

IESTI01 – TinyML

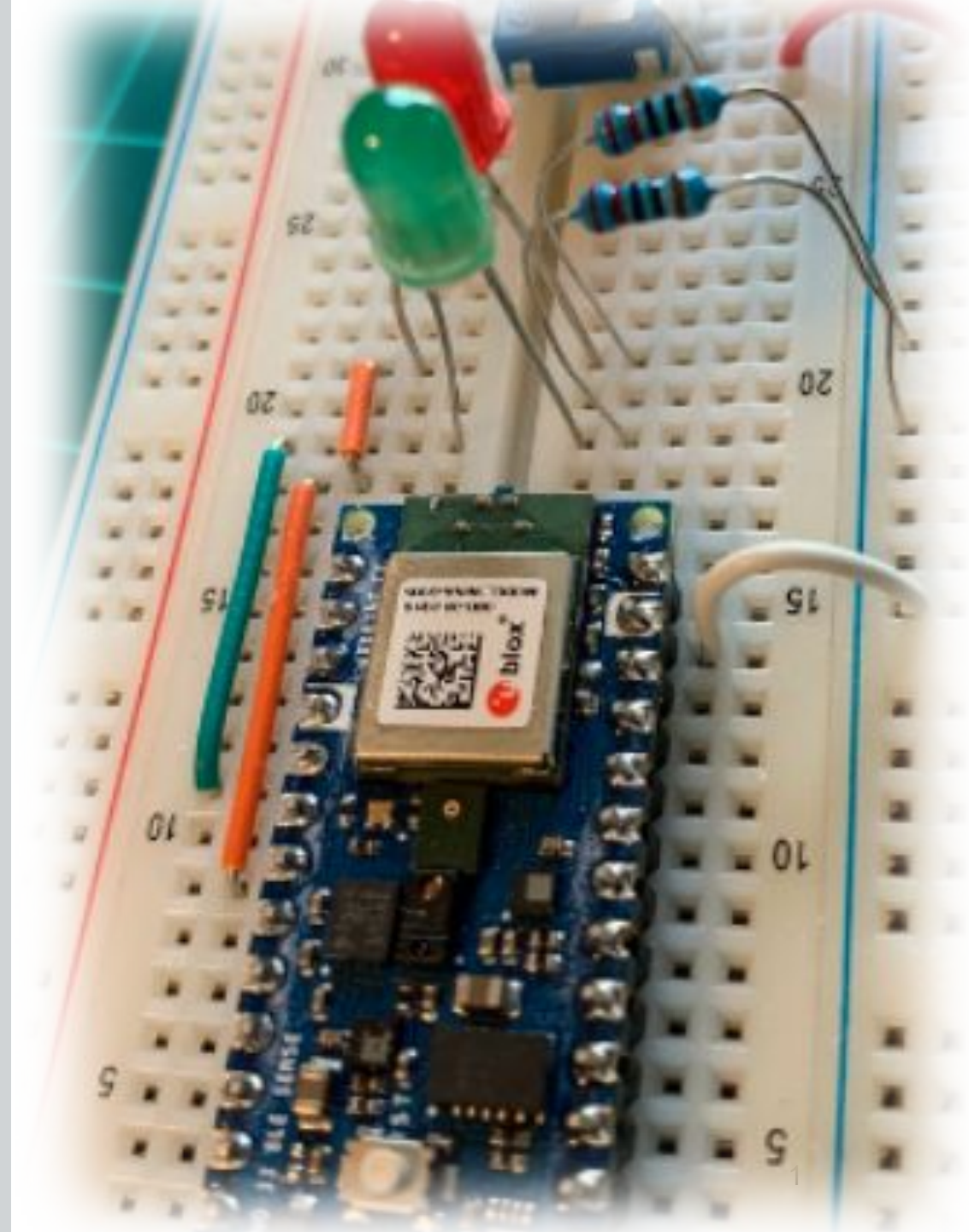
Embedded Machine Learning

10. Introducing Convolutions



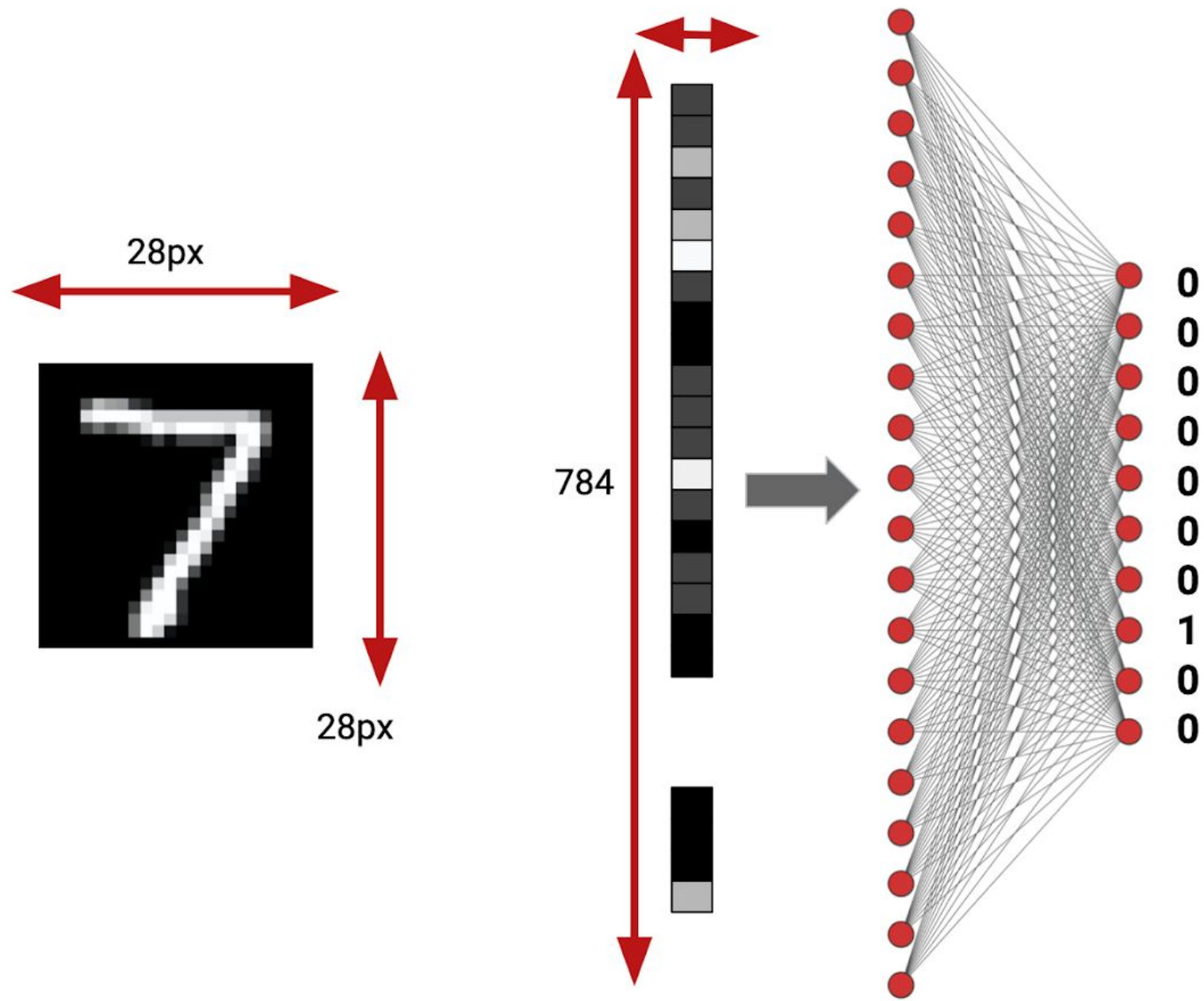
Prof. Marcelo Rovai

UNIFEI



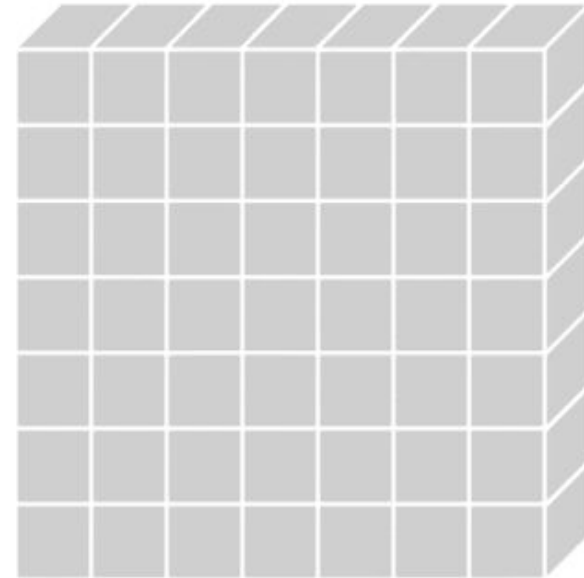
