

# Enabling machines to perceive the world



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Teradeep Inc.





e-Lab

# team



Berin Martini



Clement  
Farabet



Alfredo  
Canziani



Aysegul  
Dundar



Vinayak  
Gokhale

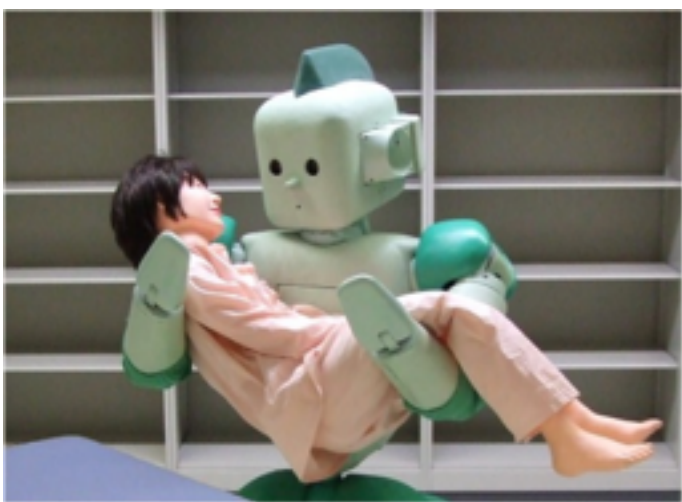


Jonghoon  
Jin



Yann LeCun  
NYU

Our Goal: enable machines to perceive the world like we do!



ROBOTICS



UAV



Autonomous cars



BUILDING SECURITY



BIOMETRICS



HOME AUTOMATION



FOOD INSPECTION



PARTS INSPECTION



PARKING MANAGEMENT

# enabling technology

70k\$



+TeraDeep HW



~100\$

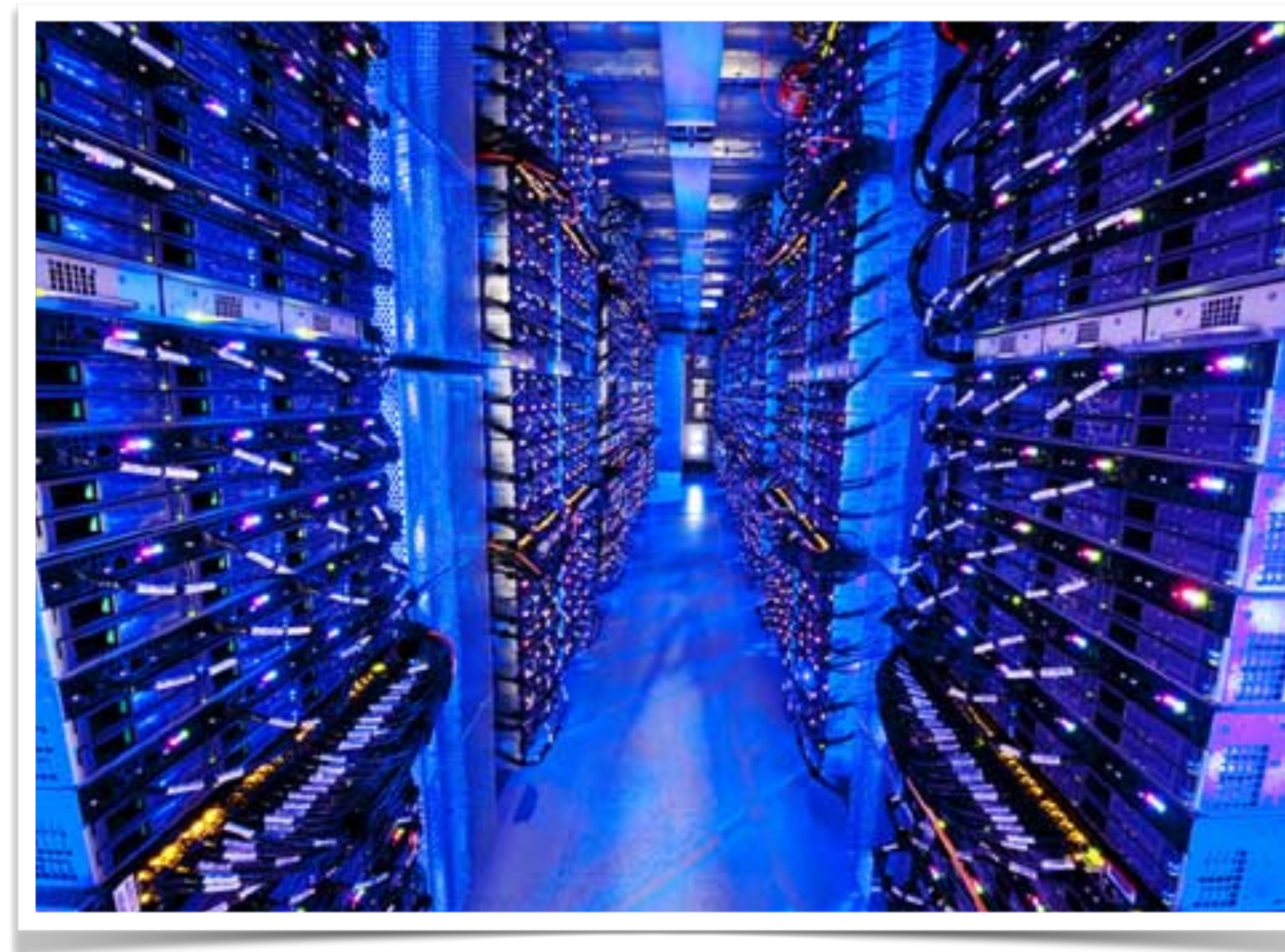
enabling low-cost  
autonomous  
driving!

# enabling technology



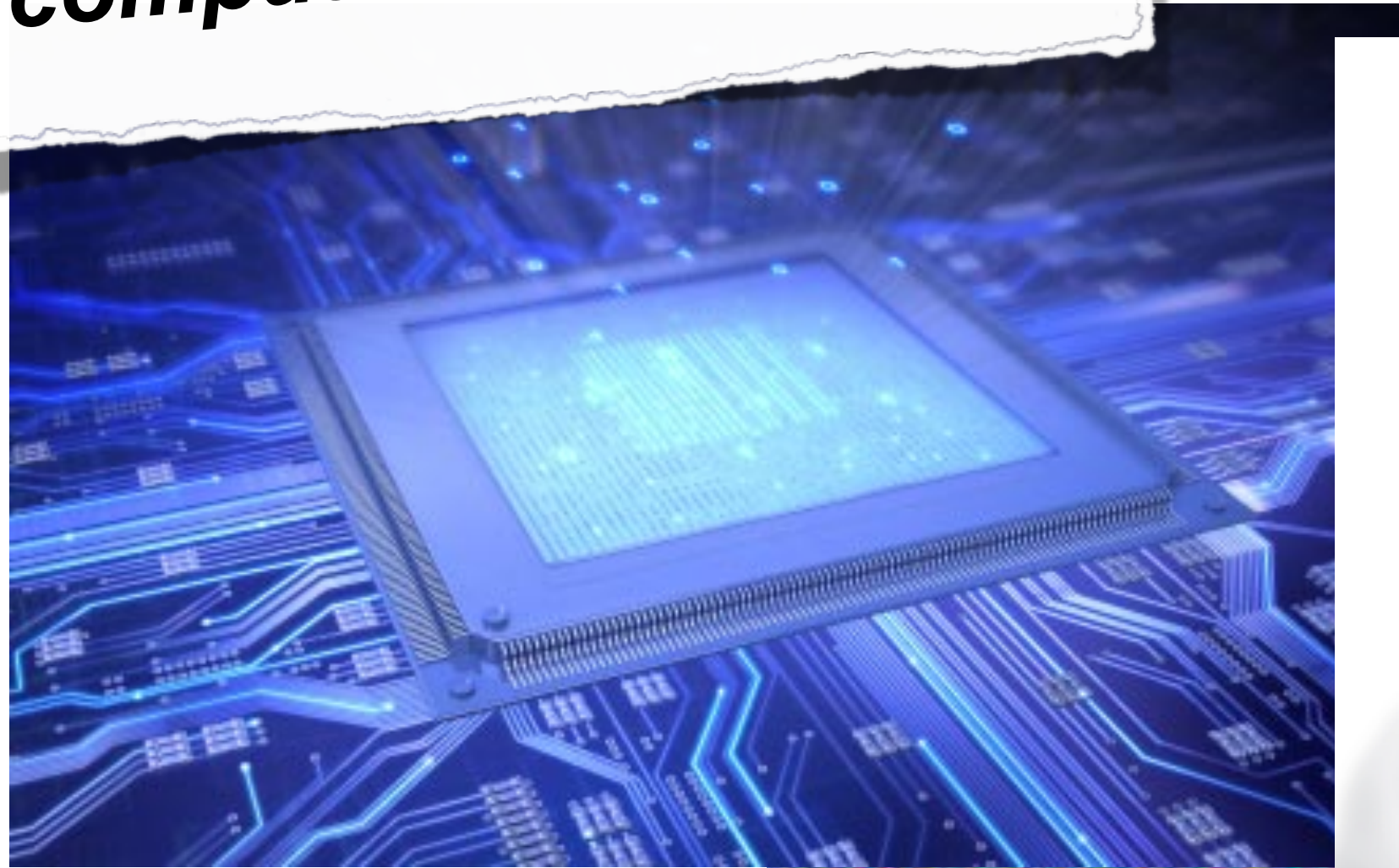
enabling **always-on real-time** hardware  
tagging videos on mobile devices!

# enabling technology

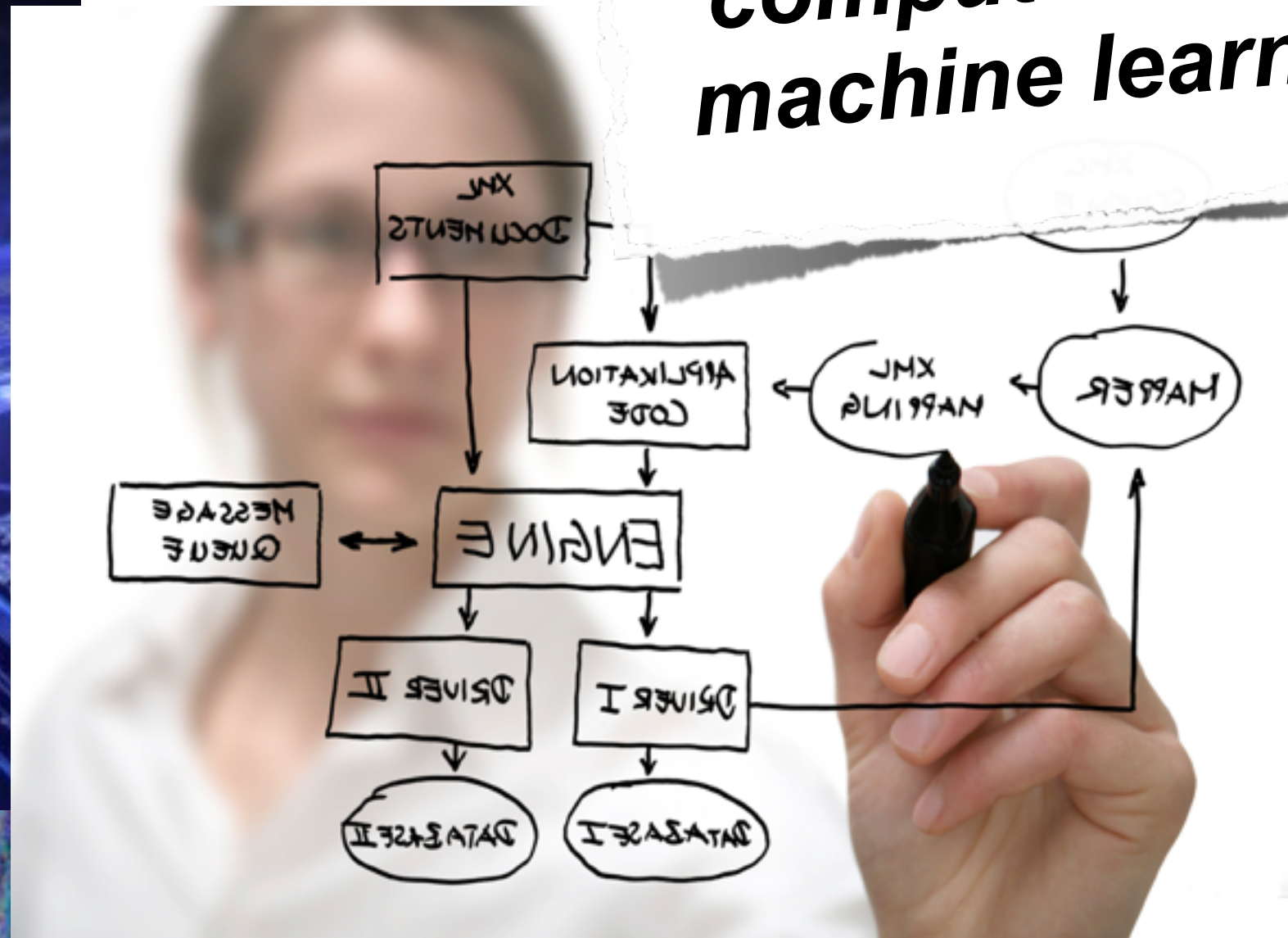


enabling **low-power** servers for parsing big data, images, videos!

