

Photonic Reservoir Computing with coupled SOAs

Kristof Vandoorne, Wouter Dierckx, David Verstraete,
Benjamin Schrauwen, Roel Baets, Peter Bienstman
and Jan Van Campenhout

OSC 2008: August 26

84520

420665

p6SY

580150

5d9dec

21407

QbQc fV

TJ2LE

63BdmL

1KHA

mejZ

28484

514207

72PYKR

ghwgr

JXR52M7

motk

ki n v r

C6Q5

RAHNU

yS5P

cavion
28rt2

VTE9HX

Gba7fE

63v6L

N9BM

Gqsz2R

zTR

6ct3zv

KeerK

WUE

j3ZG

jY2P

K8ztX

mzWYpz

idnzl

8brTAT

Intelligence is all around us,
but so far limited for computers

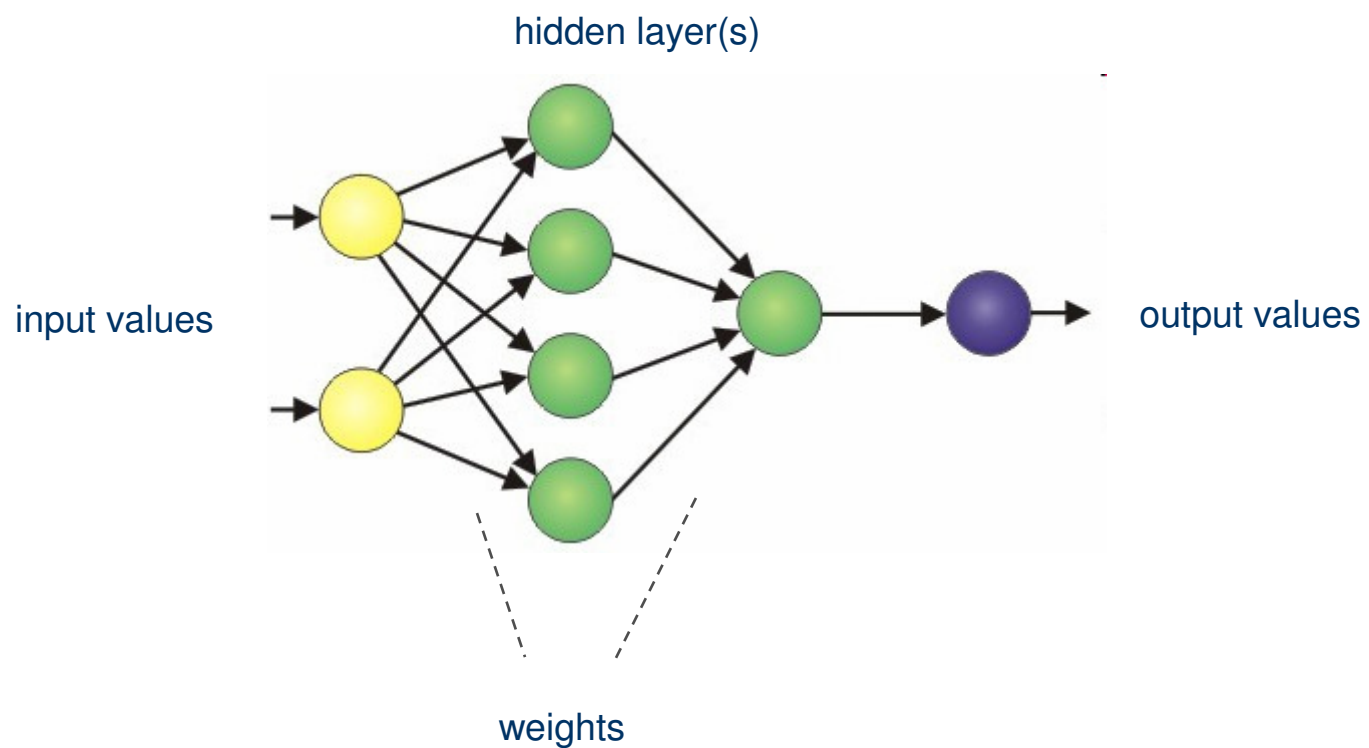
Confirmation

* Verification Code:

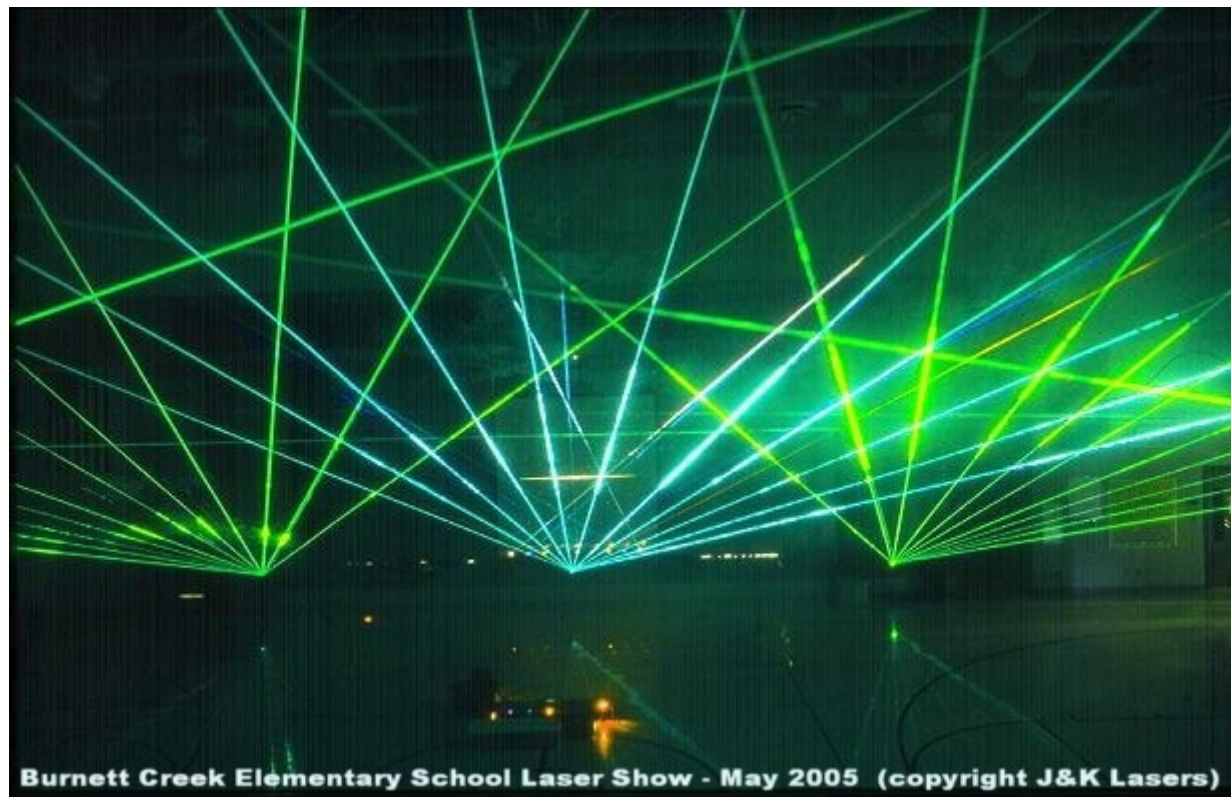
h Y^t e n

Enter case sensitive Verification Code and click **Submit Request**.

Reservoir Computing is a new approach
coming from the field of artificial neural networks



We use light because it is potentially faster and more power efficient than the present software implementations

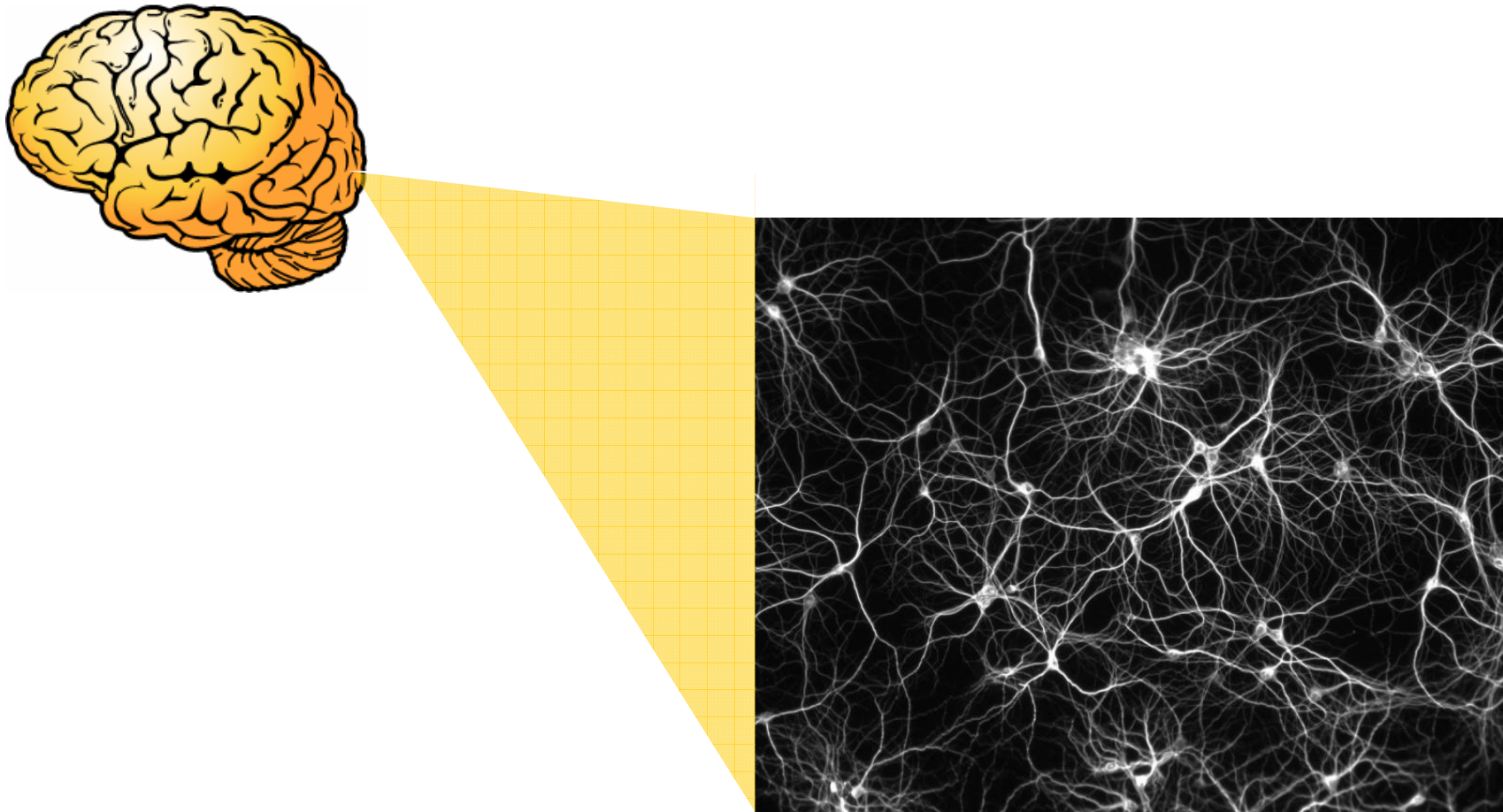


The concept works...