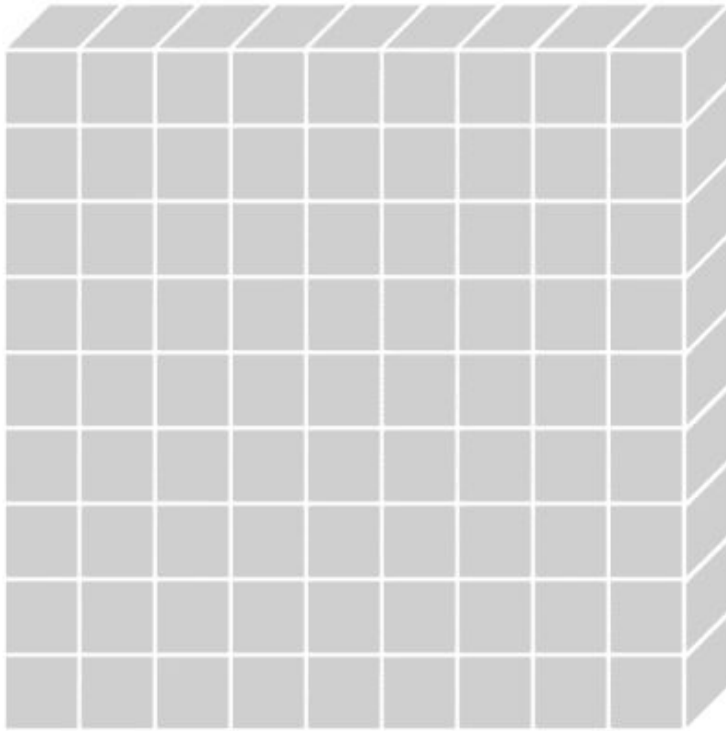
Introducing Convolutions

Beyond weights and biases...

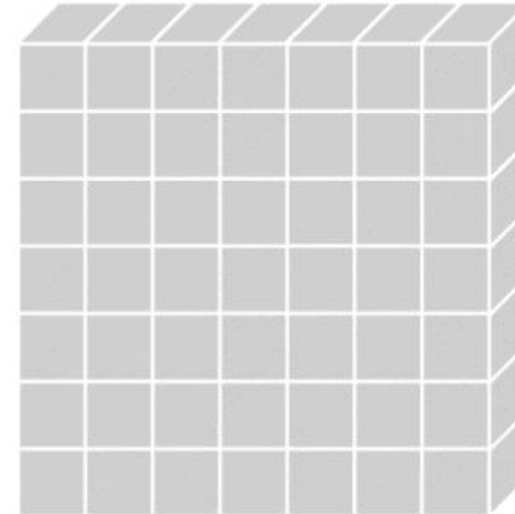
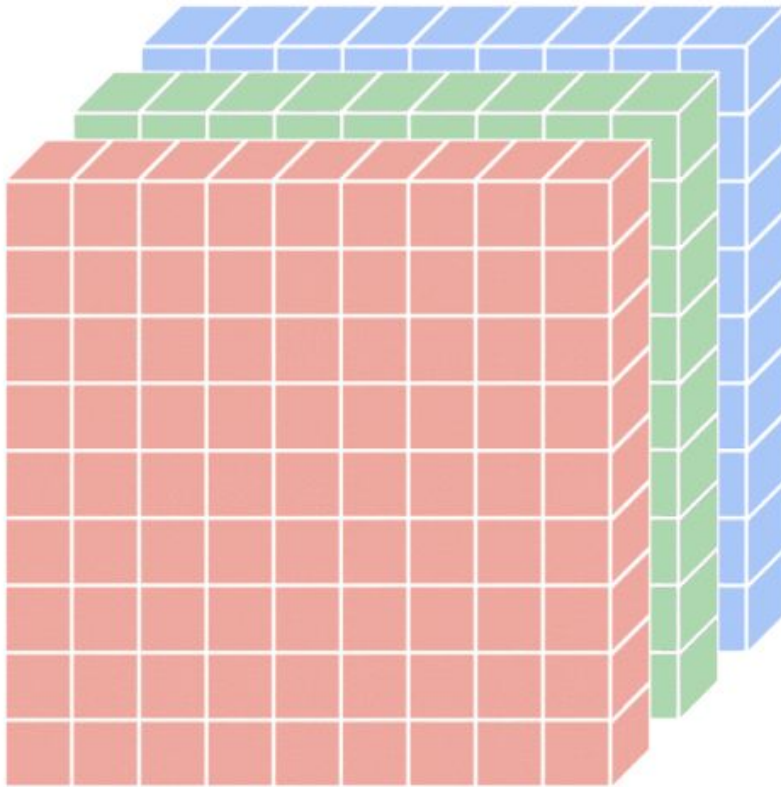




Standard Convolution (1 Channel)

