

# Compression of Spike Data using the Self-Organizing Map

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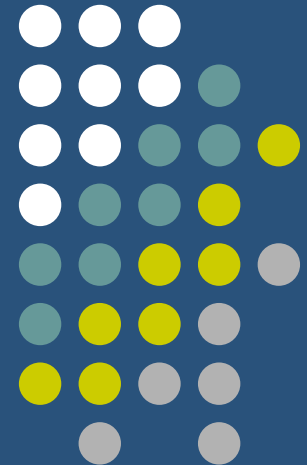
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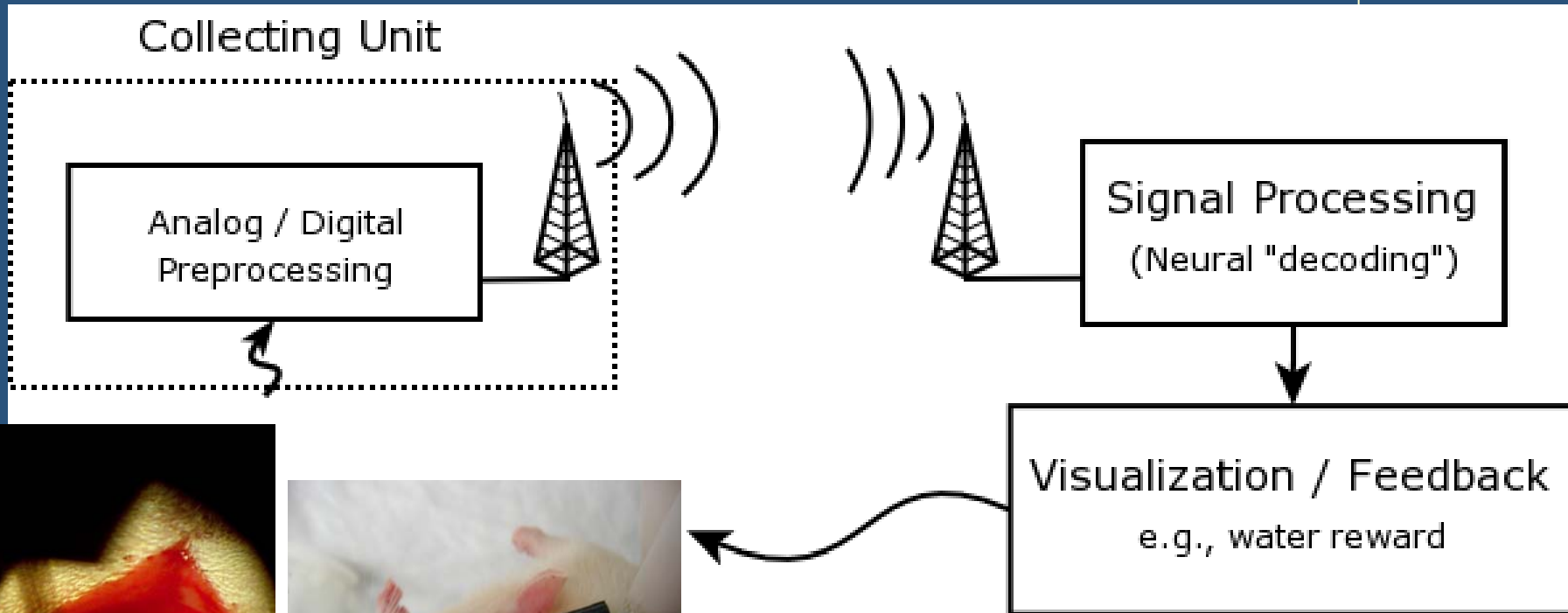




