

Bit Error Rate is Convex at High SNR

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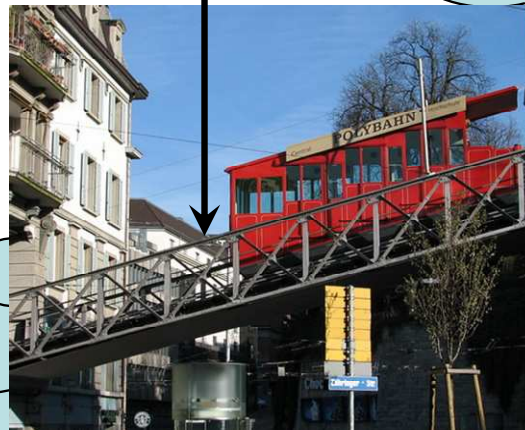
MOTIVATION

- **Importance of convexity¹**
 - strong analytical structure (convex problem ~ linear problem)
 - optimization problems: global solution
 - powerful numerical/analytical techniques
 - significant insight (even if no solution)
- **BER/SER is in the core of digital communications**
 - important performance objective
 - subject to analysis/optimization/design
- **Various applications/insights/generalizations**

¹ see e.g. S. Boyd, L. Vandenberghe, Convex Optimization, Cambridge University Press, 2004.

Digital
Communications

bridge



Convex Analysis

- strong analytical structure
- analytical/numerical techniques
- algorithms
- global solutions
- convergence

SYSTEM MODEL: AWGN CHANNEL

Baseband, discrete-time AWGN channel

$$\mathbf{r} = \mathbf{s} + \boldsymbol{\xi}$$

\mathbf{s} , \mathbf{r} are the Tx and Rx symbols,

$\mathbf{s} \in \{\mathbf{s}_1, \mathbf{s}_2, \dots, \mathbf{s}_M\}$, a set of M symbols (constellation/codebook)

$\boldsymbol{\xi} \sim \mathcal{CN}(\mathbf{0}, \sigma_0^2 \mathbf{I})$ is AWGN,

$$p_{\boldsymbol{\xi}}(\mathbf{x}) = \left(\frac{1}{2\pi\sigma_0^2} \right)^{\frac{n}{2}} e^{-\frac{|\mathbf{x}|^2}{2\sigma_0^2}}$$

Fading channel can be considered as well.

MAXIMUM LIKELIHOOD (ML) DETECTOR

The maximum likelihood (ML) = minimum distance detector,

$$\hat{\mathbf{s}} = \arg \min_{\mathbf{s}} |\mathbf{y} - \mathbf{s}|$$

Symbol error rate ($\text{SER}_i / \text{SER}$),

$$P_{ei} = \text{SER}_i = \Pr[\hat{\mathbf{s}} \neq \mathbf{s}_i | \mathbf{s} = \mathbf{s}_i], \quad P_e = \text{SER} = \sum_{i=1}^M P_{ei} \Pr[\mathbf{s} = \mathbf{s}_i]$$

Bit error rate (BER):

$$\text{BER} = \sum_{i=1}^M \sum_{j \neq i} \frac{h_{ij}}{\log_2 M} \Pr\{\mathbf{s} = \mathbf{s}_i\} \Pr\{\mathbf{s}_i \rightarrow \mathbf{s}_j\}$$

PRIOR RESULTS²: CONVEXITY OF SER

Theorem 1:

- $n \leq 2$: SER is *convex in low dimensions*, (any constellation/coding), $d^2 \text{SER} / d\text{SNR}^2 > 0$,
- $n > 2$:
 - SER is *convex at high SNR*, $\text{SNR} \geq (n + \sqrt{2n}) / d_{\min}^2$
 - *concave at low*, $\text{SNR} \leq (n - \sqrt{2n}) / d_{\max}^2$
 - odd number of inflection points in-between.

² S. Loyka, V. Kostina, F. Gagnon, Error Rates of the Maximum-Likelihood Detector for Arbitrary Constellations: Convex/Concave Behavior and Applications, IEEE Trans. Info. Theory, 2009, accepted.

RECENT IMPROVEMENTS

Theorem 1a: High/low SNR bounds can be strengthened as

High SNR: $\text{SNR} \geq (n - 2)/d_{\min}^2$

Low SNR: $\text{SNR} \leq (n - 2)/d_{\max}^2$

and no further improvement is possible.