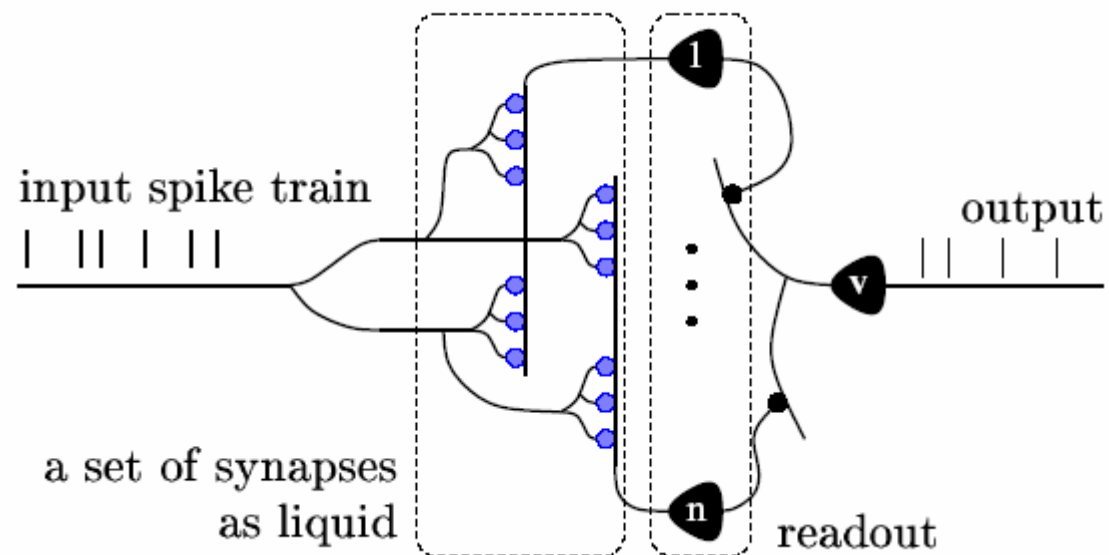
(in simulation)

1. Reservoir Computing
2. How to do it with photonics
3. Simulation results
4. Future

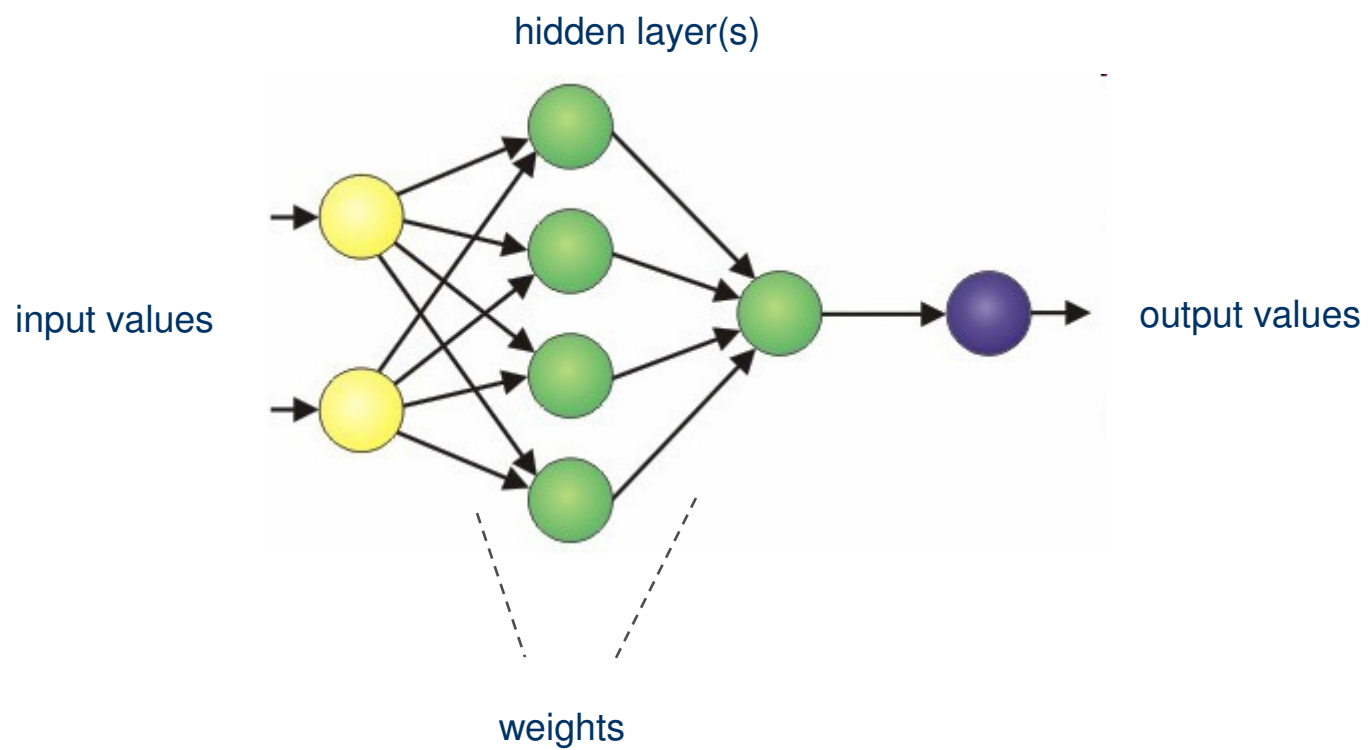
Learn from the best: the brain



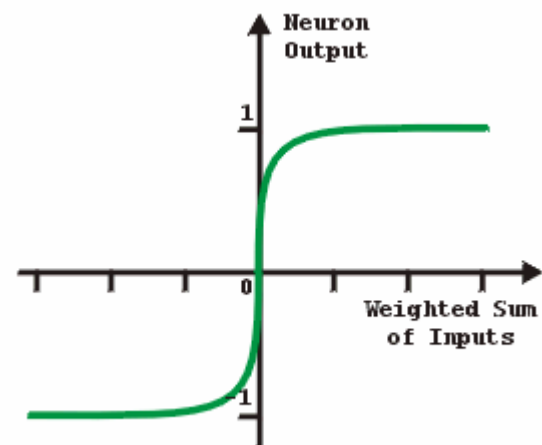
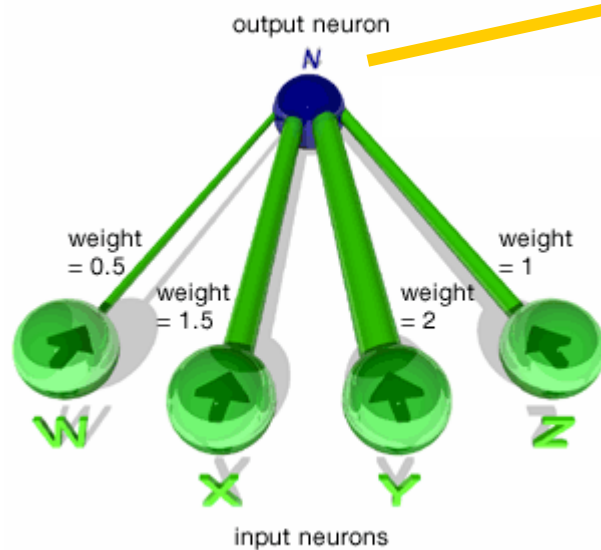
And mimic its behaviour



