

Mathematics of the Optical Fiber Communication:

Nonlinear Fourier Transforms and Information Theory

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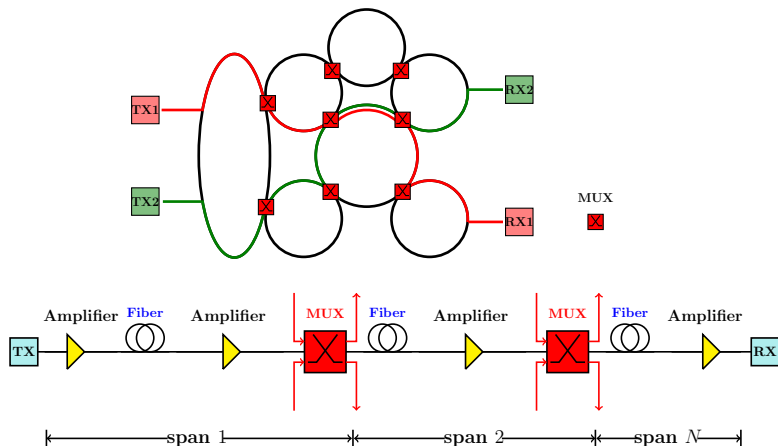
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Wired Communication Networks



(1) Multiple users with add-drop multiplexers (**ADMs**); (2) **interference unknown** to the user-of-interest (UOI); (3) network topology unknown

Optical Fiber

- **Advantages:** Low-loss (~ 0.2 dB/km), huge bandwidth (~ 5 - 10 THz bandwidth), all-optical processing (laser sources, amplifiers, detectors)
- **Challenges:** Refractive index is a function of **frequency** and **intensity**

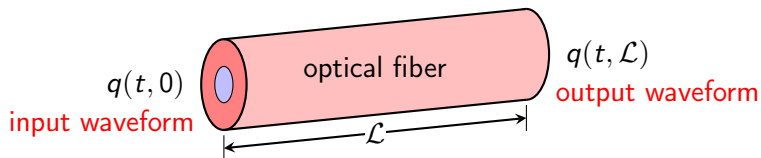
$$n(\omega, |q|^2) = \underbrace{n_0 + n_1(\omega - \omega_0) + n_2(\omega - \omega_0)^2 + \dots}_{\text{dispersion}} + \underbrace{\gamma_0 |q|^2}_{\text{Kerr nonlinearity}} + \dots$$



intuition

- 1 **Dispersion:** n depends on frequency
- 2 **Kerr nonlinearity:** *the intensity of the signal modifies the refractive index!*
- 3 High reliability: $P_e = 10^{-15}$
- 4 High speed: 400 Gb/s

Stochastic Nonlinear Schrödinger Equation



Pulse propagation in optical fibers can be modeled by the *stochastic nonlinear Schrödinger (NLS) equation*:

$$\frac{\partial q(t, z)}{\partial z} = \underbrace{\frac{\partial^2 q(t, z)}{\partial t^2}}_{\text{dispersion}} \pm \underbrace{2j|q(t, z)|^2 q(t, z)}_{\text{nonlinearity}} + \underbrace{n(t, z)}_{\text{noise}}$$

- $q(t, z)$ is the signal, t is time, z is distance
- Distributed white Gaussian noise
- + focusing regime, – defocusing regime
- Vectorial generalizations exist

Fourier Analysis of the NLS Equation

Assume a Fourier series with variable coefficients for $q(t, z)$ at $z > 0$

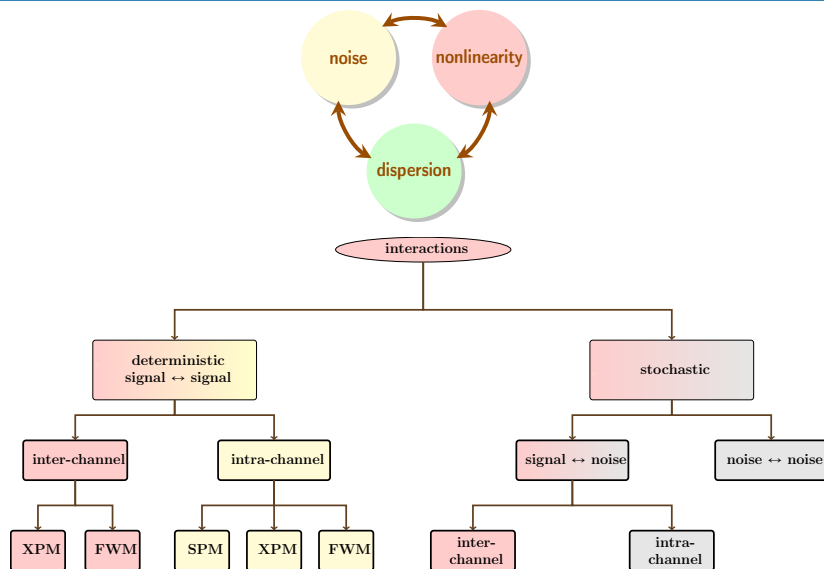
$$q(t, z) = \sum_{k=0}^{N-1} q_k(z) e^{j2\pi k W t}.$$

