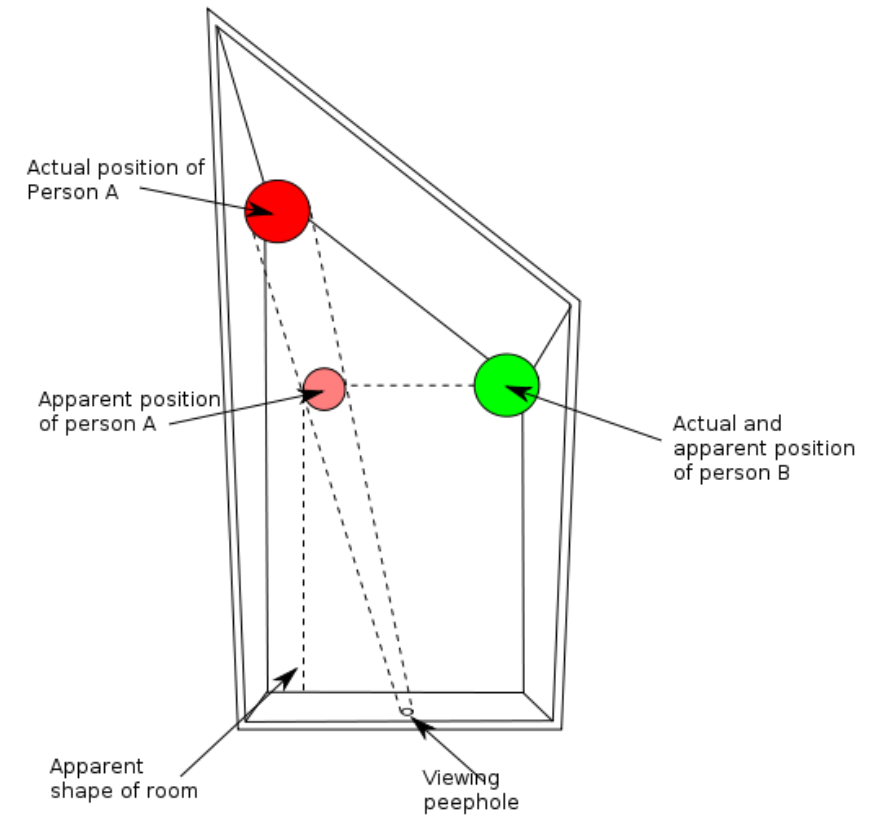
- What about perspective?



AMES ROOM



DataLab
Community



AMES ROOM



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Community



VIRTUAL REALITY LIGHTING



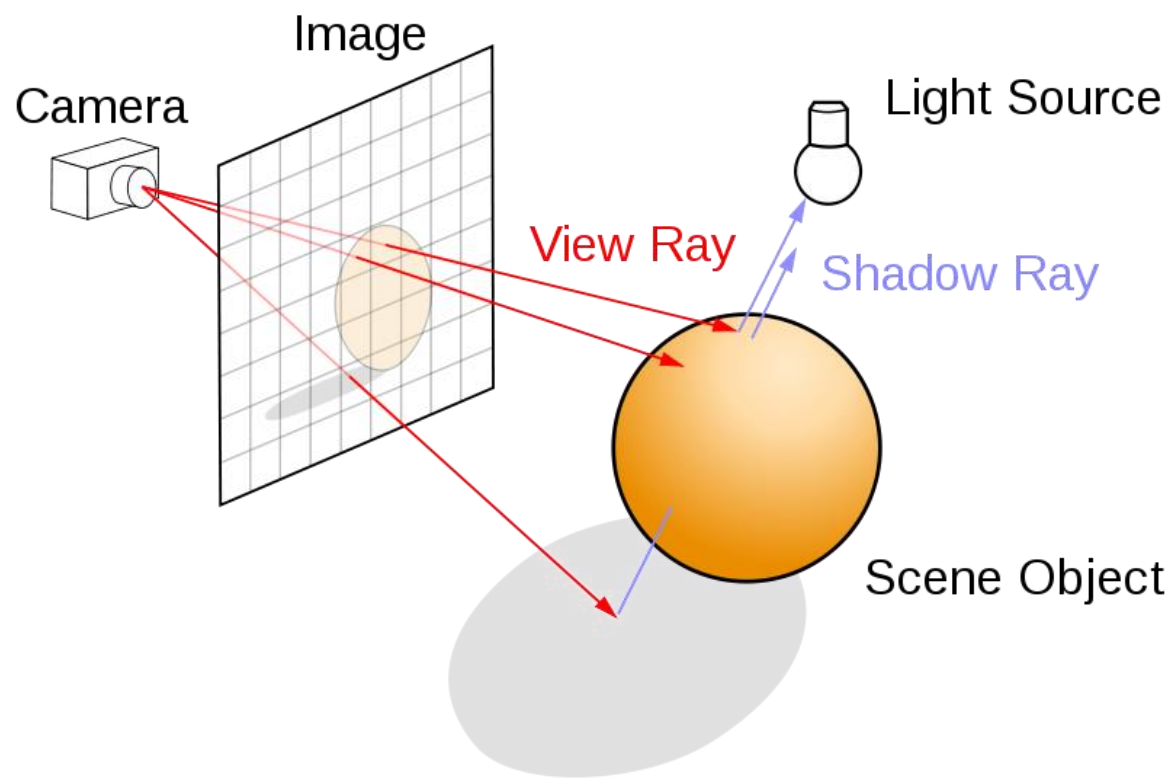
DataLab
Community

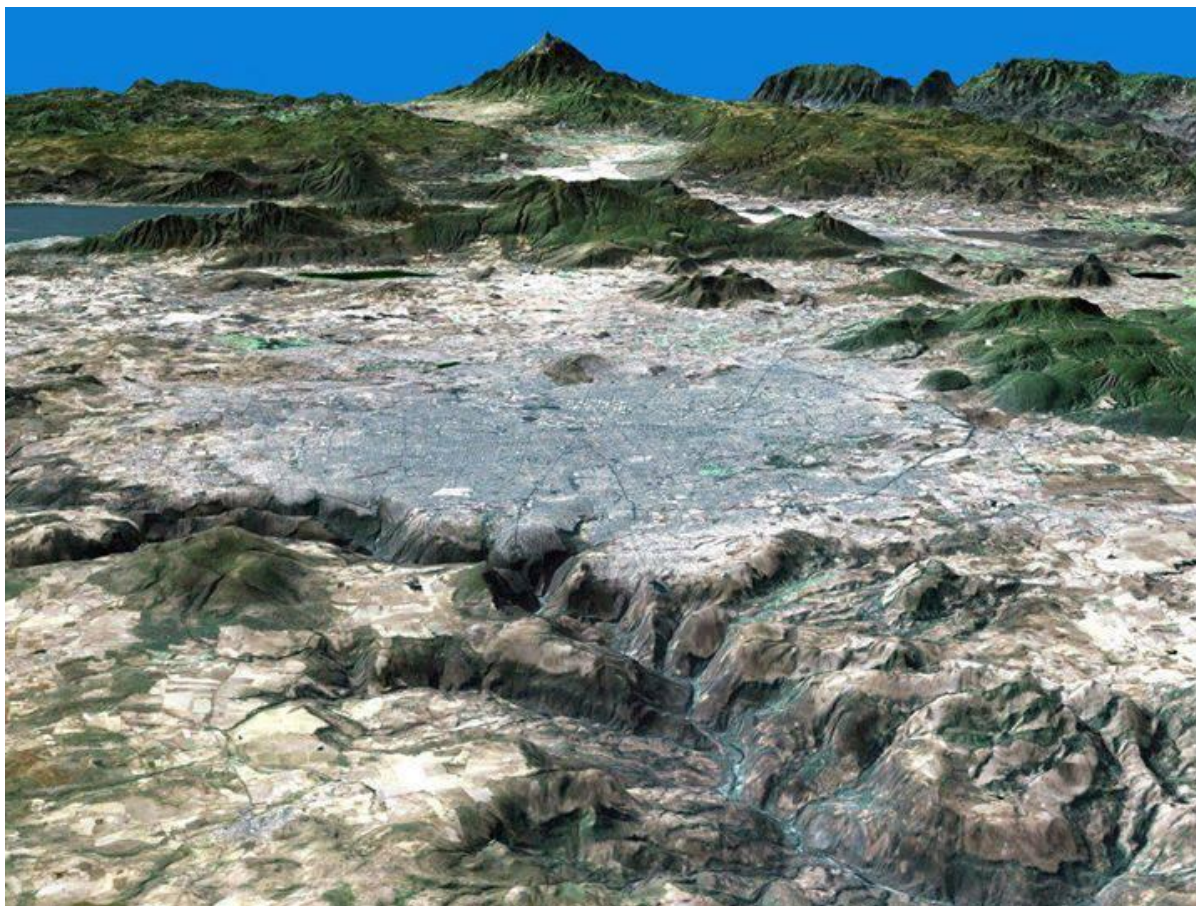


RAY TRACING



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WHAT IF WE JUST WORK WITH THE CENTER OF THE LENS?



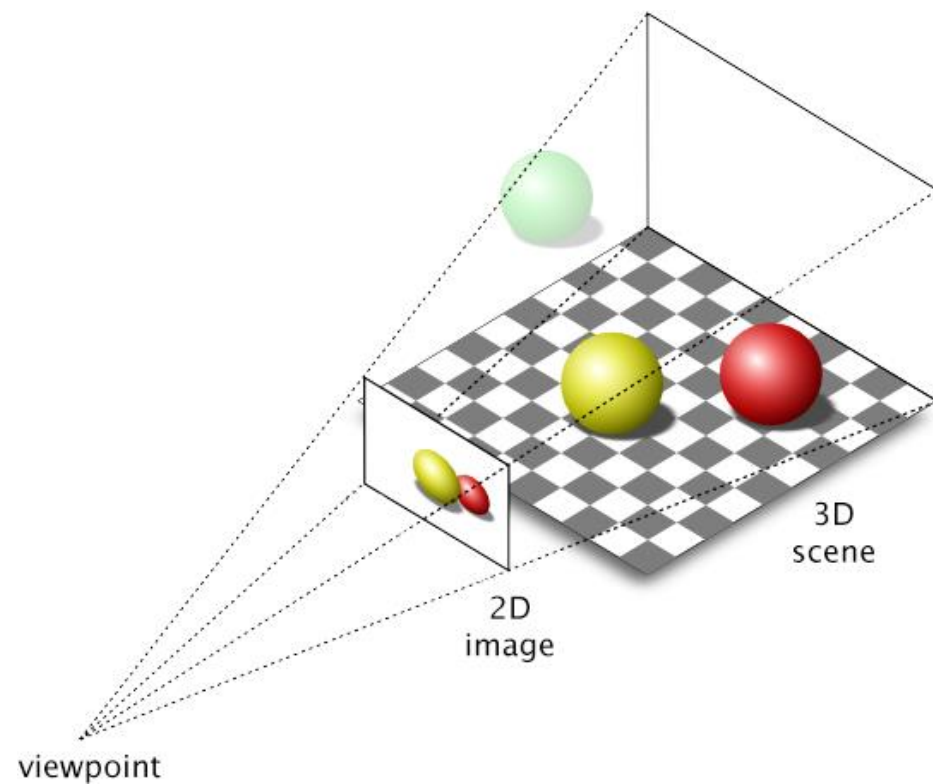
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MOVE INFORMATION FROM 3D TO 2D