**Electrical Engineering**  
neuromorphic engineering  
synthetic vision  
computer architecture



**Computer Science**  
computer vision  
machine learning



**Neuroscience**  
vision  
learning  
neural networks

# our vision system

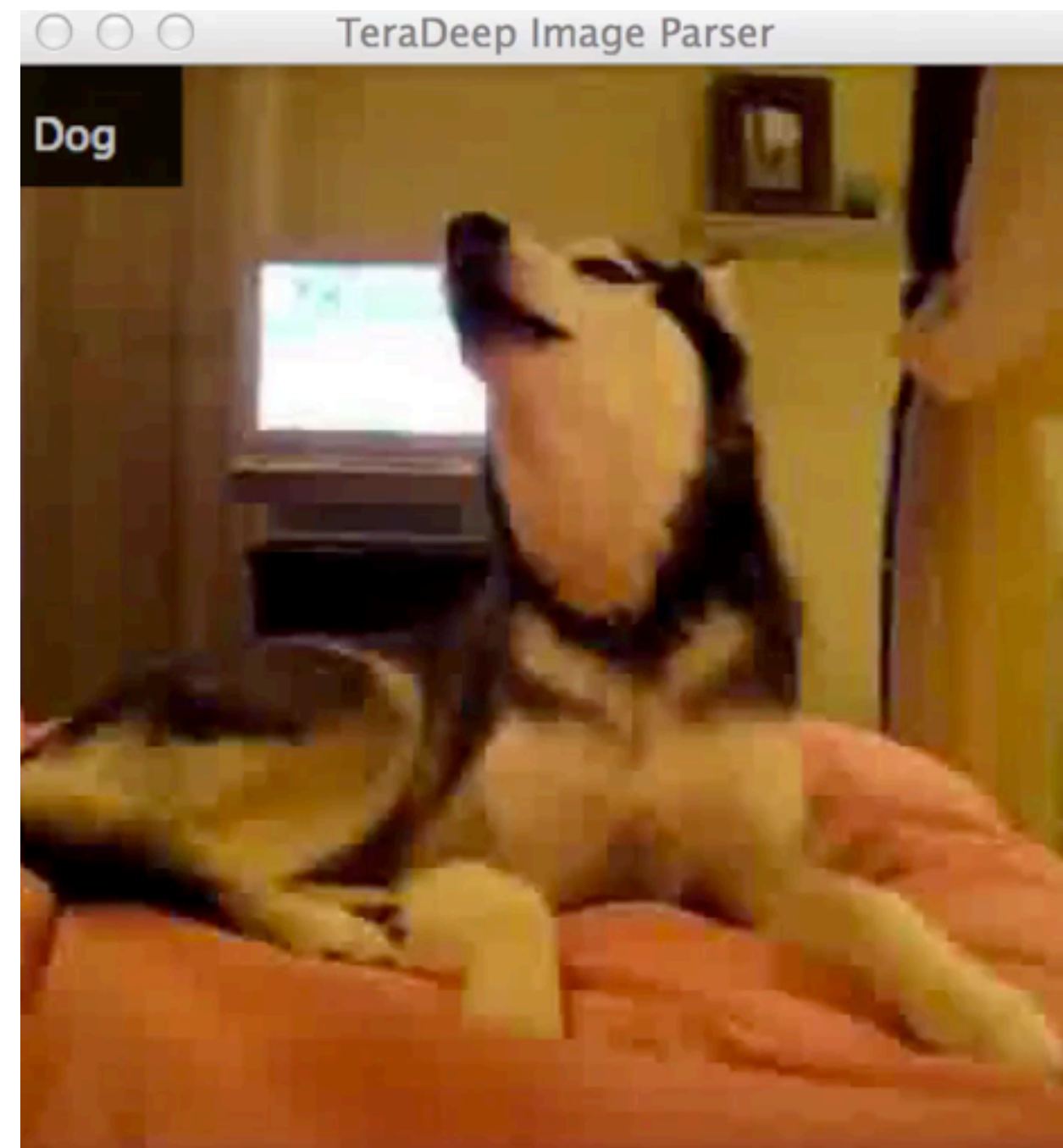
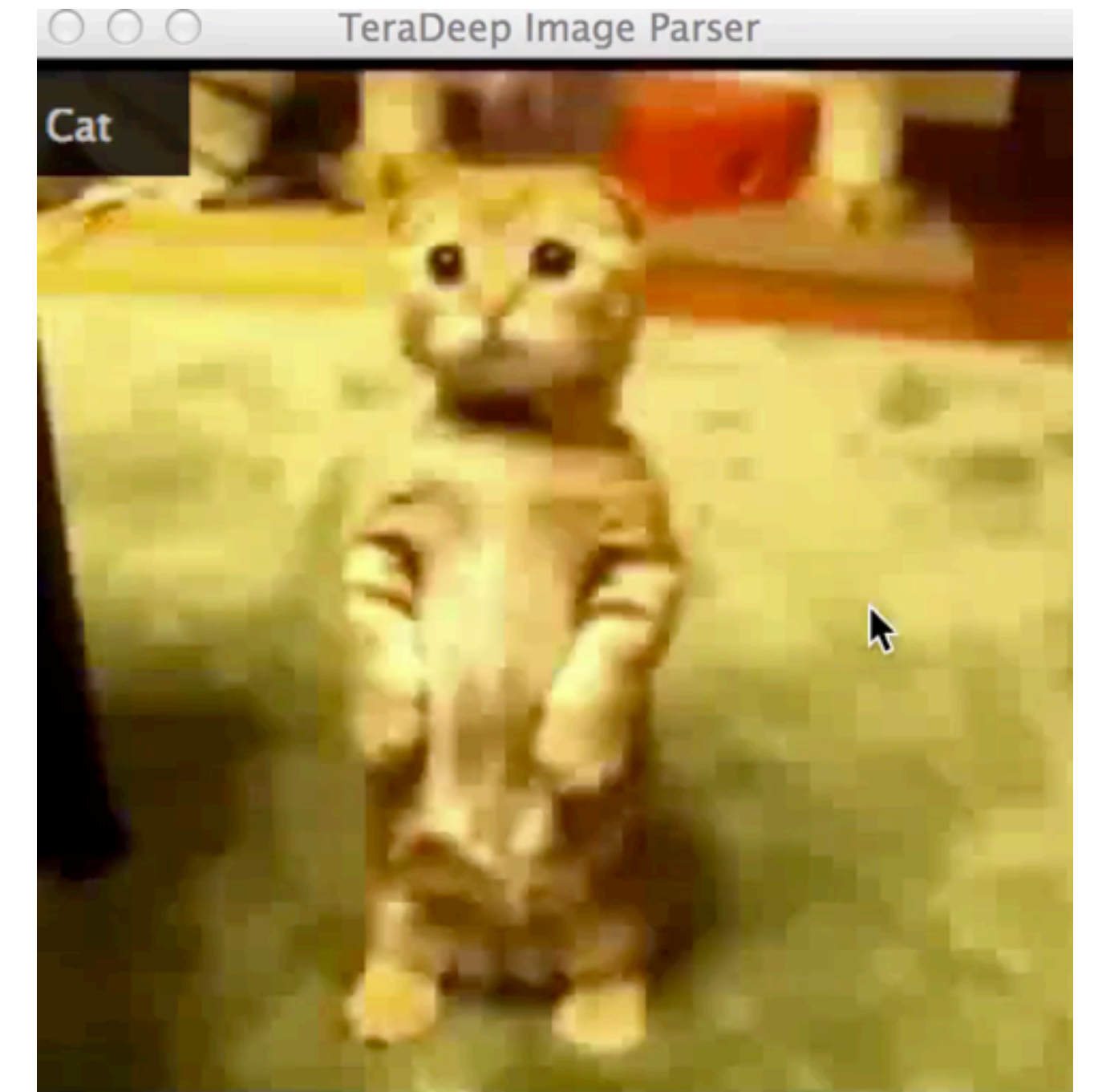


**palm sized!**



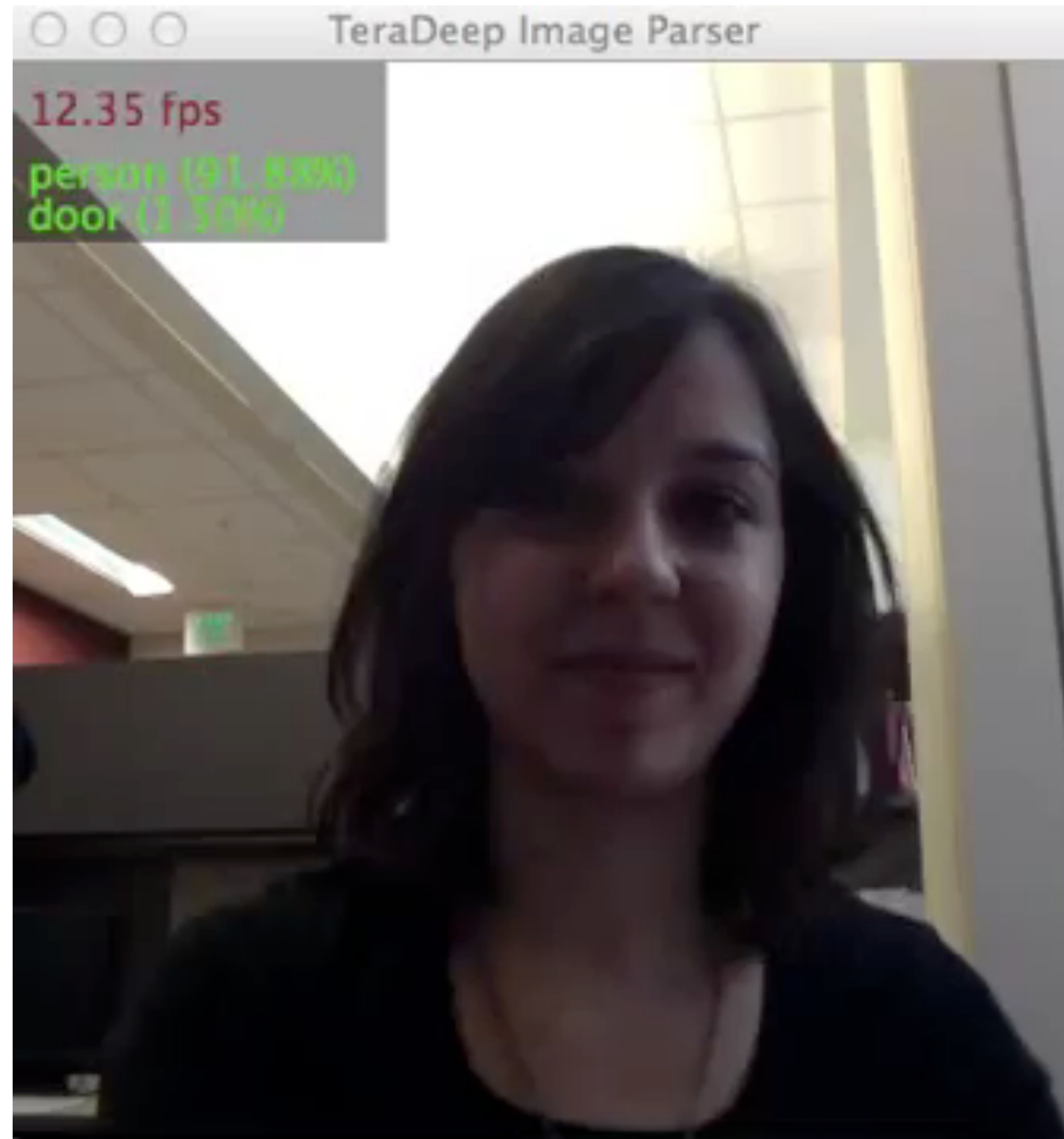


# demos



state-of-the-art

demos



state-of-the-art

# demos

TeraDeep Learner

	Object 1		0
	Object 2		0
	Object 3		0
	Object 4		0
	Object 5		0

error bars: lower is better

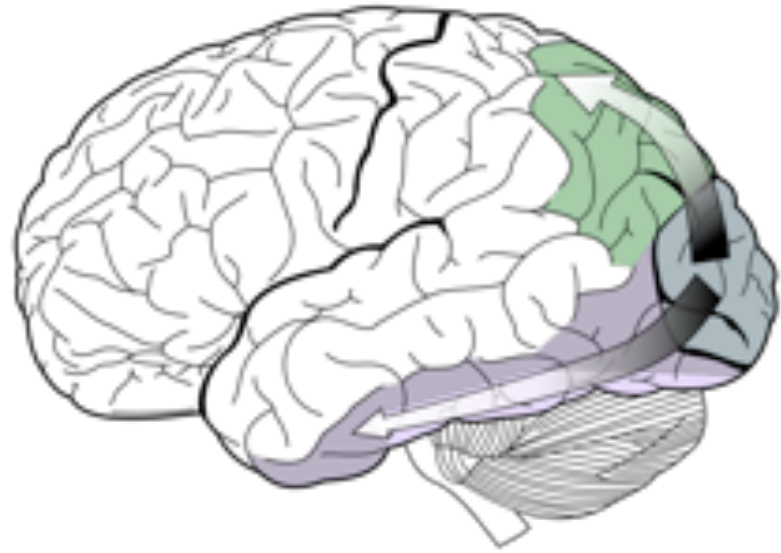
Forget All      Save Session      Load Session

state-of-the-art

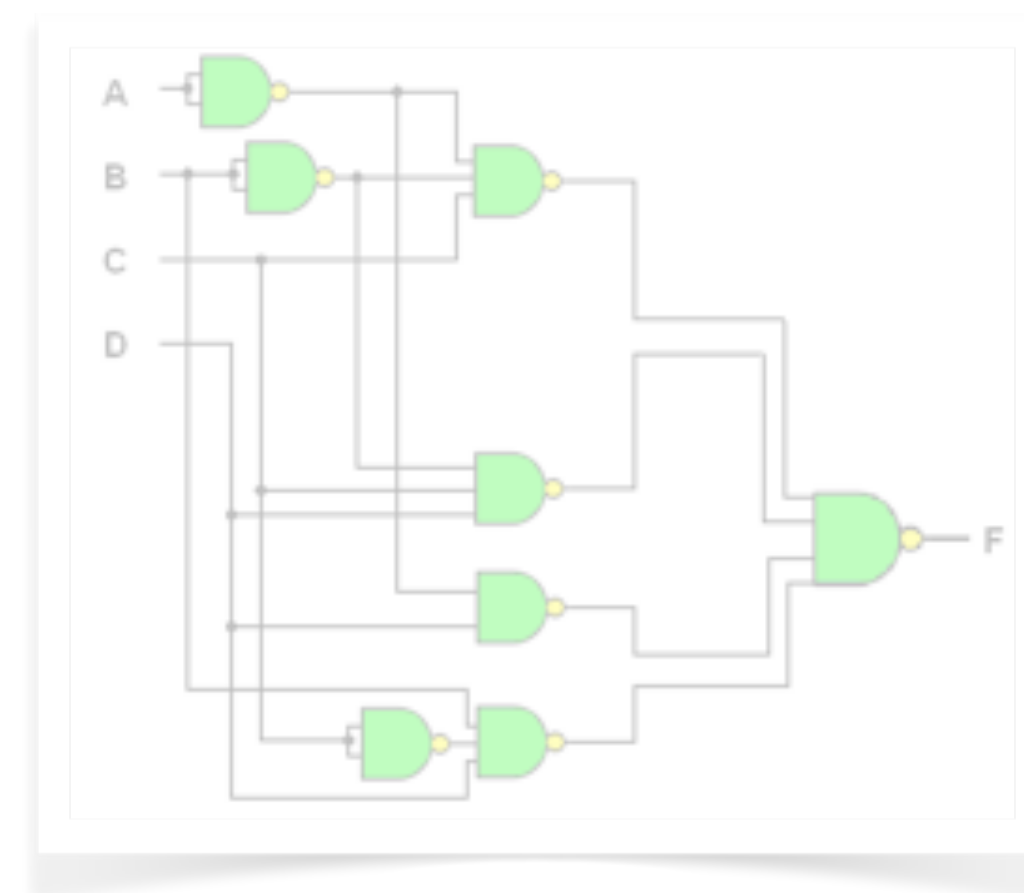
# how did we do this?

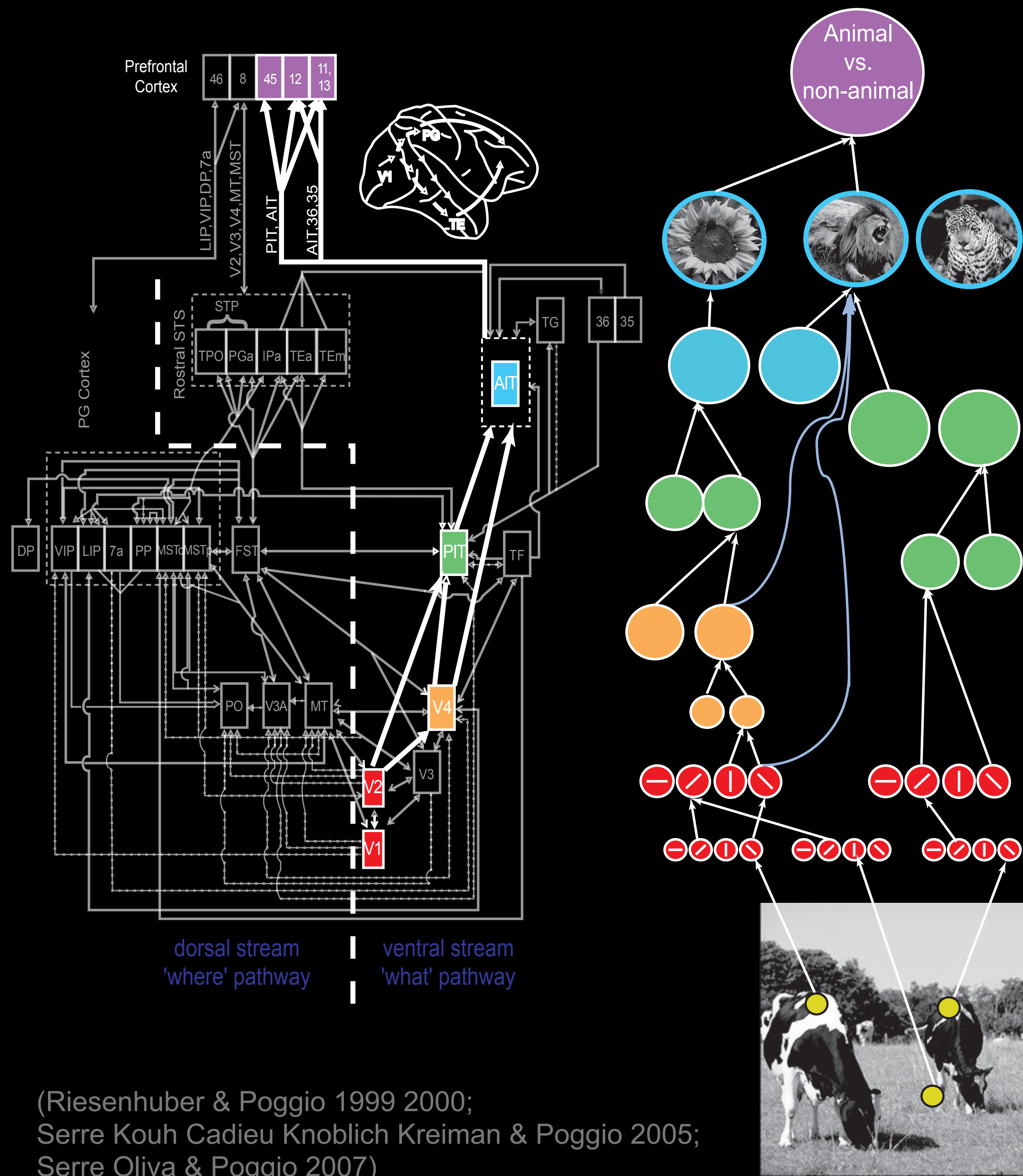
- 1: deep neural networks
- 2: streaming processor architecture
- 3: assemble vision systems
- 4: training deep networks

