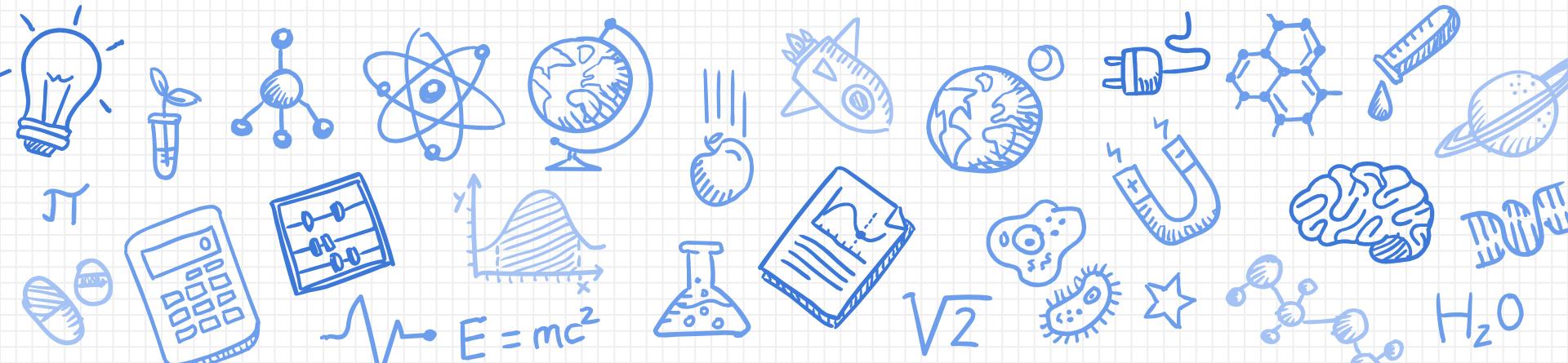


# AI in Computer Vision & Facial Recognition

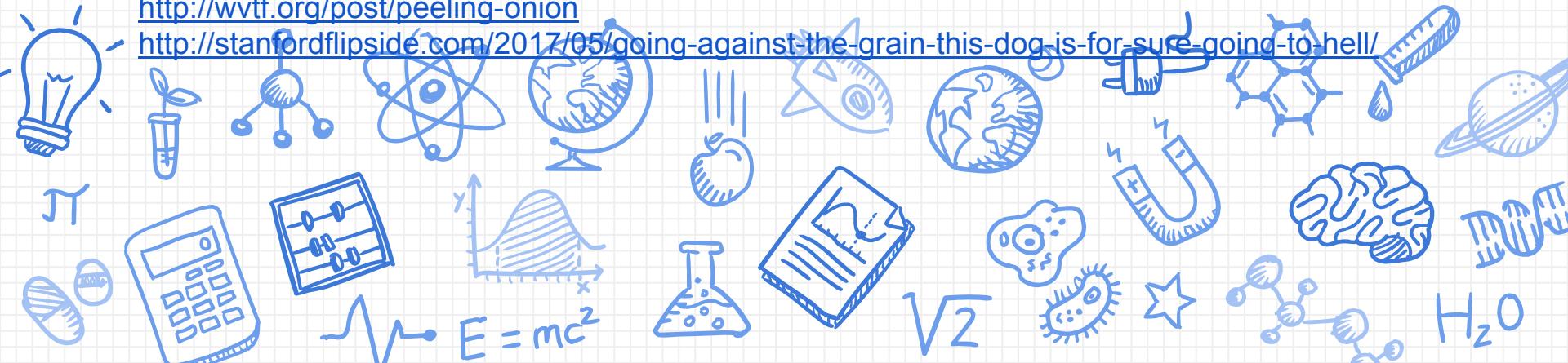
CSE 352  
Prof. Anita Wasilewska

Team 9  
Haoran Yang  
Wangshu Luo  
Xiaohan Chen  
Zijun He



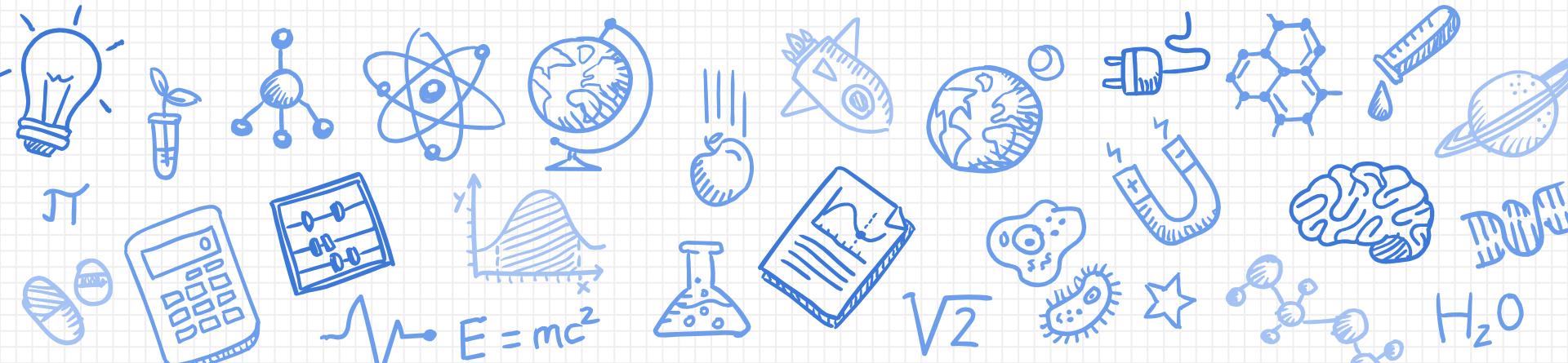
# Sources

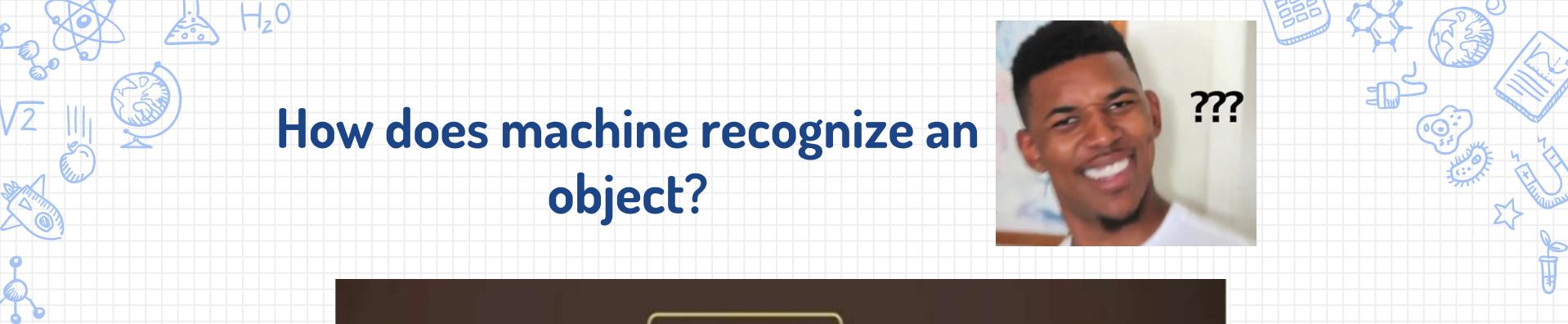
- <https://www.youtube.com/watch?v=40riCqvRoMs&t=936s>
- <https://research.googleblog.com/2016/10/how-robots-can-acquire-new-skills-from.html>
- [http://cs231n.stanford.edu/slides/2017/cs231n\\_2017\\_lecture1.pdf](http://cs231n.stanford.edu/slides/2017/cs231n_2017_lecture1.pdf)
- <https://www.youtube.com/watch?v=FmpDlaiMleA>
- <http://www3.cs.stonybrook.edu/~cse352/L12NN.pdf>
- <https://www.newscientist.com/article/2114900-concerns-as-face-recognition-tech-used-to-identify-criminals/>
- <https://www.oreilly.com/ideas/the-ethics-of-face-recognition>
- <https://www.cnet.com/news/iphone-xs-face-id-supposedly-got-hacked-we-have-questions/>
- <https://www.appannie.com/en/apps/ios/app/b612-selfie-from-the-heart/>
- <https://blog.emojiipedia.org/apples-new-animoji/>
- <http://wvtf.org/post/peeling-onion>
- <http://stanfordflipside.com/2017/05/going-against-the-grain-this-dog-is-for-sure-going-to-hell/>



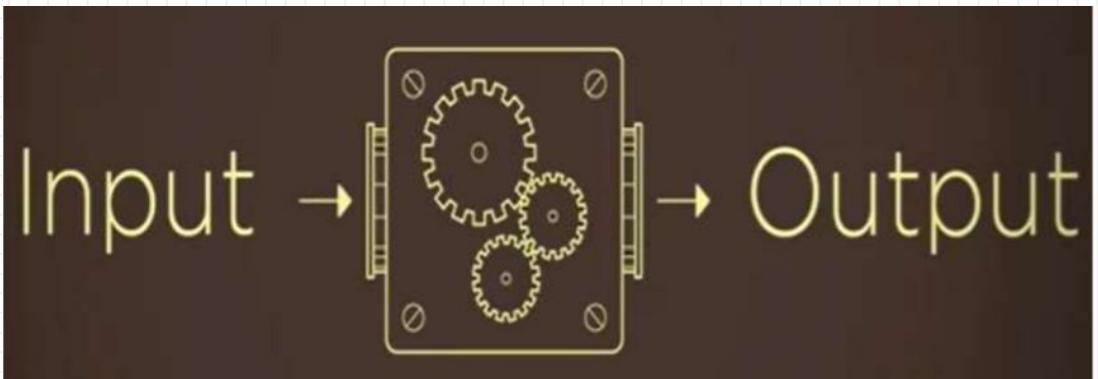
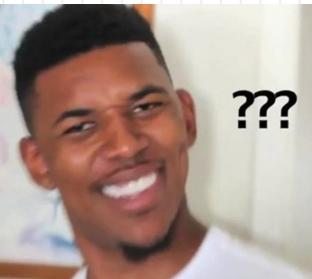
# Overview

- Object Recognition
- Algorithm in Facial Recognition
- Applications in Facial Recognition
- Concerns in Facial Recognition