Or simplify it, by removing feedback

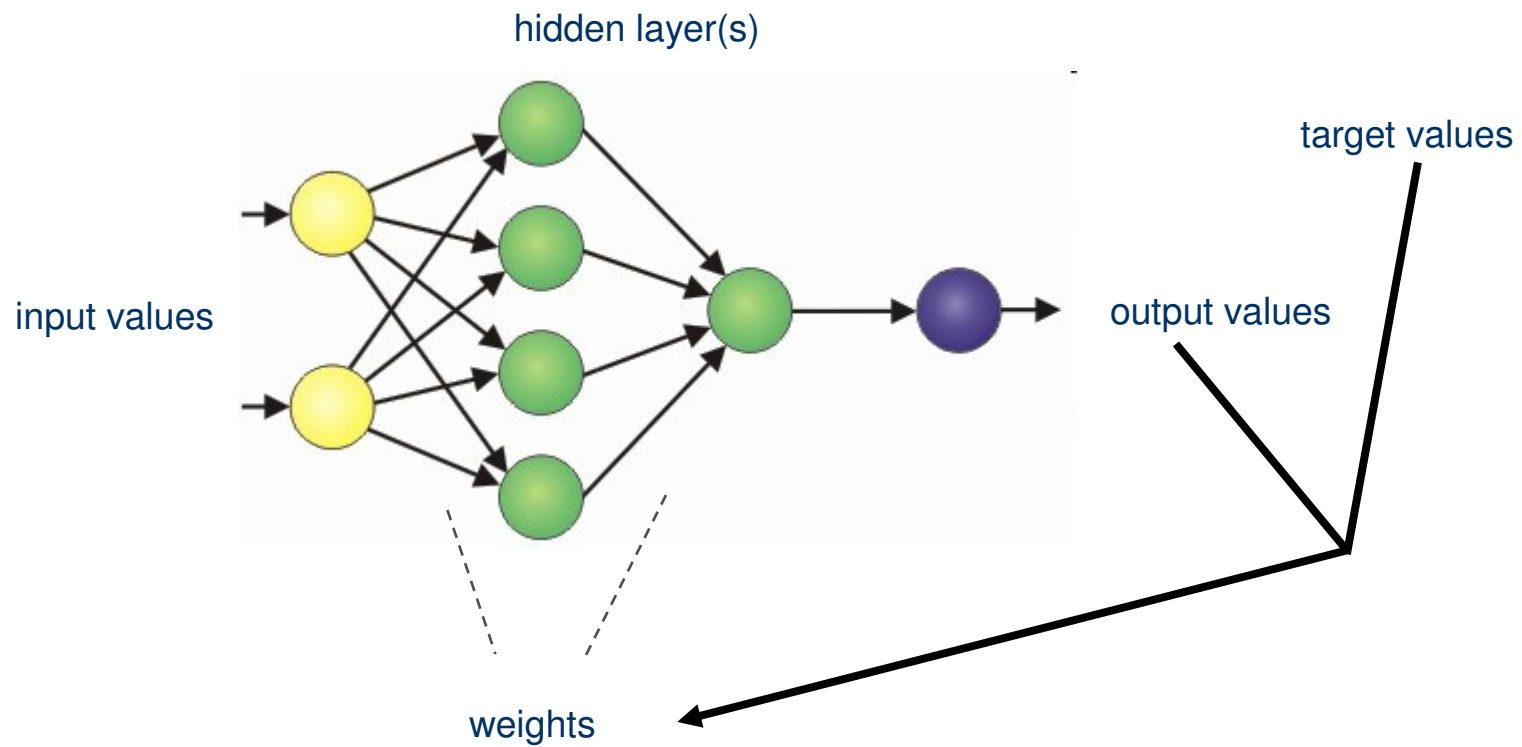


Each node applies a nonlinear function over the weighed sum of its inputs

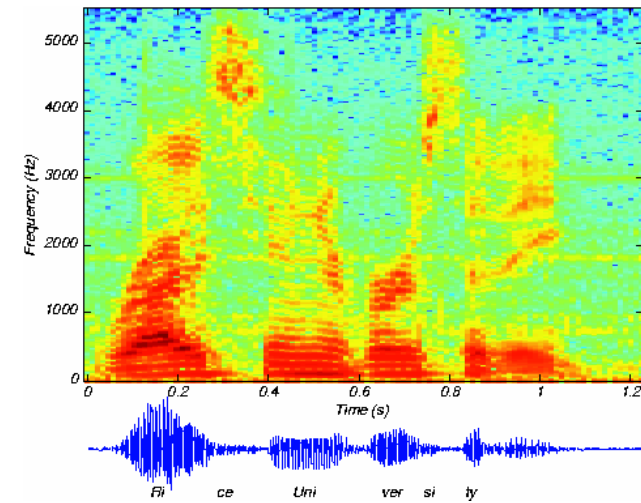
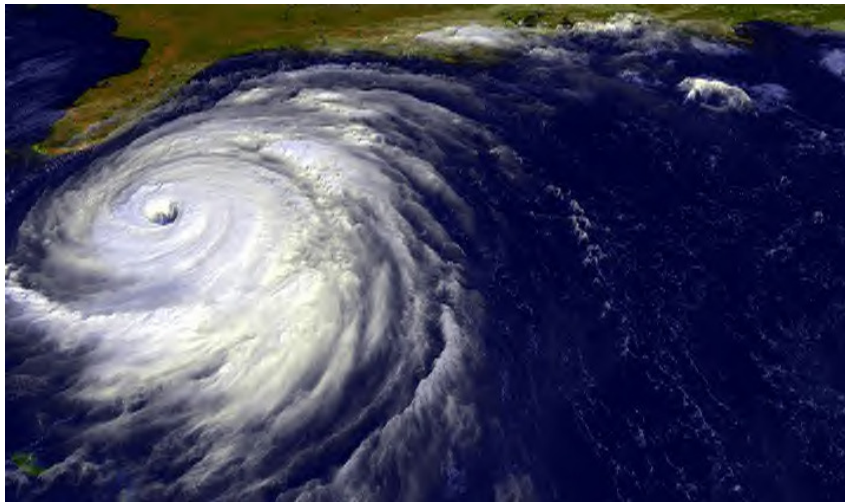


Pictures from <http://logicalgenetics.com>

The network is trained by tuning the weights of the connections



But many of the real-world problems are temporal,
calling for some memory in our system
(e.g. through feedback)



Networks with feedback are hard to train.

Reservoir Computing offers an elegant solution by splitting up the network

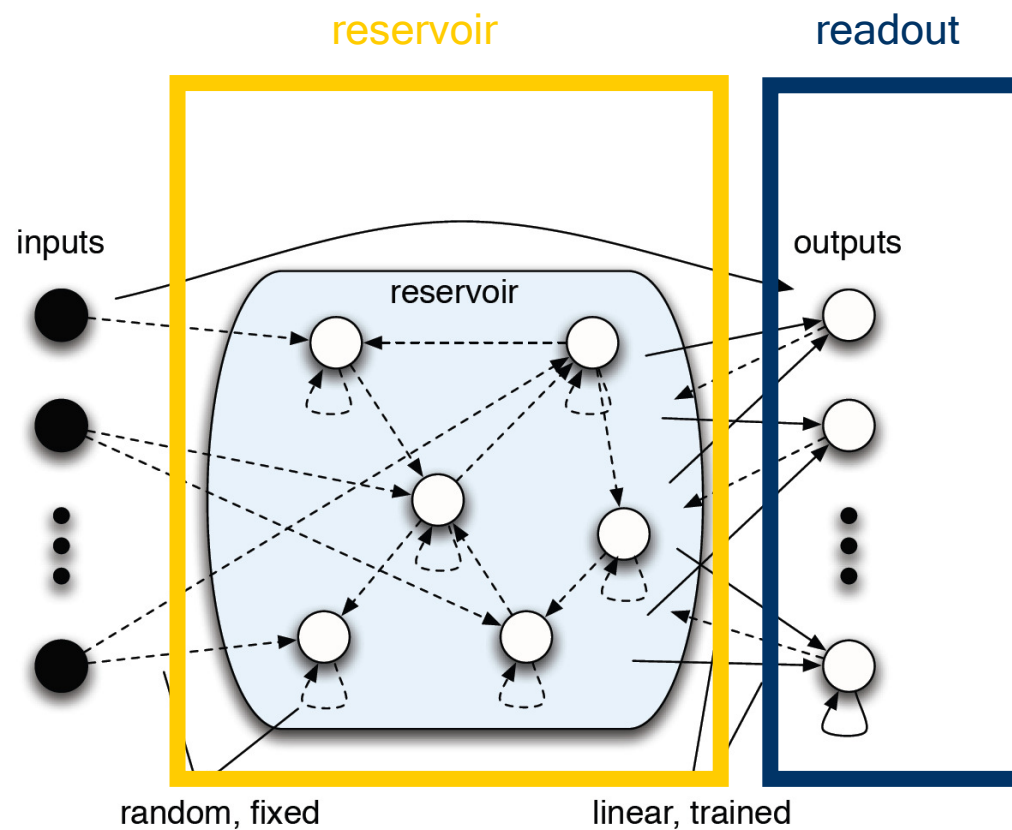
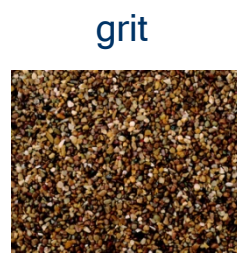
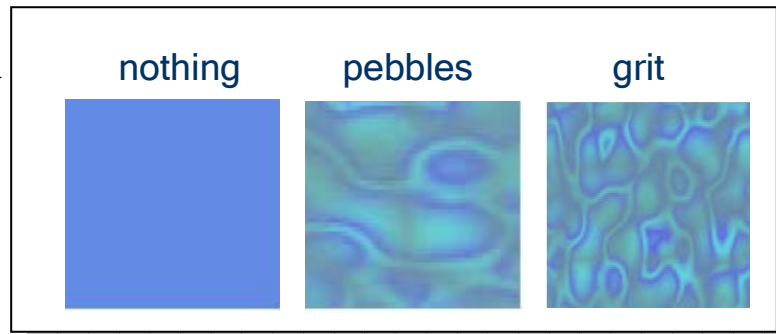


figure taken from B. Schrauwen



reservoir state



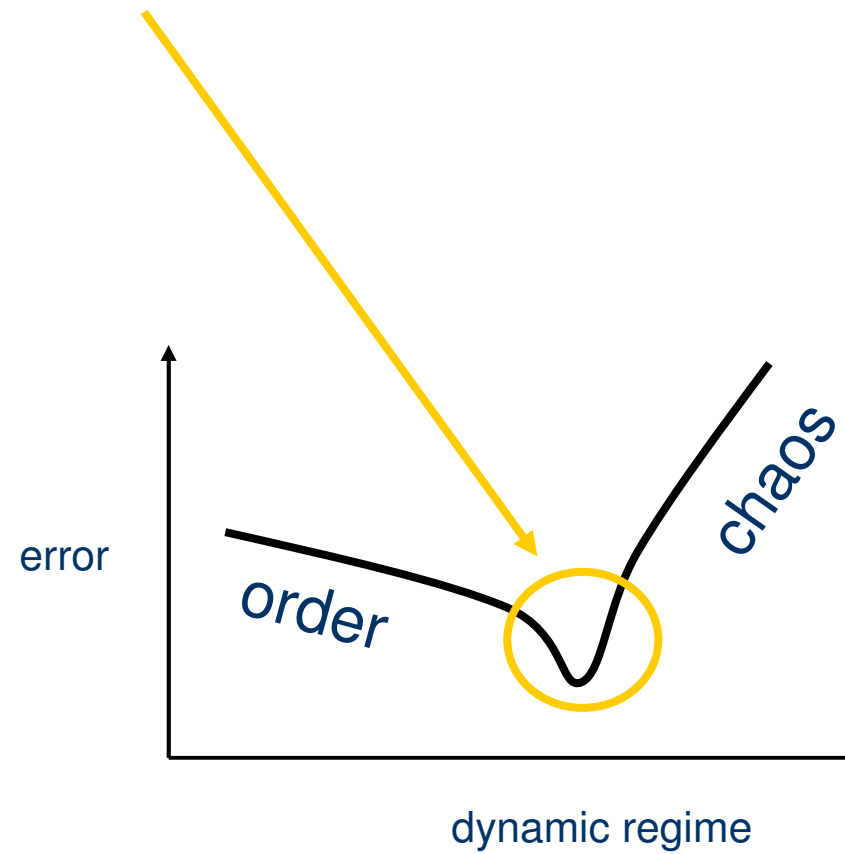
reservoir

readout



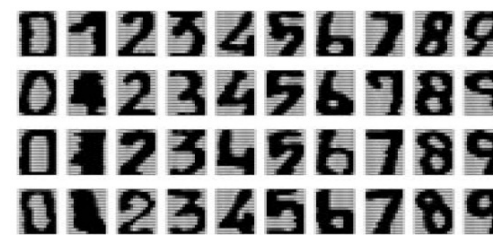
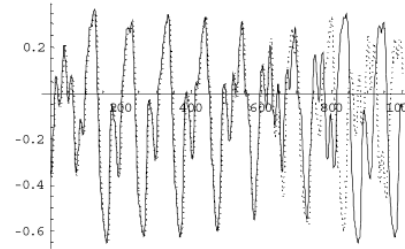
Figure taken from J. Dambre

