

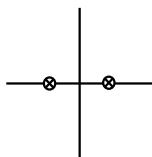
Random Matrices in Communication Theory

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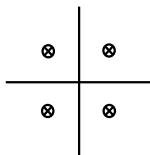
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Message transmission: an abstraction

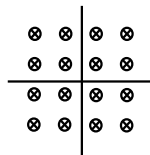
- ▶ K transmit antennas sending messages
- ▶ Message is a signal $x_k \in S \subset \mathbb{C}^K$ carried on antenna k



1 bit



2 bits



4 bits

- ▶ Each message is equiprobable
- ▶ Message requires energy $\mathbb{E}[|x_k|^2]$
- ▶ Can scale the constellation S

Impairments to transmitted signal

- ▶ L receive antennas
- ▶ Channel gain from transmit antenna k to receive antenna l is H_{lk}
- ▶ Received signal:

$$y = Hx + z$$

- ▶ $H \in \mathbb{C}^{L \times K}$, $z \in \mathbb{C}^L$
- ▶ H : random - iid entries distributed as $\mathbb{CN}(1)$, but realisation learnable at the receiver
- ▶ z : random - $\mathbb{CN}(I_L)$

- ▶ x : random - equiprobable in the message set with $\mathbb{E}[\|x\|^2] \leq P$ (battery life versus performance)
- ▶ Same constellation on each antenna $\Rightarrow \mathbb{E}[|x_k|^2] \leq P/K = p$.

How do these impairments arise

Random H with iid $\mathbb{CN}(1)$ entries:

- ▶ Arises because of scattering and many multipaths, with no dominant contribution from a single path
- ▶ No direct line of sight path
- ▶ If direct line of sight, mean is nonzero – also of interest, but not considered here

Random z : Receiver thermal noise in the receiving antenna

I: Receiver problem - MMSE

- ▶ Given y and H , process it and identify x
- ▶ Test each hypothesis for x , pick the most likely one (after having observed y and H). Exponential complexity. Want simpler receivers
- ▶ E.g., A linear receiver that minimises mean squared error (MMSE)

$$\rho := \min_{M \in \mathbb{C}^{K \times L}} \mathbb{E} [\|x - My\|^2 \mid H]$$

Questions:

Find the argmin M_{opt} . It can depend on H .

Find ρ

The solution

Proposition:

- ▶ $M_{\text{opt}} = \rho H^* [I_L + \rho H H^*]^{-1}$,
- ▶ $\rho = \rho \text{tr} \left\{ (I_L + \rho H H^*)^{-1} \right\}$.

Observations:

- ▶ One random environment H . The optimal receiver can depend on H .
- ▶ Scaling very natural for the problem:

$$\rho = \frac{P}{K} \sum_{k=1}^K \frac{1}{1 + P \lambda_k \left(\frac{H^* H}{K} \right)}$$

- ▶ Just a scaled Stieltjes transform evaluated at a point.

Proof steps: standard fare in estimation

- ▶ If x were Gaussian, the best estimate is $\mathbb{E}[x|y]$, a linear estimate.
- ▶ Even otherwise, look for the best estimate within the affine family (Wiener-Kolmogorov filtering)
- ▶ Projection of x onto the affine family
- ▶ Let $\mathbb{E}[x] = 0$. Then

$$\begin{aligned}M_{\text{opt}} &= (\mathbb{E}[xy^*]) (\mathbb{E}[yy^*])^{-1} \\ &= \rho H^* [I_L + \rho H H^*]^{-1}\end{aligned}$$

- ▶ MMSE (ρ) evaluation is by direct substitution.

The Marcenko-Pastur law

Theorem: Let L_K be the ESD of H^*H/K , and \bar{L}_K its expectation. The entries are iid, zero mean, with variance 1 and bounded fourth moments. Let $L/K \rightarrow \beta$. Then

$$\begin{aligned} L_K &\xrightarrow{P} \mu_{MP} \\ \bar{L}_K &\rightarrow \mu_{MP} \end{aligned}$$

Proof: The tough exercise via Stieltjes transforms with the antidiagonal trick that I didn't do.

The law μ_{MP} has generalised density

$$f_{\beta}(x) = (1 - \beta)^+ \delta(x) + \frac{\sqrt{(x - a)^+(b - x)^+}}{2\pi x}$$

where $a = (1 - \sqrt{\beta})^2$ and $b = (1 + \sqrt{\beta})^2$.

Observations: Mass at 0 and bounded support

II : CDMA for Code division multiple access

- ▶ K mobiles, each having a message to transmit
- ▶ The message x_k for the mobile k is “spread” over L symbols
- ▶ Transmit $x_k[H_{1k} H_{2k} \dots H_{Lk}]$ in the L symbols (signature)
- ▶ Superposition:

$$y = \sum_{k=1}^K \begin{bmatrix} H_{1k} \\ H_{2k} \\ \vdots \\ H_{Lk} \end{bmatrix} x_k + z = Hx + z$$

The twist: H_{lk} randomly picked from ± 1 with equal probability, independent of all others. Then normalised.

- ▶ Receiver informed (*pseudo noise* random bit generator)
- ▶ Helps hide information if seed is not known
- ▶ Use MMSE again. The same as the previous problem, with expectation over H as well.