# 1: deep neural networks



```
begin
  make QUEUE empty;
  for j ← 0 until m - 1 do make Q[j] empty;
  for l ← lmax step -1 until 1 do
    begin
      concatenate LENGTH[l] to the beginning of QUEUE;
      while QUEUE not empty do
        begin
          let Ai be the first string on QUEUE;
          move Ai from QUEUE to bucket Q[Ai];
        end;
      for each j on NONEMPTY[l] do
        begin
          concatenate contents of Q[j] to the end of QUEUE
          make Q[j] empty
        end
      end
    end
  end
end
```





◆ V1:

- Simple and complex cells tuning properties (Schiller et al 1976; Hubel & Wiesel 1965; Devalois et al 1982)
- MAX operation in subset of complex cells (Lampl et al 2004)

◆ V4:

- Tuning for two-bar stimuli (Reynolds Chelazzi & Desimone 1999)
- MAX operation (Gawne et al 2002)
- Two-spot interaction (Freiwald et al 2005)
- Tuning for boundary conformation (Pasupathy & Connor 2001)
- Tuning for Cartesian and non-Cartesian gratings (Gallant et al 1996)

◆ IT:

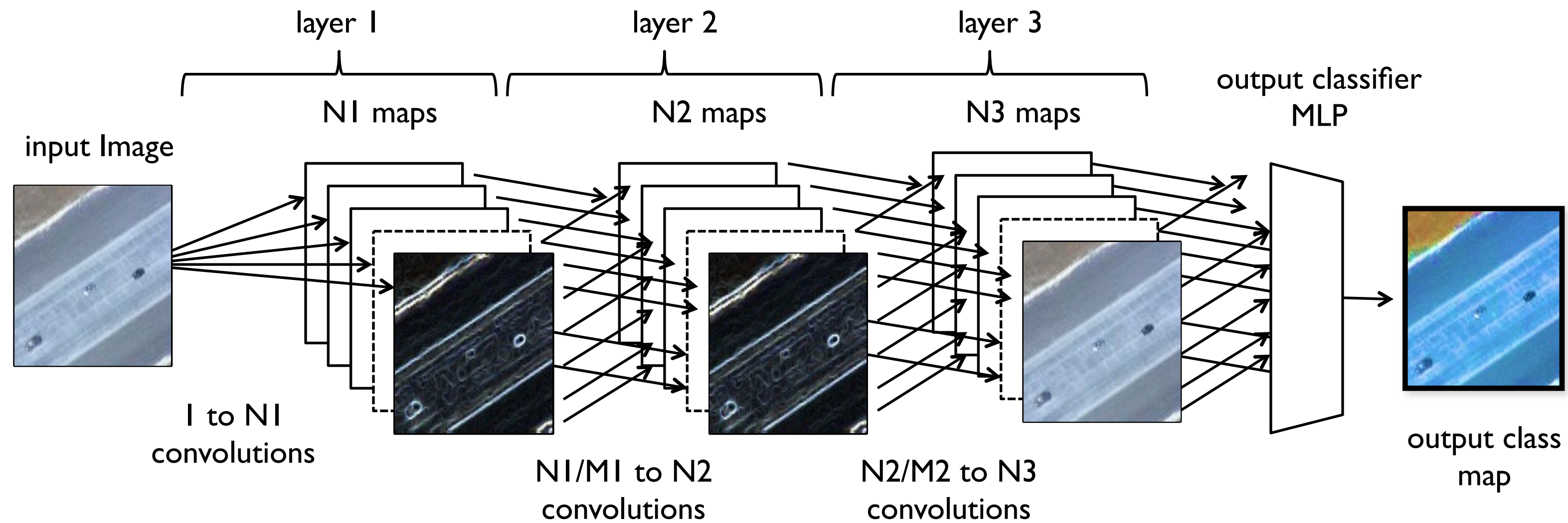
- Tuning and invariance properties (Logothetis et al 1995)
- Differential role of IT and PFC in categorization (Freedman et al 2001 2002 2003)
- Read out data (Hung Kreiman Poggio & DiCarlo 2005)
- Average effect in IT (Zoccolan Cox & DiCarlo 2005; Zoccolan Kouh Poggio & DiCarlo in press)

◆ Human behavior:

- Rapid animal categorization (Serre Oliva Poggio 2007)

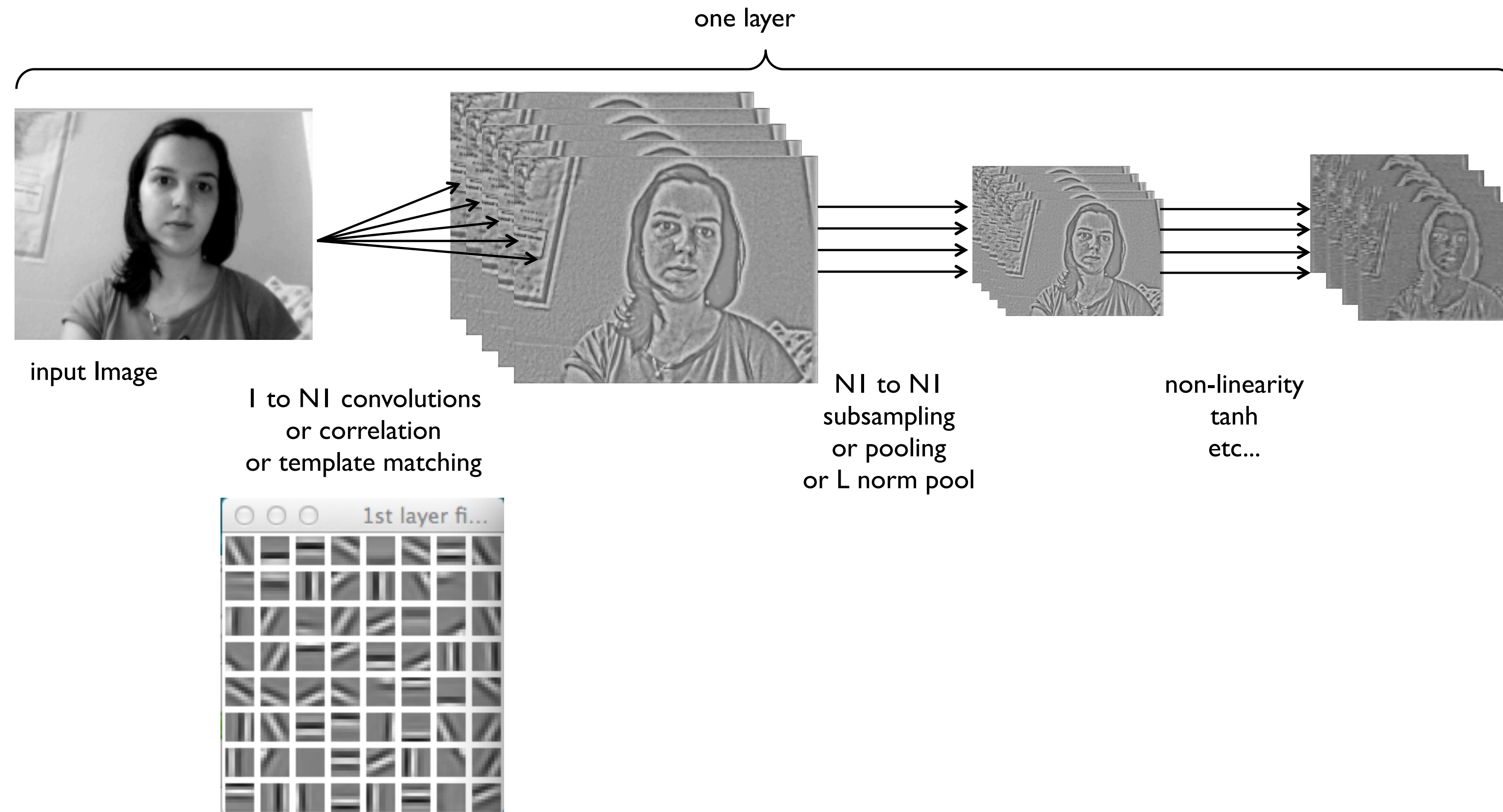


# deep networks



deep, invariant,  
trained on data,  
model of visual system,  
state-of-art

# deep networks

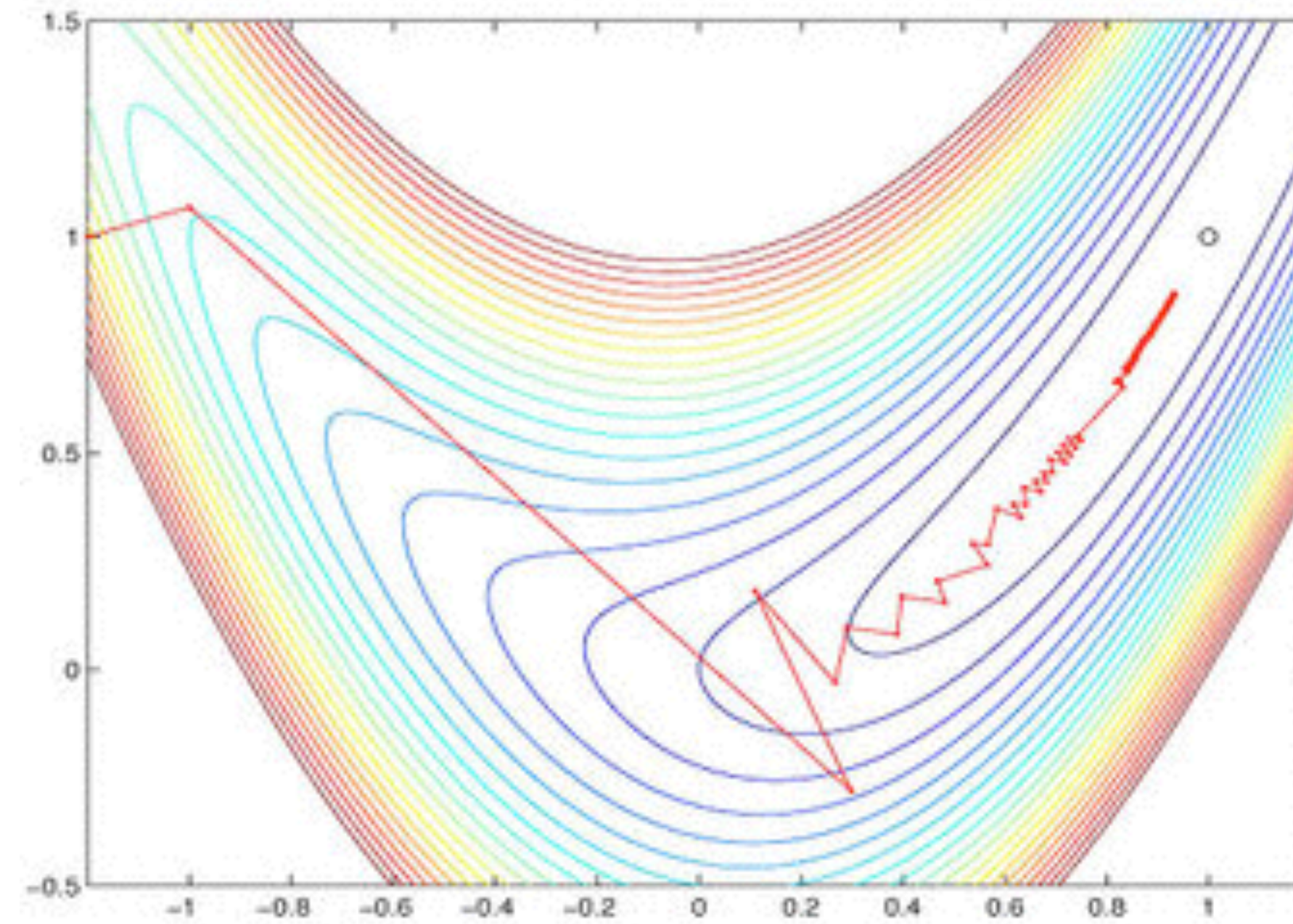




# deep network training



dataset



stochastic gradient descent

more on training later...

we use:

Deep Learning  
algorithms

multi-layer neural  
networks trained on  
large amounts of data



10 BREAKTHROUGH  
TECHNOLOGIES

### Deep Learning

With massive amounts of computational power, machines can now recognize objects and translate speech in real time. Artificial intelligence is finally getting smart.



### Temporary Social Media

Messages that quickly self-destruct could enhance the privacy of online communications and make people freer to be spontaneous.

# Our workhorse: deep learning algorithms



## Facebook's Quest to Build an Artificial Depends on This Guy

BY DANIELA HERNANDEZ 08.14.14 | 6:30 AM | PERMALINK

Share 470 Tweet 790 +1 164 in Share 160 Pin it



Yann LeCun, the new head of artificial intelligence at Facebook. © Josh Valc



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P2P LENDING + INVESTING OCTOBER 6-7, 2014 NEW YORK CITY

INDUSTRY TRENDSETTERS

JIM FRANCIS Union Bank



GEAR SCIENCE ENTERTAINMENT BUSINESS SECURITY DESIGN OP

The IBM Cloud can help businesses take the next step.