Substituting (1) into the NLS equation

$$\begin{aligned} j \frac{\partial q_k(z)}{\partial z} = & \underbrace{-4\pi^2 W^2 k^2 q_k(z)}_{\text{dispersion}} + \underbrace{2|q_k(z)|^2 q_k(z)}_{\text{SPM}} \\ & + \underbrace{4q_k(z) \sum_{\ell \neq k} |q_\ell(z)|^2}_{\text{XPM}} \\ & + \underbrace{2 \sum_{\substack{\ell \neq m \\ \ell \neq k}} q_\ell(z) q_m^*(z) q_{k+m-\ell}(z)}_{\text{FWM}} + n_k(z), \end{aligned}$$

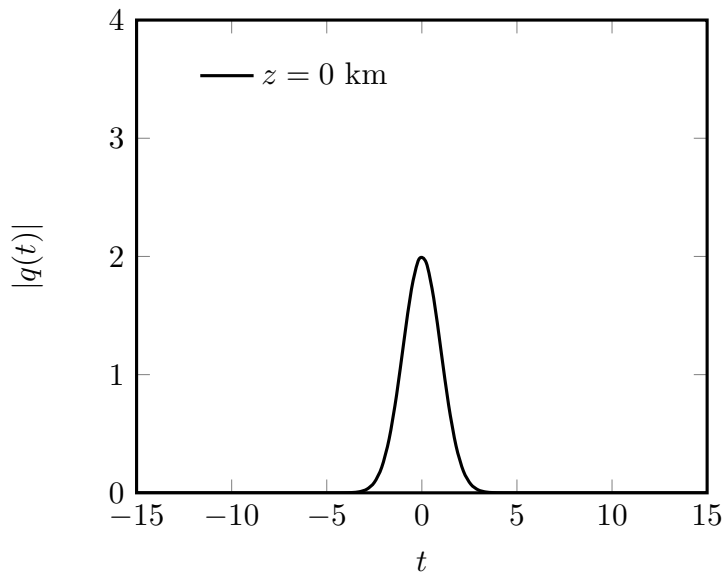
in which n_k are the noise coordinates and where we have identified the dispersion, self-phase modulation (SPM), cross-phase modulation (XPM) and four-wave mixing (FWM) terms.

Nonlinear Effects in Fibers

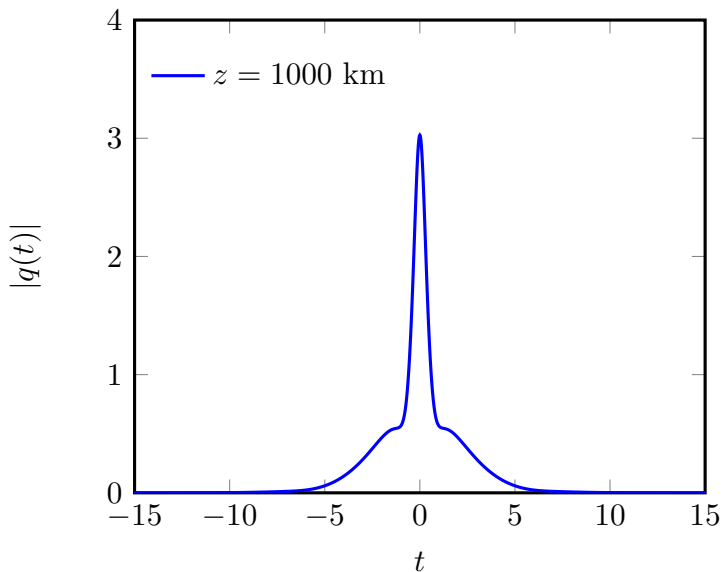


SPM & XPM = self- & cross- phase modulation; FWM = four-wave mixing.

