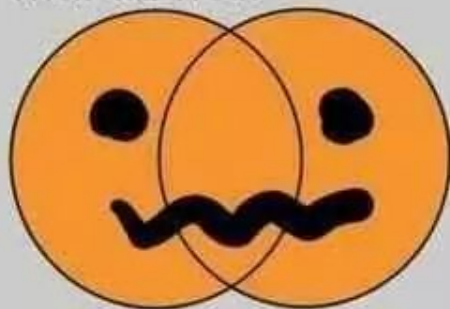




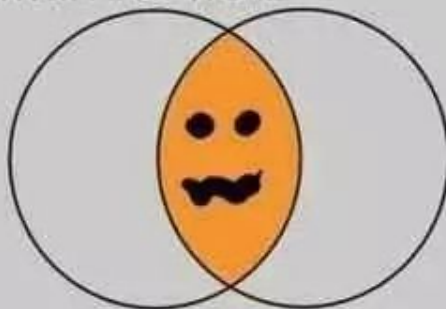
# Welcome!

GM **#8** | HHS CS

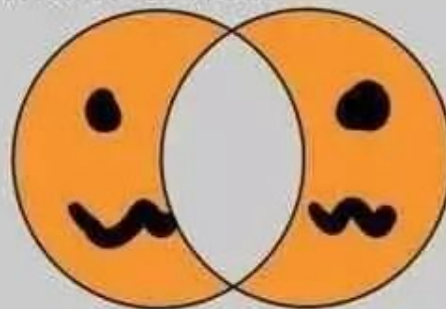
Trick OR Treat



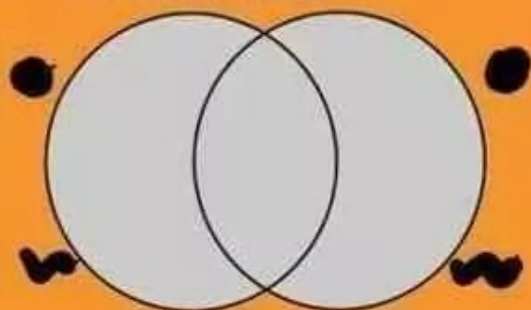
Trick AND Treat



Trick XOR Treat



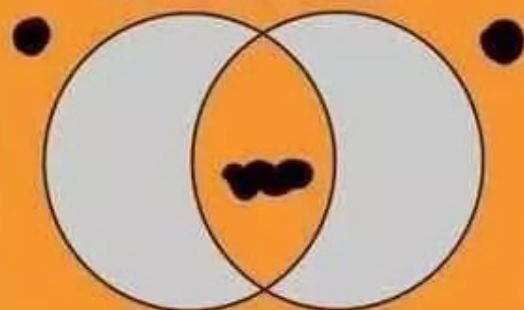
Trick NOR Treat



Trick NAND Treat

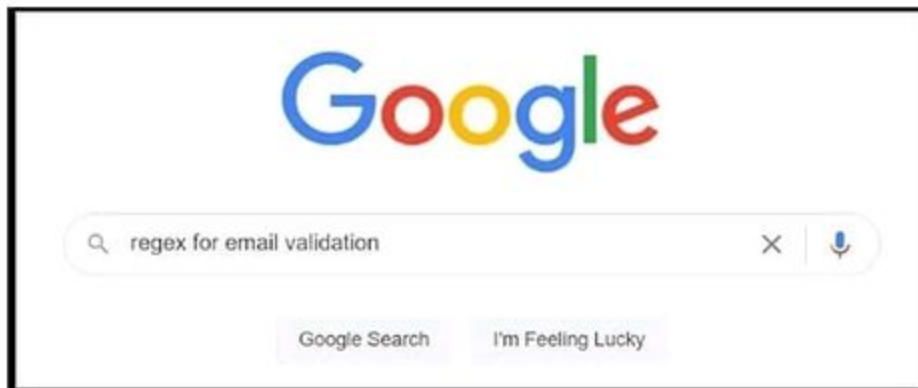


Trick XNOR Treat



@38mo1

## DAY1 OF PROGRAMMING



## 10 YEARS OF PROGRAMMING





**MOST OF MY CODE**

**THE PARTS COPIED  
FROM STACK OVERFLOW  
AND GENERATED WITH AI**

**// DON'T TOUCH IT WORKS**

> **When do we meet?**

Every **Tuesday**

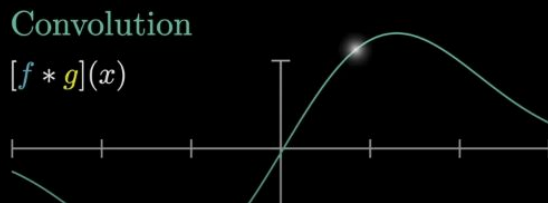
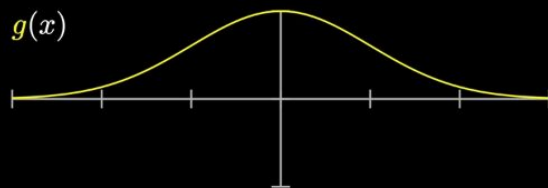
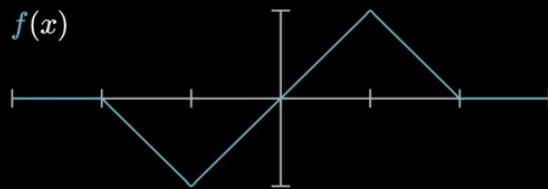
At lunch (**1:30-2:10**)

in **i5** (here!)