ENTERPRISE analytics google machine learning neural networks

Big Data

## Why a deep-learning genius left Google & joined Chinese tech st Baidu (interview)



## Google Hires Brains that Helped Supercharge Machine Learning

BY ROBERT MCMILLAN 03.13.13 | 6:30 AM | PERMALINK

Share 0 Tweet 1 +1 140 in Share 1 Pin it 3



Geoffrey Hinton (right), Alex Krizhevsky, and Ilya Sutskever (left) will do machine learning work

# Our workhorse: deep learning algorithms



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Updated AOL Privacy Policy ar

Europe acquisition Google DeepMind Technologies DeepMind

## Google Acquires Artificial Intelligence Startup DeepMind For More Than \$500M

Posted Jan 26, 2014 by Catherine Shu (@catherineshu)

75 Share 8.3k Share 678 Tweet 2,209

## Google Buys Jetpac: Former iOS App's Deep Learning Artificial Intelligence Useful for Maps, Advertising



WIRED

GEAR SCIENCE ENTERTAINMENT BUSINESS SECURITY DESIGN

The IBM Cloud is the cloud for business. Business on the cloud is made with IBM. Tap into the IBM Cloud

ENTERPRISE

deep learning Deepminds google Madbits

## Buying Madbits, Twitter Wants Image-Search Super Powers

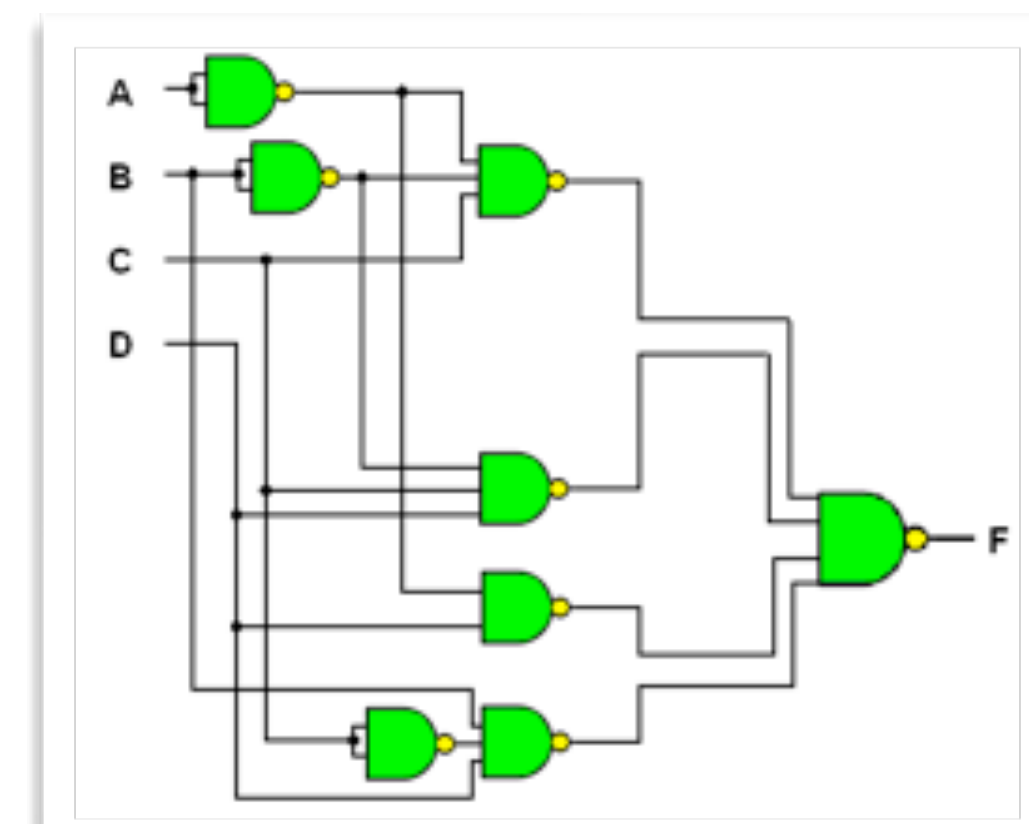
BY CADE METZ 07.30.14 | 4:57 PM | PERMALINK



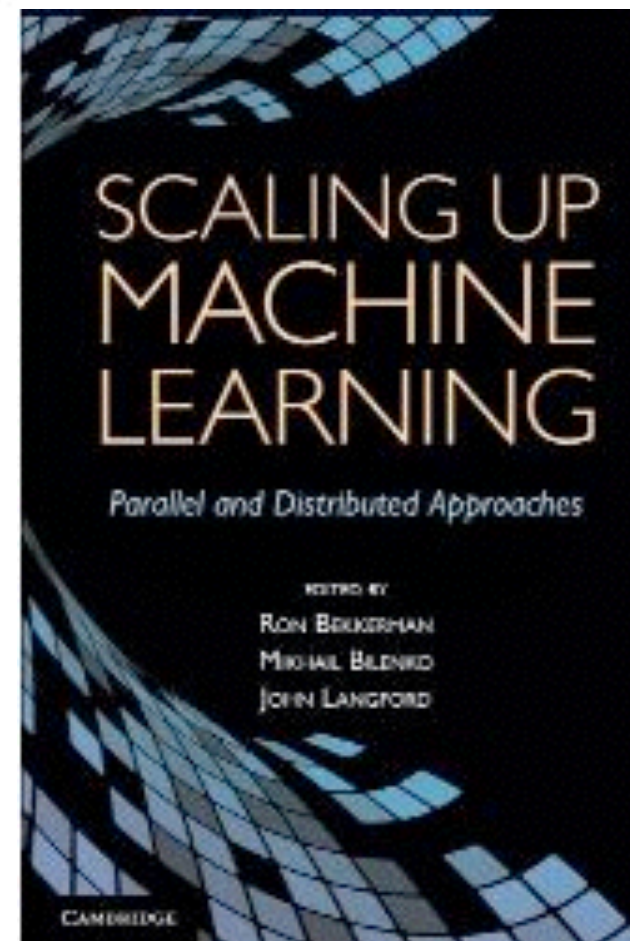
## 2: streaming processor architecture



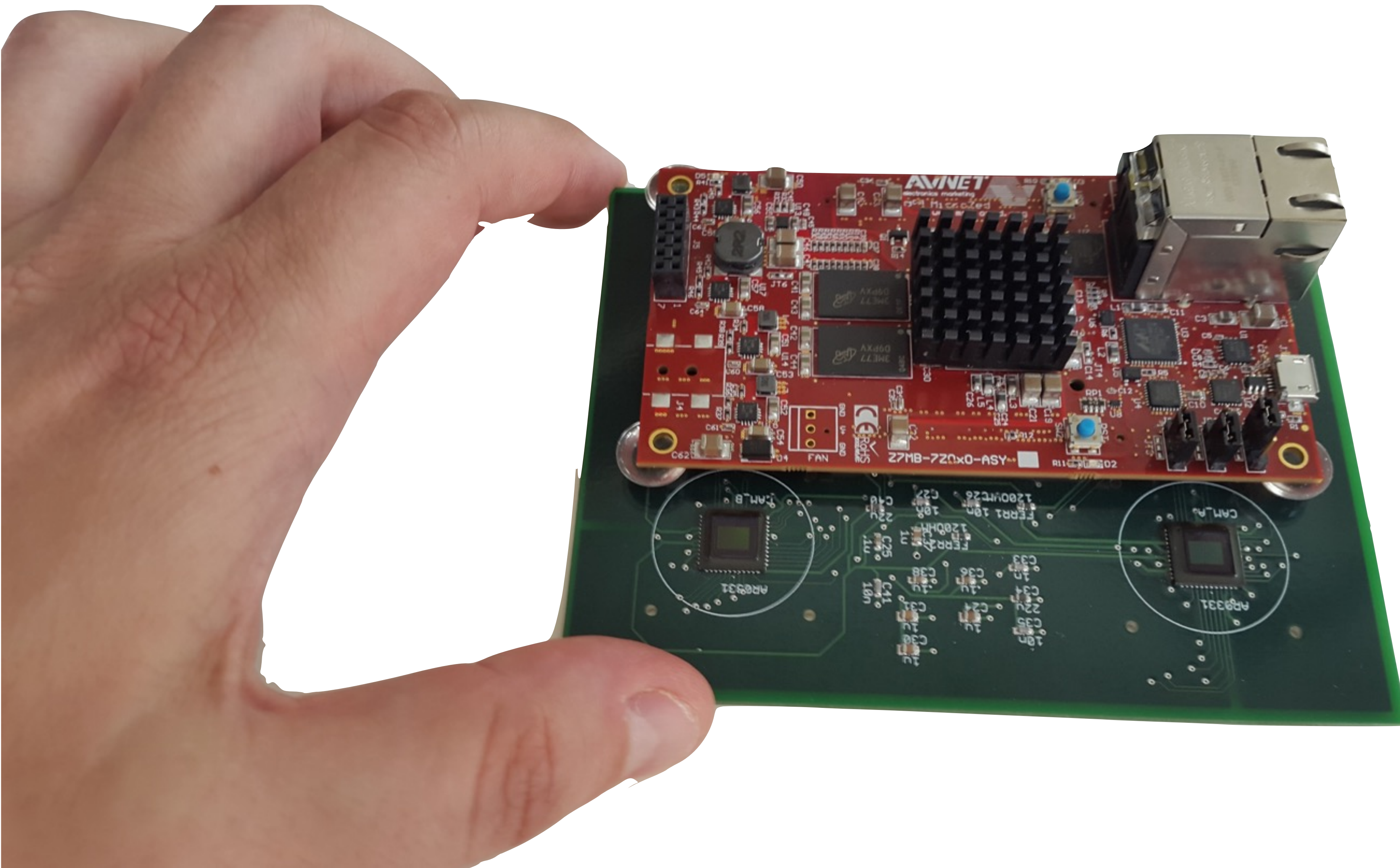
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```