For spherical decision regions: sufficient *and* necessary.

Extensions:

- not only ML, any detector such that
 $\text{Sphere}(R = d_{\min}) \in \Omega_i$ (simply-bounded)
- any unimodal noise power density (e.g. Laplace)

(BEAUTIFUL) BOUNDS ON SER DERIVATIVES IN SNR

$$-\frac{c_n}{\gamma} \leq P'_{e|\gamma} \leq 0, \quad , \quad c_1 = \frac{1}{\sqrt{2\pi e}}, \quad c_2 = \frac{1}{e}$$

$$\frac{\beta_l}{\gamma^2} \leq P''_{e|\gamma} \leq \frac{\beta_u}{\gamma^2}, \quad n = 2: \quad 0 \leq P''_{e|\gamma} \leq \left(\frac{2}{e\gamma}\right)^2$$

c_n, β_l, β_u are constants (depend on dimensionality, but not const. geometry).

MORE PRIOR RESULTS

- SER(noise power):
 - convex at high SNR
 - concave at low
 - odd number of inflex. points in-between
- Fading channels: average SER is
 - convex in low dimensions (mild condition)
 - “*Fading is Never Good*” in low dimensions
 - convex at high SNR (Rayleigh/Rice/polynomial at 0)

NUMEROUS ADVANTAGES/APPLICATIONS

- Strong analytical structure of convex problems³
- Efficient numerical algorithms, convergence, global solution
- Examples⁴ (many more in the literature):
 - BLAST optimization
 - Time/power sharing in decrease/increase SER
 - $n=1,2$: “fading is never good”
 - convexity: power/time sharing is no good (Tx)
 - concavity: power/time sharing is no good (jammer)

³ S. Boyd, L. Vandenberghe, Convex Optimization, Cambridge University Press, 2004.

⁴ S. Loyka, V. Kostina, F. Gagnon, Error Rates of the Maximum-Likelihood Detector for Arbitrary Constellations: Convex/Concave Behavior and Applications, IEEE Trans. Info. Theory, 2009, accepted.

NEW RESULTS: PEP IS CONVEX AT HIGH SNR

Theorem 3:

- Any n : the pairwise error probability (PEP) $\Pr\{\mathbf{s}_i \rightarrow \mathbf{s}_j\}$ is *convex at high SNR*, $\gamma \geq (n + \sqrt{2n}) / d_{\min,i}^2$,
- $n = 1, 2$: *concave at low SNR*, $\gamma \leq (n + \sqrt{2n}) / (d_{ij} + d_{\max,j})^2$,
inflections (odd) in-between,
- $n > 2$: *convex at low SNR*, $\gamma \leq (n - \sqrt{2n}) / (d_{ij} + d_{\max,j})^2$,
inflections (even) in-between.

MAIN RESULT: BER IS CONVEX AT HIGH SNR

Theorem 4:

- For *any* constellation and bit mapping (also coding), *BER is convex at high SNR* (under ML detection),

$$\text{SNR} \geq \frac{n + \sqrt{2n}}{d_{\min}^2} \rightarrow \frac{d^2 \text{BER}}{d \text{SNR}^2} > 0$$

CONVEXITY OF PEP/BER IN NOISE POWER

Theorem 5:

- the PEP is convex in noise power at high SNR (low noise).

Corollary 5.1:

- For *any* constellation and bit mapping (also coding), *BER is convex at high SNR,*

$$\text{SNR} = \frac{1}{\sigma_0^2} \geq \frac{n + 2 + \sqrt{2(n + 2)}}{d_{\min}^2} \rightarrow \frac{d^2 \text{BER}}{d(\sigma_0^2)^2} > 0$$

CONCLUSIONS

- **Importance of convexity/concavity**
- **SER is convex in low dimensions ($n=1,2$) or high SNR**
 - dimensionality is critical, but not geometry
 - in terms of both SNR and noise power
 - universal bounds on derivatives
- **BER/PEP is convex at high SNR**
 - any constellation/bit mapping/coding
 - in SNR and noise power
- **Various applications/advantages**
 - “fading is never good” in low dimensions