DataLab
Community

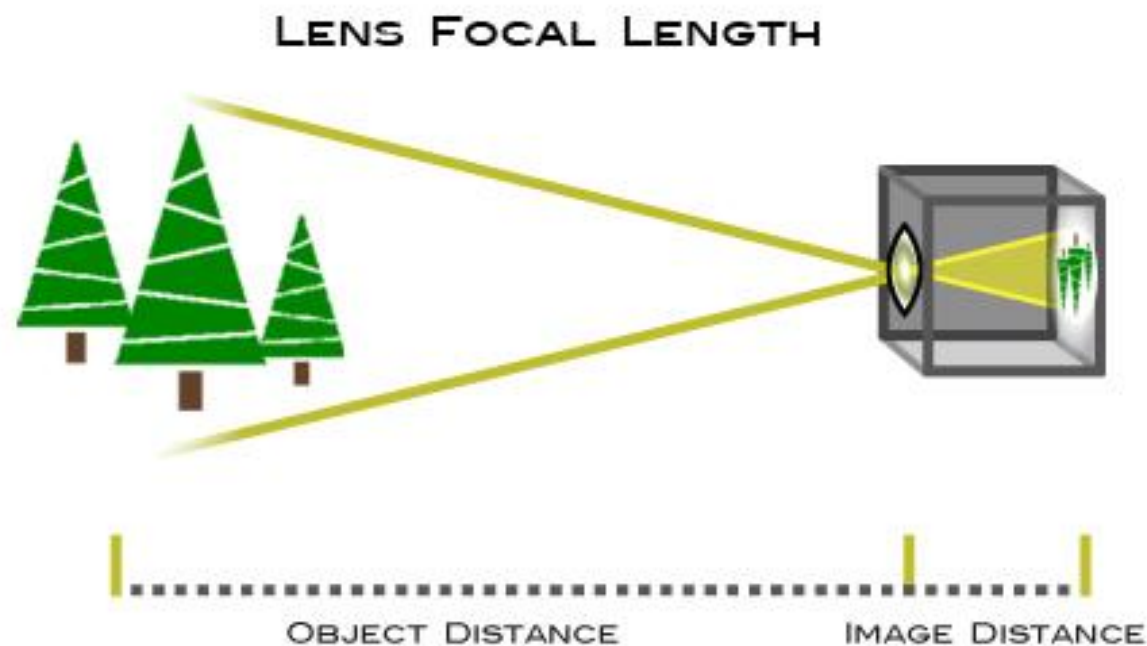


DOES OPENNESS MATTER?



DataLab
Community

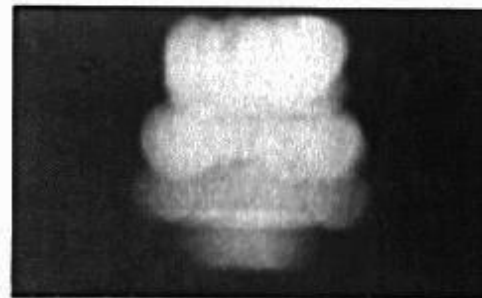
- In the case of the above drawing, is it important to enter the dimensions of that opening where the light enters the box?



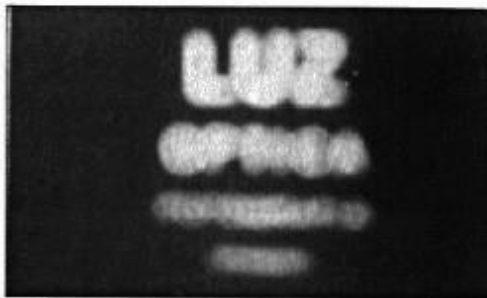
DOES OPENNESS MATTER?



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2 mm



1 mm



0.6mm

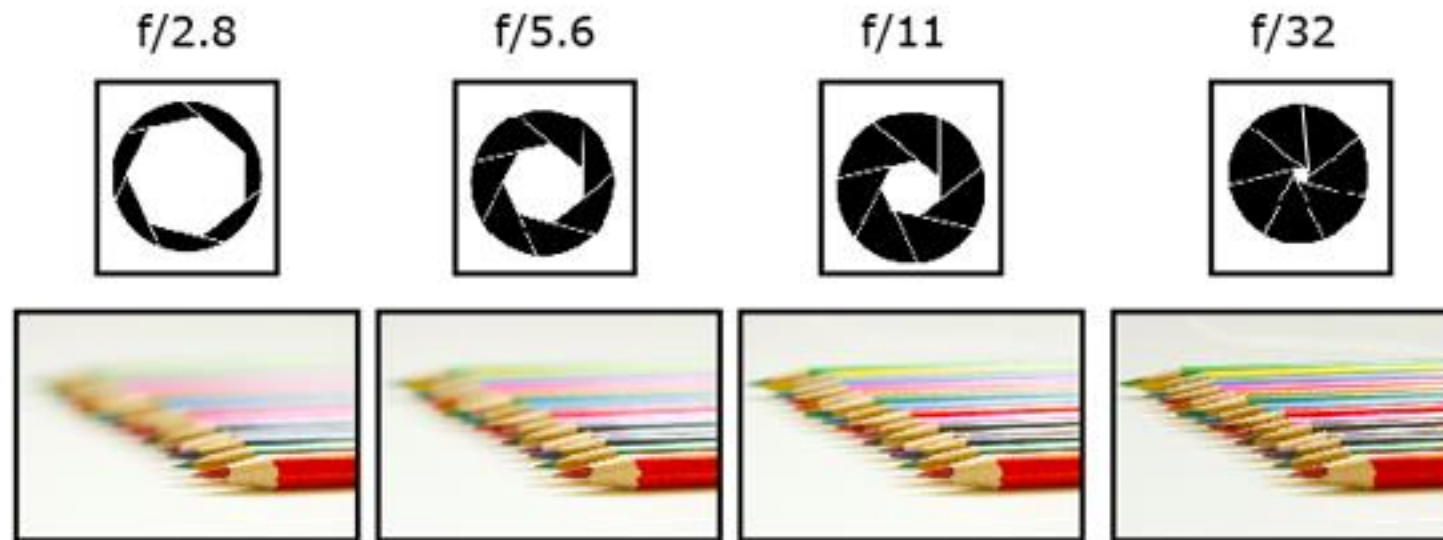


0.35 mm

OPENNESS AND DEPTH ARE CORRELATED



DataLab
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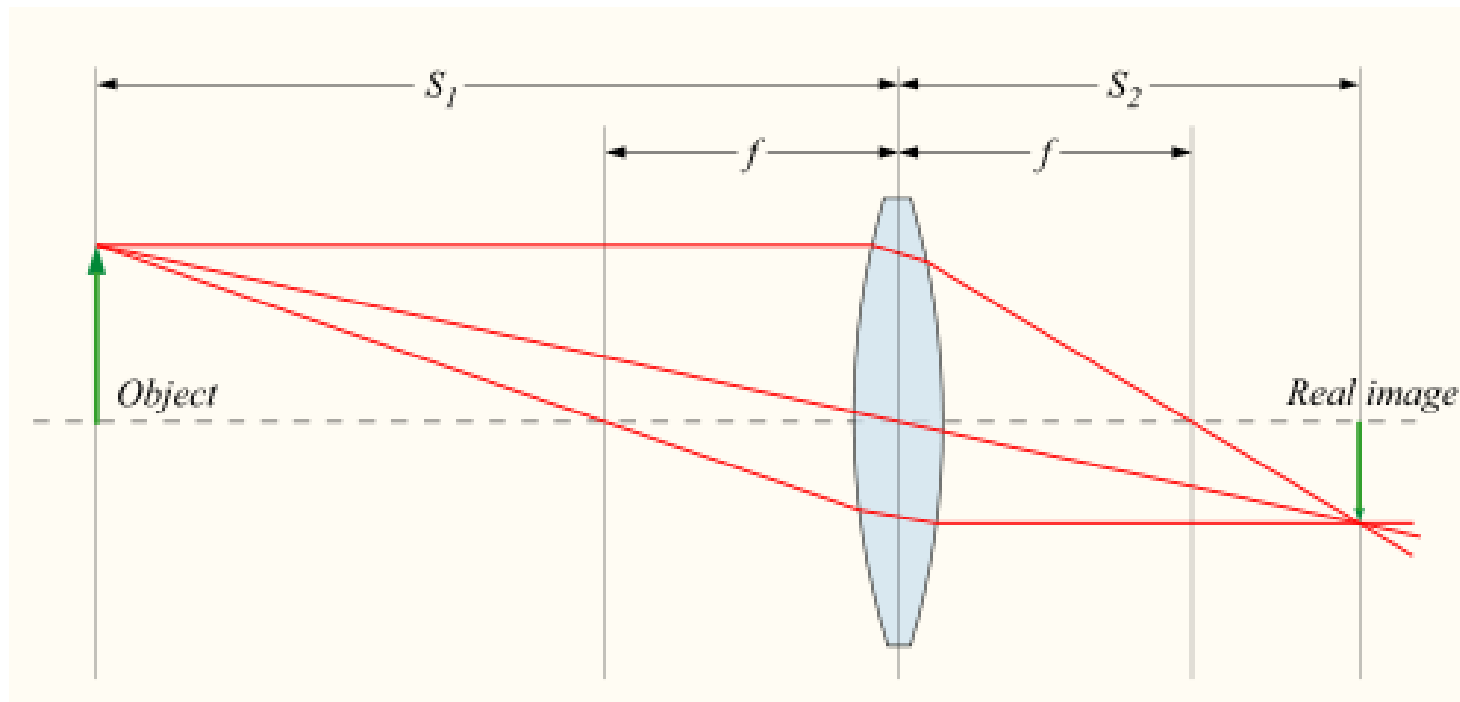


THIN LENSES



DataLab
Community

- Gaussian Law $\frac{1}{z} + \frac{1}{z} = \frac{1}{f}$



CAMERA CALIBRATION



DataLab
Community

$$\blacksquare \begin{pmatrix} p_{xi} \\ p_{yi} \\ 1 \end{pmatrix} = \begin{pmatrix} \frac{f}{dx} & -\rho \frac{f}{dy} & c_{xi} \\ 0 & \frac{f}{dy} & c_{yi} \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} X_w \\ Y_w \\ Z_w \end{pmatrix}$$

