# **Embedded Hardware Spiking Neural Network for UWB Bladder Volume Classification**

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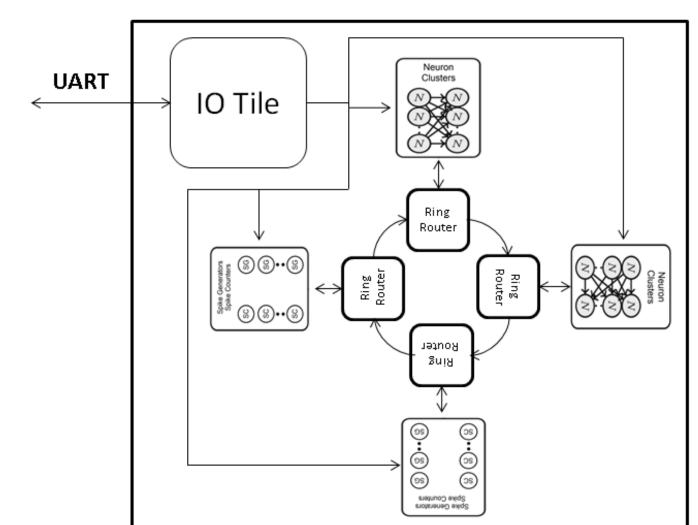
### Introduction

This project aims to develop an embedded SNN bladder volume classifier using previously reported Ultra Wideband (UWB) Radar Target Signatures of a bladder phantom [1].

The proposed UWB dataset SNN classifier is being prototyped on the EMBRACE Embedded Hardware SNN platform [4].

# **EMBRACE Embedded Hardware SNN**

EMBRACE Embedded SNN on Spartan6 FPGA



Prototyped on Xilinx • Spartan6 FPGA

### **UWB Bladder Volume Phantoms**

- Bladder widths are varied
- UWB pulse is transmitted

Fig.1. Diagram of EMBRACE [4] Embedded HW SNN on Spartan6

- Fixed local spike latency
- Configurable topology ۲
- Fully connected clusters •
- 64 LIF neurons, 32 spike generators and counters
- Classifier is evolved using a host GA (Fig. 3)
- Previously tested with the •
  - Wisconsin Cancer dataset

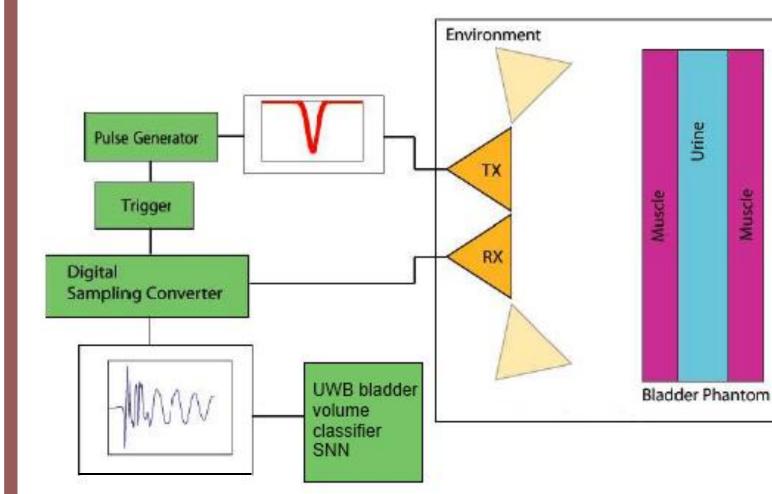
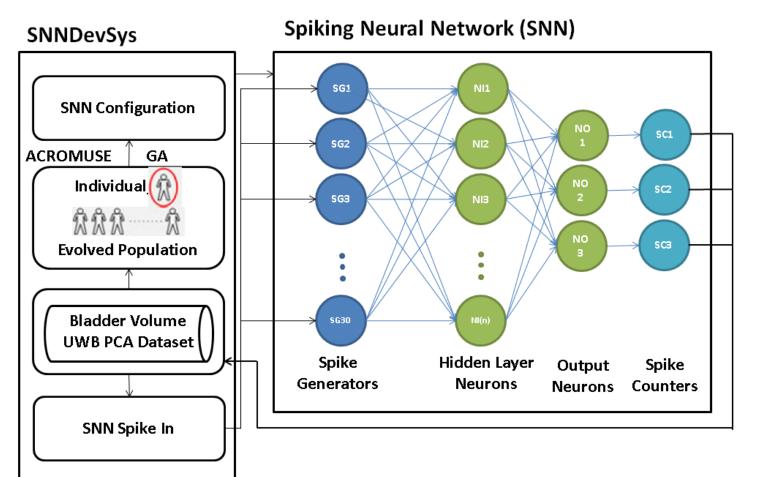


