

Nonlinear Sculpturing of Optical Pulses in Fibre Systems

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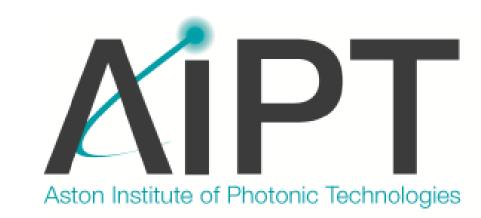
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INTRODUCTION

In recent years, there has been a growing interest from the photonics community in the generation of non-conventional optical waveforms because of their applications in all-optical signal processing and microwave signal manipulation. While sinusoidal, Gaussian and hyperbolic secant intensity profiles are now routinely produced by modulators or mode-locked lasers, other signal waveforms such as parabolic, triangular or flat-top pulse shapes remain rather hard to synthesize. The interplay among the effects of dispersion, nonlinearity and gain/loss in optical fibre systems can be efficiently used to shape the pulses and manipulate and control the light dynamics and, hence, lead to different pulse-shaping regimes. In particular, it has been demonstrated that the temporal and spectral properties of a pulse can be tailored by use of such a nonlinear shaping approach to generate ultra-short parabolic or triangular pulses [1, 2].

However, achieving a precise waveform with various prescribed characteristics, such as temporal duration and chirp properties, is a complex issue that requires careful choice of the initial pulse conditions and system parameters. The general problem of optimization towards a target operational regime in a complex multi-parameter space can be intelligently addressed by implementing machine-learning strategies. In this paper, we discuss a novel approach to the characterization and optimization of nonlinear shaping in fibre systems, in which first we perform numerical simulations of the basic propagation model to identify the relevant parameters, and then we use the machine-learning method of neural networks (NNs) to make predictions across a larger range of the data domain.

IN PASSIVE NORMALLY DISPERSIVE FIBER

NUMERICAL MODEL

Propagation in normally dispersive fiber is governed by the nonlinear Schrödinger equation (NLSE) :

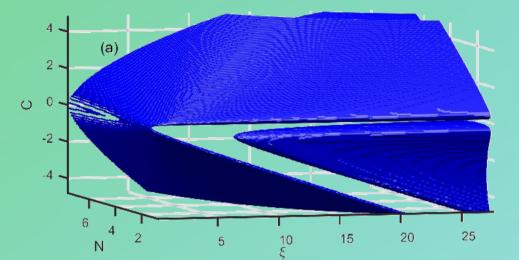
 $i\frac{\partial\psi}{\partial z} - \frac{\beta_2}{2}\frac{\partial^2\psi}{\partial t^2} + \gamma |\psi|^2 \psi = 0 \qquad \text{grown}$

group velocity dispersion Kerr nonlinearity

- $\begin{array}{c} \hline & 6 & \text{degrees of freedom}: & (C_0, T_0, P_0, \beta_2, \gamma, L) \\ & \text{input pulse} & \text{fibre} \end{array}$
- Using normalized quantities reduces the level of complexity. (C, N, ζ) : normalized input chirp, soliton number, normalized propagation length $i \frac{\partial u}{\partial \zeta} - \frac{1}{2} \frac{\partial^2 u}{\partial \tau^2} + N^2 |u|^2 u = 0$

Output pulse properties are quantified using the full-width at half maximum temporal and spectral widths, the Strelh ratio, a misfit factor or the kurtosis excess.

SEARCH FOR THE OPTIMUN WORKING PARAMETERS



The three-dimensional space of normalized quantities can be visualized using 3D plots.

