

# Least Favorable Additive Noise under a Divergence Constraint

Andrew L. McKellips and Sergio Verdú<sup>1</sup>

Department of Electrical Engineering, Princeton University, Princeton, NJ, U.S.A. 08544

Email andyman@ee.princeton.edu, verdu@ee.princeton.edu

**Abstract** — An uncertainty class of additive noise pdfs satisfying a constraint on divergence from a prescribed nominal is analyzed in the context of antipodal communication. From within this class, the noise pdf that maximizes detection error probability is determined for both zero-threshold and maximum-likelihood detection strategies. An optional additional constraint on SNR is also considered, and asymptotic behavior is studied for vanishing divergence tolerance.

## I. INTRODUCTION

Consider the hypothesis test for the normalized binary-input additive noise channel, given by

$$H_0 : Y = -1 + N \text{ versus } H_1 : Y = +1 + N$$

where  $N$  is a random variable with symmetric probability density function (pdf)  $f_N$ . The zero-threshold (ZT) detector (hard-limiter) exhibits error probability

$$P_{ZT}(f_N) = \int_1^\infty f_N(x) dx$$

while the maximum-likelihood (ML) detector achieves minimum error probability

$$P_{ML}(f_N) = \frac{1}{2} \int_{-\infty}^{\infty} \min\{f_N(x), f_N(x+2)\} dx ;$$

these quantities agree for symmetric unimodal pdfs. Often, the actual noise pdf is only known to resemble a given nominal pdf, leading us to define an additive noise uncertainty class by specifying a divergence neighborhood about a fixed pdf  $f_N$ , given by

$$\{\text{pdfs } f : \int_{-\infty}^{\infty} f(x) \log \frac{f(x)}{f_N(x)} dx \leq \delta\} .$$

We find the members of the uncertainty class that maximize  $P_{ZT}$  and  $P_{ML}$ . This formulation is of intrinsic interest in the analysis of AWGN channels subject to jamming, interference or other distortions. The results also represent the optimum strategy for a jammer who wishes to avoid detection, since discernibility between ambient noise and jammer-degraded noise is directly governed by their mutual divergence.

## II. ZERO-THRESHOLD DETECTION

The least favorable ZT pdf  $f_N^{ZT}$  is obtained from a prescribed nominal  $f_N$  through an area-preserving scalar weighting of the tails chosen to satisfy the divergence constraint. In the special case of small divergence tolerance  $\delta$ , we obtain the relation

$$P_{ZT}(f_N^{ZT}) = P_{ZT}(f_N) + \sqrt{P_{ZT}(f_N)(1 - 2P_{ZT}(f_N))\delta} + o(\sqrt{\delta}) .$$

<sup>1</sup>This work was supported in part by the U.S. Army Research Office under Grant DAAH04-96-1-0379.

## III. MAXIMUM-LIKELIHOOD DETECTION

Least favorable ML noise represents the maxmin solution to the detection error probability game of divergence-constrained noise versus unconstrained receiver. For symmetric continuous unimodal nominals  $f_N$  satisfying a local log-concavity condition (including Gaussians) the least favorable ML pdf  $f_N^{ML}$  is obtained from  $f_N$  by weighting the tails around a 2-periodic interval made up of weighted mod-2 geometric means. The weightings are mutually dependent through demonstrated continuity of  $f_N^{ML}$ , with the remaining degree of freedom chosen to preserve area. The size of the 2-periodic interval is chosen to satisfy the divergence constraint.

## IV. ADDITIONAL POWER CONSTRAINT

In both ZT and ML cases, imposing the further constraint that noise power not exceed the nominal power leads to a least favorable pdf obtained by performing the described transformations on a variance-scaled nominal pdf

$$f_N(x) = C f_N(x) \exp\{-cx^2\} .$$

An upper bound on detection error probability for least favorable ML power- and divergence-constrained noise is realized by least favorable ML power-constrained noise [1] as depicted in Figure 1; under an appropriate condition on the support of  $f_N$ , the former converges in distribution to the latter for growing divergence tolerance.

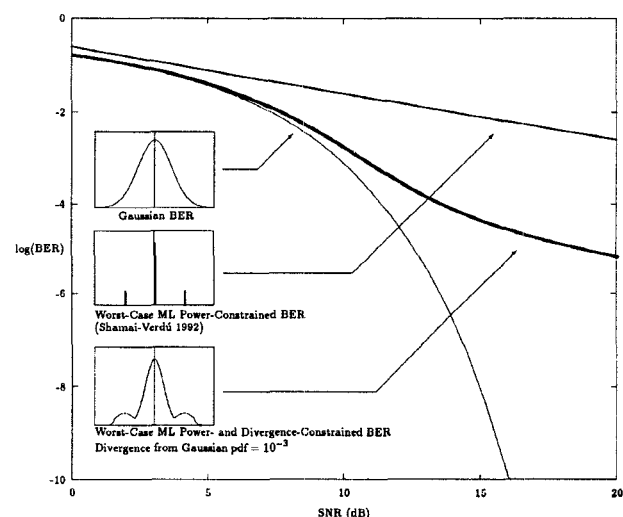


Figure 1: ML Bit-Error Rate under Various Constraints

## REFERENCES

- [1] S. Shamai (Shitz) and S. Verdú, "Worst-case power-constrained noise for binary-input channels," *IEEE Transactions on Information Theory*, vol. 38, no. 5, pp. 1494–1511, Sept. 1992.