

# Non-Linear Digital Self-Interference Cancellation for In-Band Full-Duplex Radios Using Neural Networks

Alexios Balatsoukas-Stimming

Telecommunications Circuits Laboratory

Ecole polytechnique fédérale de Lausanne (EPFL), Switzerland

## In-Band Full-Duplex Communications

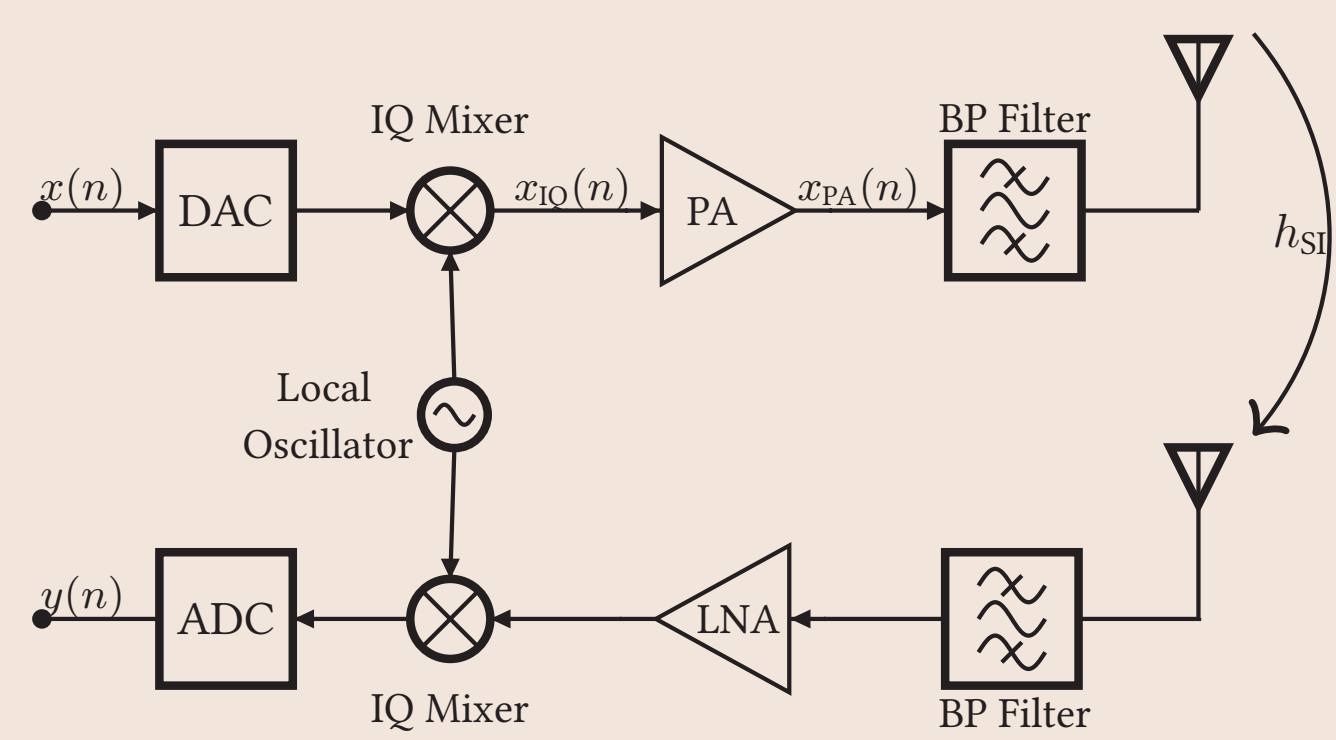
### In-band full-duplex communications:

1. Up to twice the throughput wrt TDD & FDD
2. No additional bandwidth
3. No wasted time or frequency resources

### Fundamental Challenge

Self-interference (SI) signal is **much stronger** than the desired signal and needs to be cancelled!

- **In principle**, SI cancellation should be easy since the digital transmitted signal is known
- **In practice**, the digital signal does not tell the whole story!



