










# Multimodal Neurons in Artificial Neural Networks

Florian Zogaj

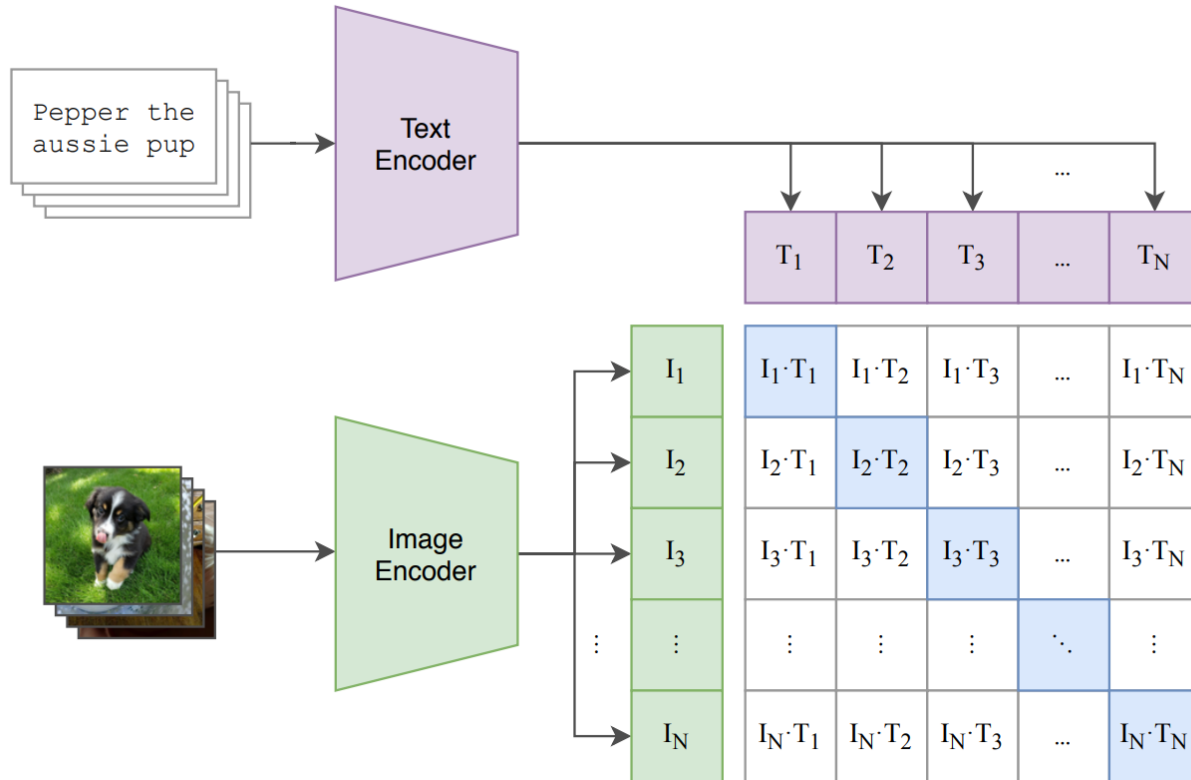


# Multimodal neurons

Biological Neuron	CLIP Neuron	Previous Artificial Neuron	
Probed via depth electrodes	Neuron 244 from penultimate layer in CLIP RN50_4x	Neuron 483, generic person detector from Inception v1	
Halle Berry	Spiderman	human face	
 <p>Responds to photos of Halle Berry and Halle Berry in costume ✓</p>	 <p>Responds to photos of Spiderman in costume and spiders ✓</p> <p><a href="#">view more</a></p>	 <p>Responds to faces of people ✓</p>	Photorealistic images
 <p>Responds to sketches of Halle Berry ✓</p>	 <p>Responds to comics or drawings of Spiderman and spider-themed icons ✓</p> <p><a href="#">view more</a></p>	 <p>Does not respond significantly to drawings of faces ✗</p>	Conceptual drawings
 <p>Responds to the text "Halle Berry" ✓</p>	 <p>Responds to the text "spider" and others ✓</p> <p><a href="#">view more</a></p>	 <p>Does not respond significantly to text ✗</p>	Images of text

