



Table of Content What will We Learn Today?

- 1. What is image classification?
- 2. What is Convolutional Neural Network?
- 3. Who is the inventor of CNN?
- 4. How CNN works?
- 5. Hands on CNN







What image is this?







How about this?











How do you know cats and dogs are different?









Image classification

Image classification is the process of categorizing and labeling groups of pixels or vectors within an image based on specific rules. The categorization law can be devised using one or more spectral or textural characteristics.

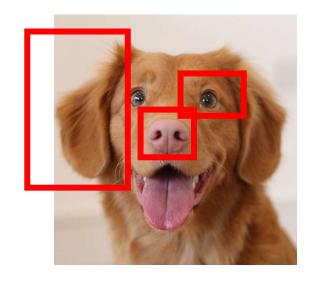


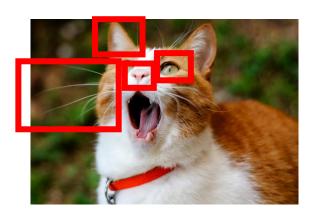




To put simply...

Image classification is grouping / labeling image from images' features or pattern.





* Red rectanges are potential features that differ dog from cat

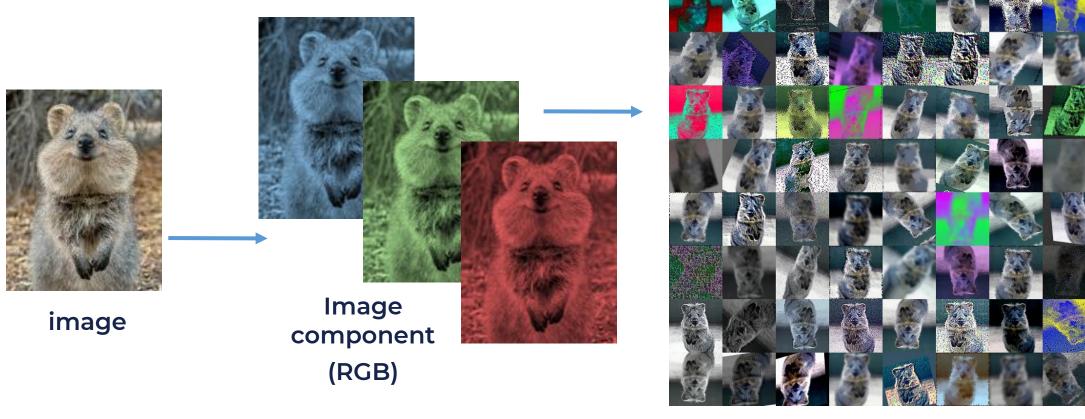


How computer extract image's feature?





Feature of images



features

Resize, filter, transform are form of feature image methods



How to extract while classify image?



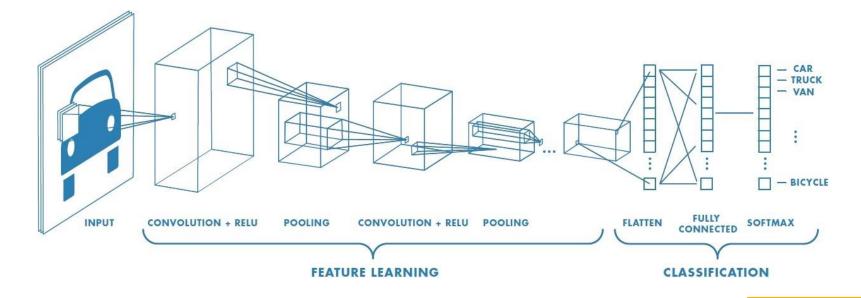




Convolutional Neural Network

Convolutional Neural Network (CNN) is a type of artificial neural network used in image recognition and processing that is specifically designed to process pixel data.

CNNs are powerful image processing, artificial intelligence (AI) that use deep learning to perform both generative and descriptive tasks







Yann LeCun

Yann André LeCun is a French computer scientist working primarily in the fields of machine learning, computer vision, mobile robotics, and computational neuroscience. He is one of Geoffrey Hinton's student (Geoffrey is godfather of Deep learning).