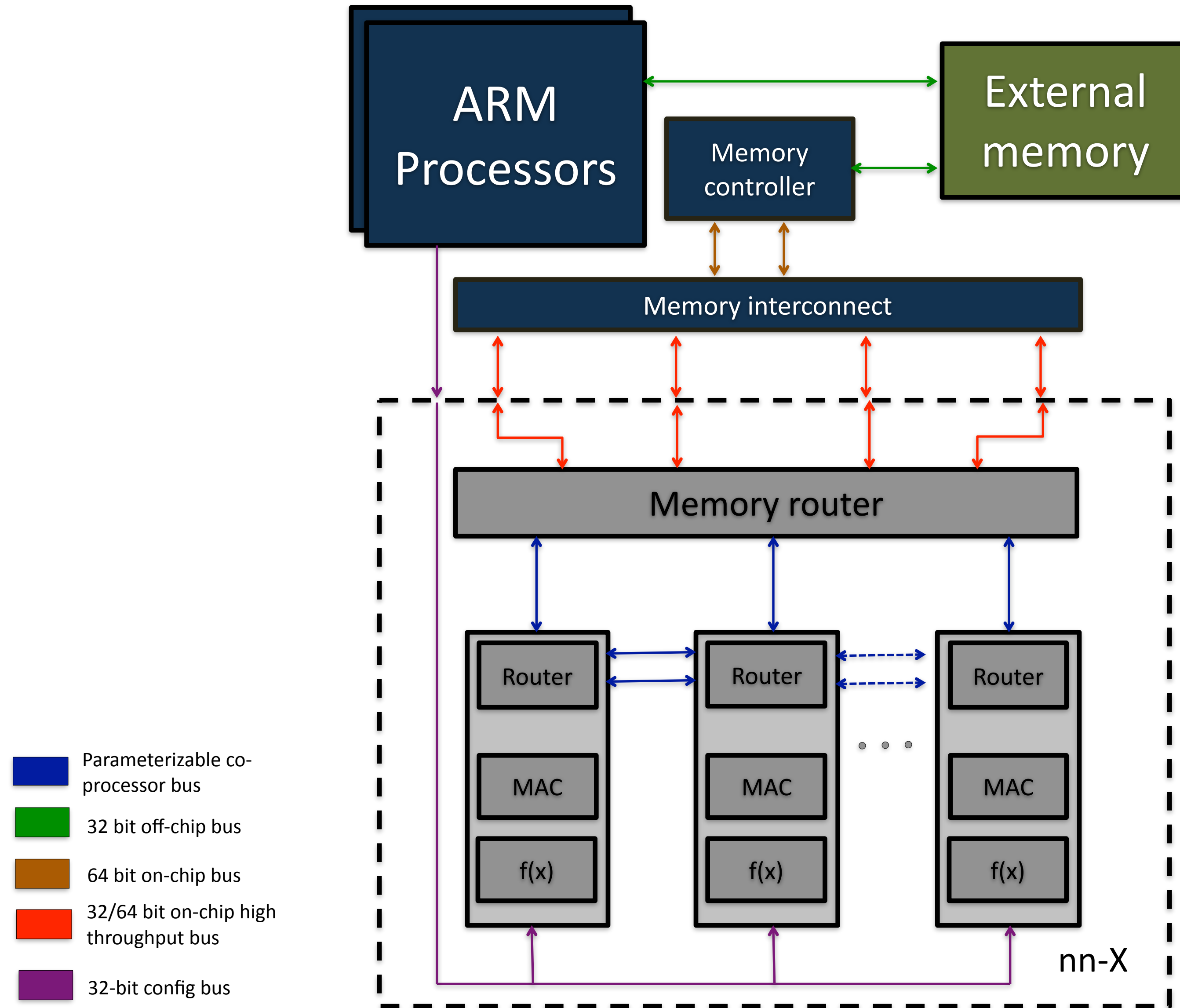
# conventional hardware



# our hardware: FPGA



# nn-X

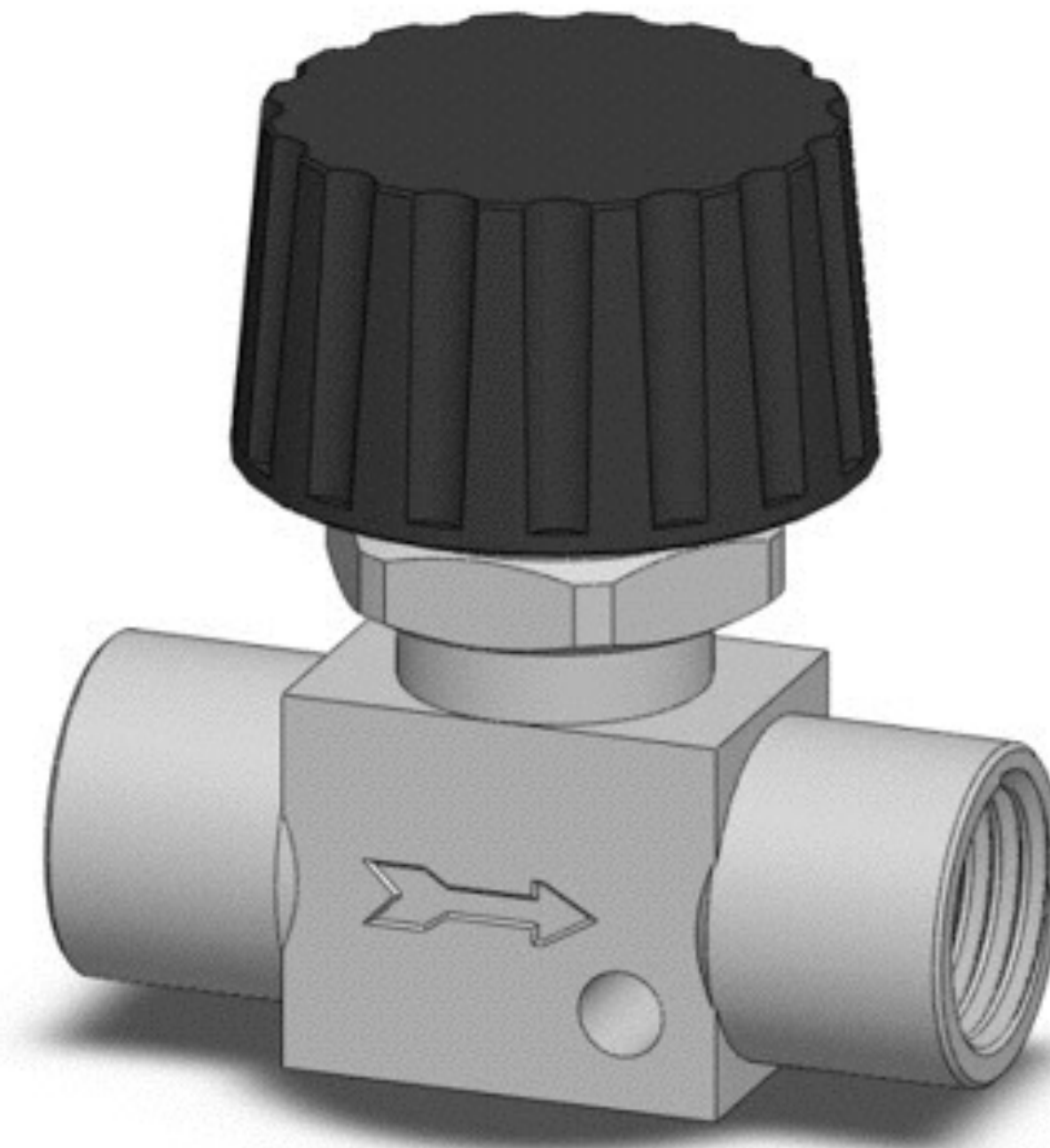




# streaming architecture



Parallel

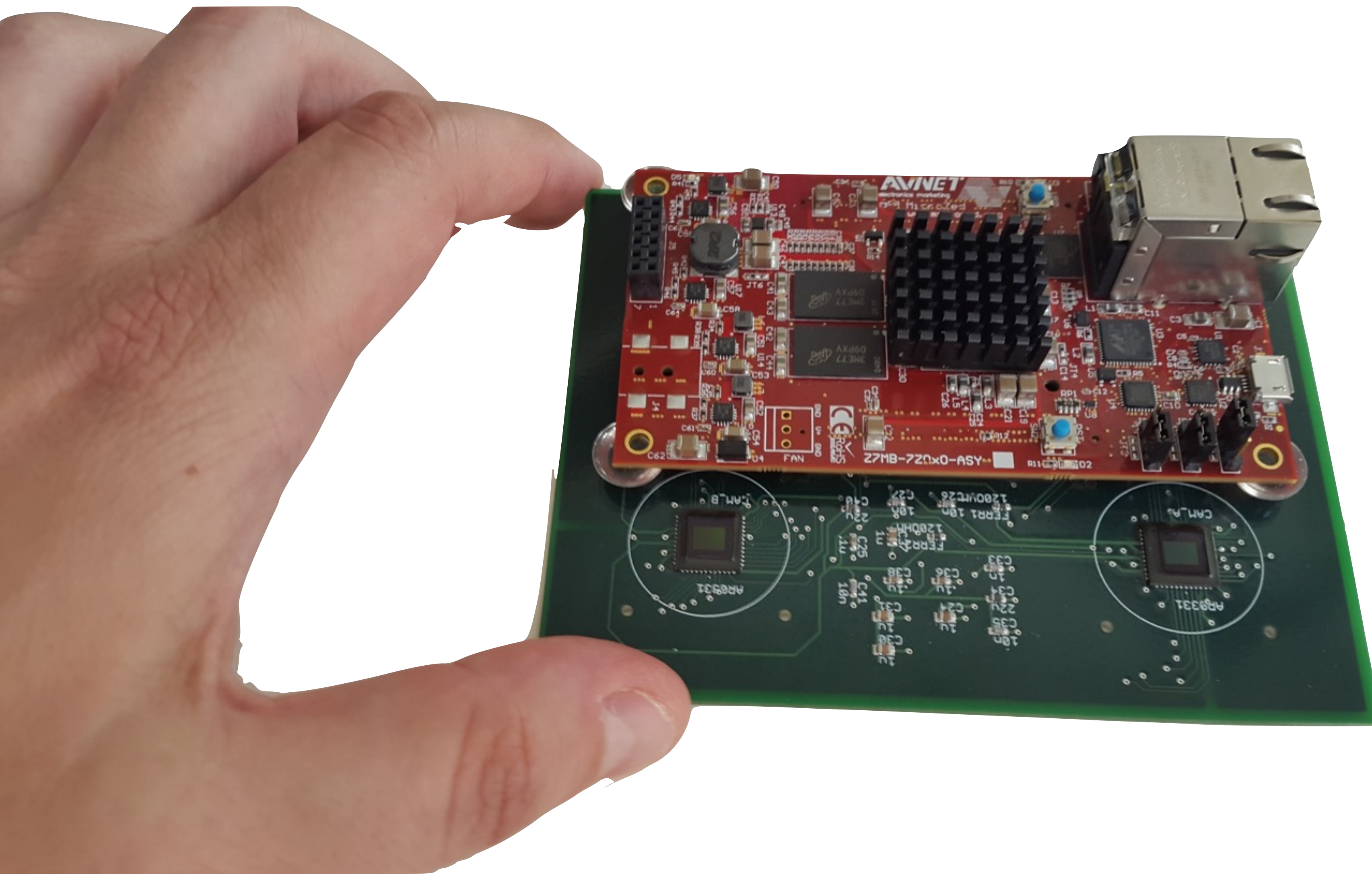


no flow control

even with a slow clock

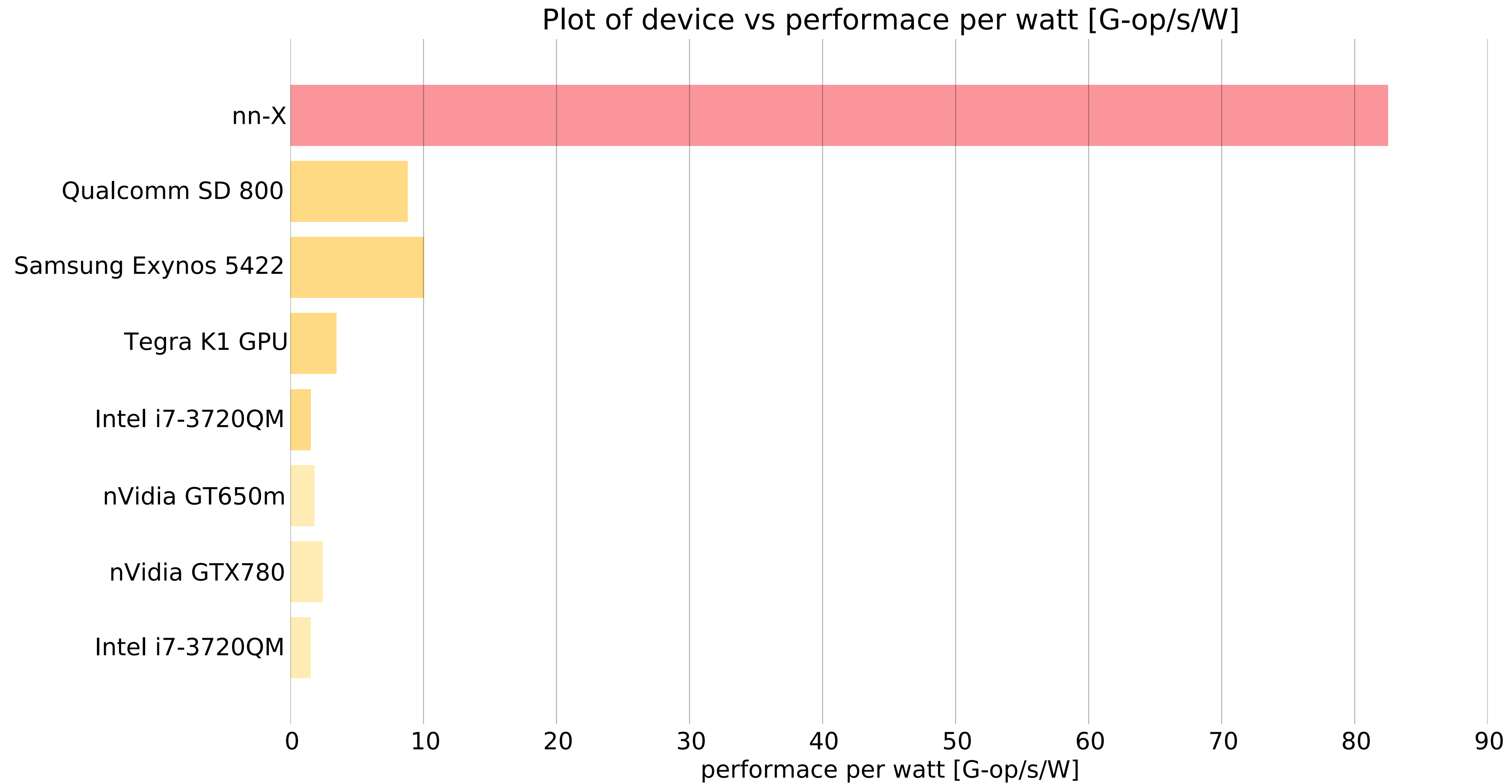
### 3: assemble vision systems

nn-X



> 360 G-ops/s on  
FPGA  
parallel: 8x  
operators collections  
includes neural  
classifier

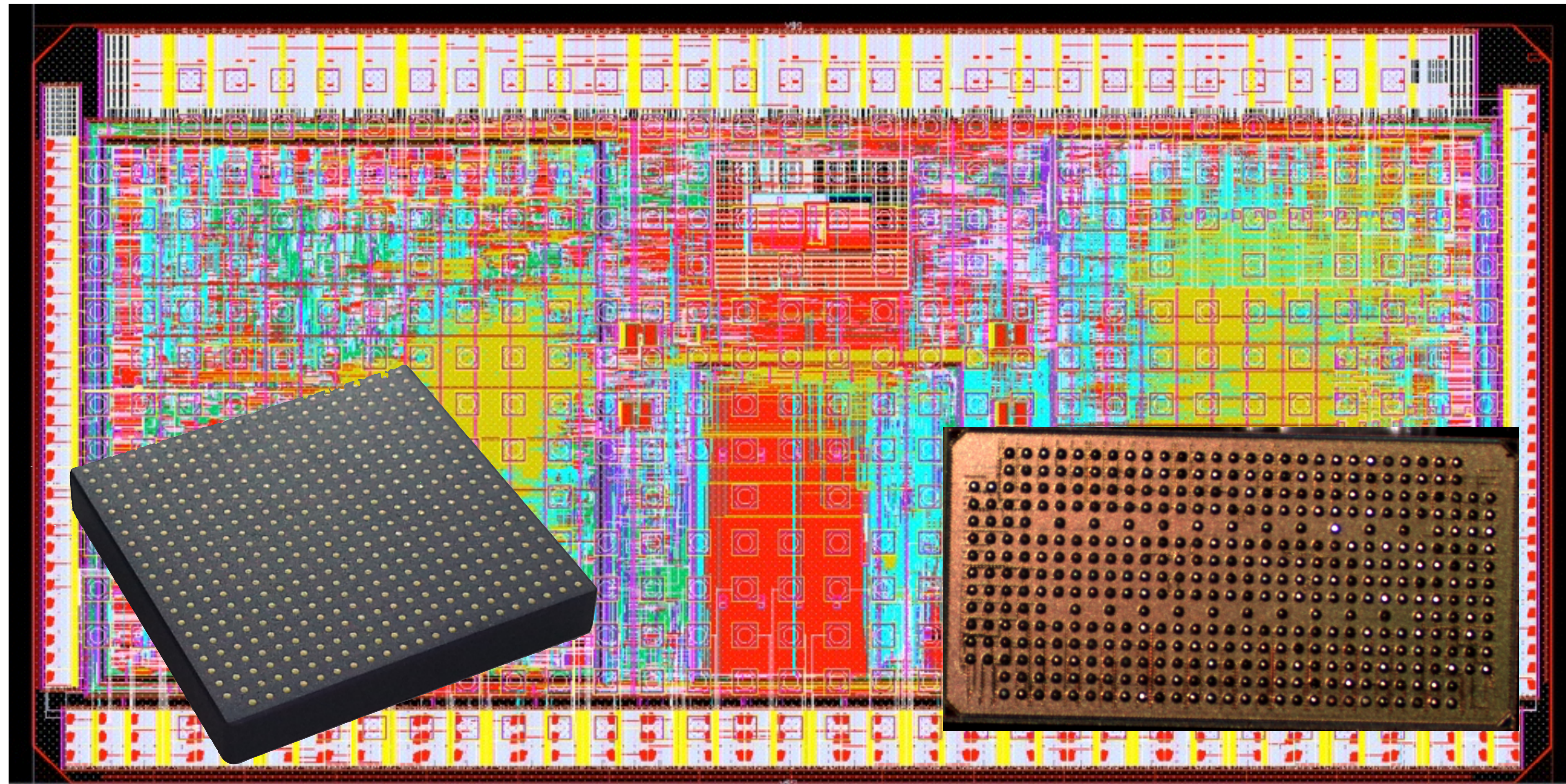
# nn-X performance



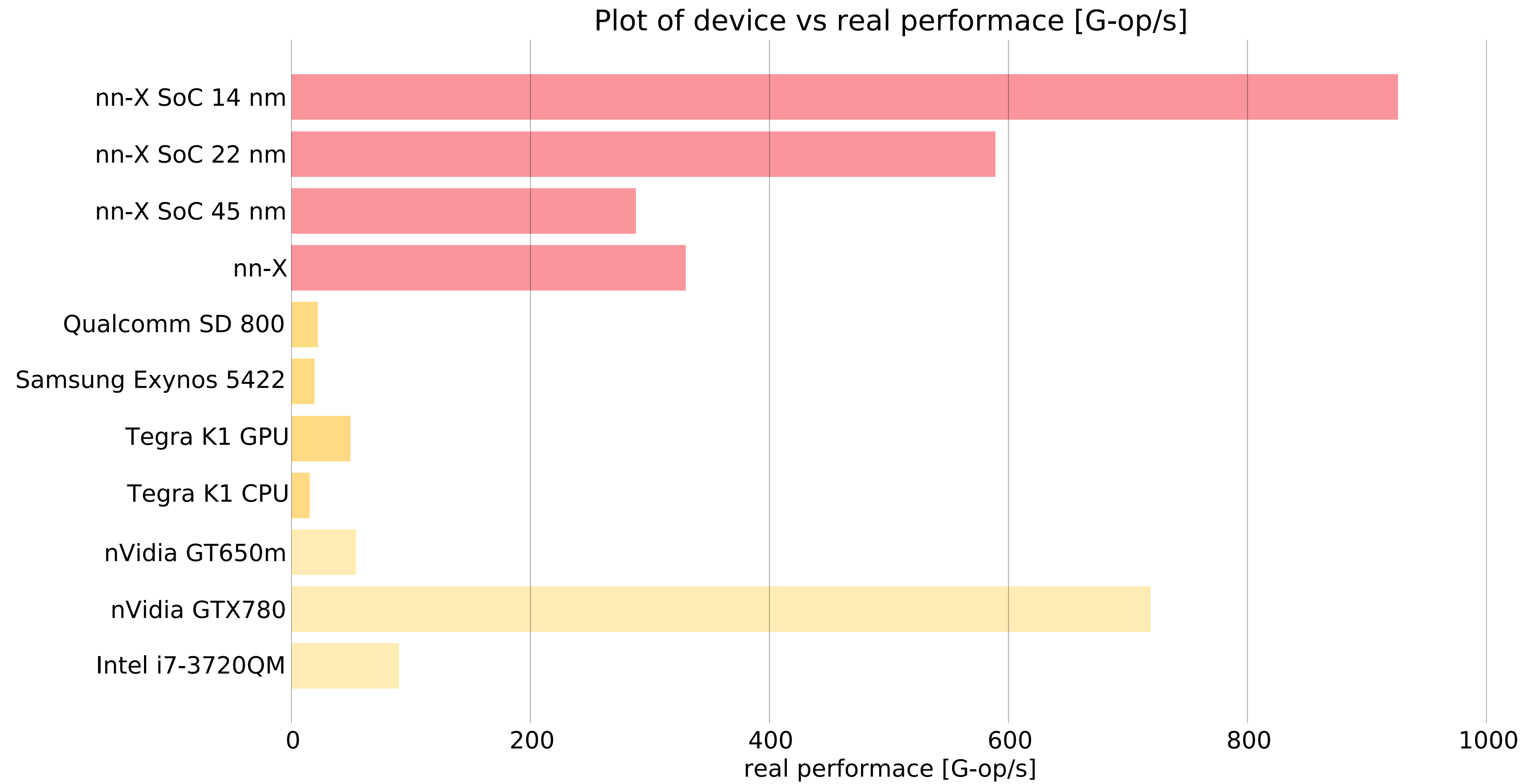
Performance and power consumption computed on a  
16x10x10 filter-bank over a 500x500 input image