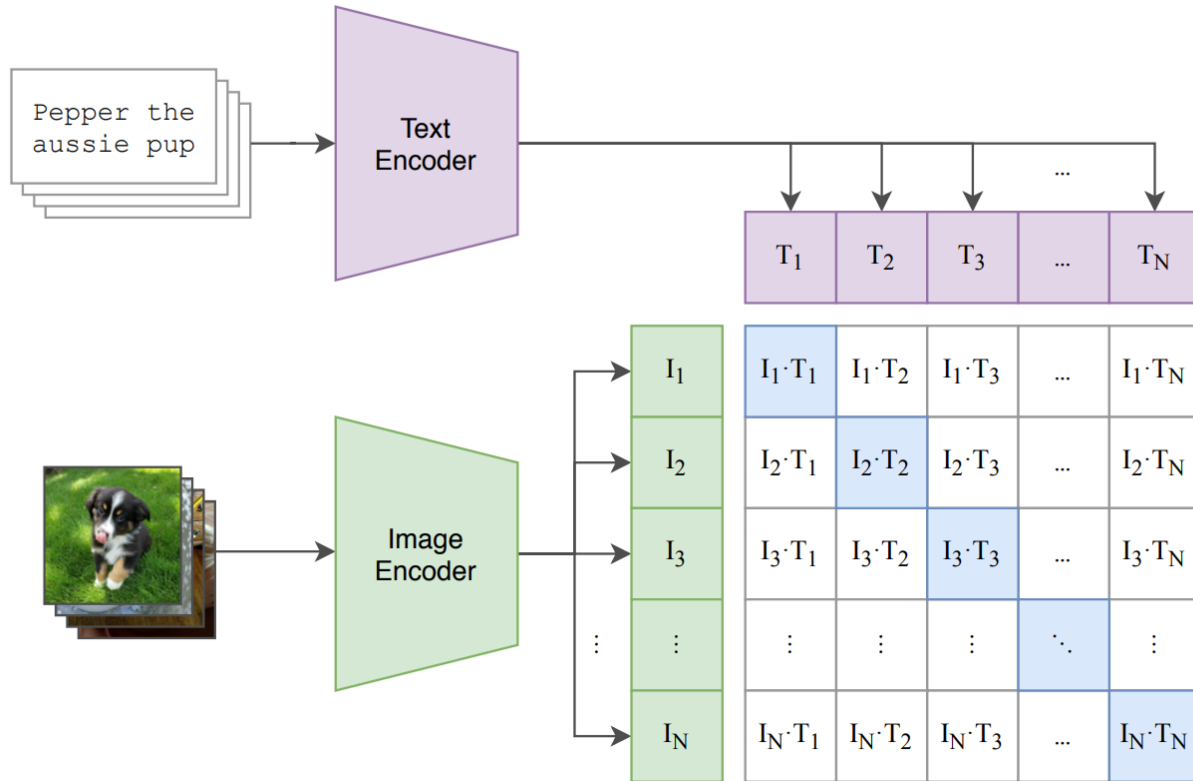
# CLIP

## (1) Contrastive pre-training

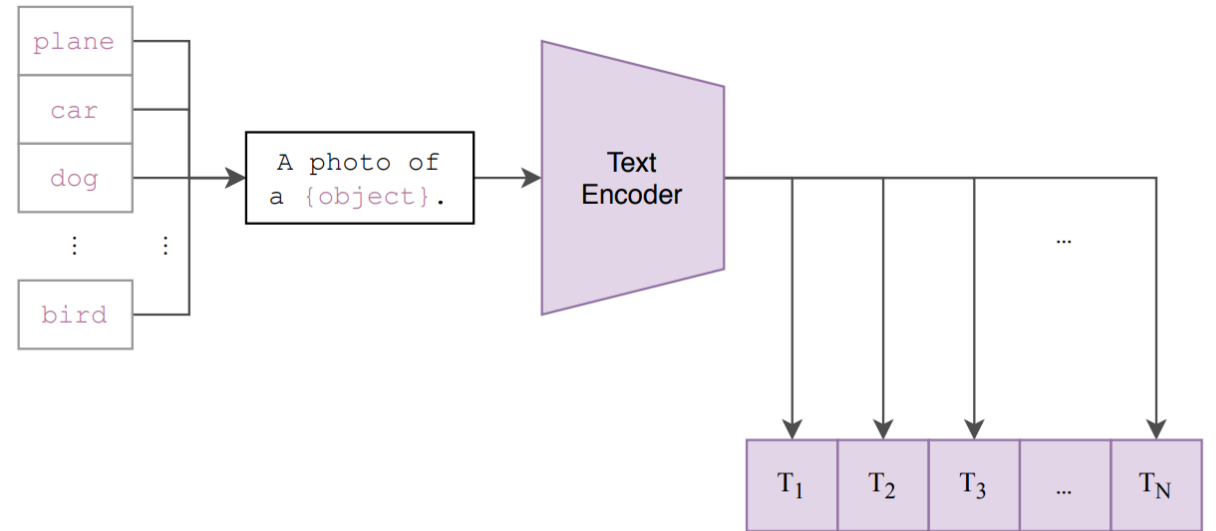


# CLIP

## (1) Contrastive pre-training

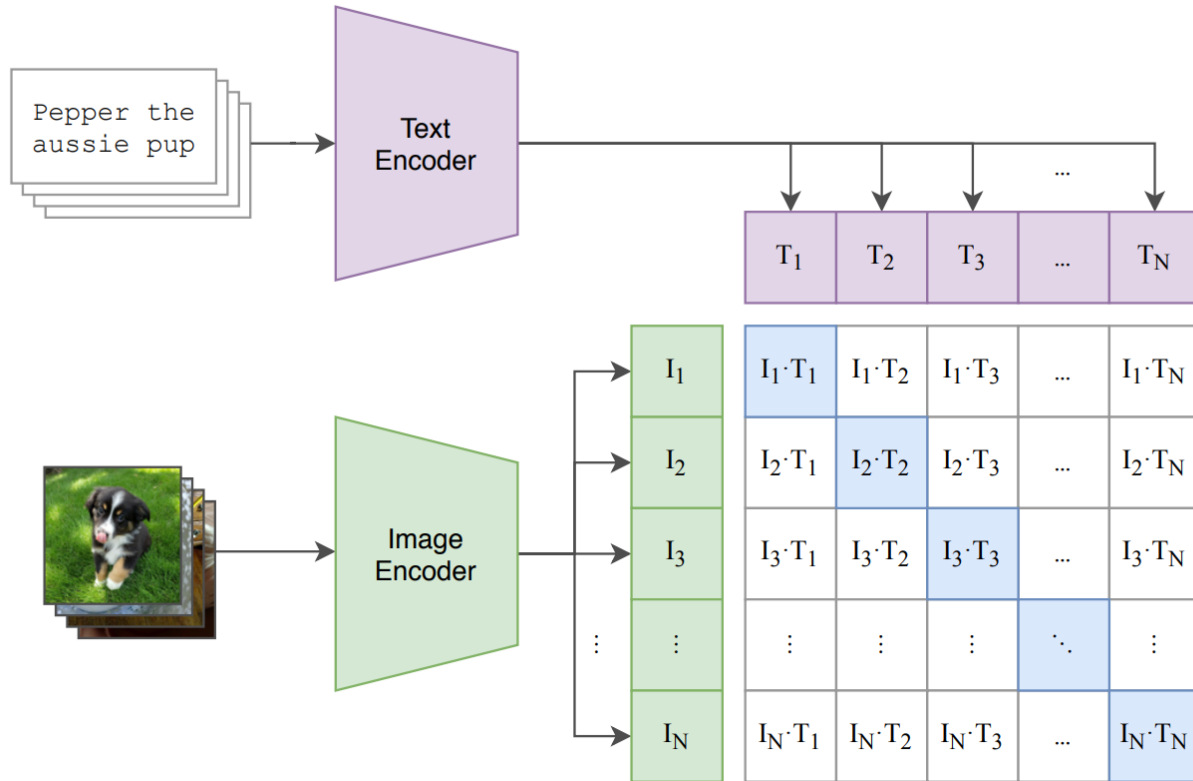


## (2) Create dataset classifier from label text

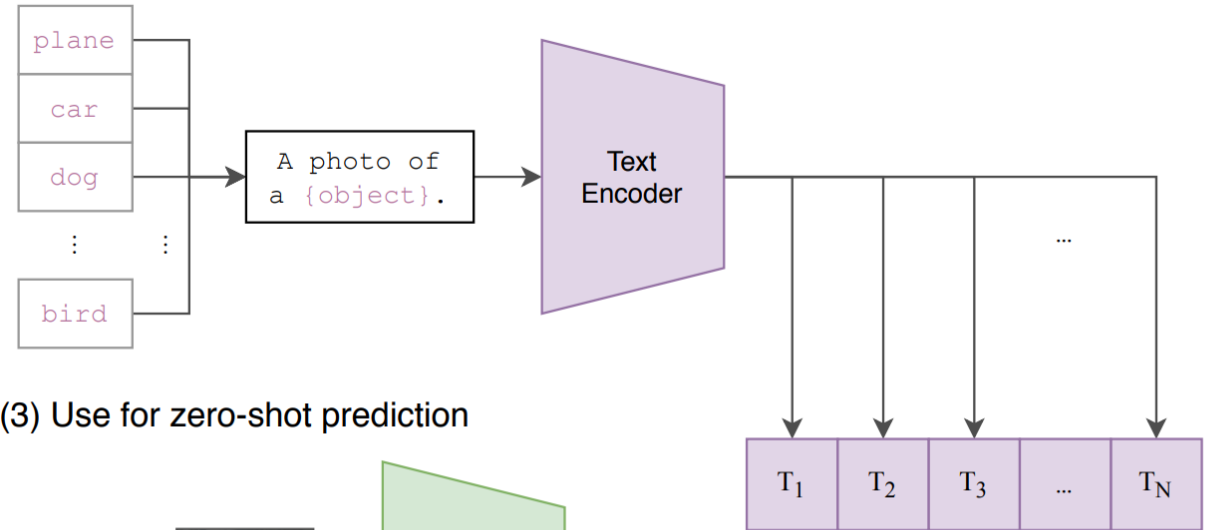


# CLIP

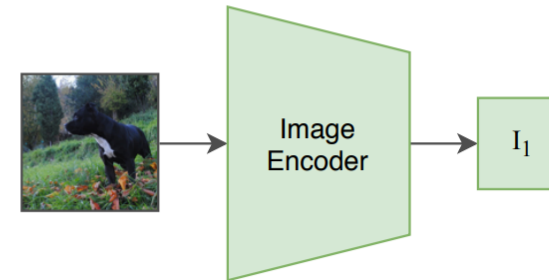
## (1) Contrastive pre-training



## (2) Create dataset classifier from label text

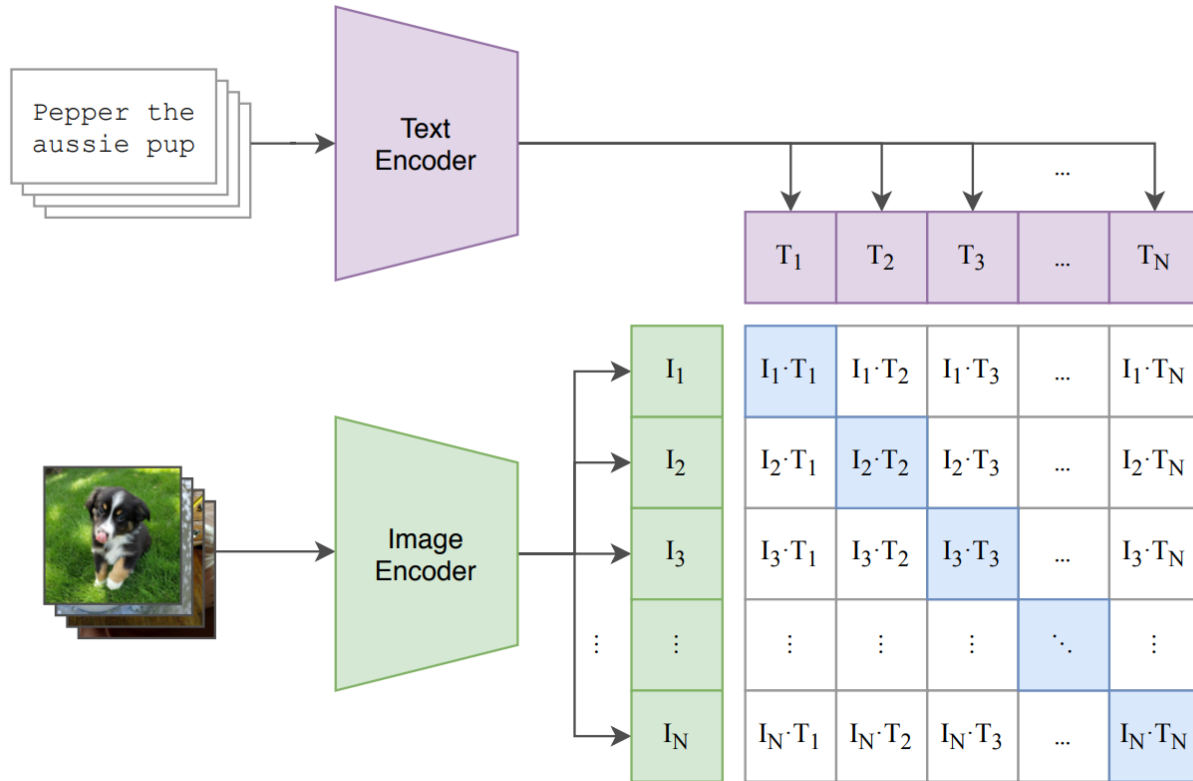


