

IESTI01 – TinyML

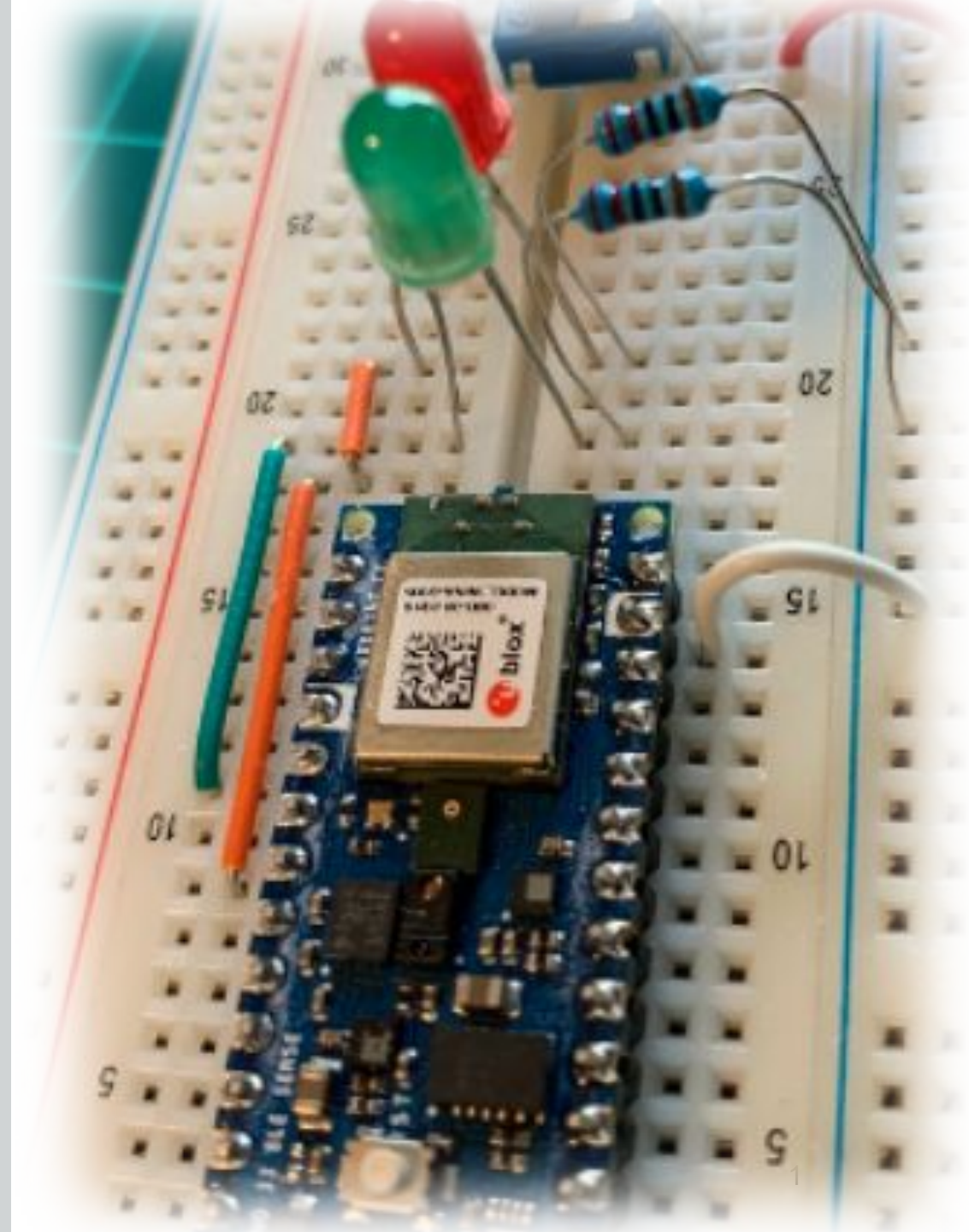
Embedded Machine Learning

10. Introducing Convolutions



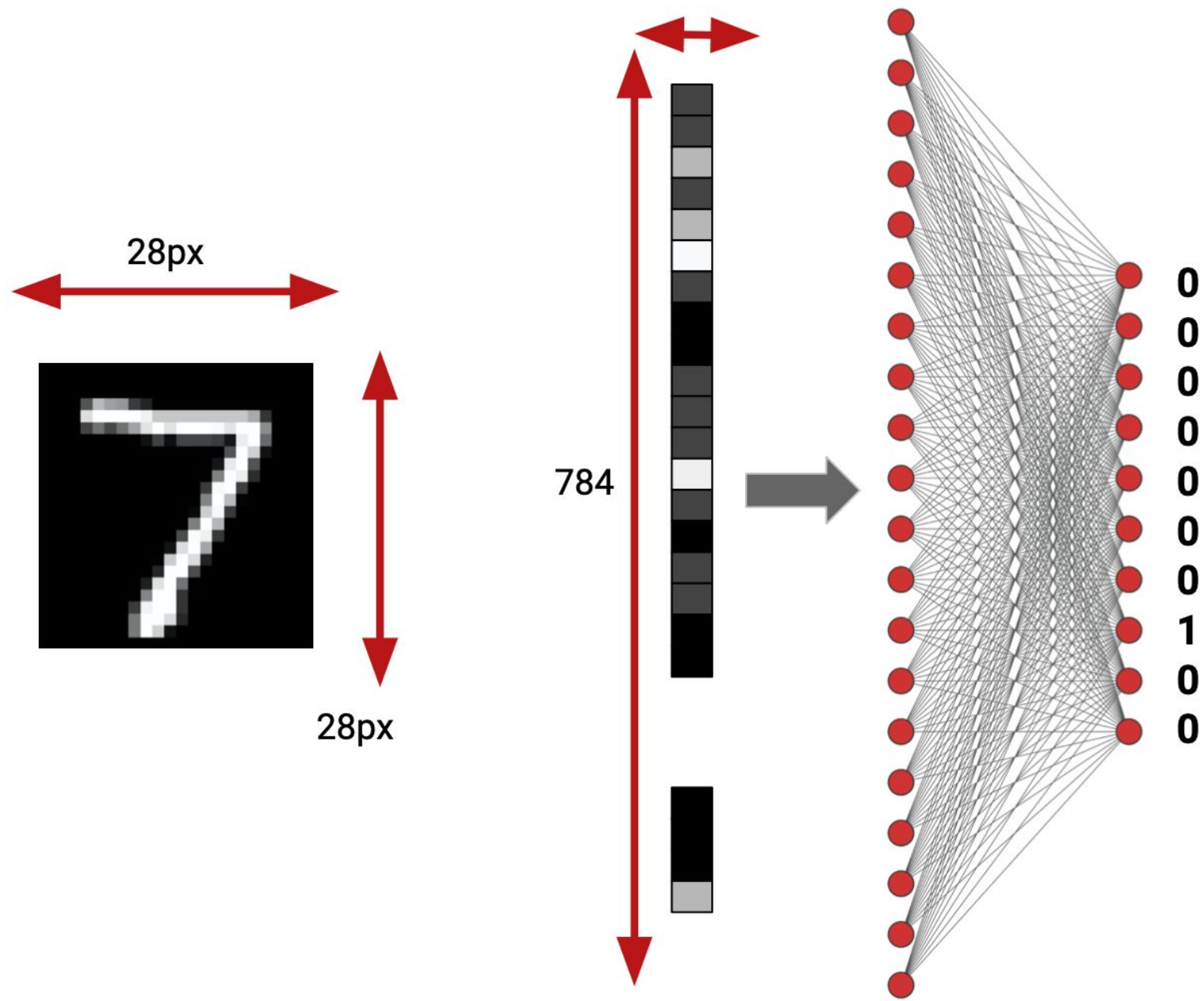
Prof. Marcelo Rovai

UNIFEI



