Applications of Random Matrix Theory on Fiber Optical Communications

A. Moustakas

A. Karadimitrakis (Athens)
P. Vivo (KCL)
R. Couillet (Paris-Central)
L. Sanguinetti (Pisa)
A. Muller (Huawei)
H. Hafermann (Huawei)



• <u>Need for high speed infrastructure</u>

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

-Wired: IPTV, telepresence,

online gaming, live streaming etc.

- "Capacity- crunch" eminent



– <u>Solutions so far</u>

- Optical Networks (WDM-DWDM) have exploited many degrees of freedom (BW, available power, polarization diversity) except one: spatial.
- Soon the online devices will exceed the number of global population!
- Use of optical Space Division Multiplexing (SDM) is compelling.
 - Ideally: without changing the already infrastructure...



• <u>What is Optical MIMO?</u>

Just like in wireless domain, in optical we can use N 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)



Problem: Crosstalk phenomenon arises



Due to :

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



Problem: Crosstalk phenomenon

- Two Approaches:
 - Fight it



Section of 7-core optical fiber



Problem: Crosstalk phenomenon

- Two Approaches:
 - Take advantage of it

"Classic" MIMO techniques required but with some twists:

- Low power constraints to avoid non-linear behavior.
- Optical channel matrix just a subset of a unitary matrix.



System Model

Consider a single-segment *N*-channel lossless optical fiber system: $N_t \le N$ transmitting channels excited, $N_r \le N$ receiving channels coherently. The $2N \times 2N$ scattering matrix is

$$\mathbf{S} = \begin{bmatrix} \mathbf{r}_{t} & \mathbf{t} \\ \mathbf{t}^{\mathrm{T}} & \mathbf{r}_{r} \end{bmatrix} \quad (\mathbf{S} = \mathbf{S}^{\mathrm{T}})$$

Only **t** (Haar-distributed $\mathbf{t}^{\dagger}\mathbf{t} = \mathbf{t}\mathbf{t}^{\dagger} = \mathbf{I}_{N}$) sub-matrix is of interest (**r** is ~0).

Generally N_r , $N_t < N$:

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



Only **t** (Haar-distributed $\mathbf{t}^{\dagger}\mathbf{t} = \mathbf{t}\mathbf{t}^{\dagger} = \mathbf{I}_{N}$) sub-matrix is of interest (**r** is ~0).

$$\mathbf{t} = \begin{bmatrix} t_{11} & \dots & t_{1N} \\ & \dots & \\ t_{N1} & \dots & t_{NN} \end{bmatrix}$$

Define $N_t \ge N_r$ matrix **U** as

$$\mathbf{U} = \mathbf{P}_{N_t}^{T} \mathbf{S} \mathbf{P}_{N_r}$$

• where P projection operator:

$$\mathbf{P}_{N_t} = \begin{bmatrix} \mathbf{I}_{N_t} \\ \mathbf{0} \end{bmatrix} = \begin{bmatrix} \mathbf{N}_{N_t} \\ \mathbf{N}_{N-N_t} \end{bmatrix}$$



Channel Equation:

$$\mathbf{y} = \mathbf{U}\mathbf{x} + \mathbf{z}$$

Assume no differential delays between channels (*frequency flat fading*)the mutual information is

$$I_N(\mathbf{U}) = \log \det \left(\mathbf{I} + \rho \mathbf{U}^{\dagger} \mathbf{U} \right) = \sum_{k=1}^{N_t} \log \left(1 + \rho \lambda_k \right)$$

- Gaussian noise z
- Receiver knows the channel (pilot)
- Transmitter does not know the channel
- w.l.o.g. $N_t \leq N_r$



– Outage Probability:

$$P_{out}(r) = \operatorname{Prob}(I_N < N_t r) = E_{\mathbf{U}}[\Theta(N_t r - I_N(\mathbf{U}))]$$

• Optimal: Assumes infinite codewords

What is the price of finite codelengths?