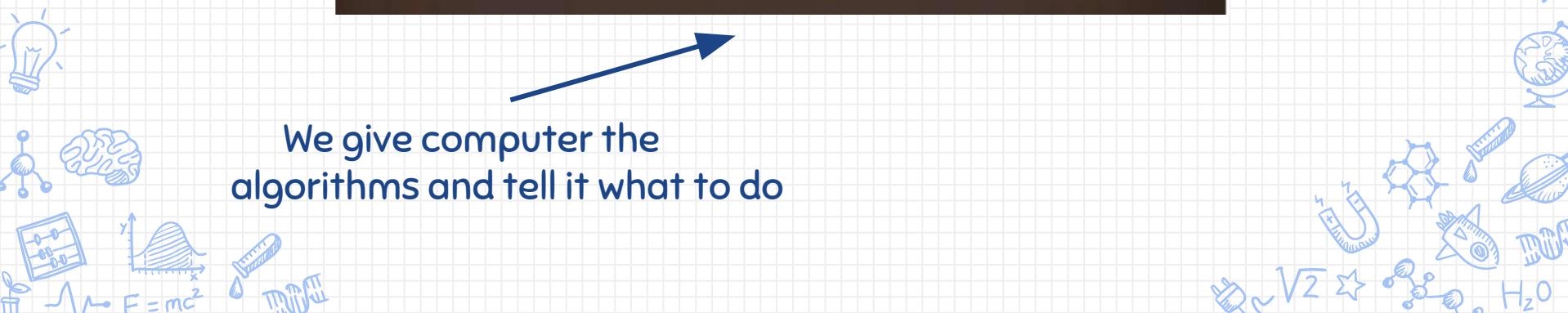


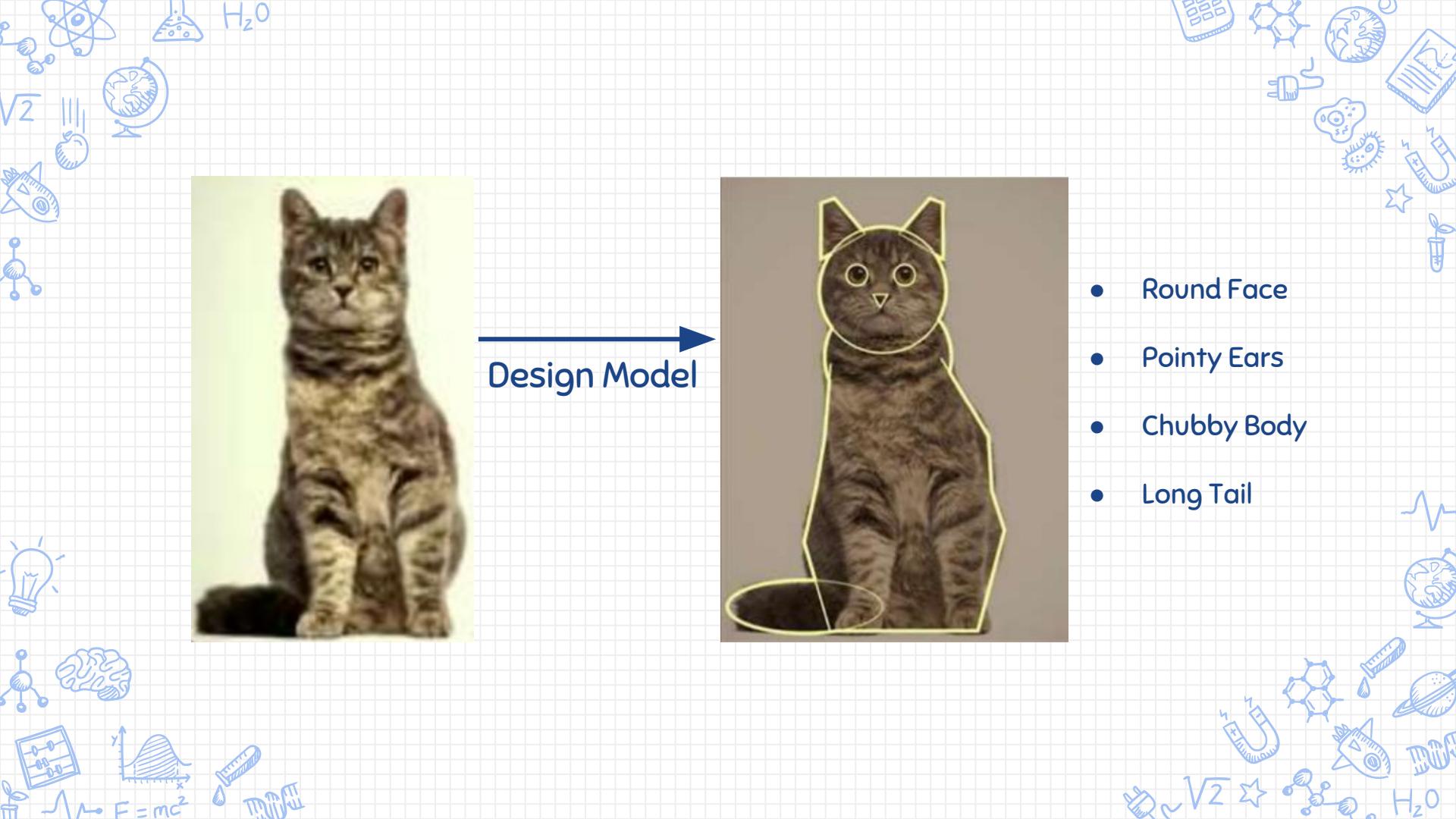


# How does machine recognize an object?



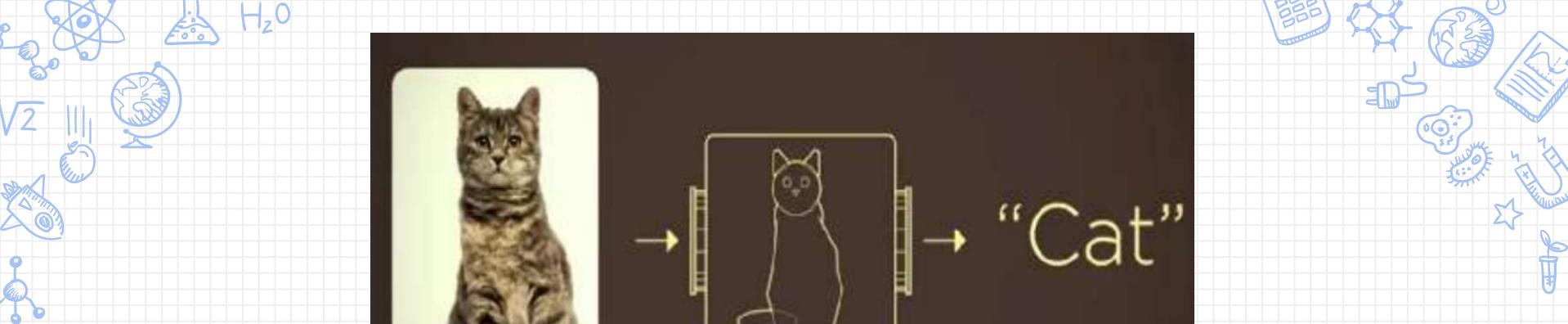
We give computer the  
algorithms and tell it what to do



$H_2O$ 

- Round Face
- Pointy Ears
- Chubby Body
- Long Tail

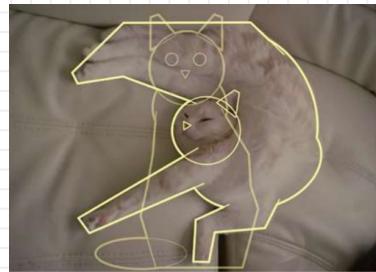
Design Model



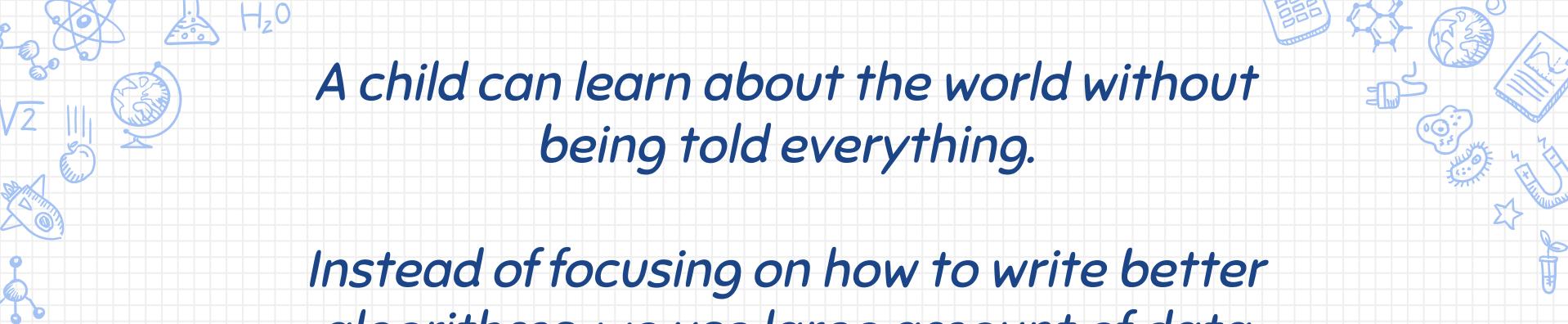
# What about these cats?



?

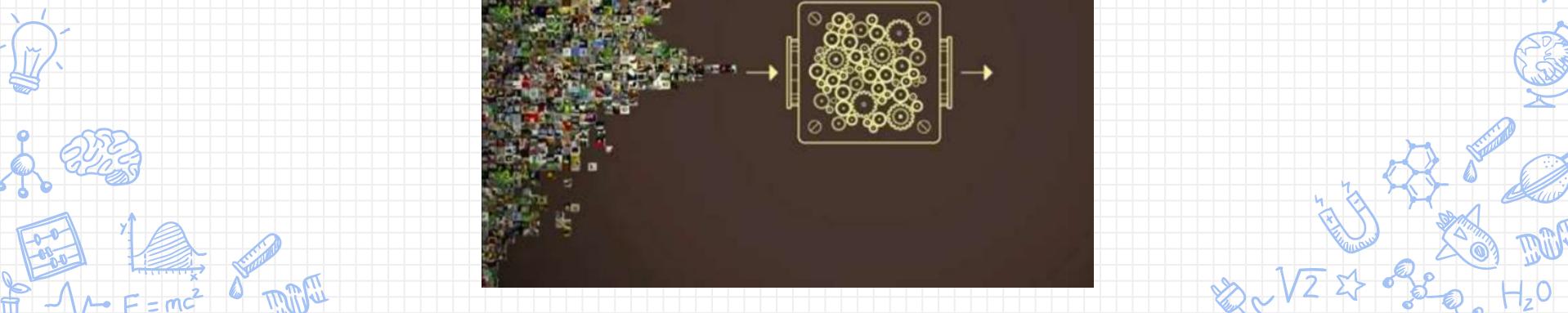
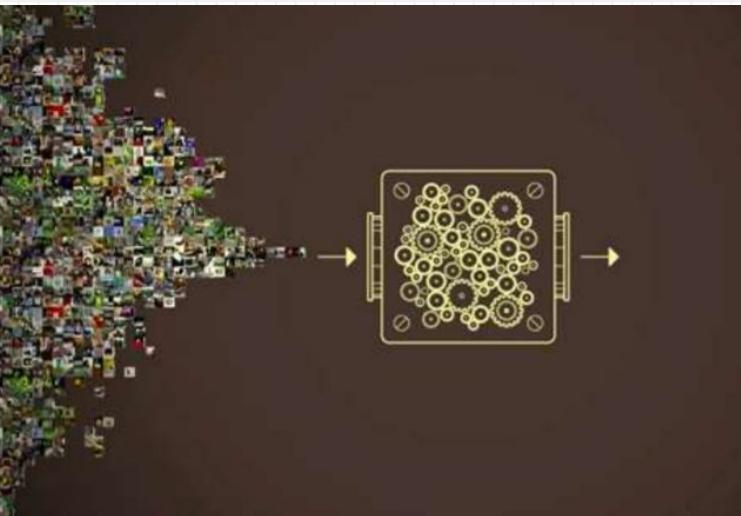


?