Computational power is the highest
on the edge of stability



Applications:

- Chaotic time series prediction
- Speech recognition on small vocabulary
- Digits recognition: better than state-of-the-art
- Robot control



figures taken from H. Jaeger

1. Reservoir Computing
2. Do it with photonics
3. Simulation results
4. Future

We should go from slow software
to fast and power efficient hardware

Photonic

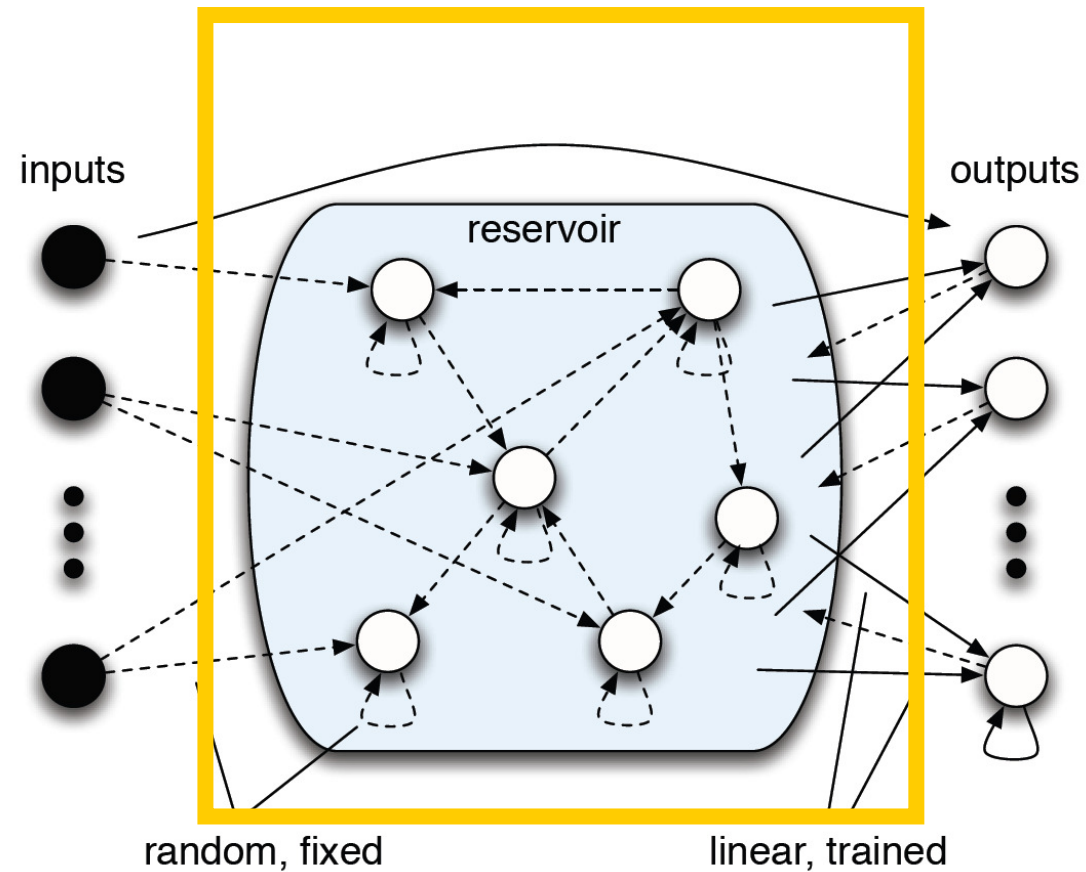
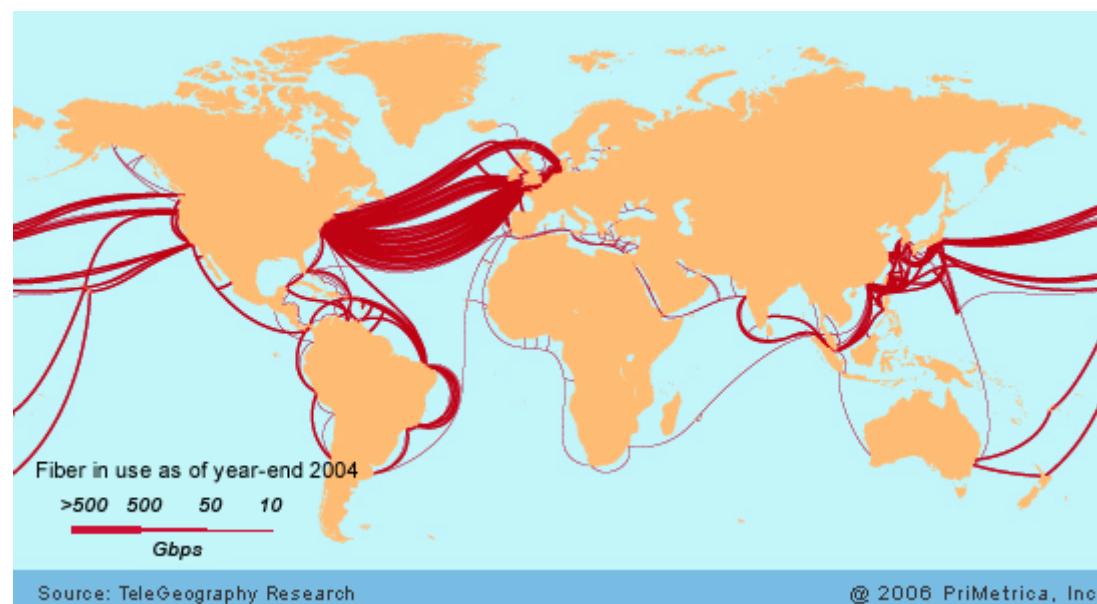


figure taken from B. Schrauwen

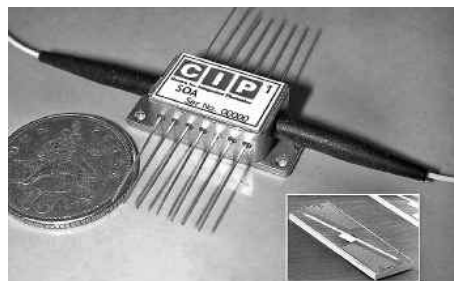
Photonics is replacing electronics for large distances



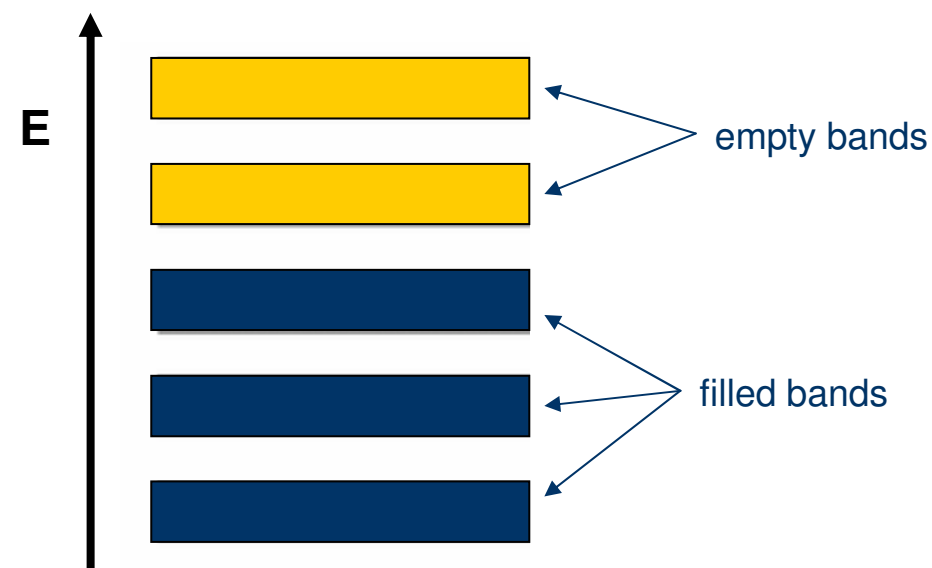
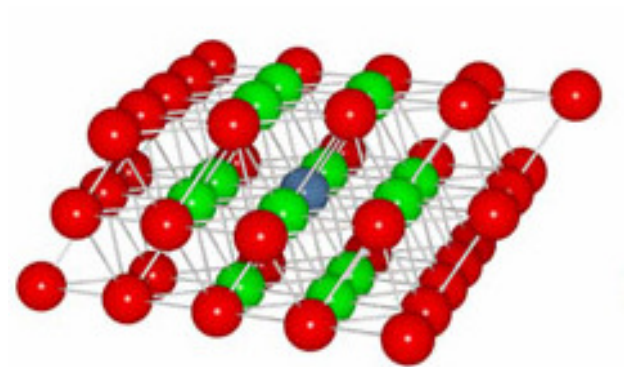
But those distances are getting smaller...