- Gallager error bound for *M*-length code:

$$P_{err}(r) < E_{\mathbf{U}}\left[\exp\left[M\max_{0 \le k \le 1}\left(N_{t}kr - k\log\det\left[\mathbf{I} + \frac{\rho}{1+k}\mathbf{U}^{\dagger}\mathbf{U}\right]\right)\right]\right]$$



10

• Joint probability distribution of eigenvalues of $\mathbf{U}^{\dagger}\mathbf{U}$

$$P(\lambda_1, \lambda_2, \dots, \lambda_{N_t}) \propto \exp\left[N_0 \sum_k \log(1 - \lambda_k) + (N_r - N_t) \sum_k \log \lambda_k + 2 \sum_{k > m} \log |\lambda_k - \lambda_m|\right]$$

$$- N_0 = N - N_1 - N_2 \ge 0$$

- Exponent is energy of point charges repelling logarithmically in the presence of external field $P \propto e^{-N_t^2 S_0[p]}$
- Large N: charges coalesce to density

$$S_{0} = \int dx \ p(x)V_{eff}(x) - \int \int dx dy \ p(y)p(x)\log|x-y|$$
$$V_{eff}(x) = -n\log(1-x) - (\beta - 1)\log x$$
$$- \text{ where } n = \frac{N_{0}}{N_{t}} \quad \beta = \frac{N_{r}}{N_{t}}$$

• Minimizing (convex) S_0 w.r.t p(x) gives the "Marcenko-Pastur" distribution

$$p_{MP}(x) = \frac{\sqrt{(x-a)(b-x)}}{2\pi x(1-x)} \qquad a,b = \left(\frac{\sqrt{n+1} \pm \sqrt{\beta(n+\beta)}}{n+\beta+1}\right)^2$$



• We need tails of distribution: Optical Comms operate at $p_{out} \approx 10^{-8}$

$$P_{out}(r) = \operatorname{Prob}(I_N < N_t r) = E_{\mathbf{U}}[\Theta(N_t r - I_N(\mathbf{U}))]$$

- Fourier transform (or use large deviations arguments) $S \rightarrow S_0 - k \int dx \ p(x) (\log(1 + \rho x) - r)$ $V_{eff}(x) \rightarrow V_{eff}(x) - k \log(1 + \rho x)$
- k plays role of strength of logarithmic attraction/repulsion at $x_0 = -\rho^{-1}$
 - k>0 shifts charge density to larger values (R>Rerg), k<0 to smaller ones (R<Rerg)
- Minimizing S w.r.t p(x) gives the *generalized* Marcenko Pastur distribution
- Equivalent to balancing forces on the charge located at *x*:

$$2P\int \frac{p(x')}{x-x'} dx = \frac{n}{1-x} - \frac{\beta - 1}{x} - \frac{k\rho}{1+k\rho}$$



Generalized MP equation

- Use Tricomi theorem to calculate p(x)
- Obtain closed form expr. for energy S[p]
- E.g. $\beta > 1; n > 0$

$$p(x) = \frac{\sqrt{(b-x)(x-a)}}{2\pi x(1+\rho x)} \left(\frac{n(1+\rho)}{(1-x)\sqrt{(1-a)(1-b)}} + \frac{\beta-1}{x\sqrt{ab}}\right)$$

- a, b, k calculated from

$$p(b) = p(a) = 0$$

$$r = \int_{a}^{b} dx p(x) \log(1 + \rho x)$$

$$1 = \int_{a}^{b} dx p(x)$$





• In general

	S _{0b}	S _{ab}	S ₀₁	S _{a1}
	a=0 b<1	a>0 b<1	a=0 b=1	a>0 b=1
n=0; β=1	r <r<sub>c1</r<sub>		$r_{c1} < r < r_{c2}$	r>r _{c2}
n>0; β=1	r <r<sub>c3</r<sub>	r>r _{c3}		
n=0; β>1		r <r<sub>c4</r<sub>		r>r _{c4}
n>0; β>1		all r		

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



Finally $f(r) \approx N_t \frac{e^{-N_t^2 \left[S(r) - S(r_{erg})\right]}}{\sqrt{2\pi v_{erg}}}$

– where

$$v_{erg} = \frac{1}{S''(r_{erg})} = \log\left[\frac{\left(\sqrt{1+\rho b_0} - \sqrt{1+\rho a_0}\right)^2}{4\sqrt{1+\rho a_0}\sqrt{1+\rho b_0}}\right]$$

is the variance at the peak of the distribution.