## (3) Use for zero-shot prediction

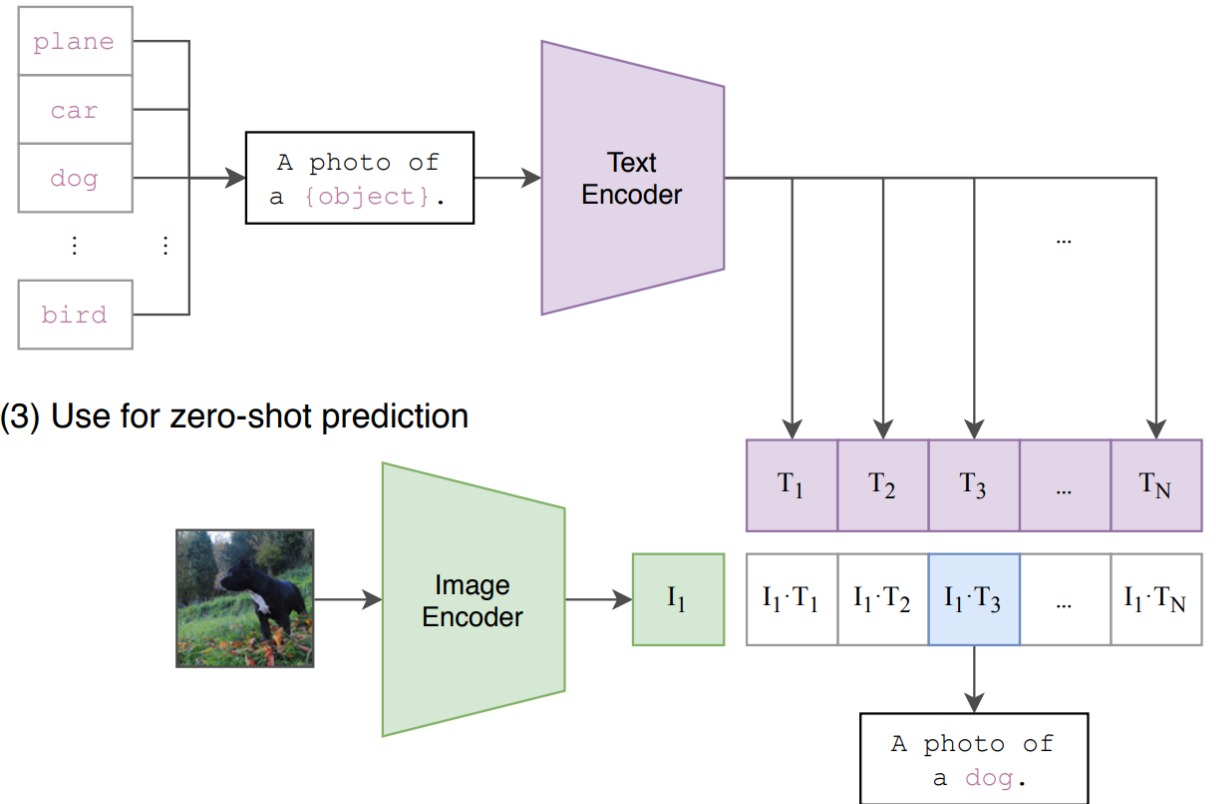


# CLIP

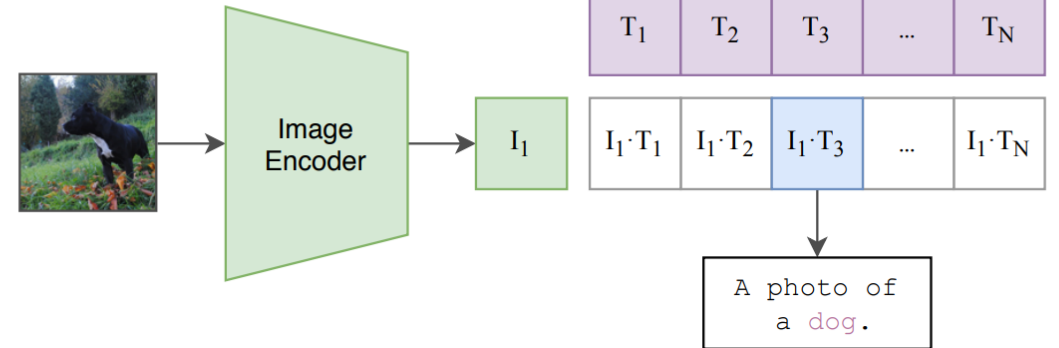
## (1) Contrastive pre-training



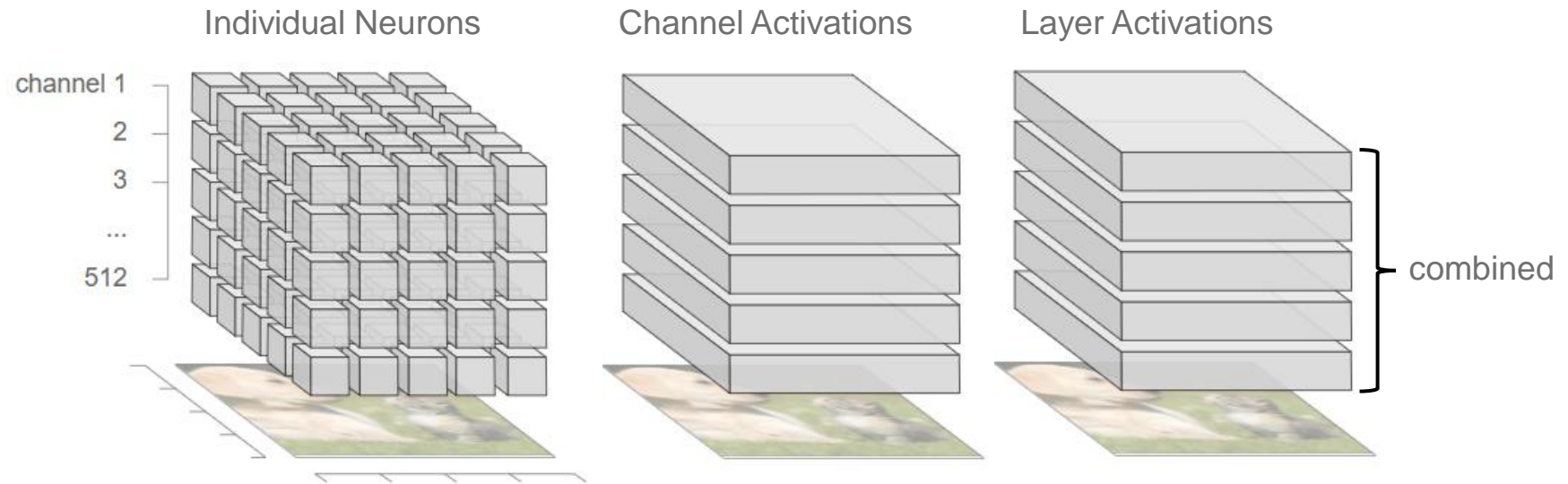
## (2) Create dataset classifier from label text



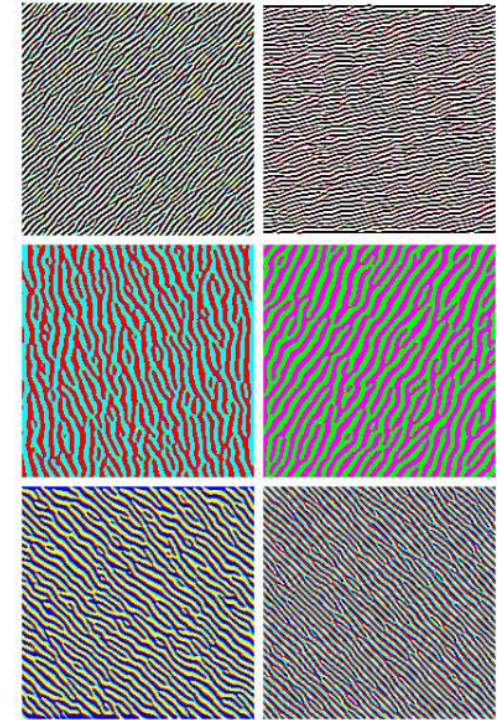
## (3) Use for zero-shot prediction



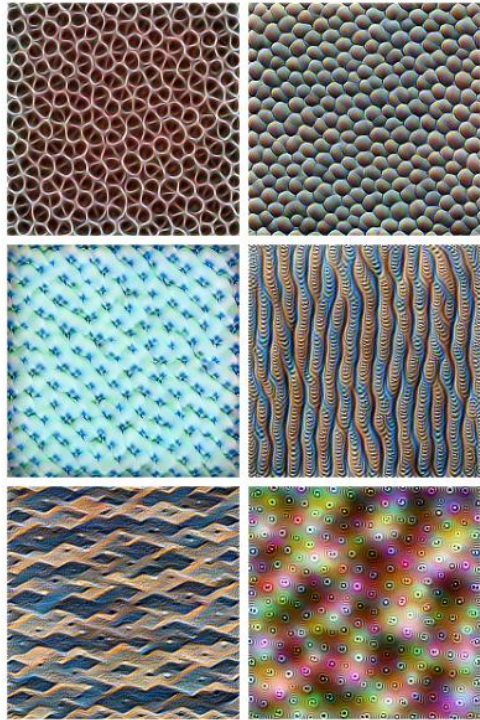
# What is a neuron, channel, layer?