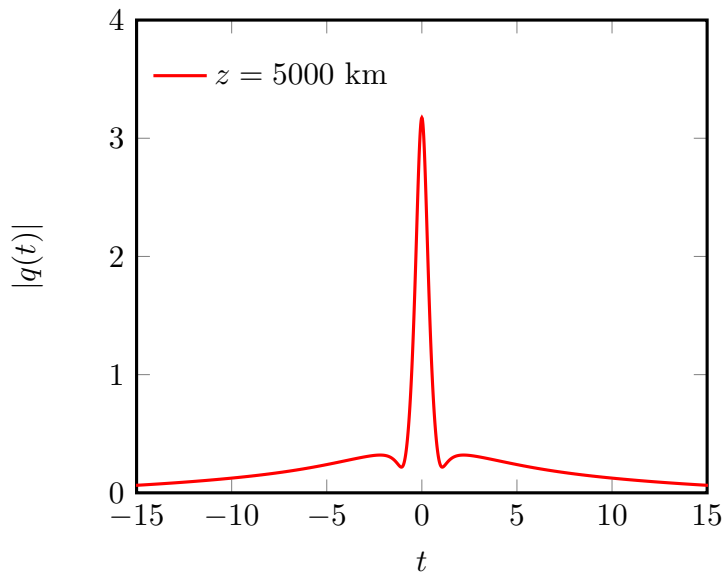
Example of Signal Propagation



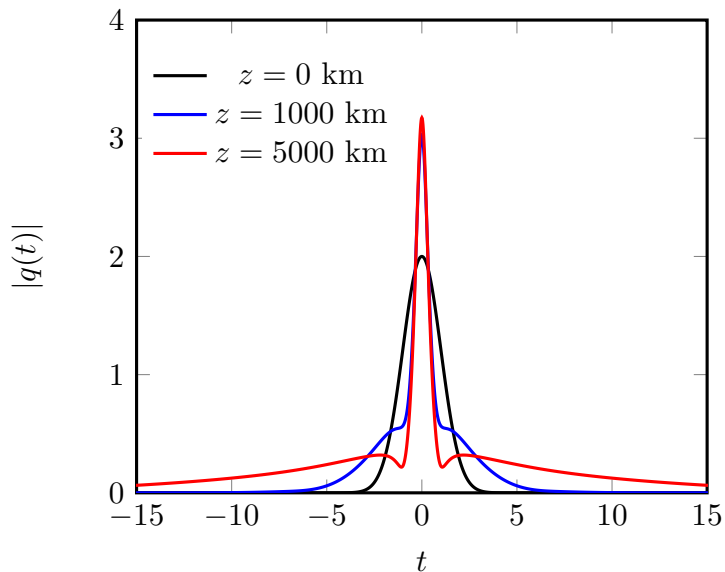
Example of Signal Propagation



Example of Signal Propagation

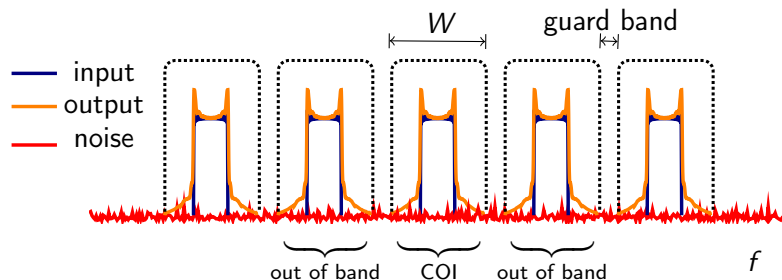


Example of Signal Propagation

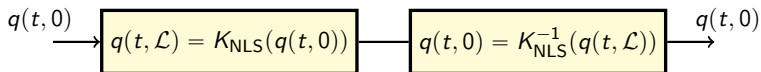


The State-of-the-Art Approach

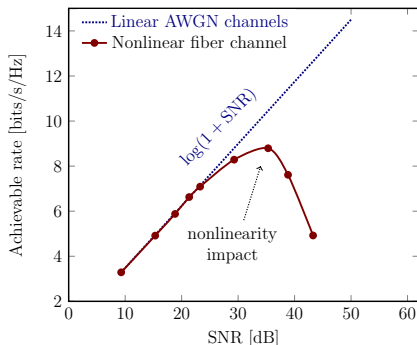
- ① **TX:** Wavelength-division multiplexing (WDM)



- ② **RX:** Digital back-propagation (BP)



Current Achievable Rates



Fiber nonlinearity places an upper limit on capacity

Nonlinear Shannon limit in fiber

Capacity crunch in fiber!

Central Question:

Does fiber nonlinearity really place an upper limit on achievable spectral efficiency?

Origin of the Capacity Limitation – 1

Let $T : \mathcal{H} \mapsto \mathcal{H}$ be a *linear map*:

$$y = T(x) + n,$$

where x and y are input and output signals and n is noise.