Fig.2. Block diagram of the UWB bladder system setup. [1]

into the phantom and the reflection recorded [1]

- Signals are preprocessed
- Features are extracted from resulting Radar Target Signatures by Principle Component Analysis • Dataset has 15 features and
  - 180 observations

# **GA for Classifier Evolution**

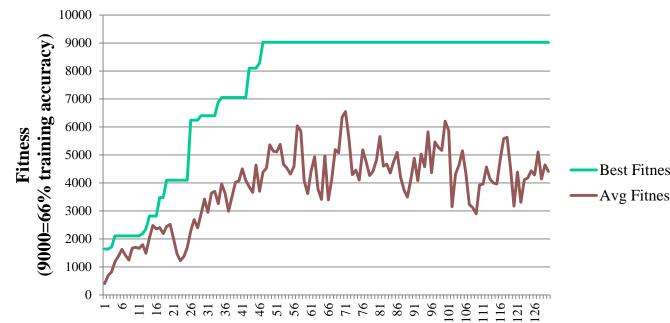


- Proof of concept SNN classifier in SW only
- 2 layer feed-forward network classifies each observation consisting of 30 inputs from dataset
- ACroMuSe [3] adaptive ٠ GA evolves a classifier
- SNNs are evaluated in

### **Preliminary Results**

SNN	<b>Fisher's IRIS</b>	<b>BVM PCA</b>
SW	91.332%	67.778%

Fig.4. Table of Software Classifier results



- Standard Fisher's Iris dataset was used as a test for the SW system
- Datasets were shuffled and split into 10 sets
- Both classifiers were evolved with 4 hidden layer neurons

#### Fig.3. Block diagram of hardware/software SNN UWB bladder volume classifier and the host ACROMUSE SNN evolution system (SNNDevSys).

a Software SNN

Spike rate coding used •

#### Generation

Fig.5. Graph of typical fitness evolution for BVM dataset

**BVM** classification

results indicate

incomplete classifier

### References

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[3] Mc Ginley, B., Maher, J., O'Riordan, C., & Morgan, F. (2011). "Maintaining healthy population diversity using adaptive crossover, mutation, and selection." IEEE Transactions on Evolutionary Computation, 15(5), 692-714.

[4] Pande, S., Morgan, F., Smit F., Bruintjes, T., Rutgers, J., Cawley, S., Harkin, J., McDaid, L., "Fixed Latency On-Chip Interconnect for Hardware Spiking Neural Network Architectures", manuscript submitted for publication to Parallel Computing July 17, 2012

# Conclusions

- IRIS classification results indicate GA and Software SNN system is functional
- UWB BVM PCA data appears to be more difficult to solve by GA and SNN • than Iris dataset with only 68% accuracy opposed to 91.3%
- BVM may require larger dataset for accurate classification

# **Future Work**

- Evolution of EMBRACE Embedded Hardware classifier
- Investigation of suitability of alternative SNN training methods including SpikeProp and SWAT for the purpose of lifelong learning
- Development of anatomically correct pelvis and bladder phantom



### Acknowledgements