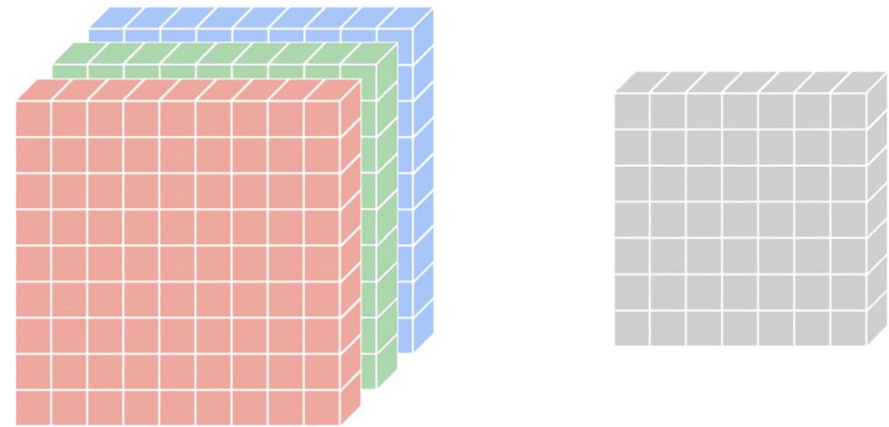


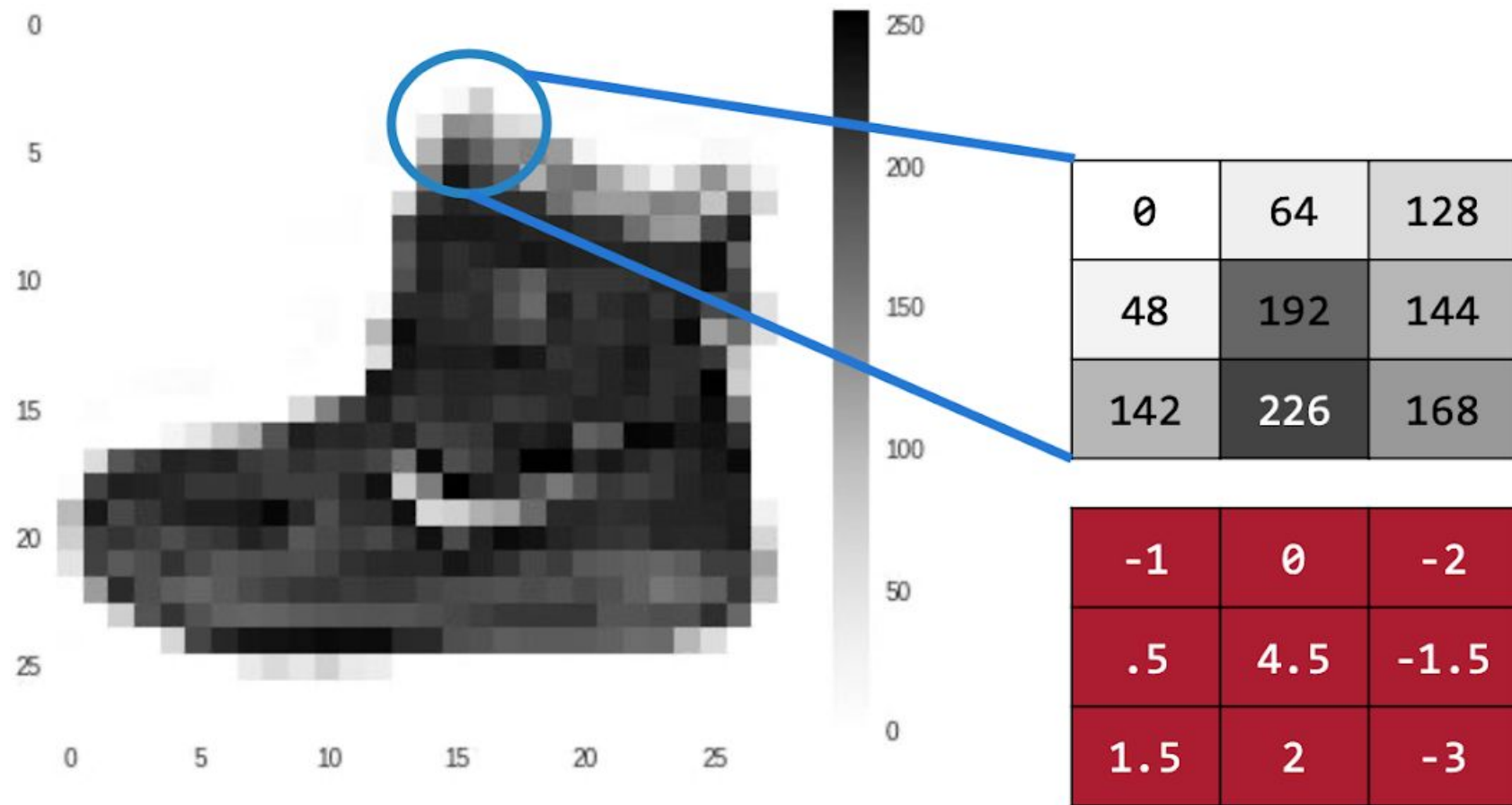
Standard Convolution (**3 Channel**—e.g., *RGB*)



Standard Convolution (**3 Channel**—e.g., *RGB*)

- Input Feature Map
 - $8 \times 8 \times 3$
 - Width \times Height \times Channels
- Kernel (*1 Filter*)
 - $3 \times 3 \times 3$





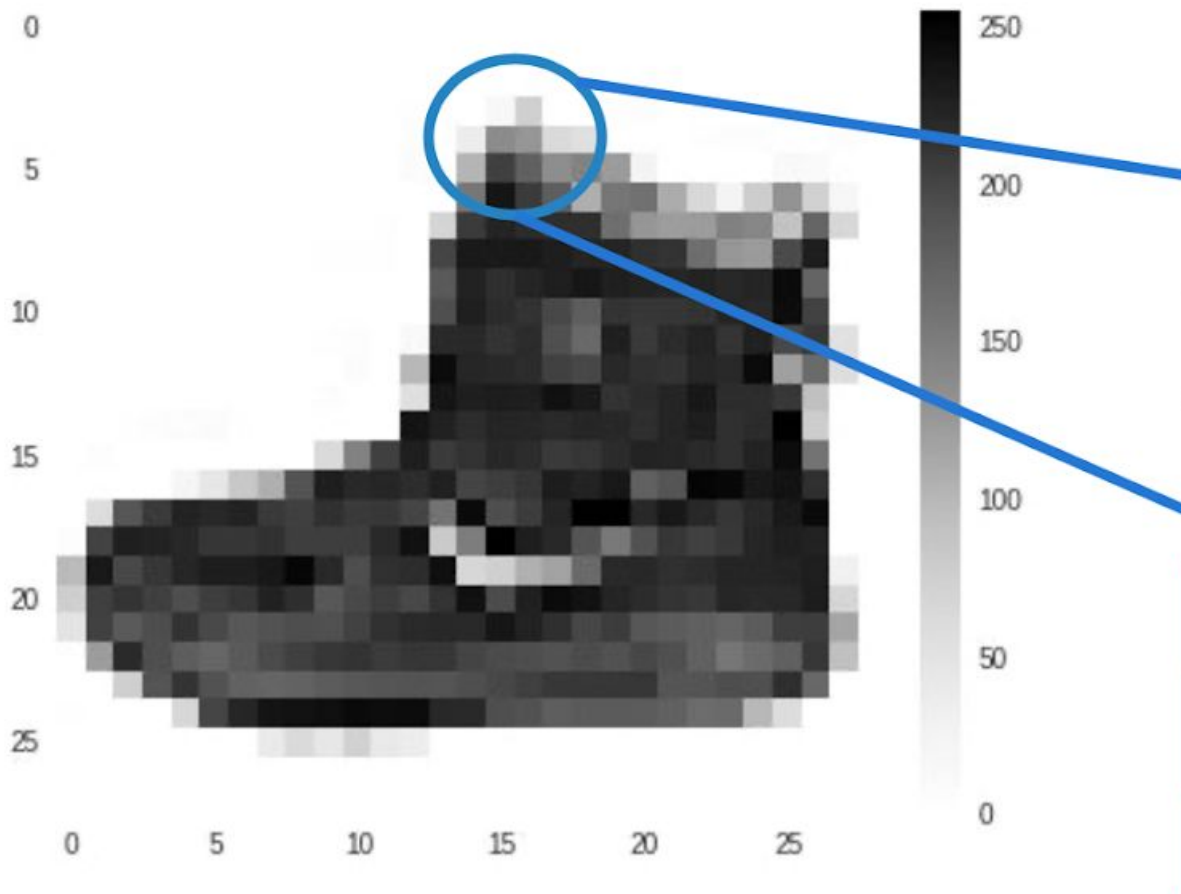
Current Pixel Value is 192

Consider neighbor Values

Filter Definition

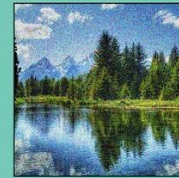
CURRENT_PIXEL_VALUE = 192

NEW_PIXEL_VALUE = $(-1 * 0) + (0 * 64) + (-2 * 128) +$
 $(.5 * 48) + (4.5 * 192) + (-1.5 * 144) +$
 $(1.5 * 42) + (2 * 226) + (-3 * 168)$