Projecting signals onto an orthonormal basis $\{\phi_k\}_{k \in \mathbb{N}}$:

$$\{x, y, n\} = \sum_{k=1}^{\infty} \{x_k, y_k, n_k\} \phi_k, \quad \text{Thus:}$$

$$y_k = x_k \langle T \phi_k, \phi_k \rangle + \underbrace{\sum_{i \neq k} x_i \langle T \phi_i, \phi_k \rangle}_{\text{linear interactions}} + n_k$$

However, if $\{\phi_k(t)\}_k$ is the set of eigenvectors of T , then

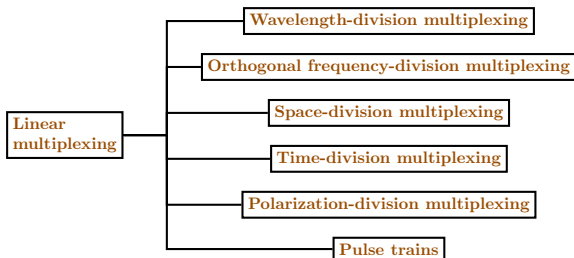
$$y_k = \lambda_k x_k + n_k$$

where λ_k is eigenvalue.

Origin of the Capacity Limitations – 2

Capacity crunch occurs if the **basis** used for communication is **not compatible** with the channel.

- Deterministic nonlinear effects are not a fundamental limitation. It is the **method of the communication** causing the problem
- After abstracting away non-essential aspects, current methods, in essence, modulate **linear-algebraic modes**
- In nonlinear channels, this introduces **interference** and **ISI**
- *BP cannot remove the interference in a network scenario*



Nonlinear frequency-division multiplexing (NFDM)

- It was realized that the NLS equation supports **nonlinear eigenfunctions** which have a crucial **independence property**, the key to build a multiuser system
- The tool necessary to reveal signal degrees of freedom is

Nonlinear Fourier Transform

- Based on NFT, we constructed an *NFDM*, which can be viewed as a generalization of OFDM to optical fiber
- Exploiting the integrability, NFDM modulates **non-interacting degrees-of-freedom**
- Capacity of the NFDM in the deterministic model is **infinite**

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Linear Convolutional Channel

Let $T_h(x) = h * x$

$$y(t) = h(t) * x(t) + n(t), \quad 0 \leq t < T.$$

The eigenvectors and eigenvalues of $T_h(x) = h * x$ are

$$\phi_k(t) = \frac{1}{\sqrt{T}} \exp(-jk\omega_0 t), \quad \omega_0 = \frac{2\pi}{T},$$

and $\lambda_k = \mathcal{F}(h(t))(k\omega_0)$.

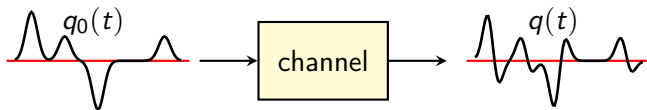
Fourier transform maps convolution into a multiplication operator

$$Y(\omega) = H(\omega) \cdot X(\omega) + N(\omega), \quad \omega = k\omega_0.$$

- 1 Frequency ω is **conserved** in the channel
- 2 Channel is decomposed into **parallel independent** channels
- 3 **OFDM**: information is encoded in spectral amplitudes $X(\omega)$

General Waveform Channels

- Instantaneous waveform channel



$$q(t) = K(q_0(t)) + n(t)$$

- Evolutionary channel.* Here the signal evolves according to an *evolution equation* in **1+1** dimensions (**time** t , **distance** z)

$$\frac{\partial q}{\partial z} = K(q(t, z)) + n(t)$$

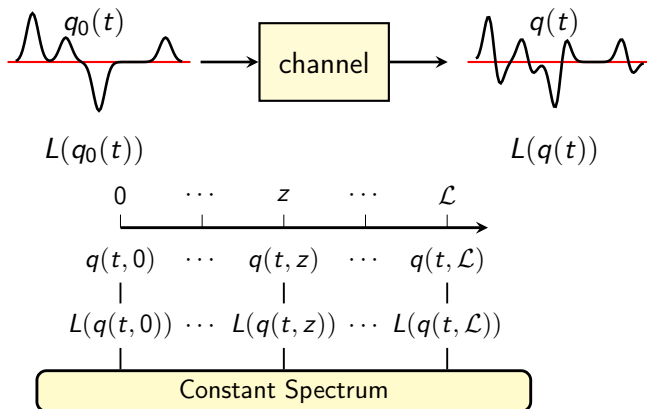
Examples: ($q_t := \partial_t q$)

- $K(q) = j|q|^2$ (memoryless)
- $K(q) = -j(q_{tt} + 2|q|^2 q)$ (NLS)
- $K(q) = q_{tt}$ (heat eq.)
- $K(q) = q_{ttt} + 6qq_t$ (KdV)

Isospectral Flow

A Key Idea

We seek an **invariant** under evolution (in the absence of noise). Let L be a linear differential operator (depending on $q(t, z)$). It **may** be possible to find an L whose (eigenvalue) **spectrum** remains constant, even as q evolves (in z).



Isospectral Families of Operators

If the eigenvalues of $L(z)$ do not depend on z , then we refer to $L(z)$ as an **isospectral family** of operators.