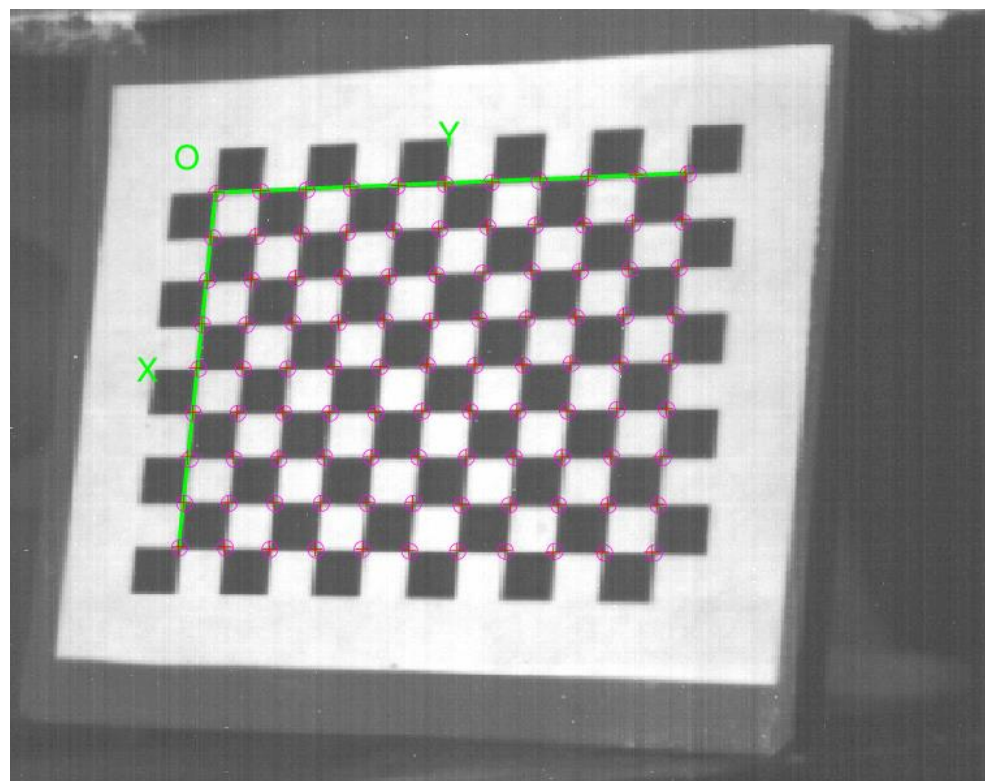
COMPUTER VISION

DEMO ARUCO

CHECKERBOARD



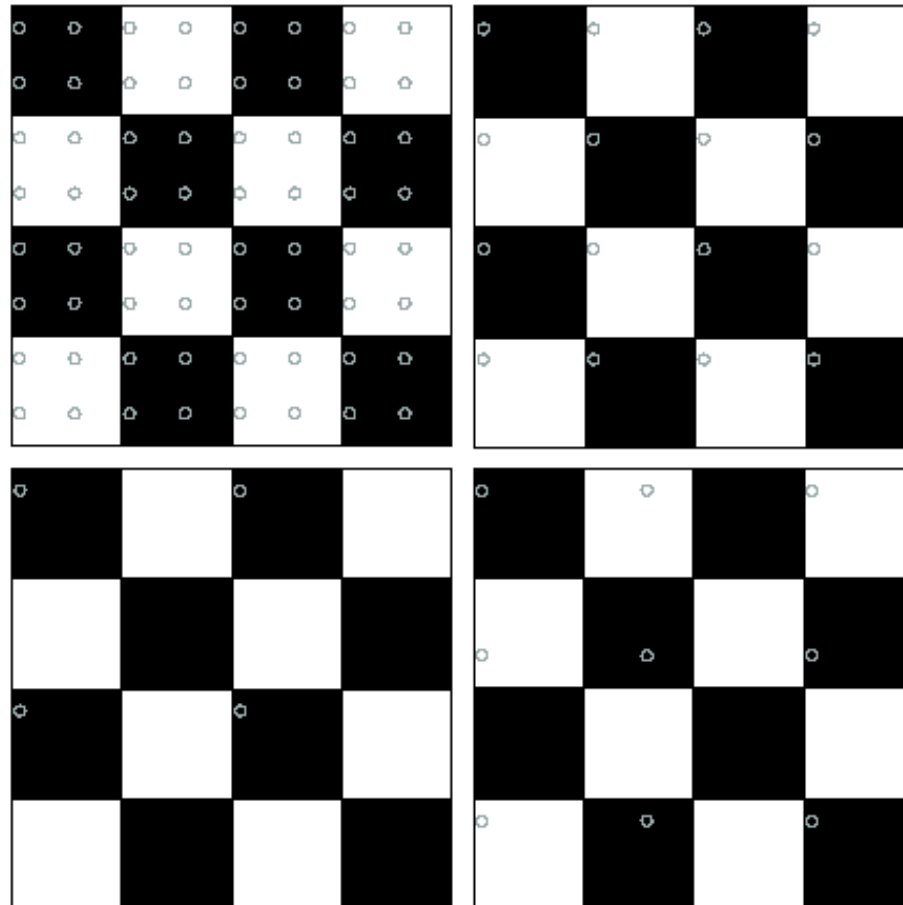
DataLab
Community

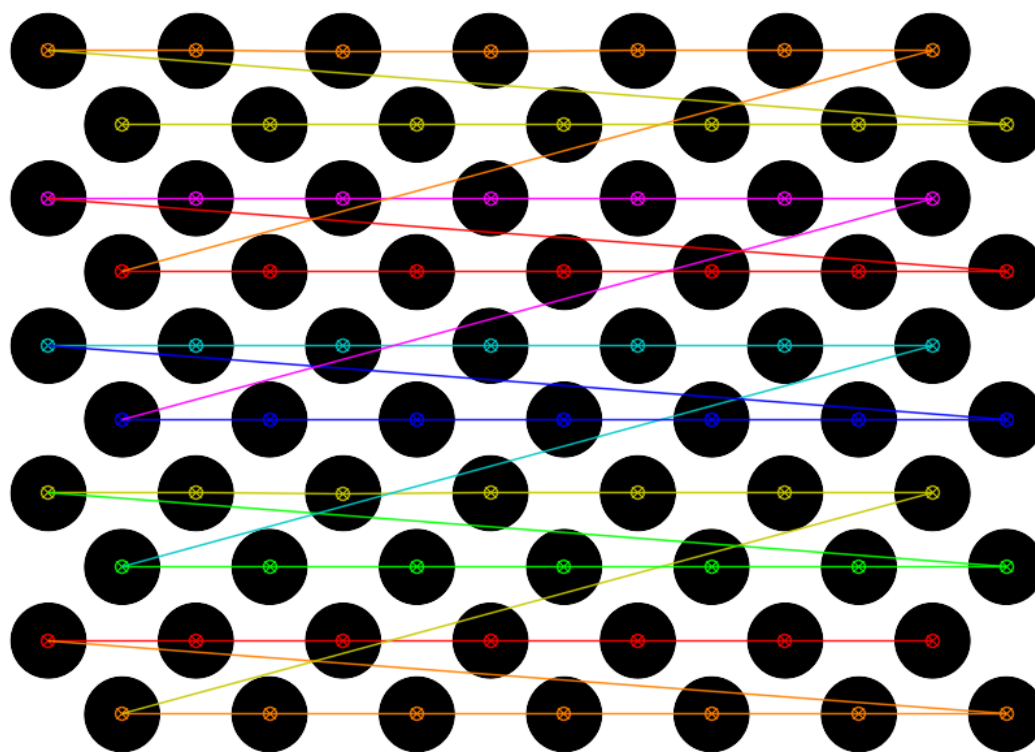


CHECKERBOARD



DataLab
Community



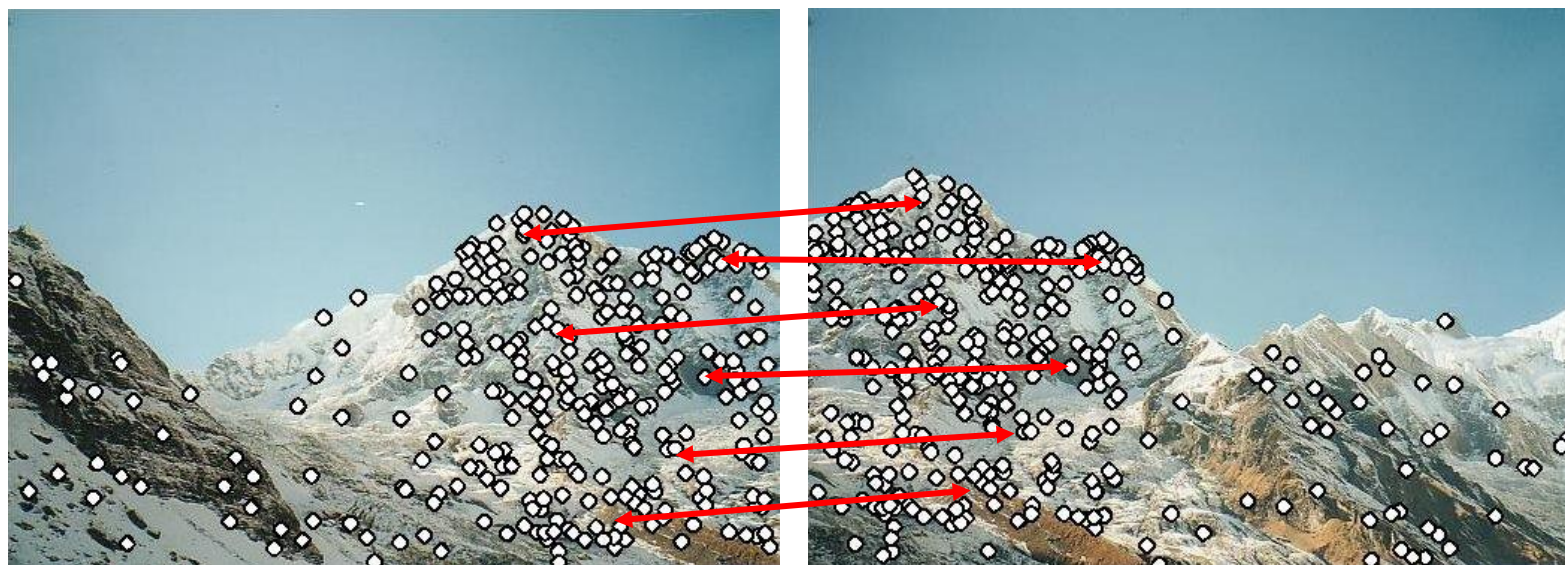


FEATURE MATCHING



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Community







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Community



COMPUTER VISION

DEMO MATCHING

WHAT IS CONVOLUTION?

REMEMBER...



AVERAGE FILTER



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Community

- 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...



DataLab
Community

$F[x, y]$

| | | | | | | | | | | |
|---|---|----|----|----|----|----|----|---|---|---|
| 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 | | | | | | | | | |
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AND IN PICTURES...



DataLab
Community

$F[x, y]$

| | | | | | | | | | |
|---|---|----|----|----|----|----|----|---|---|
| 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 | | | | | | | |
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| | | | | | | | | | |

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 | | | | | | |
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AND IN PICTURES...



DataLab
Community

$F[x, y]$

| | | | | | | | | | |
|---|---|----|----|----|----|----|----|---|---|
| 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 | | | | | |
| | | | | | | | | | |
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| | | | | | | | | | |

AND IN PICTURES...



DataLab
Community

$F[x, y]$

| | | | | | | | | | |
|---|---|----|----|----|----|----|----|---|---|
| 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 | | | | | |
| | | | | | | | | | |
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AND IN PICTURES...



DataLab
Community

$$F[x, y]$$

| | | | | | | | | | |
|---|---|----|----|----|----|----|----|---|---|
| 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 | |
| | | | | | | | | | |

WHAT IS CONVOLUTION?



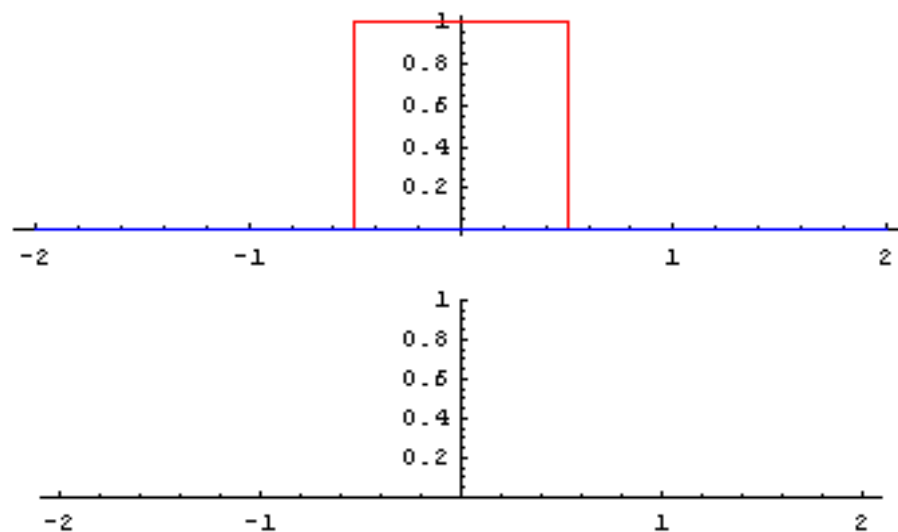
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Community

$$(f * g)(t) = \int_{-\infty}^{\infty} f(\eta)g(t - \eta) d\eta$$

WHAT IS CONVOLUTION?