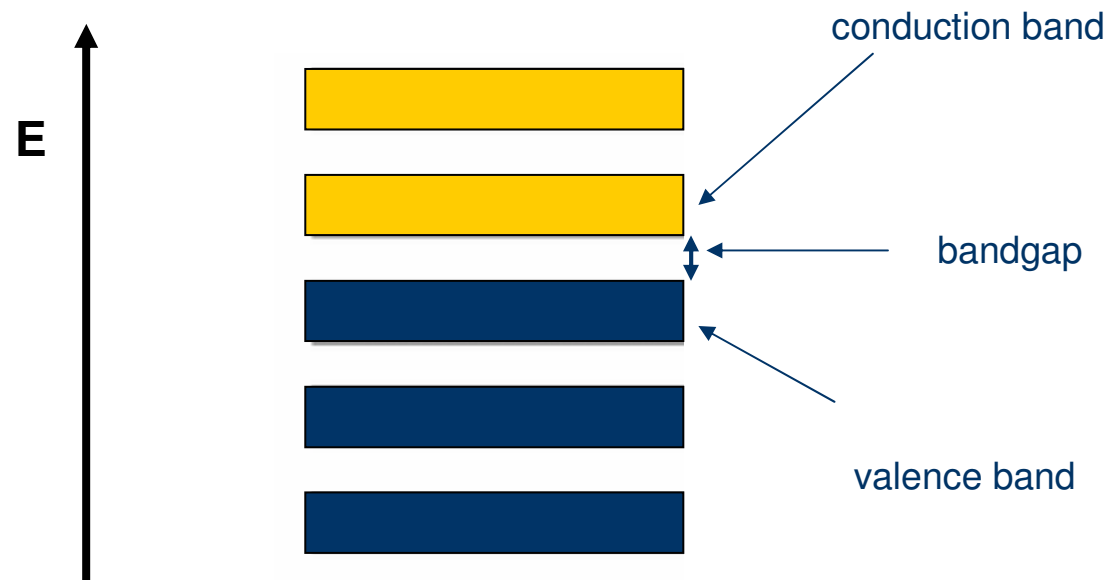
An optical chip with integrated Semiconductor Optical Amplifiers (SOAs) amplifying light



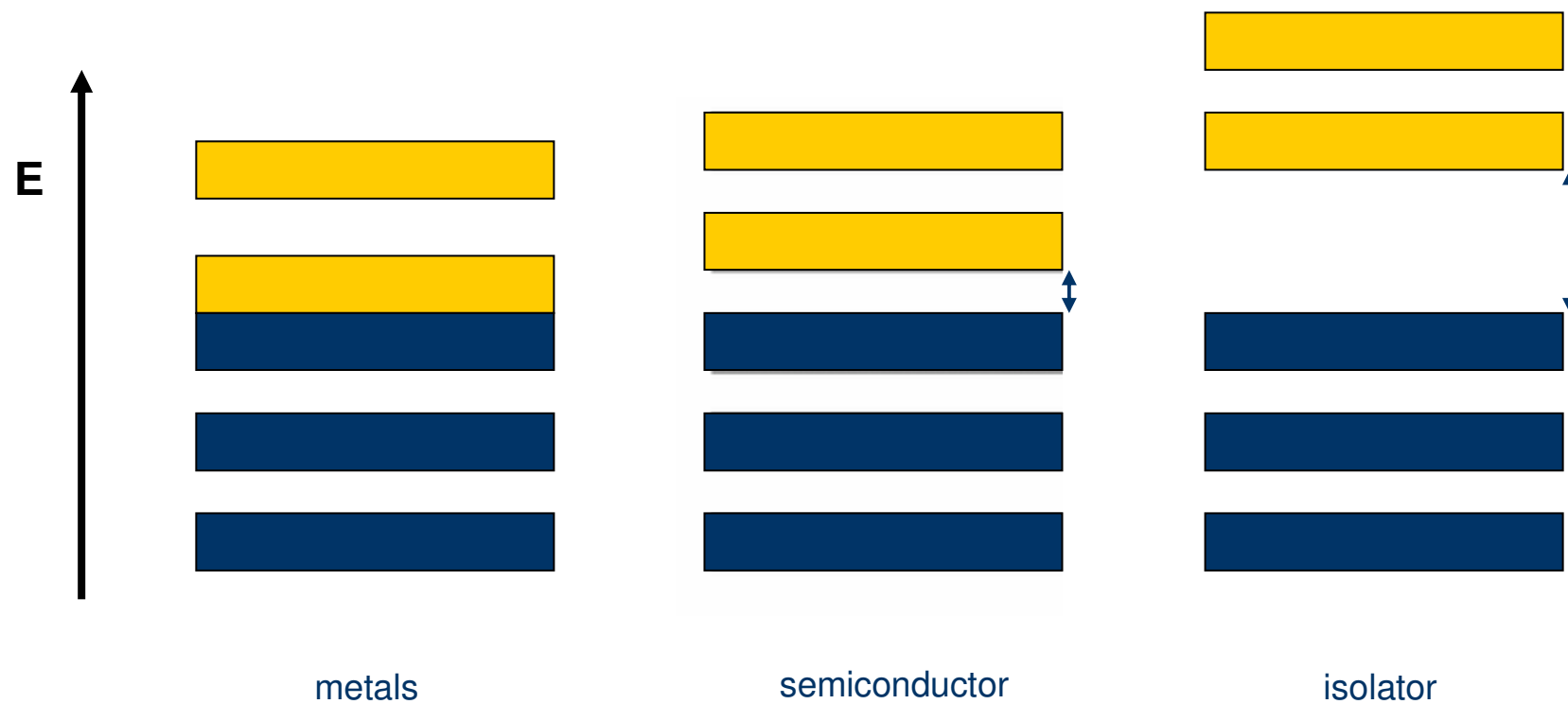
Electrons can only inhabit certain energy bands in solids



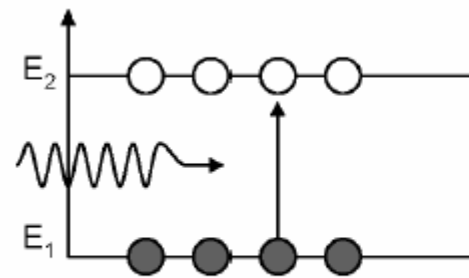
The energy gap between the top band filled with electrons and the first empty band is called the *bandgap*



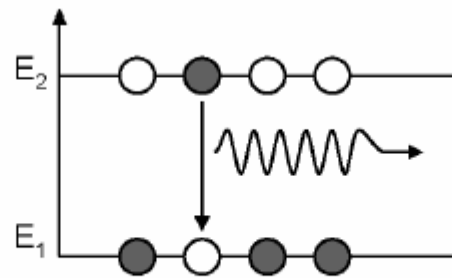
The size of the bandgap differs for
conductors (metals), semiconductors and isolators



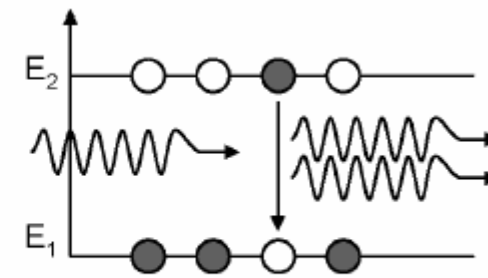
The small bandgap for semiconductors is the origin of three different interactions between electrons and photons



