Neural Networks for Images How to see like a human

Yordan Darakchiev iordan93@gmail.com



Machine Learning

- Making a program which performs a task without explicitly programming it
 - Like the way people learn



Seeing new thingsLearning

Neural Network



Neural Network Learning



Convolution

I(0,0)	I(1,0)	I(2,0)	I(3,0)	I(4,0)	I(5,0)	I(6,0)
I(0,1)	I(1,1)	I(2,1)	I(3,1)	I(4,1)	I(5,1)	I(6,1)
I(0,2)	I(1,2)	I(2,2)	I(3,2)	I(4,2)	I(5,2)	I(6,2)
I(0,3)	I(1,3)	I(2,3)	I(3,3)	I(4,3)	I(5,3)	I(6,3)
I(0,4)	I(1,4)	I(2,4)	I(3,4)	I(4,4)	I(5,4)	I(6,4)
I(0,5)	I(1,5)	I(2,5)	I(3,5)	I(4,5)	I(5,5)	I(6,5)
I(0,6)	I(1,6)	I(2,6)	I(3,6)	I(4,6)	I(5,6)	I(6,6)

	H(0,0)	H(1,0)	H(2,0)					
×	H(0,1)	H(1,1)	H(2,1)	=				
	H(0,2)	H(1,2)	H(2,2)					
Filter								

O(0,0)		

Input image

Output image

How about many convolutions?

Convolution Layer

- Many channels
- Many filters (kernels)



Looks just like a "usual" NN!But works perfectly on images!

A Complete Model

- Convolution layers
- Pooling layers
- "Standard" layers



How Deep Can We Go?

- Inception ResNet V2
- Each rectangle is a layer



Fig. 1. Architecture of Inception Resnet V2 Network

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The Gritty Details

- Activation function
- Hyperparameters
- Batch normalization
- Regularization
- Function optimization
- ... and other "-ations"

The Applications

Image classification & friends



The Applications (2)

Image captioning & generation



The Applications (3)

Keypoint detection, pose estimation, activity recognition



The Applications (4)

Robotic control, autonomous vehicles



In Real Life

- Lots of data
- Data quality
- Finding biases
- Producing metrics
- Fine-tuning models
- ExplainabilityResponsibility

Thank you!