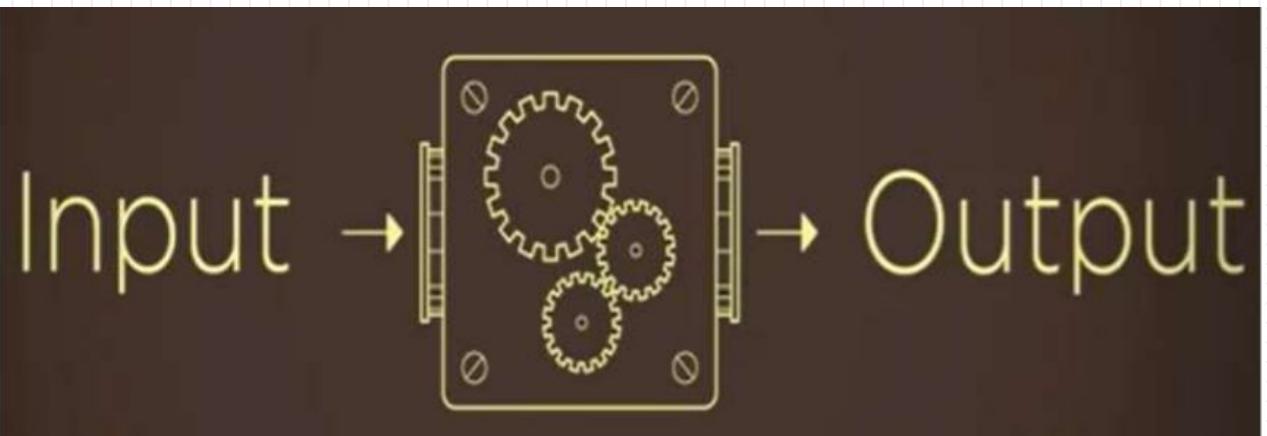
*A child can learn about the world without being told everything.*

*Instead of focusing on how to write better algorithms, we use large amount of data to train the computer.*



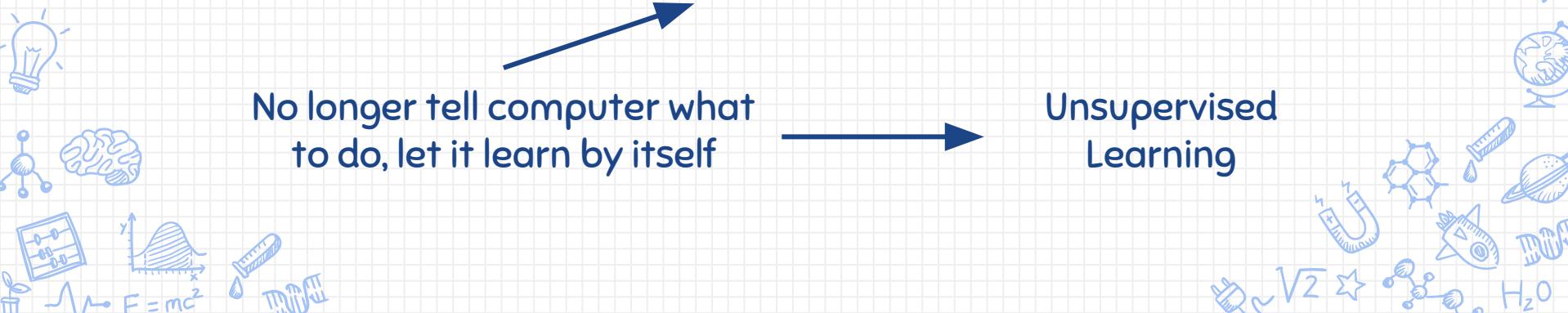
$H_2O$ 

# Deep Learning



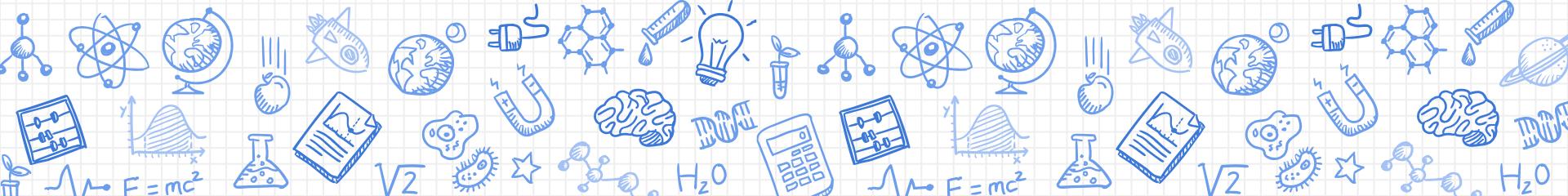
No longer tell computer what  
to do, let it learn by itself

Unsupervised  
Learning

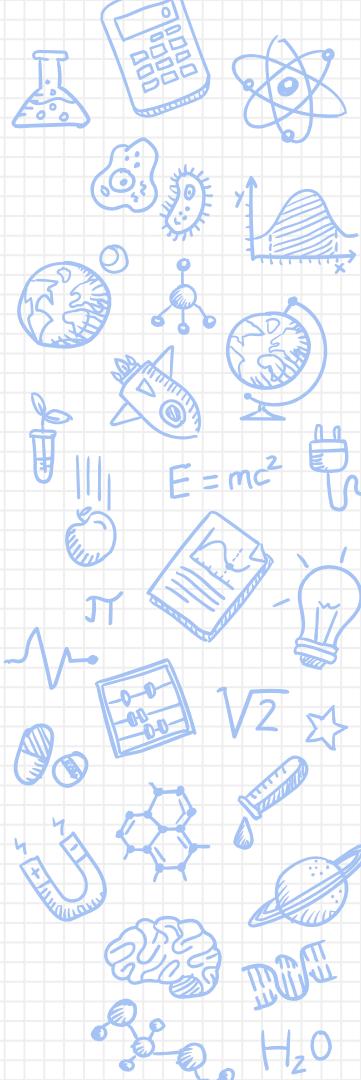


# A Quick Game!!!

Let's learn some Chinese



# Now u r an algorithm-free machine



What does 踢 mean in Chinese?

- A. Eat
- B. Fart
- C. Kick
- D. Ball

What does 跖 mean in Chinese?

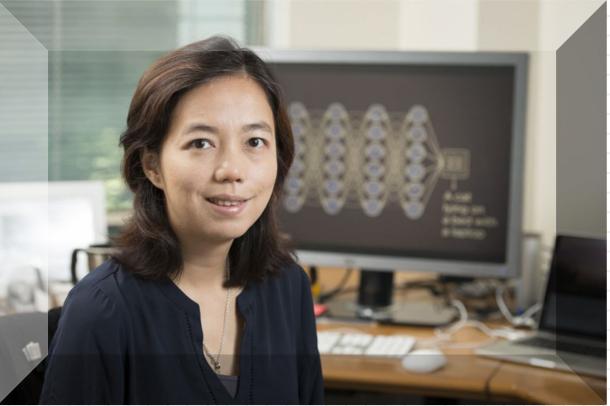
- A. Step On
- B. Knee Down
- C. Push up
- D. Cook

What does 蹤 mean in Chinese?

- A. Sneak
- B. Trample
- C. Pour
- D. Pick

What does 蹤 mean in Chinese?

- A. Wash
- B. Whisper
- C. Jump
- D. Cry



# IMAGENET

14,197,122 images, 21841 synsets indexed

[Explore](#) [Download](#) [Challenges](#) [Publications](#) [CoolStuff](#) [About](#)

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**ImageNet** is an image database organized according to the [WordNet](#) hierarchy (currently only the nouns), in which each node of the hierarchy is depicted by hundreds and thousands of images. Currently we have an average of over five hundred images per node. We hope ImageNet will become a useful resource for researchers, educators, students and all of you who share our passion for pictures.

[Click here](#) to learn more about ImageNet, [Click here](#) to join the ImageNet mailing list.



## Statistics of high level categories

High level category	# synset (subcategories)	Avg # images per synset	Total # images
amphibian	94	591	56K
animal	3822	732	2799K
appliance	51	1164	59K
bird	856	949	812K
covering	946	819	774K
device	2385	675	1610K
fabric	262	690	181K
fish	566	494	280K
flower	462	735	339K
food	1495	670	1001K
fruit	309	607	188K
fungus	303	453	137K
furniture	187	1043	195K
geological formation	151	838	127K
invertebrate	728	573	417K
mammal	1138	821	934K
musical instrument	157	891	140K
plant	1666	600	999K

Since 2010, the annual ImageNet Large Scale Visual Recognition Challenge (ILSVRC) is a competition where research teams evaluate their algorithms on the given data set, and compete to achieve higher accuracy on several visual recognition tasks



**48940 Workers  
167 Countries**

# The Image Classification Challenge:

1,000 object classes

1,431,167 images



**Output:**  
Scale  
T-shirt  
Steel drum  
Drumstick  
Mud turtle



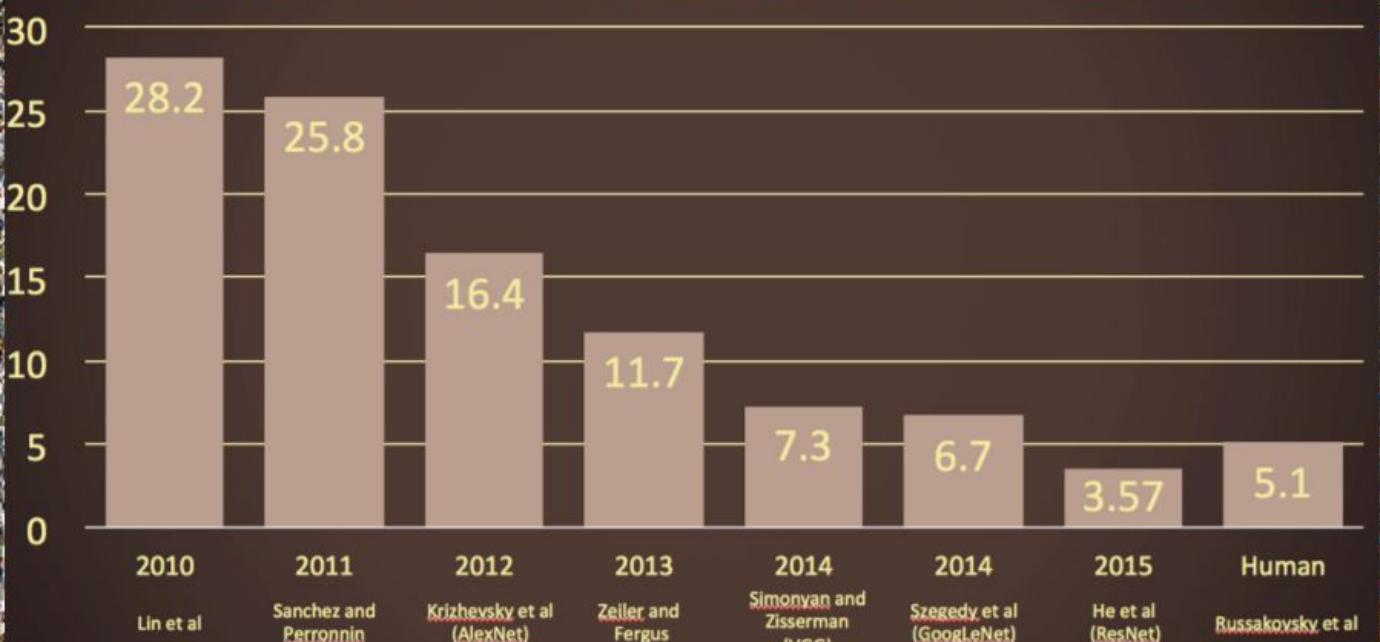
**Output:**  
Scale  
T-shirt  
Giant panda  
Drumstick  
Mud turtle