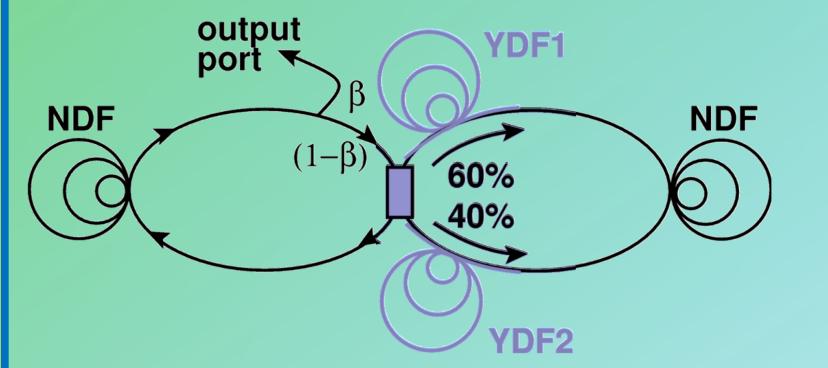
Surfaces in the three-dimensional parameter space that enable: (a) the formation of a pulse with a misfit parameter to a parabolic shape $M \le 0.045$ (b) an output temporal duration of $T_{out} = 10 T_{in}$

The intersections of different surfaces provide the means

N A DUAL-PUMP NONLINEAR AMPLIFYING LOOP MIRROR MODE-LOCKED ALL-FIBRE LASER

NUMERICAL MODEL

We numerically characterize the performance of the figure-8 laser demonstrated in [4], based on a dual-pump nonlinear amplifying loop mirror.



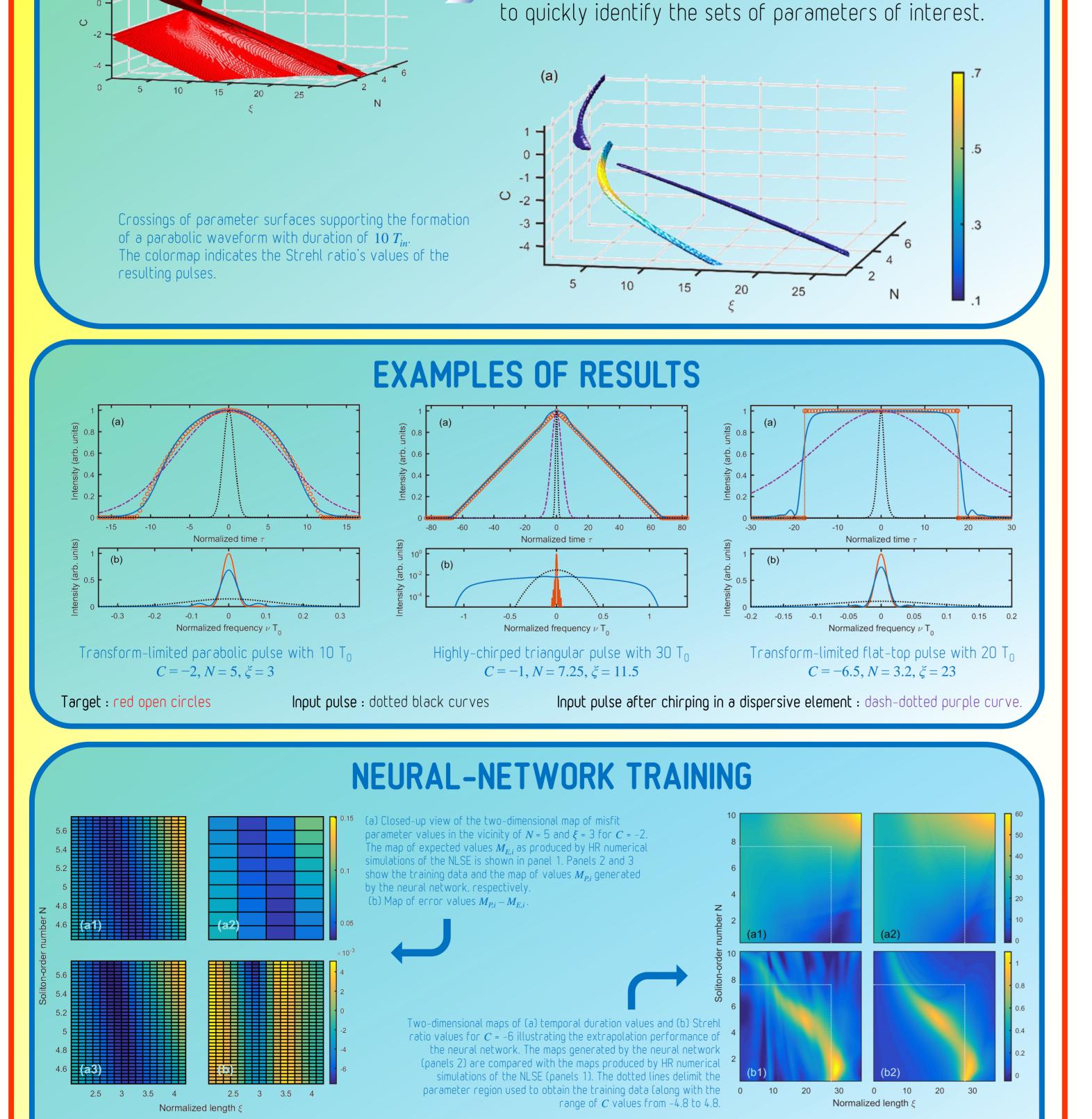
- two segments of ytterbium doped fibre are used.
- normally dispersive fibres are included.
- the output coupling ratio can be changed.
- We use a lumped model for the laser, in which propagation in the fibres follows a generalized NLSE:

 $i\psi_z - \frac{1}{2}\beta_2\psi_{tt} + \gamma|\psi|^2\psi = \frac{i}{2}g(z,\omega)\psi$

The gain g includes saturation and spectral response

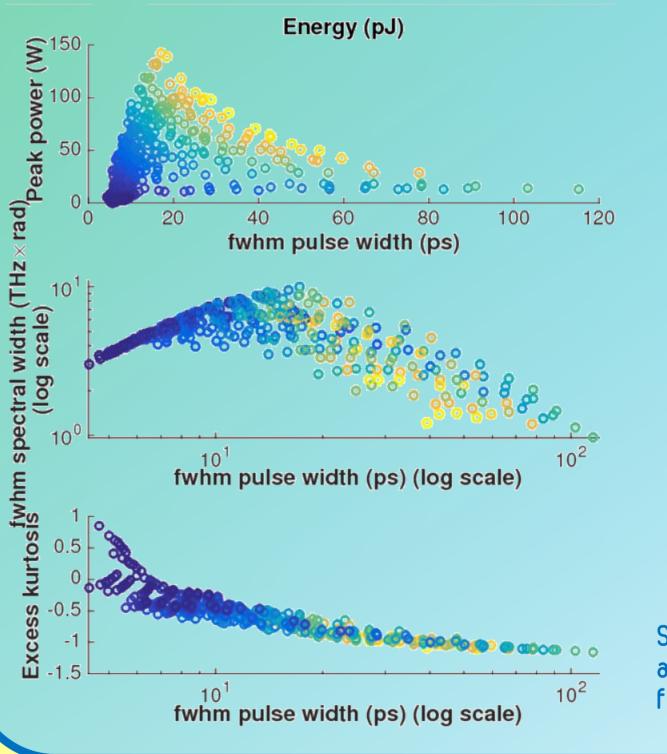
Due to the longitudinal evolution of *g* (requiring the solution of steady-state rate equations for an effective two-level system at each longitudinal step), direct simulations are time consuming.

We explore the space of operating states that can be accessed through control of the total pump power $P_T = P_1 + P_2$, the fractional pump power $R = P_1/P_T$, and the laser output coupling ratio β .