# > What are Convolutions?

- Is a mathematical function between two functions that produces a third function
- Simply put, it expresses how the shape of one function is modified by a different function



## > Mathematical Definition

- Now this is pretty complicated, but I'm not going to expect you to understand what this is/how this works!

$$(f * g)(t) \stackrel{\text{def}}{=} \int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau$$

## > Visual Explanation (3Blue1Brown)

<https://youtu.be/KuXjwB4LzSA?t=242>

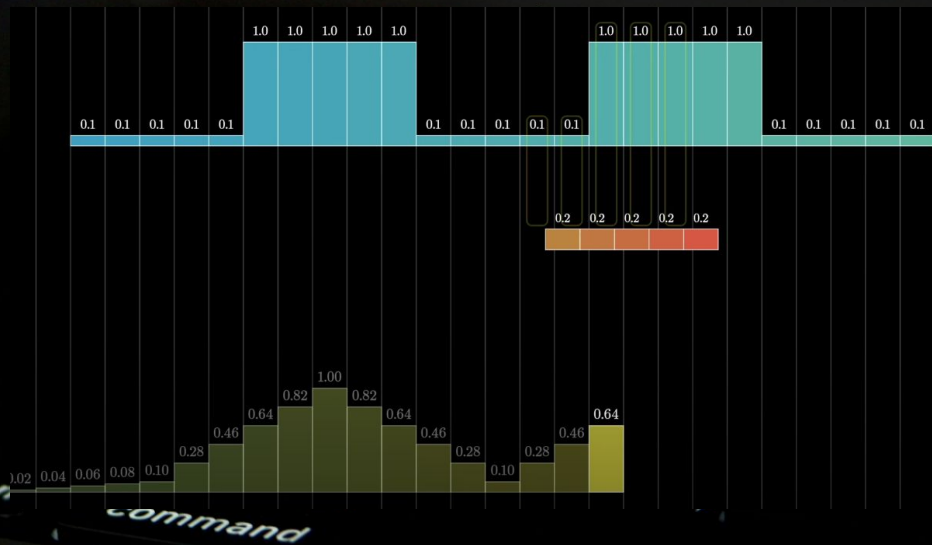
Here we are doing a convolution with 2 1d arrays to get a 3rd 1d array that is the convolution of these 2 1d arrays.



## > Convolutions (ctd)

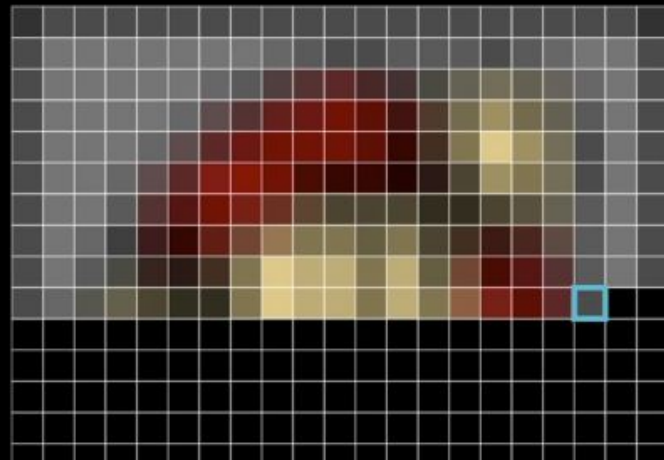
- Seems decently simple right!
- Here's an example for doing a moving average with these two arrays.