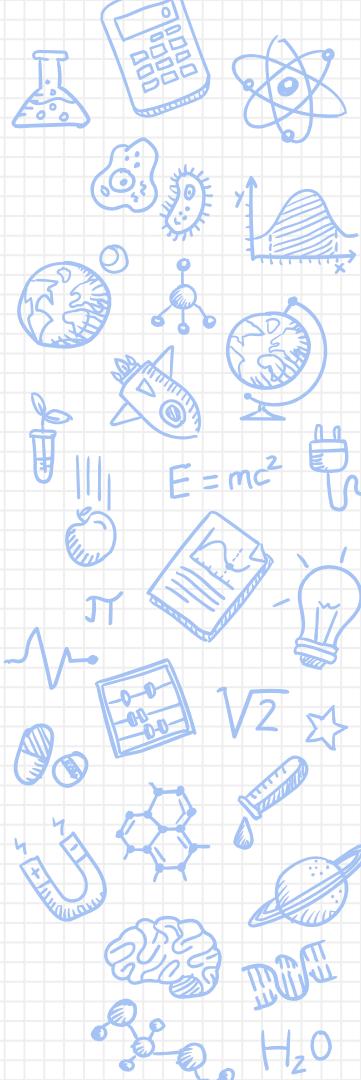
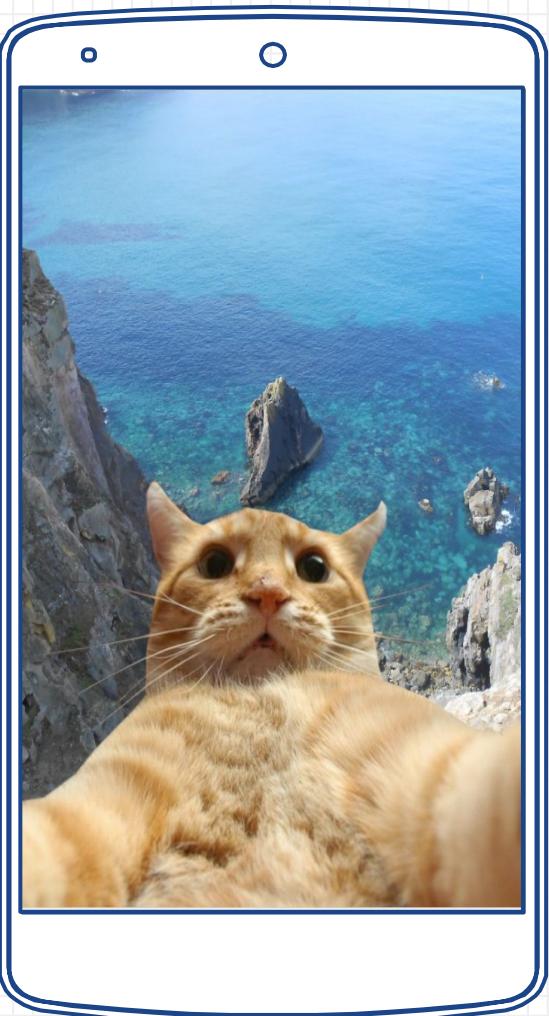
# The Image Classification Challenge:

1,000 object classes

1,431,167 images



# Facial Recognition



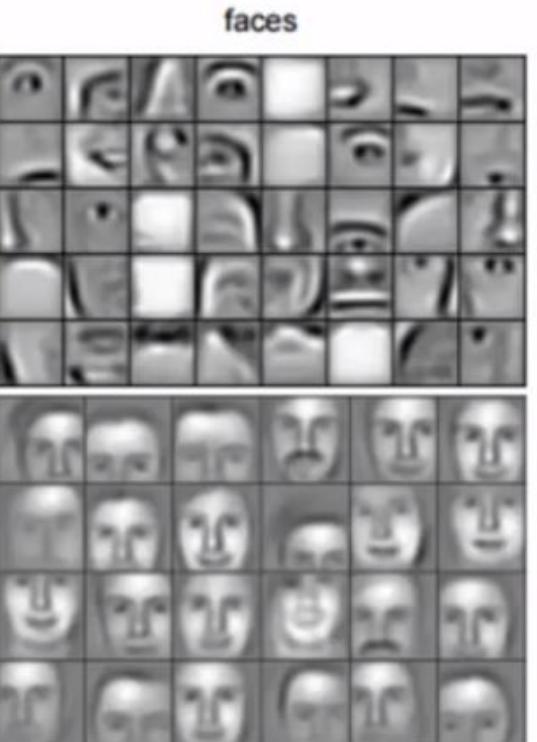
# Algorithm in Facial Recognition

## Convolutional Neural Network (CNN)

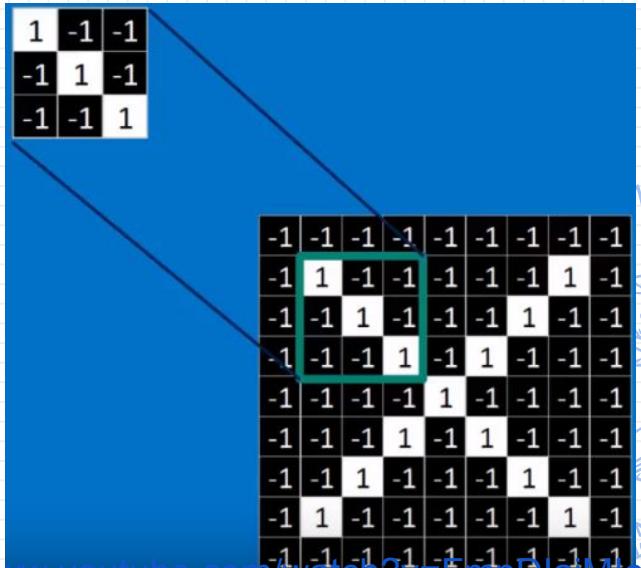
- Convolution
- Normalization
- Pooling

# Convolution

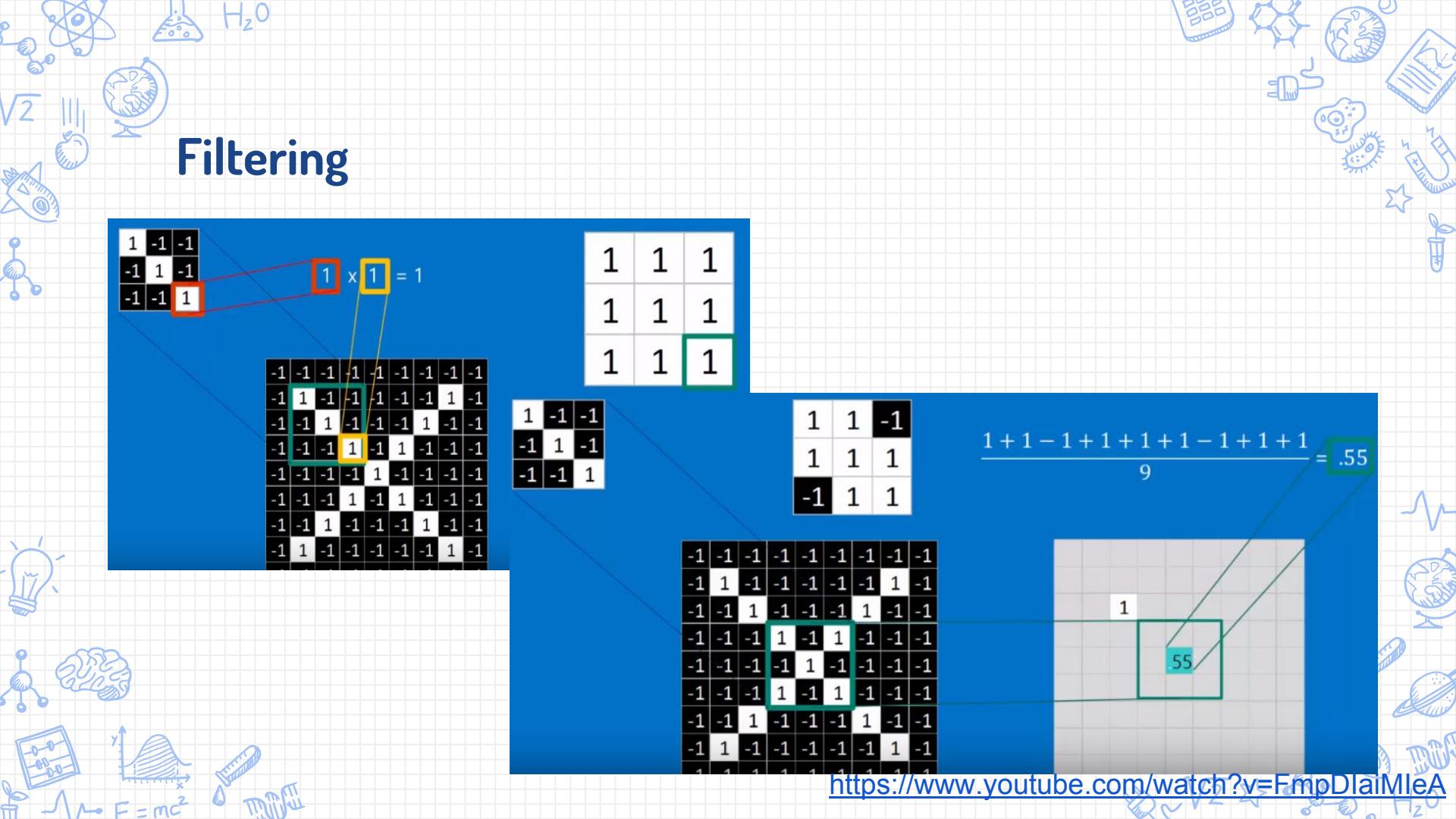
## Features

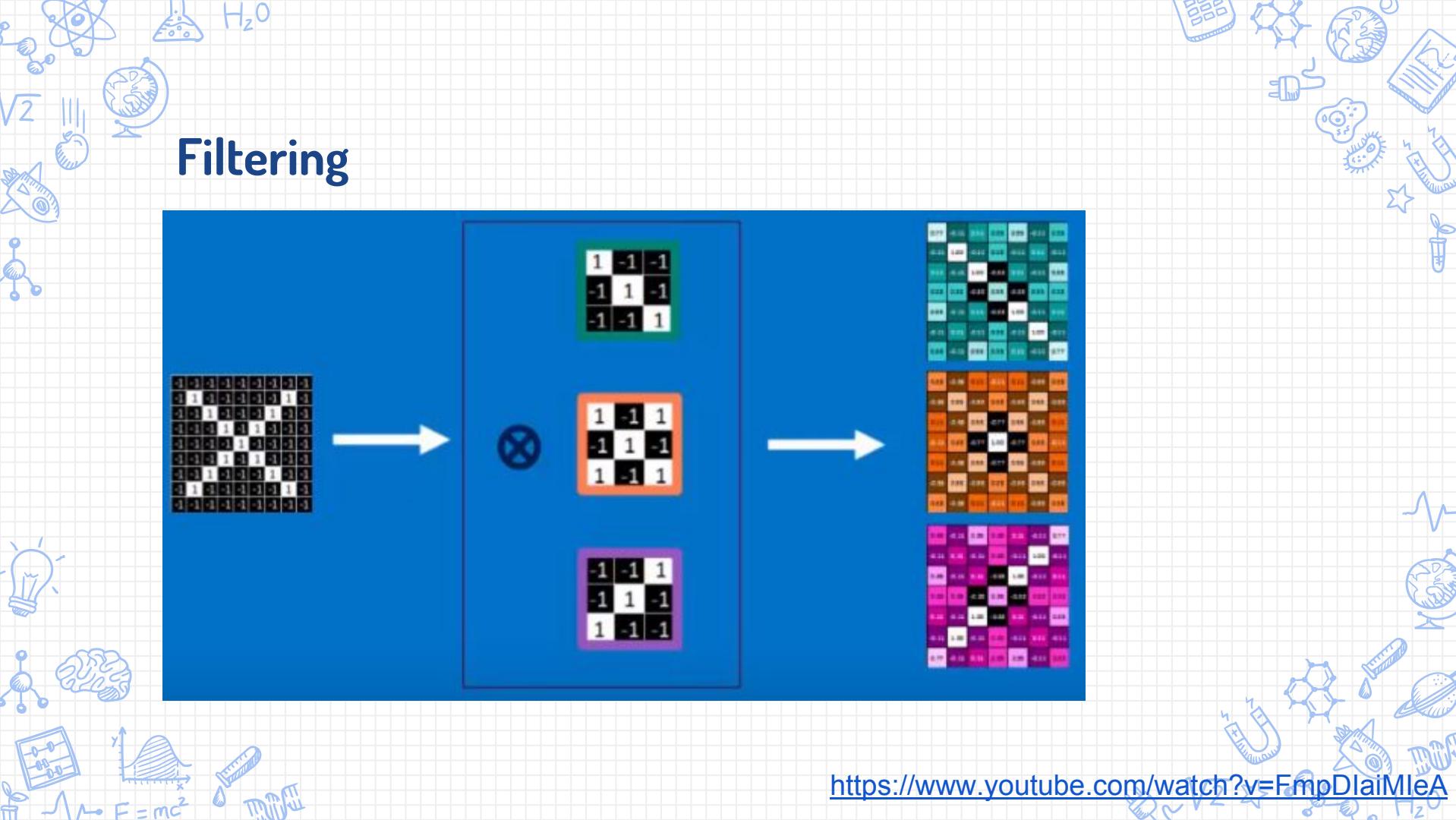


## Filtering



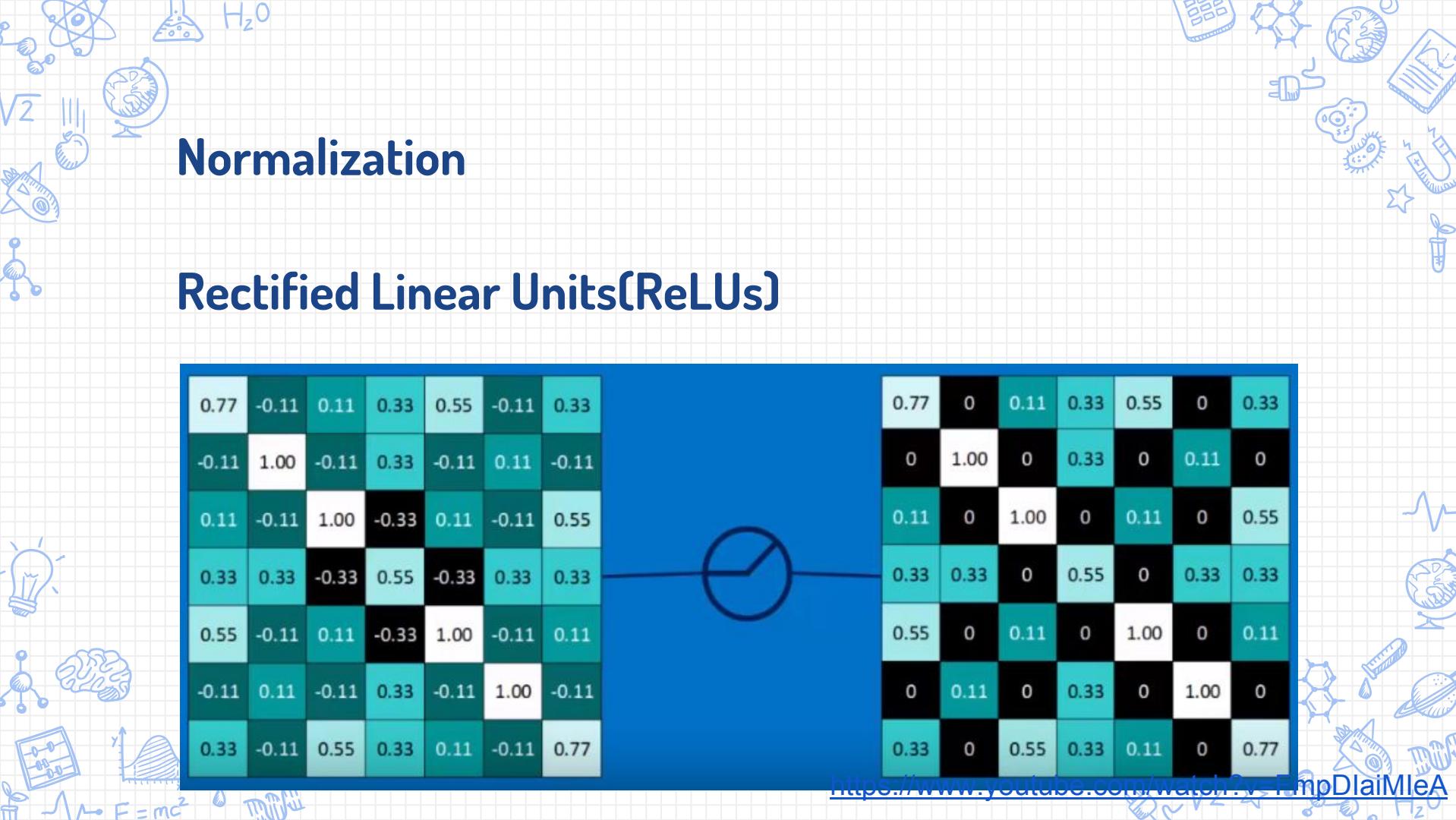
<https://www.youtube.com/watch?v=EmpDIAmIeA>





# Filtering

<https://www.youtube.com/watch?v=EmpDlaIMleA>



# Normalization

## Rectified Linear Units(ReLUs)

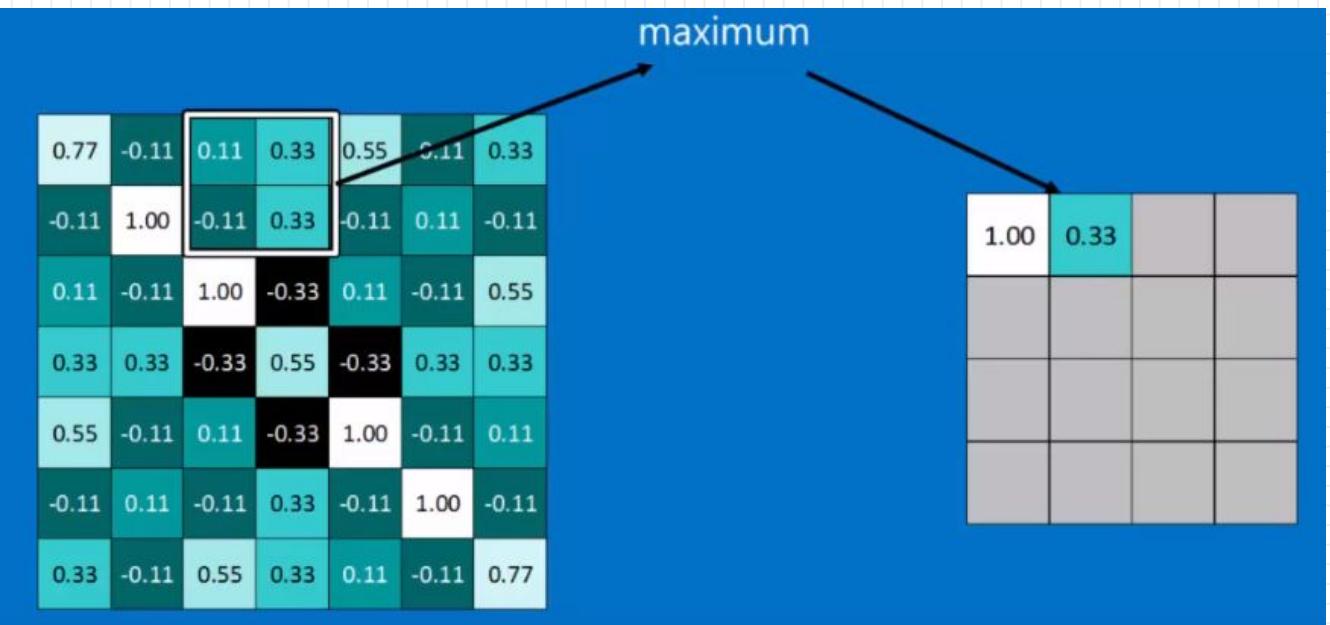
0.77	-0.11	0.11	0.33	0.55	-0.11	0.33
-0.11	1.00	-0.11	0.33	-0.11	0.11	-0.11
0.11	-0.11	1.00	-0.33	0.11	-0.11	0.55
0.33	0.33	-0.33	0.55	-0.33	0.33	0.33
0.55	-0.11	0.11	-0.33	1.00	-0.11	0.11
-0.11	0.11	-0.11	0.33	-0.11	1.00	-0.11
0.33	-0.11	0.55	0.33	0.11	-0.11	0.77



0.77	0	0.11	0.33	0.55	0	0.33
0	1.00	0	0.33	0	0.11	0
0.11	0	1.00	0	0.11	0	0.55
0.33	0.33	0	0.55	0	0.33	0.33
0.55	0	0.11	0	1.00	0	0.11
0	0.11	0	0.33	0	1.00	0
0.33	0	0.55	0.33	0.11	0	0.77

<https://www.youtube.com/watch?v=FmpDlaIMleA>

# Pooling



<https://www.youtube.com/watch?v=EmpDlaIMeA>

# Pooling

0.77	-0.11	0.11	0.33	0.55	-0.11	0.33
-0.11	1.00	-0.11	0.33	-0.11	0.11	-0.11
0.11	-0.11	1.00	-0.33	0.11	-0.11	0.55
0.33	0.33	-0.33	0.55	-0.33	0.33	0.33
0.55	-0.11	0.11	-0.33	1.00	-0.11	0.11
-0.11	0.11	-0.11	0.33	-0.11	1.00	-0.11
0.33	-0.11	0.55	0.33	0.11	-0.11	0.77

