

# Energy-efficient neuromorphic computing with magnetic tunnel junctions

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NIST, USA

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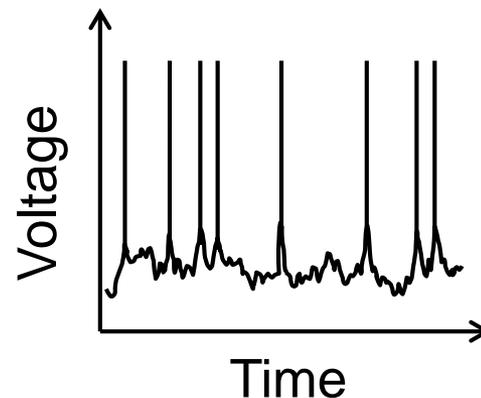
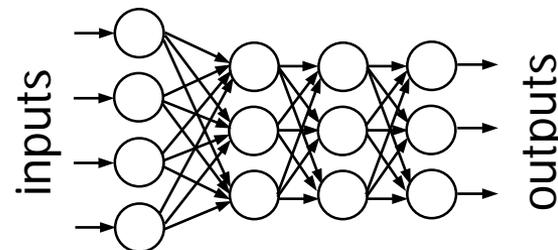


**THALES**



# WHAT: Brain-inspired computing for cognitive tasks

- Neural networks.
  - Already in heavy commercial application.
  - Quasi-static.
- **Dynamic processing.**
  - **Make use of dynamical properties of devices.**
  - **Spikes, rates, oscillations, timing, ...**
- Simulate brain function



# Sophisticated CMOS-based neuromorphic chip development

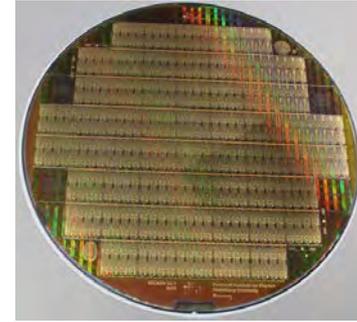
Loihi (Intel)



TrueNorth (IBM)



BrainScales  
(Human Brain Project)



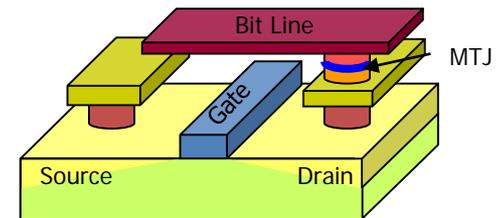
- Brain-like features natural in CMOS
  - local energy source
    - Incoming signal does not power outgoing signal.
  - Connectivity
    - Digital spikes – shared communication channels

**Energy efficiency and complexity is still far from the brain!**

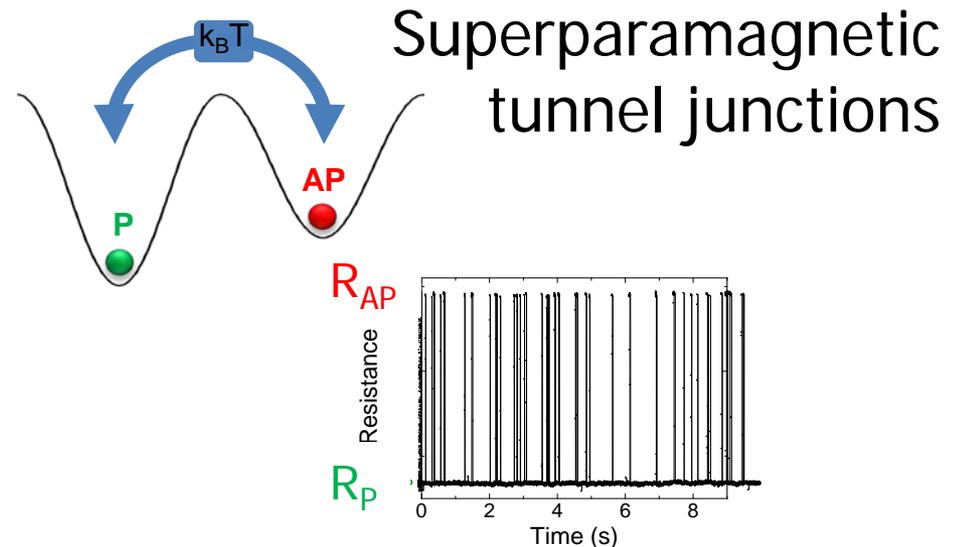
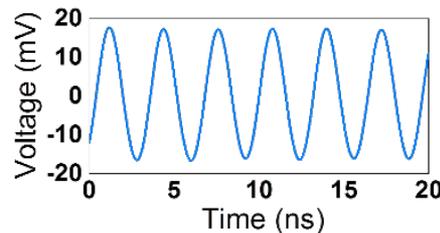
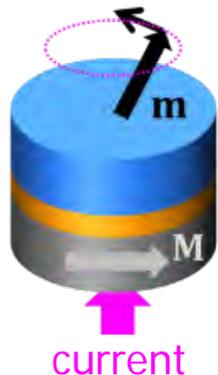
# HOW: Augment CMOS with efficient neural devices (MTJs)

- Features for which CMOS may be inefficient (energy and/or device area)
  - Non-volatility
  - Plasticity (local learning)
  - Stochasticity
  - **Oscillators**

## MRAM



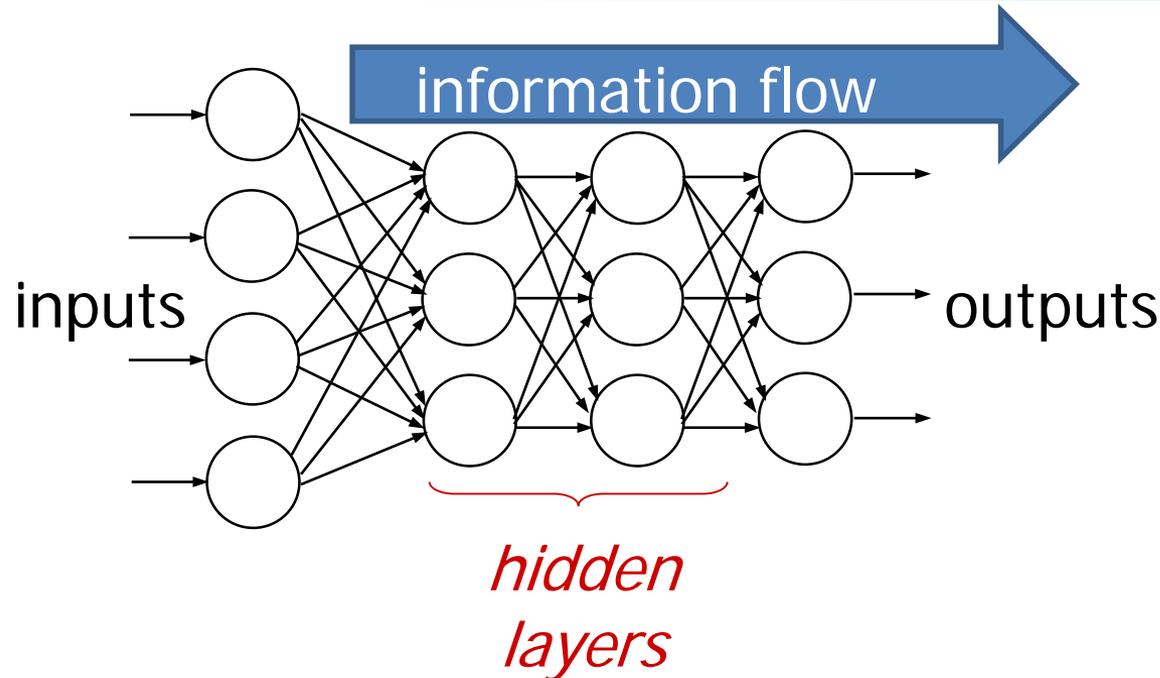
## Nano-oscillators



# Single oscillator reservoir computing

- **Time multiplexed reservoir computing**
- Sine/Square Identification (Intrinsic memory)
- Spoken digit recognition (Non-linearity)
- Sine/Square Identification (Delayed feedback memory)

# Feed-forward networks – one direction of information flow



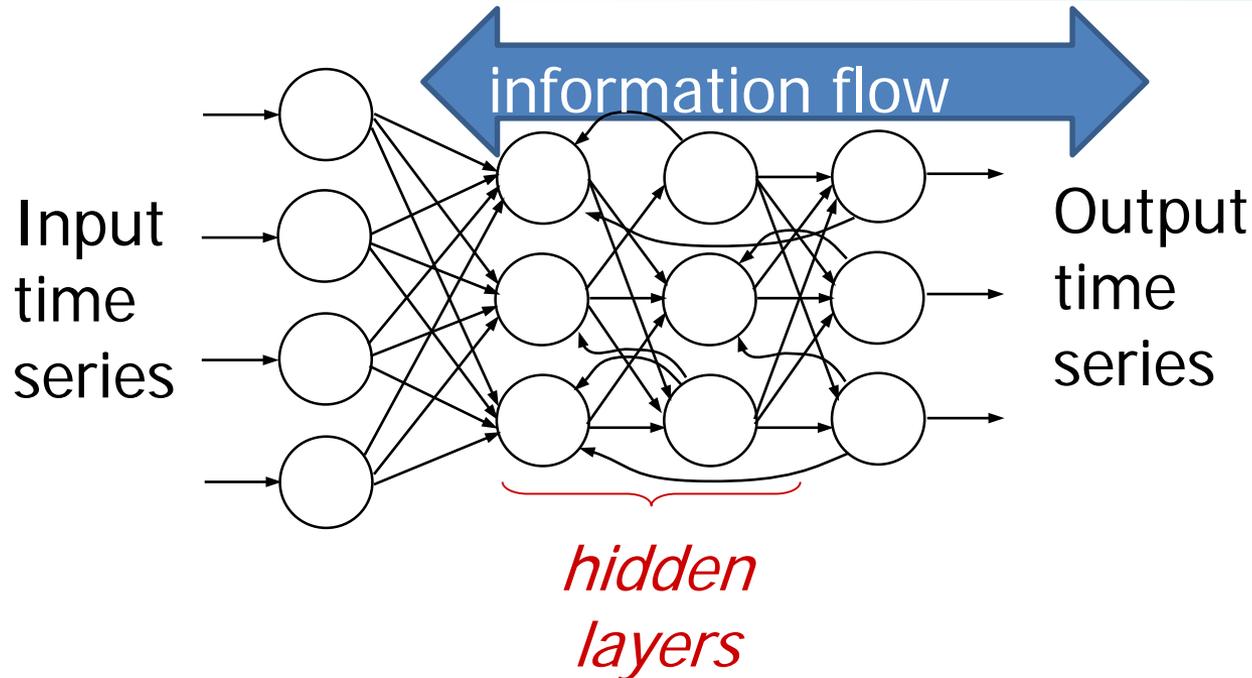
Non-linear nodes (neurons)