Kernels = Filters

Different Filters



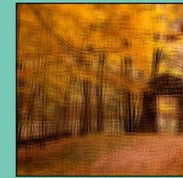
Noise



Gaussian Blur



Sharpen More



Fragment



Facet



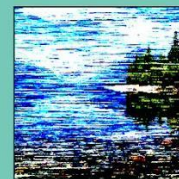
Pointillize



Mosaic



Tiles



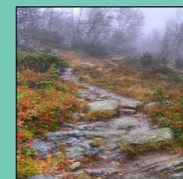
Mezzotint



Solarize



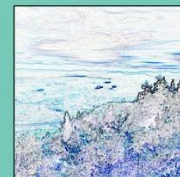
Trace Contour



Wind



Clouds



Find Edges



Shape Blur



Fibers

Image Kernels

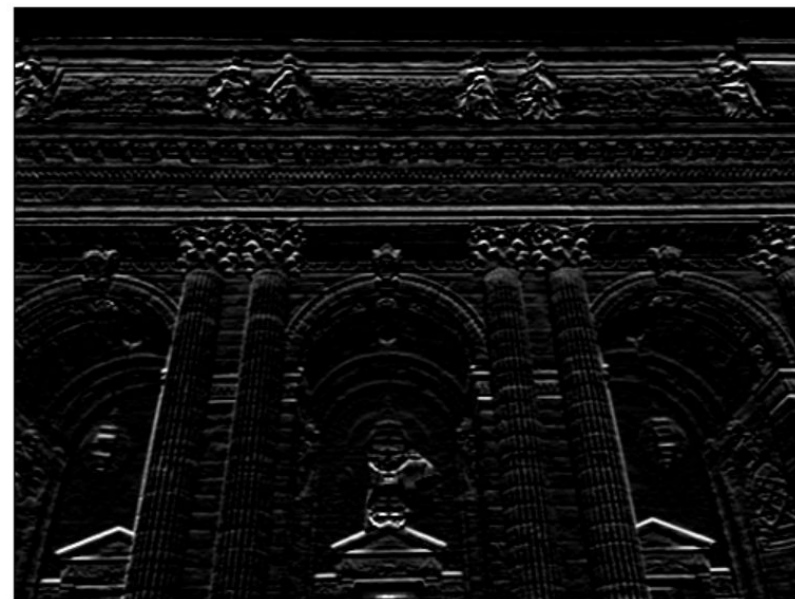


-1	0	1
-2	0	2
-1	0	1

custom

-1	-2	-1
0	0	0
1	2	1

custom



<https://setosa.io/ev/image-kernels/>

0	64	128	128
48	192	144	144
142	226	168	0
255	0	0	64

0	64
48	192

192

128	128
144	144

144

142	226
255	0

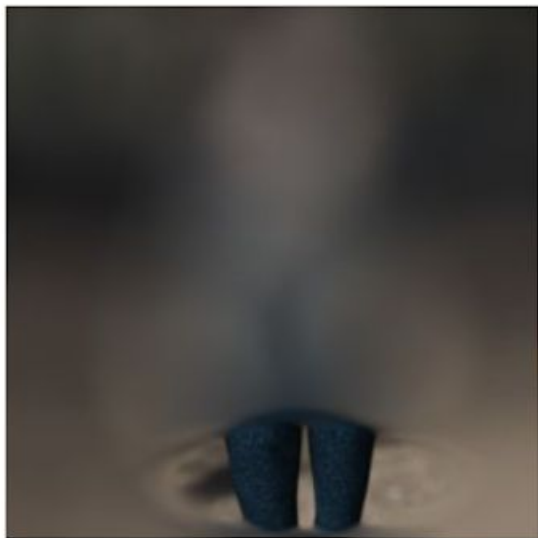
255

168	0
0	64

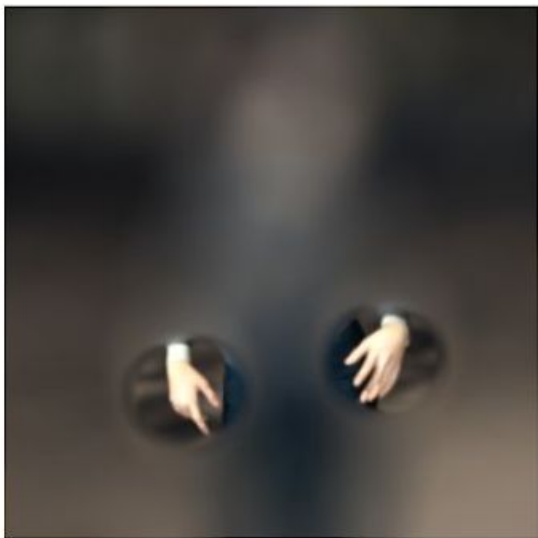
168

192	144
255	168

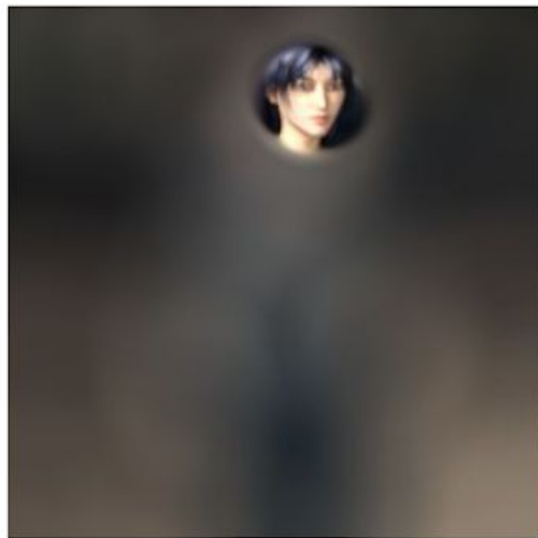
Max Pooling



+

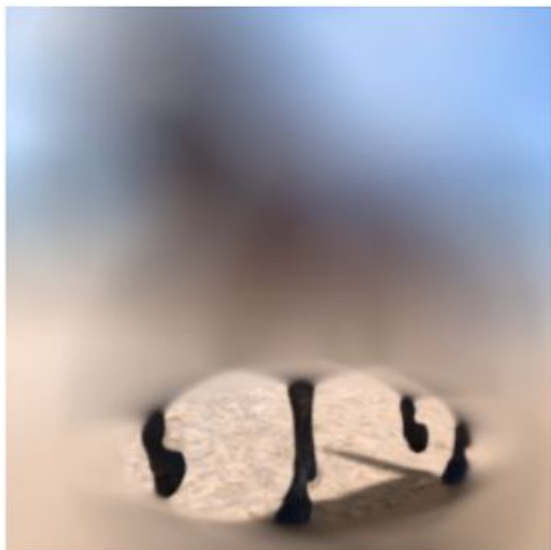


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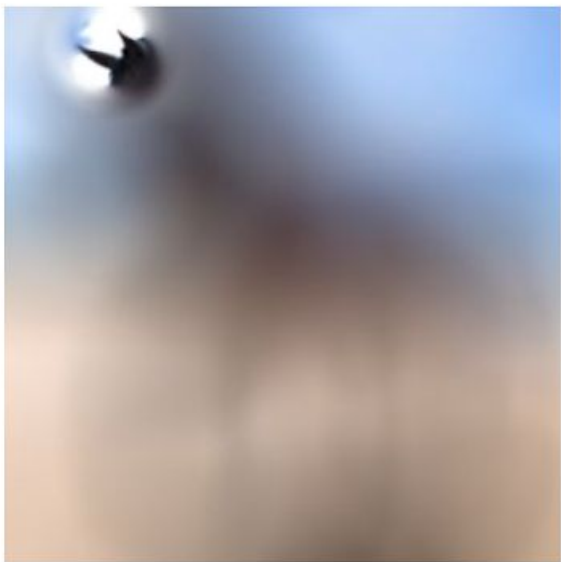