max pooling

1.00	0.33	0.55	0.33
0.33	1.00	0.33	0.55
0.55	0.33	1.00	0.11
0.33	0.55	0.11	0.77

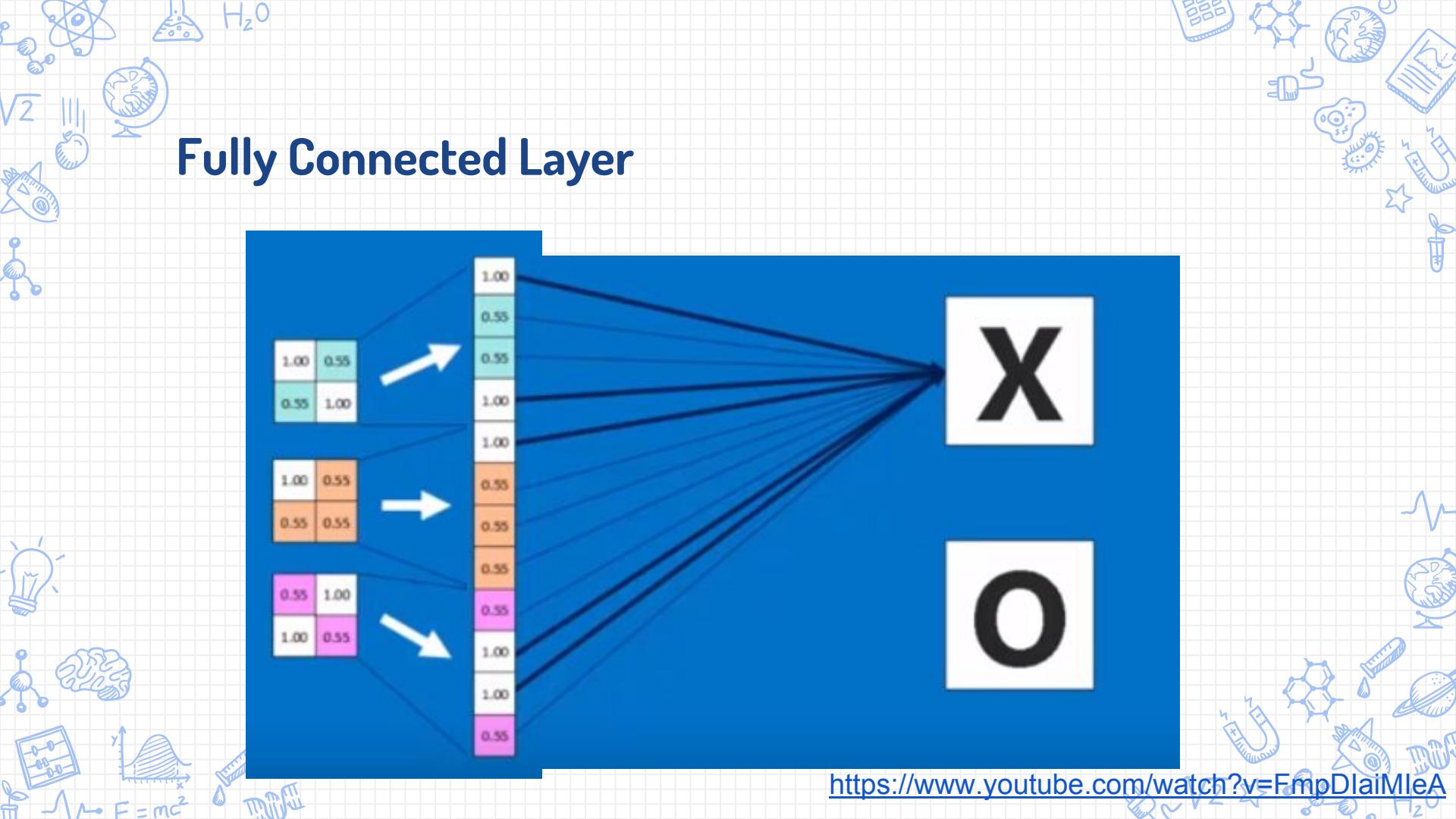
# Stacking

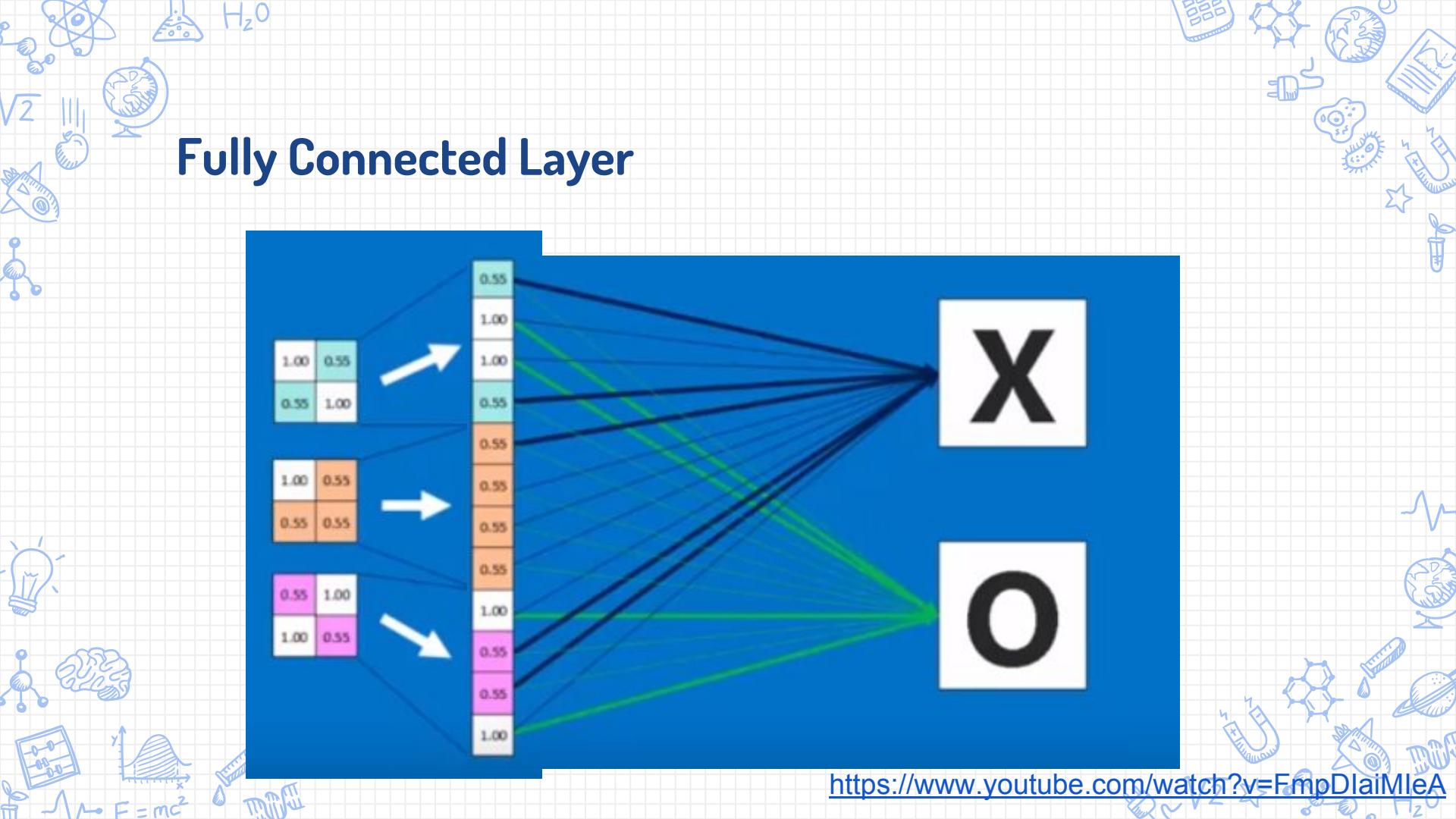


# Fully Connected Layer

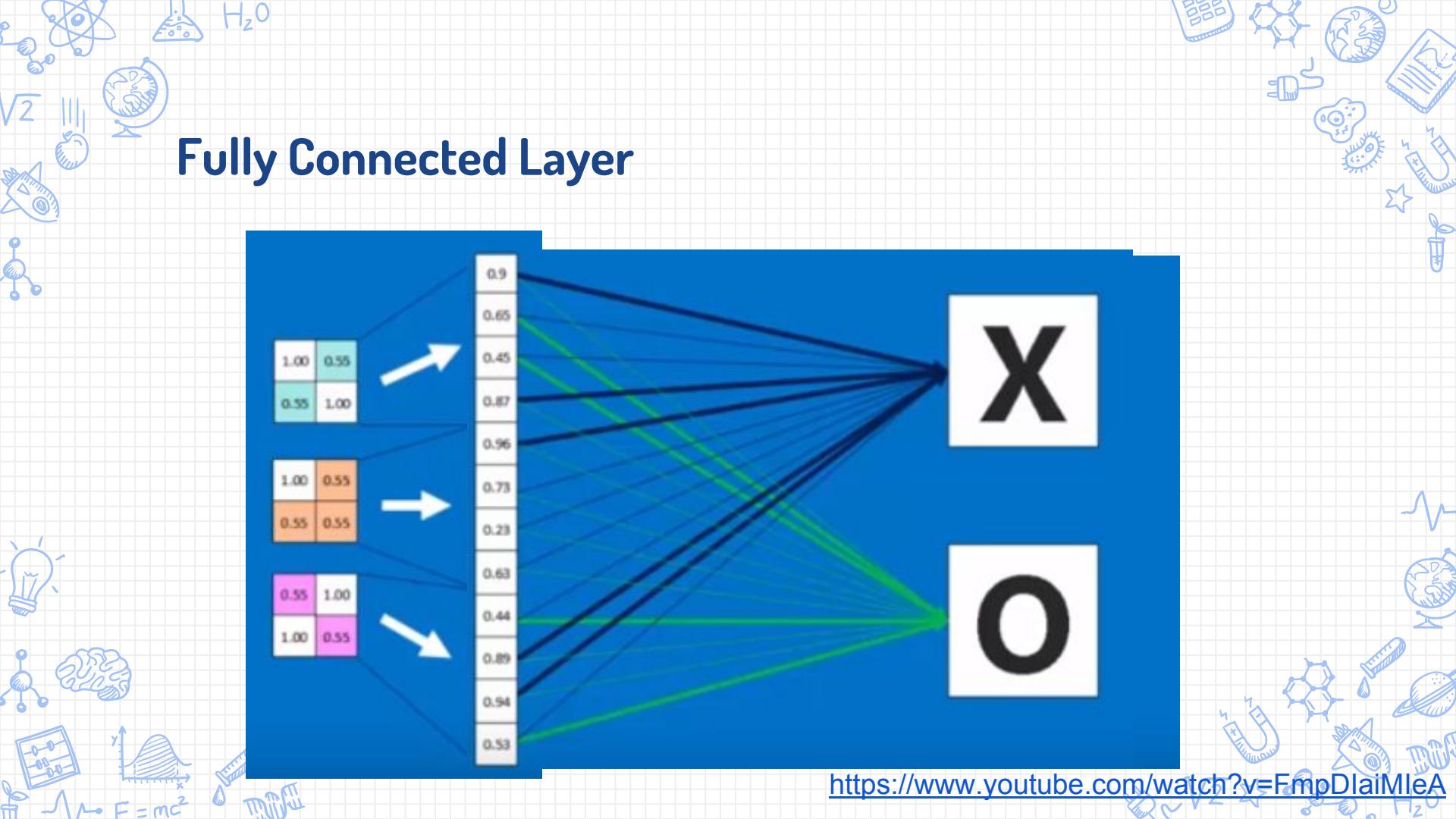


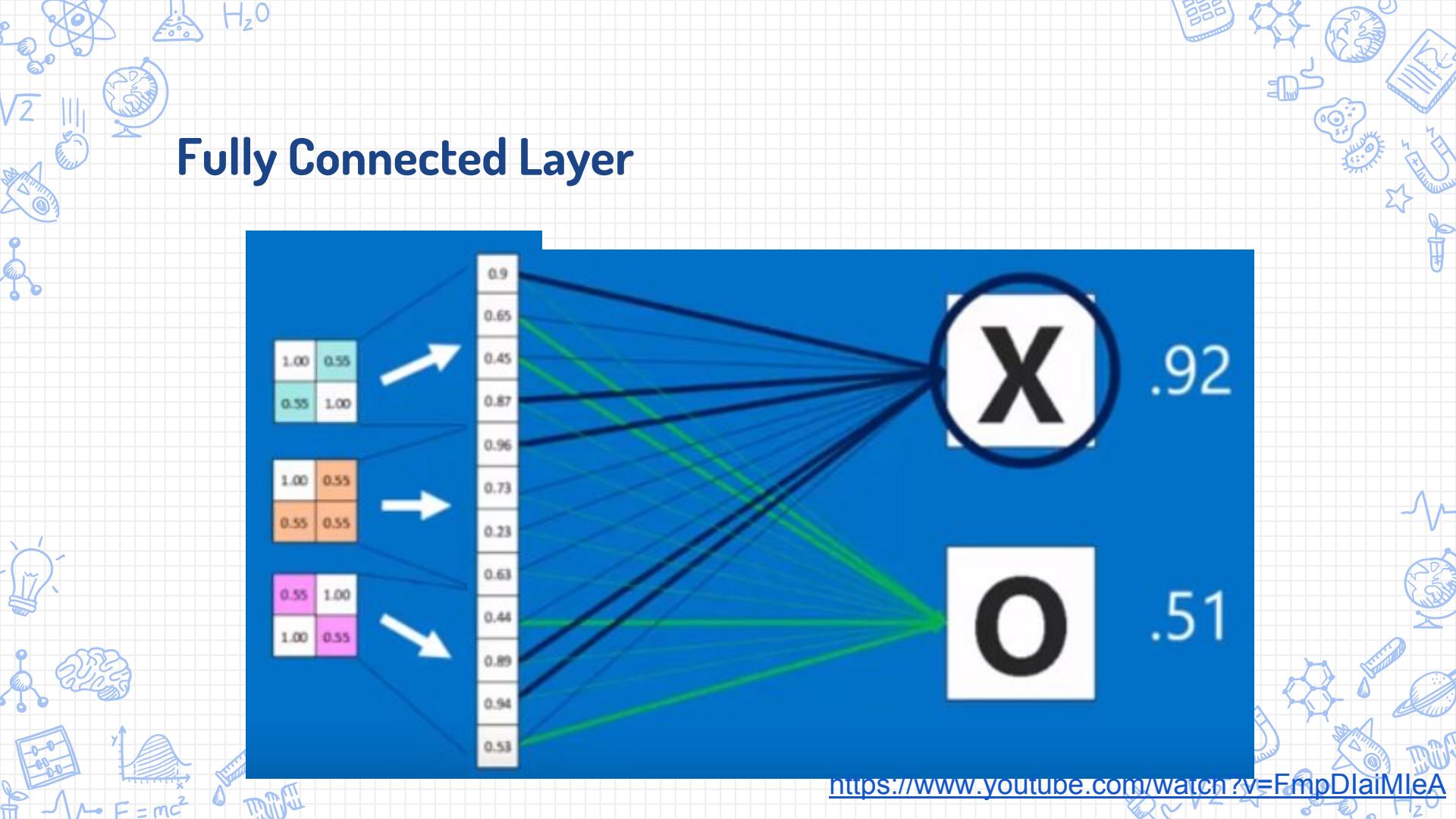
<https://www.youtube.com/watch?v=EmpDlaIMeA>