Example: The operator L can be a matrix

$$L(z) = \begin{pmatrix} \cos(z) & \sin(z) \\ \sin(z) & -\cos(z) \end{pmatrix}, \quad \lambda = \pm 1, \quad L(z) = G(z)\Lambda G^{-1}(z),$$

where $\Lambda = \text{diag}(1, -1)$.



Compact self-adjoint operators can be diagonalized similarly, via *Hilbert-Schmidt Spectral Theorem*. Here, Λ is a *multiplication operator*.

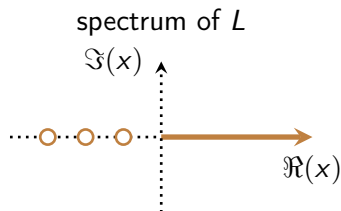
Spectrum of Bounded Linear Operators

- **Spectrum** of an operator is defined as

$$\sigma(L) = \{\lambda \mid L - \lambda I \text{ is not invertible}\}$$

Example: Linear Schrödinger operator

$$L(q(t, z)) = -\frac{\partial^2}{\partial t^2} q(t, z).$$



- **Classification:** Spectrum can be *discrete* (like matrices), *continuous*, residual, essential, etc.

The Lax Equation

We have $L(z) = G(z)\Lambda G^{-1}(z)$, where Λ does not depend on z . Assuming that $L(z)$ varies smoothly with z , we can form

$$\begin{aligned}\frac{dL(z)}{dz} &= G' \Lambda G^{-1} + G \Lambda (-G^{-1} G' G^{-1}) \\ &= \underbrace{G' G^{-1}}_{M(z)} \underbrace{(G \Lambda G^{-1})}_{L(z)} - \underbrace{(G \Lambda G^{-1})}_{L(z)} \underbrace{G' G^{-1}}_{M(z)} \\ &= M(z)L(z) - L(z)M(z) = [M, L],\end{aligned}\tag{1}$$

where $[M, L] \triangleq ML - LM$ is the *commutator bracket*. In other words, every diagonalizable isospectral operator $L(z)$ satisfies the differential equation (1).

The converse is also true.

Lemma

Let $L(z)$ be a diagonalizable family of operators. Then $L(z)$ is an isospectral family if and only if it satisfies

$$\frac{dL}{dz} = [M, L], \quad (2)$$

for some operator M , where $[M, L] = ML - LM$.

Definition

The operators L and M satisfying (2) are called a *Lax Pair* (after Peter D. Lax, who introduced the concept [1968]).



Integrable System

Let L and M be operators (depending on $q(t, z)$).

$$\begin{array}{ccc} \frac{\partial L}{\partial z} = [M, L] & \iff & \frac{\partial q}{\partial z} = K(q) \\ \text{operator form} & & \text{signal form} \end{array}$$

Example [KdV]: Let $q(t, z)$ be a real-valued function and choose

$$L = \partial_t^2 + q/3, \quad M = 4\partial_t^3 + q_t + q\partial_t.$$

Then:
$$\frac{\partial L}{\partial z} = [M, L] \iff \boxed{q_z = q_{ttt} + qq_t}.$$

NLS Equation

For the normalized nonlinear Schrödinger equation

$$jq_z = q_{tt} + 2|q|^2q,$$

Zakharov and Shabat (1972) found a Lax pair:

$$L = j \begin{pmatrix} \frac{\partial}{\partial t} & -q(t, z) \\ -q^*(t, z) & -\frac{\partial}{\partial t} \end{pmatrix},$$

$$M = \begin{pmatrix} 2j\lambda^2 - j|q(t, z)|^2 & -2\lambda q(t, z) - jq_t(t, z) \\ 2\lambda q^*(t, z) - jq_t^*(t, z) & -2j\lambda^2 + j|q(t, z)|^2 \end{pmatrix}.$$

As $q(t, z)$ evolves according to the NLS equation, the spectrum of L is preserved.

Thus the NLS equation is indeed generated by a Lax pair!

Nonlinear Fourier Transform. Summary-1

$$L = j \begin{pmatrix} \frac{\partial}{\partial t} & -\mathbf{q}(\mathbf{t}) \\ -\mathbf{q}^*(\mathbf{t}) & -\frac{\partial}{\partial t} \end{pmatrix}$$

- Generalized frequencies: **eigenvalues λ of L**
- Nonlinear Fourier coefficients: **\mathbf{a} , \mathbf{b}** where