# nn-X ASIC

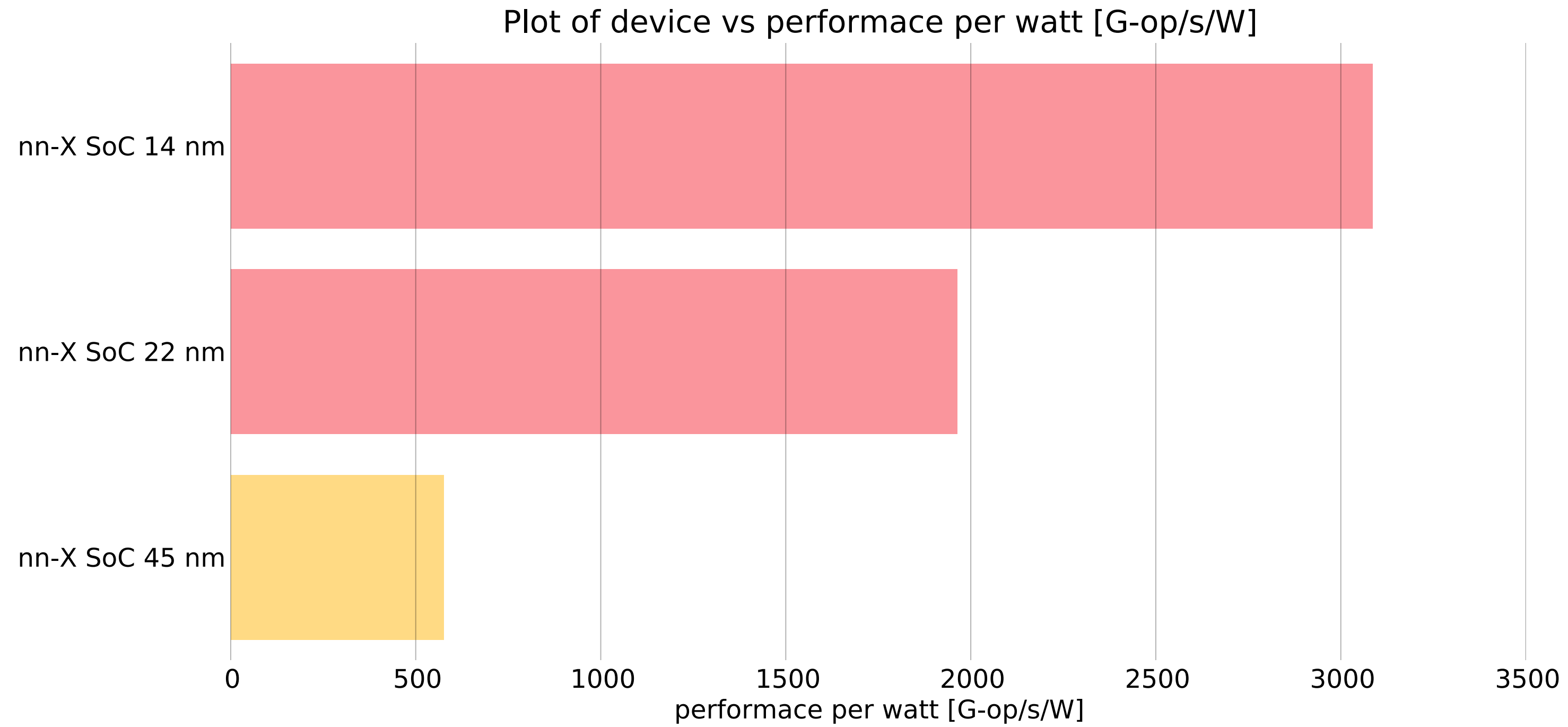
- IBM 45 nm high-density technology, 2.5x5mm
- 400MHz system clock, 200MHz CPU clock
- 4 convolvers, combiners, mappers
- Performance 12x better than FPGA (G-ops/Watt)



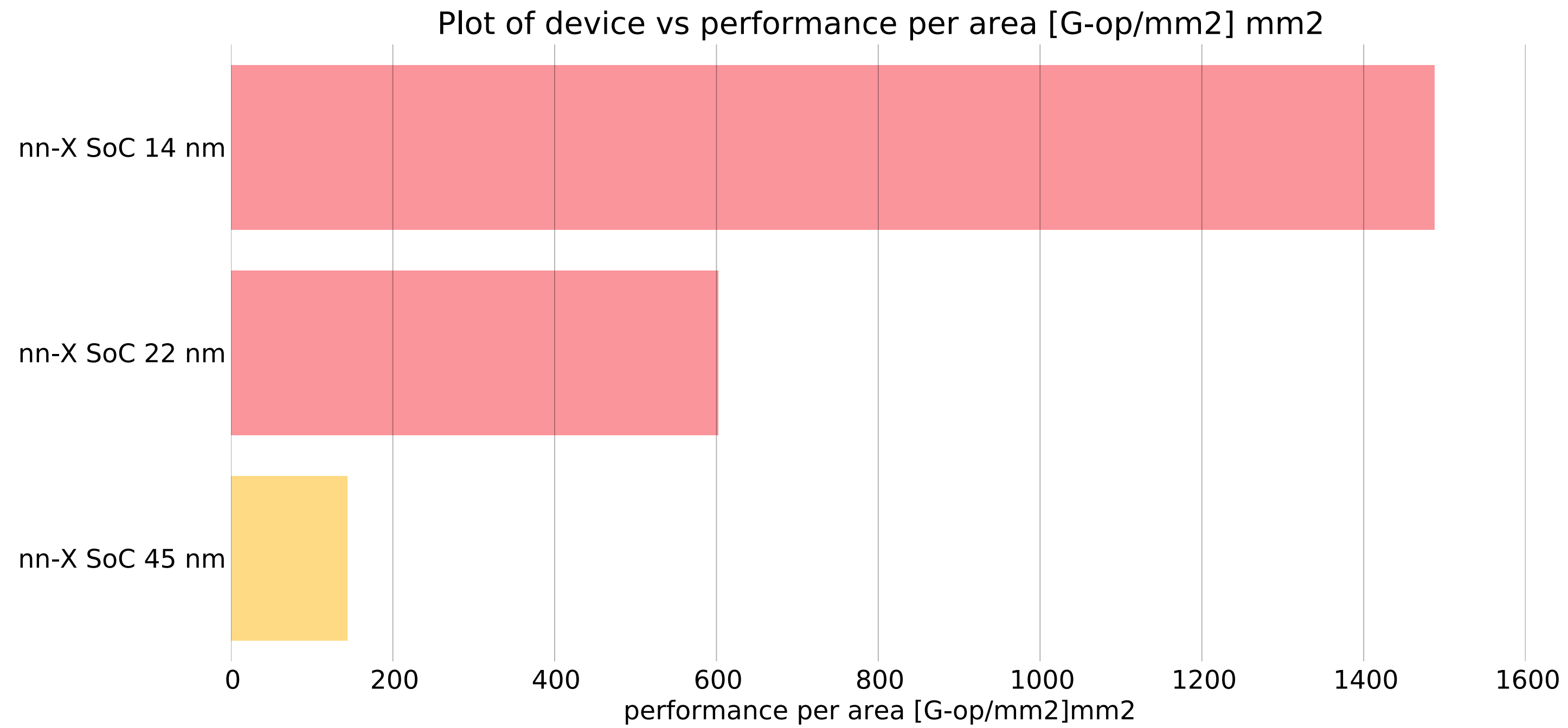
# nn-X performance



# nn-X performance



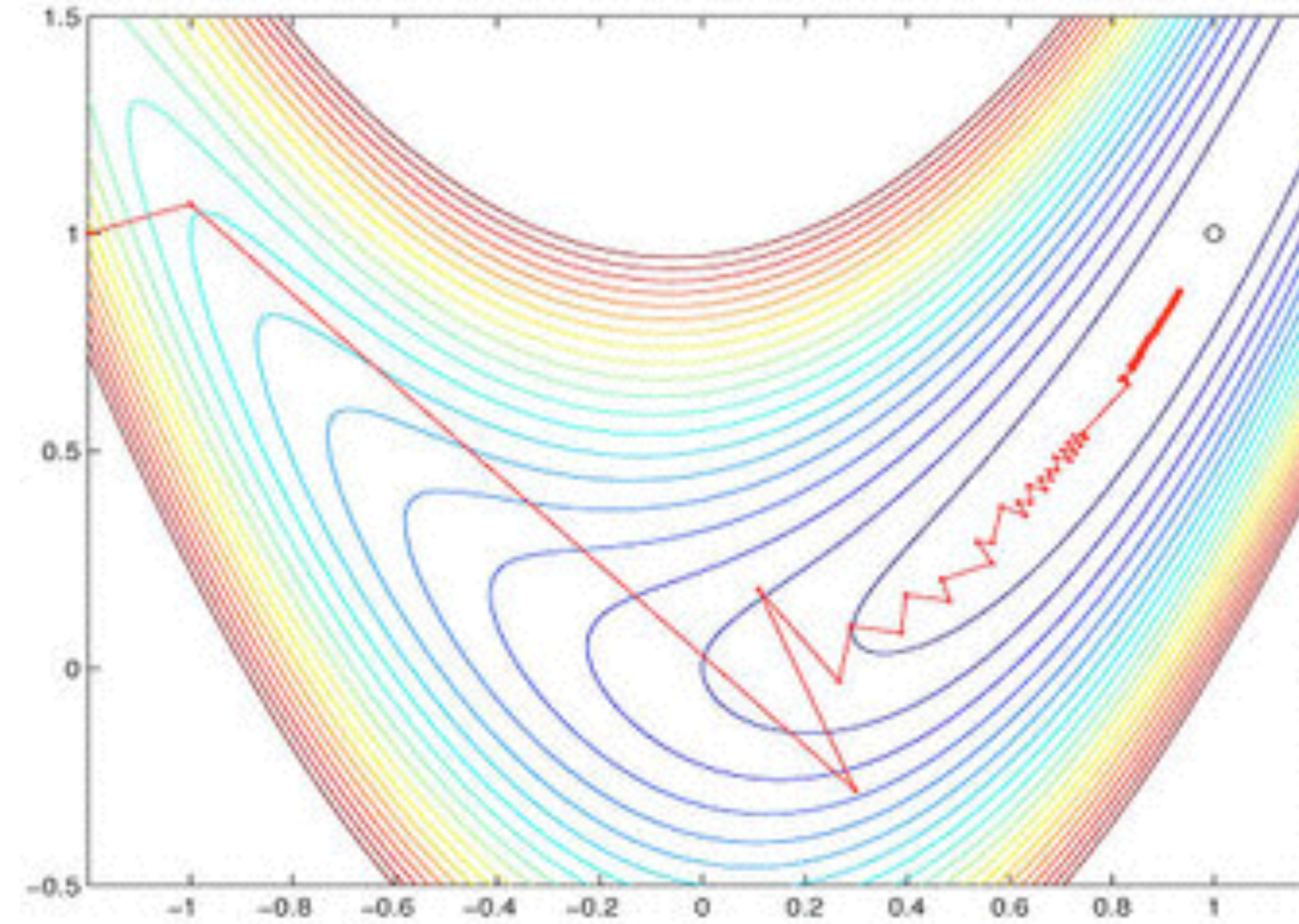
# nn-X performance



# 4: training deep networks



dataset



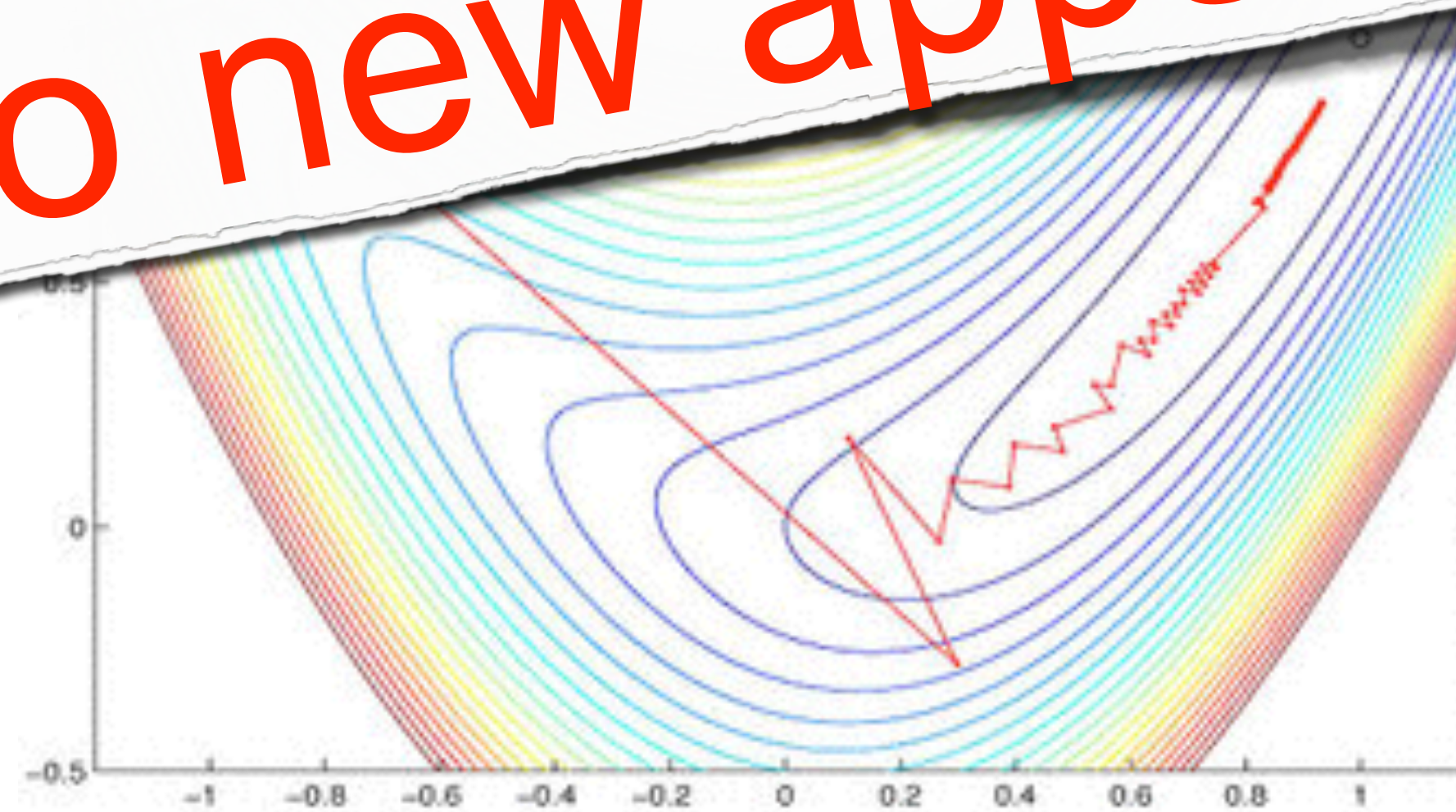
stochastic gradient descent



expensive, not scalable for  
videos, imprecise, not  
“portable” to new apps

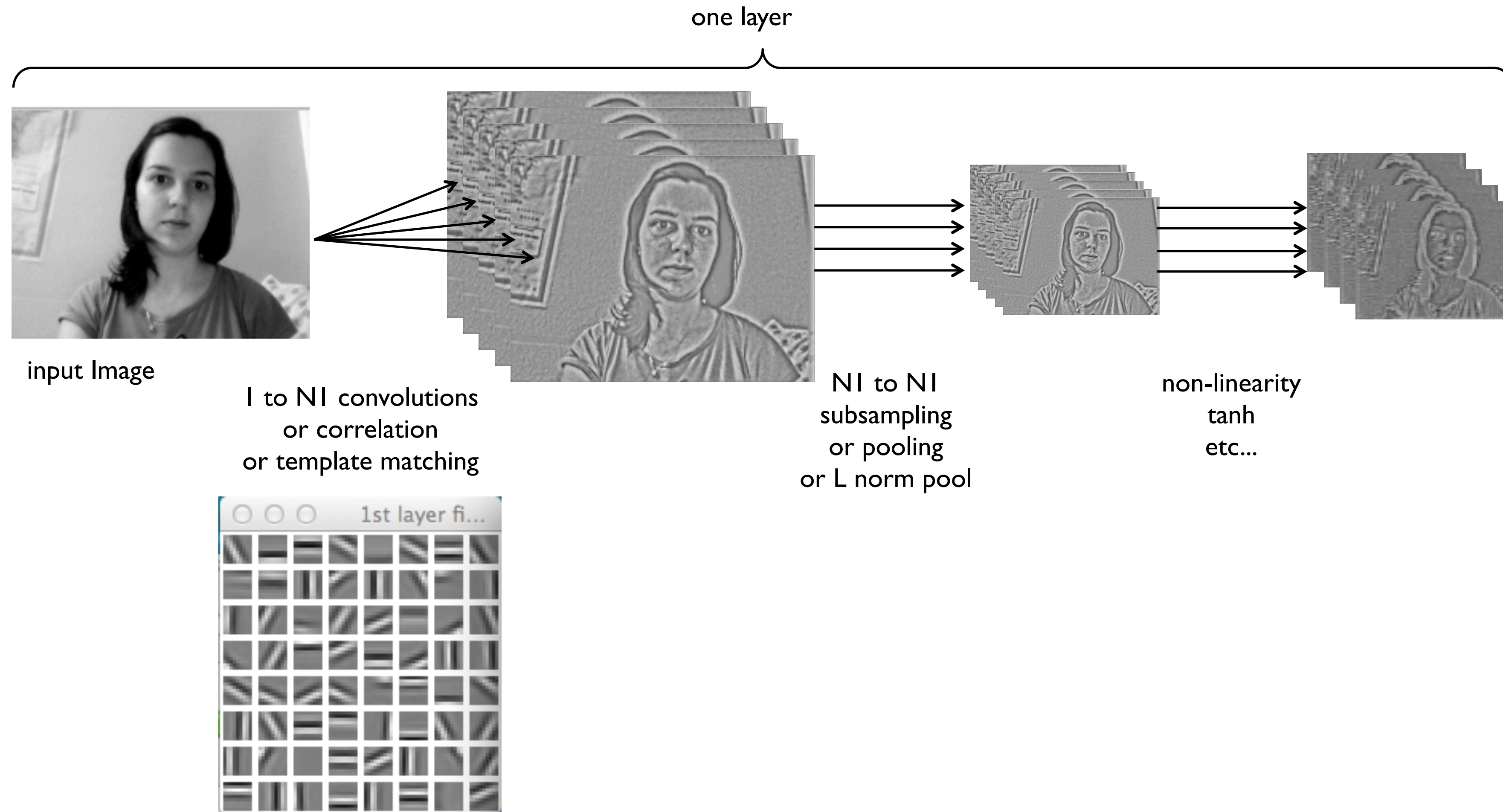


dataset



stochastic gradient descent

# deep networks training



It is all about:  
[filters], connections!!!