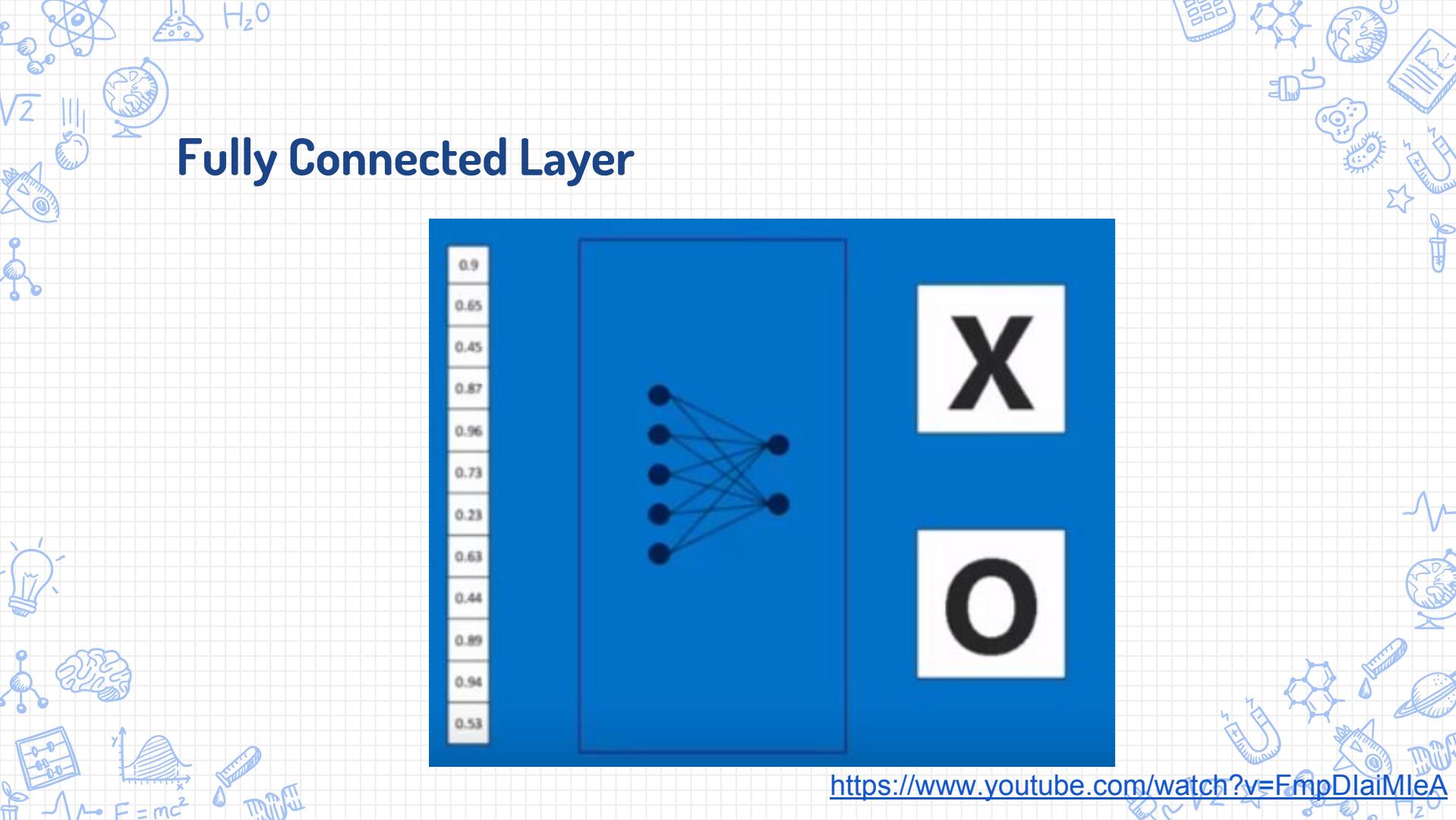


# Fully Connected Layer



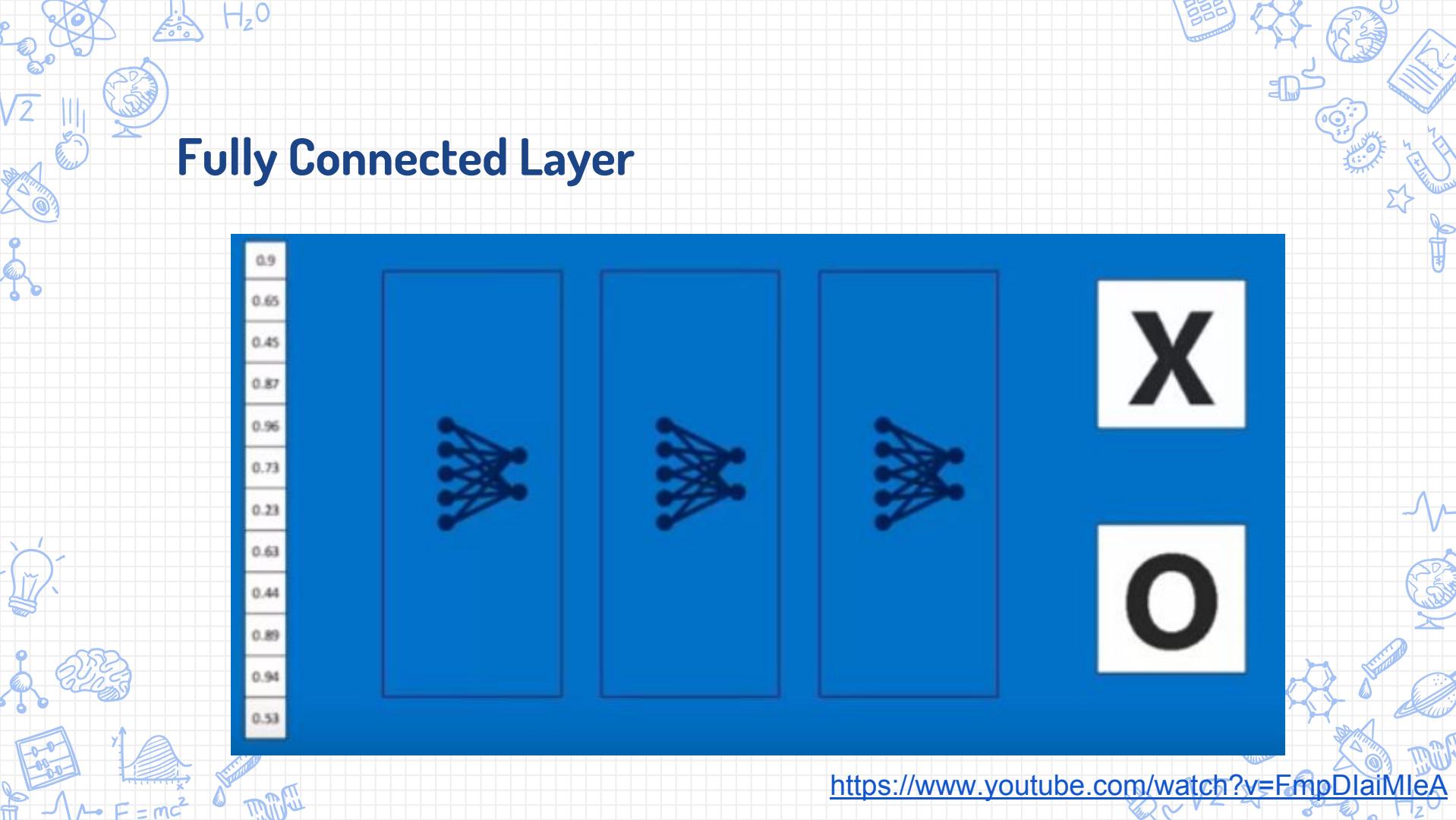


<https://www.youtube.com/watch?v=EmpDlaIMeA>



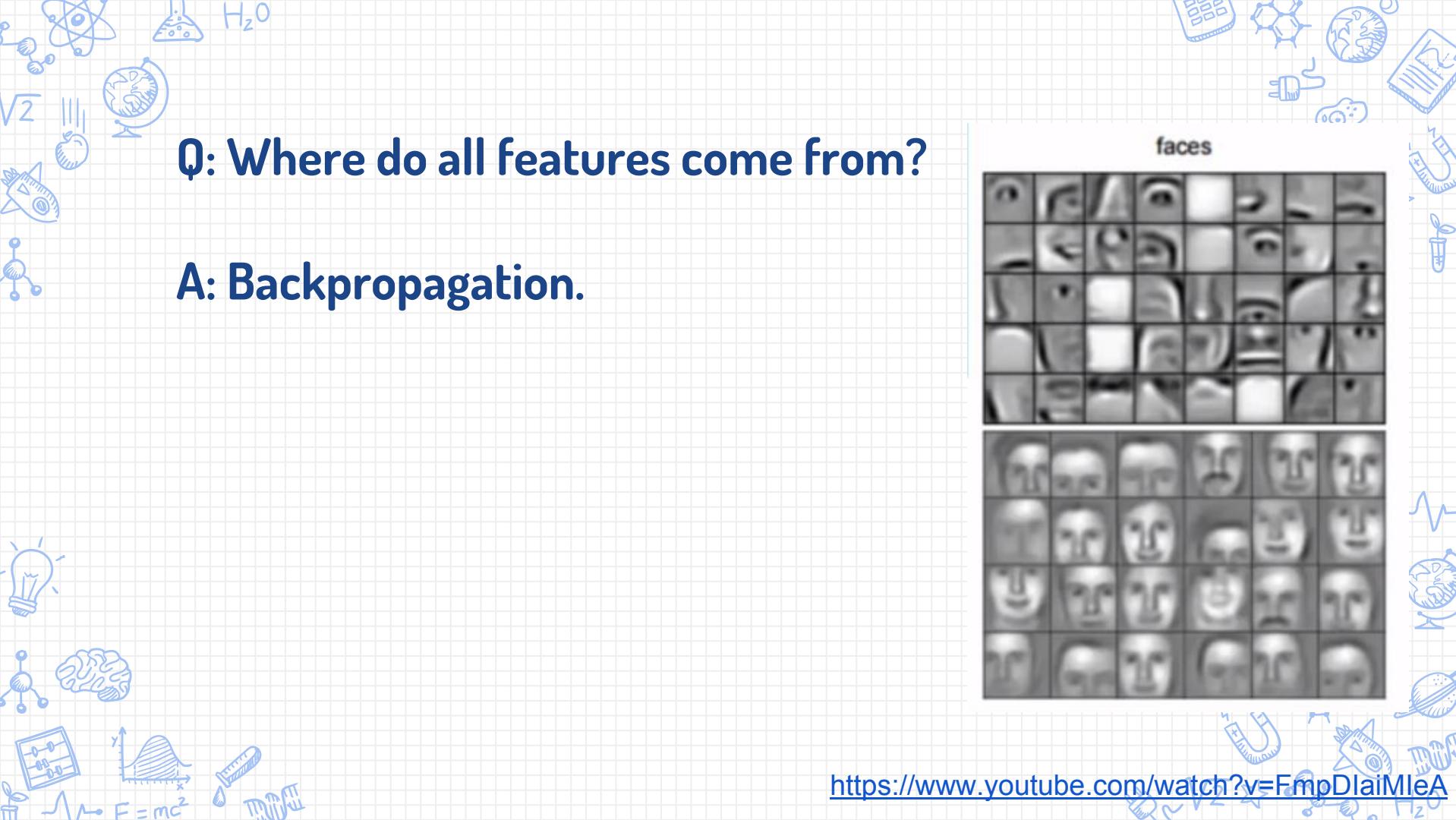
# Fully Connected Layer

<https://www.youtube.com/watch?v=EmpDlaIMeA>



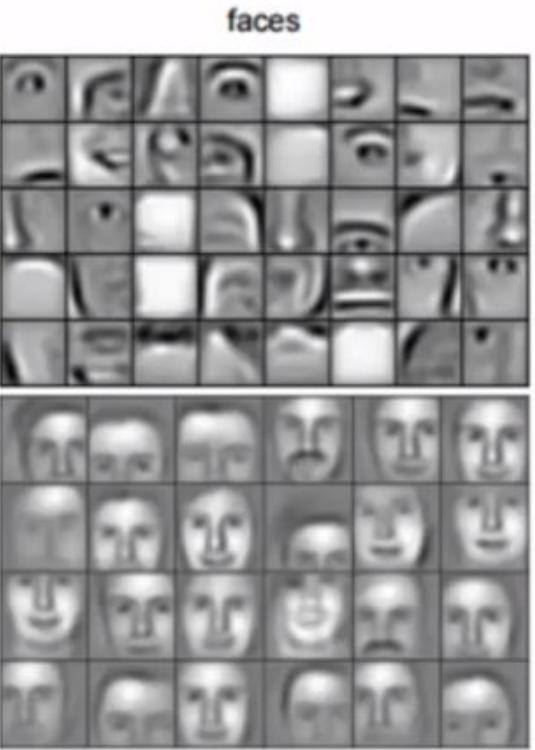
# Fully Connected Layer

<https://www.youtube.com/watch?v=EmpDlaIMeA>



**Q: Where do all features come from?**

**A: Backpropagation.**



<https://www.youtube.com/watch?v=EmpDlaIMeA>

# Backpropagation

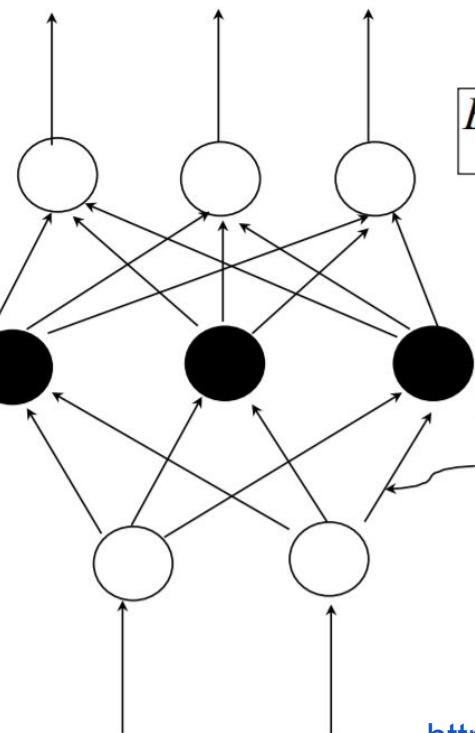
**Output vector**

**Output nodes**

**Hidden nodes**

**Input nodes**

**Input vector:  $x_i$**



$$Err_j = O_j(1 - O_j) \sum_k Err_k w_{jk}$$

$$\theta_j = \theta_j + (l) Err_j$$

$$w_{ij} = w_{ij} + (l) Err_j O_i$$

$$Err_j = O_j(1 - O_j)(T_j - O_j)$$

$$w_{ij}$$

$$O_j = \frac{1}{1 + e^{-I_j}}$$

$$I_j = \sum_i w_{ij} O_i + \theta_j$$

# Application

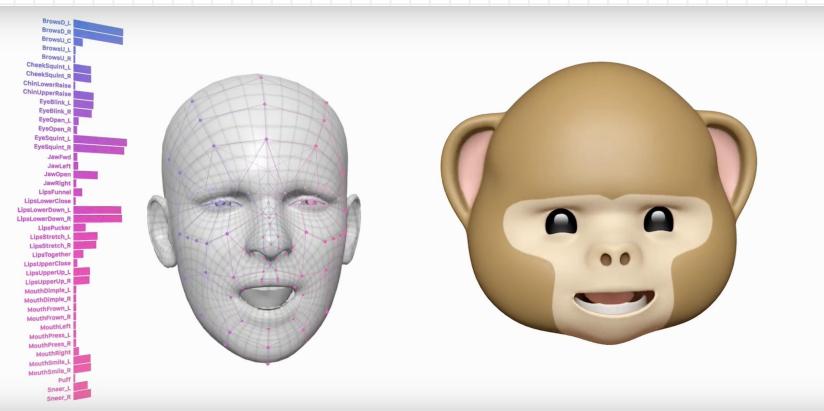
- B612 APP: express yourself daily through fun stickers and amazing AR filters.



# Application

- Animoji