### Why is Cancellation Difficult?

Analog components introduce strong **non-linear** effects!

## Polynomial Non-Linear Digital Cancellation

Captures IQ imbalance, PA non-linearities (up to order  $P$ , memory  $M$ ), and channel memory ( $L$ ) using a complex polynomial model:

$$y(n) = \sum_{\substack{p=1, \\ p \text{ odd}}}^P \sum_{q=0}^p \sum_{m=0}^{M+L-1} h_{p,q}(m) x(n-m)^q x^*(n-m)^{p-q}$$

Number of parameters:  $n_{\text{poly}} = (M+L) \left(\frac{P+1}{2}\right) \left(\frac{P+1}{2} + 1\right)$

### Complexity Analysis

#### Best-case scenario assumptions:

- Basis function computation is free
- Complex mults require 3 real mults and 0 real additions

#### Complexity:

1. Real additions:  $n_{\text{ADD,poly}} = 2(n_{\text{poly}} - M - L - 1)$
2. Real multiplications:  $n_{\text{MUL,poly}} = 3(n_{\text{poly}} - M - L)$

## Neural Network Non-Linear Digital Cancellation

### Two-step digital cancellation:

1. Linear digital cancellation:  $\hat{y}_{\text{lin}}(n) = \sum_{m=0}^{M+L-1} \hat{h}_{1,1}(m) x(n-m)$
2. Train a neural network to cancel:  $y_{\text{nl}}(n) \approx y(n) - \hat{y}_{\text{lin}}(n)$

Single-layer NN with  $n_h$  neurons and ReLU activation functions.