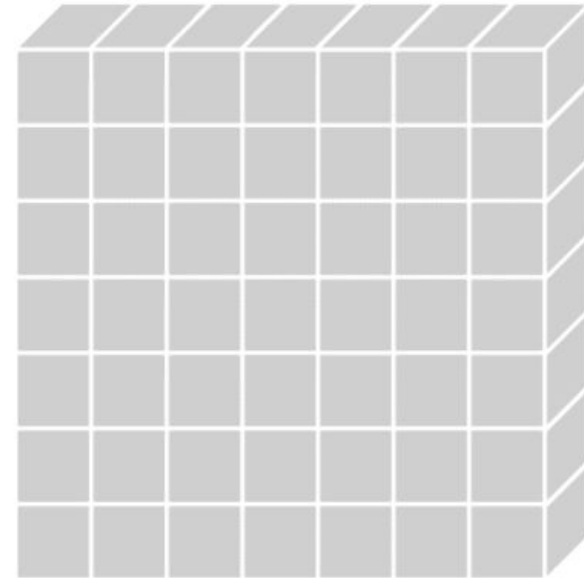
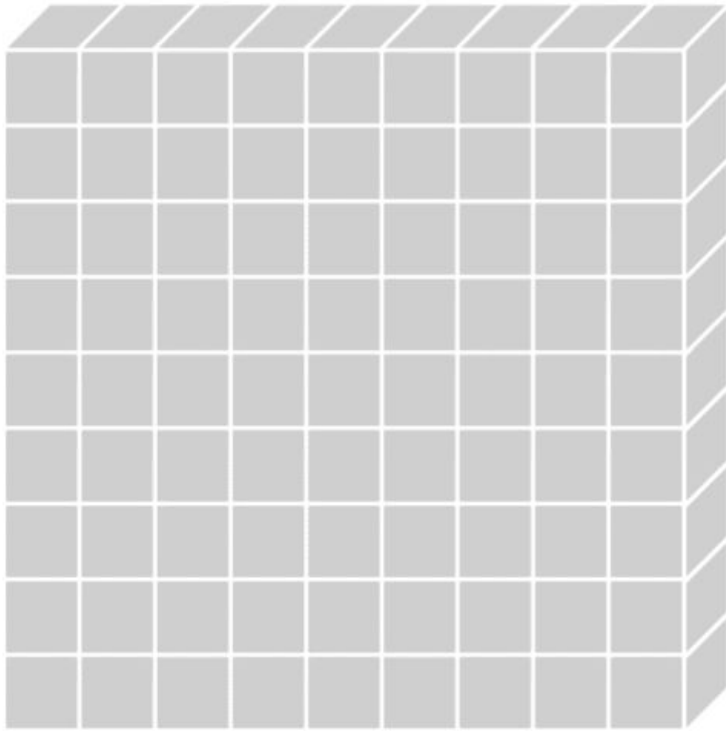
Introducing Convolutions

Beyond weights and biases...

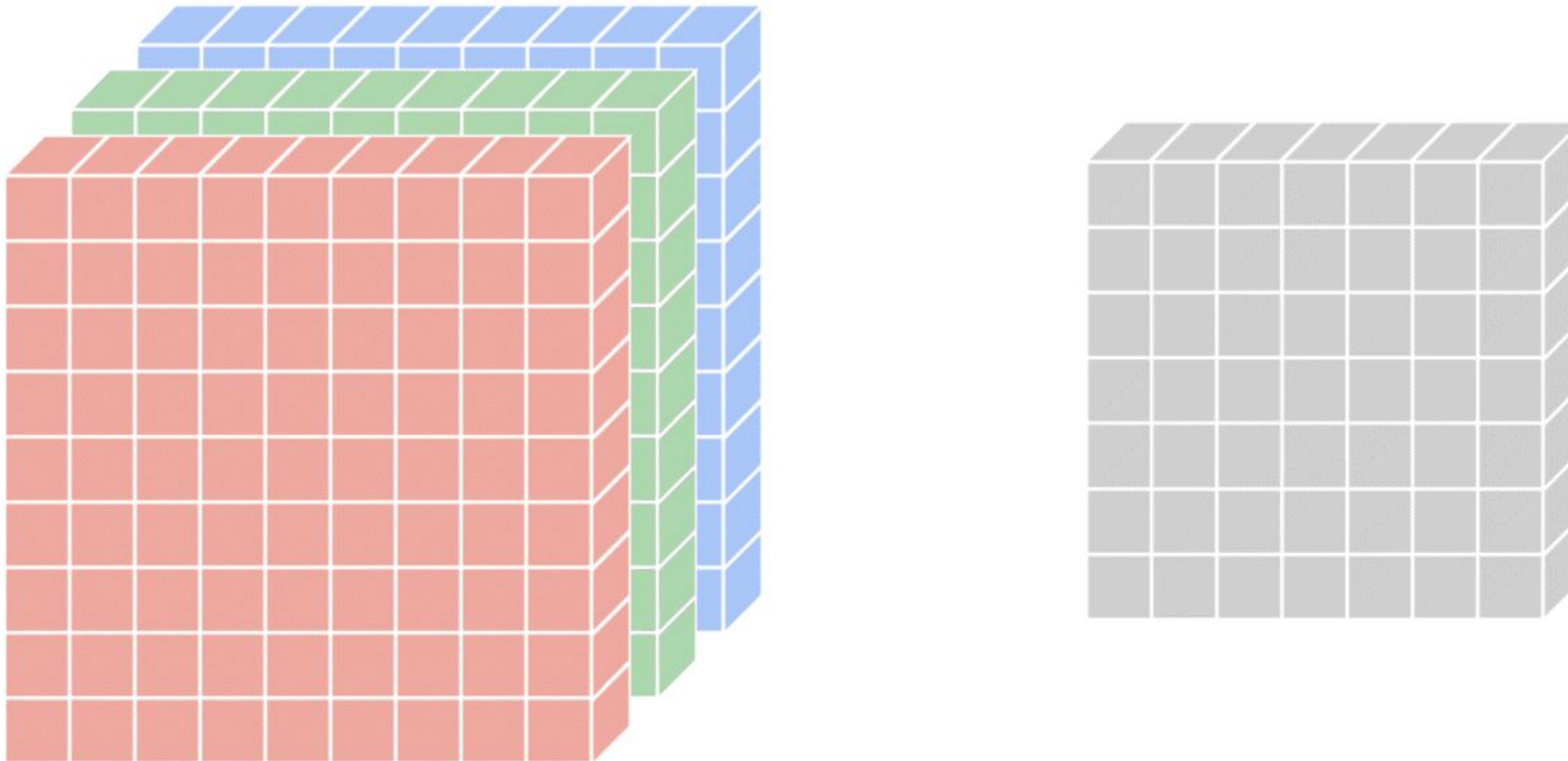




Standard Convolution (1 Channel)