New feature in iOS 11. It is an animated versions of emoji. It uses facial recognition to create 3D versions based on facial expression.



<https://blog.emojipedia.org/apples-new-animoji/>

# Application

- Taking Attendance



<http://wvtf.org/post/peeling-onion>

# Application

- Finding Rover



<http://stanfordflipside.com/2017/05/going-against-the-grain-this-dog-is-for-sure-going-to-hell/>

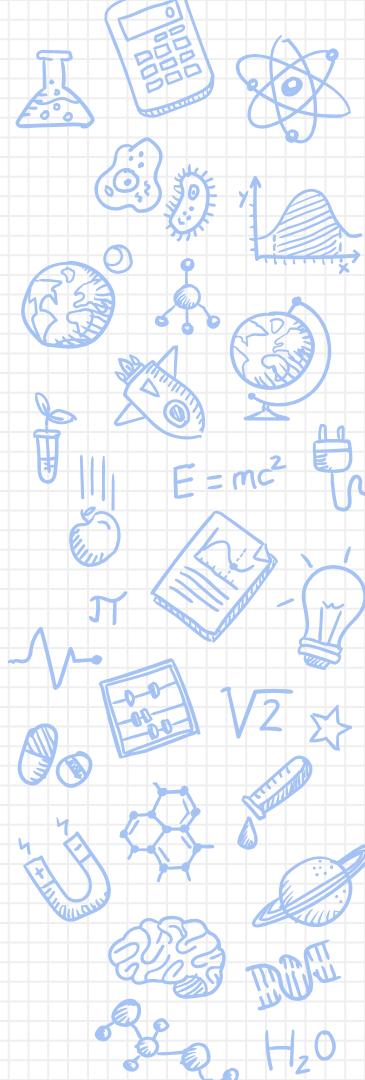
# Some Concerns About Facial Recognition

# Concerns

---

What can your face say about you?

- ✗ Age
- ✗ Gender
- ✗ Mood

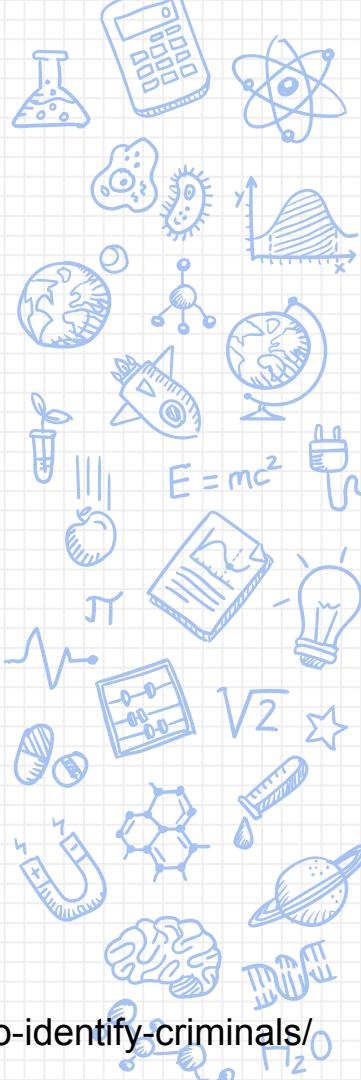


# Concerns

---

**Face recognition tech used to ‘identify’ criminals**

- ✗ Accuracy of up to 90%
- ✗ Different sources

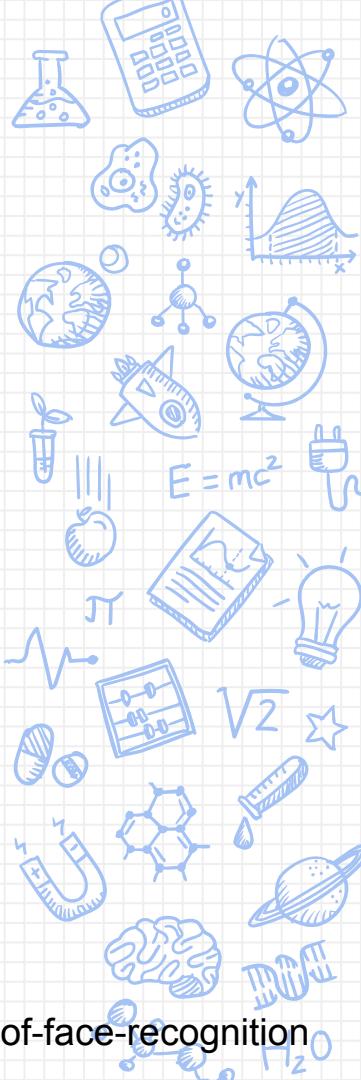


# Concerns

---

An error-prone system would be worse than useless

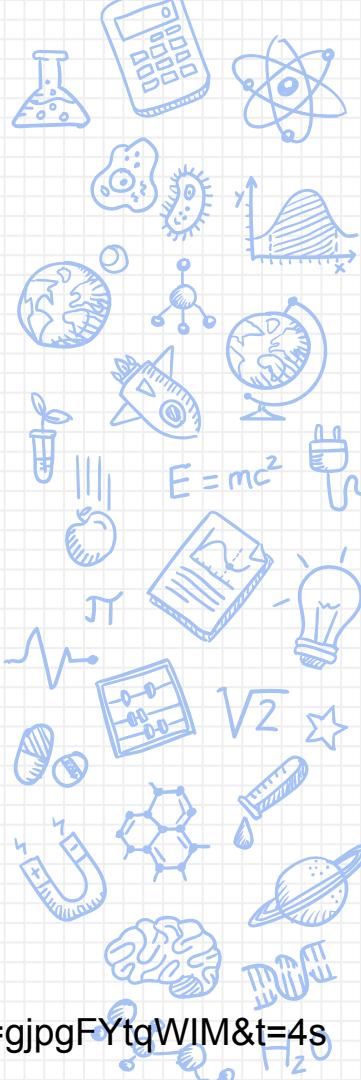
- ✗ Hard to build a good training data
- ✗ Very hard to decreasing the false positives



# Concerns



<https://www.youtube.com/watch?v=gjgFYTqWIM&t=4s>



# THANKS!

## Any questions?