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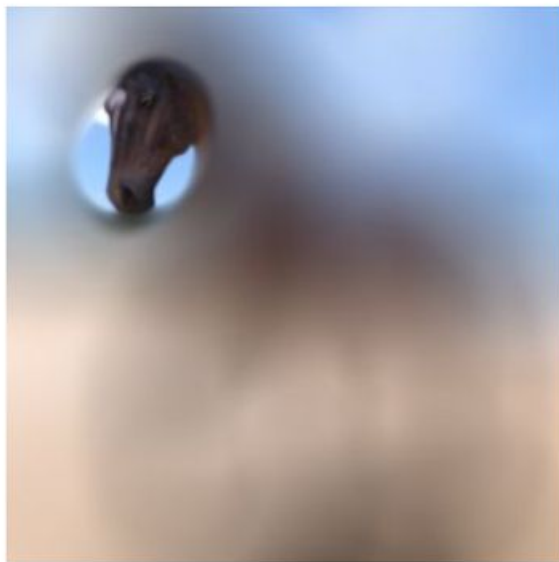
HUMAN



+

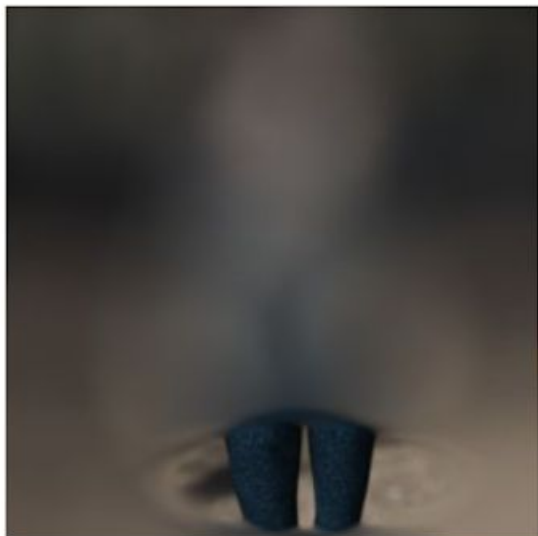


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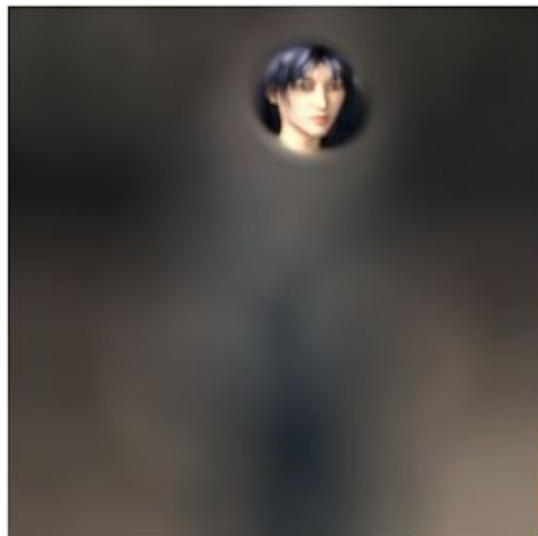
HORSE



+

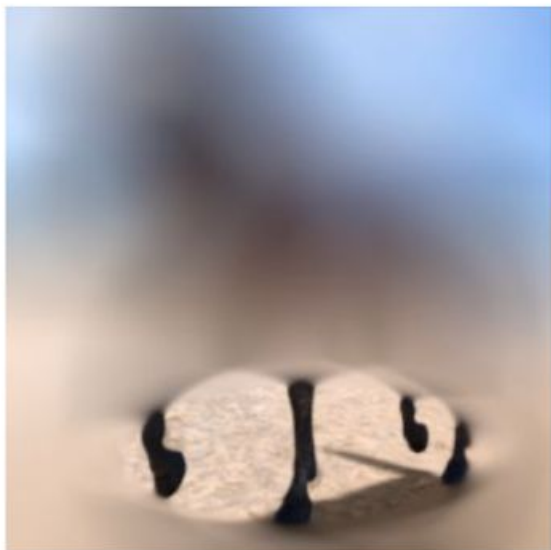


+



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HUMAN

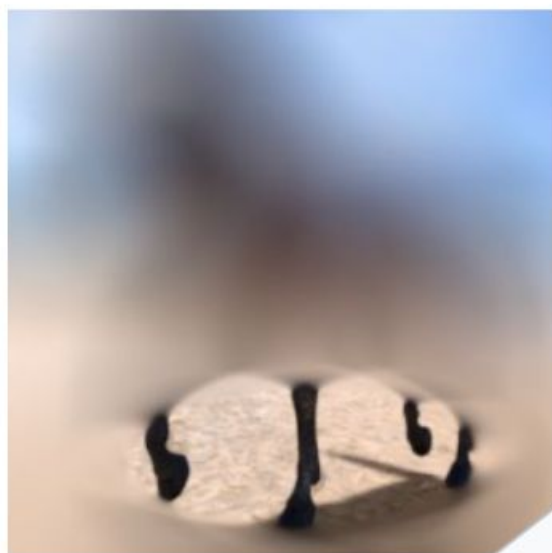


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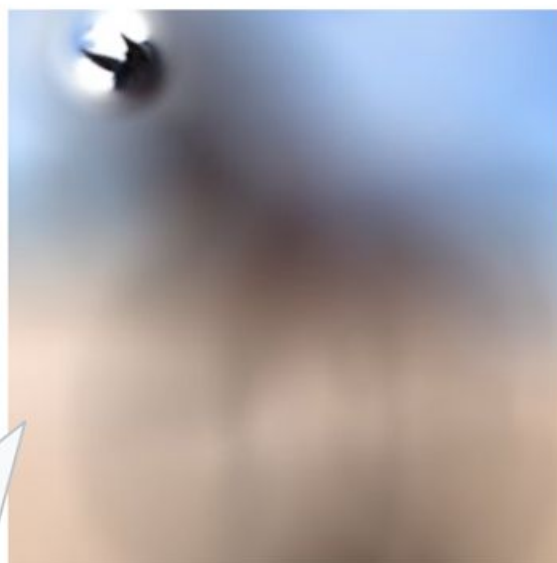


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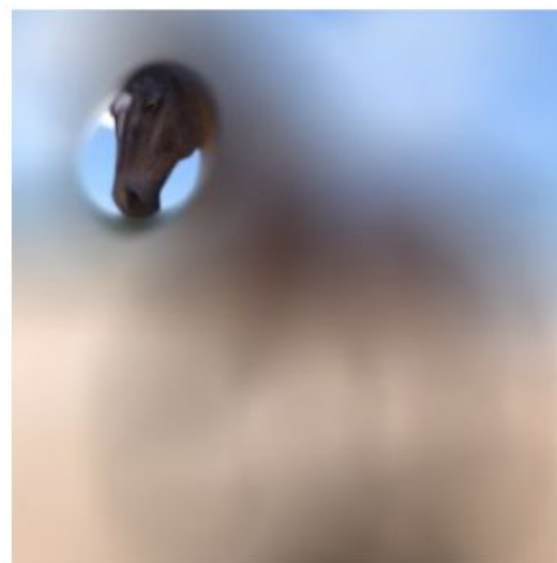
HORSE



