



Design and Control of a Photonic Neural Network Applied to Low-Latency Classification

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Motivation: MDM Benefits

- Increases Aggregate Bandwidth:** Allows more information to be carried in a single wave-guide.
- Increases Network Size:** Allows for more neuron channels in a single sub-network.

MDM Challenges

- Precise Geometry:** Coupling requires a certain waveguide width and length.
- Weighting Complexity** to cancel Intermodal Mixing.

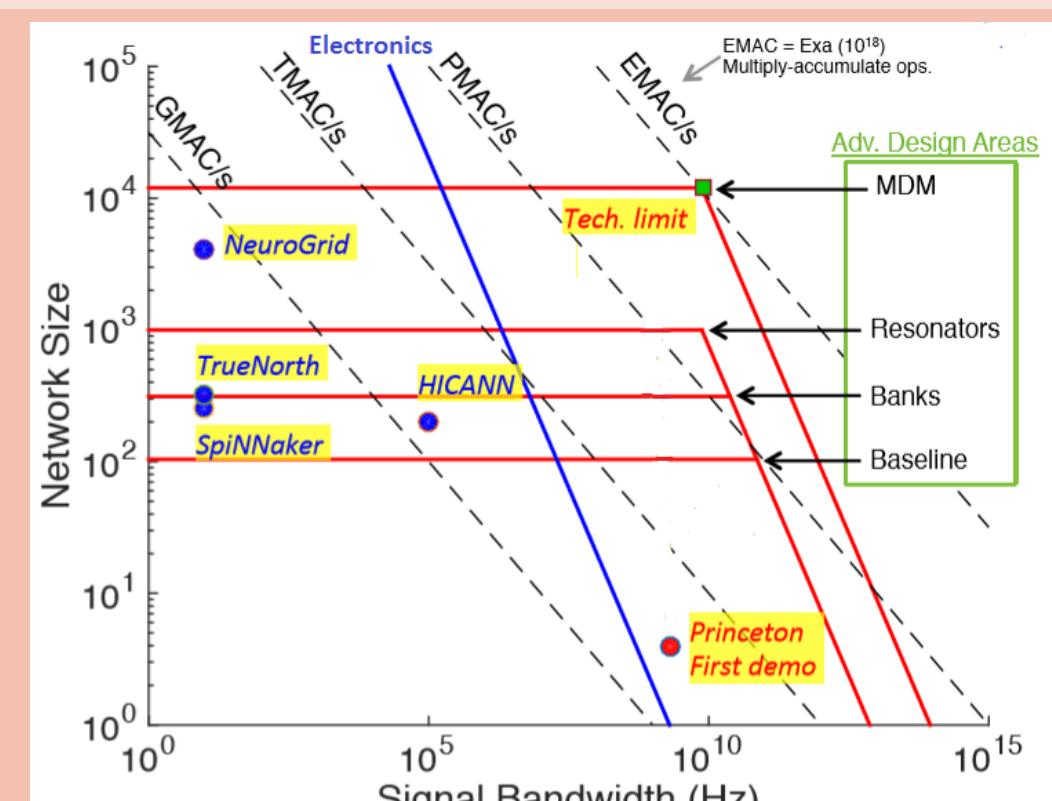


Figure 1: Network Size / Bandwidth Tradeoff for previous neural networks.