# Feature Visualization



**Edges** (layer conv2d0)



**Textures** (layer mixed3a)



**Patterns** (layer mixed4a)



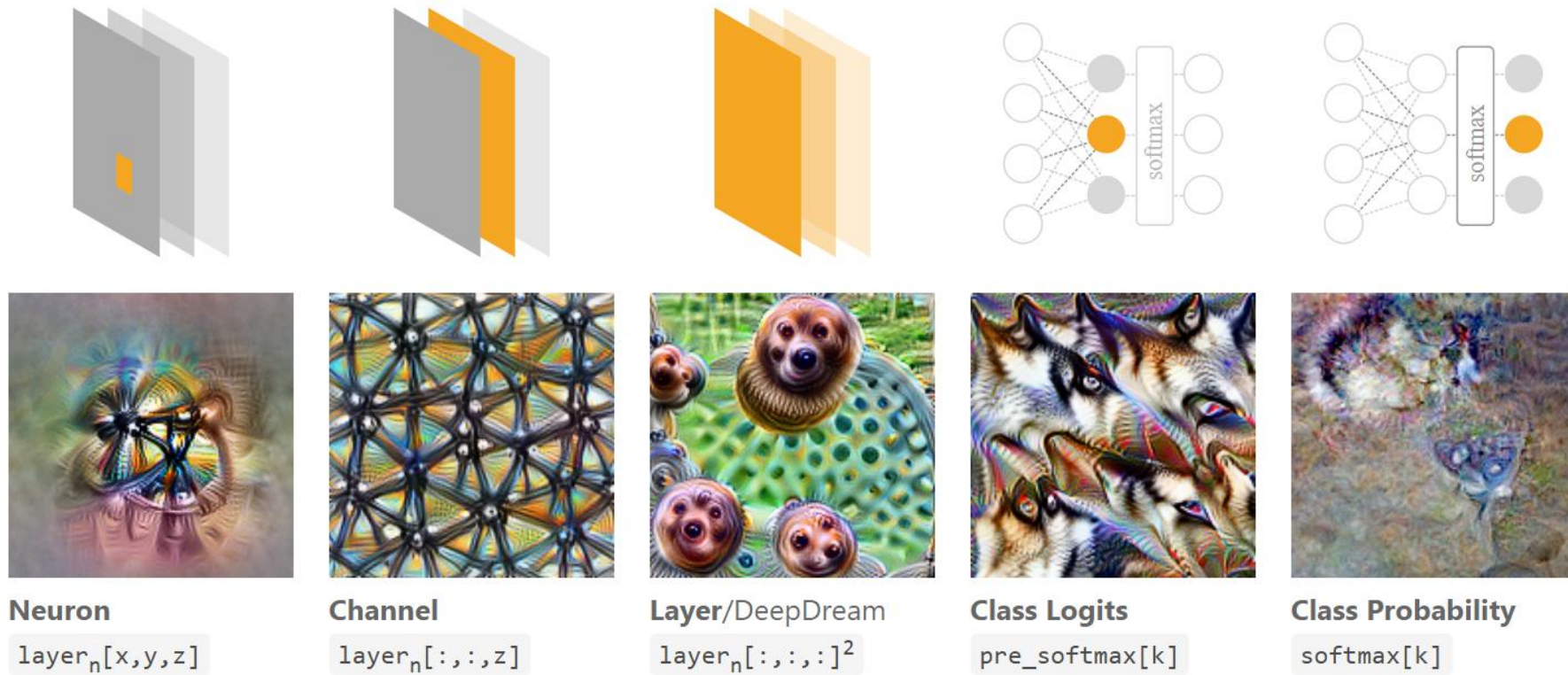
**Parts** (layers mixed4b & mixed4c)



**Objects** (layers mixed4d & mixed4e)



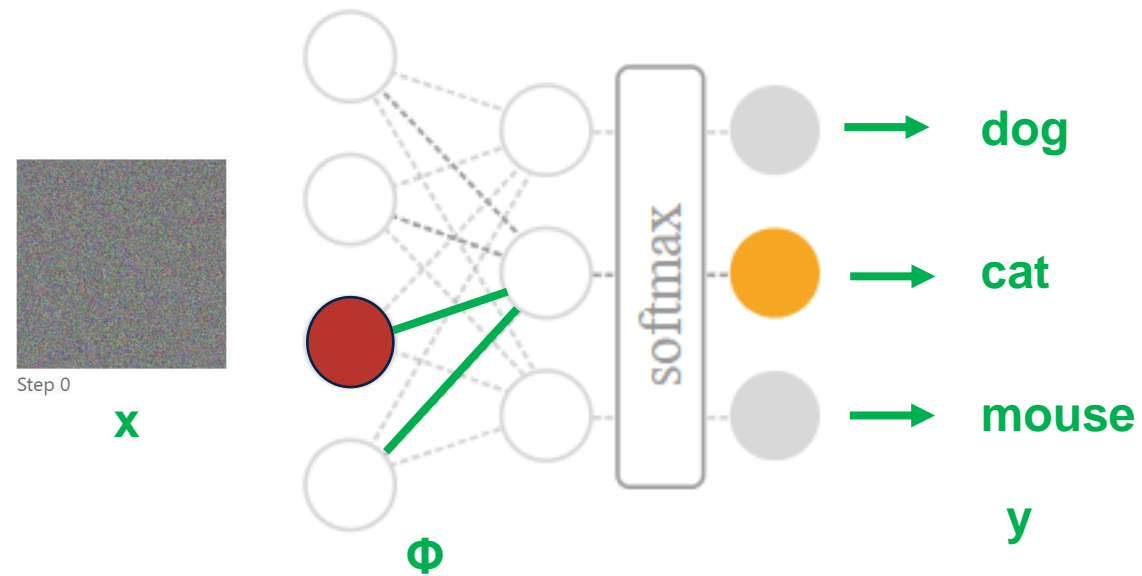
# Neuron, Channel and Layer activations



Feature visualization is a 'powerful tool for understanding what our models are doing' - Geoffrey Hinton, Google Researcher and Pioneer in Neural Networks.

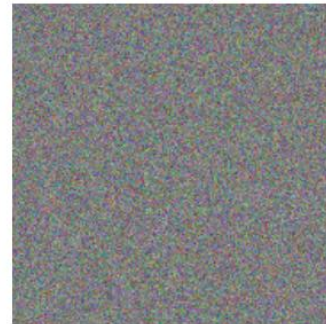
# Feature Visualization

- Optimize for neuron



# Feature Visualization

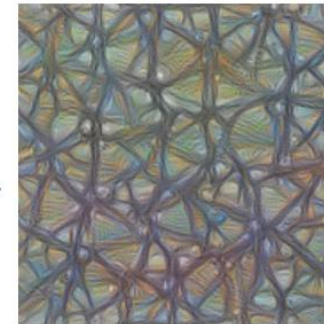
Starting from random noise, we optimize an image to activate a particular neuron (layer mixed4a, unit 11).



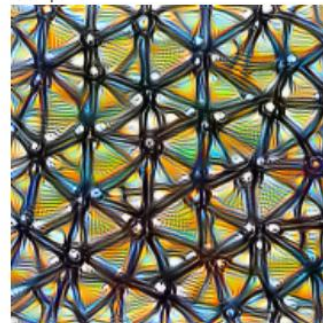
Step 0



Step 4



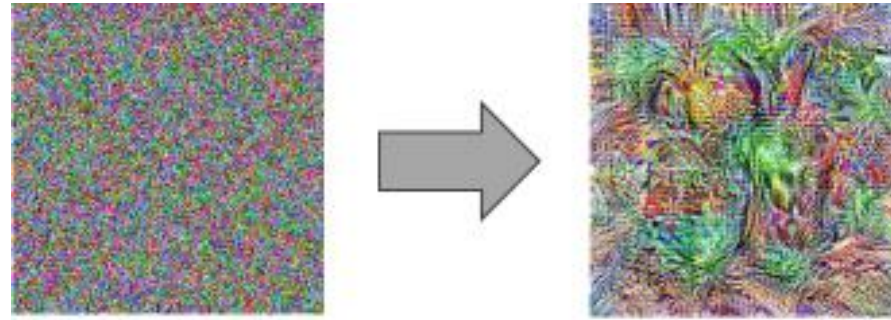
Step 48



Step 2048

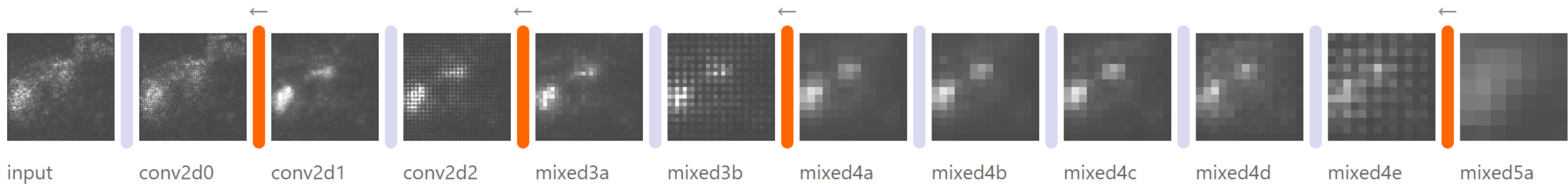
# Activation Maximization

$$x_{t+1} = x_t + \varepsilon_1 \cdot \frac{\partial_a(\theta, x_t)}{\partial_{x_t}}$$



a) Random initialization

b) Adversarial example



# Activation Maximization - Regularization

- Frequency penalization
- Transformation robustness
- Learned priors

$$x_{t+1} = r(x_t) + \varepsilon_1 \cdot \frac{\partial_a(x_t)}{\partial_{x_t}}$$