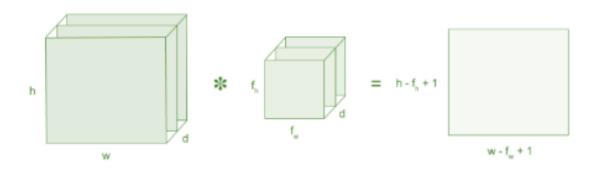






Convolution Layer

- Convolution is the first layer to extract features from an input image.
- Convolution preserves the relationship between pixels by learning image features using small squares of input data.
- It is a mathematical operation that takes two inputs such as image matrix and a filter or kernel.
 - An image matrix (volume) of dimension (h x w x d)
 - A filter (f_h x f_w x d)
 - Outputs a volume dimension (h f_h + 1) x (w f_w + 1) x 1







Convolution Layer (2)

Consider a 5 x 5 whose image pixel values are 0, 1 and filter matrix 3 x 3 as shown
 in below

1	1	1	0	0	
0	1	1	1	0	
0	0	1	1	1	
0	0	1	1	0	
0	1	1	0	0	

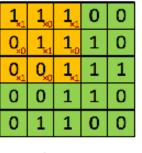




5 x 5 - Image Matrix

3 x 3 - Filter Matrix

 Consider a 5 x 5 whose image pixel values are 0, 1 and filter matrix 3 x 3 as shown in below





Image

Convolved Feature





Convolution Layer (3)

- Convolution of an image with different filters can perform operations such as edge detection, blur and sharpen by applying filters.
- Example beside shows various convolution image after applying different types of filters (Kernels).

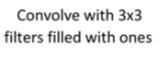
Operation	Filter	Convolved Image	
Identity	$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$		
	$\begin{bmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{bmatrix}$		
Edge detection	$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$		
	$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$		
Sharpen	$\begin{bmatrix} 0 & -1 & 0 \\ -1 & 5 & -1 \\ 0 & -1 & 0 \end{bmatrix}$		
Box blur (normalized)	$\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$		
Gaussian blur (approximation)	$\frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$	9	

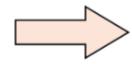




Stride is the number of pixels shifts over the input matrix. When the stride is 1 then we move the filters to 1 pixel at a time. When the stride is 2 then we move the filters to 2 pixels at a time and so on. The below figure shows convolution would work with a stride of 2.

1	2	3	4	5	6	7
11	12	13	14	15	16	17
21	22	23	24	25	26	27
31	32	33	34	35	36	37
41	42	43	44	45	46	47
51	52	53	54	55	56	57
61	62	63	64	65	66	67
71	72	73	74	75	76	77





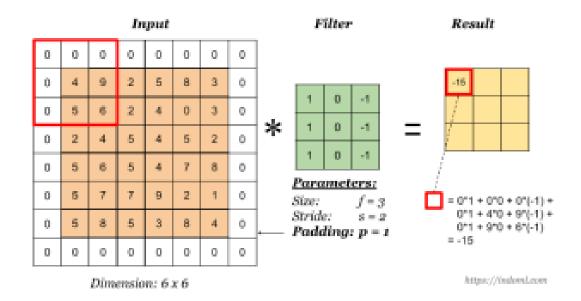
108	126	
288	306	





Sometimes filter does not fit perfectly fit the input image. We have two options:

- Pad the picture with zeros (zero-padding) so that it fits
- Drop the part of the image where the filter did not fit. This is called valid padding which keeps only valid part of the image.







Non Linearity (ReLU)

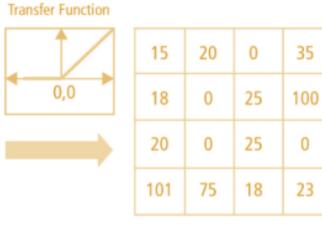
- ReLU stands for Rectified Linear Unit for a non-linear operation.
- The output is f(x) = max(0,x).
- Why ReLU is important: ReLU's purpose is to introduce non-linearity in our ConvNet.
 Since, the real world data would want our ConvNet to learn would be non-negative linear values.