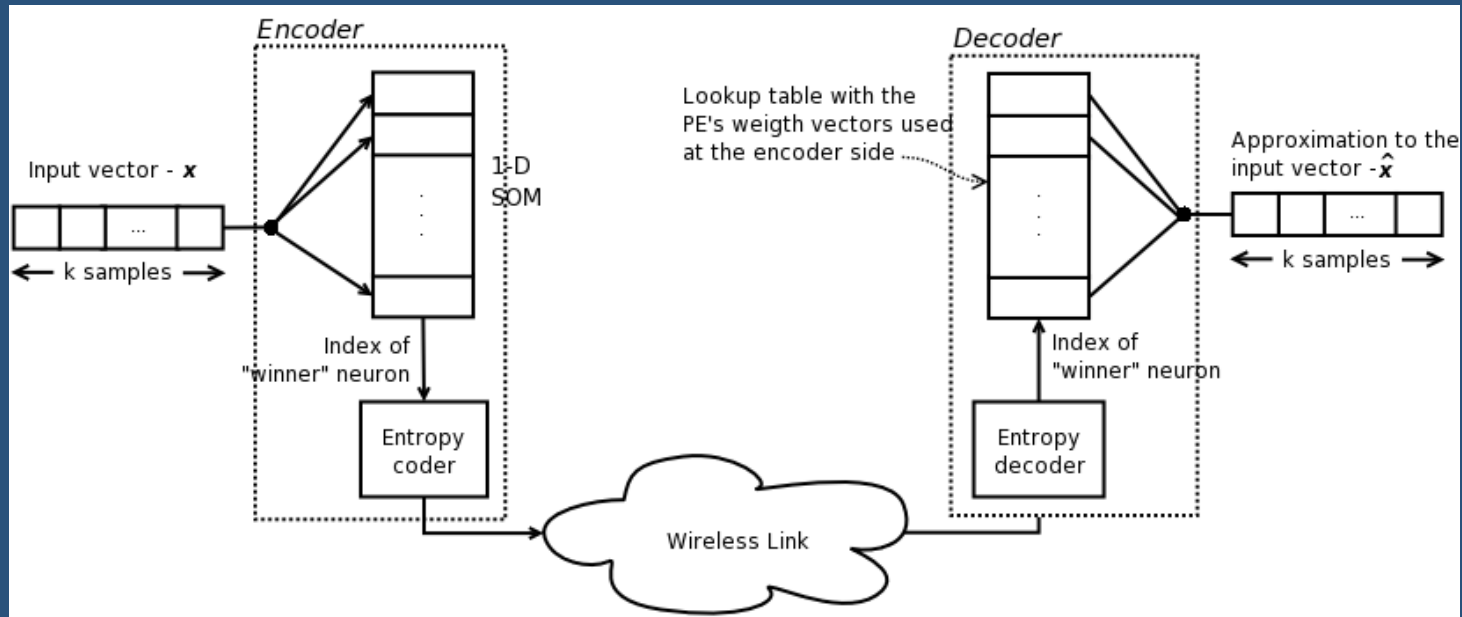
# Motivation

- Wireless links for electrophysiology have limitations on the available bandwidth.
  - 32ch @ 20kHz, 16 bits ~ 10Mbps
  - 32ch snips, 100 spikes/s, 32 samples, 16 bits ~ 1.6Mbps
  - 32ch bin, 8 bits ~ 240bps
- Gap in approaches: high or low data rates
- Preserve as much information as possible for sorting at a later time.

# The wireless Brain Machine Interface Paradigm



# Conceptual diagram of Encoding/Decoding process



- In this method, compression is achieved in two steps:
  - Quantization through the application of the SOM;  $k$  samples is represented by an index
  - Entropy coding of the indices.



# Compression through the SOM

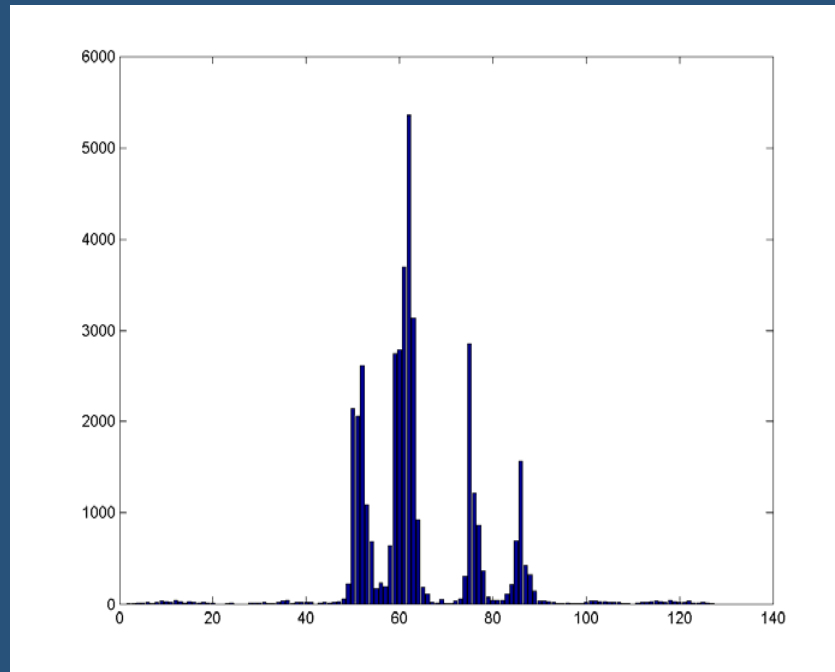
- Local framework - preserve the topology of the data using neighborhood links between PEs
- Compression through the SOM is *lossy*:
  - Many-to-one mapping;
  - Choose nearest neighbor to minimize the quantization error.
- Compression ratio is constant:
  - If each sample has  $b$  bits, with  $k$  samples per vector, and a SOM with  $N$  vectors, the compression ratio (just due to the SOM) is

$$\frac{kb}{\lceil \log_2 N \rceil} : 1$$

# Entropy coding



- Why use entropy coding?
  - Because of non-uniform distribution of the *indices* of the firing PEs.





