

# Deep Convolutional Networks

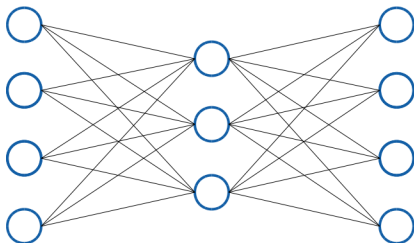
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Study Project

24th of February 2016

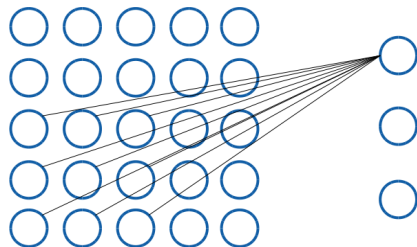
# Convolutional Neural Networks

- Learns the weights of convolutional filters
- Exploits spatial structure in the input
- Convolution of entire input with filter implies shared weights
- Reduced amount of weights allows lots of filters
- Filters specific to color channels



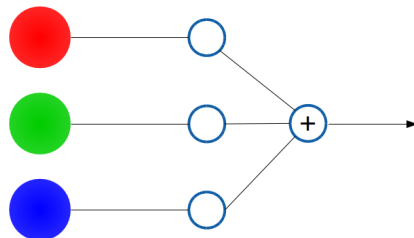
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# Network Structure

Layer	Type	Configuration	Activation function
0	Convolutional	100 filters of size $7 \times 7$ per channel	tanh
1	Max Pooling	Pool size $2 \times 2$	-
2	Convolutional	150 filters of size $4 \times 4$ per channel	tanh
3	Max Pooling	Pool size $2 \times 2$	-
4	Convolutional	250 filters of size $4 \times 4$ per channel	tanh
5	Max Pooling	Pool size $2 \times 2$	-
6	Dense	300 neurons	tanh
7	Dense	43 neurons	softmax

# German Traffic Sign Recognition Benchmark

**RUB**

- Dataset of traffic signs taken while on the road
- 39209 training and 12630 test images in 43 classes
- Images contain a 10% border around the sign
- Annotations provide precise sign locations
  - We cropped out just the sign
- Show some images



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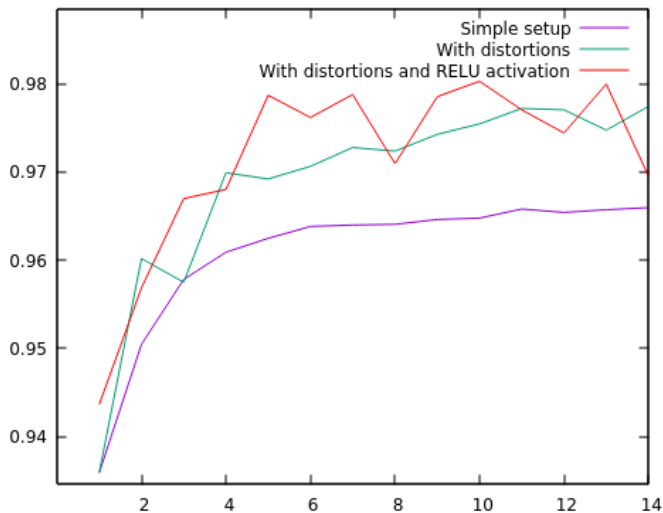


# Simple Setup

- Input size  $48 \times 48$
- Contrast normalization
- Describe Simple Setup
- Present Results



# Results on GTSRB



# Input Distortions

- Mention input distortions
- Explain them
- Present distortion parameters
- Maybe add one or two images before and after the transformations

# Results with RELU

- Add RELU image
- Present results with RELU activation function

# Missclassified images



Input



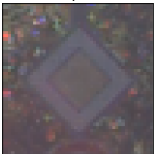
0.5814



0.2840



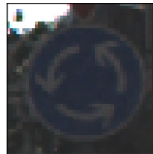
0.0401



Input



0.8513



0.0923



0.0493



Input



0.4296



0.2681

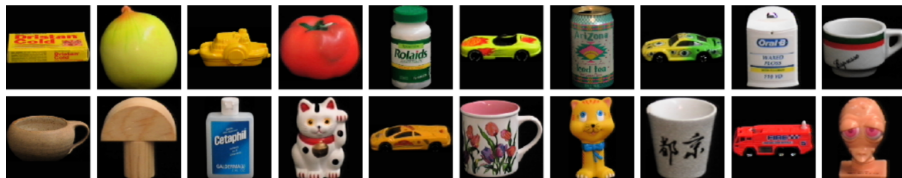


0.2390

# Filter Reuse

- How well do the GTSRB filters generalize?
- Initialize new network with same structure randomly
- Copy GTSRB filters to the new network
- Train only the fully connected layers!

## COIL100



- Columbia Object Image Library 100  $\Rightarrow$  COIL100
- 100 different objects
- Objects turning on a black turntable
- One foto each time the object has turned by  $5^\circ$
- 72 images per object, 7200 images in total
- Random separation into 58 training and 14 test images per object

- Describe INRIA dataset
- Show image
- Show results with reused filters
- Show results with original filters

# Conclusion

- Summarize results



# Questions?

**RUB**

Questions?