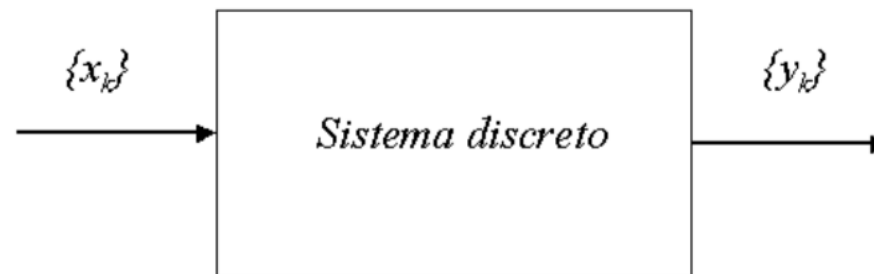
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WHY THE CONVOLUTION?



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WHY THE CONVOLUTION?



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- Filtros FIR (Finite Impulse Response)
- Filtros IIR (Infinite Impulse Response)

CONVOLUTION THEOREM



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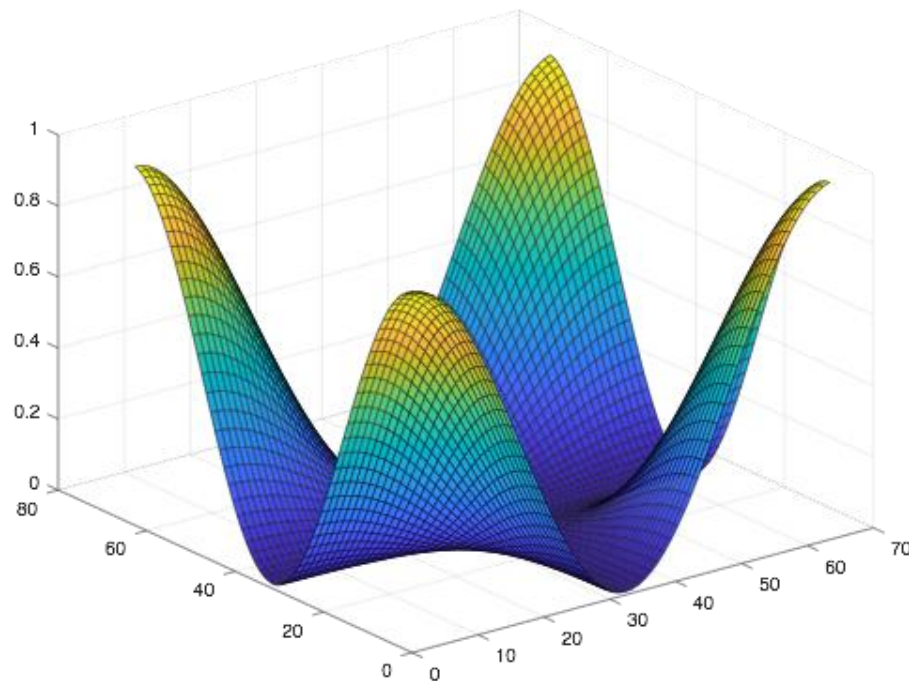
$$\mathcal{F}(f * g) = F(\omega)G(\omega)$$