# Multi-faceted neurons

- Diverse images as seed
- Add diversity term to optimization
- Faceted feature visualization

Reconstructions of multiple feature types (facets) recognized by the same "grocery store" neuron



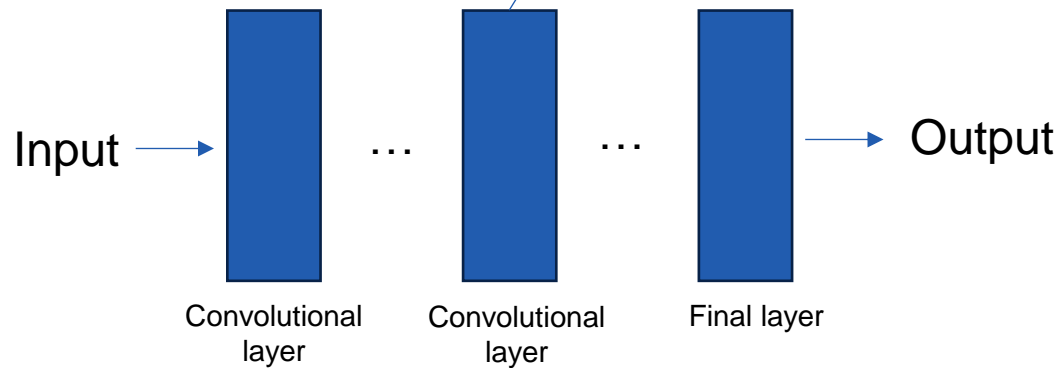
Corresponding example training set images recognized by the same neuron as in the "grocery store" class



# Faceted Feature Visualization



Train linear classifier (distinguish)



$$f(g(x)) + w^T \left( g(x) \odot \nabla f(g(x)) \right)$$

- $g(x)$  activations up to intermediate layer
- $f(g(x))$  activations of all layers
- $w$  trained weights from linear classifier