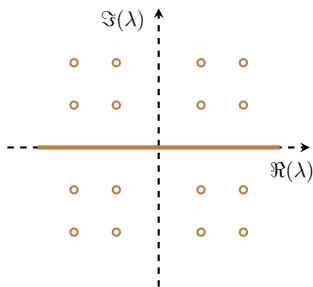
$$V(\lambda) = \begin{pmatrix} a \\ b \end{pmatrix}$$

is a normalized eigenvector of L

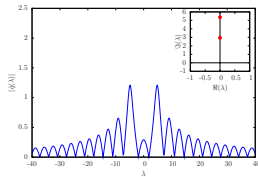
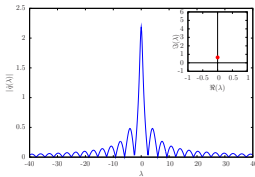
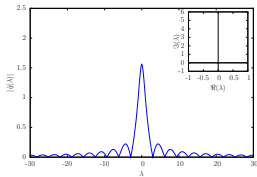
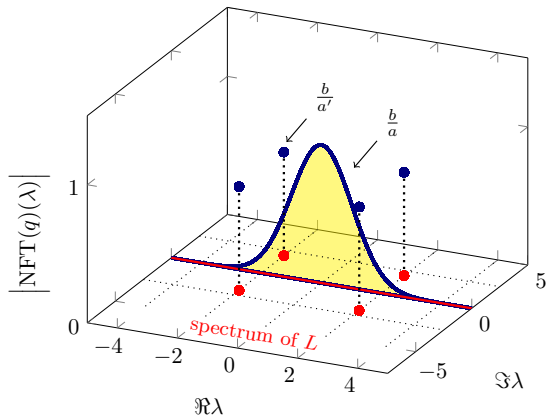
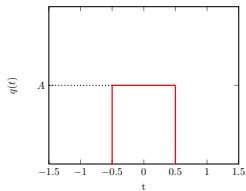
Nonlinear Fourier Transform. Summary-2

The Zakharov-Shabat operator has two types of spectra:

- A *discrete (or point) spectrum* which occurs in \mathbb{C}^+ and corresponds to *solitons*
- A *continuous spectrum*, which in general includes the whole real line \mathbb{R}



Nonlinear Fourier Transform. Summary-3



Properties of the NFT

- 1 The NFT shares some of properties of the Fourier transform (FT)
- 2 FT is a special case of the NFT if $\|q\|_{L_1} \ll 1$

3

$$\text{Linear: } y(t) = h(t) * x(t) \longleftrightarrow Y(\omega) = H(\omega)X(\omega)$$

$$\text{Integrable: } y(t) = x(t) * (L, M; \mathcal{L}) \longleftrightarrow \text{NFT}(y)(\lambda) = H(\lambda, \mathcal{L}) \text{NFT}(x)(\lambda)$$

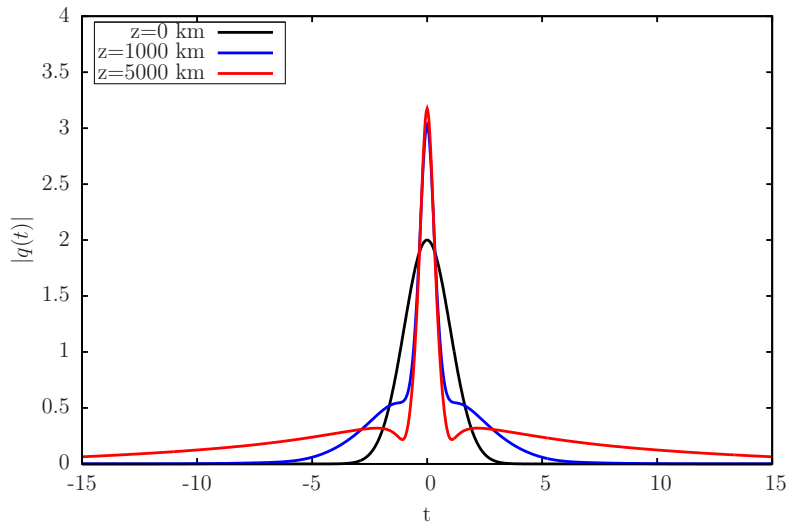
where $H(\lambda, \mathcal{L}) = e^{-4j\lambda^2 \mathcal{L}}$ is the channel filter. *The generalized frequencies are invariant in the channel.*

- 4 When there are a finite number of parameters, the solutions can be expressed via theta functions.