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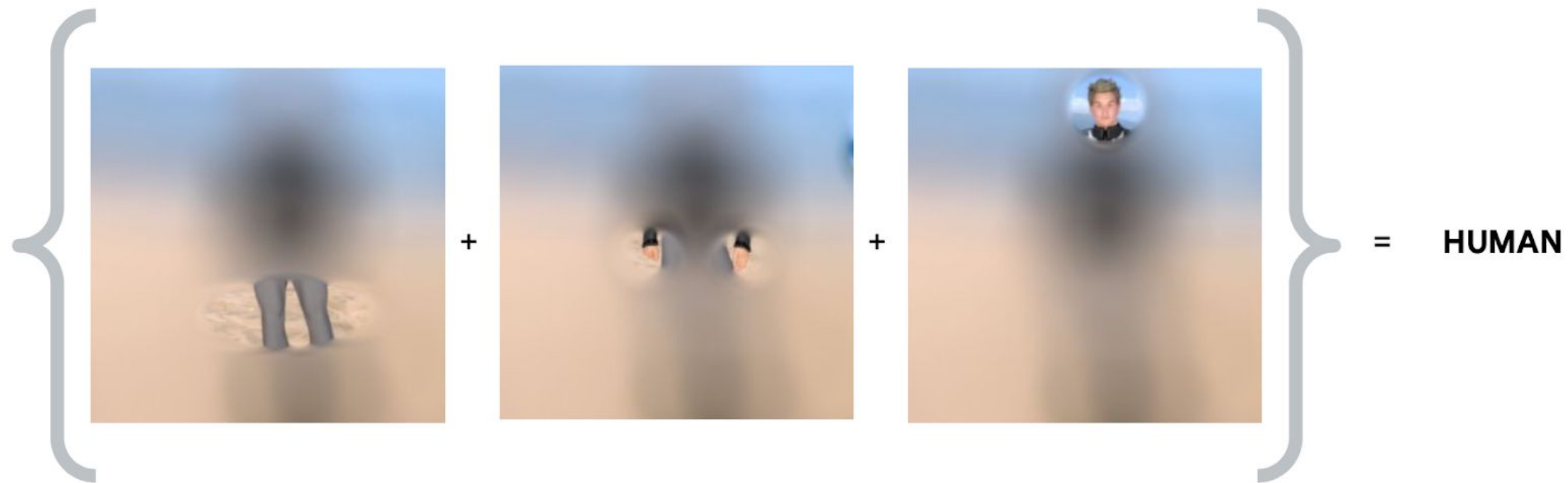
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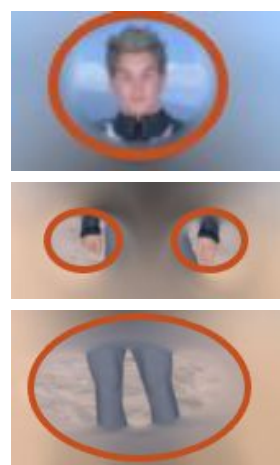
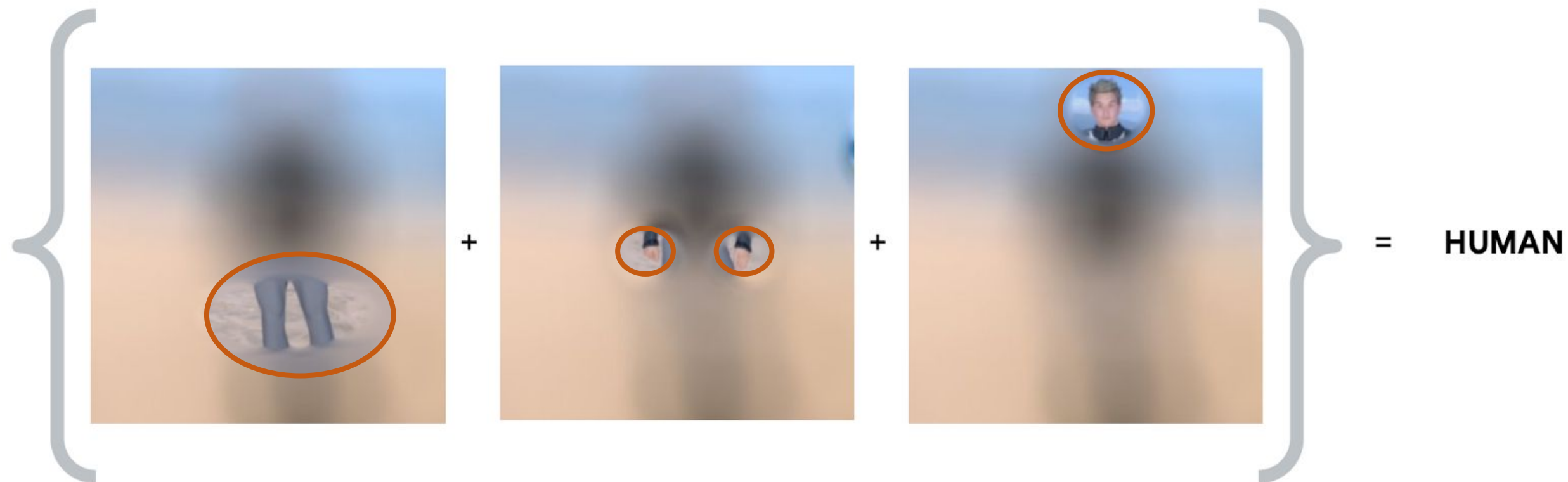


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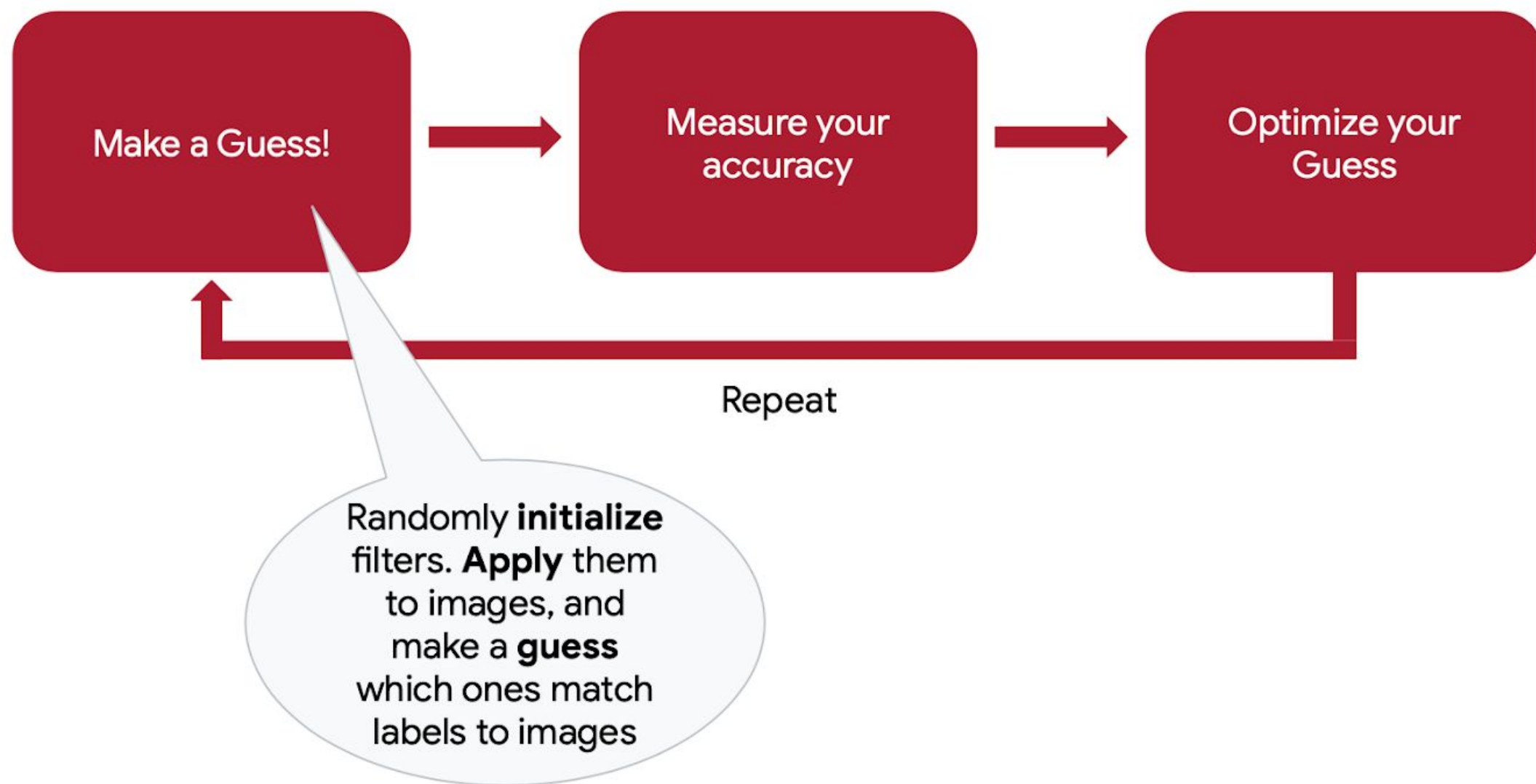
HORSE

Filters can then be combined with **labels** to make a **prediction** of the image contents...