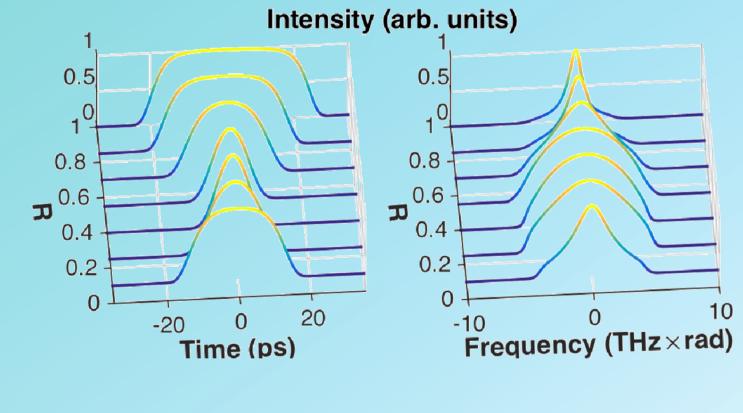


DIRECT NUMERICAL SIMULATIONS





(a1)



Evolution of the pulse temporal and spectral intensity profiles with R for $\beta = 0.75$ and $P_T = 11.0$ W.

Scatter plots of output pulse characteristics from the laser as obtained from NLSE numerical simulations for variable (β , P_T , R) parameters

NEURAL-NETWORK TRAINING

>>> A NN regression model provides a rapid and precise identification of the attainable output properties.

	NN output, E _N (pJ)	(a2) Expected values, E _E (pJ)	(b1) NN output, E _N (pJ)	(b2) Training data (pJ)
)		10 2500 (a) Predictions on a new value of the output coupling coefficient (8 - 0.7)	10	10 2500

A NN regression can reproduce the observed trends.

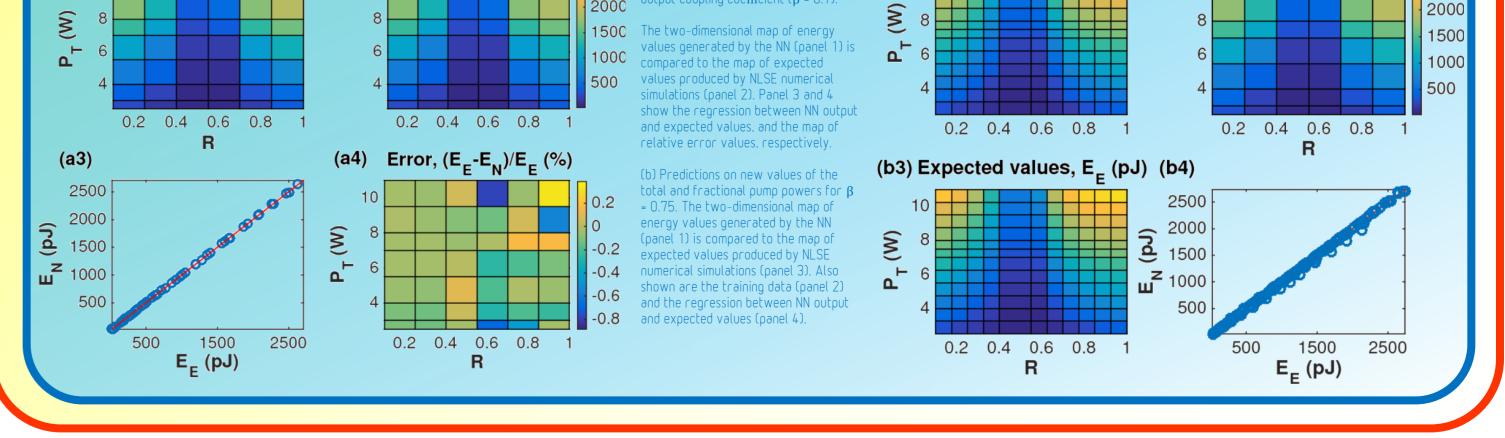
The reduced-size data set used to train the NN enables a reduction of the CPU time by a factor of 16 relative to the generation of the full data set.

> The predictions from the NN are found to be sufficiently accurate for the targeted application.











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[2] S. Boscolo, A. I. Latkin, and S. K. Turitsyn, "Passive nonlinear pulse shaping in normally dispersive fiber systems," IEEE J. Quantum Electron. 44, 1196–1203 (2008).

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[4] S. Kobtsev, A. Ivanenko, A. Kokhanovskiy, and S. Smirnov, "Electronic control of different generation regimes in mode-locked all-fibre F8 laser," Laser Phys. Lett. 15, 045102 (2018).

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