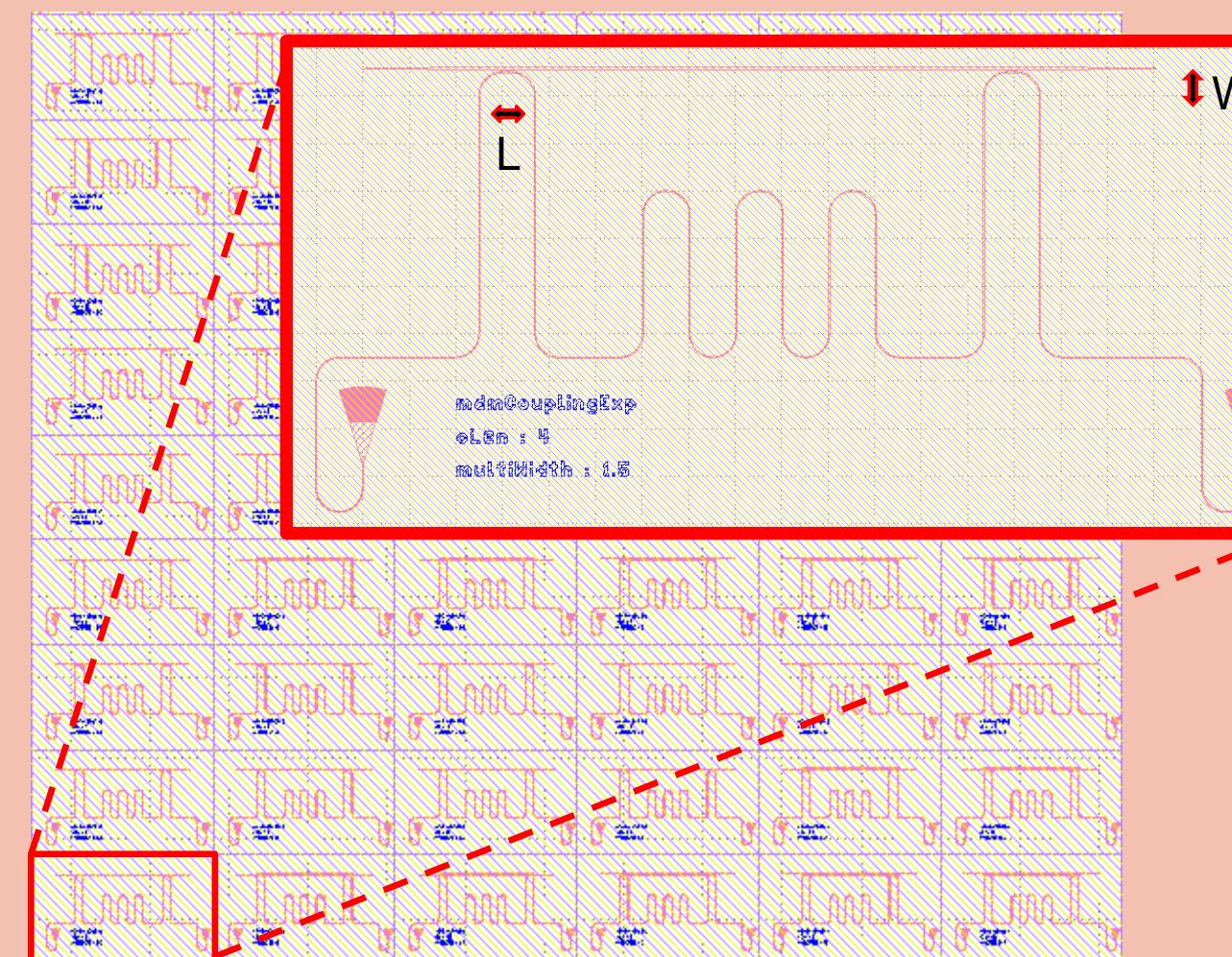


Figure 2 (above): Asymmetric Mach-Zehnder Interferometer Design. Width (W) and Coupling Length (L) need to be optimized for maximum coupling.

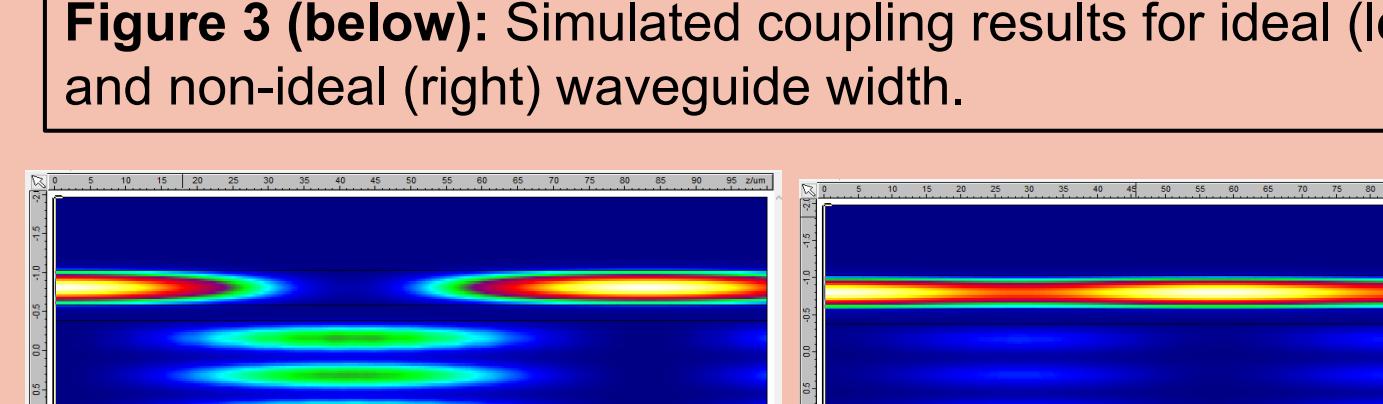


Figure 3 (below): Simulated coupling results for ideal (left) and non-ideal (right) waveguide width.