rearranges spaces to allow classification

Train off-line

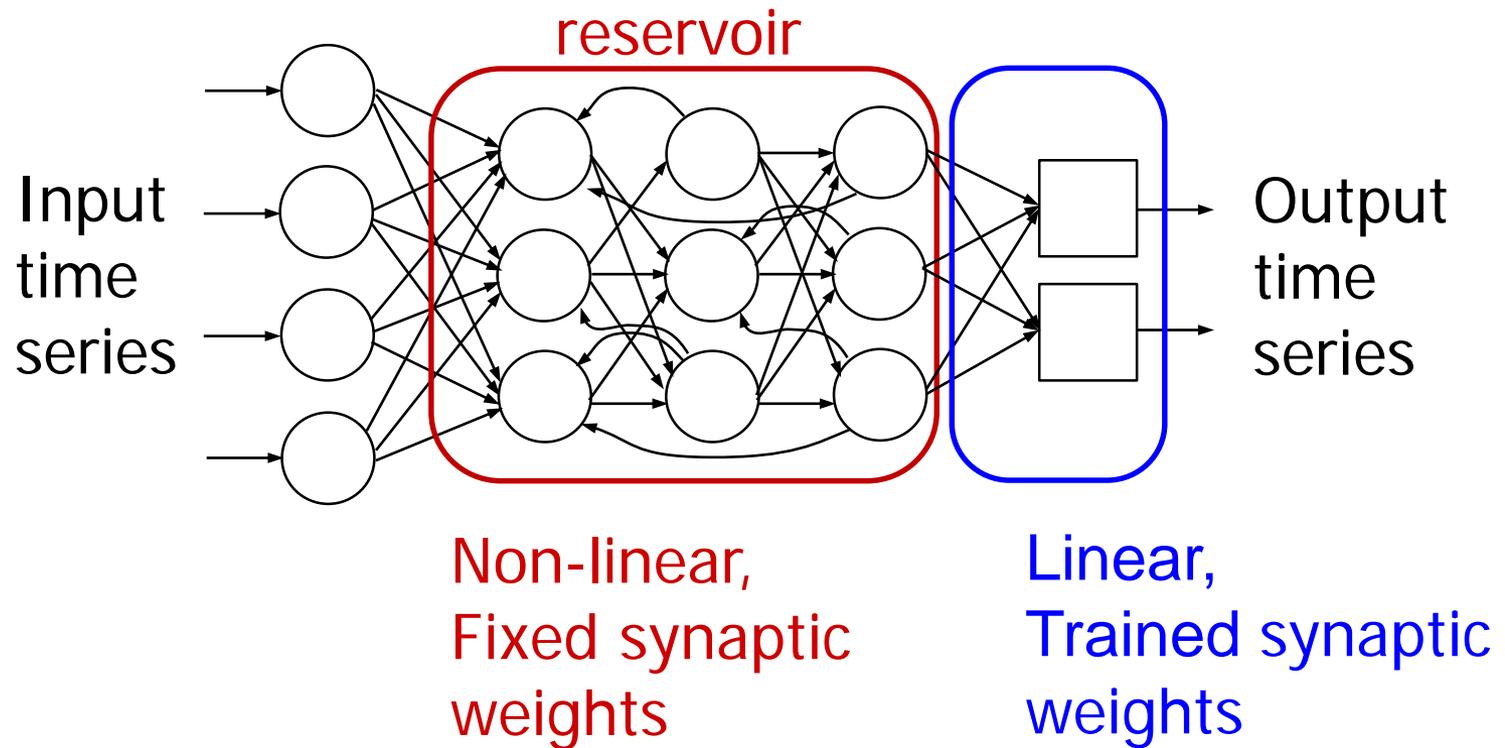
adjust synaptic weights to optimize fit to test data

# Recurrent networks – have intrinsic time scales

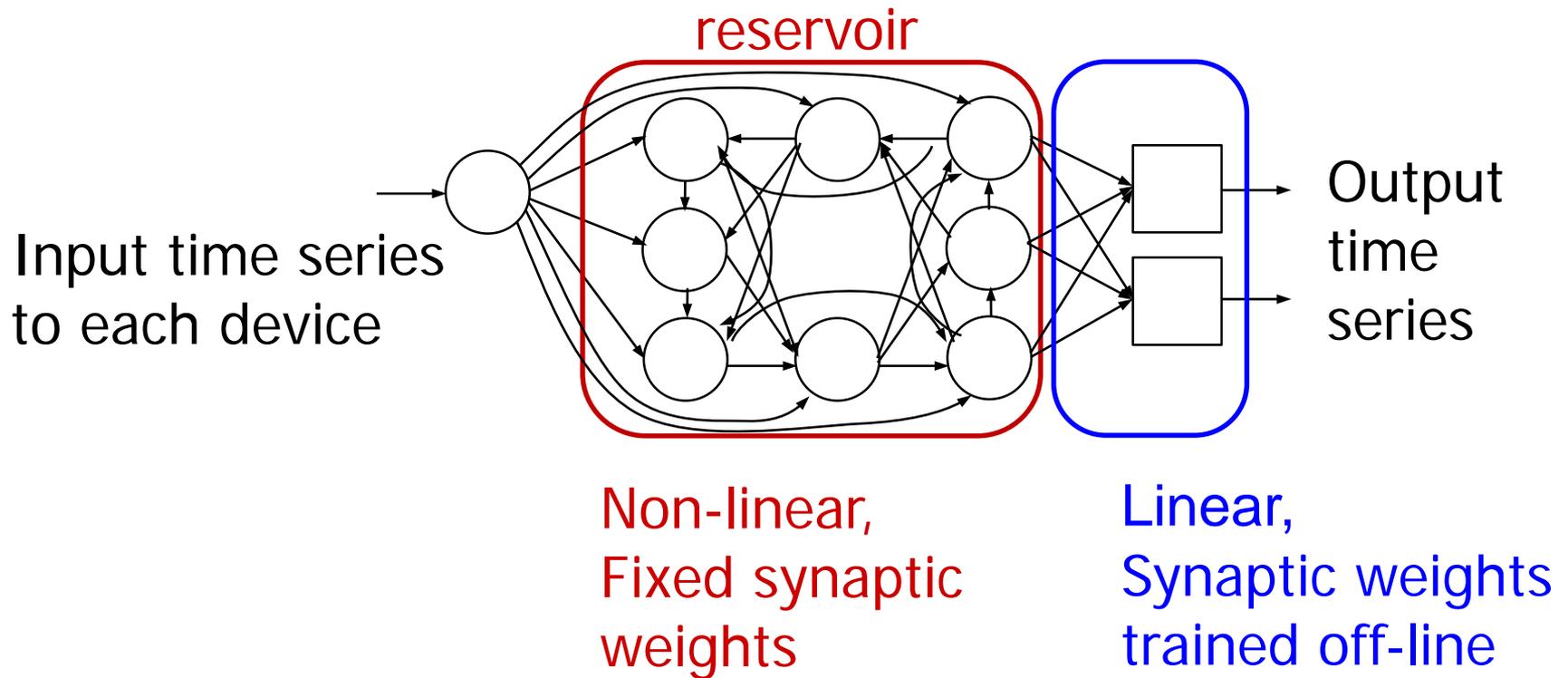


Output time dependent  
input becomes time series  
Training protocol not simple

# Reservoir computing – a simply trainable recurrent network

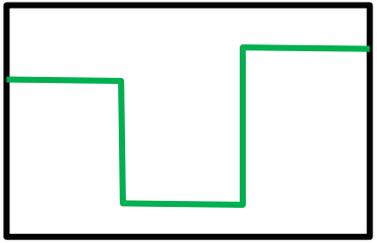


# Reservoir computing – ring geometry



# Reservoir computing – time multiplexed single device

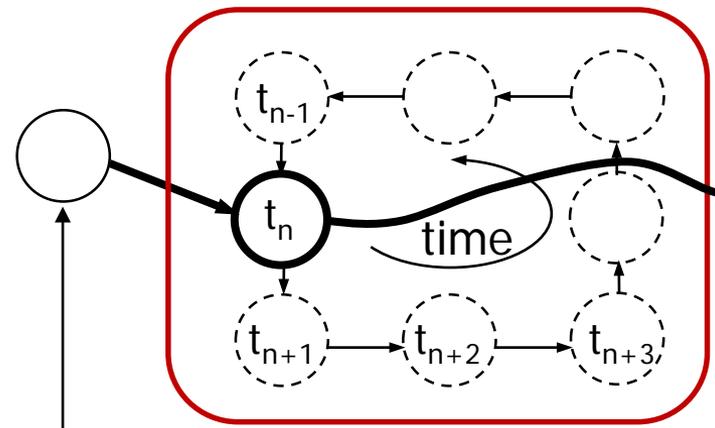
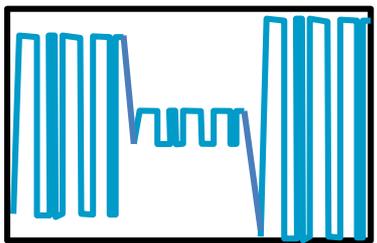
Input time series



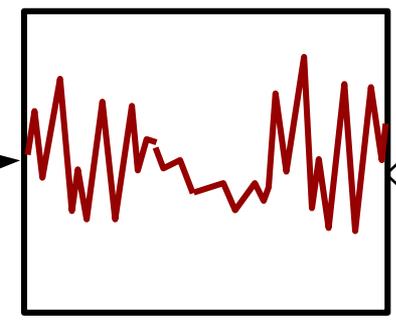
times mask



Time multiplexed input to reservoir

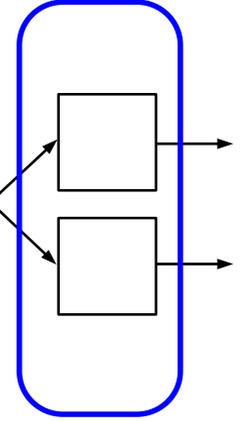


reservoir –  
Device at different  
times gives virtual  
Devices.