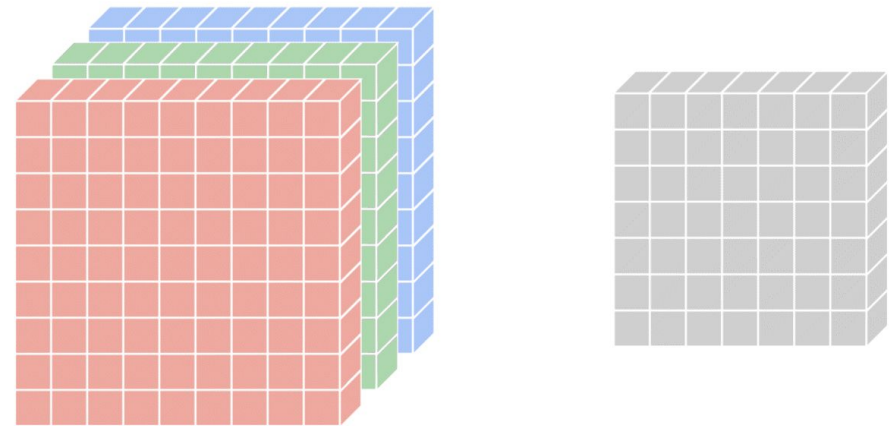


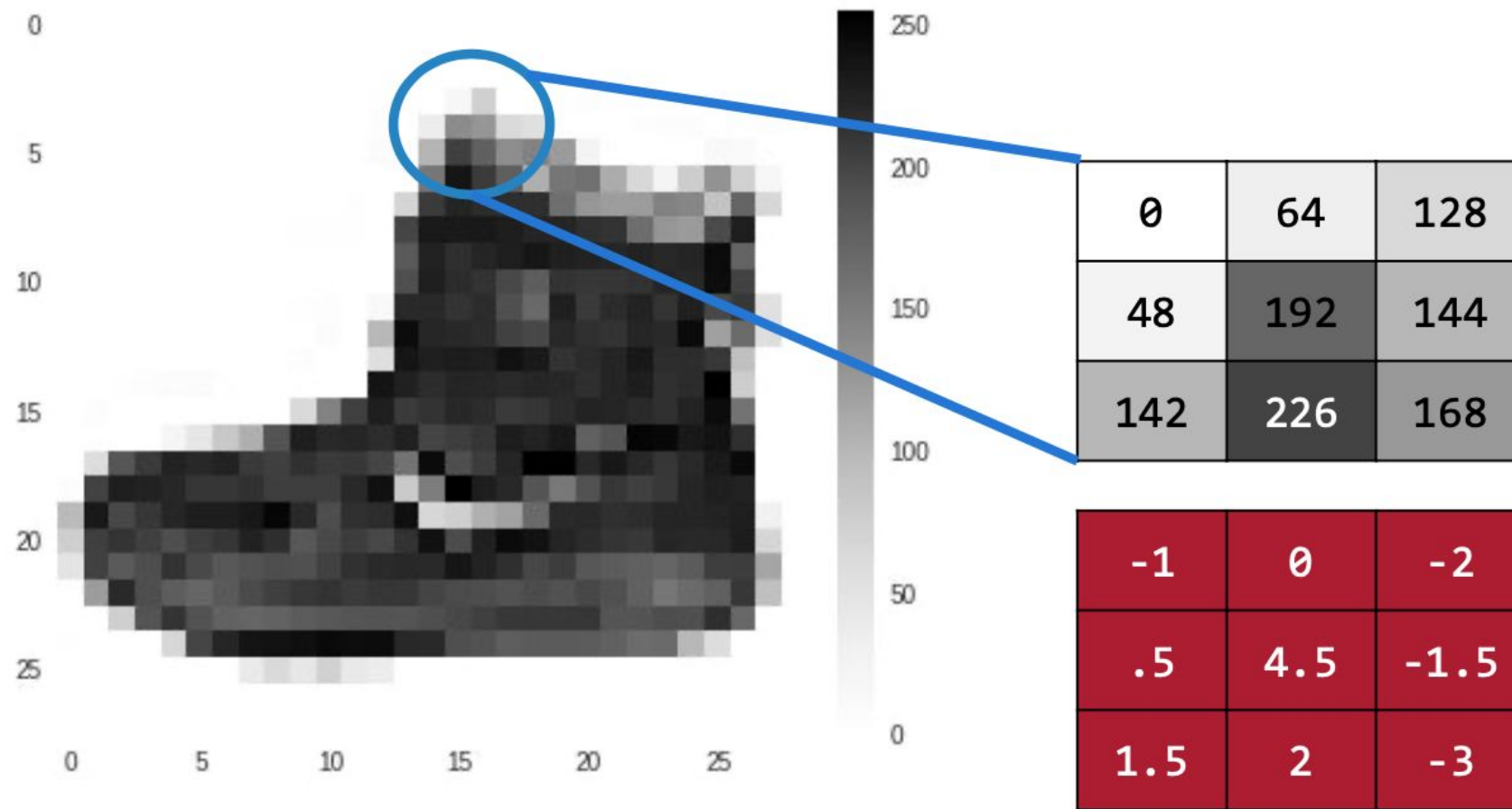
Standard Convolution (**3 Channel**—e.g., *RGB*)