Mode Division Multiplexing

Experimental Design

- Goal:** Determine optimal coupling width and length for each mode.
- Principle:** Asymmetric Mach-Zehnder Interferometer ($r = \text{coupling percentage}$)
- $T = r^2 + (1-r)^2 + 2r(1-r)\cos(\lambda\Delta L)$
- Extinction Ratio: $2^*r*(1-r)$,
- $r=0$ when W is not optimal: 0% Coupling
- Adjust W to maximize extinction ratio:** $r=0.5$
- Adjust L to minimize extinction ratio:** $r=1$

Simulated Results

Since a null extinction ratio could actually correspond with 100% or 0% coupling, we can use simulation data to determine which is more likely.

(All μm)	TE_0	TE_1	TE_2	TE_3
W	0.5	0.95	1.4	1.9
L	25	35	41	47

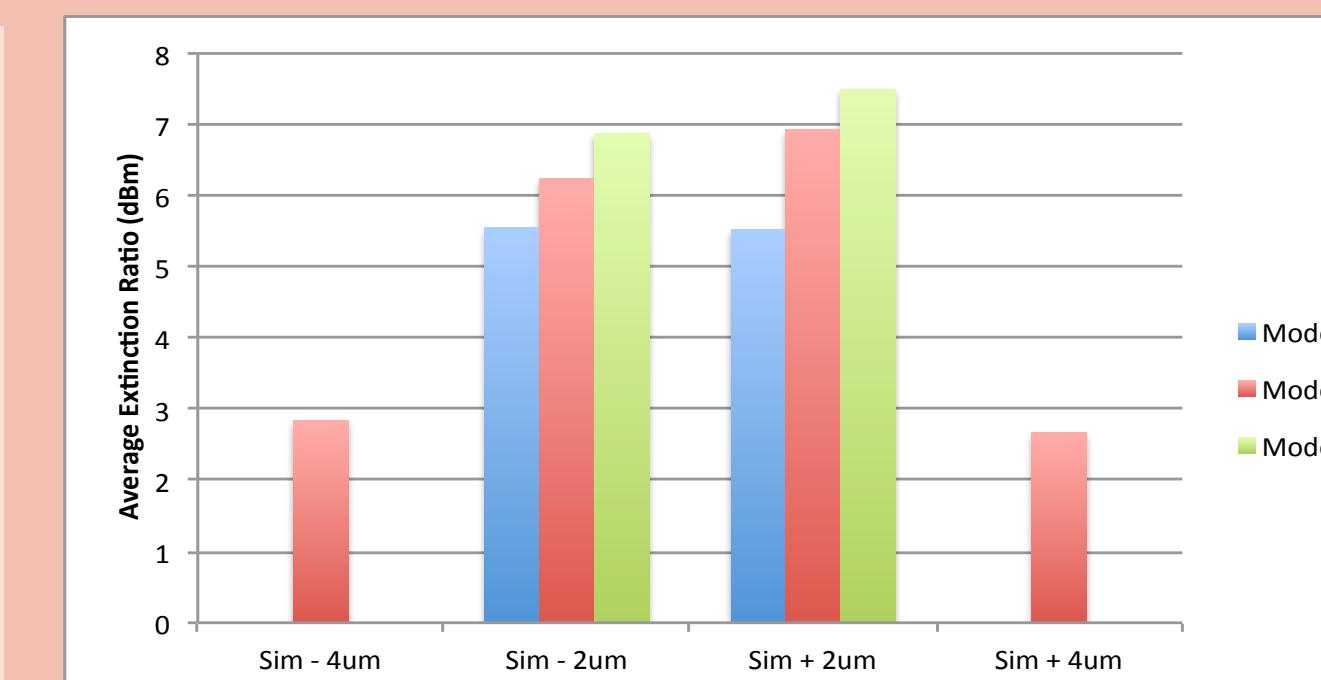
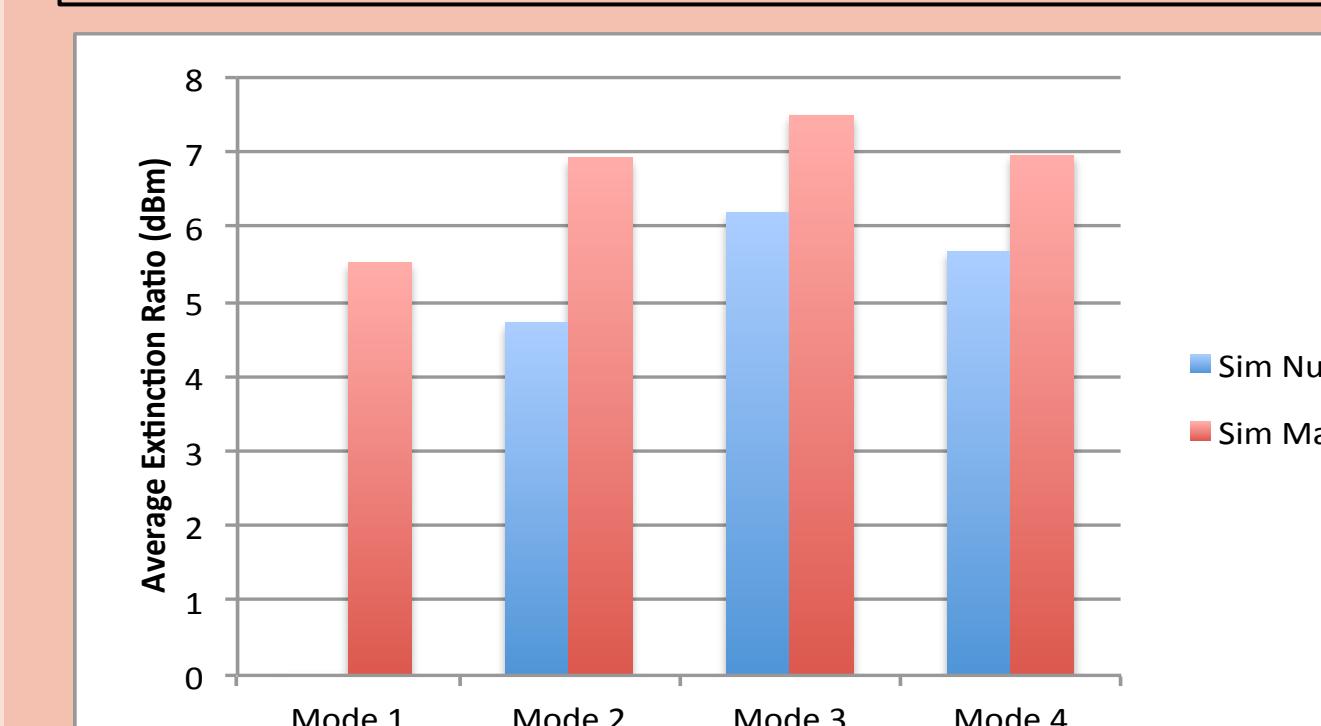