output –  
time series

outputs

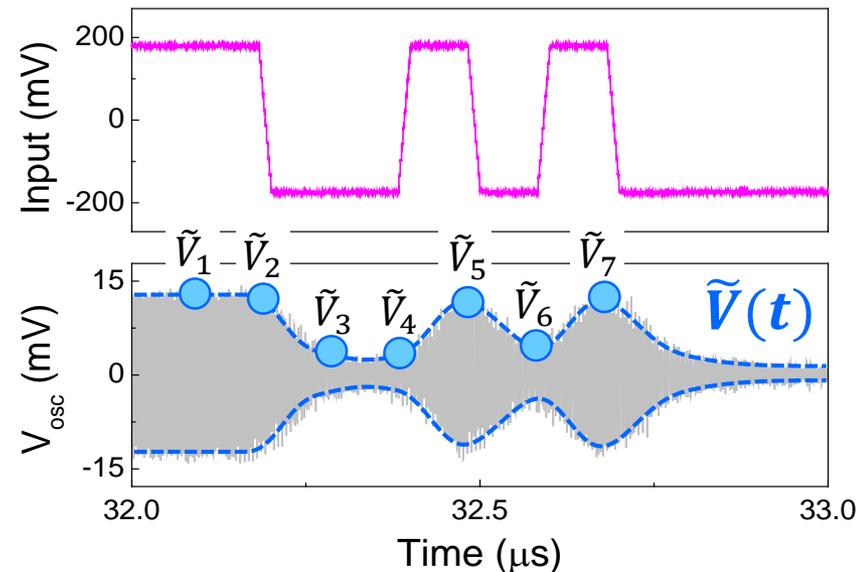
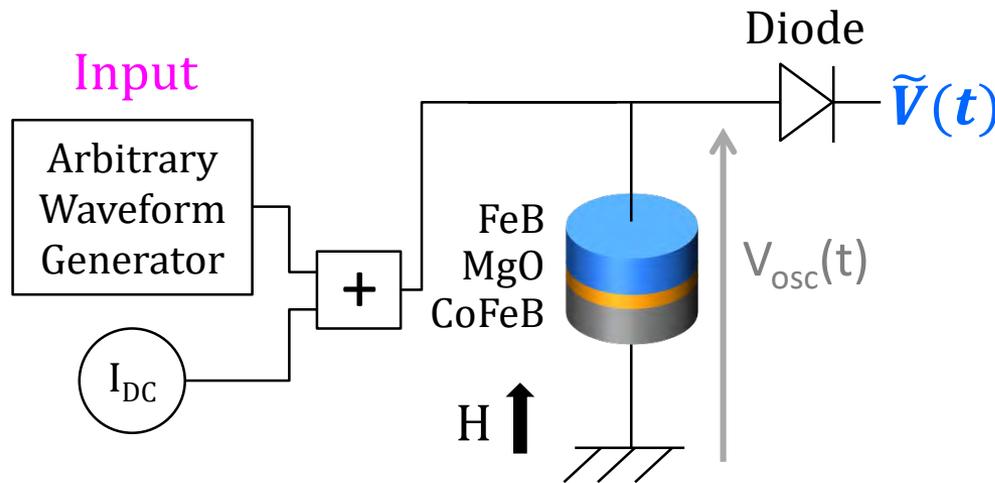
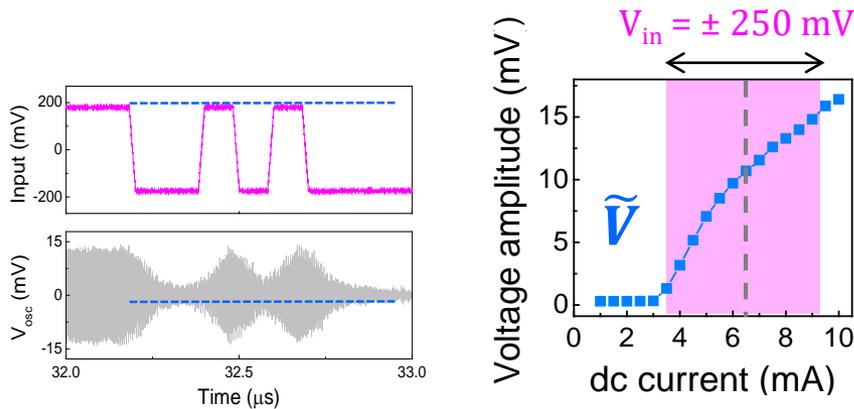


Output weights,  
time dependent,  
accumulated.

# Spin-torque nano-oscillators: non-linear amplitude dynamics and memory

Torrejon *et al.*, Nature 547, 428 (2017).

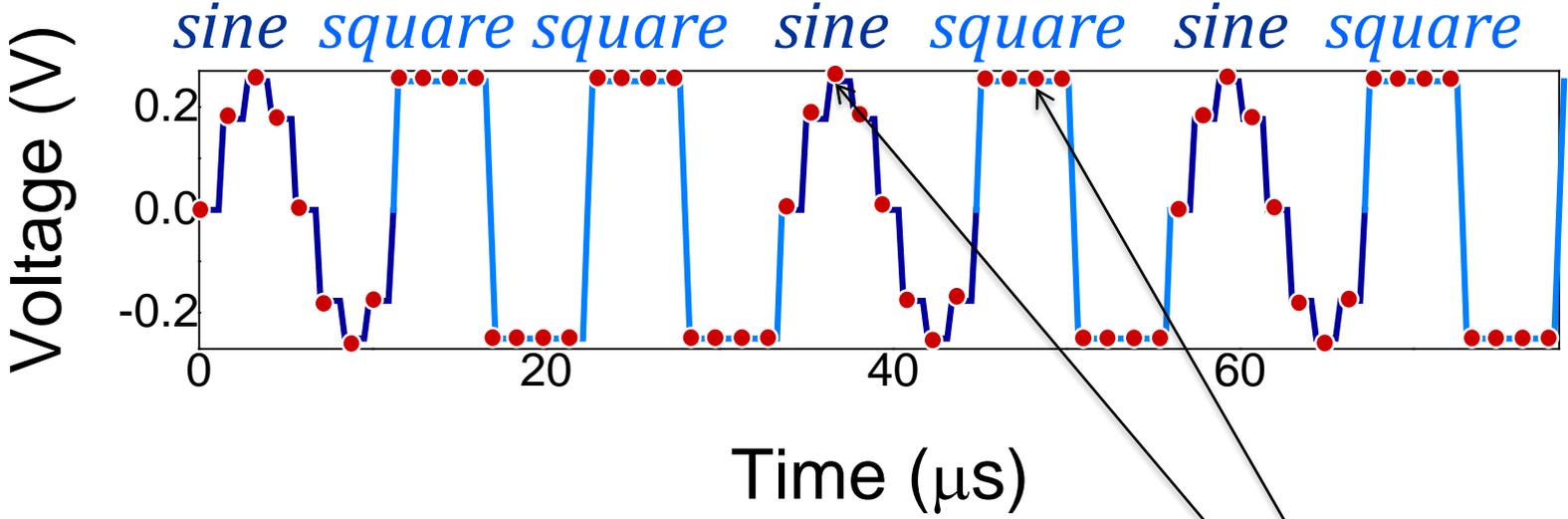
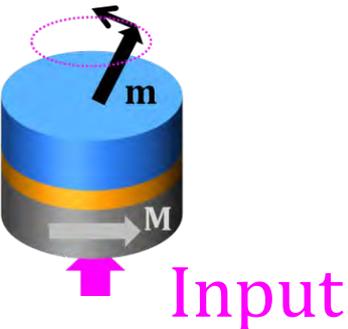
Input: current  
Output: amplitude of the oscillator's voltage



# Single oscillator reservoir computing

- Time multiplexed reservoir computing
- **Sine/Square Identification (Intrinsic memory)**
- Spoken digit recognition (Non-linearity)
- Sine/Square Identification (Delayed feedback memory)

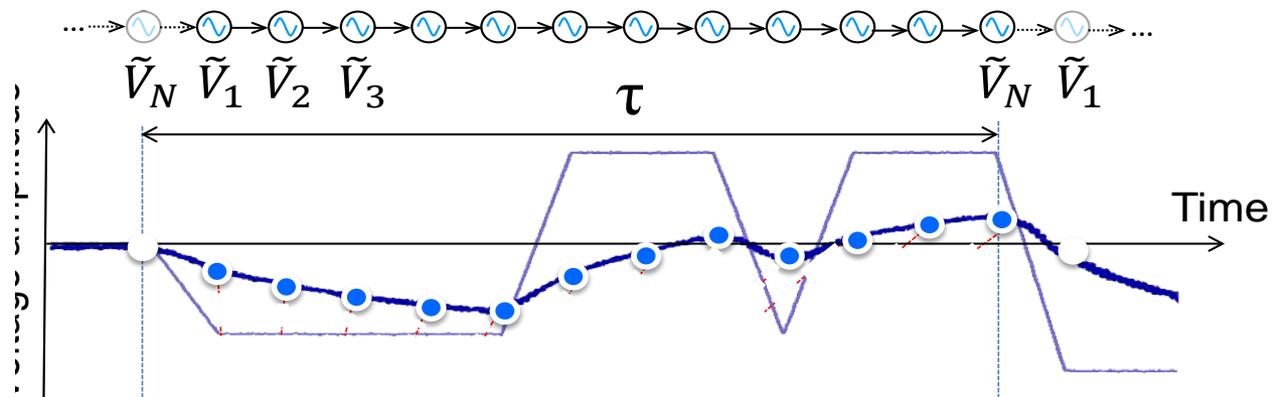
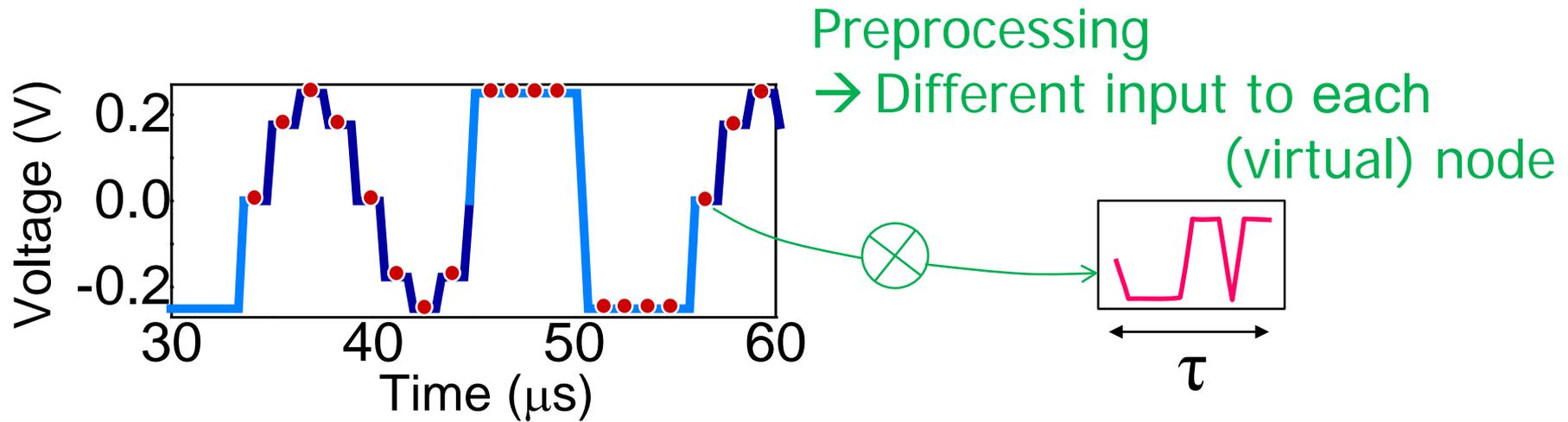
# Task: recognizing sines from squares at each point in time with a single oscillator



Paquot et al, Scientific Reports 2:287 (2012)