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

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





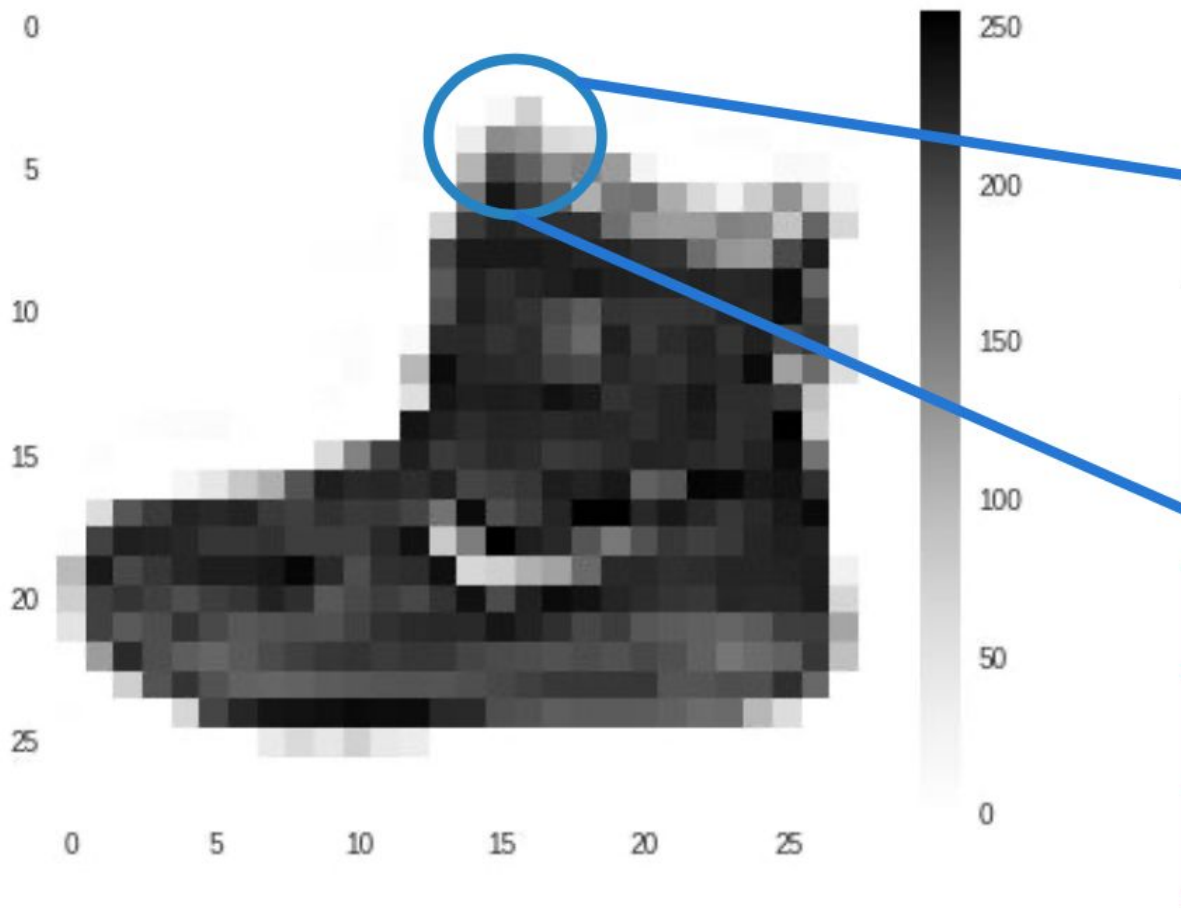
Current Pixel Value is 192

Consider neighbor Values

Filter Definition

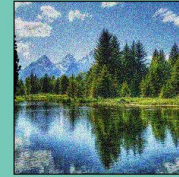
CURRENT_PIXEL_VALUE = 192

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

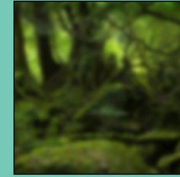


Kernels = Filters

Different Filters



Noise



Gaussian Blur



Sharpen More



Fragment



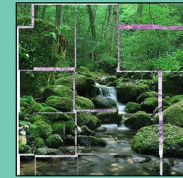
Facet



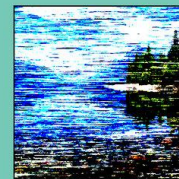
Pointillize



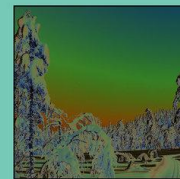
Mosaic



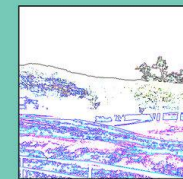
Tiles



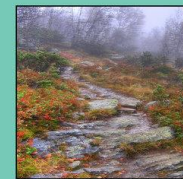
Mezzotint



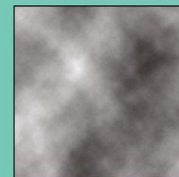
Solarize



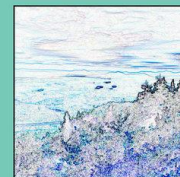
Trace Contour



Wind



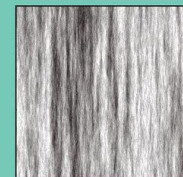
Clouds



Find Edges



Shape Blur



Fibers

Image Kernels

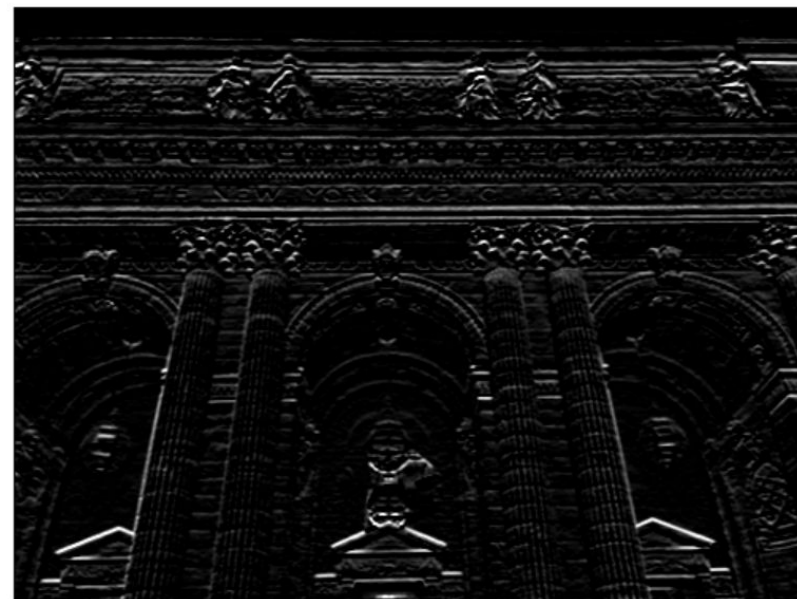


-1	0	1
-2	0	2
-1	0	1

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

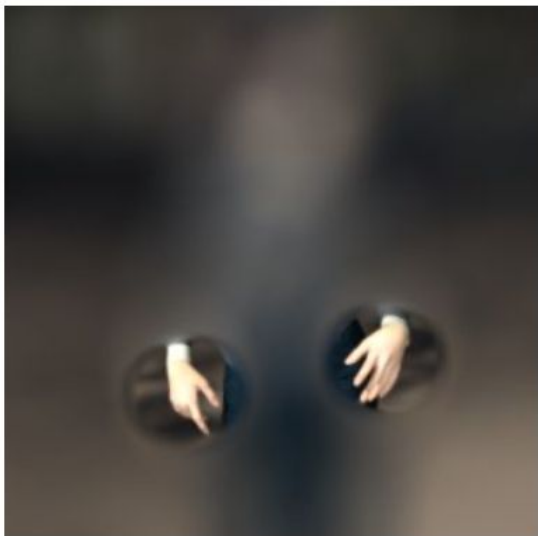
168

192	144
255	168

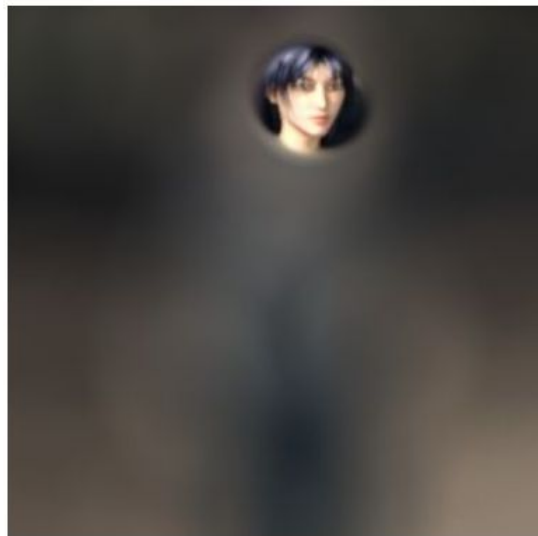
Max Pooling



+

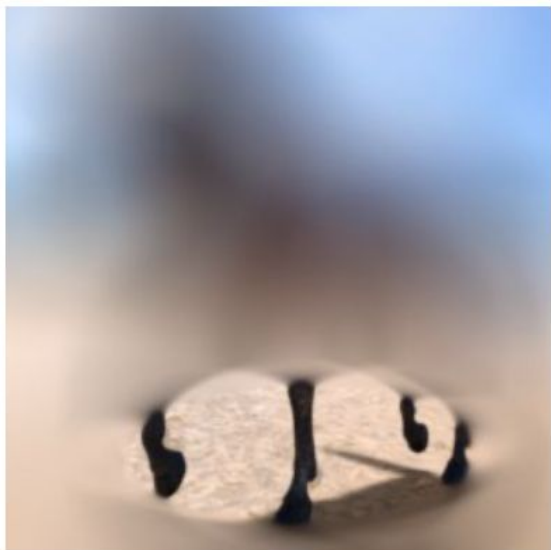


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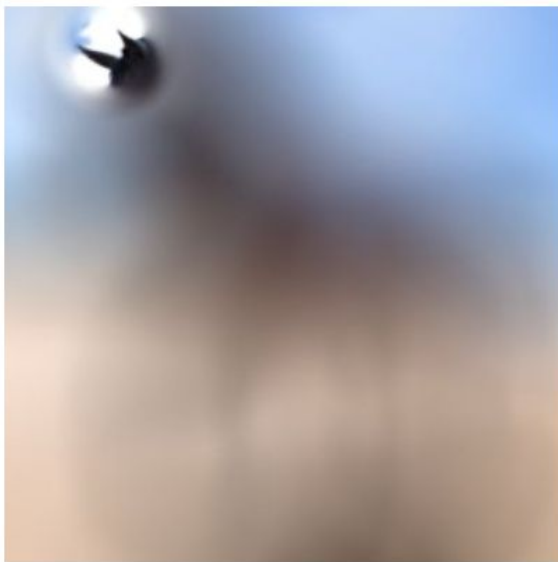