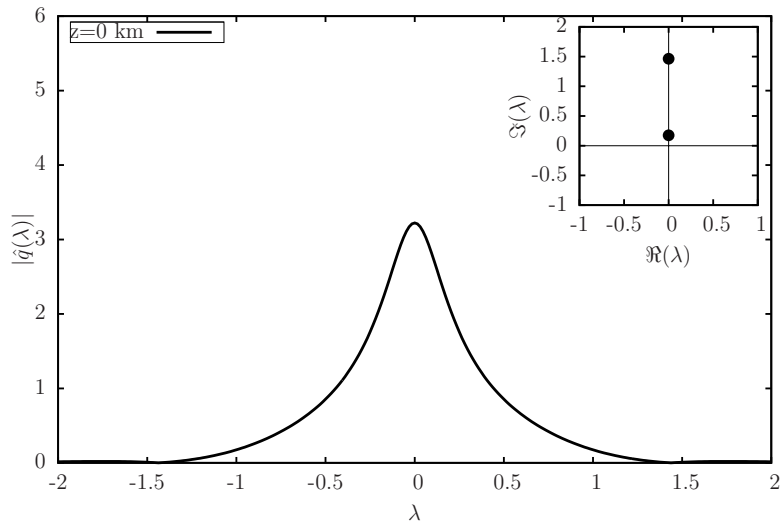
Let \mathbb{K} be an $N \times N$ complex matrix with $\Im(\mathbb{K}) > 0$. The Riemann theta function is defined by

$$\theta(\mathbf{t}|\mathbb{K}) = \sum_{\mathbf{m} \in \mathbb{Z}^N} \exp\left(2\pi j(\mathbf{m}^T \mathbf{t} + \frac{1}{2} \mathbf{m}^T \mathbb{K} \mathbf{m})\right), \quad \mathbf{t} \in \mathbb{C}^N.$$

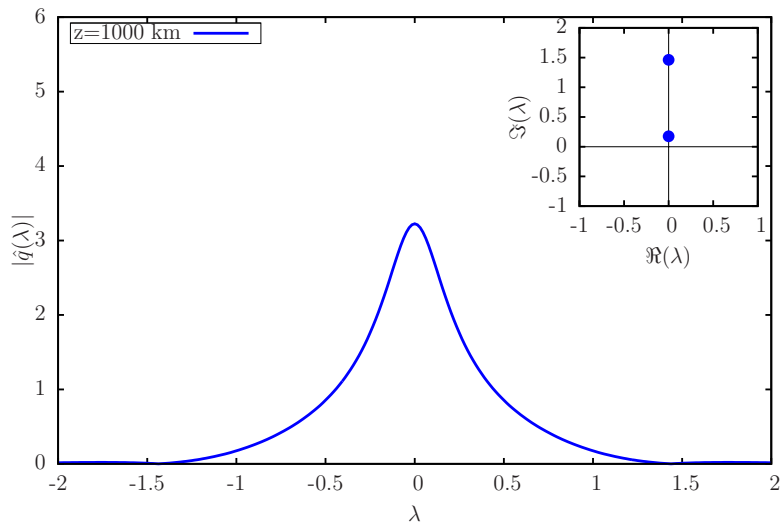
Launching a Pulse (revisited)



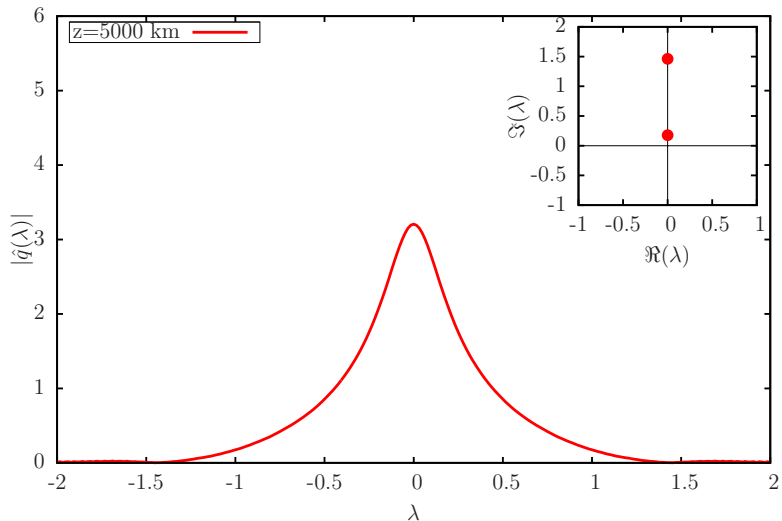
Launching a Pulse (revisited)



Launching a Pulse (revisited)

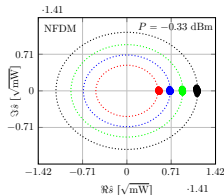
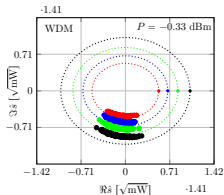
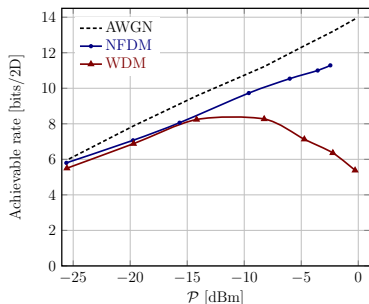
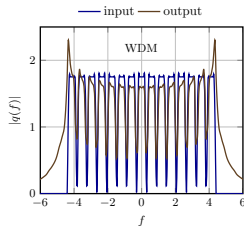
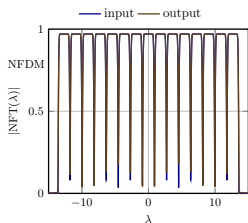


Launching a Pulse (revisited)



NFDM vs WDM

$W = 60$ GHz, $z = 2000$ km, 15 users, one symbol per user, defocusing.