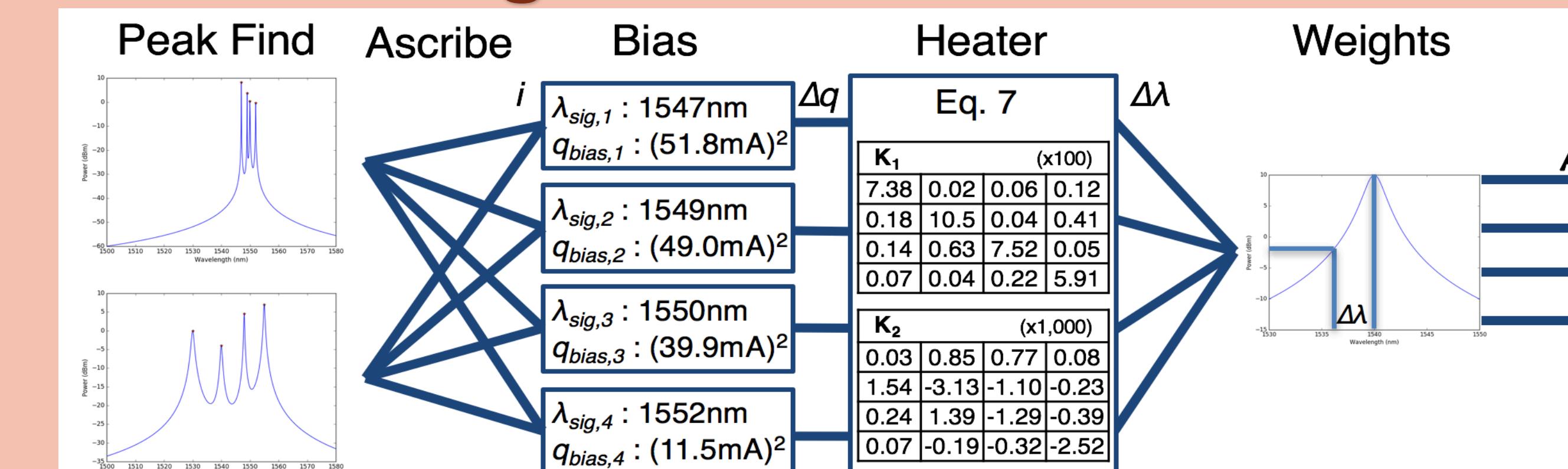
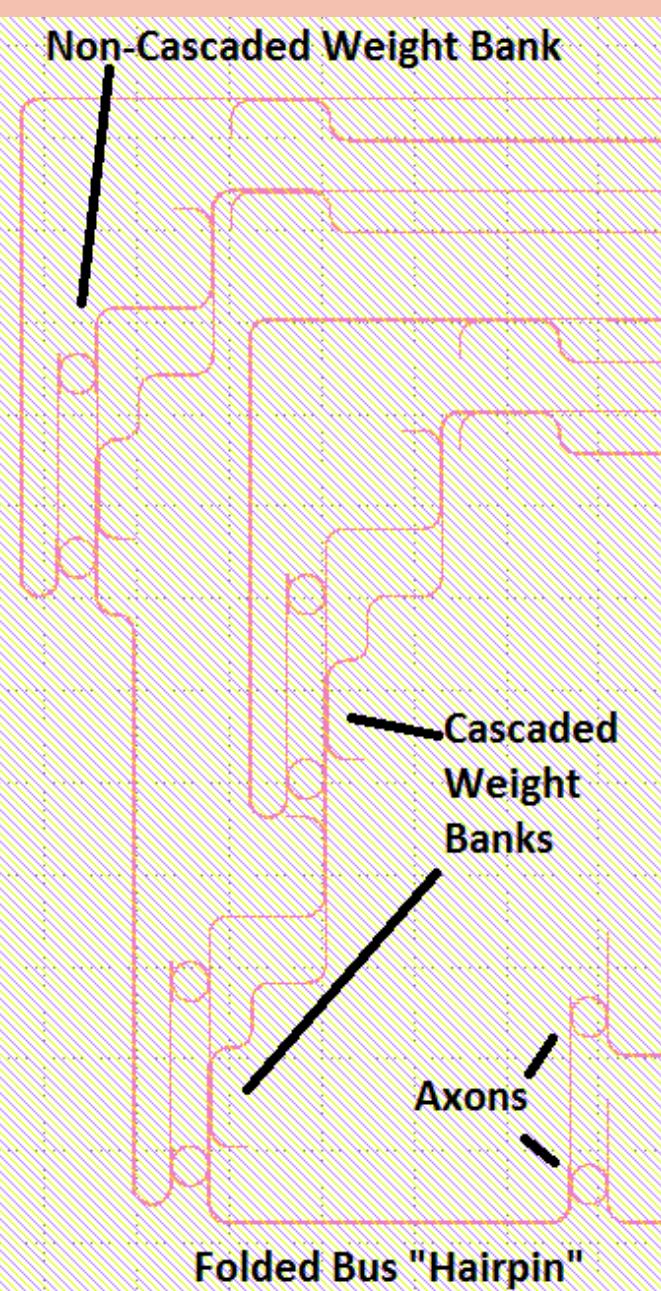
Figure 4: Experimental results. Above shows maximum extinction ratio near optimal width. Below shows lower extinction ratio near optimal coupling length. No data = no extinction ratio above noise floor.