(a) Absorption

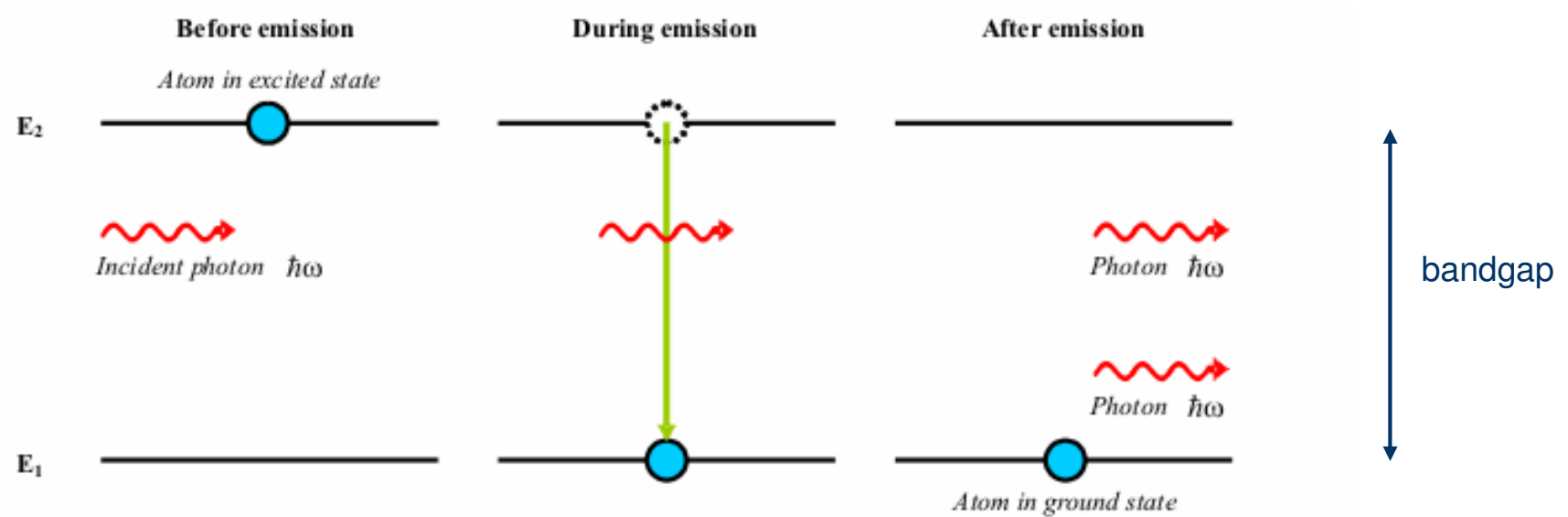


(b) Spontaneous emission

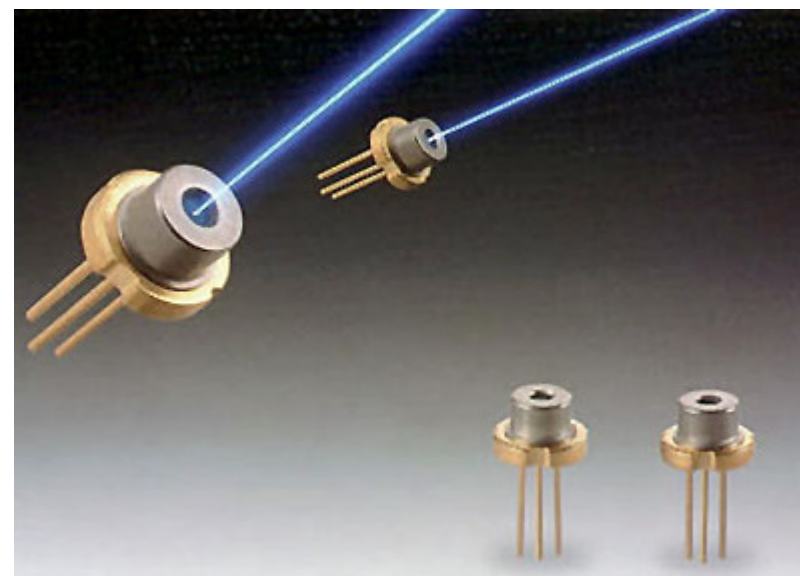


(c) Stimulated emission

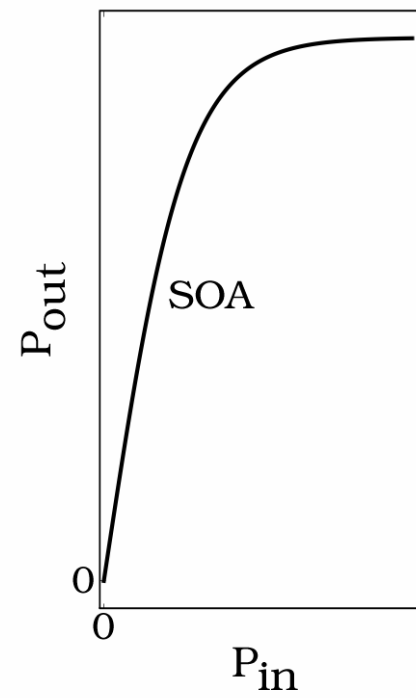
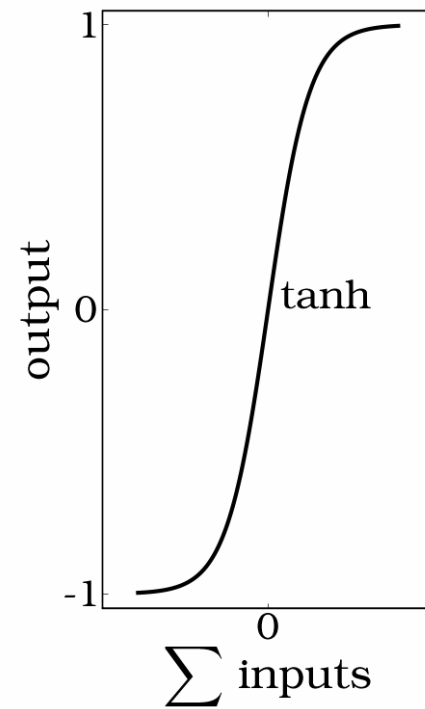
Efficient stimulated emission requires more excited states than ground states (population inversion)



Therefore they need to be pumped (current, light,...).
They are not particularly energy efficient or fast,
but they are broadband



SOAs are a bridge between
the reservoir and the photonic world



The gain in the SOA model is dependent on the **input power** and its own history

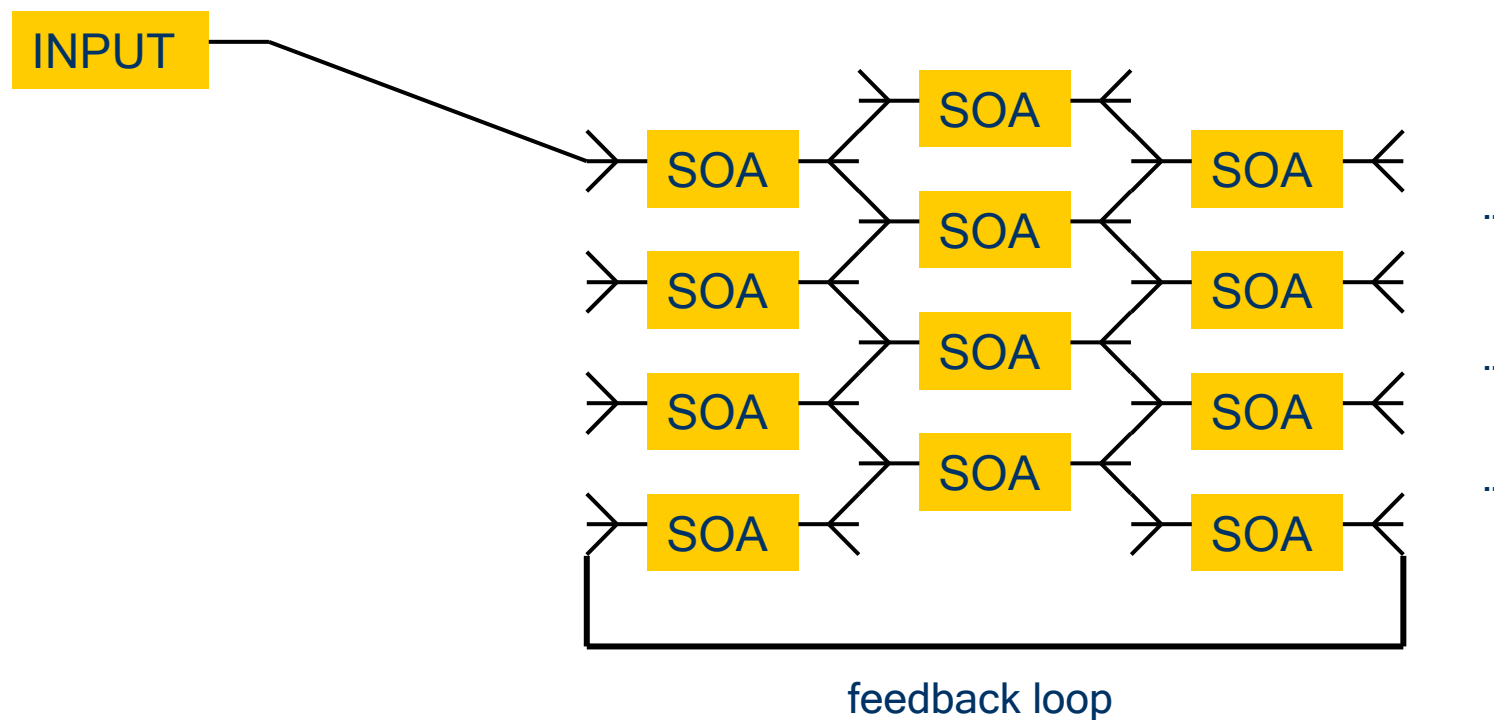
$$P_{out}(\tau) = P_{in} \exp[h(\tau)]$$

$$\phi_{out}(\tau) = \phi_{in} - \frac{1}{2} \alpha h(\tau)$$

$$h(\tau) = \int_0^L g(z, \tau) dz$$

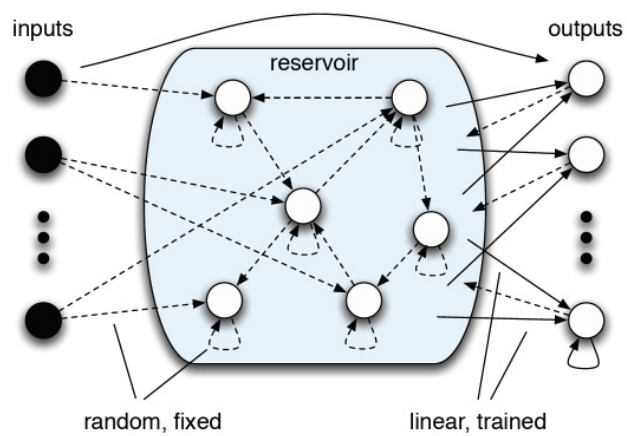
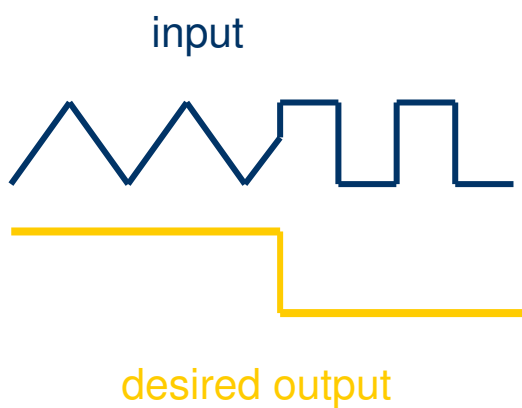
$$\frac{dh}{d\tau} = \frac{g_0 L - h}{\tau_c} - \frac{P_{in}(\tau)}{P_{sat} \tau_c} [\exp(h) - 1]$$