 15
 20
 -10
 35

 18
 -110
 25
 100

 20
 -15
 25
 -10

 101
 75
 18
 23



There are other non linear functions such as tanh or sigmoid that can also be used instead of ReLU. Most of the data scientists use ReLU since performance wise ReLU is better than the other two.

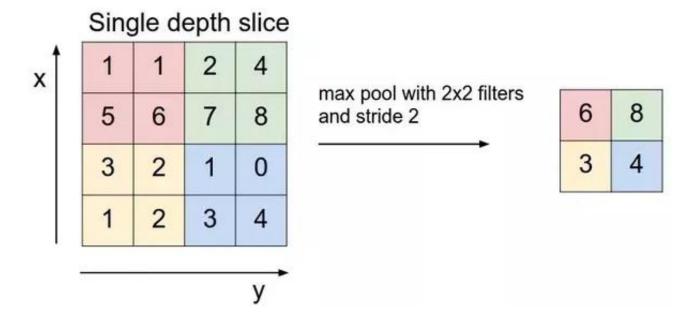
ReLU Layer





Pooling layers section would reduce the number of parameters when the images are too large. Spatial pooling also called subsampling or downsampling which reduces the dimensionality of each map but retains important information. Spatial pooling can be of different types:

- Max Pooling
- Average Pooling
- Sum Pooling



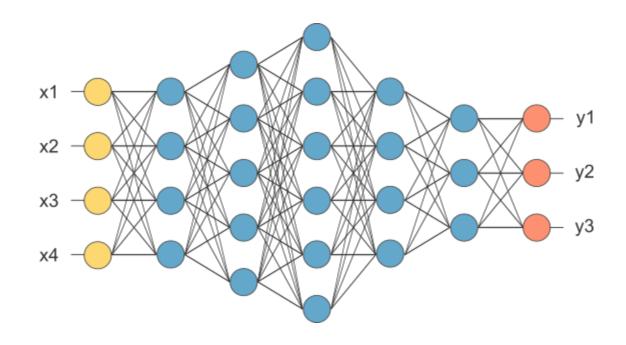




Fully Connected Layer (NN)

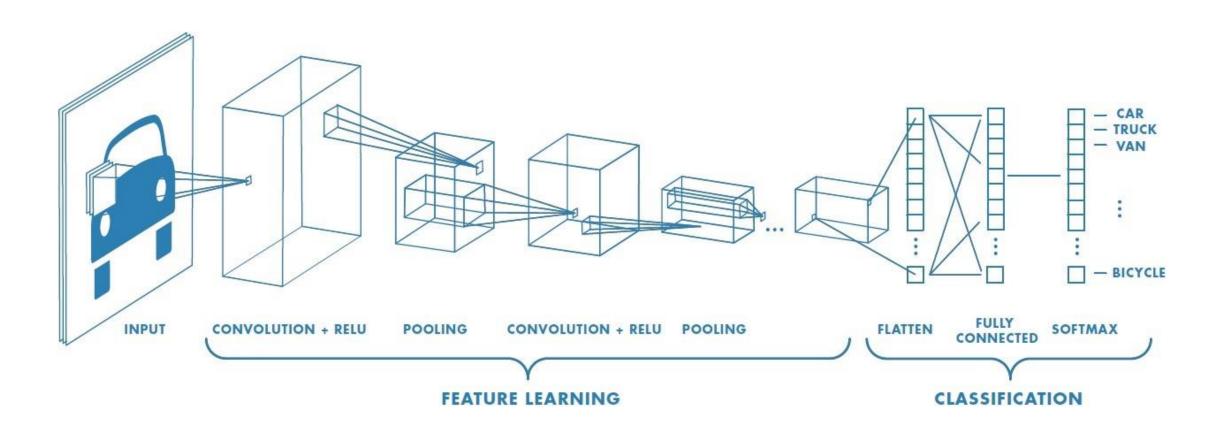
The layer we call as FC layer, we flattened our matrix into vector and feed it into a fully connected layer like a neural network.

With the fully connected layers, we combined these features together to create a model. Finally, we have an activation function such as softmax or sigmoid to classify the outputs as cat, dog, car, truck etc.





Final Architecture





Let's have a hands-on





Thank YOU