# CLIP RN50-x4 – What kind of neurons were found?





# Person Neurons

Logo



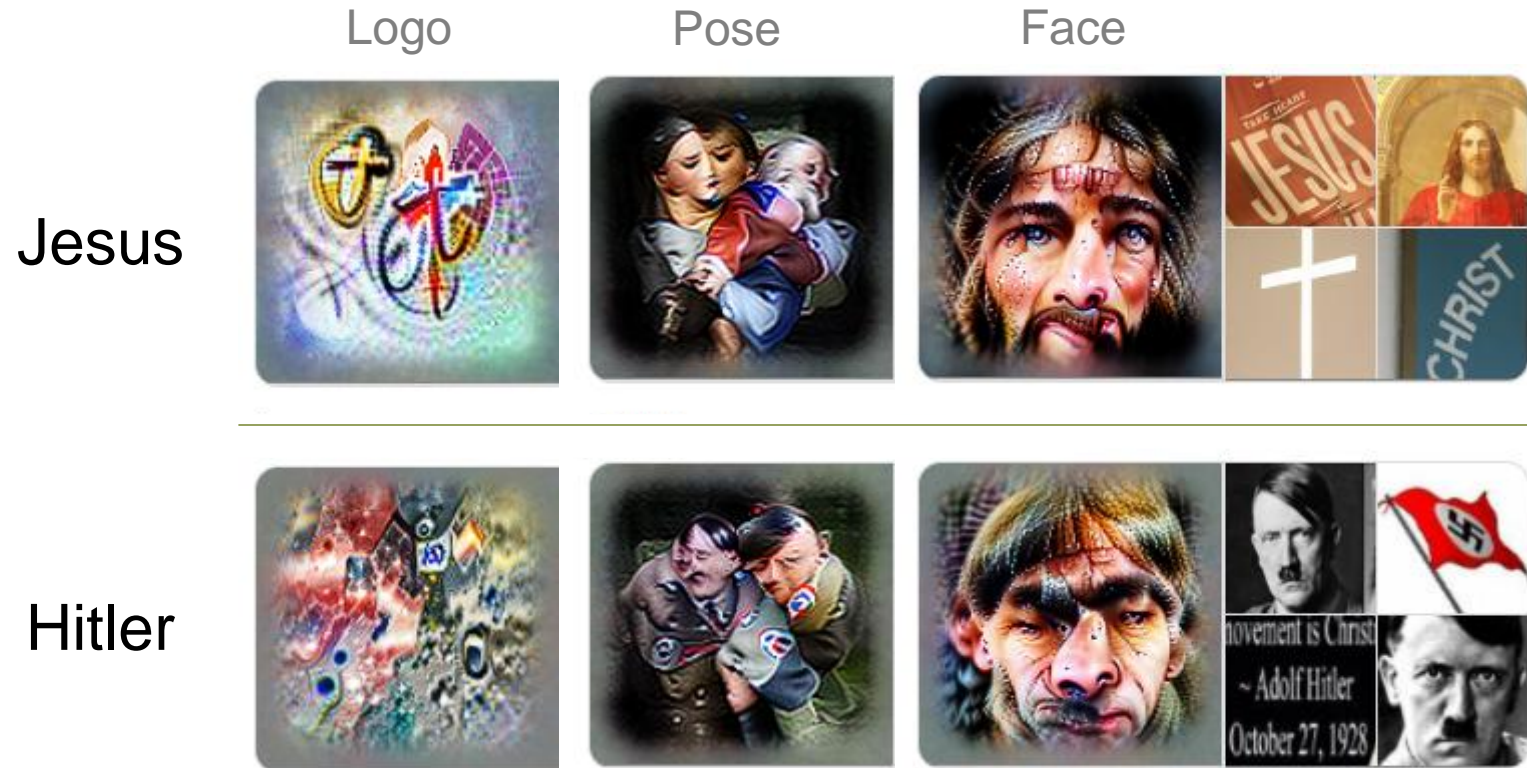
Pose



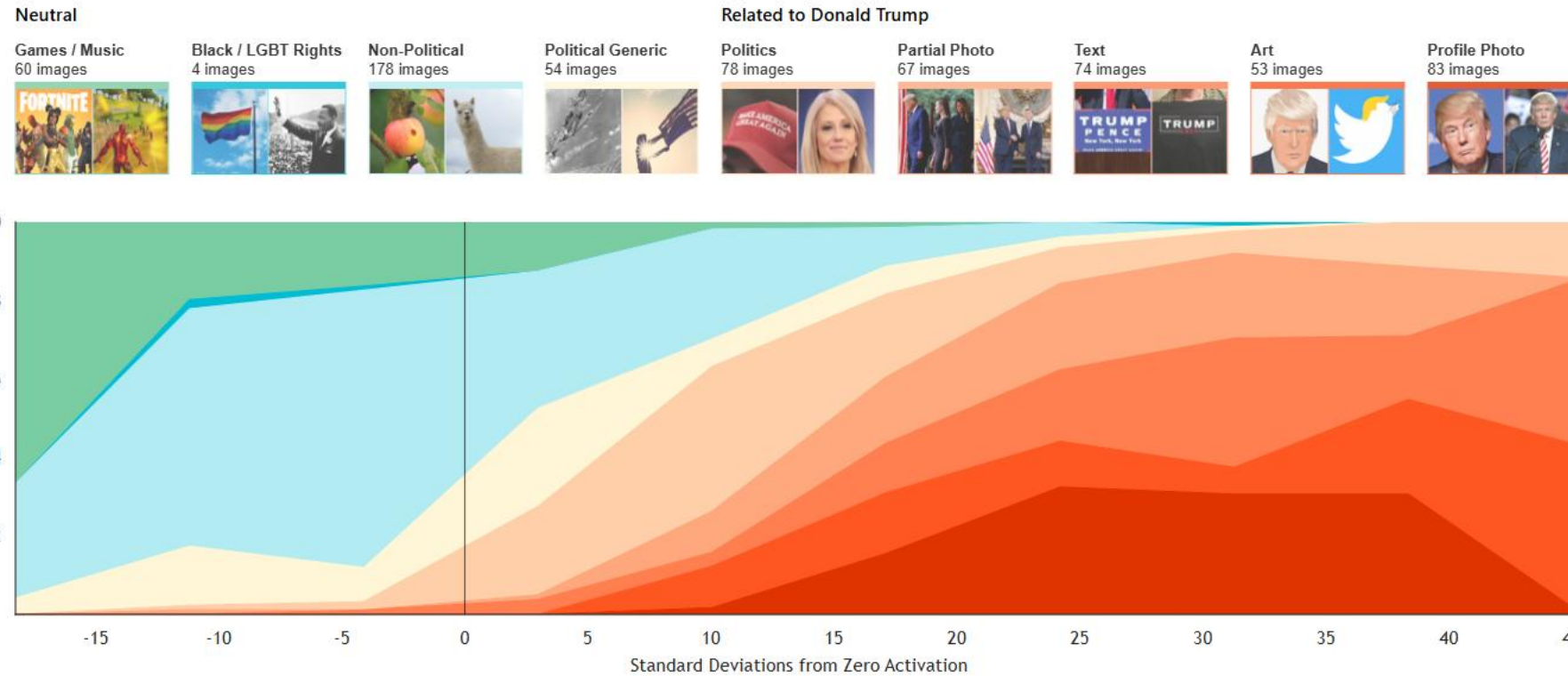
Face



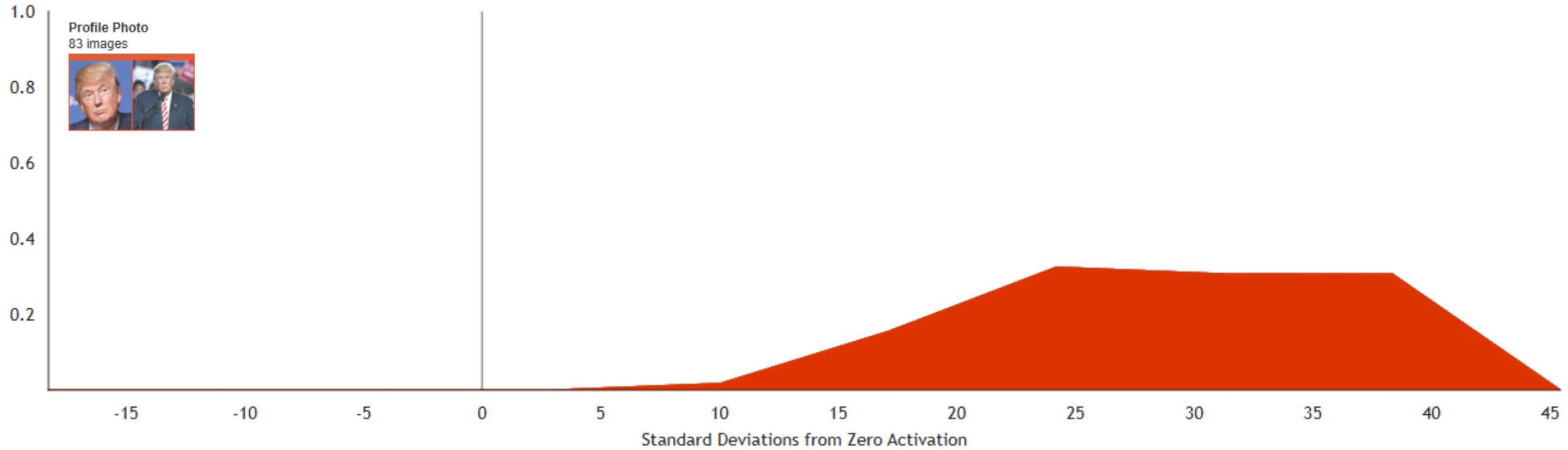
# Person Neurons



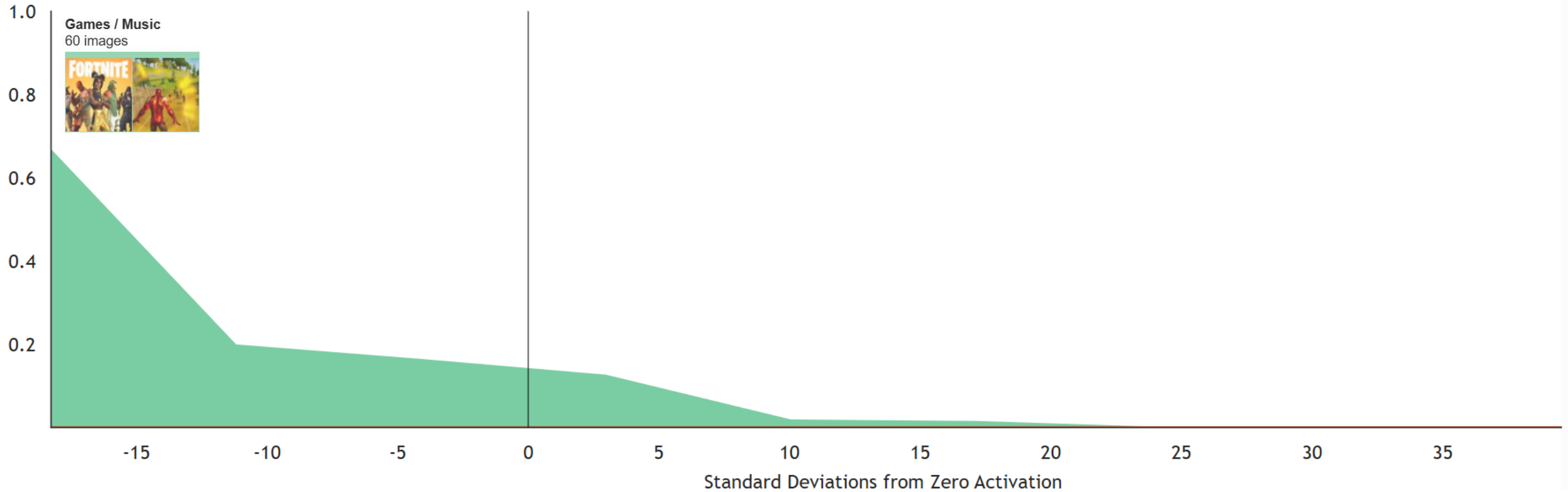
# Case study – Donald Trump neuron activation