Focusing regime, vectorial models, experiments, robustness to perturbations, ...

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Shannon's Formula for Linear Channels

The capacity of a channel $p_{Y|X}(y|x)$

$$C = \sup_{p_X(x)} I(X; Y)$$

The mutual information is defined as

$$I(X; Y) = h(Y) - h(Y|X)$$

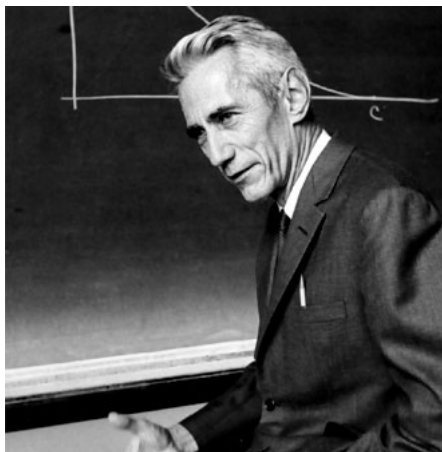
where

$$h(X) = - \int p_X(x) \log(p_X(x)) dx.$$

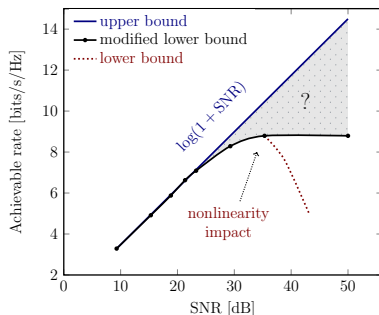
For a linear channels

$$C = \log(1 + \text{SNR}), \quad \text{bits/s/Hz}$$

The capacity of optical fiber is unknown, for about 50 years. Even $p(y|x)$ is unknown!



Upper Bound



Theorem

Consider the discrete-time periodic model $\mathbb{C}^n \mapsto \mathbb{C}^n$. We have

$$\mathcal{C}(\mathcal{P}) \leq \log(1 + \text{SNR}).$$

The proof combines:

- Energy and entropy conservation
- Shannon's entropy power inequality

Proof: Invariant Measures for PDEs

Lemma (Volume Preservation in NLS)

Let $\Omega = (\ell^2, \mathcal{E}, \mu)$ be a measure space, where $\ell^2 \triangleq \{\mathbf{q}^n \mid \sum |q_k|^2 < \infty\}$ and

$$\mu(A) = \text{vol}(A) = \int_A \left(\prod_{k=1}^n dq_k dq_k^* \right), \quad \forall A \in \mathcal{E},$$

is the Lebesgue measure. Transformation T_z underlying the NLS equation, as a dynamical system on Ω , is **measure-preserving**. That is to say

$$\mu(T_z(A)) = \mu(A), \quad \forall A \in \mathcal{E}.$$

Application 1: Theorem. The flow of T_z is entropy preserving!

Application 2: There are invariant measures. Gibbs measure:

$$d\mu_x = \frac{1}{Z} \exp \left\{ -\alpha \left(\sum_{i=1}^m |q_i|^4 - |q_i - q_{i-1}|^2 \right) \right\} \prod_{i=1}^m dq_i \chi_{\|q\| \leq 1}$$

where $\alpha > 0$, Z is the partition function, and χ_S is the indicator function.

Asymptotic Capacity

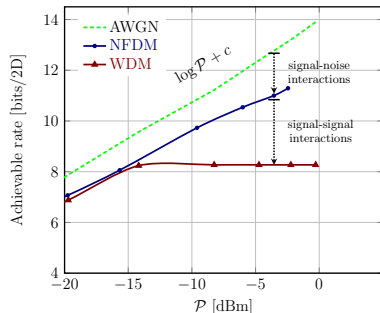
Theorem

Discrete-time periodic model $\mathbb{C}^n \mapsto \mathbb{C}^n$:

$$\mathcal{C}(\mathcal{P}) = \frac{1}{n} \log(\log \mathcal{P}) + c,$$

where $c \triangleq c(n, \mathcal{P}) < \infty$.

With n signal DOFs, $n - 1$ DOFs are asymptotically lost to **signal-noise interactions**.



Conclusions

We showed examples where advanced mathematics help make progress in long-standing engineering problems.

- Nonlinear Fourier transforms could be used for data transmission
- The **growth** of the capacity is **too small** compared to the linear channel

$$C = \frac{1}{n} \log(\log \mathcal{P}) + c$$