$$(a * b)_n = \sum_{\substack{i,j \\ i+j=n}} a_i \cdot b_j$$



## > 2D Convolutions (blurring)

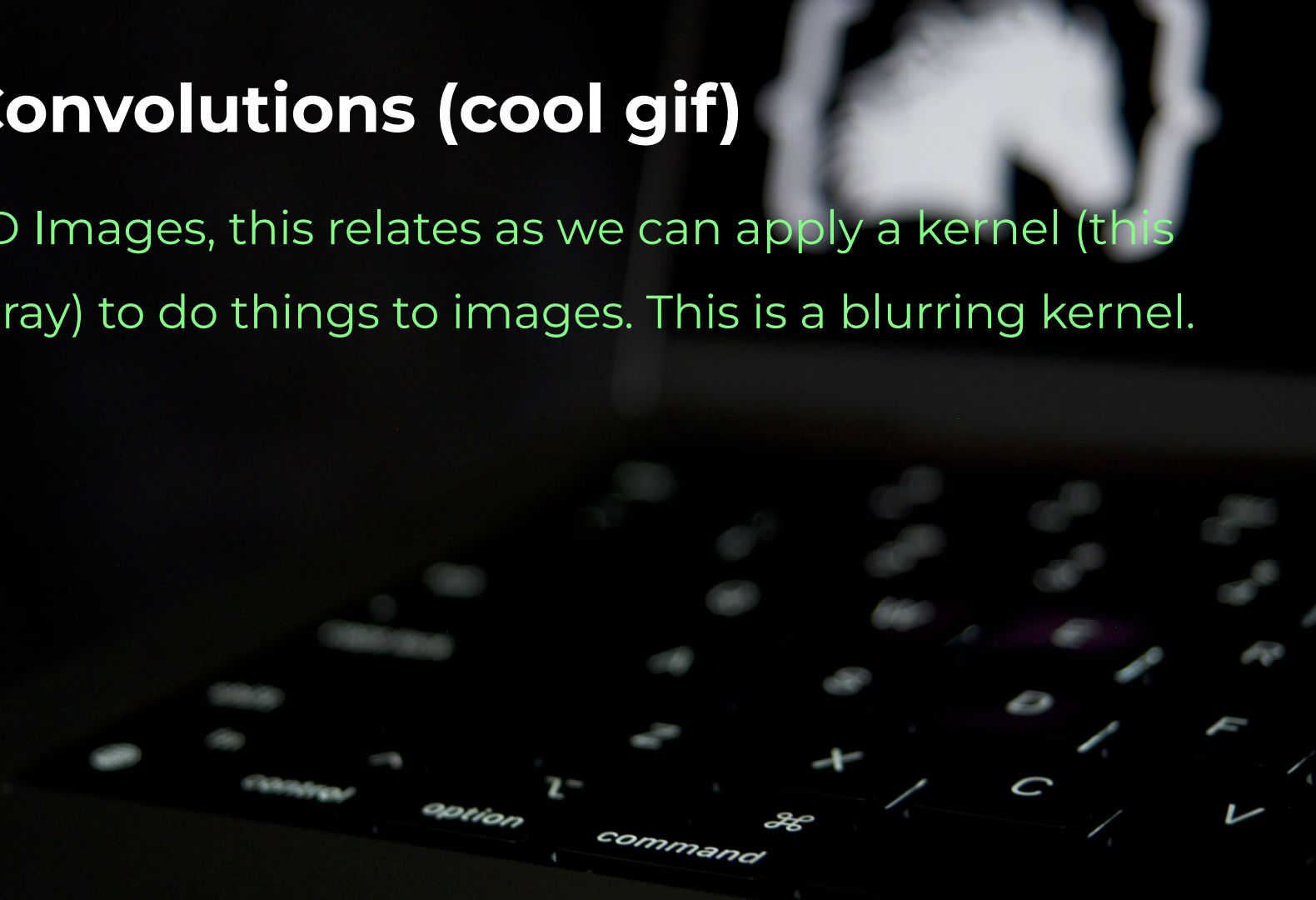
- For 2D Images, this relates as we can apply a kernel (this 3x3 array) to do things to images. This is a blurring kernel.