# Case study – Donald Trump neuron activation



# Case study – Donald Trump neuron activation



# Emotion Neurons

- Facial expression
- Body language
- Text



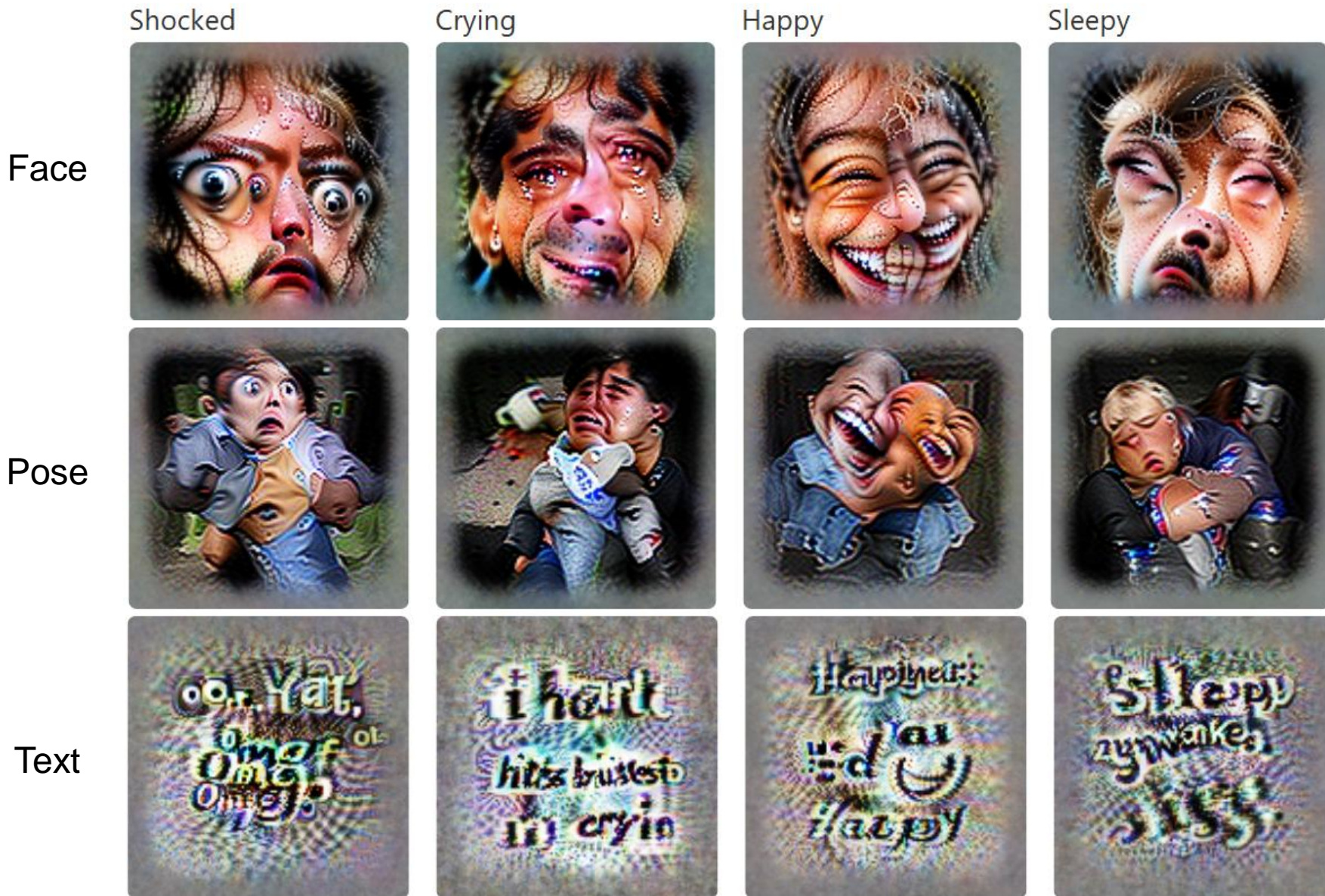
# Emotion Neurons

- Facial expression
- Body language
- Text



Surprise / Shock

# Emotion Neurons





# Region Neurons

Text



Face



Architecture



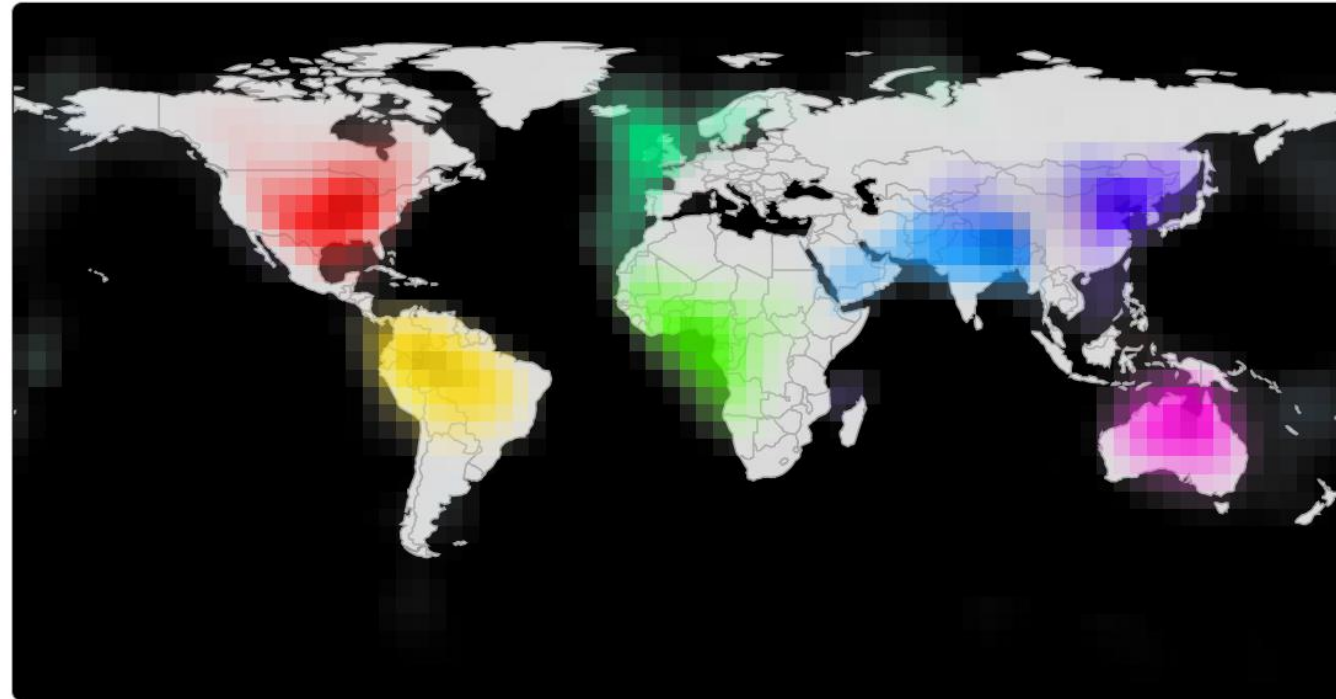
Logo



# Region Neurons

	Text	Face	Architecture	Logo
USA				
West Africa				

# Region Neurons

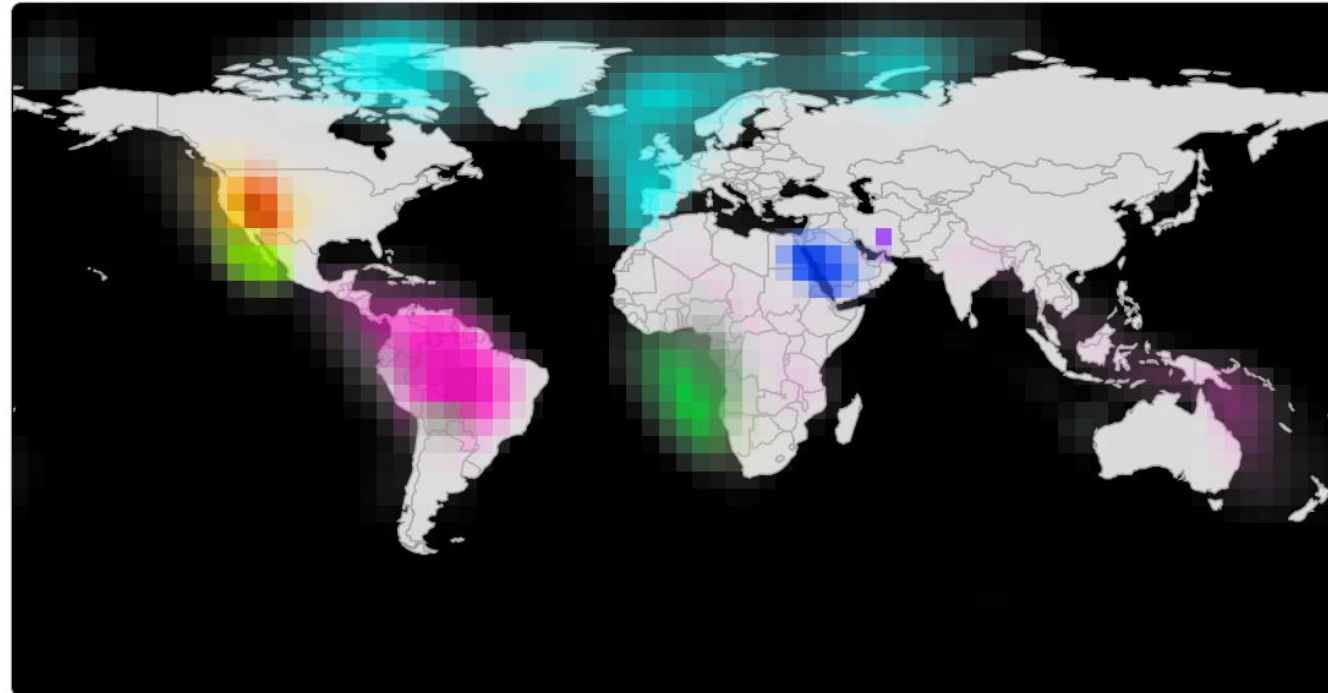


## Most Activating Words

Words which most activate these neurons when rastered into images, out of 10,000 most common English words.

americans, american, america, usa, americas	portuguese, eu, madrid, argentina, portugal	ghana, uganda, africa, tanzania, african	netherlands, luxembourg, stockholm, amsterdam, switzerland	mumbai, singh, pakistan, afghanistan, bangladesh	shanghai, asian, vietnamese, cambodia, chinese	australian, australia, adelaide, nsw, queensland
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# Region Neurons



## Most Activating Words

Words which most activate these neurons when rastered into images, out of 10,000 most common English words.

angel, angels, wings, heaven, angeles	entrepreneurs, entrepreneur, founder, startup, starter	immigrants, immigration, borders, border, refugees	lions, jaguar, tigers, eagles, tiger	blanket, jackets, jacket, wrap, arctic	saudi, terrorists, terrorism, terrorist, allah	muslim, muslims, allah, islamic, islam	muslims, somalia, ethiopia, aboriginal, muslim
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# Typographic attacks – Zero shot

NO LABEL



# Typographic attacks – Zero shot

NO LABEL



LABELED "IPOD"



LABELED "LIBRARY"



# Typographic attacks – Zero shot

NO LABEL



LABELED "IPOD"



LABELED "LIBRARY"




# Typographic attacks – Zero shot

NO LABEL

	Granny Smith	85.61%
	iPod	0.42%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

	laptop computer	15.98%
	iPod	0%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

	coffee mug	61.71%
	iPod	0%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

LABELED "IPOD"

	Granny Smith	0.13%
	iPod	99.68%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

	laptop computer	4.03%
	iPod	78.2%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

	coffee mug	2.97%
	iPod	95.43%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

LABELED "LIBRARY"

	Granny Smith	1.14%
	iPod	0.08%
	library	90.53%
	pizza	0%
	rifle	0%
	toaster	0%

	laptop computer	37.6%
	iPod	0%
	library	5.24%
	pizza	0%
	rifle	0%
	toaster	0%

	coffee mug	2.13%
	iPod	0%
	library	80.77%
	pizza	0%
	rifle	0%
	toaster	0%



# Typographic attacks – Linear probes

NO LABEL

	Granny Smith	26.9%
	iPod	0%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

LABELED "IPOD"

	Granny Smith	16.1%
	iPod	1%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%


LABELED "LIBRARY"

	Granny Smith	2.9%
	iPod	0%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%


	laptop computer	68.1%
	iPod	0%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

	laptop computer	58%
	iPod	0.3%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