Because chips are planar,
we try to avoid too many crossings

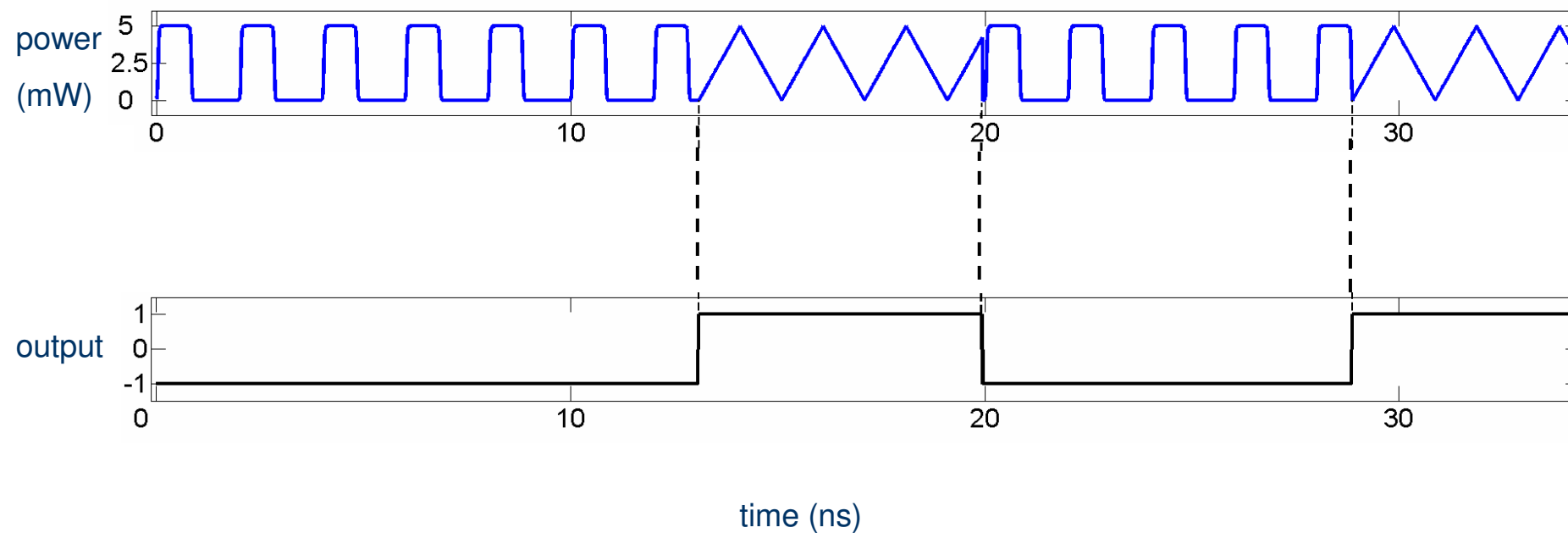


1. Reservoir Computing
2. Do it with photonics
- 3. Simulation results**
4. Future

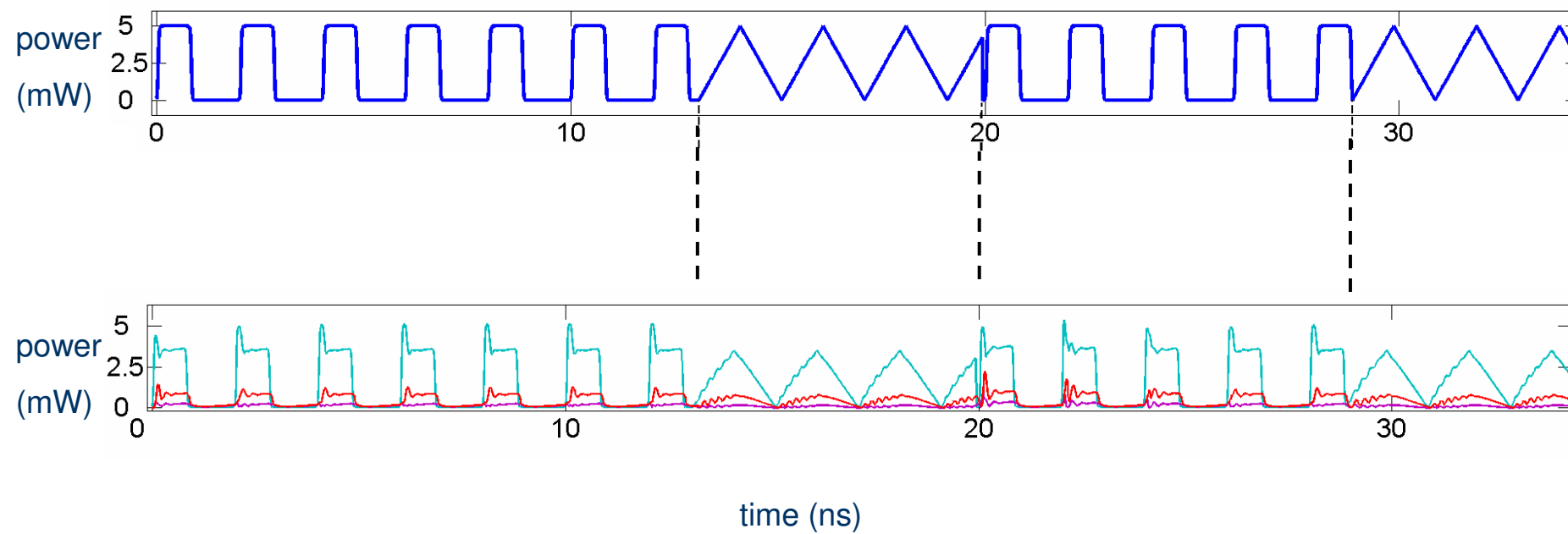
Train the network to distinguish between
a triangular and a rectangular function



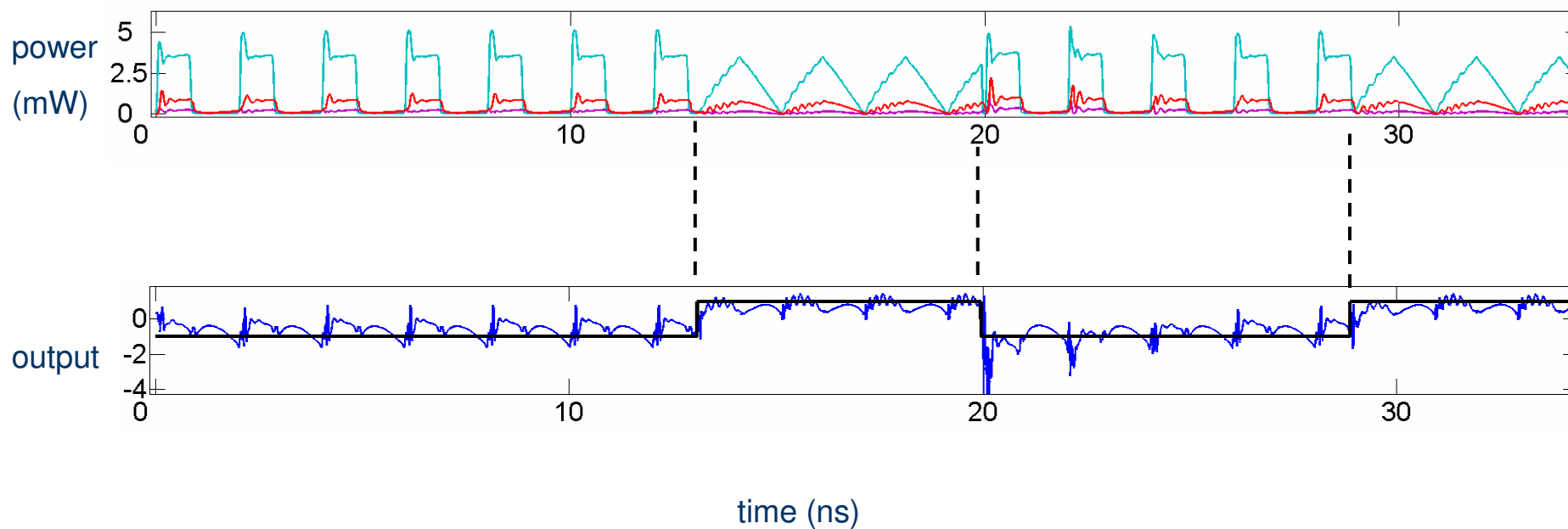
First the input signals and their desired output are defined



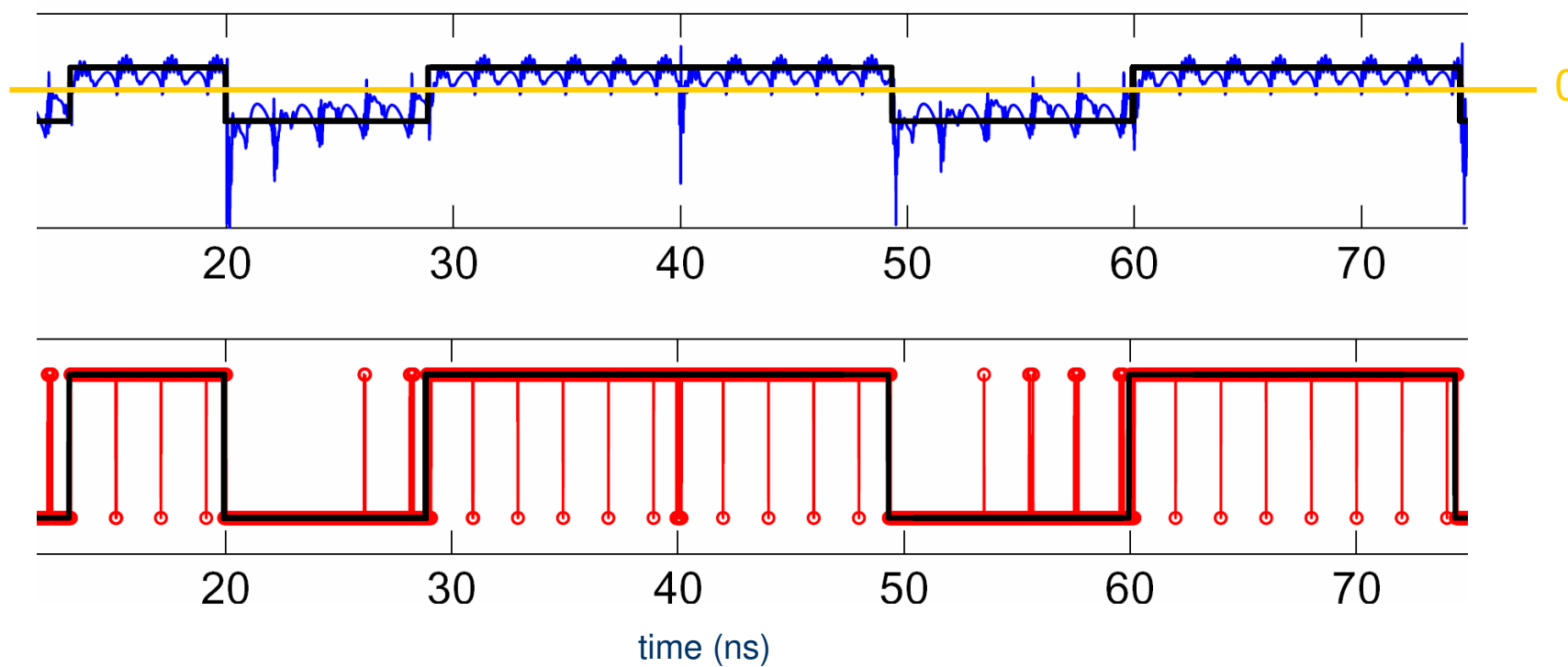
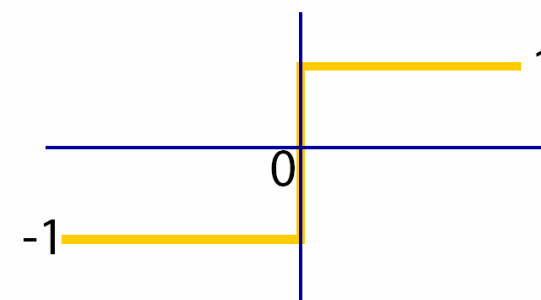
The input is fed into the reservoir



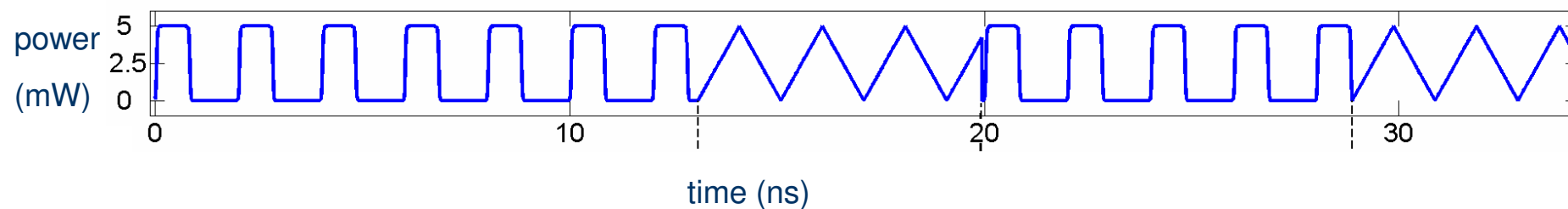
Train the readout function to follow transitions, by using a linear combination of reservoir states