FILTERS IN WAVE NUMBER DOMAIN



DataLab
Community

$$\frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

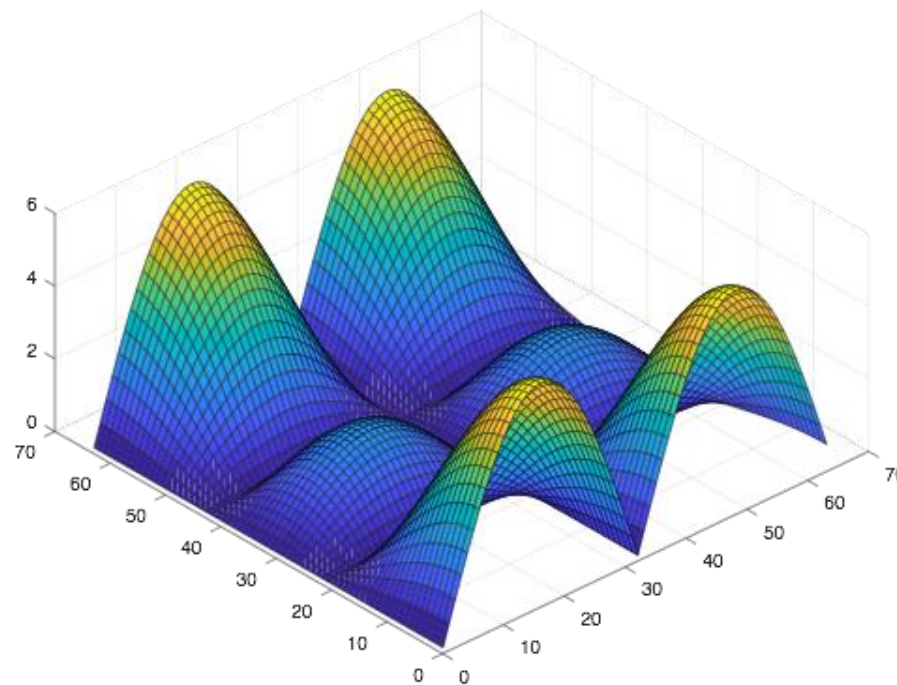


FILTERS IN WAVE NUMBER DOMAIN



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Community

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

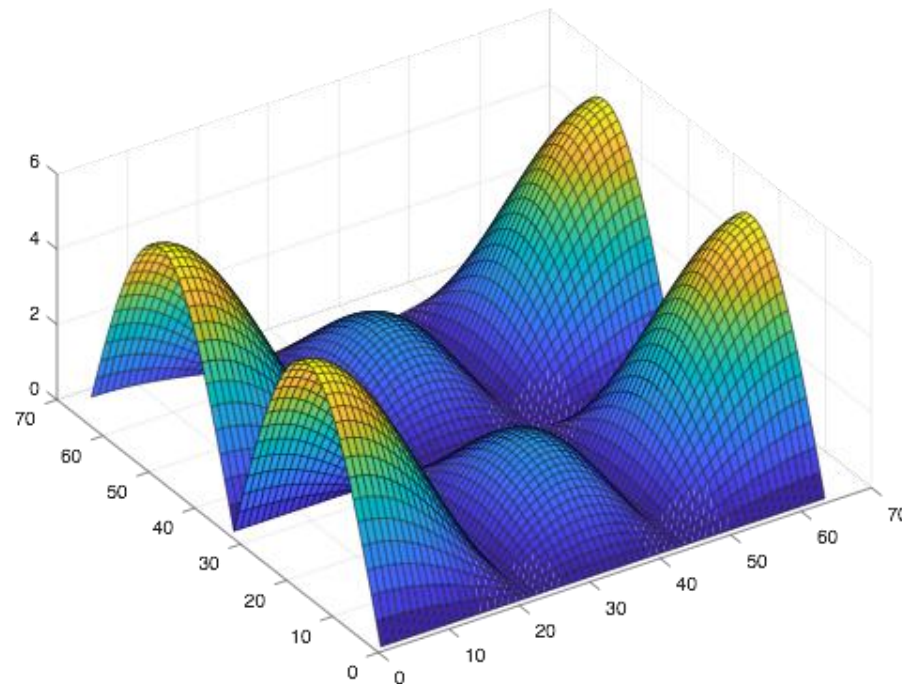


FILTERS IN WAVE NUMBER DOMAIN



DataLab
Community

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



COMPUTER VISION

DEMO EDGES

IMAGE PROCESSING VS CNN

COMPUTER VISION

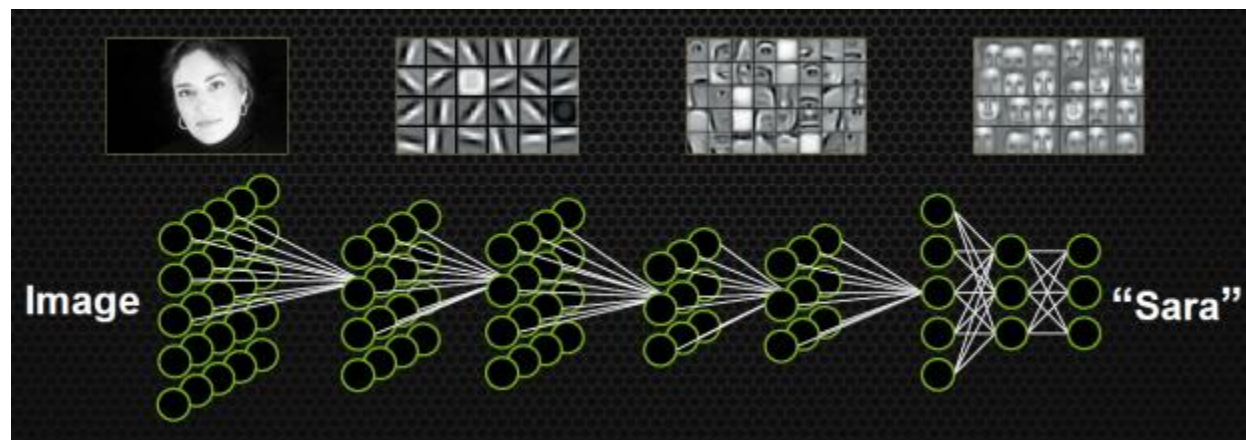


DataLab
Community

CONVOLUTIONAL NEURAL NETWORKS



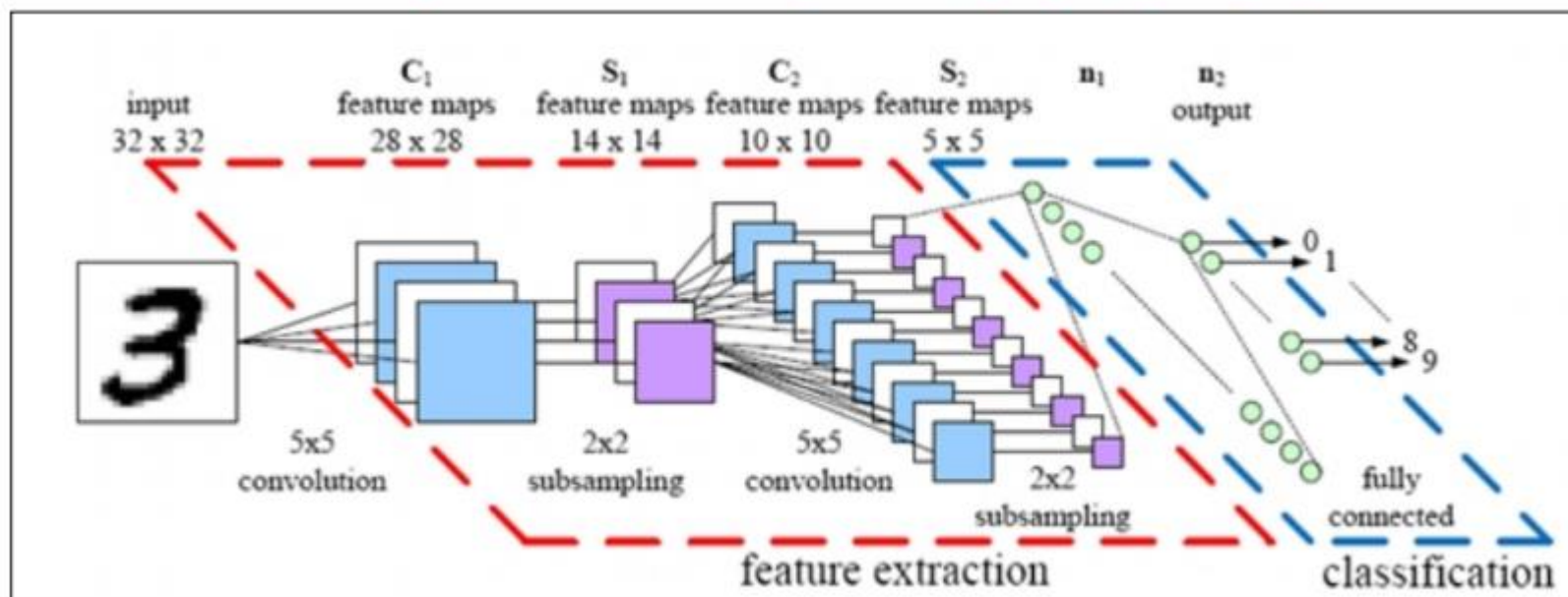
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CNN



DataLab
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ARCHITECTURE OF A CNN



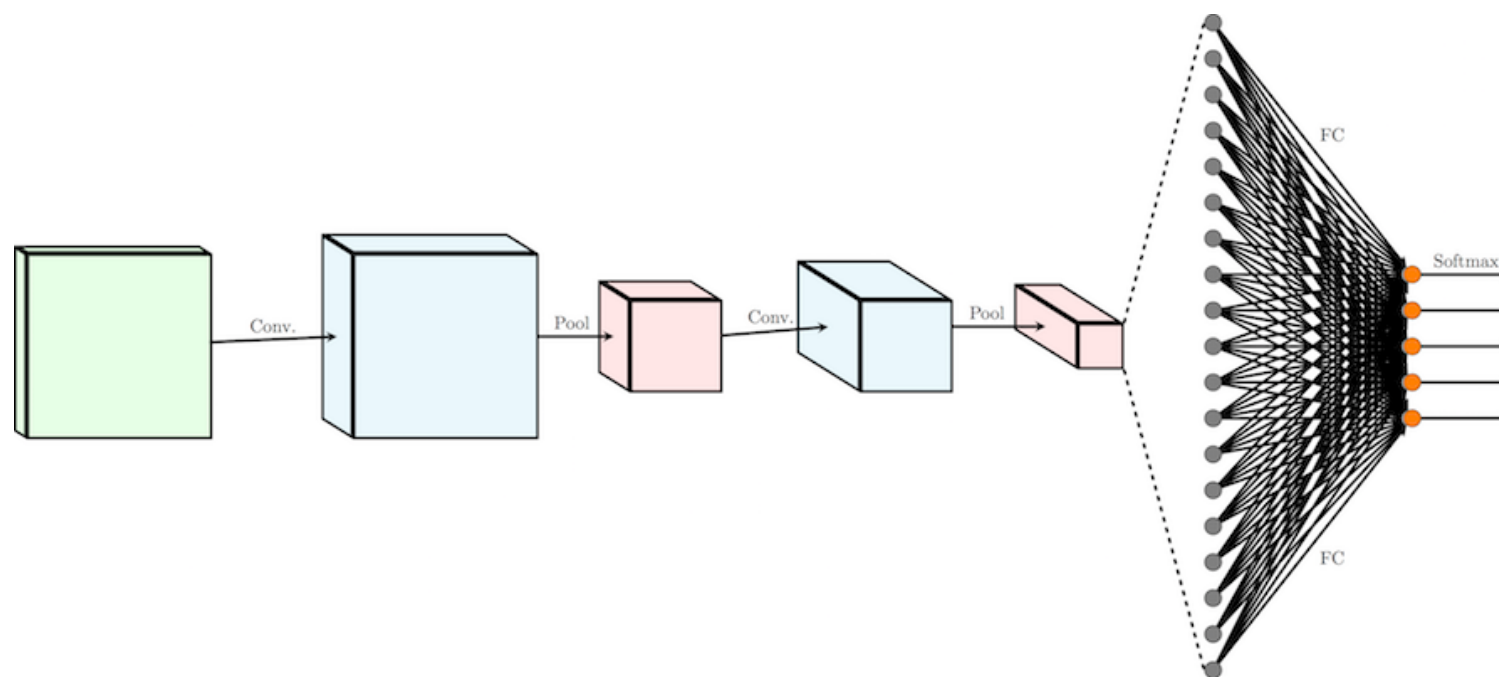
DataLab
Community

- Convolutional Layer
- Pooling Layer
- Flatten Layer
- Activation Layer
- Fully Connected Layer
- ...

CONVOLUTIONAL NEURAL NETWORKS



DataLab
Community

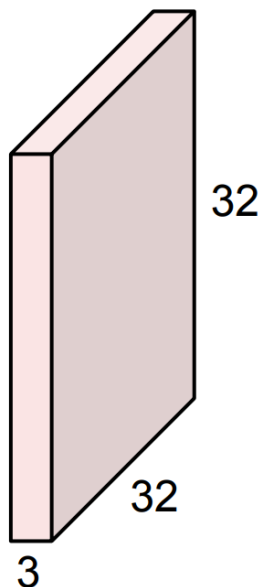


CONVOLUTIONAL LAYER



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Community

32x32x3 image



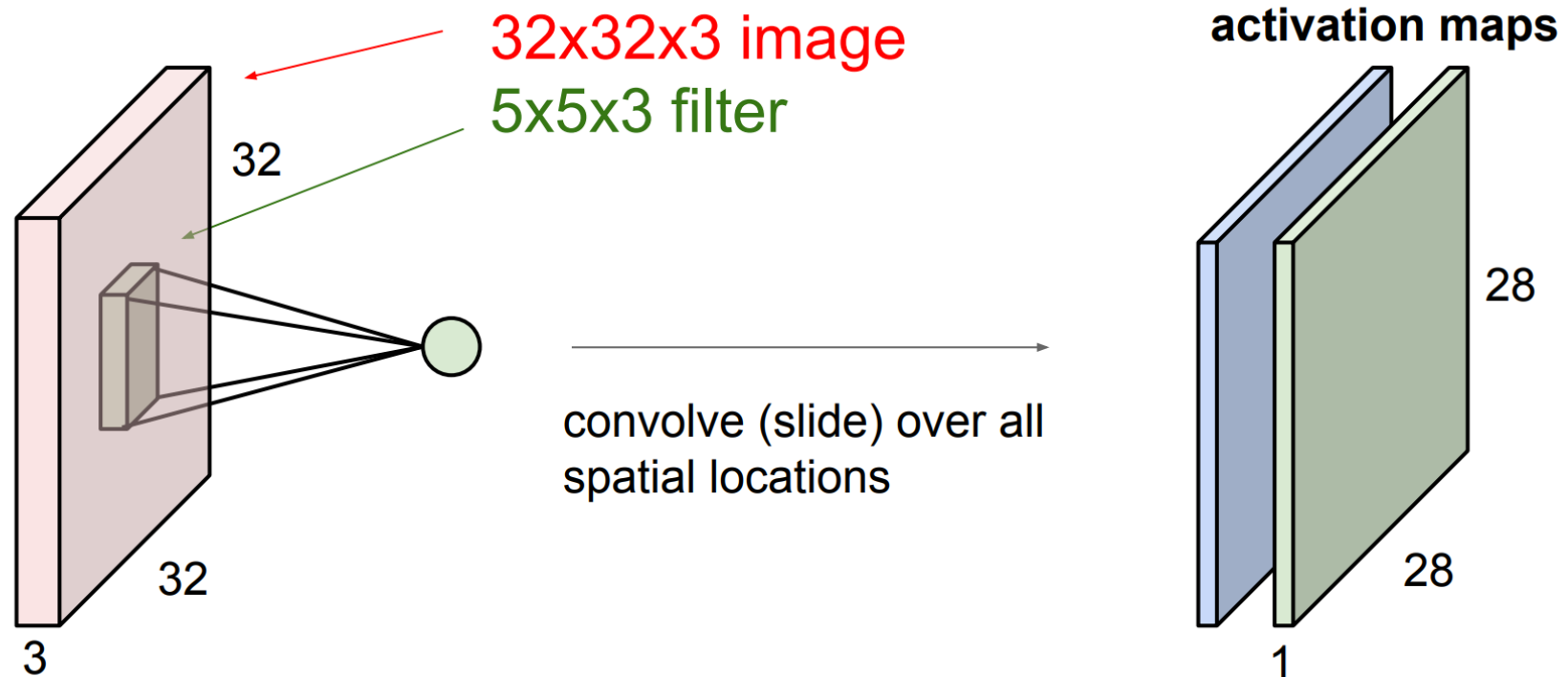
5x5x3



ACTIVATION MAP CNN



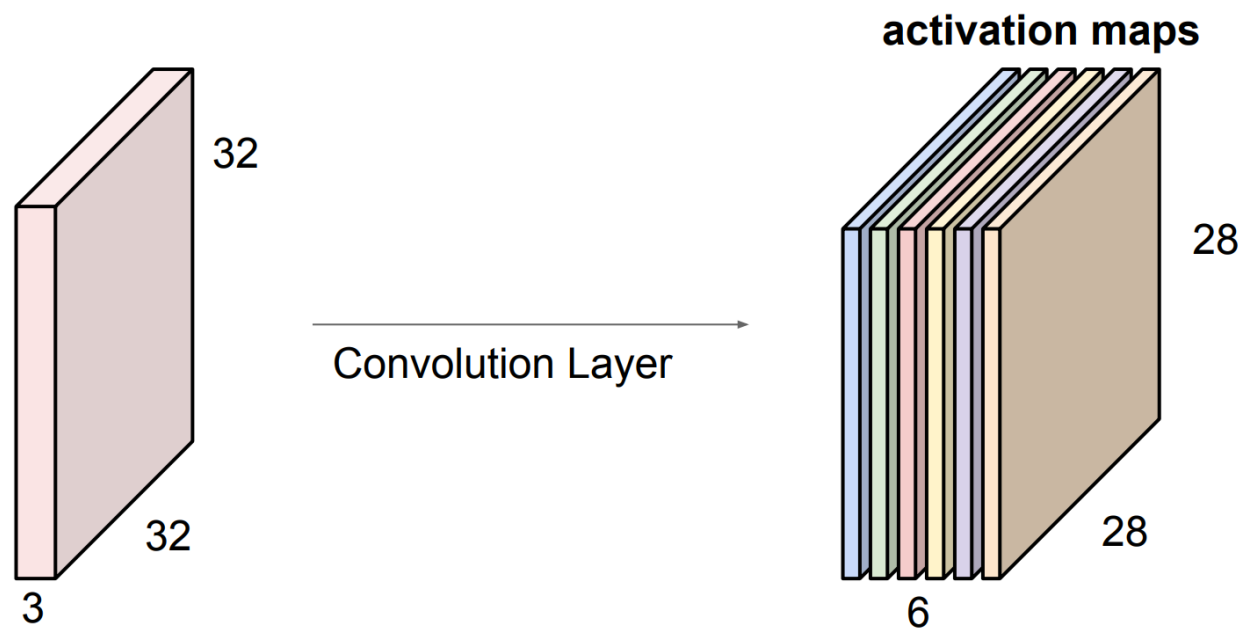
DataLab
Community



ACTIVATION MAP



DataLab
Community



We stack these up to get a “new image” of size 28x28x6!