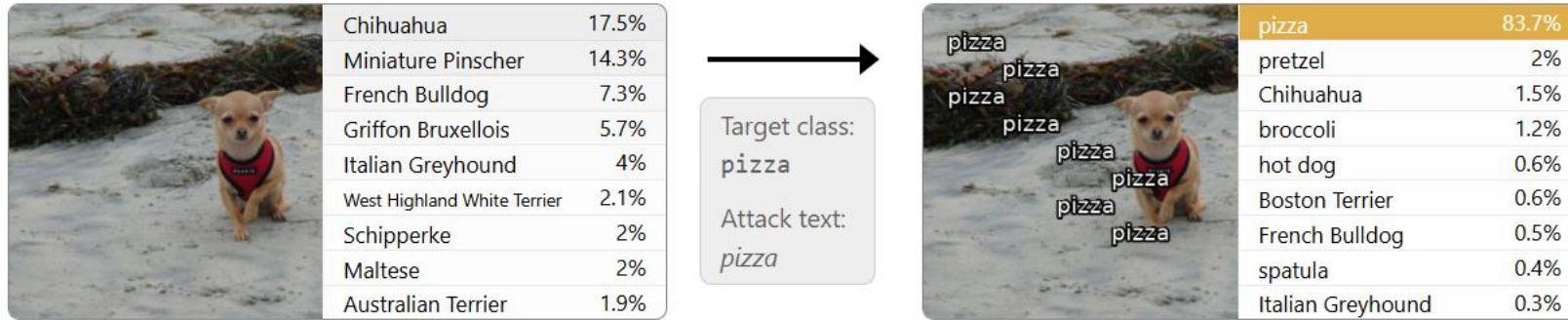
	laptop computer	72.9%
	iPod	0%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

	coffee mug	70.7%
	iPod	0%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

	coffee mug	76%
	iPod	0.1%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

	coffee mug	12.3%
	iPod	0%
	library	0%
	pizza	0%
	rifle	0%
	toaster	0%

# Typographic attacks



Target class	Attack text	Pixel cover	Success Linear probes
waste container	<i>trash</i>	7.59%	95.4%
iPod	<i>iPod</i>	6.8%	94.7%
rifle	<i>rifle</i>	6.41%	91%
pizza	<i>pizza</i>	8.11%	92.3%
radio	<i>radio</i>	7.73%	77%
great white shark	<i>shark</i>	8.33%	62.2%
library	<i>library</i>	9.95%	75.9%
Siamese cat	<i>meow</i>	8.44%	46.5%
piggy bank	<i>\$\$\\$\\$\\$\\$</i>	6.99%	36.4%

# Opinion & Discussion

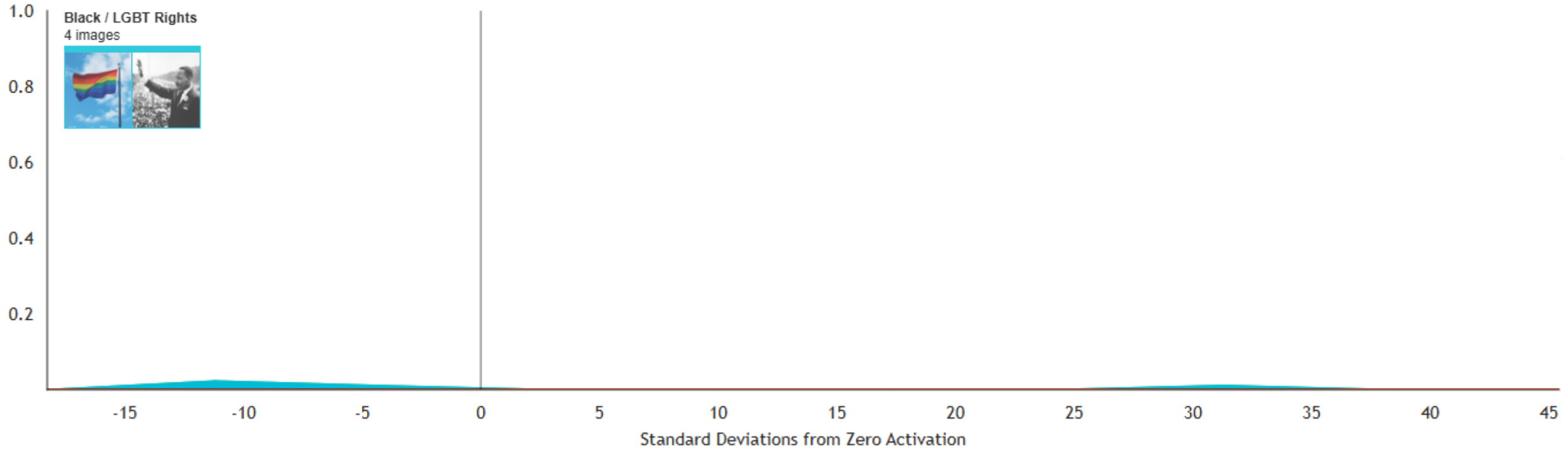
- + Reveals properties in CLIP
- + Faceted Feature Visualization
- + Mentions bias and weaknesses
- Political bias
- Dependent on human interpretation
- Not open source, little explanation

<https://openai.com/index/microscope/>

# References

- Goh, Gabriel, et al. "Multimodal neurons in artificial neural networks." Distill 6.3 (2021): e30
- Quiroga, R. Quian, et al. "Invariant visual representation by single neurons in the human brain." Nature 435.7045 (2005): 1102-1107
- Radford, Alec, et al. "Learning transferable visual models from natural language supervision." International conference on machine learning. PmLR, 2021
- Olah, Chris, et al. "The building blocks of interpretability." Distill 3.3 (2018): e10
- Olah, Chris, Alexander Mordvintsev, and Ludwig Schubert. "Feature visualization." Distill 2.11 (2017): e7
- Qin, Zhuwei, et al. "How convolutional neural network see the world-A survey of convolutional neural network visualization methods." arXiv preprint arXiv:1804.11191 (2018)
- Olah, Chris, et al. "Zoom in: An introduction to circuits." Distill 5.3 (2020): e00024-001.

# Case study – Donald Trump neuron activation



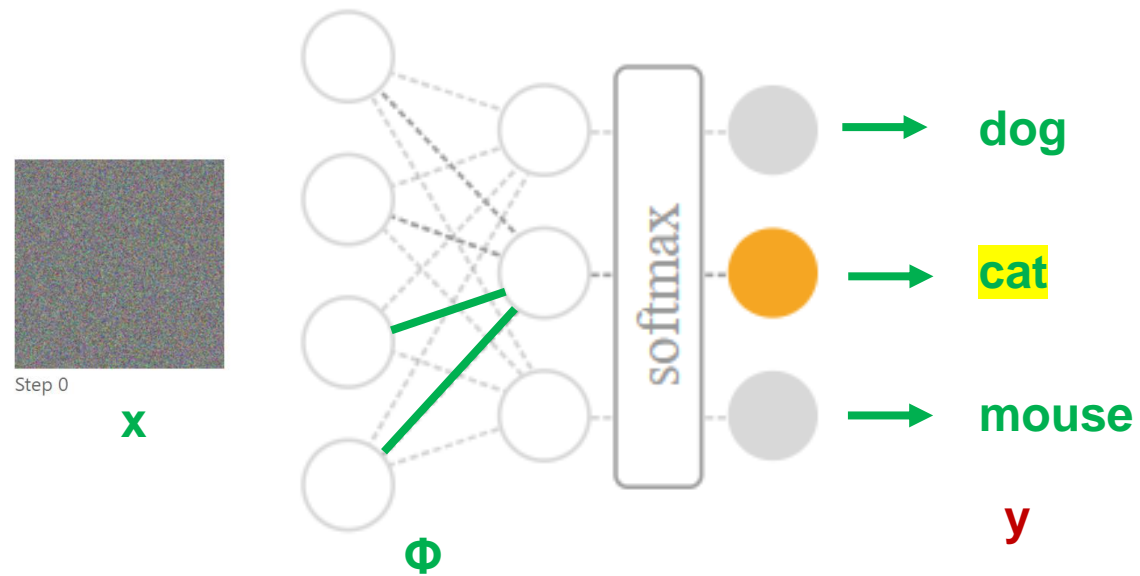
# Feature Visualization



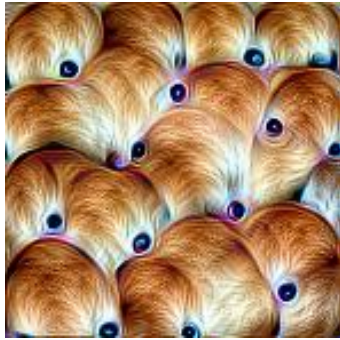
A **car detector** (4c:447) is assembled from earlier units.

# Feature Visualization

- Optimize for label



# Multi-faceted neurons - Diversity



Simple optimization



Optimization with diversity



Dataset examples



# Multi-faceted neurons - Diversity

from style transfer:  $G_{ij} = \sum_{x,y} layer_n[x, y, i] * layer_n[x, y, j]$

$$C_{diversity} = - \sum_a \sum_{b \neq a} \frac{vec(G_a) * vec(G_b)}{\| vec(G_a) \| \| vec(G_b) \|}$$