15

Numerical Simulations ($\beta > 1$ and $n_0 > 0$)



The LD approach demonstrates better behavior, following Monte Carlo.





Numerical Simulations ($\beta > 1$ and $n_0 > 0$)



The LD approach demonstrates better behavior, following Monte Carlo.

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



$$P_{err}(r) < E_{\mathbf{U}}\left[\exp\left[M\max_{0 \le k \le 1}\left(N_{t}kr - k\sum_{j}\log\left(1 + \frac{\rho}{1+k}\lambda_{j}\right)\right)\right]\right]$$

• Here we need $S \to S_0 + k\alpha \int dx \ p(x) \left[\log \left(1 + \frac{\rho}{1+k} x \right) - r \right]$

$$V_{eff}(x) \rightarrow V_{eff}(x) + k\alpha \log\left(1 + \frac{\rho}{1+k}x\right)$$

- where
$$\alpha = \frac{M}{N_t}$$

- Now k is bounded in [0,1]
- Also when k < 1

$$r = \int_{a}^{b} dx p(x) \left[\log \left(1 + \frac{\rho}{1+k} x \right) + \frac{k}{1+k} \frac{\rho x}{1+\rho x} \right]$$

- otherwise no constraint







• Chaotic cavity picture:



• Assuming very low backscattering at edges we obtain

$$\mathbf{U} = \mathbf{P}_r^{+} \mathbf{S} \mathbf{P}_t = -c \mathbf{P}_r^{+} (\mathbf{H}_0 + \gamma \mathbf{G} + i \mathbf{\Gamma})^{-1} \mathbf{P}_t$$

- Deterministic (mode energy) \mathbf{H}_0
- Random (complex Gaussian) **G**
- Diagonal loss matrix Γ



• Mutual Information metric:

$$I(\mathbf{G}) = \log \det \left[\mathbf{I} + \rho \mathbf{P}_r^{+} (\mathbf{H}_0 + \gamma \mathbf{G} + i\Gamma)^{-1} \mathbf{P}_t \mathbf{P}_t^{+} (\mathbf{H}_0 + \gamma \mathbf{G} - i\Gamma)^{-1} \mathbf{P}_r \right]$$

- I Distribution (mean-variance) can be obtained using replica theory $E[I(\mathbf{G})] = I_1 - I_2$ $I_1 = \log \det[(\rho + \delta + \gamma t)(1 + \gamma r) + (\mathbf{H}_0 - \gamma p)^2] - N(tr - p^2)$ $r = \frac{1}{N}Tr[\frac{\gamma(1 + \gamma r)}{(\rho + \delta + \gamma t)(1 + \gamma r) + (\mathbf{H}_0 - \gamma p)^2}]$ $p = \frac{1}{N}Tr[\frac{\gamma(\mathbf{H}_0 - \gamma p)}{(\rho + \delta + \gamma t)(1 + \gamma r) + (\mathbf{H}_0 - \gamma p)^2}]$ $t = \frac{1}{N}Tr[\frac{\gamma(\rho + \delta + \gamma t)}{(\rho + \delta + \gamma t)(1 + \gamma r) + (\mathbf{H}_0 - \gamma p)^2}]$
 - I_2 same with $\rho=0$ also expressions for variance



Numerical Simulations







- MIMO: promising idea in Optical Communications
- In this work:
 - Simple model for Optical MIMO channel
 - Large Deviation Approach provides tails for MIMO mutual information
 - Method provides metric for outage throughput and finite blocklength error
- Many issues still open:
 - Channel modeling still at its infancy
 - Transmitter/Receiver Architectures
 - Multiple fiber segments
 - Nonlinearities: Signal becomes interference

- ...