CONVOLUTIONAL LAYER



DataLab
Community

| | | | | |
|-----------------|-----------------|-----------------|---|---|
| 1 _{x1} | 1 _{x0} | 1 _{x1} | 0 | 0 |
| 0 _{x0} | 1 _{x1} | 1 _{x0} | 1 | 0 |
| 0 _{x1} | 0 _{x0} | 1 _{x1} | 1 | 1 |
| 0 | 0 | 1 | 1 | 0 |
| 0 | 1 | 1 | 0 | 0 |

Image

| | | |
|---|--|--|
| 4 | | |
| | | |
| | | |

Convolved
Feature

CONVOLUTIONAL LAYER



DataLab
Community

| | | | | |
|-----------------|-----------------|-----------------|---|---|
| 1 | 1 | 1 | 0 | 0 |
| 0 | 1 | 1 | 1 | 0 |
| 0 _{x1} | 0 _{x0} | 1 _{x1} | 1 | 1 |
| 0 _{x0} | 0 _{x1} | 1 _{x0} | 1 | 0 |
| 0 _{x1} | 1 _{x0} | 1 _{x1} | 0 | 0 |

Image

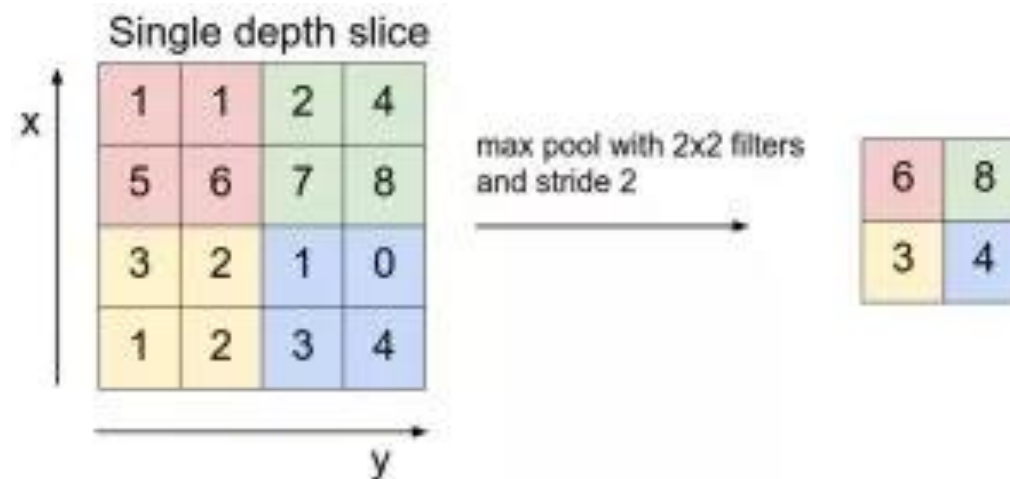
| | | |
|---|---|---|
| 4 | 3 | 4 |
| 2 | 4 | 3 |
| 2 | | |