command

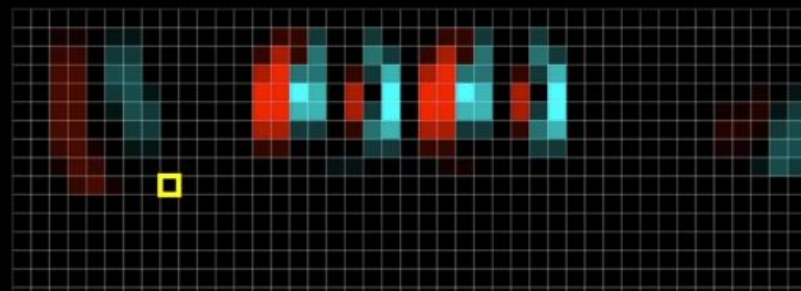
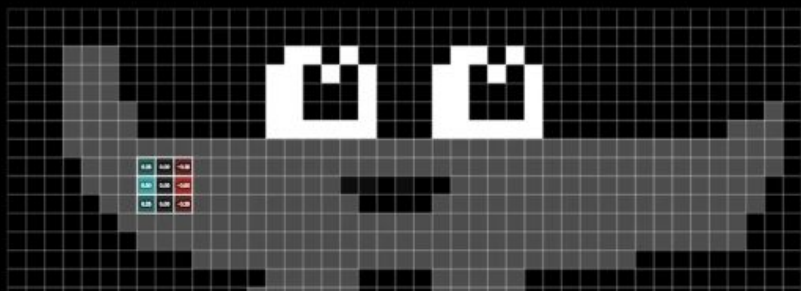
## > 2D Convolutions (cool gif)

- For 2D Images, this relates as we can apply a kernel (this 3x3 array) to do things to images. This is a blurring kernel.



## > 2D Convolutions (Edge Detection)

- Another type of kernel is one that can do edge detection.



- Sideways edge detection kernel!

	0.25	0.00	-0.25
	0.50	0.00	-0.50
	0.25	0.00	-0.25



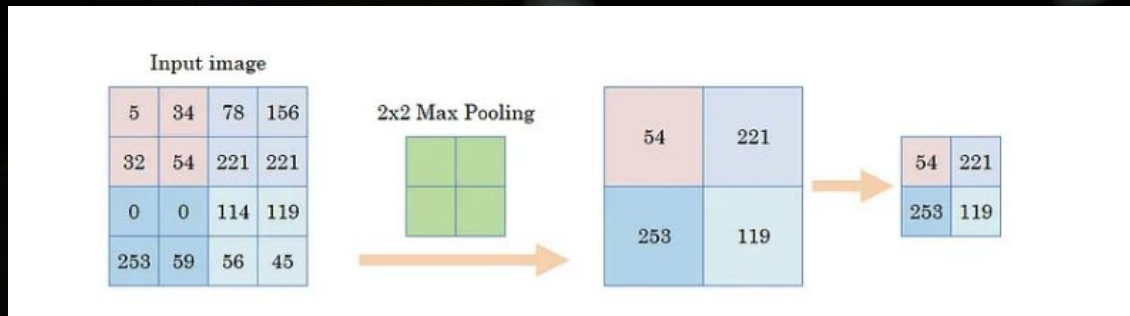
## > **Convolutions take a lot of time :(**

- So how can we make it faster?
- Two ways (that we are going to cover today): pooling & fast fourier transforms