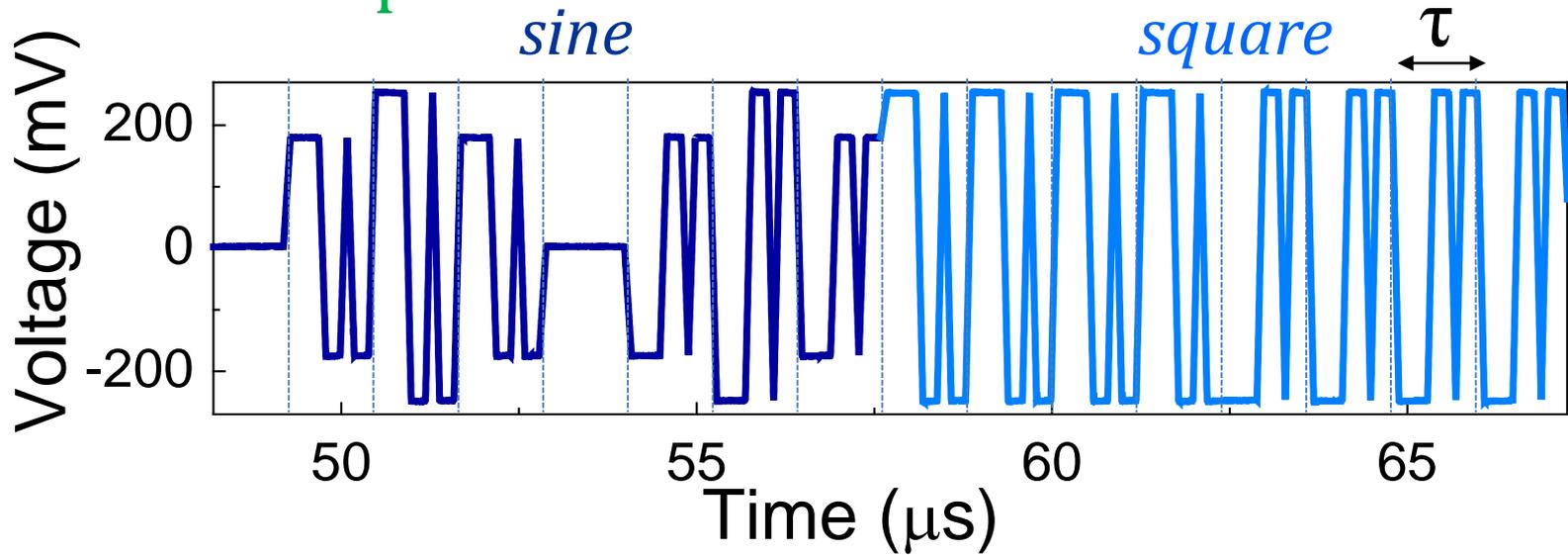
*Same inputs*

For extrinsic memory, each node should couple to a few other nodes.