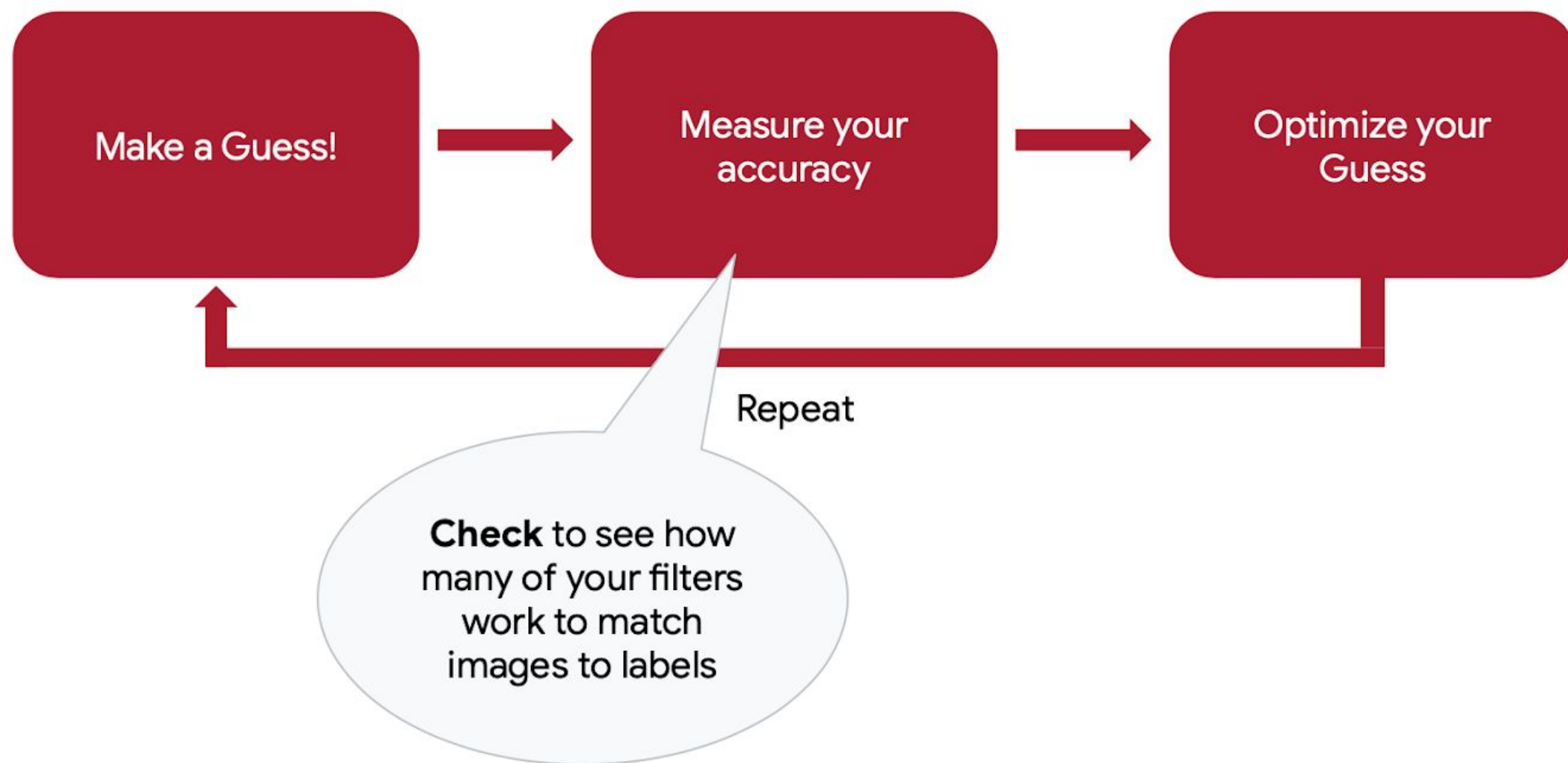




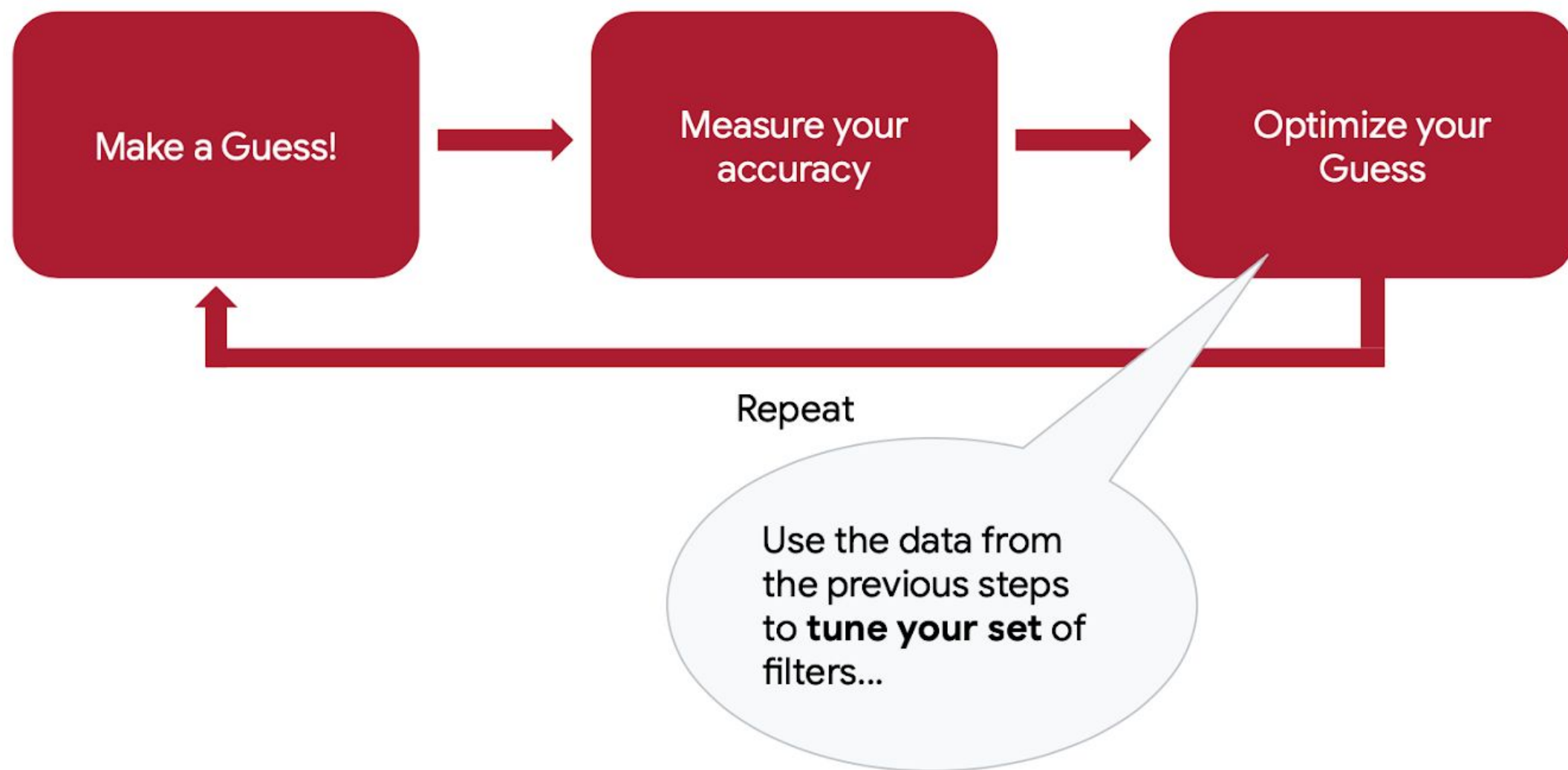
The Machine Learning Paradigm



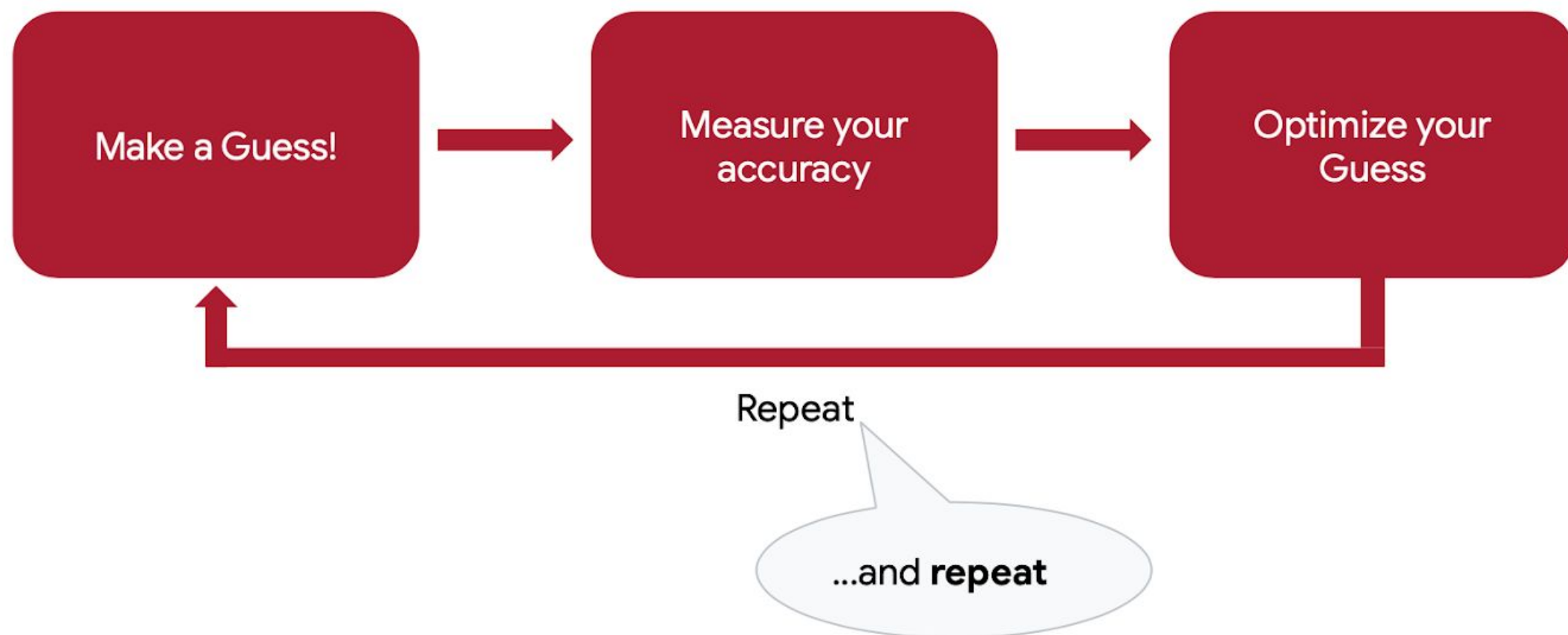
The Machine Learning Paradigm


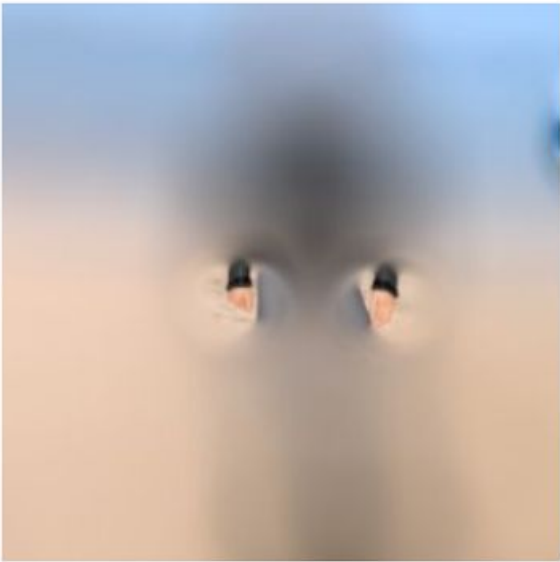
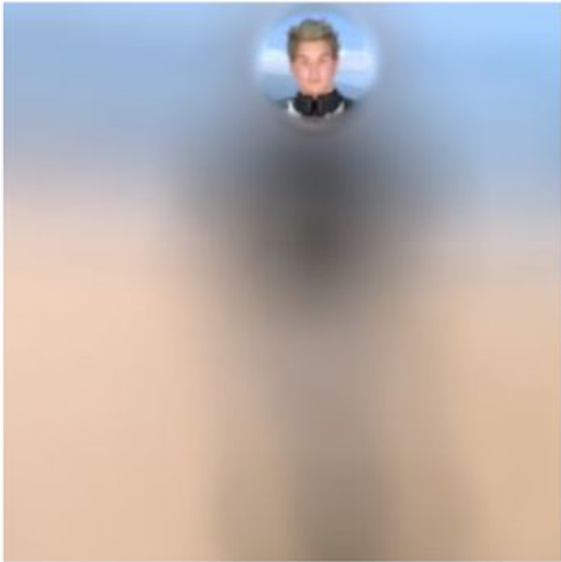


The Machine Learning Paradigm



The Machine Learning Paradigm



{  +  +  } = HUMAN



Exploring CNN

CNN Explainer

<https://poloclub.github.io/cnn-explainer/>

ConvNetJS MNIST demo

<https://cs.stanford.edu/people/karpathy/convnetjs/demo/mnist.html>

ConvNetJS CIFAR-10 demo

<https://cs.stanford.edu/people/karpathy/convnetjs/demo/cifar10.html>

Reading Material

Main references

- [Harvard School of Engineering and Applied Sciences - CS249r: Tiny Machine Learning](#)
- [Professional Certificate in Tiny Machine Learning \(TinyML\) – edX/Harvard](#)
- [Introduction to Embedded Machine Learning - Coursera/Edge Impulse](#)
- [Computer Vision with Embedded Machine Learning - Coursera/Edge Impulse](#)
- Fundamentals textbook: [“Deep Learning with Python” by François Chollet](#)
- Applications & Deploy textbook: [“TinyML” by Pete Warden, Daniel Situnayake](#)
- Deploy textbook [“TinyML Cookbook” by Gian Marco Iodice](#)

I want to thank **Shawn Hymel** and Edge Impulse, **Pete Warden** and **Laurence Moroney** from Google, Professor **Vijay Janapa Reddi** and **Brian Plancher** from Harvard, and the rest of the **TinyMLedu** team for preparing the excellent material on TinyML that is the basis of this course at UNIFEI.

The IESTI01 course is part of the **TinyML4D**, an initiative to make TinyML education available to everyone globally.

Thanks



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