III: The best code and Shannon capacity

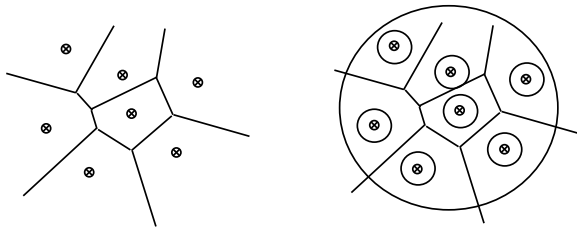
Code across time and exploit the law of large numbers. (Shannon 1948)

- ▶ $y(t) = H(t)x(t) + z(t)$, for $t = 1, 2, \dots, T$.
- ▶ For now, fix $H(t) = H$, fixed for all t , known.
- ▶ Code: Messages are $\{1, 2, \dots, M\}$. Each message maps to a position that the transmitter takes.

$$w \mapsto x^{(w)} \in (\mathbb{C}^K)^T$$

- ▶ Noise corrupts this and receiver gets corrupted $y \in (\mathbb{C}^L)^T$.
- ▶ Receiver should make probability of error arbitrarily small.
- ▶ What is $\log_2 M$ (bits per T symbols)? What is the rate $T^{-1} \log_2 M$? Maximum asymptotic rate? (Capacity)?

The scalar case $H \in \mathbb{C}$: $K = L = 1$



- ▶ Noise vector concentrates with a radius \sqrt{T}
- ▶ All noise spheres around message points disjoint
- ▶ All noise spheres are within radius $\sqrt{T(|H|^2P + 1)}$ (Near orthogonality)
- ▶
$$M \leq \frac{\alpha_{2T}(\sqrt{T(|H|^2P + 1)})^{2T}}{\alpha_{2T}(\sqrt{T})^{2T}} = (1 + |H|^2P)^T$$
- ▶ Yields rate is at most $T^{-1} \log M \leq \log(1 + |H|^2P)$.
- ▶ Can show that this is indeed *achievable*

Multiple antennas?

Theorem: Let $\{H(t)\}$ is stationary and ergodic. Let H denote $H(1)$. Under sufficient regularity on this process, the capacity per transmit antenna is given by

$$\begin{aligned} C_K &= \mathbb{E} \left[K^{-1} \log \det(I_L + pHH^*) \right] \\ &= \mathbb{E} \left[K^{-1} \sum_{k=1}^K \log \left(1 + P \lambda_k \left(\frac{HH^*}{K} \right) \right) \right] \end{aligned}$$

for a scenario where receiver knows $H(\cdot)$ and transmitter knows only its distribution. The unit is bits/symbol/antenna.

Evaluation

For Gaussian entries and finite K :

- ▶ Density $\propto \Delta(\lambda)^2 \exp\{-\sum_i \lambda_i\} \prod_i \lambda_i^{|K-L|}$
- ▶ Laguerre polynomials play the role of Hermite polynomials
- ▶ $p_{\lambda_1}(\lambda_1) = \frac{1}{m} \sum_{i=1}^m \phi_i(\lambda_1)^2 \lambda_1^{|K-L|} e^{-\lambda_1}$.
- ▶ Enables evaluation.

Asymptotic case as $K \rightarrow \infty$, we have

$$C_K = \int_0^\infty \log(1 + P\lambda) d\bar{L}_K(\lambda) \rightarrow \int_0^\infty \log(1 + P\lambda) d\mu_{MP}(\lambda).$$

$\log(1 + \cdot)$ is not bounded, but exploit monotonicity and concavity.
See a later slide.

IV: The random environment case

- ▶ Design a code with a certain rate R .
- ▶ Encounter a random environment H .
- ▶ Can have a failed transmission with probability

$$\begin{aligned} & \Pr \left\{ \int \log(1 + P\lambda) dL_K(\lambda) < R \right\} \\ &= \Pr \left\{ K^{-1} \sum_{k=1}^K \log \left(1 + P\lambda_k \left(\frac{HH^*}{K} \right) \right) < R \right\}. \end{aligned}$$

For this to converge to 0, need convergence in probability to a constant c , and $R < c - \epsilon$.

Convergence in probability

Theorem: Let $\xi_K = \int \log(1 + P\lambda) dL_K(\lambda)$. Then

$$\xi_K \xrightarrow{P} c = \int \log(1 + P\lambda) d\mu_{MP}(\lambda).$$

Proof: There is a common prob. space where $L_K \rightarrow \mu_{MP}$ a.s. (Skorohod). Consider the good set.

Let $\lambda^{(K)}$ be according to prob. measure L_K with distribution F_K .

Let G_K be distribution of $h(\lambda^{(K)}) := \log(1 + P\lambda^{(K)})$.

$$\begin{aligned}\xi_K &= \int g dG_K(g) = \int [1 - G_K(g)] dg = \int [1 - F_K(h^{-1}(g))] dg \\ &= \int [1 - F_K(\lambda)] h'(\lambda) d\lambda\end{aligned}$$

Apply BCT and ($F_K \rightarrow F$ a.e) to get $\xi_K \rightarrow c$ a.s.

So $\xi_K \rightarrow c$ in distribution, which also holds in the original space.

Convergence in distribution to a constant implies convergence in probability.

Variations

- ▶ Fluctuation result:

$$\Pr \left\{ \sum_{k=1}^K \log \left(1 + P \lambda_k \left(\frac{HH^*}{K} \right) \right) < Kc - u \right\} \rightarrow \Phi(u; \sigma^2)$$

- ▶ $y = HAx + z$, where A is a diagonal matrix. Correlations introduced.
- ▶ $H = U_r H_w U_t^*$
- ▶ Instead of $K_X = (P/K)I_K$, can tune K_X to H , if transmitter also has knowledge of the channel condition. This is not absurd in some cases.

References

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