Results and Next Steps

- Successfully showed good coupling near simulated results.
- Parameter space was too coarse: Steep drop-off in extinction ratio by $4\mu\text{m}$ off optimal width.
- Next Step:** Good enough to begin test on full MDM Weight Banks on-chip.
- Next Step:** Re-run experiment with finer parameter space.

Cascaded Weight Bank Control



Motivation: Network Topology

- Folded Bus "Hairpin" Topology:** Space-efficient layout of MDM networks.
- Cascaded Bank:** Allows each neuron to pull a percentage of power off the bus.
- Challenge:** How to adjust weights in network given cross-talk and indirect sensing.

Figure 6: Two MDM neurons in a hairpin topology.

Control Process

- Peak Detection:** Detect inputs and the Lorentzian Peak of each micro-ring.
- Ascription:** Assign each micro-ring to an input.
- Bias:** Bias each input to micro-ring peak using PI Loop.
- Heater:** Measure thermal cross-talk between rings.
- Weights:** Use peak shape to convert to weight vector.

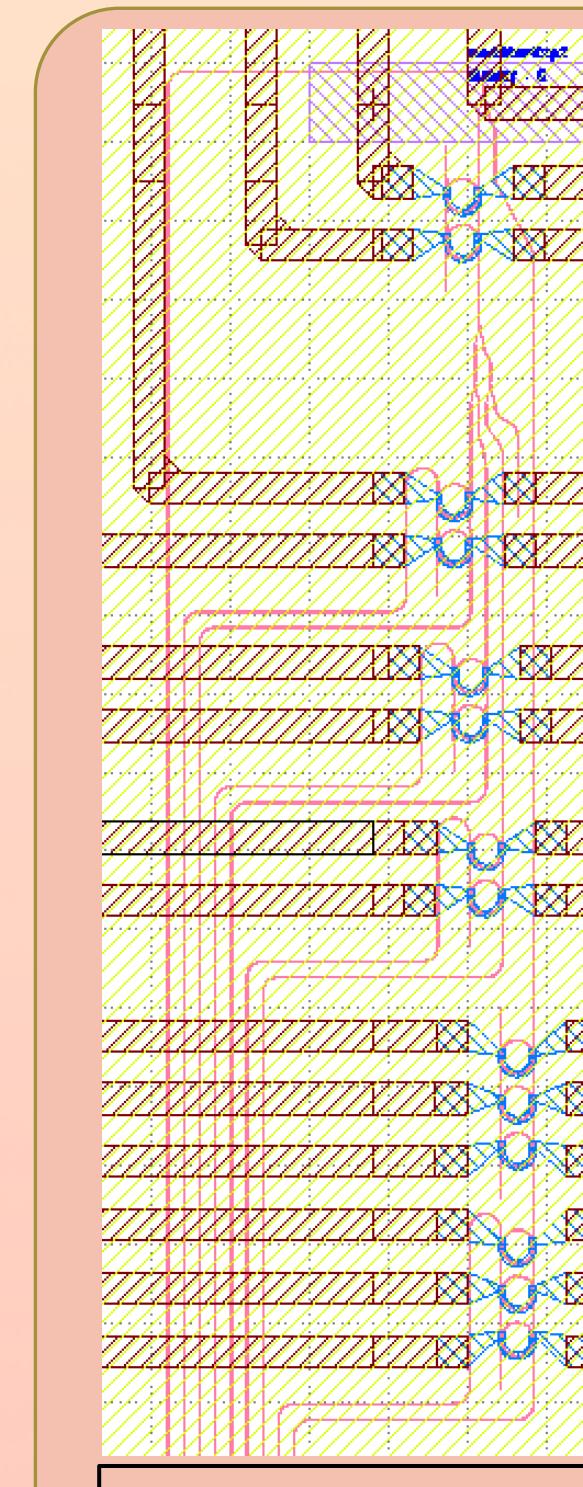


Figure 7: 2-3-1 Layer Feed Forward Neural Network

Classification Applications

Negative Result: Quantum Bit Classification

- Question:** Could a recurrent neural network, such as a Long Short-Term Memory (LSTM) network, increase the fidelity of quantum bit read-outs compared to current thresholding techniques?
- No, because the read-out trajectories are completely linearly separable.** (Fig. 8)
- Transient effects are drowned out by high levels of white noise.
- With these problems combined, averaging and thresholding is close to optimal.