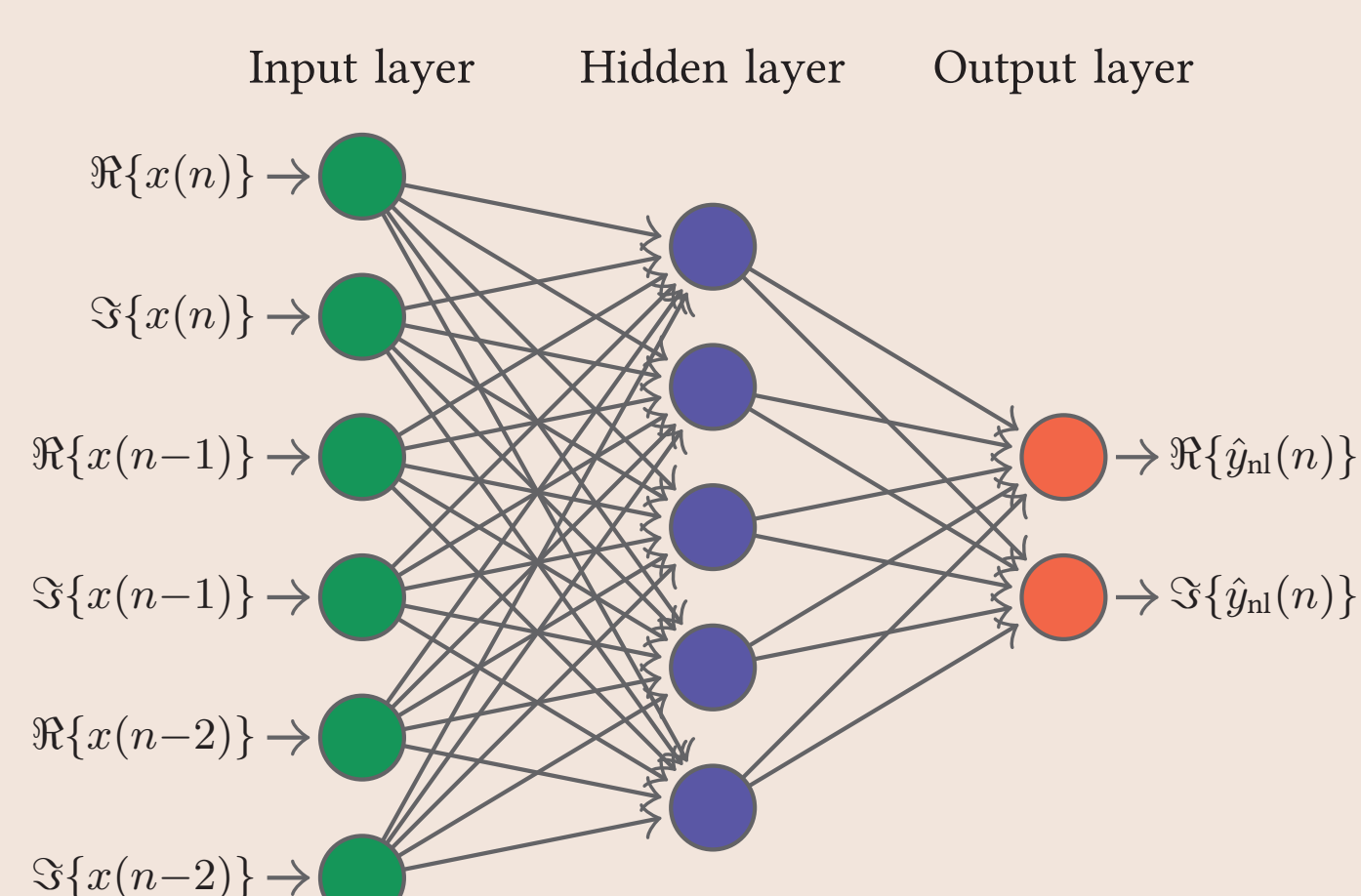
### Complexity Analysis

1. Real additions:

$$n_{\text{ADD,NN}} = (2M+2L+3)n_h$$