# Hebbian Learning

# Clustering Learning



fire together  
wire together

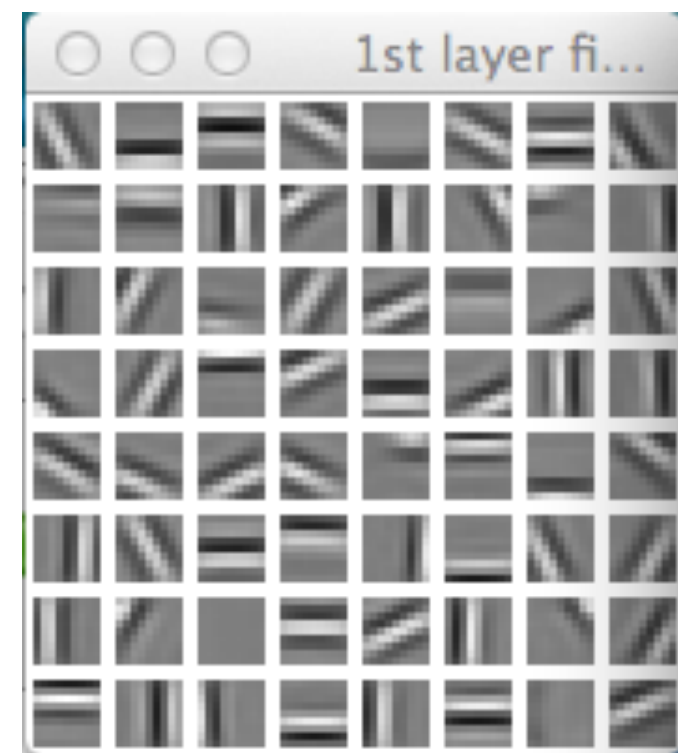
respond  
together  
cluster together

compatible with STDP learning

# clustering learning



random patches of images



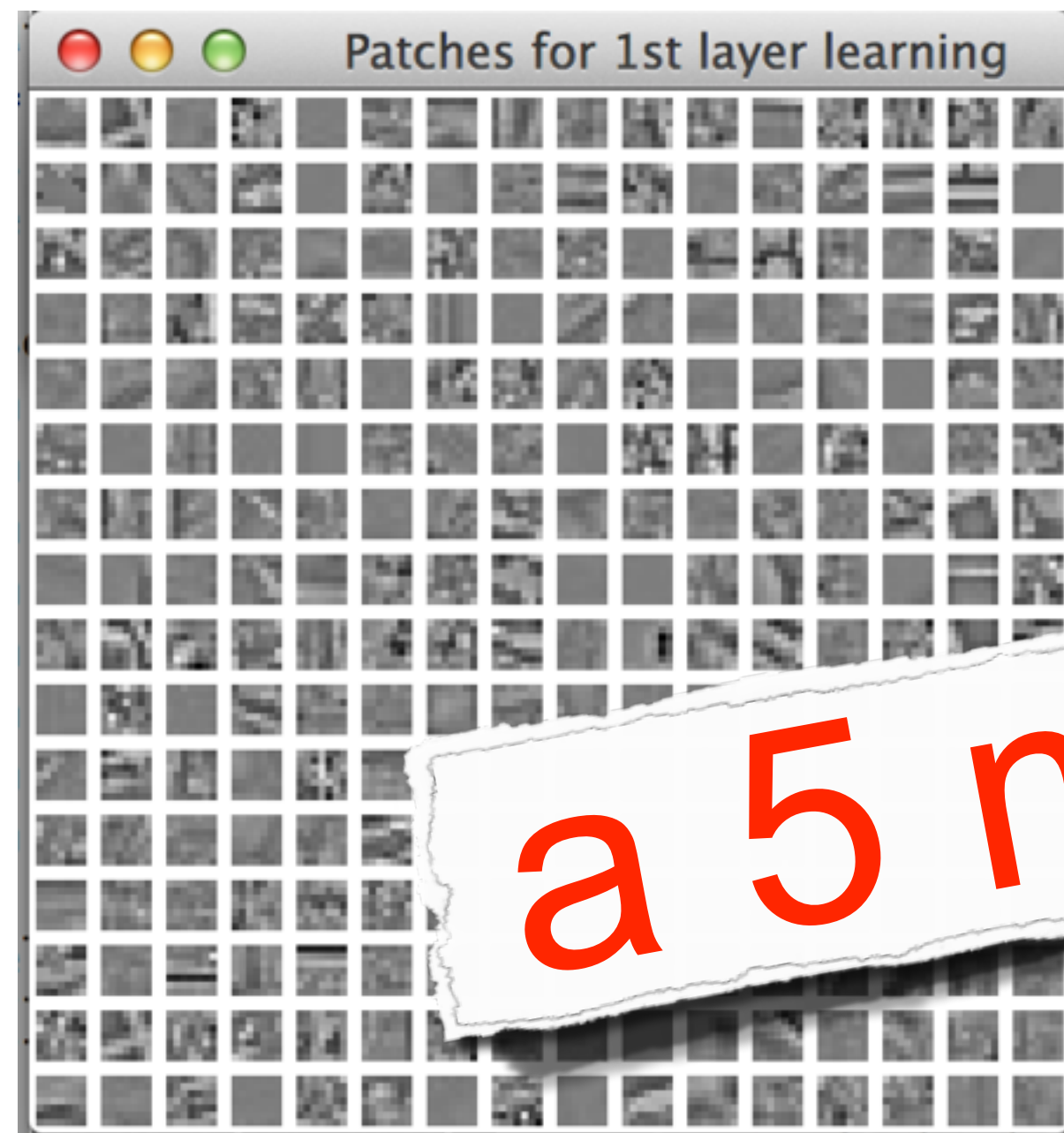
clustered means



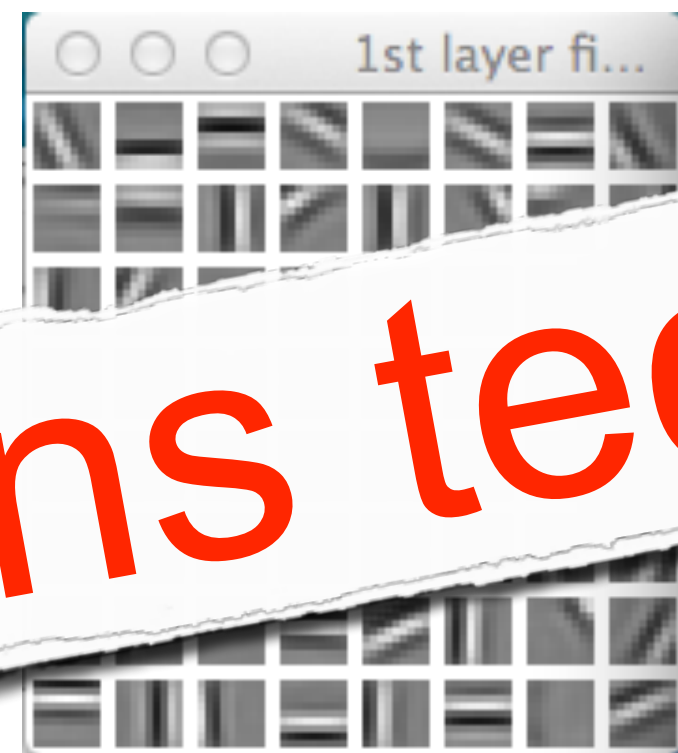
use filters as  
parameters for  
this layer



# clustering learning



random patches of images

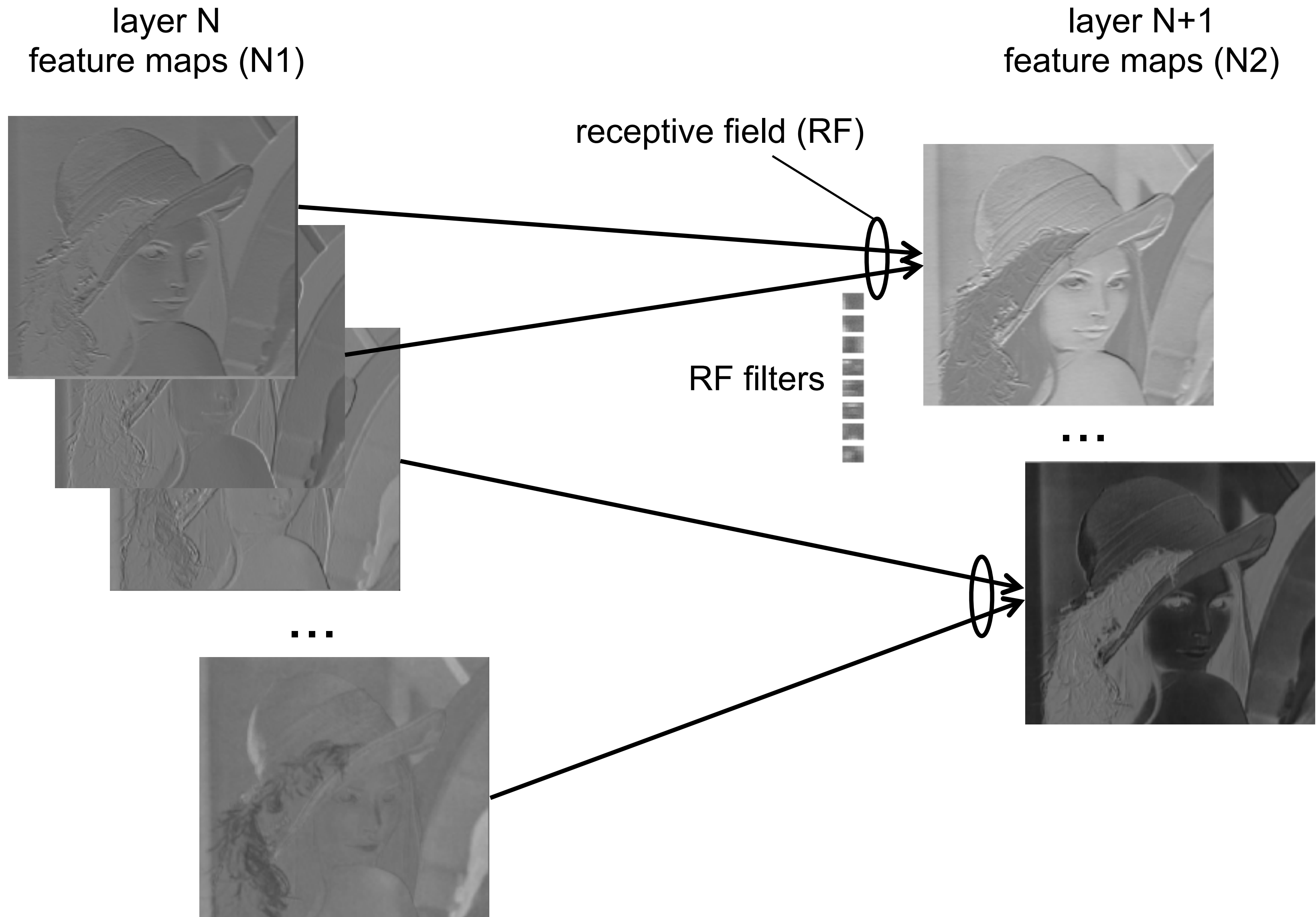


clustered means

**a 5 mins technique!**

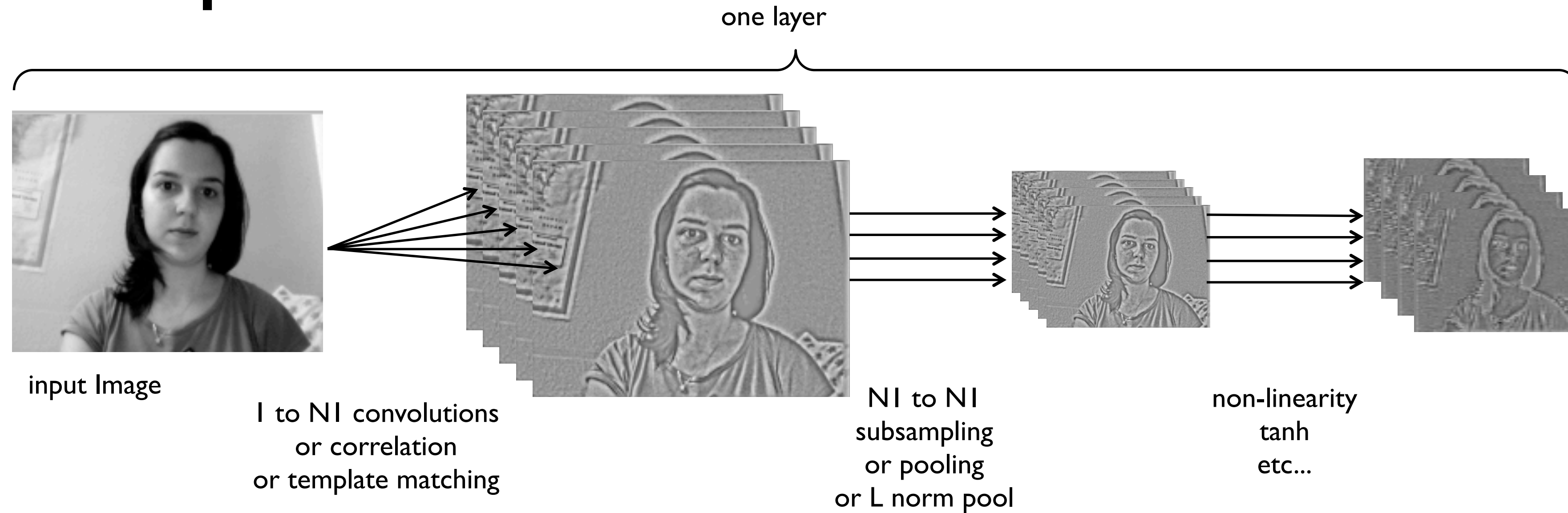
use filters as parameters for this layer