# Entropy coding

- Calculate the entropy of the indices, at the output of the SOM, with

$$h = -\sum_{i=1}^N p_i \log_2 p_i$$

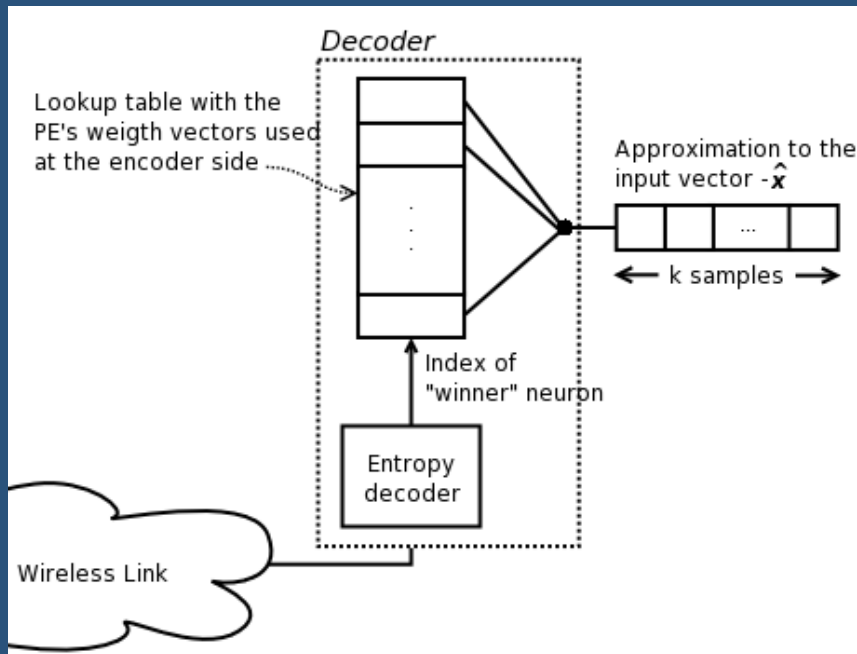
where  $p$  is the “probability” of the  $i^{\text{th}}$  index.

- Final compression ratio is:

$$(k \times N \times 8) / h$$

Vector length ( $k$ ), # of PEs ( $N$ )

# Decoding process



1. Entropy decode;
2. Use index to fetch vector of winning PE in table;
3. Concatenate vector to the output.

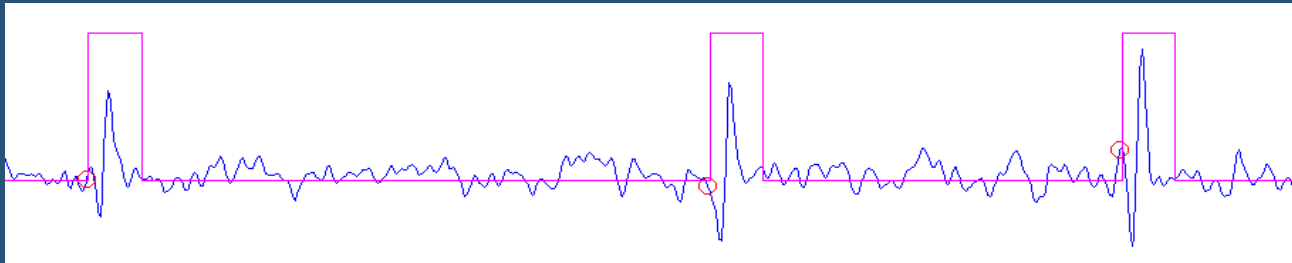




# Training/Testing Sets

- Training:

- Equal amount of training samples from each channel;
- Selected *spikes only* assuming width of spike of 27 samples;



- $2 \times (4000 \text{ spikes}) \times 27 + (2\% \text{ noise samples}) \approx 110,000$  samples;
- Designed to reconstruct spikes, but be noise aware.

