

yacoub: a Python package for Simulating Generalized Fading Channels

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Abstract—We present a well tested Python-based library for simulating and computing generalized fading channels, named **yacoub**. We describe the applicability of **yacoub** using examples in recent communications systems challenges, namely: spectrum sensing, bit error rate computation, and parameter estimation in generalized fading channels. For the latter, a Majorization-Minimization algorithm is derived. The development of **yacoub** open source and its code is available at <http://github.com/mirca/yacoub>.

I. INTRODUCTION

A. Note on notation

Scalars and random variables are denoted as *italic* small-case letters *e.g.* x ; vectors and random vectors are denoted as *italic*, boldface, small-case letters *e.g.* \mathbf{x} . The n -th component of a vector \mathbf{x} is denoted as x_n . A complex vector of length n is defined as $\mathbf{x} \in \mathbb{C}^{n \times 1}$. All vectors are column vectors. Matrices are denoted as *italic*, boldface, capital letters as in \mathbf{X} ; the identity matrix of order n is denoted as \mathbf{I}_n . We define a discrete-time circularly symmetric Gaussian process \mathbf{z} as any (finite or infinite) collection of random variables $\mathbf{z} = \mathbf{x} + j\mathbf{y}$, $j \triangleq \sqrt{-1}$, such that \mathbf{x} and \mathbf{y} are iid jointly Gaussian with zero mean vector and covariance matrix given by $\mathbb{E}[\mathbf{z}\mathbf{z}^\dagger]$, in which \mathbf{z}^\dagger means the conjugate transpose of \mathbf{z} . The expectation value wrt a random variable x is denoted as \mathbb{E}_x . The probability of an event A is denoted as $\mathbb{P}(A)$. The indicator function is denoted as $\mathbb{I}(\cdot)$ and it evaluates to 1 if its argument is true and 0 otherwise.

II. THE ACCEPTANCE-REJECTION SAMPLER IN LOG-SPACE

III. EXAMPLES

A. Spectrum Sensing in Complex Generalized Fading Channels

The spectrum sensing problem consists in deciding whether or not a given channel frequency band is being occupied by a licensed (primary) user and, in case that such frequency band is available, how to opportunistically allocate secondary users such that the interference on the primary user is negligible.

From a probabilistic point of view, the spectrum sensing problem may be framed as a decision theory problem, as

follows

$$H_0 : \mathbf{y} = \mathbf{w}, \quad (1)$$

$$H_1 : \mathbf{y} = h\mathbf{s} + \mathbf{w}, \quad (2)$$

in which $\mathbf{y} \in \mathbb{C}^{n \times 1}$ is the decoded received vector signal, $\mathbf{w} \in \mathbb{C}^{n \times 1}$ is complex Gaussian noise process with zero mean vector and covariance matrix given as $\sigma^2 \mathbf{I}_n$, and h is the channel gain.

In [?], the authors have shown that the probability distribution of the energy statistic $\tilde{y} \triangleq \mathbf{y}^\dagger \mathbf{y}$ conditioned on the knowledge of h , in case that \mathbf{s} is an M -PSK signal such that every symbol has the same probability of occurrence, $\mathbb{P}(s_n = s) = \frac{1}{M}$, is given as

$$p(\tilde{y}|h, H_1) = 1 - Q_n \left(\sqrt{\frac{2n|h|^2 E_s}{\sigma^2}}, \sqrt{\frac{2\tilde{y}}{\sigma^2}} \right), \quad (3)$$

in which Q_n is the Marcum-Q function and E_s is the energy per symbol.

The pdf of \tilde{y} can be written using the Law of Total Expectation

$$p(\tilde{y}|H_1) = \mathbb{E}_h [p(\tilde{y}|h, H_1)] = \int_{-\infty}^{+\infty} p(\tilde{y}|h, H_1) p(h) dh. \quad (4)$$

Recall that the energy detection rule can be expressed as

$$d_\delta(\tilde{y}) = \mathbb{I}(\tilde{y} > \delta) \quad (5)$$

in which δ is a strictly positive real number known as energy threshold, and $d_\delta(\tilde{y}) = j$, $j \in \{0, 1\}$, means that the detector has decided in favor of the hypothesis H_j .

As a result, the probabilities of false alarm and miss detection can be written as

$$p_f \triangleq \mathbb{P}(d_\delta(\tilde{y}) = 1|H_0) = 1 - p(\delta|H_0), \quad (6)$$

$$p_d \triangleq \mathbb{P}(d_\delta(\tilde{y}) = 0|H_1) = \mathbb{E}_h [p(\delta, h|H_1)], \quad (7)$$

B. Parameter Estimation in α - μ fading

The α - μ fading probability density is given as

$$p(h) = \frac{\alpha \mu^\mu h^{\alpha\mu-1}}{\Gamma(\mu)} \exp(-\mu h^\alpha), \quad (8)$$

$h \in \mathbb{R}_+$, $\alpha \in \mathbb{R}_+$, and $\mu \in \mathbb{R}_+$.

Given a vector $\mathbf{h} \in \mathbb{R}^{n \times 1}$ of iid samples from the $\alpha - \mu$ distribution, we would like to estimate the value of α . The pdf of \mathbf{h} is given as

$$p(\mathbf{h}) = \prod_{i=1}^n p(h_i) = \prod_{i=1}^n \frac{\alpha \mu^\mu h_i^{\alpha \mu - 1}}{\Gamma(\mu)} \exp(-\mu h_i^\alpha) \quad (9)$$

$$= \left(\frac{\alpha \mu^\mu}{\Gamma(\mu)} \right)^n \exp \left(-\mu \sum_{i=1}^n h_i^\alpha \right) \prod_{i=1}^n h_i^{\alpha \mu - 1} \quad (10)$$

and the log likelihood function (up to an additive constant that does not depend upon α) is given as

$$\log p(\mathbf{h}) \cong n \log \alpha - \mu \sum_{i=1}^n h_i^\alpha + \alpha \mu \sum_{i=1}^n \log h_i \quad (11)$$

Noting that

$$\sum_{i=1}^n h_i^\alpha \geq \sum_{i=1}^n h_i^{\alpha_t} + (\alpha - \alpha_t) \sum_{i=1}^n h_i^{\alpha_t} \log h_i \quad (12)$$

and therefore

$$\log p(\mathbf{h}) \leq n \log \alpha - \mu \left(\sum_{i=1}^n h_i^{\alpha_t} + (\alpha - \alpha_t) \sum_{i=1}^n h_i^{\alpha_t} \log h_i \right) + \alpha \mu \sum_{i=1}^n \log h_i \quad (13)$$

C. BER in Complex $\alpha - \mu$ Fading

IV. CONCLUSIONS

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