2. Real multiplications:

$$n_{\text{MUL,NN}} = (2M+2L+2)n_h$$

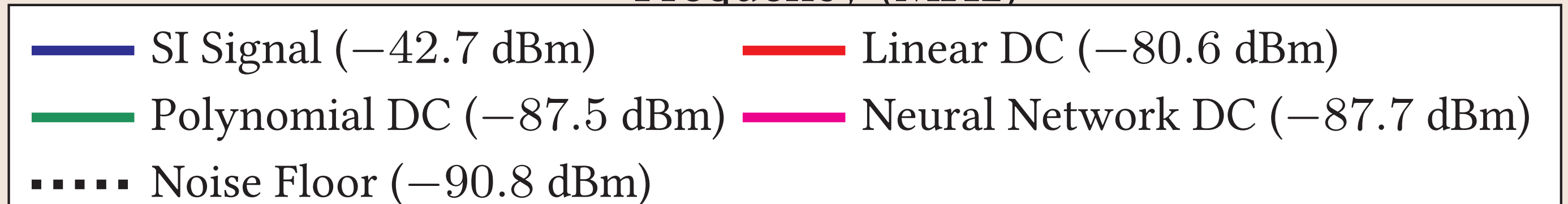
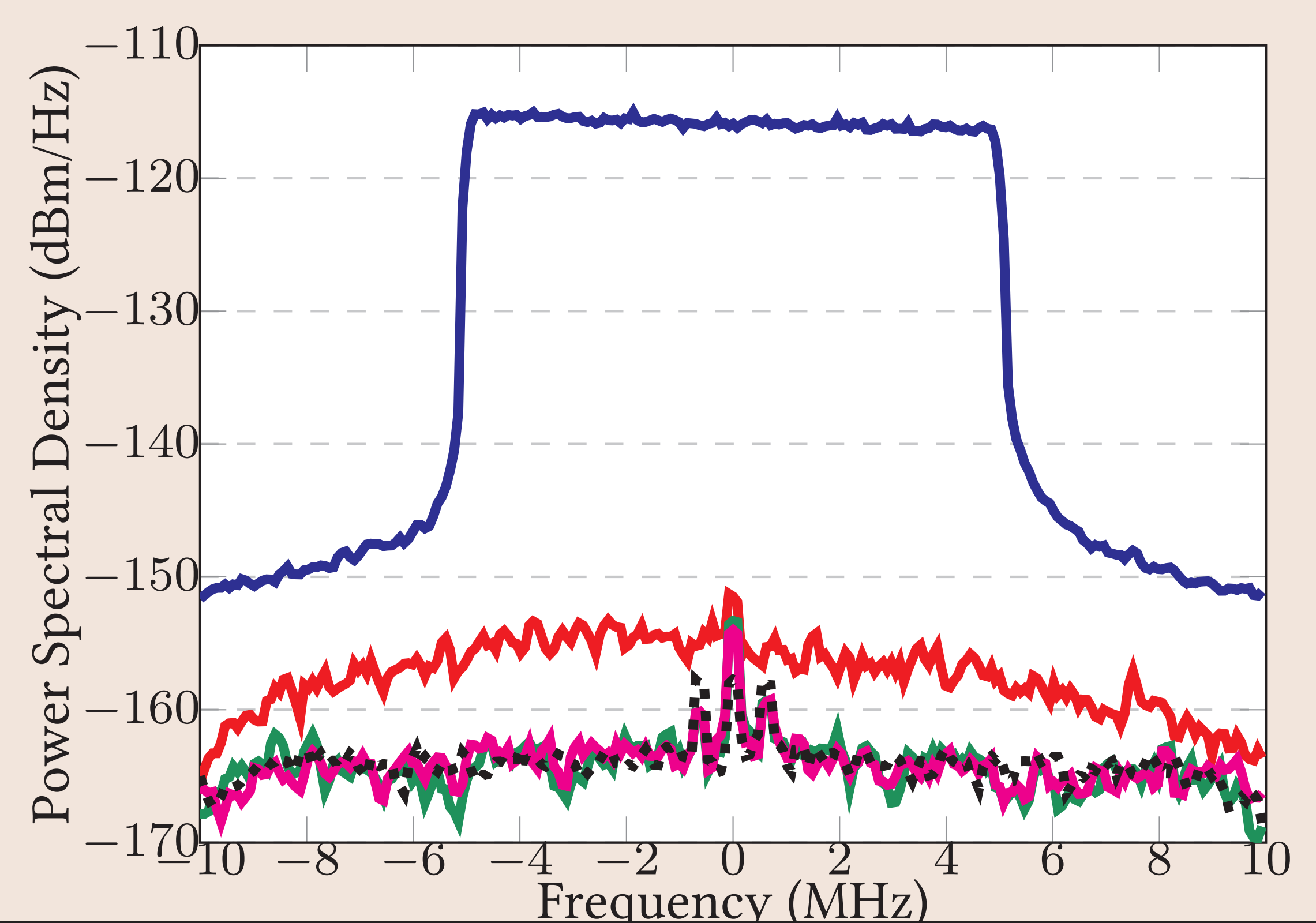