Convolved
Feature

POOLING LAYER



DataLab
Community



POOLING LAYER

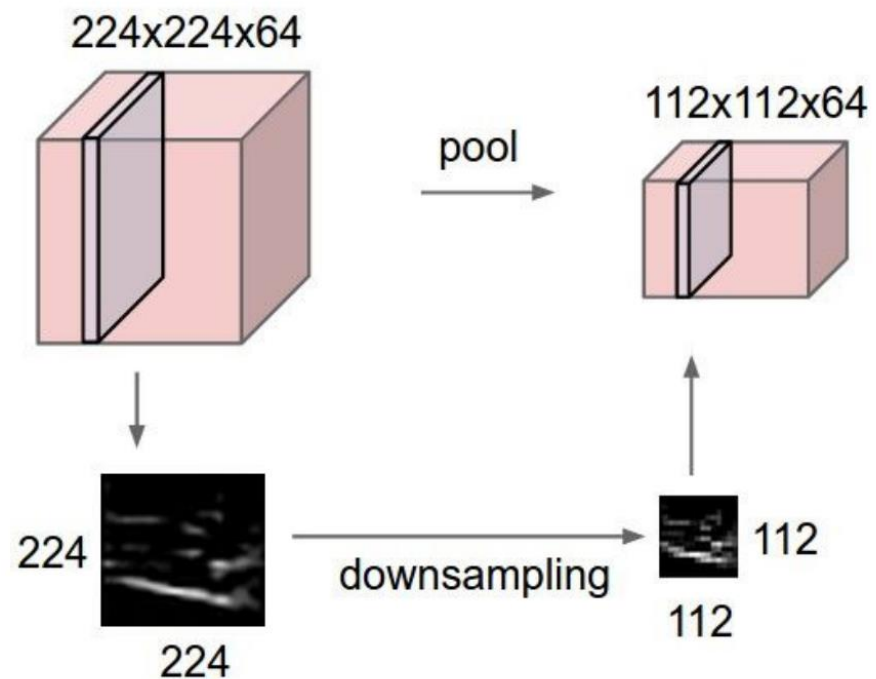
IMAGE WITH SALT AND PEPPER NOISE



POOLING LAYER



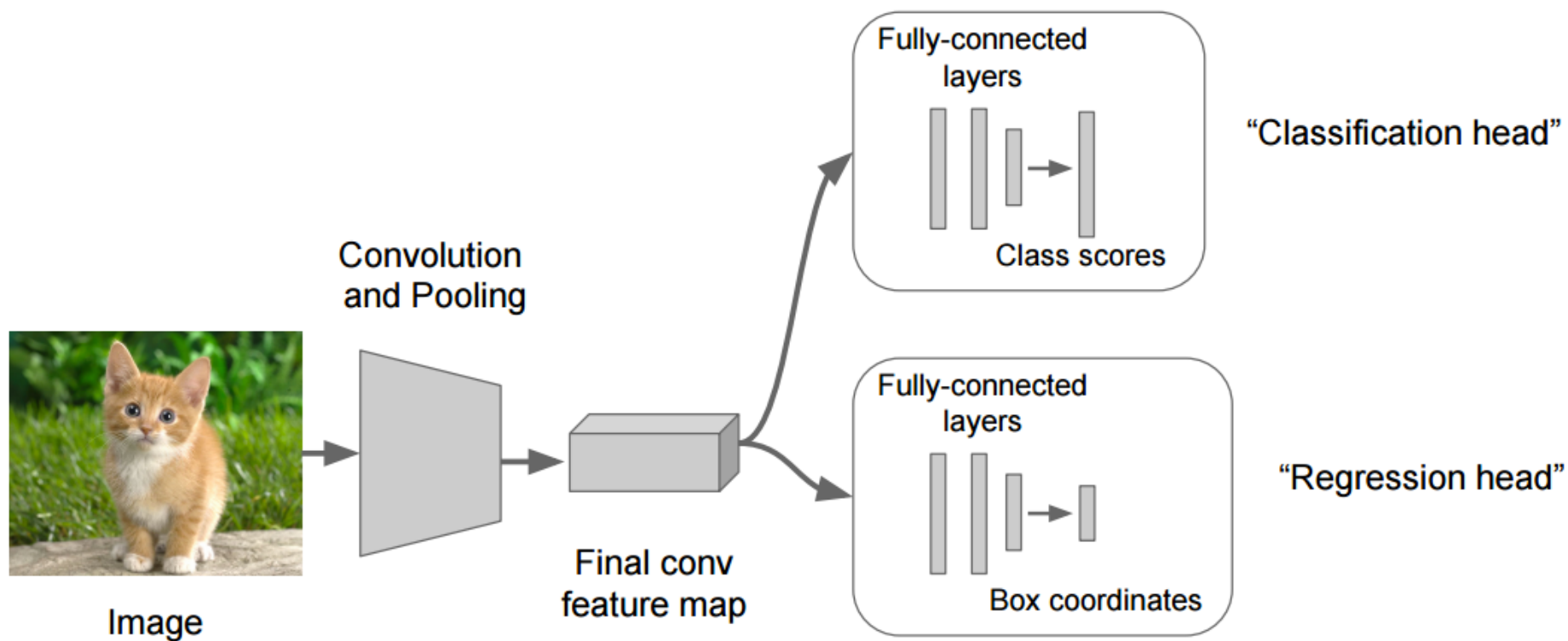
DataLab
Community



FULLY CONNECTED LAYER



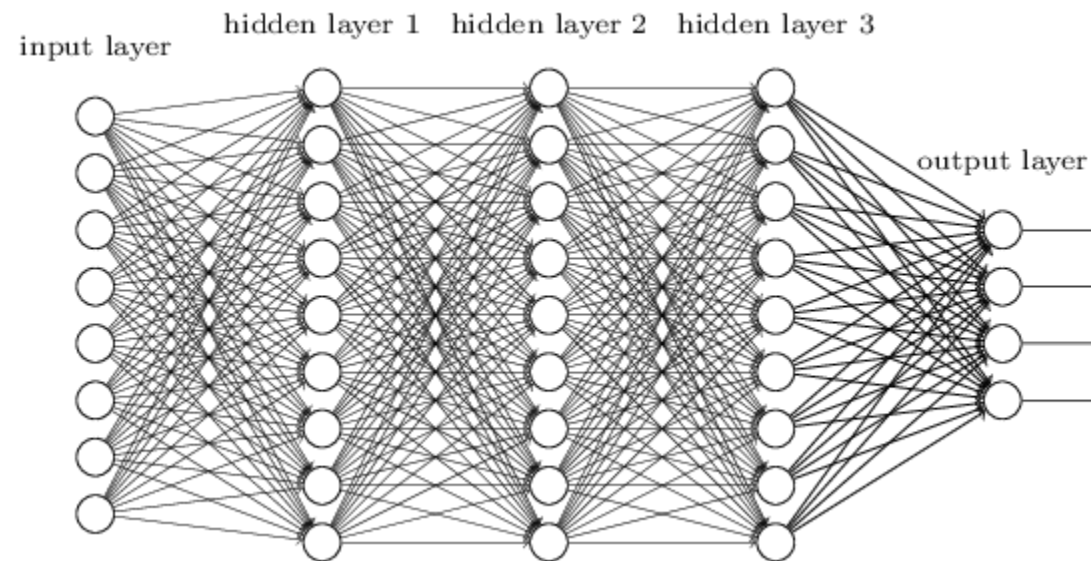
DataLab
Community



FULLY CONNECTED LAYER



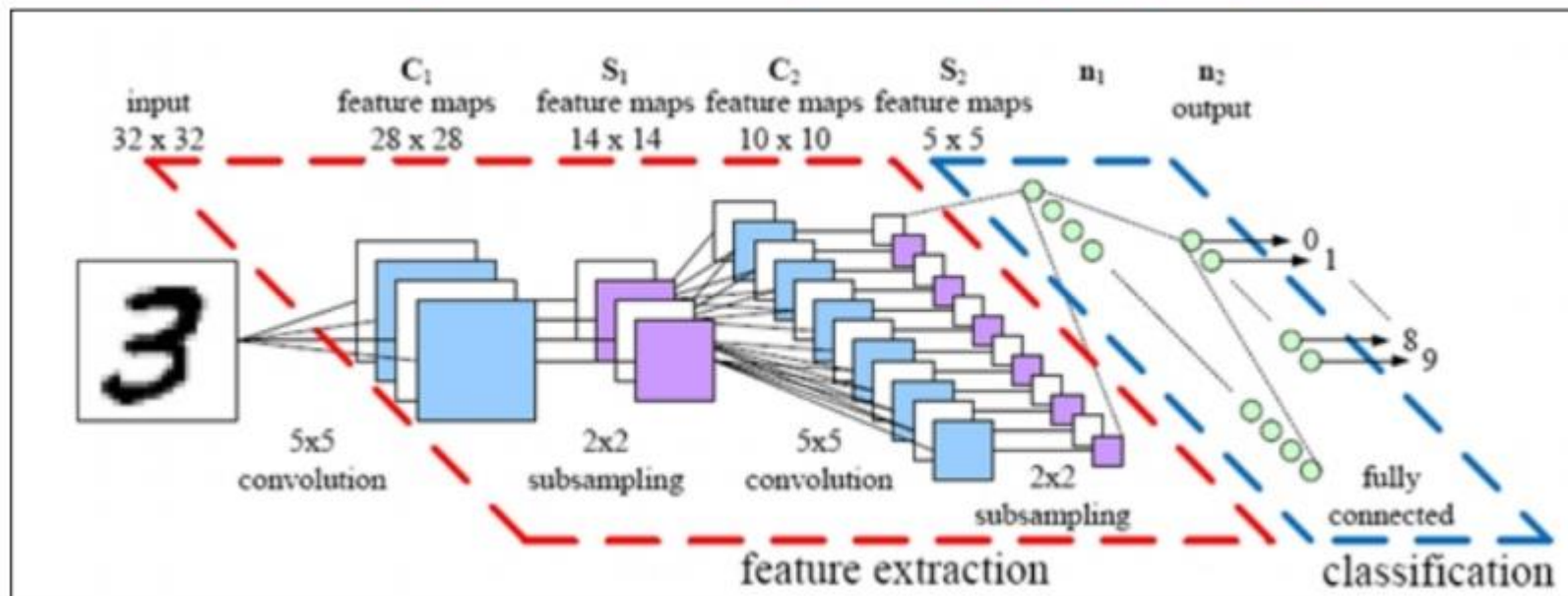
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DEEP LEARNING



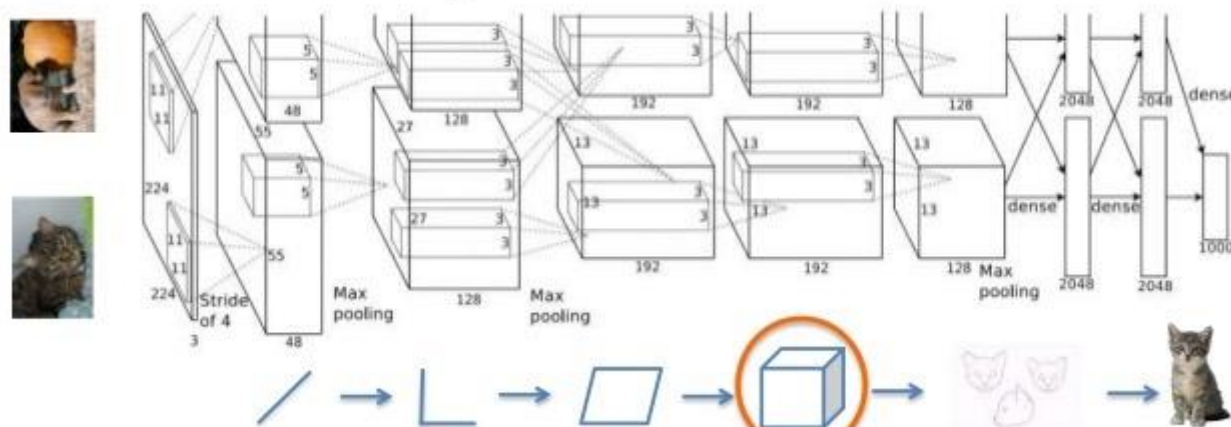
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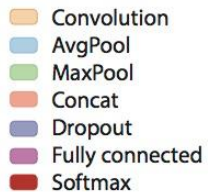
AlexNet (Krizhevsky et al. 2012)

The class with the highest likelihood is the one the DNN selects



When AlexNet is processing an image, this is what is happening at each layer.

INCEPTION V3



COMPUTER VISION

DEMO CNN



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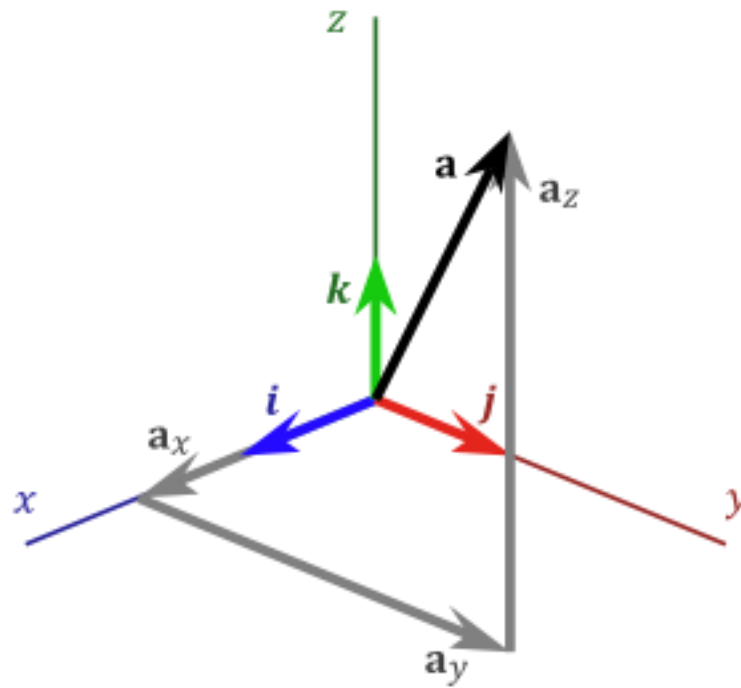
COMPUTER VISION

WHAT IS A TENSOR?

WHAT IS A TENSOR?



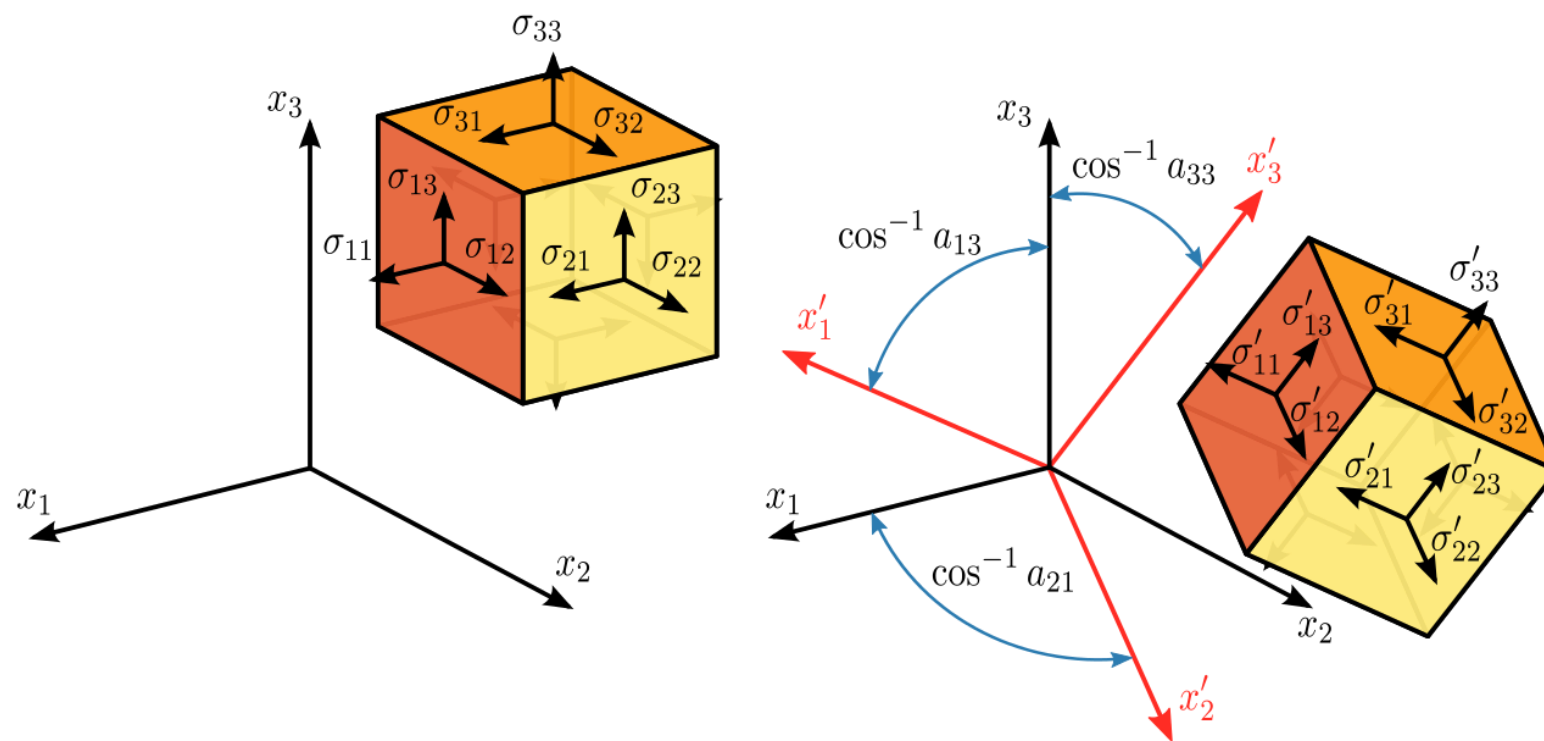
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WHAT IS A TENSOR?



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LILIAN LIEBER



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- Tensors
- “The facts of the universe”.



• *1886-1986*

LILIAN LIEBER



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- From “The Einstein Theory of Relativity”
- In n -dimensional space,
- a VECTOR has n components,
- a TENSOR of rank TWO has n^2 comp
- a TENSOR of rank THREE has n^3 comp
- and so on.