Possible Next Steps

- RF Modulation Scheme Classification:** Track non-linearly separable trajectories through I-Q space.
Challenge: LSTM networks are non-trivial to implement in our network architecture.
- Low-Latency RF Demodulation:** Can be done with a Feed-Forward Network (Fig. 7).

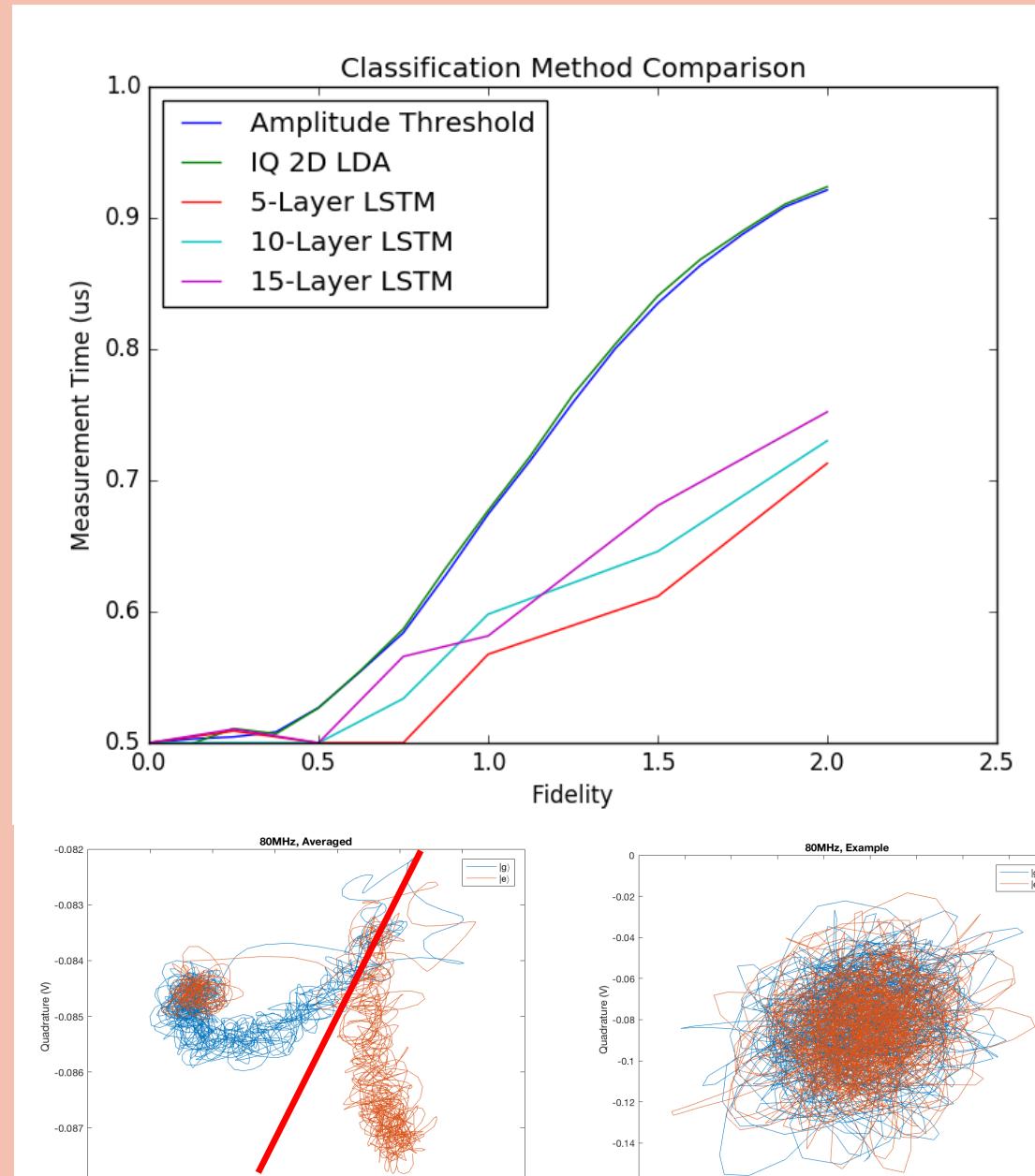


Figure 8: (top) Linear Thresholding vs. Neural Network Classification. (left) Averaged Qubit Data, showing linear separation. (right) Example of single trajectory, showing low SNR.