## Experimental Setup

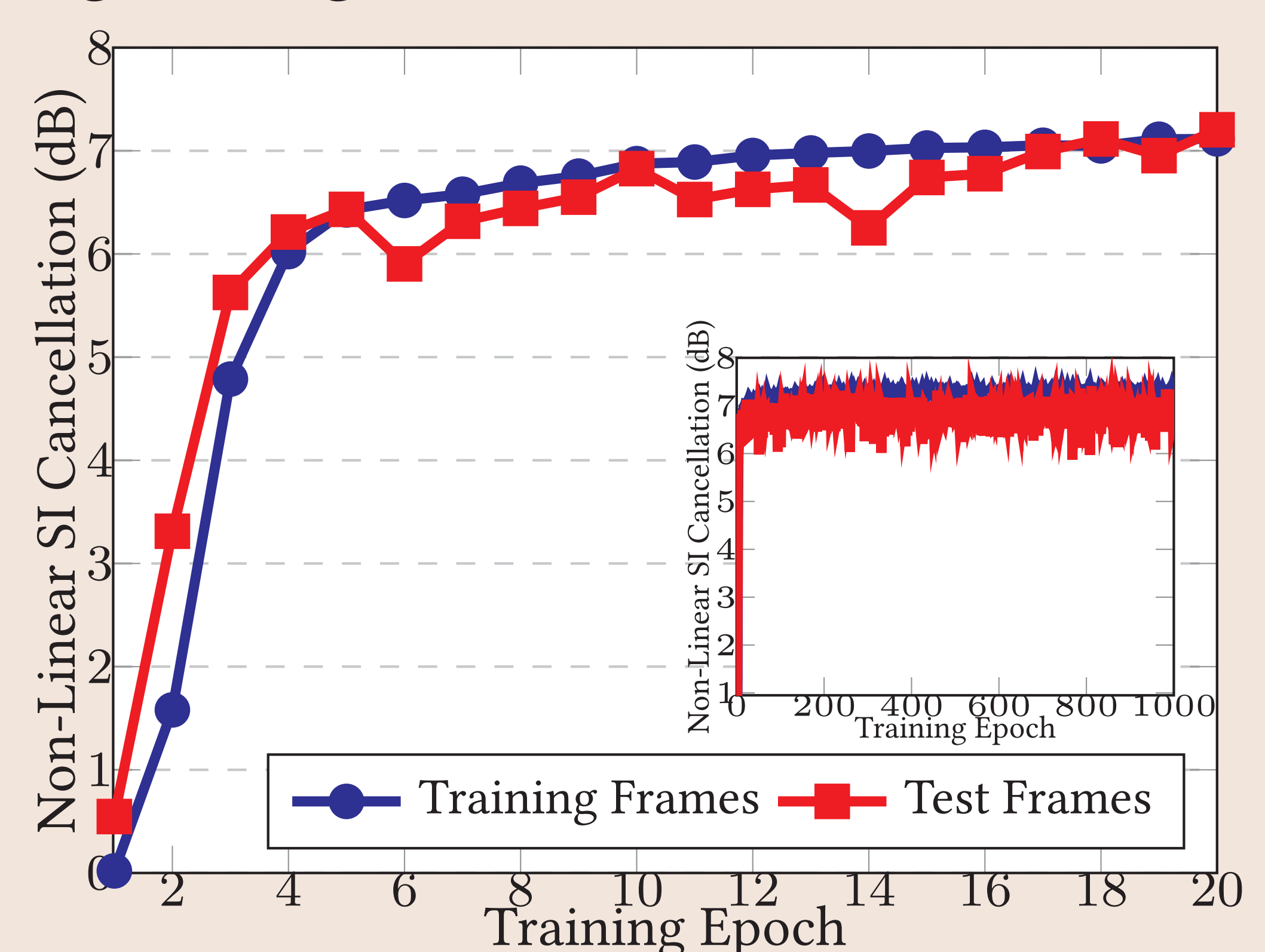
- **Dataset:** Full-duplex testbed with 53 dB analog cancellation
- **General:** 10 MHz OFDM signal, 10 dBm transmit power, 20,000 samples (90% training, 10% test),  $M+L=13$
- **Polynomial:**  $P=7$ , LS training
- **NN:**  $n_h=17$ , MSE cost, Adam optimizer ( $\lambda=0.004$ ,  $B=32$ )

## Self-Interference Cancellation Results

### Cancellation performance:



### NN training convergence:



### Complexity Comparison

	Polynomial	NN	Improvement
Additions	492	493	0%
Multiplications	741	476	<b>36%</b>

## Conclusions & Future Work

- Neural network seems **very promising**: same performance as the complex polynomial model, but with **lower complexity**
- Convergence and complexity of training need to be compared
- Scenarios with higher non-linear cancellation have to be examined
- Hardware implementations have to be compared

