
Applications of Random Matrix Theory on Fiber Optical Communications

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Introduction

- Need for high speed infrastructure

Bandwidth (BW) hungry technologies emerging (~2 dB increase per year 1

- What is Optical MIMO?

Just like in wireless domain, in optical we can use parallel transmission paths to greatly enhance the capacity of the system.

- First paper on Optical MIMO: H.R. Stuart, “Dispersive multiplexing in multimode optical fiber,” Science 289(5477), 281–283 (2000)
- Multi-Mode Fibers (MMF): Use multiple modes to carry information
- Multi-Core Fibers (MCF): Utilize different optical paths of different cores in the same fiber (within the same cladding)

Introduction

Problem: Crosstalk phenomenon arises

Crosstalk between adjacent cores in MCF

Light beam scattering resulting in crosstalk in MMF

Due to :

- Extensive fiber length
- Bending of fiber
- Limited area with multiple power distributions
- Light beam scattering
- Non-linearities

Introduction

System Model

Consider a single segment N -F K D Q Q H O O R V V O H V W_t R S W L F D O \hat{c} transmitting channels excited, " 1 receiving channels coherently. $2N \times 2N$ scattering matrix is

$$S = \begin{pmatrix} t & 0 \\ r_r & r_t \end{pmatrix} \quad (S=S^T)$$

Only t (Haar-distributed $t^T t = t t^T = I_N$) sub-matrix is of interest ($r \sim 0$).

Generally $N_r, N_t < N$:

- Other channels may be used from different, parallel transceivers
- Modelling of loss: additional energy lost during propagation

System Model

Only t (Haar-distributed $t^T t = I_N$) sub-matrix is of interest (is ~ 0).

$$t = \begin{matrix} & \overbrace{\hspace{2cm}}^{N_t} & & & \\ & \left[\begin{array}{cc} \mathbf{a}_{11} & t_{1N} \\ \ll & \gg \\ \ll & \gg \\ t_{N1} & t_{NN} \end{array} \right] & & \begin{matrix} 0 \\ \gg \\ \gg \\ \gg \end{matrix} \end{matrix}$$

Define $N_t \times N_r$ matrix U as

$$U = P_{N_t}^T S P_{N_r}$$

- where P projection operator:

$$P_{N_t} = \begin{matrix} \left[\begin{array}{cc} \mathbf{a}_{N_t} & 0 \\ \ll & \gg \\ \mathbf{0} & \mathbf{1} \end{array} \right] & \begin{matrix} N_t \\ N - N_t \end{matrix} \end{matrix}$$

Information Metrics

- Outage Probability:

$$P_{\text{out}}(r) = \text{Prob} \{ I_N \leq N_t r \} = E_U \{ \mathbb{1}_{\{I_N \leq N_t r\}} \}$$

- Optimal: Assumes infinite codewords

What is the price of finite codelengths

- Gallager error bound for M-length code:

$$P_{\text{err}}(r) \leq E_U \left[\exp \left(-M \max_{\mathbf{u}} \left[N_t k r - k \log \det \left(\frac{U U^\dagger}{1 - k} \right) \right] \right) \right]$$

Coulomb– Gas Analogy

- Joint probability distribution of eigenvalues of \mathbf{U}

–

Generalized MP equation

- Use Tricomittheorem to calculate $p(x)$
- Obtain closed form expr. for energy $S[p]$
- E.g. $!; n > 0$

$$p(x) = \frac{\sqrt{(b-x)(x-a)} \int_a^b \frac{n(1-y)}{(1-x)\sqrt{(1-a)(1-b)}} \frac{E-1}{x\sqrt{ab}} dy}{2 \int_a^b \sqrt{(b-x)(x-a)} dx}$$

– a, b, k calculated from

$$\begin{aligned} p(b) - p(a) &= 0 \\ r &= \int_a^b 3dx p(x) \log 1 \\ 1 &= \int_a^b 3dx p(x) \end{aligned}$$

Generalized MP equation

- In general

	S_{0b}	S_{ab}	S_{01}	S_{a1}
	$a=0b<1$	$a>0b<1$	$a=0b=1$	$a>0b=1$
$n=0;$	$r < r_{c1}$	--	$r_{c1} < r < r_{c2}$	$r > r_{c2}$
$n>0;$	$r < r_{c3}$	$r > r_{c3}$	--	--
$n=0; >1$	--	$r < r_{c4}$	--	$r > r_{c4}$
$n>0; >1$	--	all r	--	--

- Phase transitions ($a=0$ to $a>0$) are third order
 - discontinuous $S'''(r_c)$
 - Relation to Tracy-Widom(?)

Distribution Density of r

Finally

$$f(r) \mid N_t \frac{e^{-N_t^2 S(r) - S(r_{\text{erg}})}}{\sqrt{2 S_{\text{erg}}}}$$

– where

$$v_{\text{erg}} = \frac{1}{S''(r_{\text{erg}})} \log \left\{ \frac{\sqrt{1 + b_0} \sqrt{1 + b_0}^2}{4 \sqrt{1 + b_0} \sqrt{1 + b_0}} \right\}$$

is the variance at the peak of the distribution.

Numerical Simulations (β and $n_0 > 0$)

The LD approach demonstrates better behavior, following Monte Carlo.

Numerical Simulations (β and $n_0 > 0$)

The LD approach demonstrates better behavior, following Monte Carlo.

For small values of N , N_t and N_0 , the discrepancy is minimal

- Here we need

- where

-

Numerical Simulations (! and $n_0 = 0$)

More Realistic Channel

- Chaotic cavitypicture:



More Realistic Channel

- Mutual Information metric:

$$I(G) = \log \det \left(\mathbf{P}_r \mathbf{H}_0 \mathbf{G} \mathbf{I} + \mathbf{P}_t \mathbf{H}_0 \mathbf{G} \mathbf{I} + \mathbf{P}_r \right)$$

- I – Distribution (mean/variance) can be obtained using replica theory

$$E \{ I(G) \} \approx I_1 - I_2$$

$$I_1 = \log \det \left(\mathbf{U} \mathbf{G} \mathbf{I} (1 - \beta) \mathbf{H}_0 \mathbf{I} + \mathbf{I} \right) \quad \text{at } N \rightarrow \infty$$

$$r = \frac{1}{N} \text{Tr} \left[\frac{\mathbf{I} (1 - \beta)}{\mathbf{U} \mathbf{G} \mathbf{I} (1 - \beta) \mathbf{H}_0 \mathbf{I} + \mathbf{I}} \right]$$

$$p = \frac{1}{N} \text{Tr} \left[\frac{\mathbf{H}_0 \mathbf{I}}{\mathbf{U} \mathbf{G} \mathbf{I} (1 - \beta) \mathbf{H}_0 \mathbf{I} + \mathbf{I}} \right]$$

$$t = \frac{1}{N} \text{Tr} \left[\frac{\mathbf{U} \mathbf{G} \mathbf{I}}{\mathbf{U} \mathbf{G} \mathbf{I} (1 - \beta) \mathbf{H}_0 \mathbf{I} + \mathbf{I}} \right]$$

- I_2 same with! – also expressions for variance