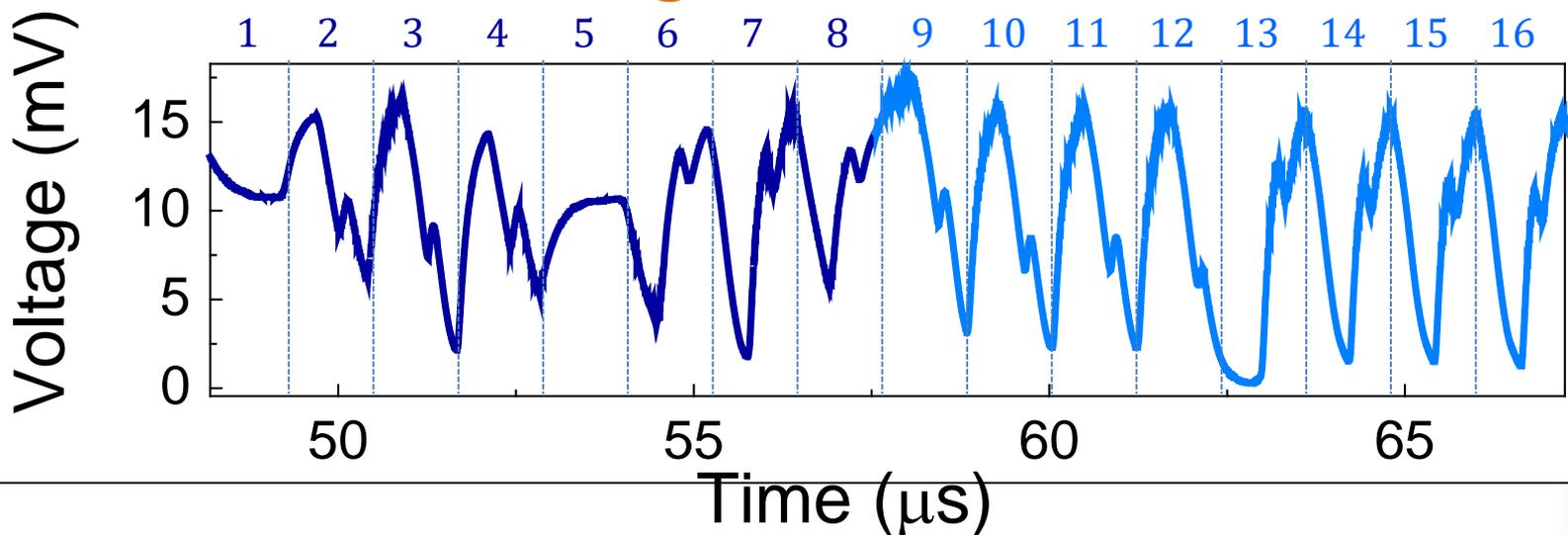


# Experimental trajectories of oscillators' amplitude

## Preprocessed Input

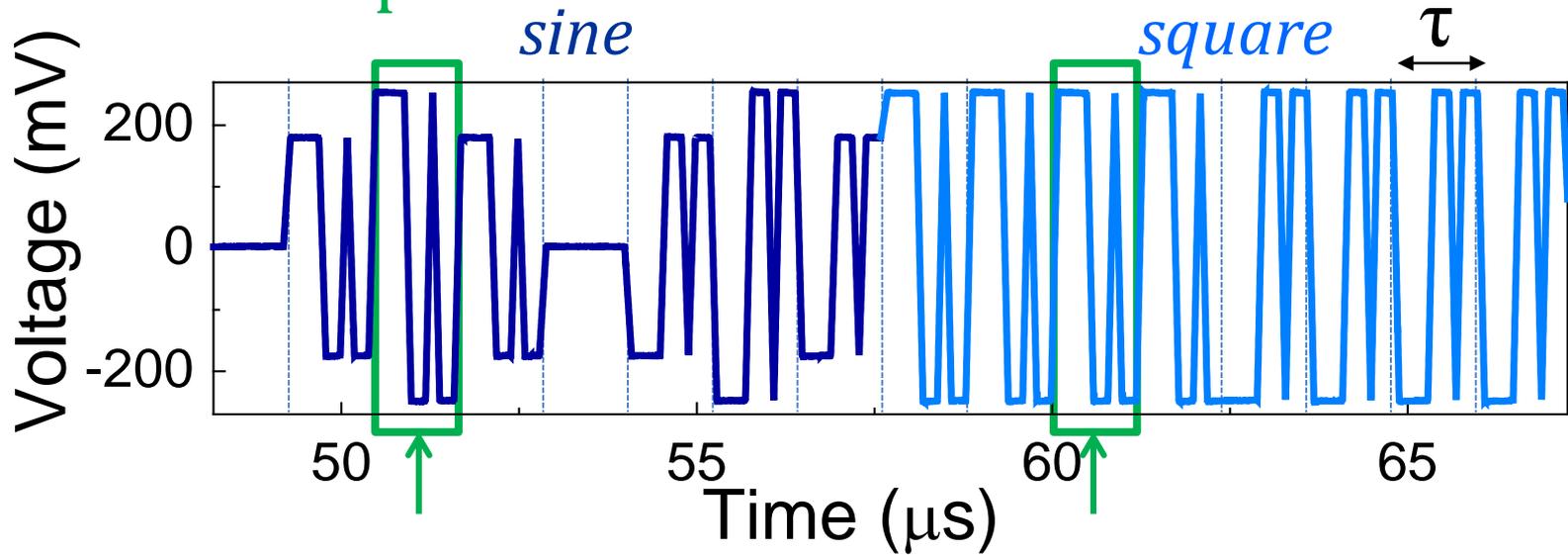


## Oscillator's emitted voltage $\tilde{V}(t)$

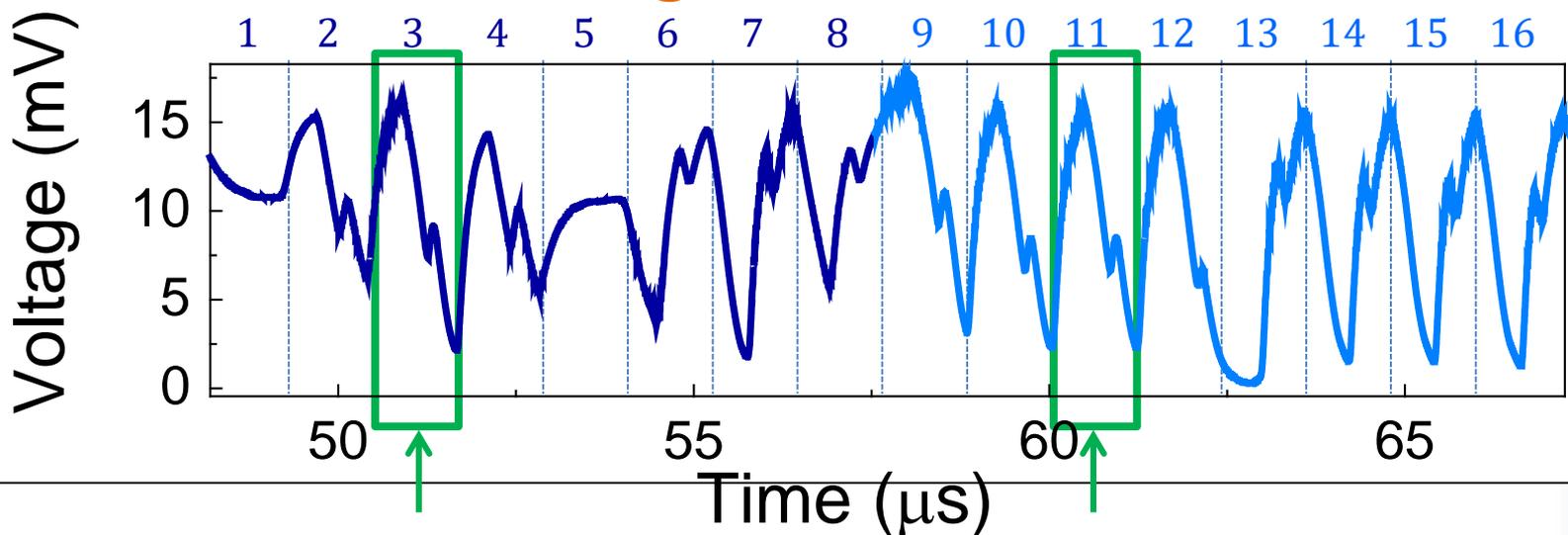


# Experimental trajectories of oscillators' amplitude

## Preprocessed Input

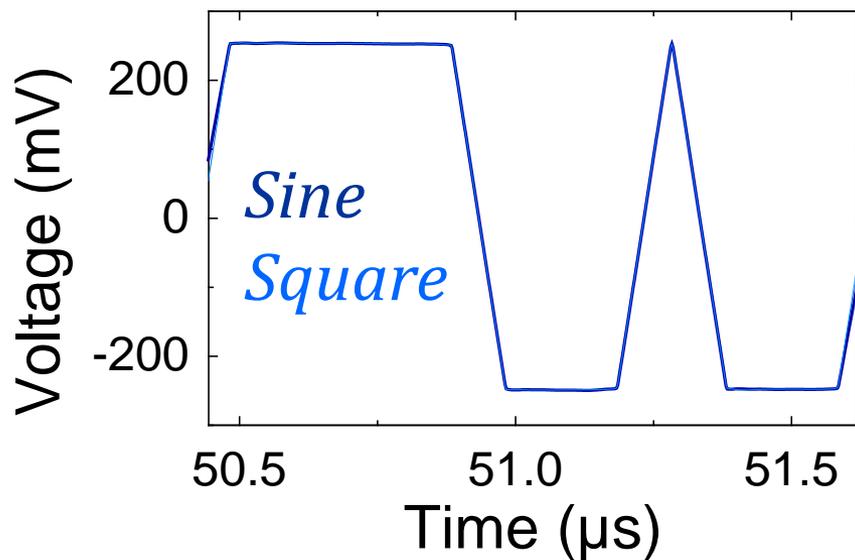


## Oscillator's emitted voltage $\tilde{V}(t)$

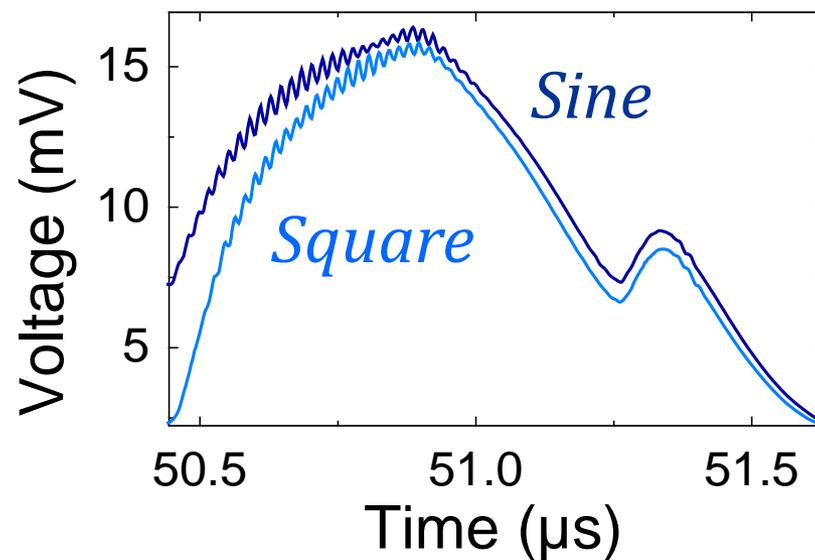


# Different trajectories: data separation is achieved

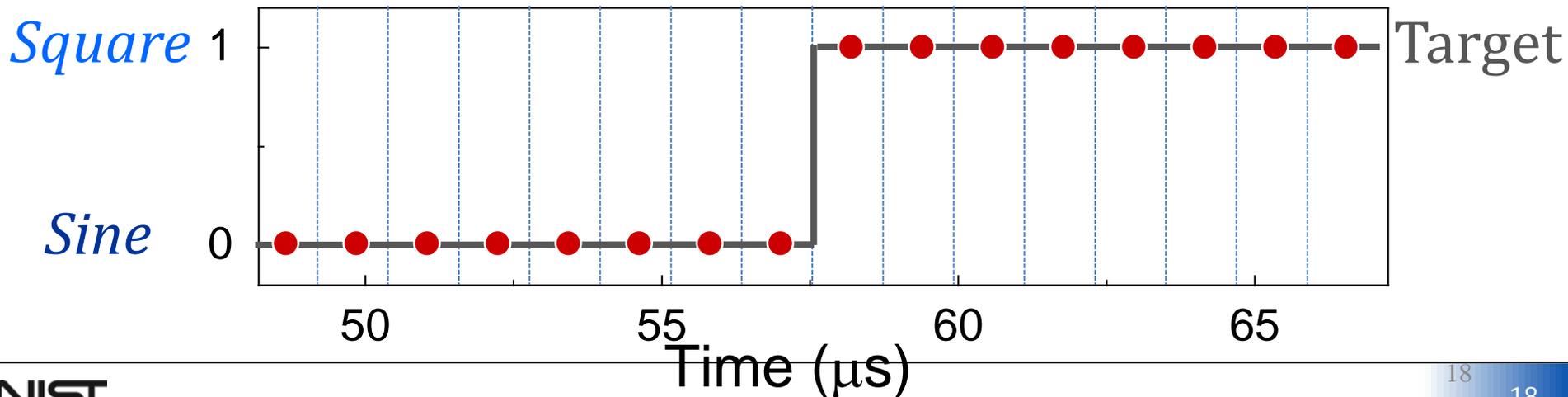
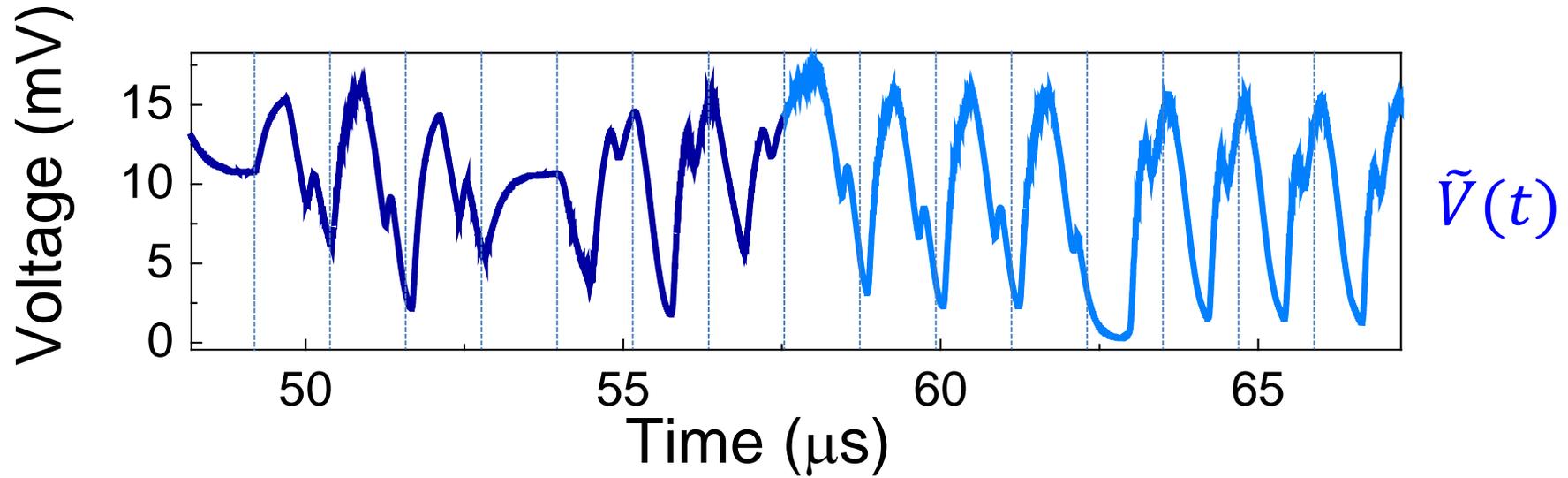
Same input for *sine* and *square*