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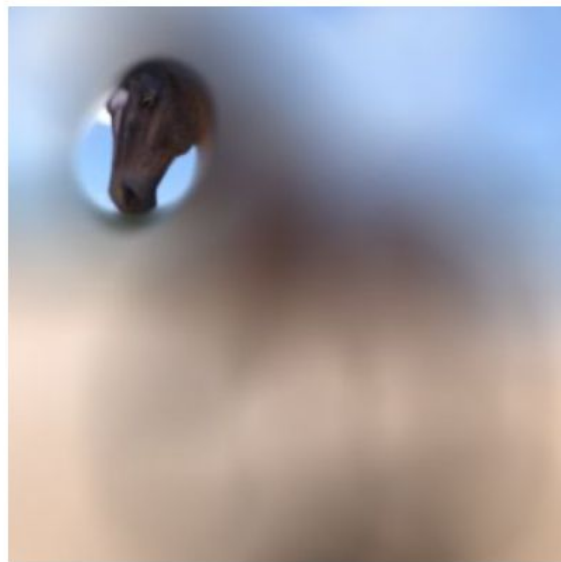
HUMAN



+



+



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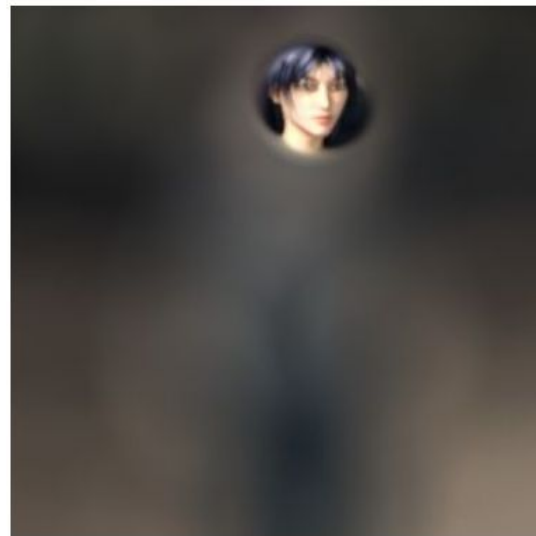
HORSE



+

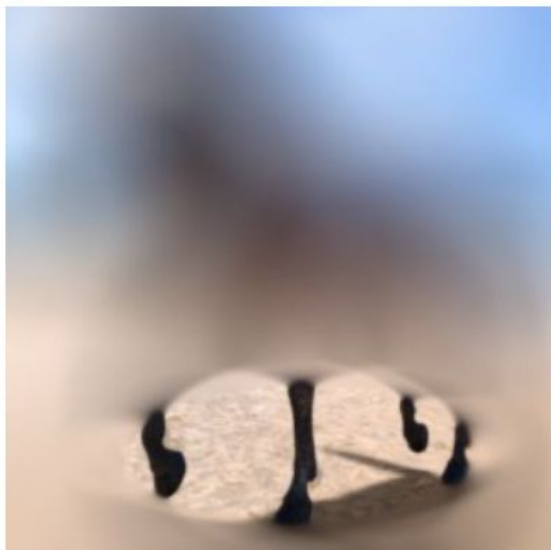


+



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HUMAN

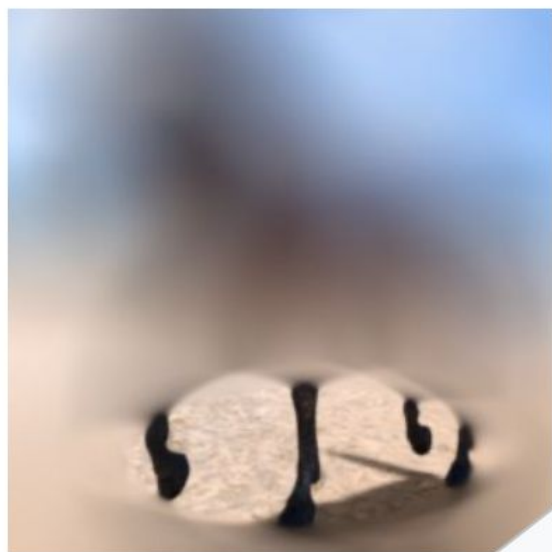


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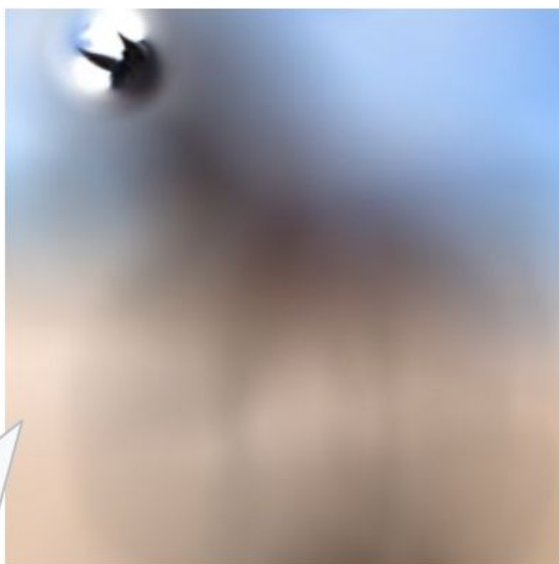


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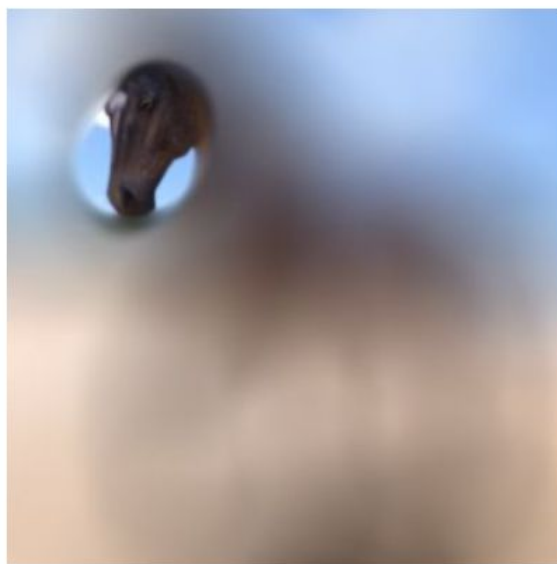
HORSE