**main idea:** learn to CLUSTER inputs based on co-occurrence of features

# deep networks



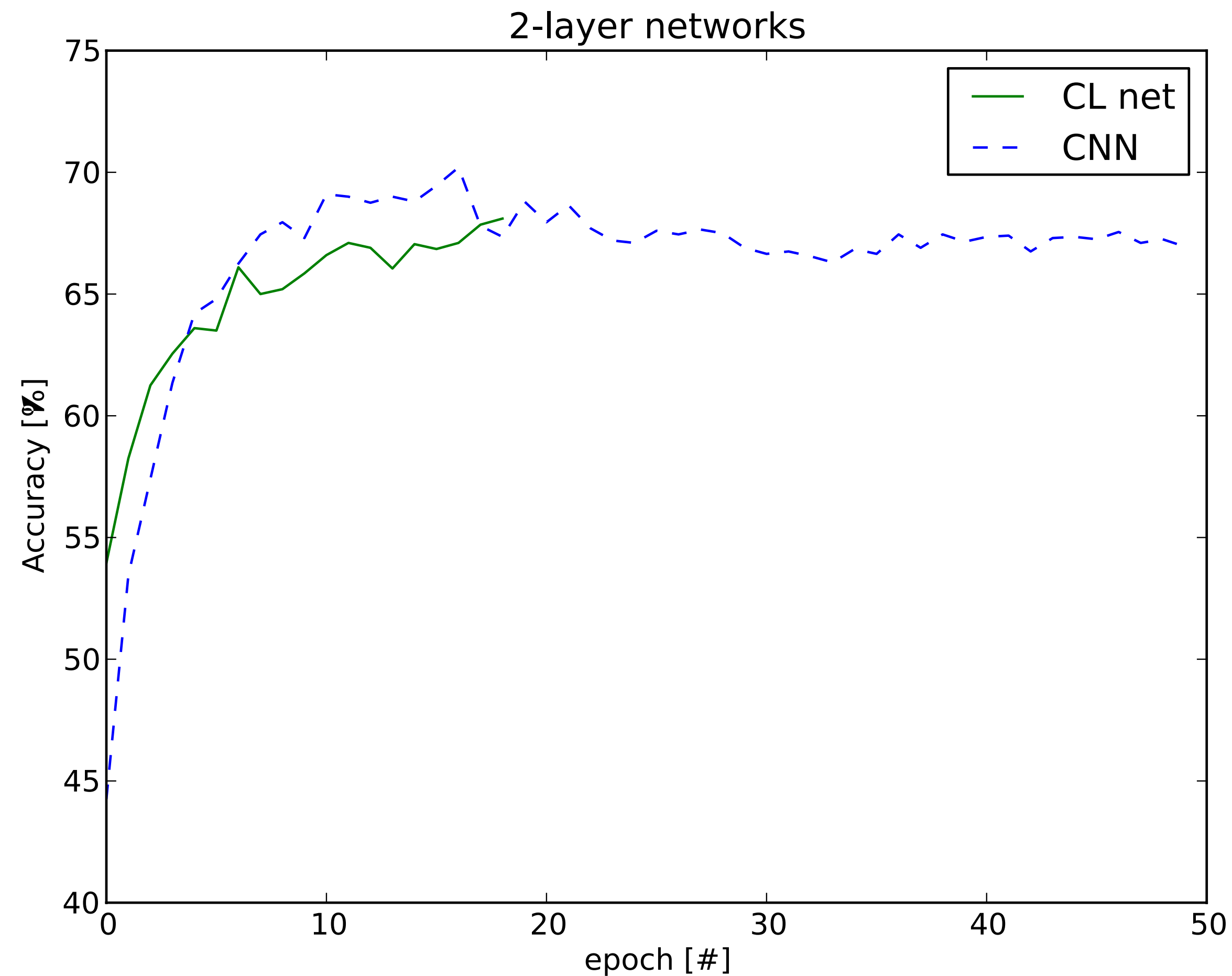
**Multiple layers of deep network:**

**Repeat for each layer:**

- 1- sample output of previous layer (new input)
- 2- cluster these inputs = filters
- 3- use filters to generate outputs

# clustering learning

as well as standard convnet/CNN!!!



CIFAR10

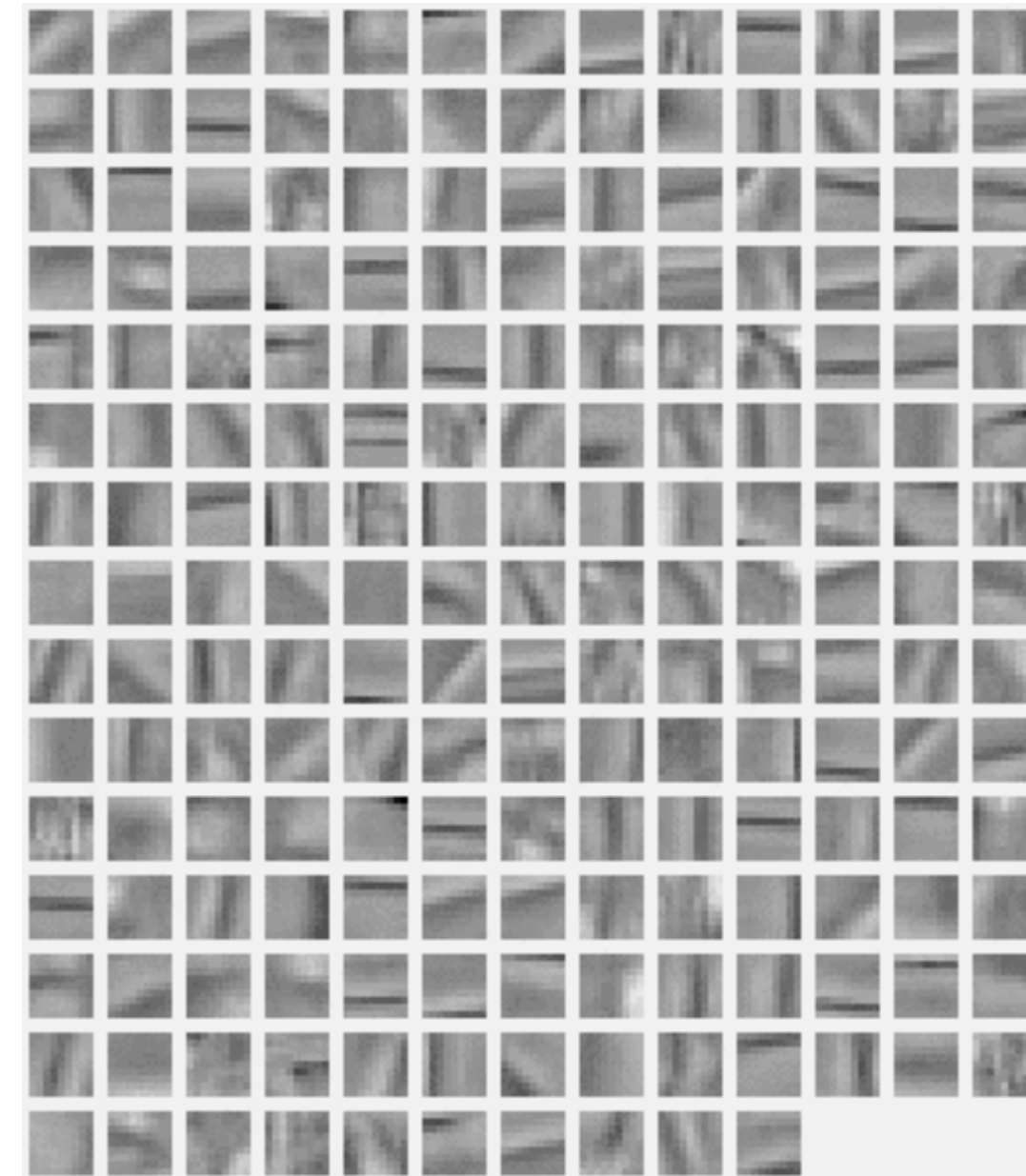


# clustering learning: motion filters

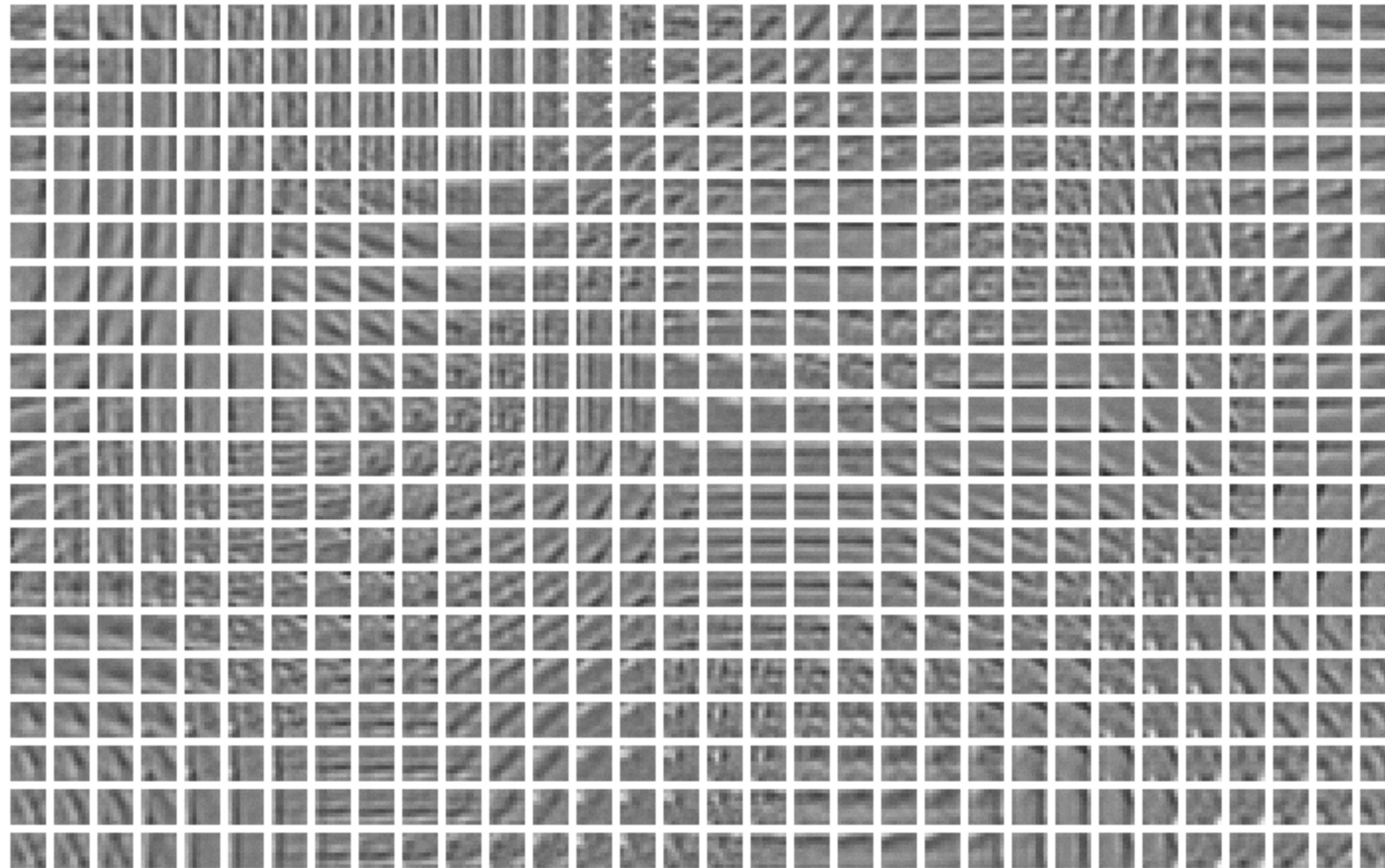


same patch location for multiple frames

run k-means  
on group of  
patches

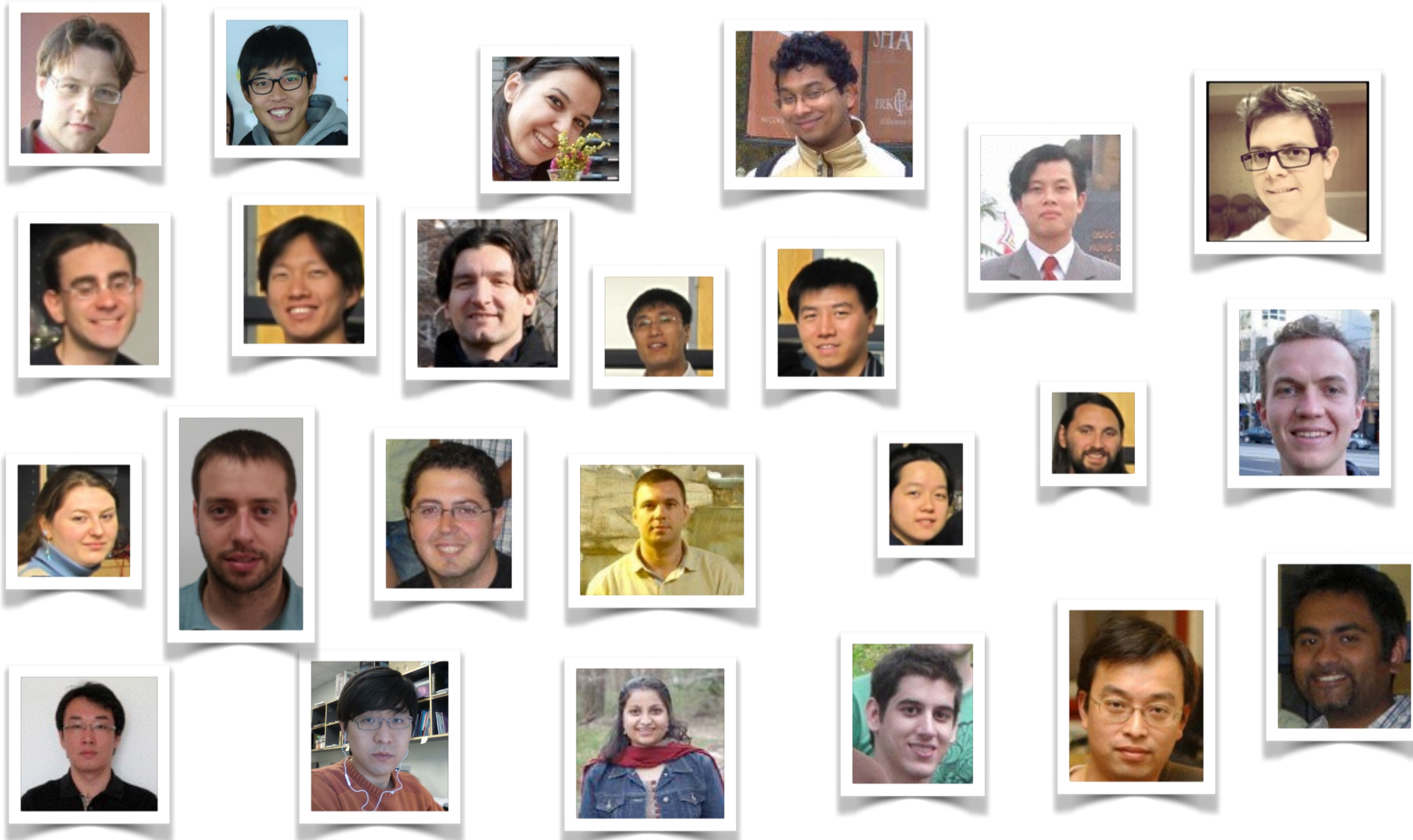


# clustering learning: topographic filters



# Summary of results

- **state-of-the-art architecture** for complex understanding of the environment
- fully **programmable**
- customized for **neural networks**, is extendable to any **state-of-the-art algorithms**
- **scalable** to larger networks and applications
- To do: extend it to **learning from videos**, unsupervised multi-layer learning





# Thanksgiving

## E-Lab current members:

Aysegul Dundar  
Vinayak Gokhale  
Jonhoon Jin  
Alfredo Canziani  
Berin Martini

and with:  
Clement Farabet

## Visitors:

Jose Carrasco  
Angelo Rottigni  
Alejandro Linares-Barranco  
Rafael Paz

## Past members:

Zhengming Fu  
Pujitha Weerakoon  
Shoushun Chen  
Farah Laiwalla  
Huang Chenxi  
Dongsoo Kim  
Andrew Kunil Choe  
Selcuk Talay  
Polina Askelrod  
Faye Zhao  
Ifigeneia Derekli  
Hazael Montanaro  
Darko Jelaca  
Jonathan McMillan  
Evan JoonHuyk Park  
Wei Tang  
Phi-Hung Pham  
Jordan Bates



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