Different outputs  $\tilde{V}(t)$

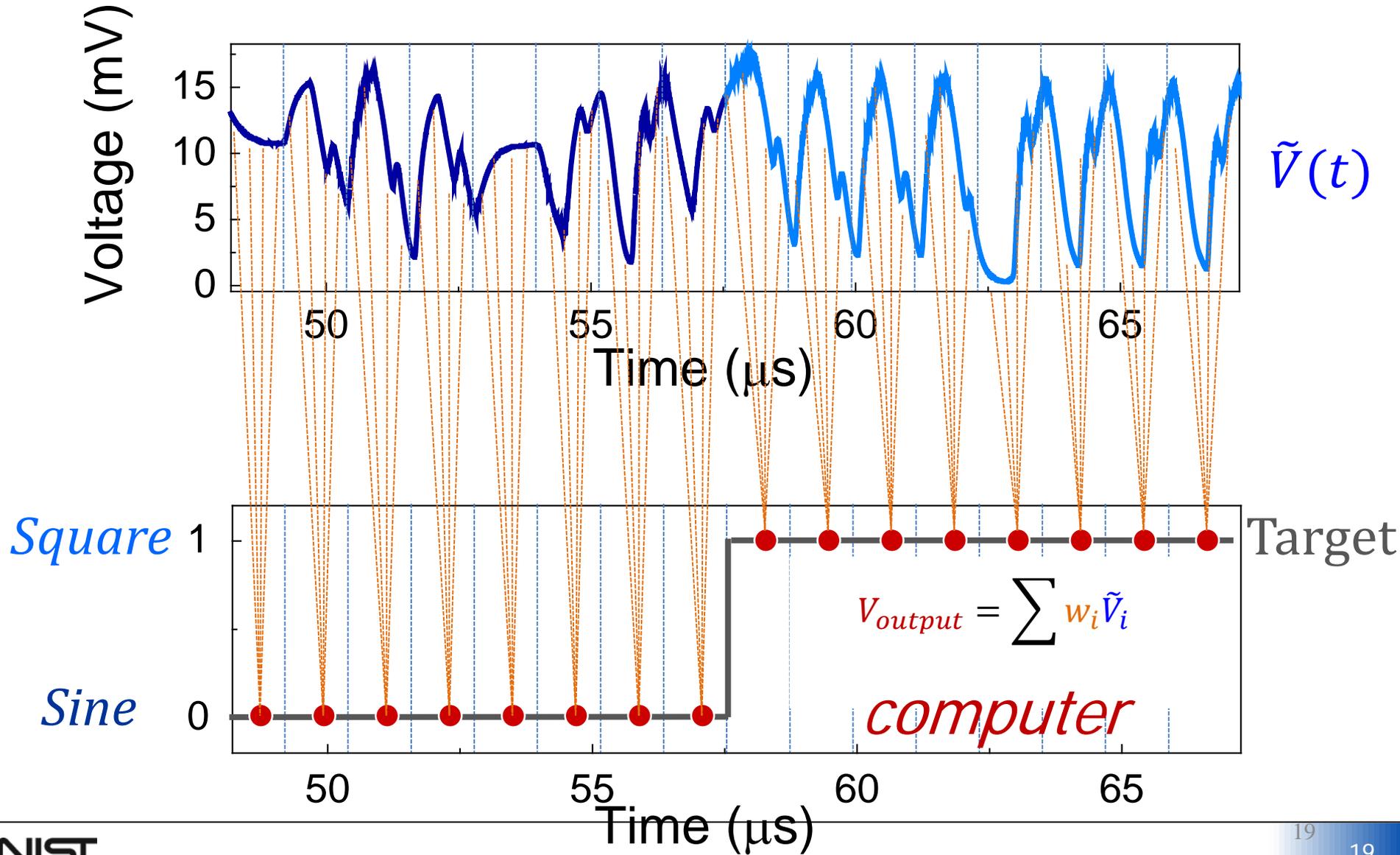


# Trajectories need to be grouped to be classified in sines and squares



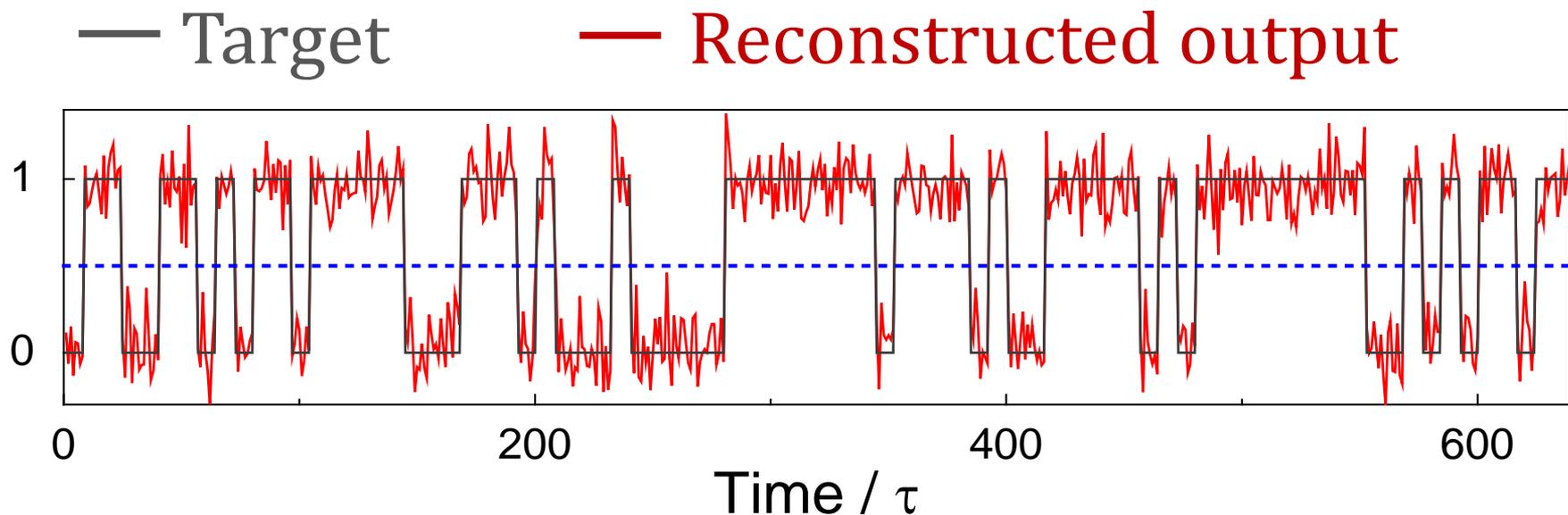
# Classification: constructing the output

0 for sine, 1 for square



Experimental result : RMS = 10%  
perfect classification of sines and squares

waveform with 80 randomly arranged sines and squares



$H = 3800 \text{ Oe}$ ,  $I_{\text{DC}} = 6.4 \text{ mA}$

$8 \tau$  per period, 24 nodes,  $\theta = 100 \text{ ns}$

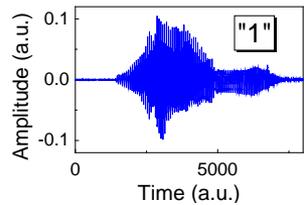
640 first  $\tau$  for training, 640 next  $\tau$  for classification

# Single oscillator reservoir computing

- Time multiplexed reservoir computing
- Sine/Square Identification (Intrinsic memory)
- **Spoken digit recognition (Non-linearity)**
- Sine/Square Identification (Delayed feedback memory)

# Spoken digit recognition (NIST TI-46 corpus)

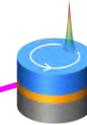
Input: audio file



Spectrogram  
or Cochlear

Pre-processed  
input

Oscillator



Recorded  
trace

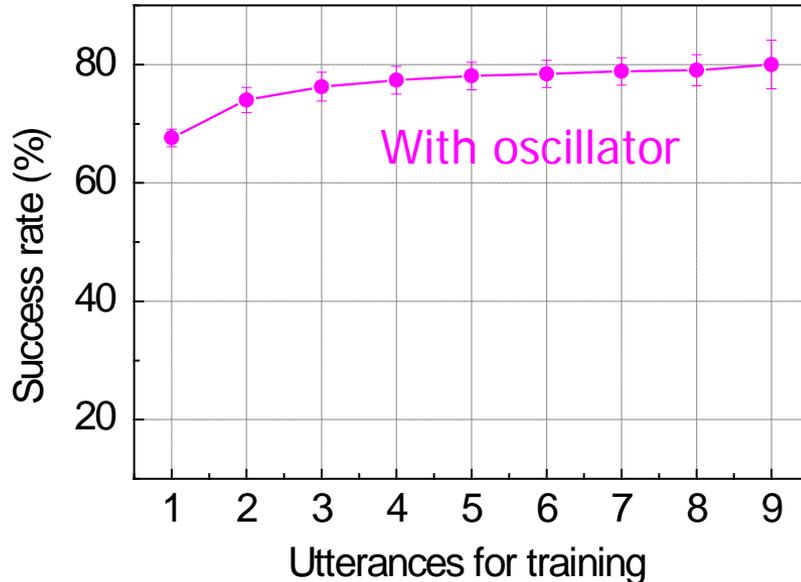
Output

"1"

*Acoustic features*    *Pre-processing*

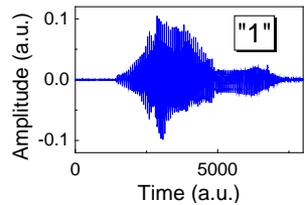
*Computer*

## Spectrogram



# Spoken digit recognition (NIST TI-46 corpus)

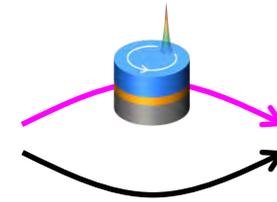
Input: audio file



Spectrogram  
or Cochlear

Pre-processed  
input

Oscillator



Recorded  
trace

Output

"1"

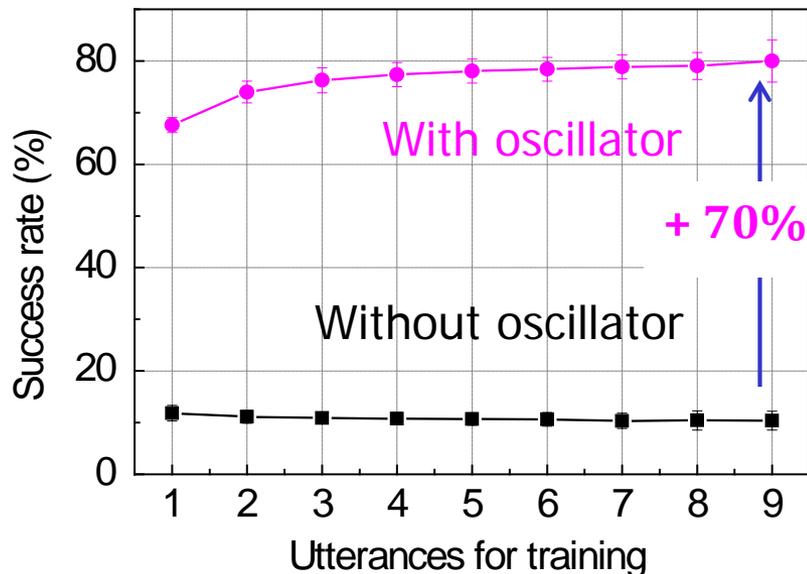
*Acoustic features*

*Pre-processing*

*No oscillator*

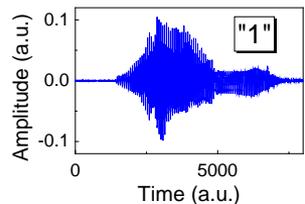
*Computer*

## Spectrogram



# Spoken digit recognition (NIST TI-46 corpus)

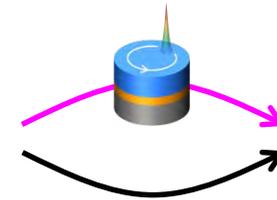
Input: audio file



Spectrogram  
or Cochlear

Pre-processed  
input

Oscillator



Recorded  
trace

Output

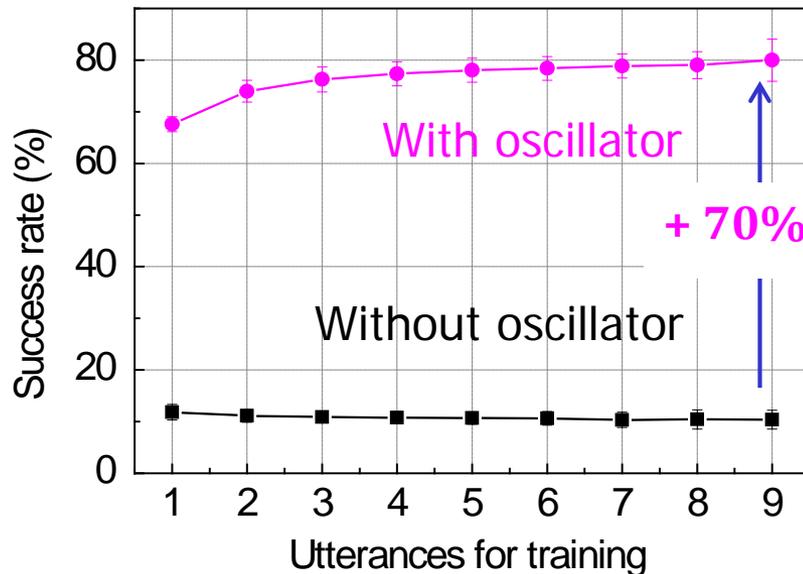
"1"