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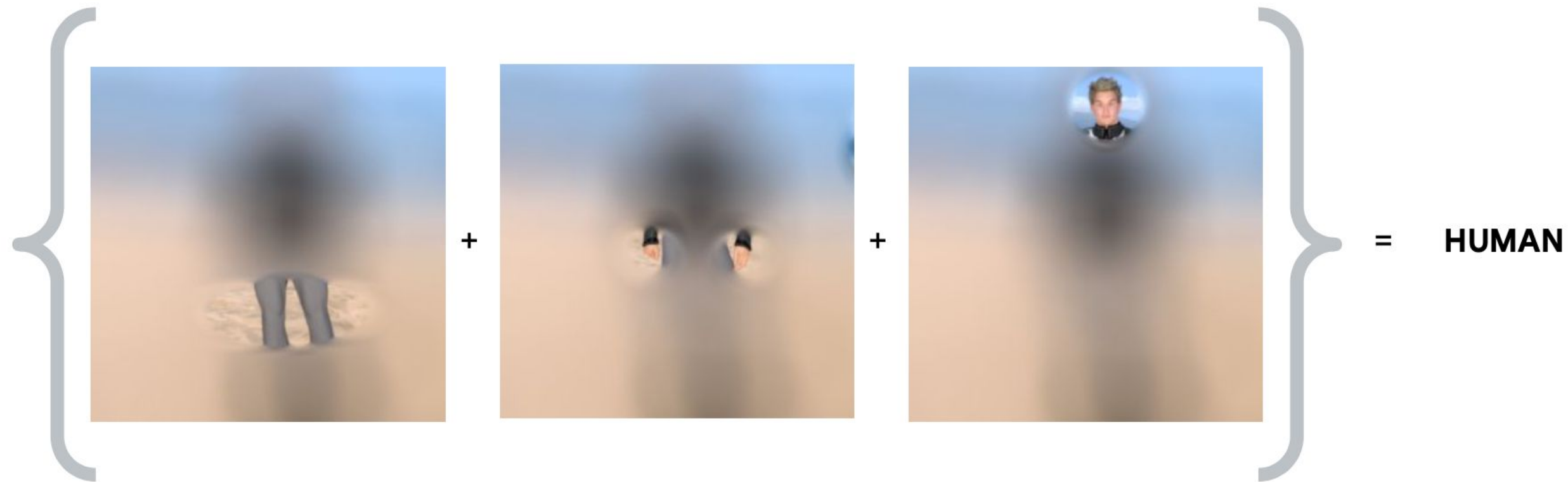
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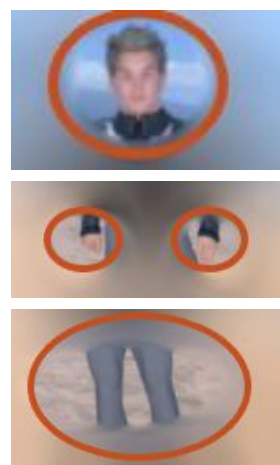
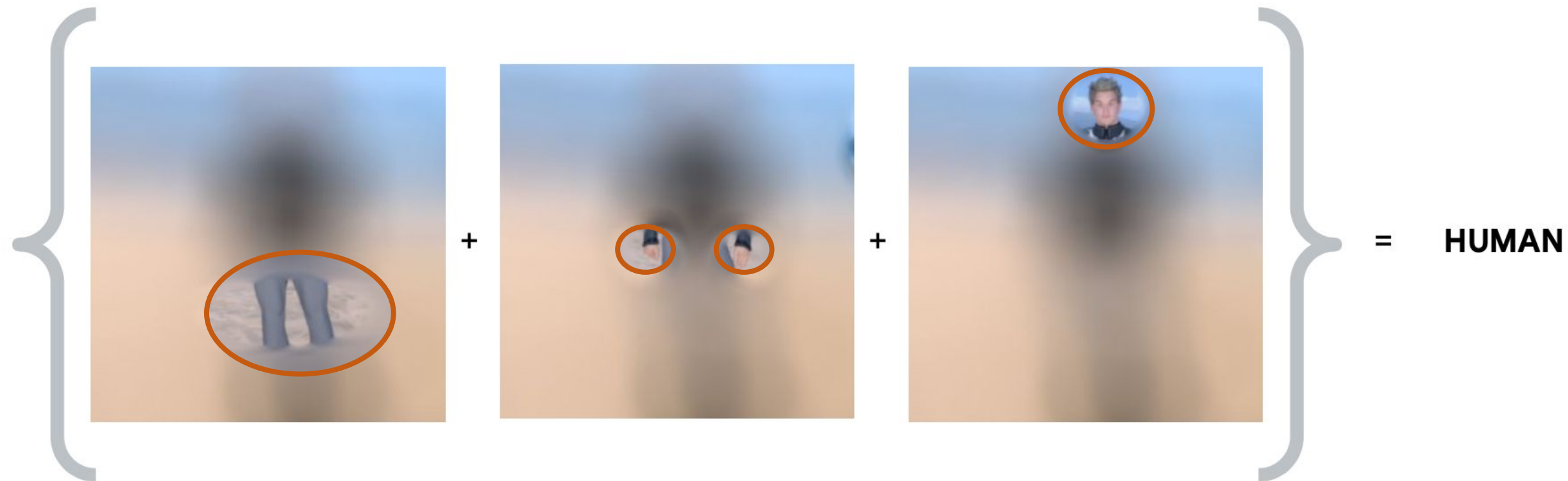


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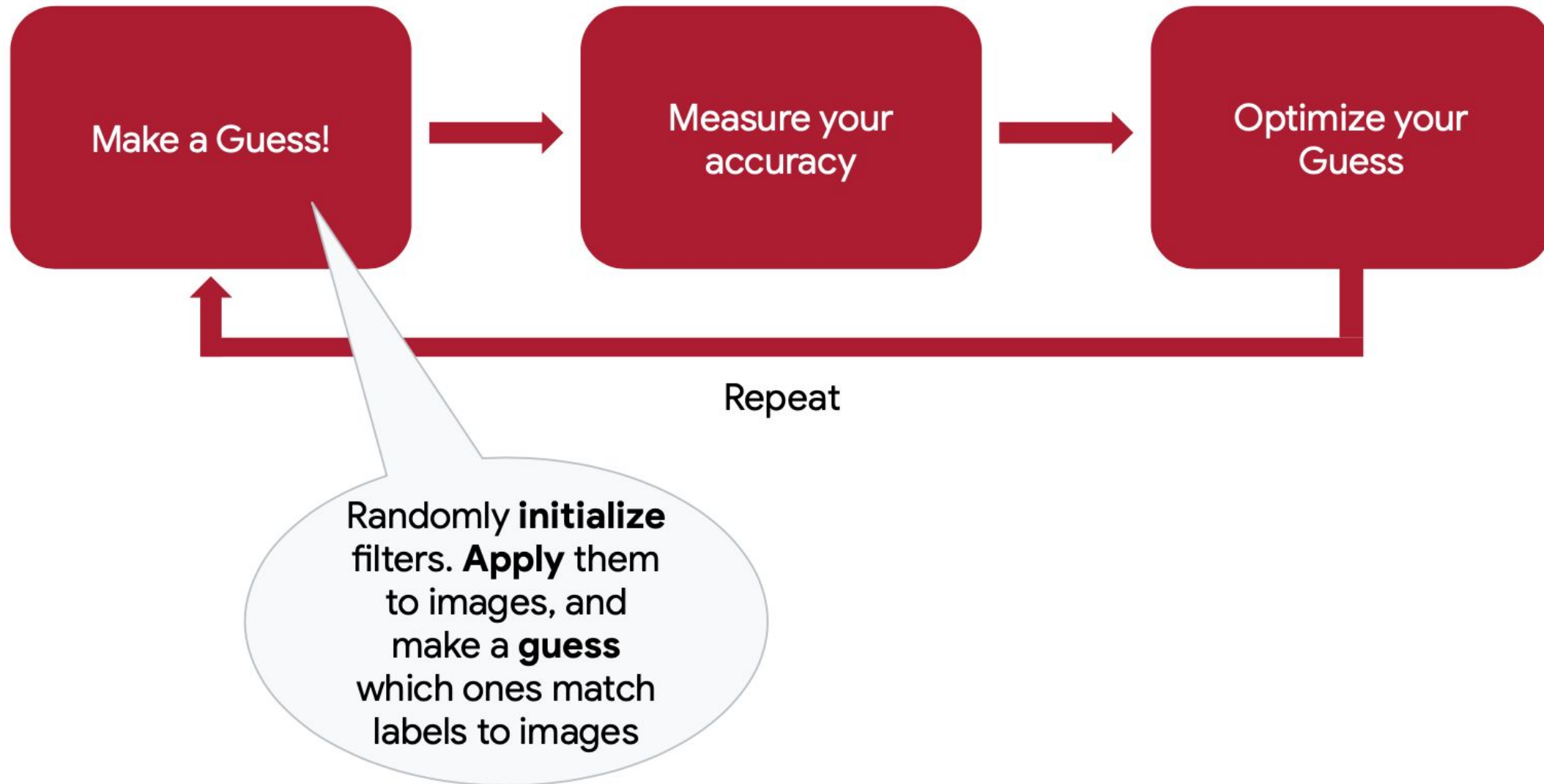
HORSE