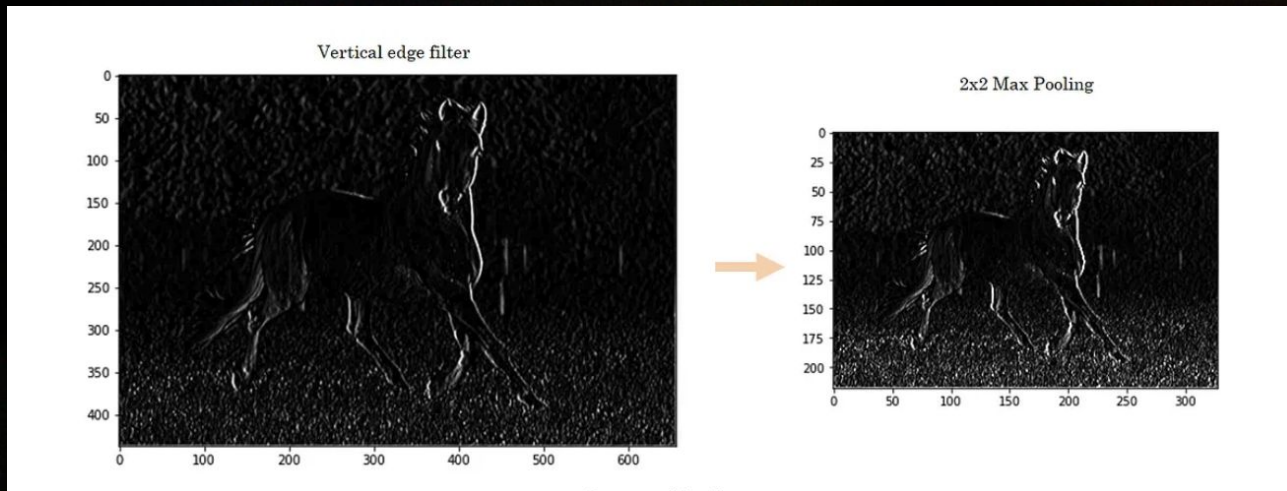
## > Pooling

- Essentially is extracting the important features from the image; Shrinking down the resolution to care only about the important things
- Max pooling (below) vs average pooling



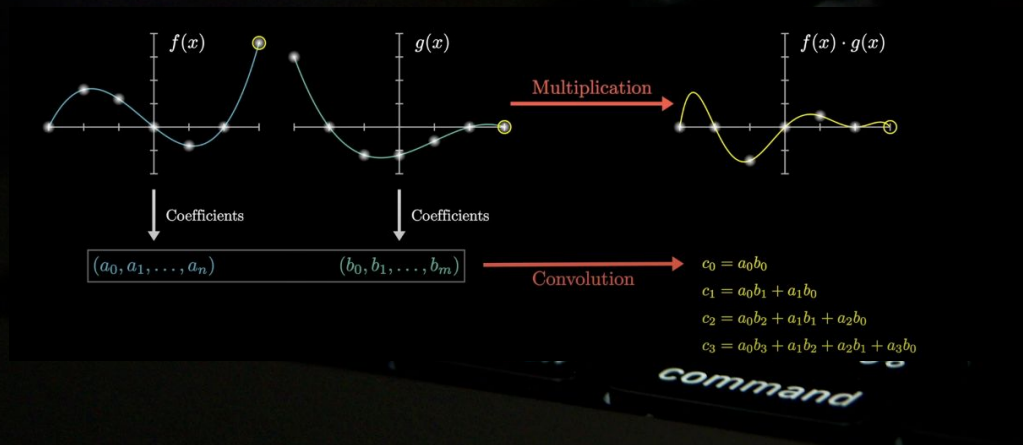
## > Pooling

- After that max pool, this is what an image will look like
- Notice resolution shrunk by two
- This stuff is useful for CNN's so try to get it down!



# > Fast Fourier Transforms

- It's a whole whole whole bunch of complicated calculus!
- But essentially... it means that I can extract the important sample coefficients from a simple multiplication for a convolution.



## > Fast Fourier Transforms

- So then you're left with a system of equations to solve for these coefficients (which will take  $O(n^2)$  to solve)
- A discrete fast fourier transform can take a finite sample of points and find the function that it takes in  $O(n \log n)$ .
- This is a very very important algorithm in today's world (used in telephone lines to convert digital to analog).





Google Colab  
Time!  
[hhscs.club](https://hhscs.club)



## > Socials

**Website:** [hhscs.club](https://hhscs.club)

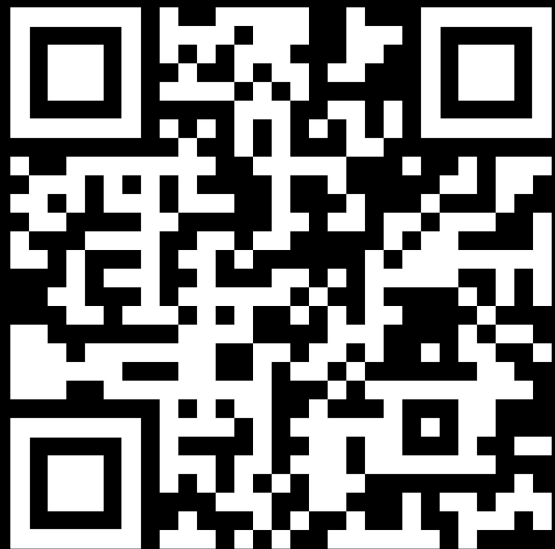
**Email List:**



**Insta:**

[@hhscomputerscience](https://www.instagram.com/hhscomputerscience)

**Discord:**





Next Meeting:  
Tuesday (11/7) Lunch