Acoustic features    Pre-processing

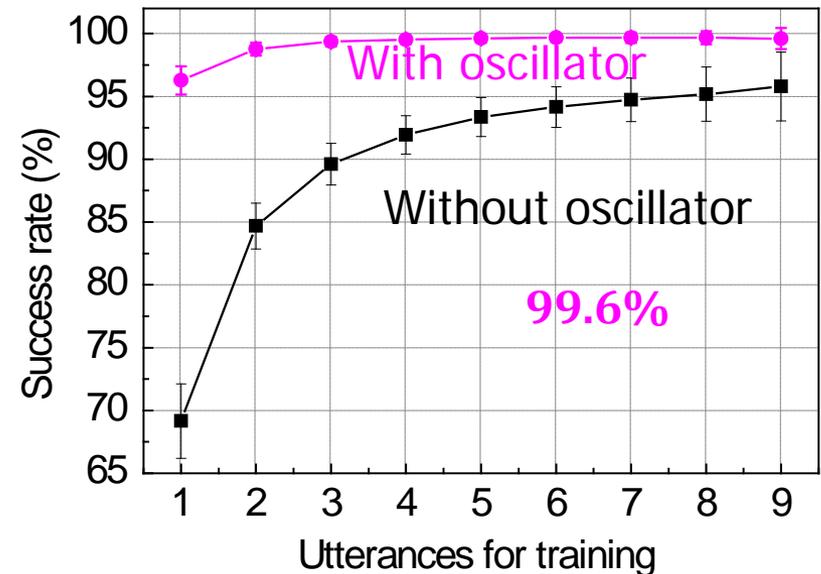
No oscillator

Computer

## Spectrogram



## Cochlear

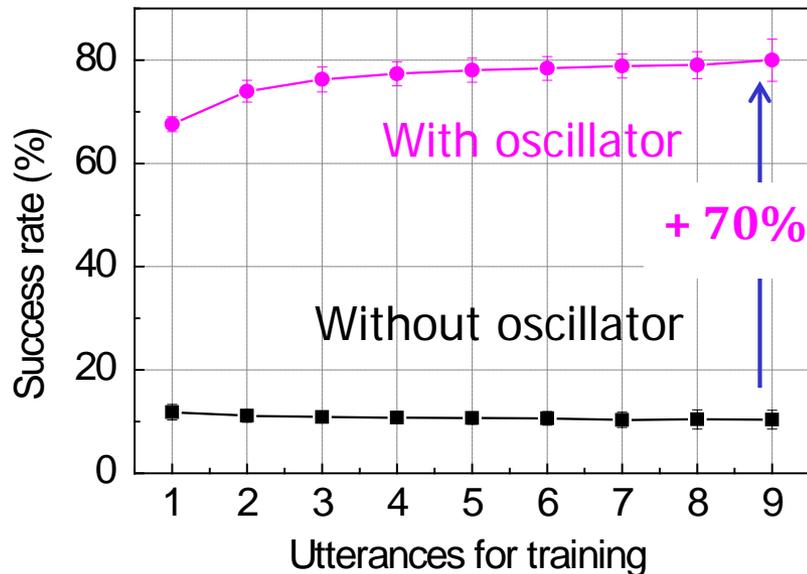


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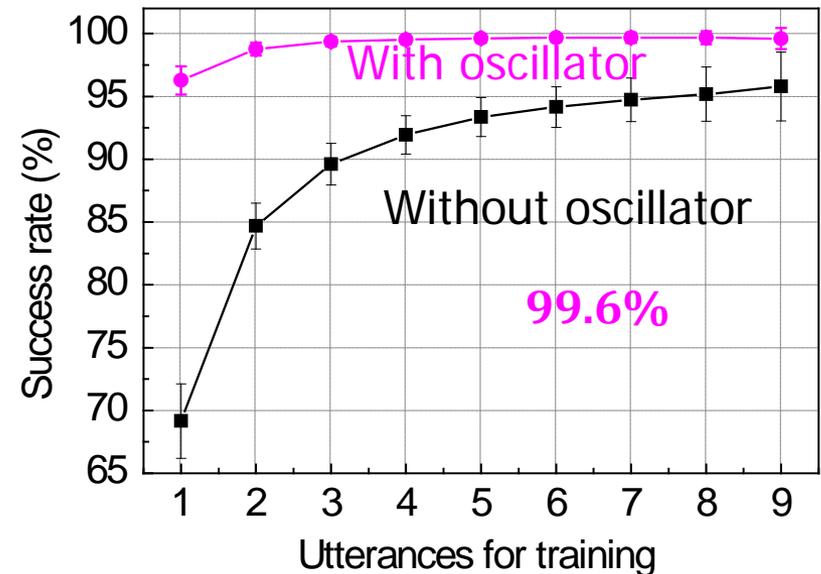
Jacob Torrejon-Diaz, Mathieu Riou, Flavio Abreu-Araujo, et al, Nature 547, 428-431 (2017)

**State of the art: 96 to 99.8 %**

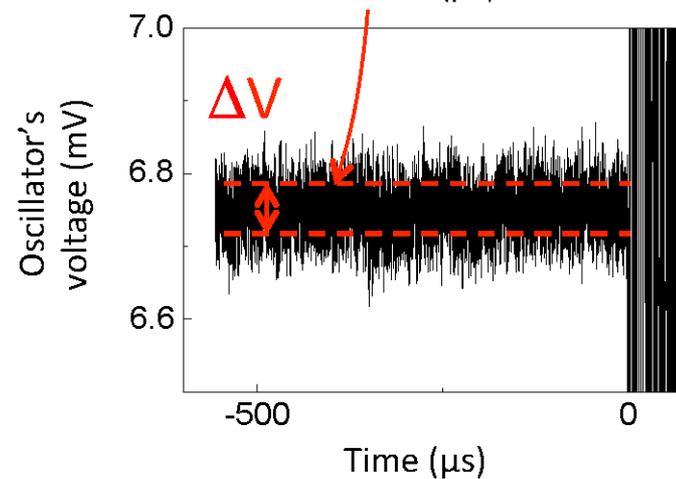
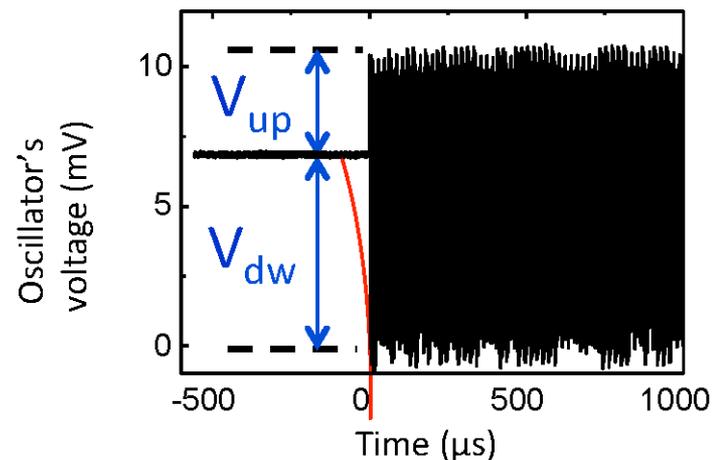
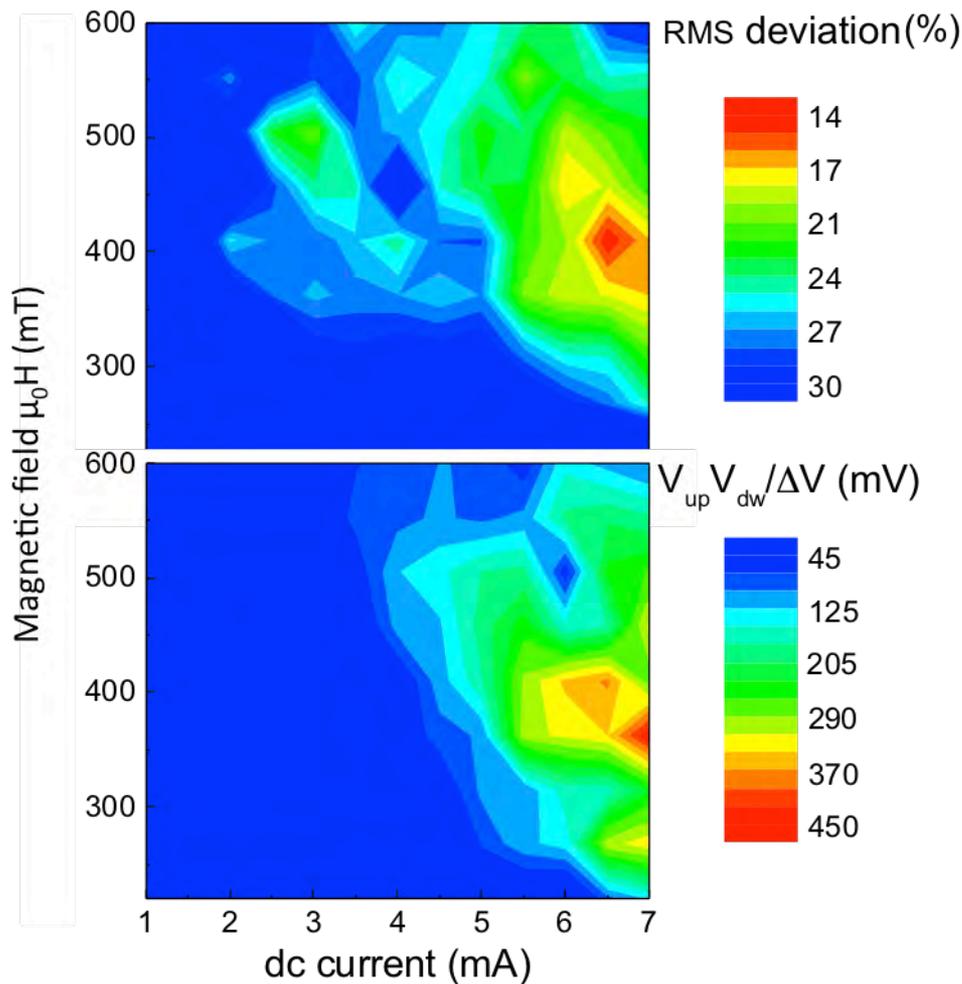
## Spectrogram



## Cochlear



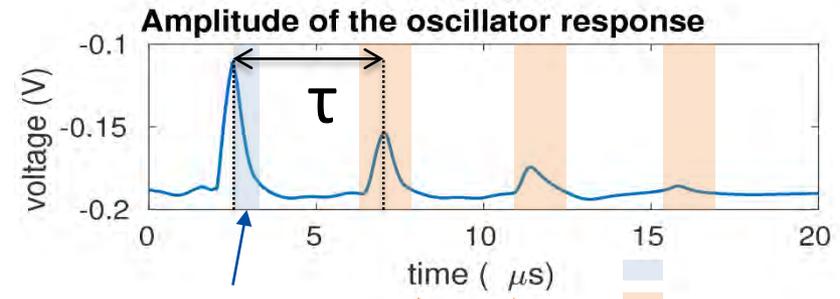
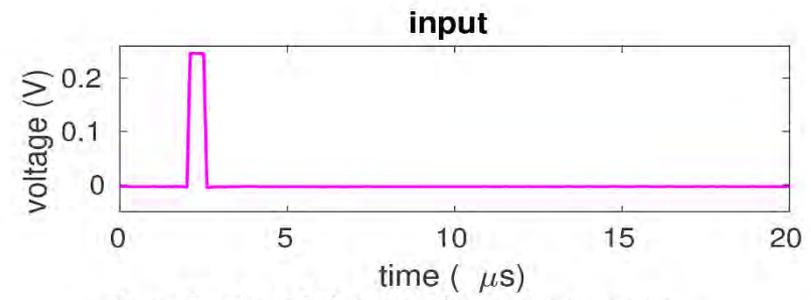
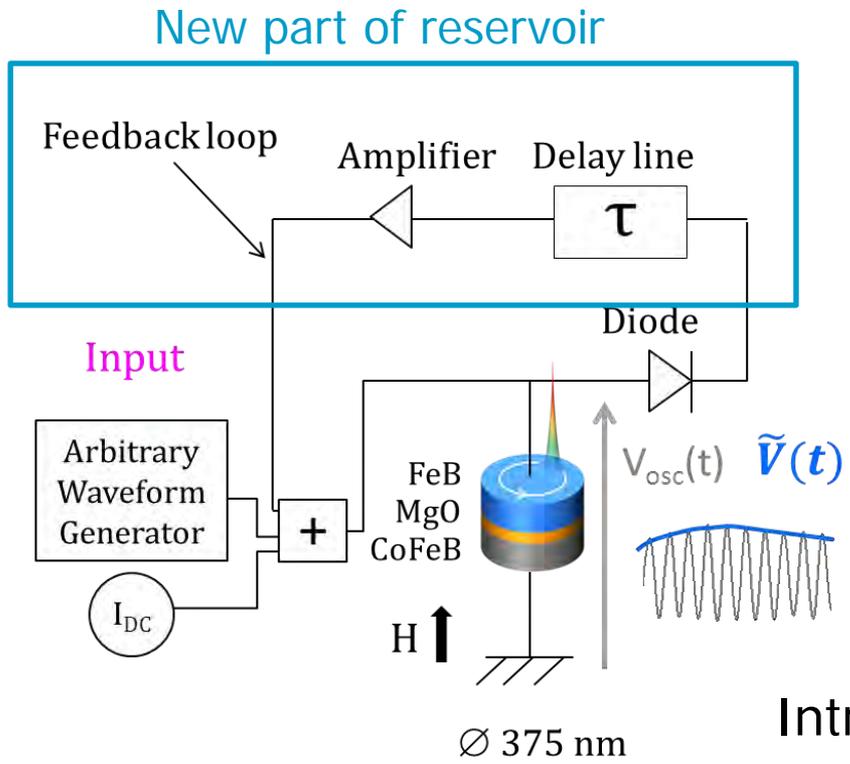
# Recognition results are sensitive to the noise, an optimal bias area is found