- Testing:

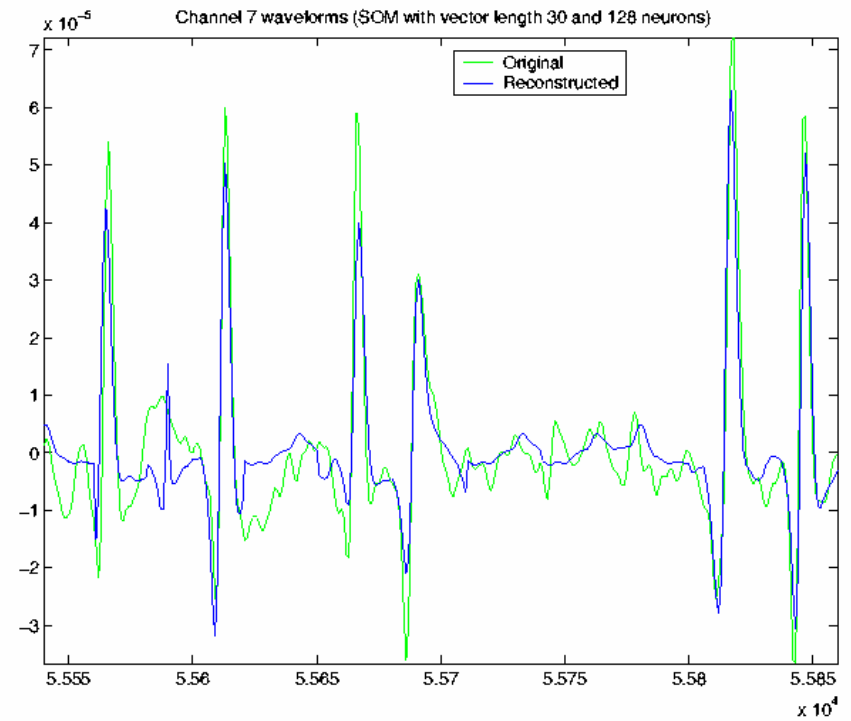
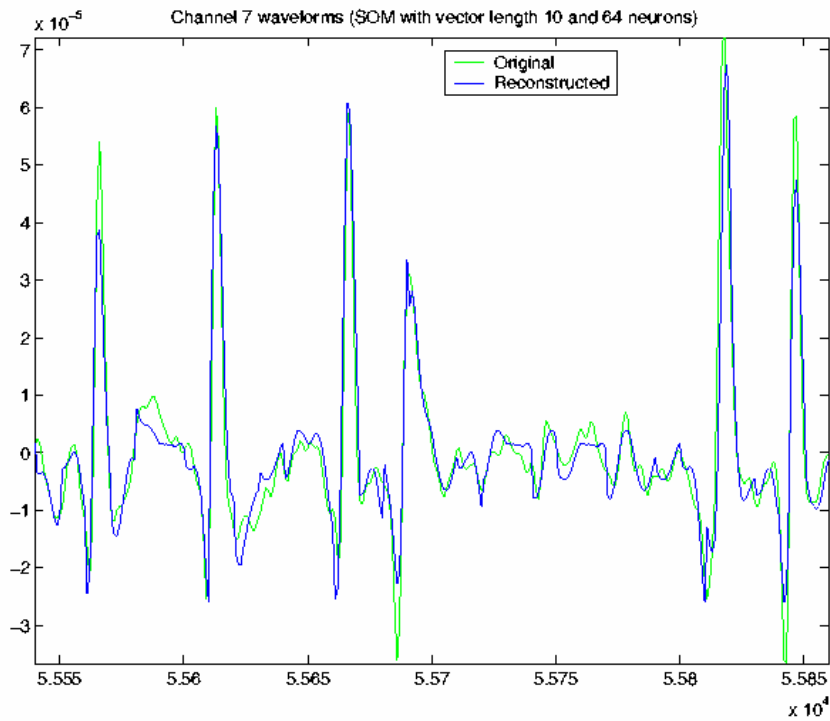
- 400,000 samples from each channel  
(outside the training set).



# Results

Input dimension	No. of PEs	Channel 6		Channel 7	
		Bitrate (kbits/sec)	Compression ratio	Bitrate (kbits/sec)	Compression ratio
10	32	4.14	37.7 : 1	3.76	41.6 : 1
	64	5.44	28.7 : 1	4.96	31.5 : 1
	128	6.73	23.2 : 1	6.18	25.3 : 1
20	64	1.71	91.6 : 1	1.27	123.2 : 1
	128	2.21	70.7 : 1	1.95	80.0 : 1
	256	2.75	56.9 : 1	2.34	66.9 : 1
30	128	1.02	152.8 : 1	0.84	186.7 : 1
	256	1.44	108.7 : 1	1.22	128.4 : 1

# Reconstructions vs Vector Length





# Conclusions

- Compression ratio provides a balance between the fidelity of the reconstruction versus the desired/allowable bitrate in the communication link.
- Future work:
  - Exploring changes to the SOM to minimize reconstruction error;
  - Quantify the effects of quantization in spike detection and spike sorting.