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

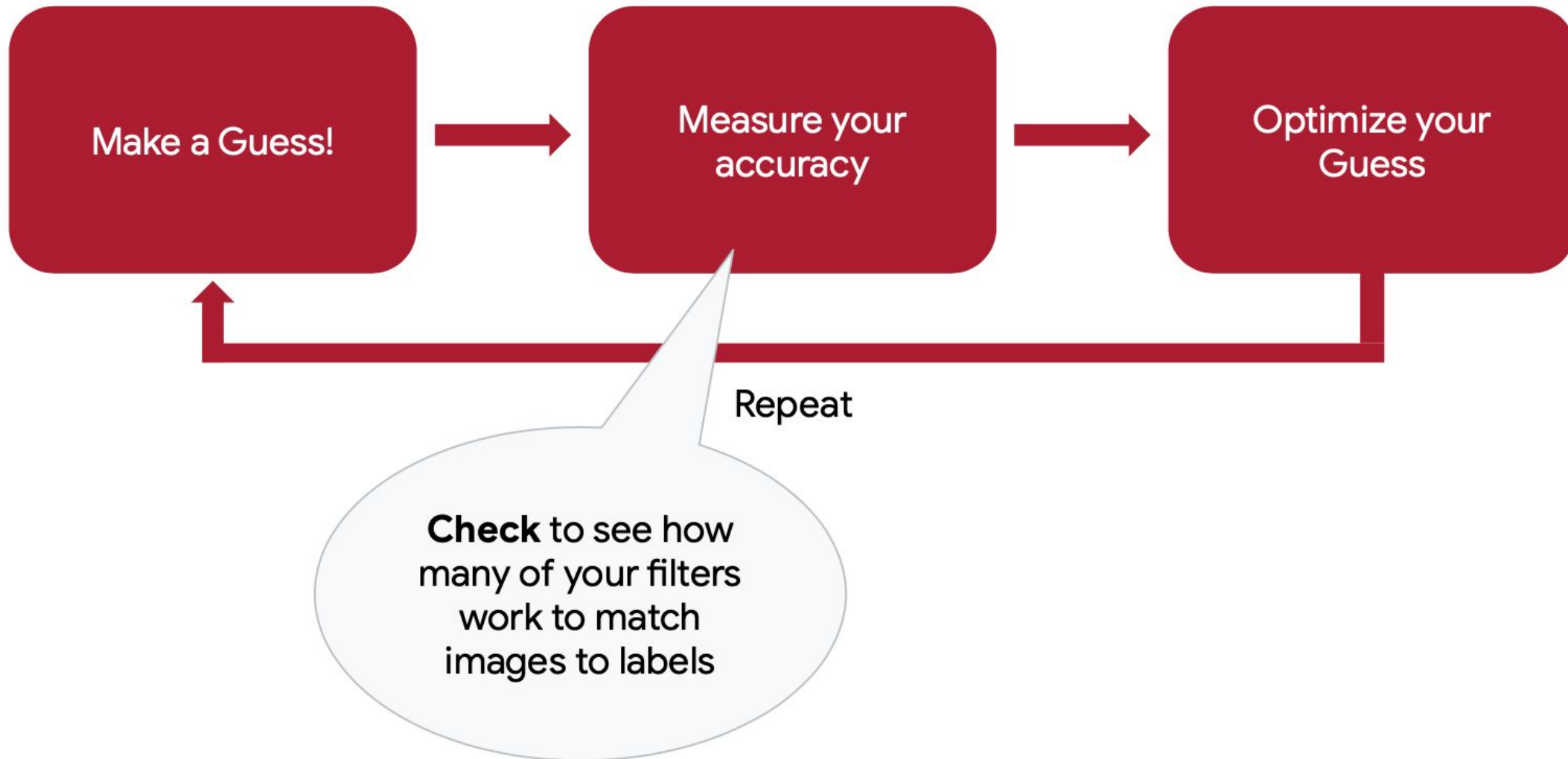




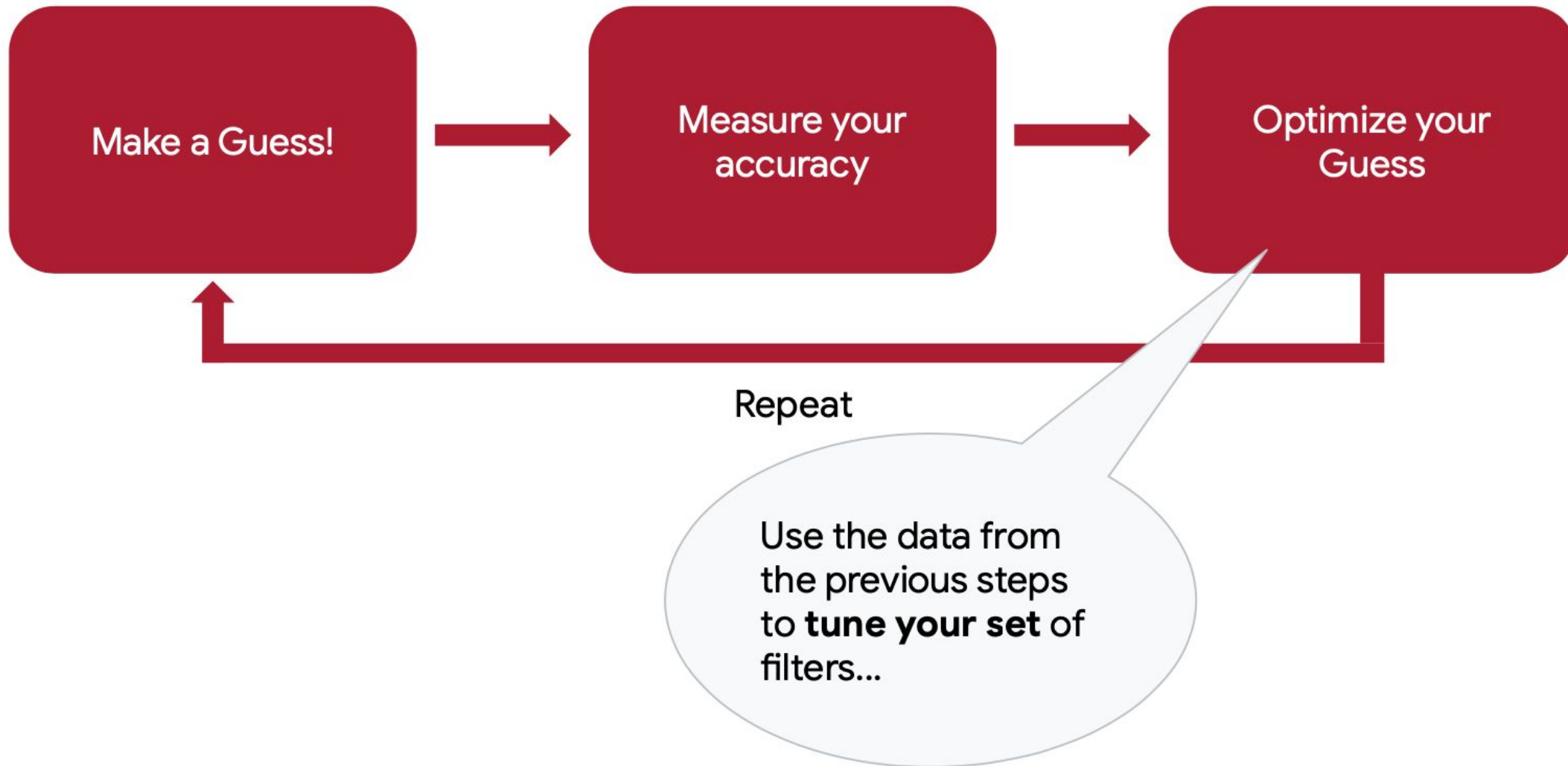
The Machine Learning Paradigm



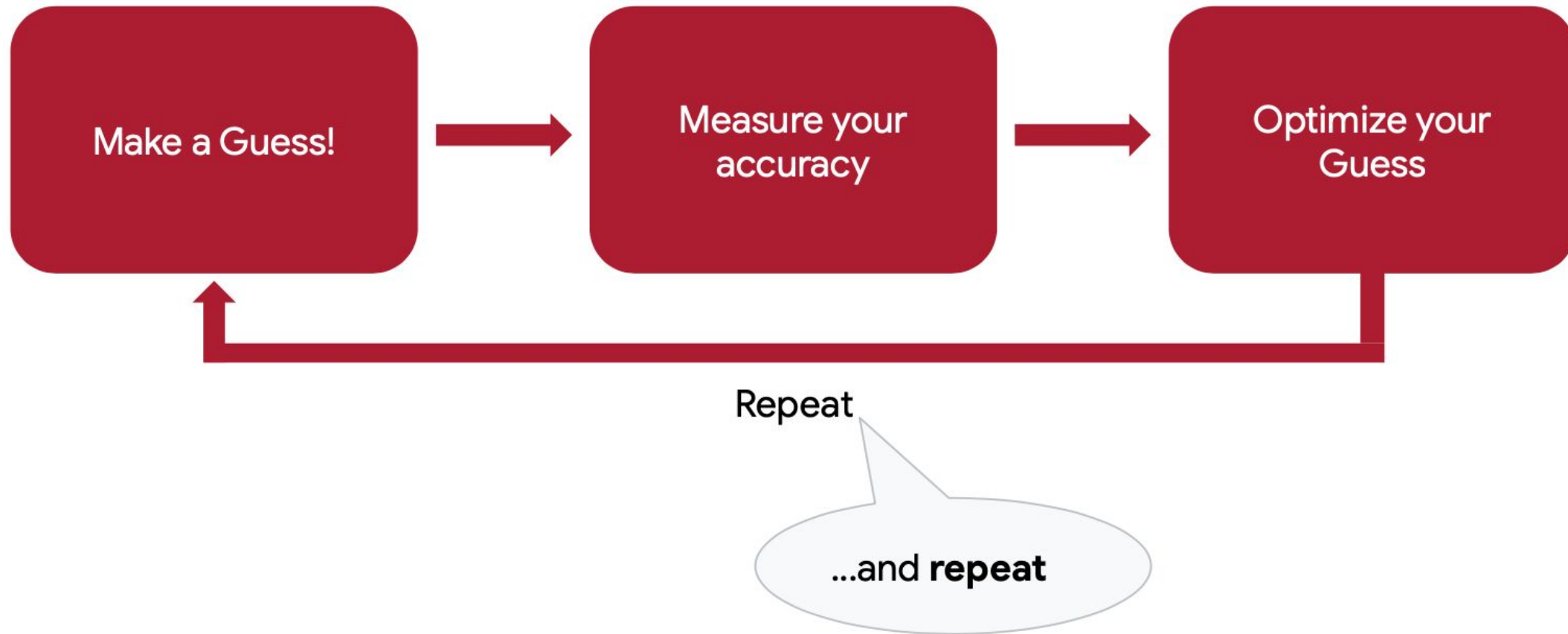
The Machine Learning Paradigm


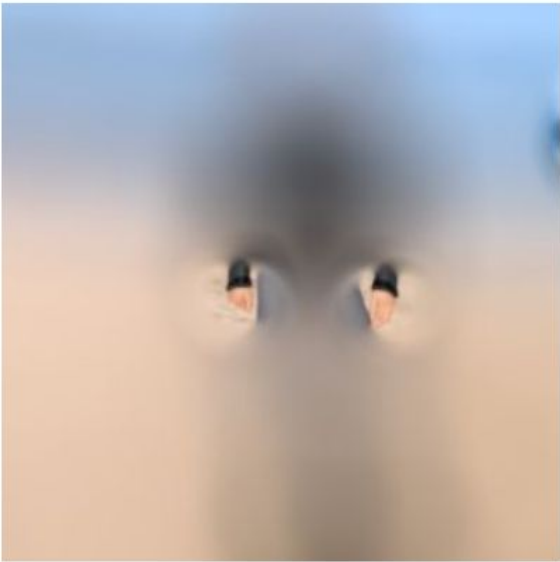
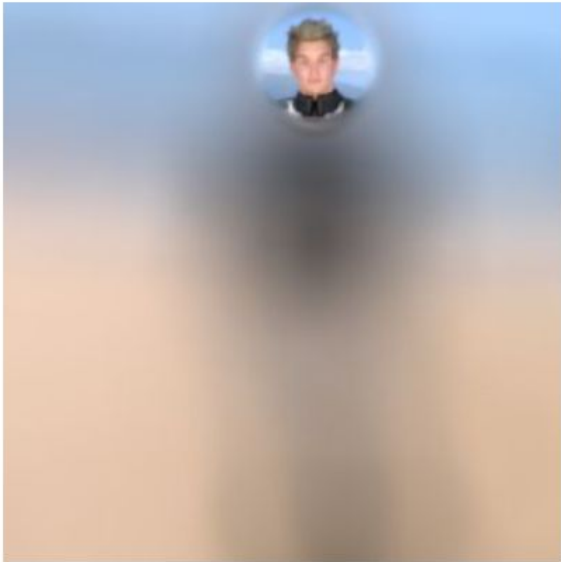


The Machine Learning Paradigm



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 \(Coursera\)](#)
- [Text Book: "TinyML" by Pete Warden, Daniel Situnayake](#)

I want to thank Shawn Hymel and Edge Impulse, Pete Warden and Laurence Moroney from Google, and especially Harvard professor Vijay Janapa Reddi, Ph.D. student Brian Plancher and their staff 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
And stay safe!



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