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Deep Neural Network Assisted Second-Order Perturbation-Based Nonlinearity Compensation

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Abstract: We propose a fiber nonlinearity post-compensation technique using the DNN and the second-order perturbation theory. We achieve ~ 1 dB Q-factor improvement for a 32 Gbaud PDM-64-QAM at 1200 km compared to the linear dispersion compensation. © 2021 The Author(s)

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1. Introduction

Fiber nonlinearity limits the transmission performance of the high baud-rate long-haul optical communication systems [1], [2]. Recently, model-based machine learning approaches have been considered for the fiber nonlinearity compensation (NLC) [3], for example, the neural network (NN)-assisted NLC (NN-NLC) [4]. The NN-NLC technique uses the first-order (FO) perturbation triplets as the input features to the deep neural network (DNN). That helps the DNN to effectively learn the underlying physical processes that govern the nonlinear interactions in the optical fiber [4]. Nowadays, there is an increasing thrust towards 800G systems using higher-order modulation formats. It should be noted that, in such systems, the FO perturbation approximation becomes inaccurate to model the nonlinear distortion field due to the increase in the transmit launch power. In other words, the higher-order perturbation terms become significant in such high data-rate systems. In this paper, unlike the NN-NLC technique, we propose using both FO triplets and second-order (SO) quintuples as the input features to the DNN to improve the NLC performance. This technique is referred to as the DNN-assisted SO-PB-NLC (DNN-SO-PB-NLC).

2. The DNN-SO-PB-NLC Technique

The underlying premise of our proposed approach is that we add the SO perturbation term to the FO term in the series expansion-based approximation of the nonlinear distortion field to improve the approximation accuracy, thereby improving the NLC performance [5]. In conformance with theory, we propose to use an augmented DNN structure: one DNN to estimate the FO nonlinear distortion field and the other to estimate the SO distortion field. The estimated SO field is then added FO field. That improves the estimation accuracy of the overall nonlinear distortion field. to the

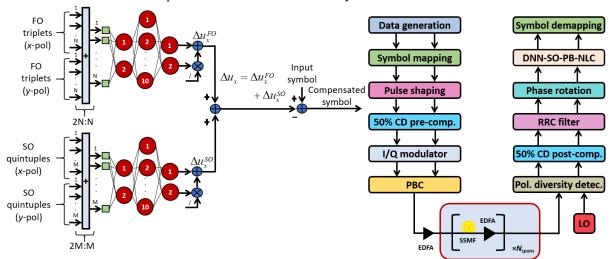


Fig. 1: The DNN-SO-PB-NLC technique. $\Delta u_x^{\rm FO}$ is the FO nonlinear Fig. 2: The simulation setup. PBC: polariza-

tion beam combiner, N_{spans} : number of spans, and LO: local oscillator.

Fig. 1 shows the network model of the DNN-SO-PB-NLC technique. The DNNs in Fig. 1 have 2 hidden layers consisting of 2 and 10 nodes, respectively. The output nodes provide the real and imaginary parts of the estimated FO/SO nonlinear distortion fields. In this technique, after generating 2N FO triplets, the intra- and interpolarization triplets with the same indices are added to generate N twin-triplets and given to the input layer of the upper DNN of Fig. 1. Similarly, M twin-quintuples are generated from 2M SO quintuples and given to the

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input layer of the lower DNN in Fig. 1. In implementing the DNN-SO-PB-NLC technique, the estimated FO/SO nonlinear distortion field is subtracted from the symbol of interest to compensate for the nonlinearity effect.

3. Numerical Simulation and the Results

Fig. 2 shows the numerical simulation setup for the DNN-SO-PB-NLC technique. The transmitter processing consists of the generation of single-channel 32 GBaud polarization-division multiplexed (PDM)-64-quadrature amplitude modulation (QAM) signal, root-raised cosine (RRC) pulse shaping filtering, 50% chromatic dispersion compensation (CDC), and the signal up-conversion using in-phase/quadrature-phase (I/Q) modulator. The upconverted signal is then transmitted through a 1200 km standard single-mode fiber (SSMF). The SSMF parameters considered are as given in [2]. An erbium-doped fiber amplifier (EDFA) with a gain of 16 dB and a noise figure of 6 dB is used for the periodic amplification. At the receiver, after coherent detection, 50% CDC, RRC filter, and phase rotation, the proposed DNN-SO-PB-NLC technique is applied as post-compensation. It is evident from Fig. 3(a) that the DNN-SO-PB-NLC technique improves the Q-factor performance by ~ 0.27 dB and ~ 1 dB when compared to the NN-NLC and the CDC techniques, respectively. It also shows a significant performance improvement for the DNN-SO-PB-NLC technique when compared to the conventional (CONV) FO-/SO-PB-NLC techniques. To reduce the computational complexity, we trimmed off all the tensor weights that are less than a specified threshold value, and the network is re-trained and estimates the nonlinear distortion field. The results in Fig. 3(b) indicate that the pruning at the weight magnitude threshold of -15 dB results in 138 FO triplets and 268 SO quintuples, and the corresponding penalty ΔQ , defined as the Q-factor difference between the case with no pruning and with pruning at the optimum launch power, is less than 0.2 dB when compared to the case without pruning. Table 1 shows the computational complexity, and the results indicate that the real-valued multiplications per symbol for the proposed DNN-SO-PB-NLC is reduced by 23.75% compared to the CONV digital back-propagation (DBP) technique.

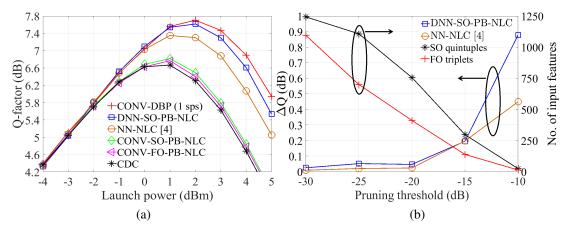


Fig. 3: Simulation results. (a) Q-factor as a function of launch power and (b) ΔQ and the no. of input features as a function of the pruning threshold. sps: step/span

Table 1: Computational	complexity in to	erms of the number of	t real-valued mu	iltiplications/symbol.
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NN-NLC [4]		DNN-SO-PB-NLC				
Compu. in the FO DNN part Compu. in the l		O+SO DNN parts				
DNN w/o prun.	DNN w/ prun.	DNN w/o prun.	DNN w/ prun.	CONV-FO-	CONV-SO-PB-	CONV-DBP
	@-15 dB		@-15 dB	PB-NLC	NLC	
5,296	592	10,592	1,704	119	1,547	2,235

4. Conclusion

We proposed a DNN-assisted SO-PB-NLC post-compensation technique to compensate for the fiber nonlinearity effect. We demonstrated that the proposed technique improves the Q-factor performance by ~ 0.27 dB and ~ 1 dB when compared to the NN-NLC and the CDC techniques, respectively. Further, we showed that the weight magnitude pruning of the DNN-SO-PB-NLC technique at a threshold of -15 dB reduces the number of real-valued multiplications per symbol by 531 when compared to the CONV-DBP technique.

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