# Single oscillator reservoir computing

- Time multiplexed reservoir computing
- Sine/Square Identification (Intrinsic memory)
- Spoken digit recognition (Non-linearity)
- **Sine/Square Identification (Delayed feedback memory)**

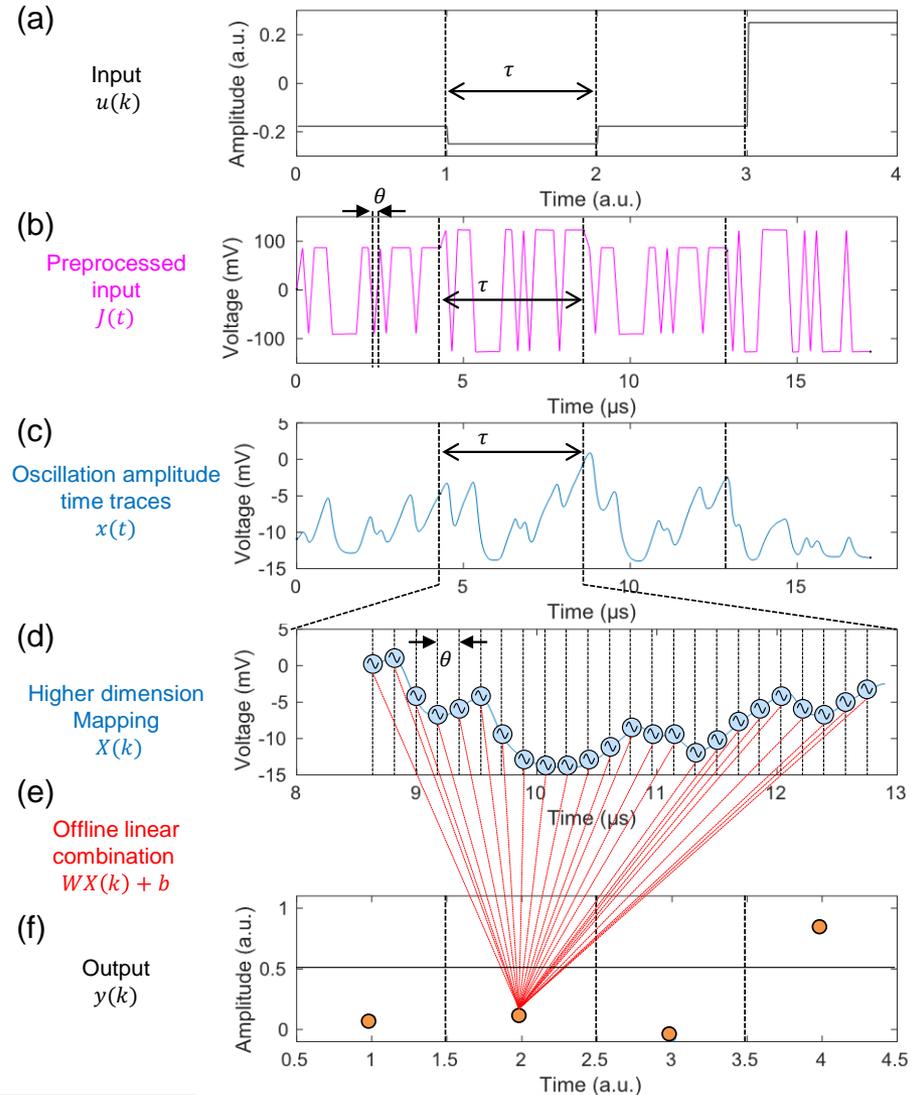
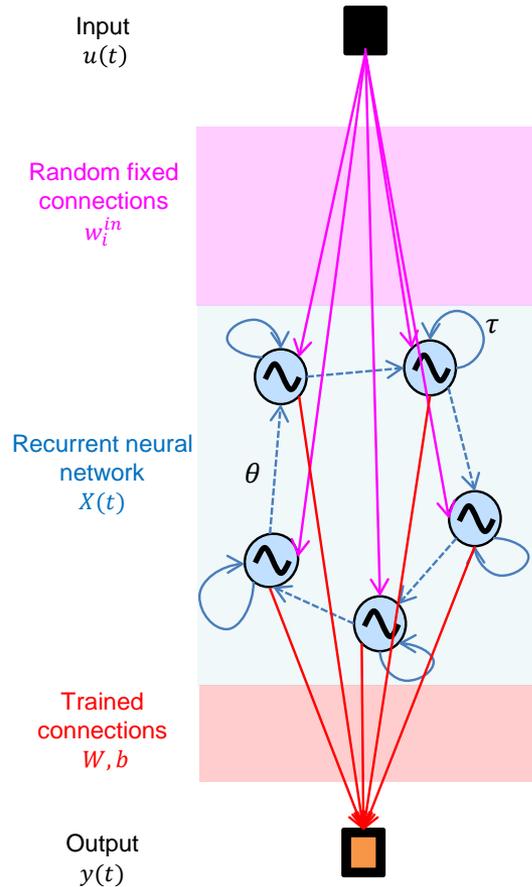
# Replace intrinsic memory with delayed feedback



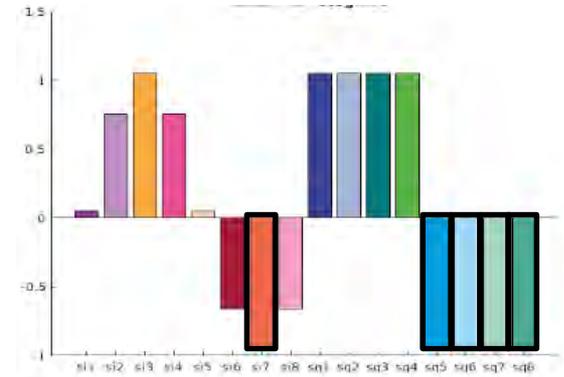
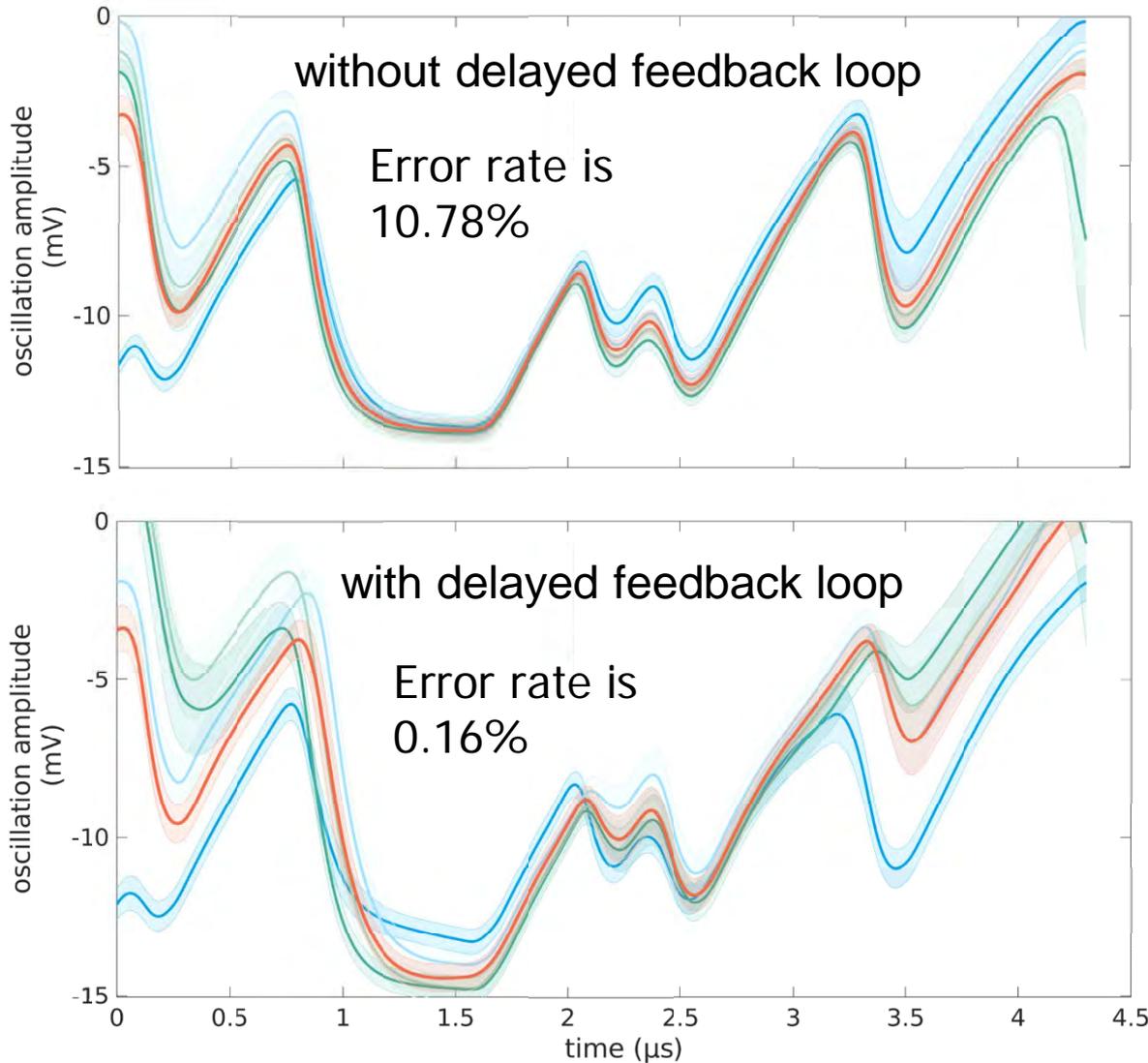
Intrinsic memory

Feedback memory

# To focus on feedback, nodes only couple to past states of themselves



# Feedback allows separation of similar inputs



# Single oscillator reservoir computing

- Augmenting CMOS based neuromorphic circuits with energy-efficient spintronic devices.
  - MTJs already accessible in stat-of-the-art BEOL CMOS
- Time multiplexed reservoir computing.
  - Memory (intrinsic or delayed feedback) allows context dependent discrimination.
  - Single oscillator achieves state of the art at spoken digit recognition – non-linearity.
- Where to?
  - Small low power oscillators.
  - Efficient coupling.
  - Appropriate algorithms