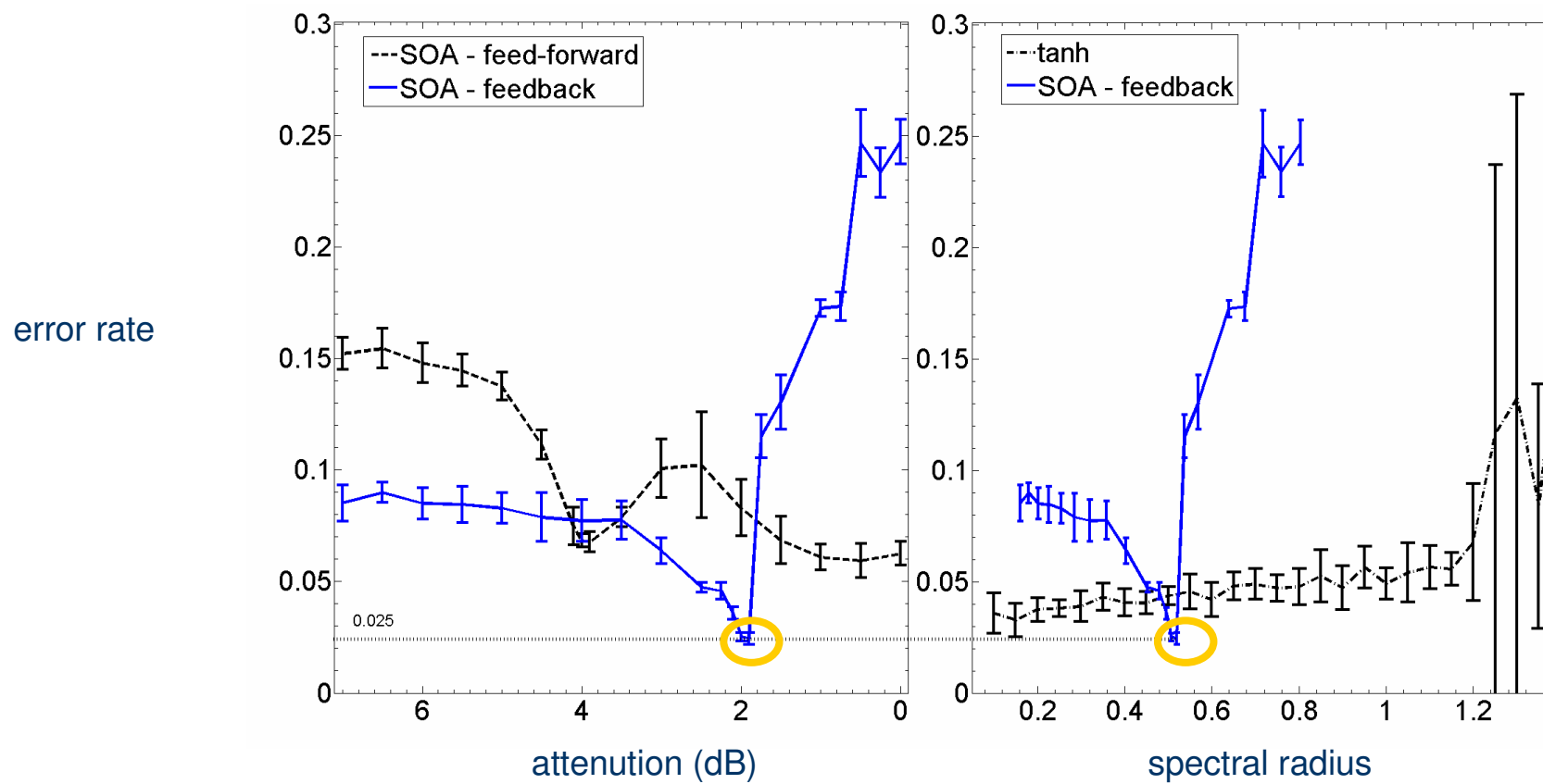
We use a sign function to establish the final output



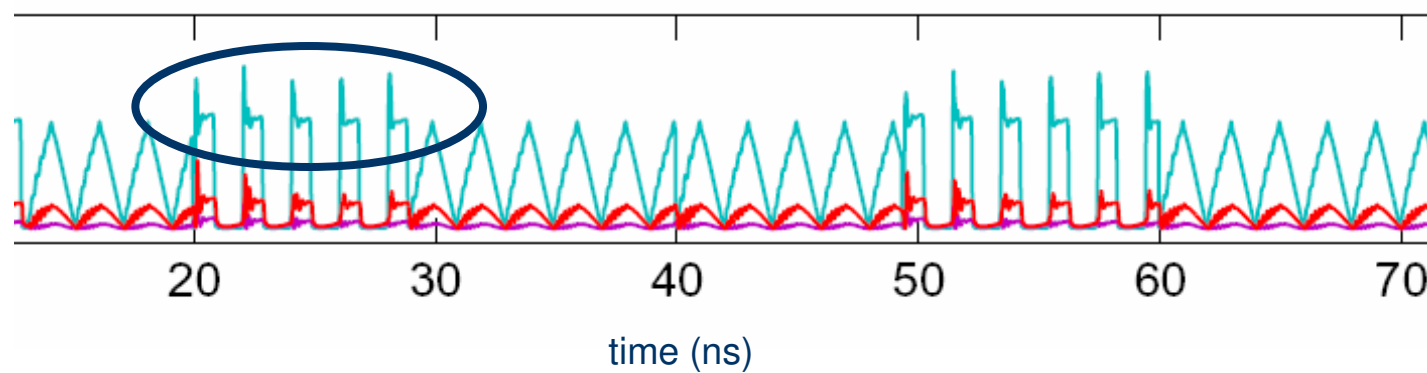
Different input signals are created with transitions on different instants. Some are used to train the network; the rest to test it



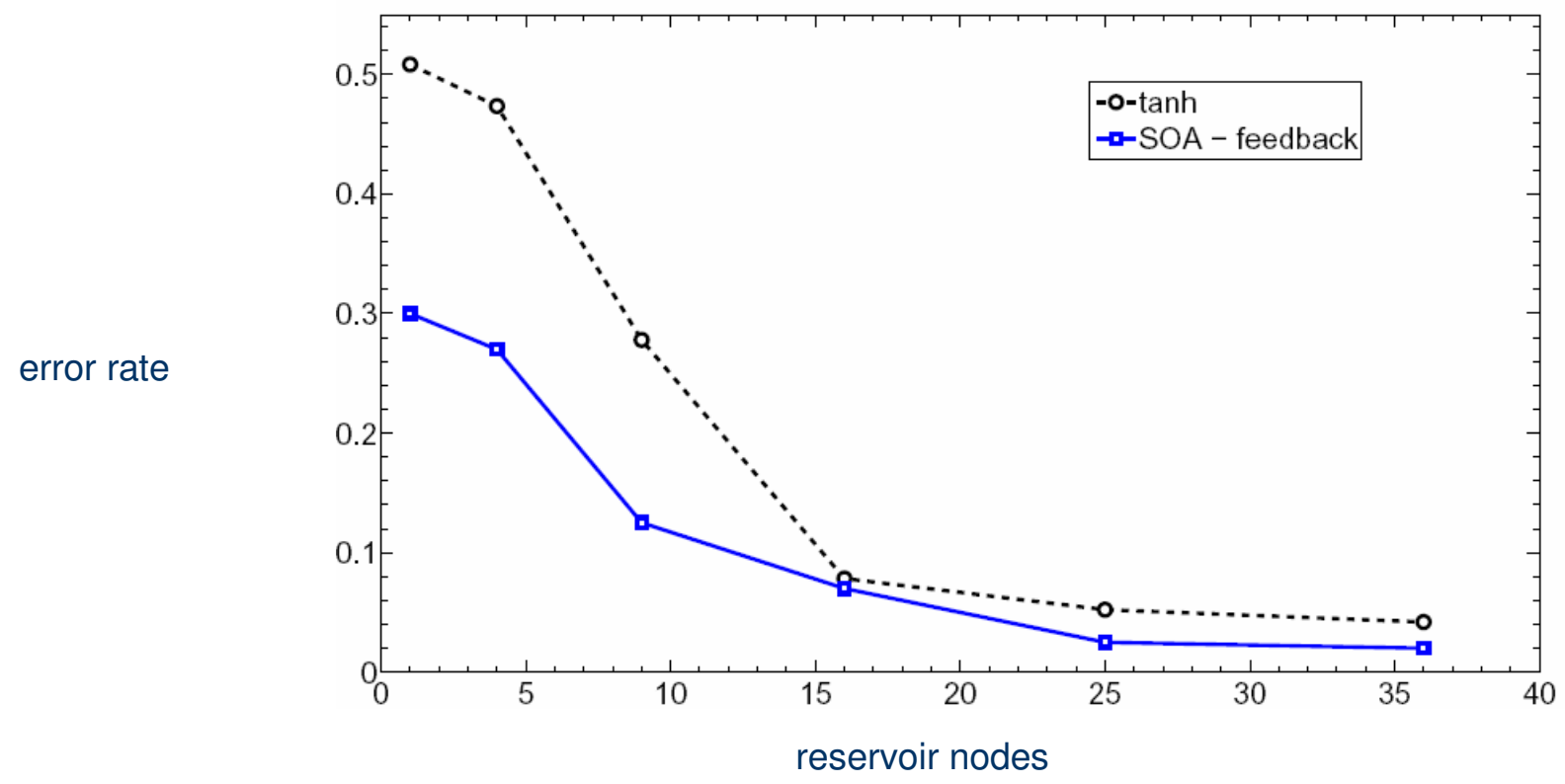
The optimum result is a correct classification 97% of the time



The reason could be the dynamic behavior of the SOA
with **transients** for fast-rising slopes



The results saturate for larger reservoirs



1. Reservoir Computing
2. Do it with photonics
3. Simulation Results
4. Future

Perhaps this won't do any more in the future

Confirmation

* Verification Code:

h Y^t e n

Enter case sensitive Verification Code and click **Submit Request**.

A price I am willing to pay

Confirmation

* Verification Code:

h Y^t e n

Enter case sensitive Verification Code and click **Submit Request**.