

## Convolution

Kernel is a grid of weights

Kernel is applied to the image

Used in image processing

Image

4	5	6	9	2	4
3	4	6	8	8	1
2	4	7	9	5	0
7	9	0	4	3	6
0	7	2	4	6	8
7	8	3	5	2	9

Kernel

1	0	-1
0	3	2
-2	1	0

Kernel Applied to image

$$\text{sum}(\text{image}_i * K_i = 4 * 1 + 5 * 0 + 6 * (-1) + 3 * 0 + 4 * 3 + 6 * 2 + 2 * (-2) + 1 * 1 + 7 * 0 = 19$$

Resulting Matrix = is output

	19				

## Kernels

### Feature Detectors

Kernels are learned

Oblique line detector

1	0	0
0	1	0
0	0	1

Vertical Line detector

0	1	0
0	1	0
0	1	0

CNN learns from many Kernels different features.CNN learns the weight of the Kernel

## Architectural decisions for convolution

Grid size

Number Of pixels for height/width

Odd Numbers used

Stride

Step size used for sliding kernel on image

Indicated in pixels

Depth

Kernel =  $3 \times 3 \times 3$ , Number Of weight = 27

Number Of Kernels

A conv layer has multiple kernels

Each kernel outputs a single 2D array

Output from a layer Number of 2d arrays = number of kernels

## Pooling

- Downsample the image
- Overlaying grid on image
- Max/average pooling
- No parameters

## Pooling Settings

- Grid Size
- Stride
- Type(max ,average)

## Example

### Max pooling (2x2, stride 2)

Input					Output	
-1	2	0	2		10	18
3	18	10	-3		12	7
2	12	5	2			
1	3	7	4			

Input ->

Feature Learning =

[Convolution + relu -> pooling -> Convolution + relu -> pooling]->

Classification =[ Flatten -> Fully Connected -> Softmax]

## CNN apply to audio image

Spectrogram/MFCC = 2D image

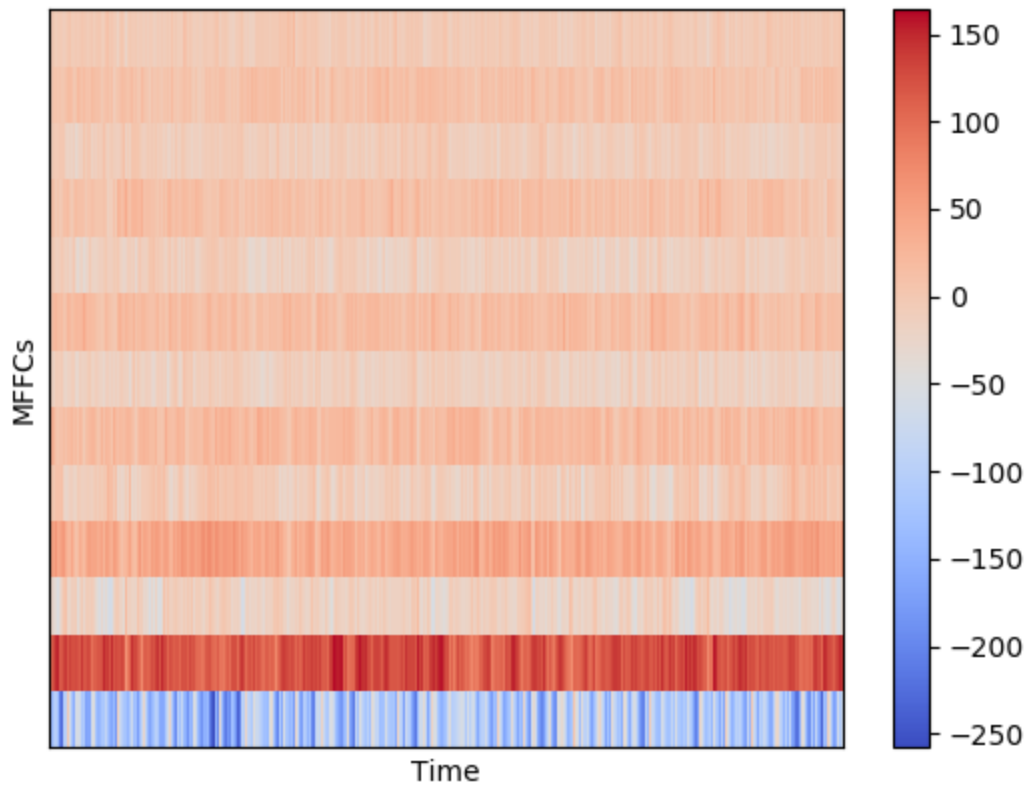
Time, frequency = x, y

Amplitude = pixel value

Example 13 MFCC

Hop length = 512 sample

Number Of samples in audio file = 51200



Data shape = 100 x 13 x 1 , 1 means gray scale image => input to Tensorflow

## Output from CNN

**Accuracy on test set is: 0.7490909099578857**

**Expected index: 6, Predicted index: [6]**

**Expected Value: metal, Predicted Value: ['metal']**