• *1886-1986*

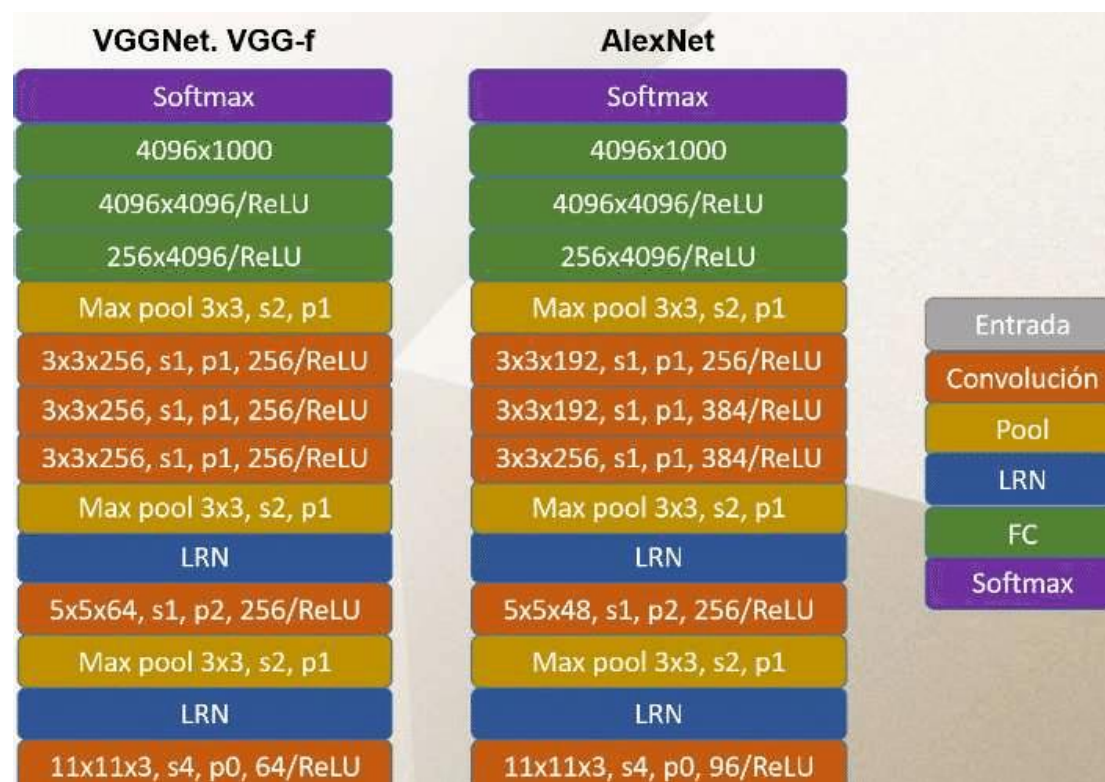
IMAGE PROCESSING VS CNN

COMPUTER VISION

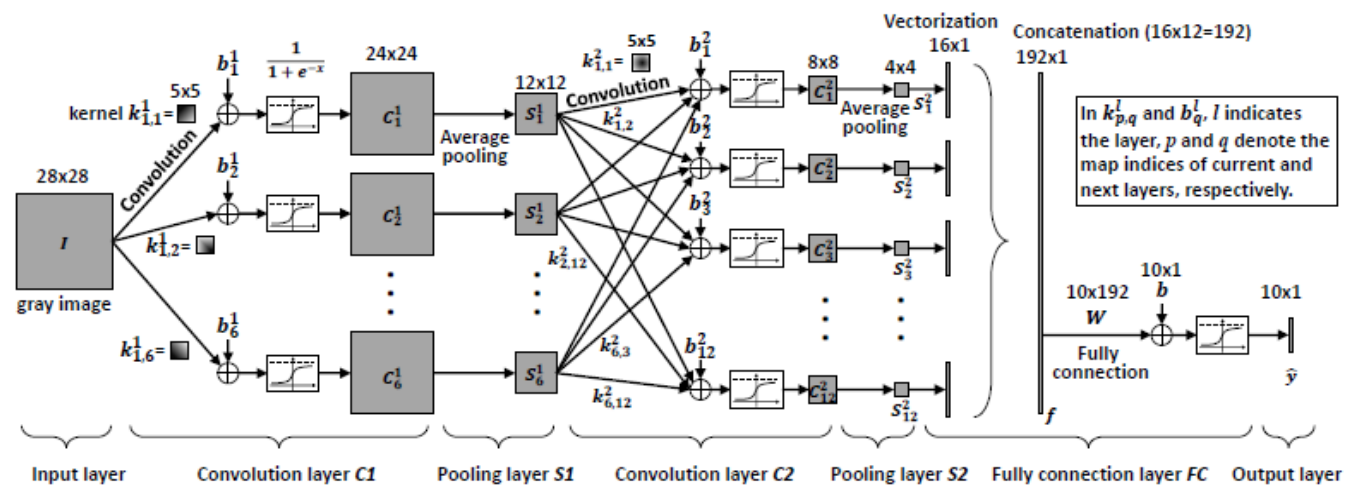


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CNN ARCHITECTURE



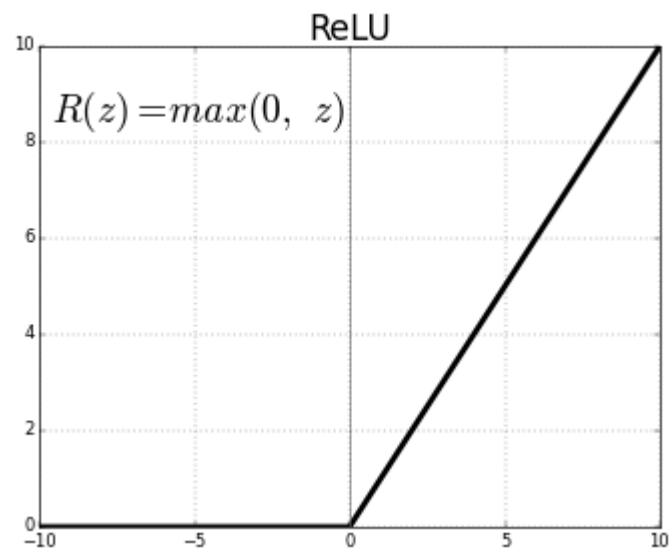
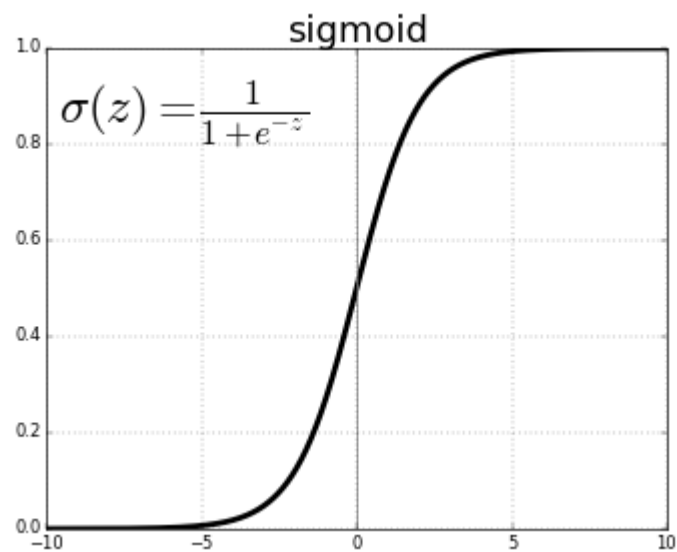
CNN ARCHITECTURE



RELU



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IF WE LOOK CLOSELY AT THE “CONVOLUTION LAYER”



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$$C_p^1(i, j) = \sigma \left(\sum_{u=-2}^2 \sum_{v=-2}^2 I(i-u, j-v) k_{1,p}^1(u, v) + b_p^1 \right)$$

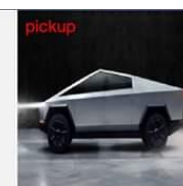


COMPUTER VISION

WHY DO YOU NEED SO MANY IMAGES FOR TRAINING?



VGG 16



GoogLeNet



InceptionV3



ResNet-101





How to confuse machine learning



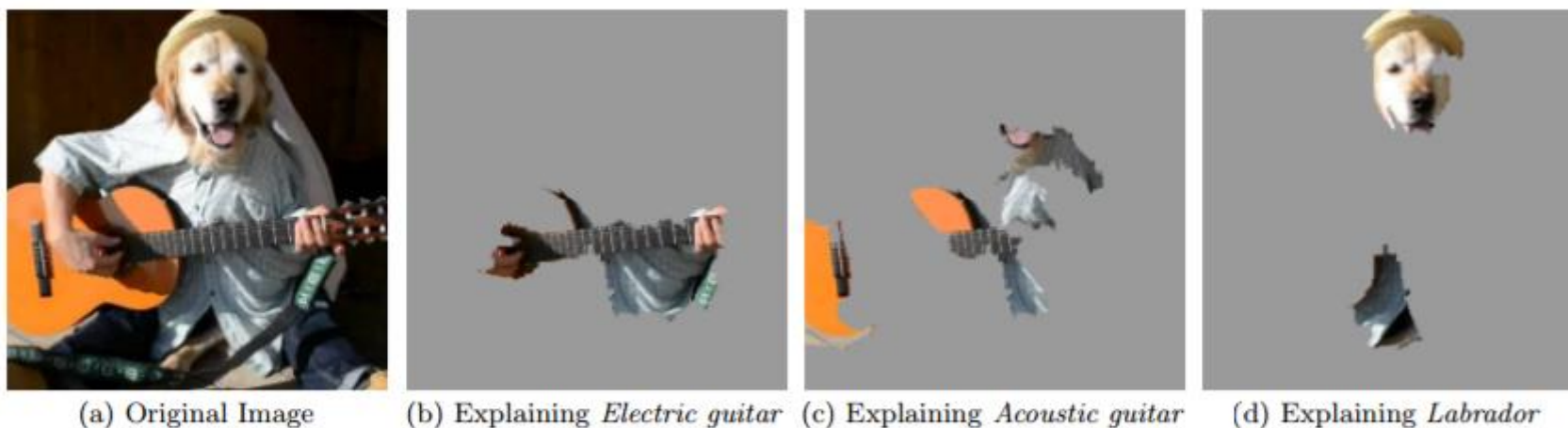
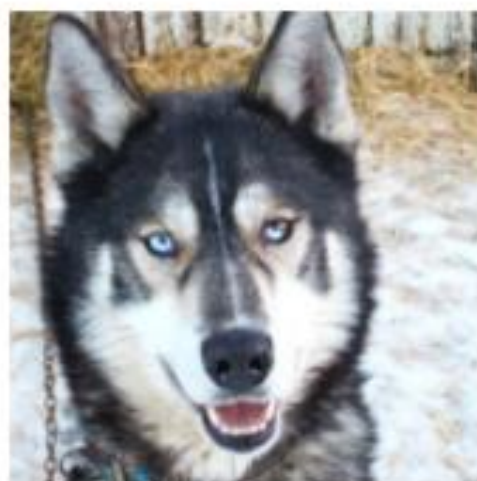
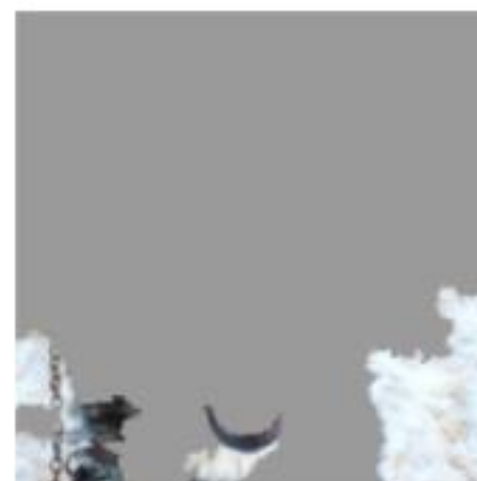


Figure 4: Explaining an image classification prediction made by Google's Inception neural network. The top 3 classes predicted are "Electric Guitar" ($p = 0.32$), "Acoustic guitar" ($p = 0.24$) and "Labrador" ($p = 0.21$)



(a) Husky classified as wolf



(b) Explanation

Figure 11: Raw data and explanation of a bad model's prediction in the "Husky vs Wolf" task.



**“Those who can
imagine
anything, can
create the
impossible.”**

-Alan Turing



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Questions?

Rubén Alvarez - @bio_ruben

@datalabmx