PREDICTION OF FADING ENVELOPES WITH DIFFUSE SPECTRA

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ABSTRACT

We show how to compute the minimum mean squared prediction error of a mobile-radio fading envelope modeled as a stationary process arising from a diffuse set of local scatterers. Our approach reduces the prediction problem to an eigenvalue decomposition, and is appropriate when the prediction is to be based on errorcorrupted estimates of a narrowband fading process. Such prediction can improve the performance of adaptive modulation techniques that require up-to-date channel-state information for optimal performance.

1. INTRODUCTION

In mobile radio, adaptive modulation requires the use of fading-envelope prediction to overcome delays in feeding channel-state information back from receiver to transmitter [1,2]. In this paper, we show how to compute the minimum mean squared prediction error of a fading envelope modeled as a stationary process arising from a diffuse set of local scatterers. The prediction is based on estimates of the fading envelope over a finite interval of past values, with the estimation errors on this interval modeled as white noise. Our approach reduces the prediction problem to an eigenvalue decomposition [3, p. 242]. This has advantages over the morecommon spectrum-factorization approach [3, Sec. 11-6] when the prediction is to be based on error-corrupted estimates of a narrowband fading process, because the spectrum of such a process is nearly discontinuous, and is thus difficult to approximate with a rational function. Example curves are shown for a fading envelope having a Clarke-type Doppler spectrum [4].

2. PREDICTION OF THE FADING ENVELOPE

Let x(t) be a quadrature component of a fading envelope, modeled as a zero-mean, wide-sense stationary

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random process, and let

$$\tilde{x}(t) = x(t) + v(t), \tag{1}$$

where $\tilde{x}(t)$ is an estimate of past values of the fading envelope and v(t) represents the estimation errors. These errors are modeled as zero-mean white Gaussian noise with variance σ_v^2 , and are uncorrelated with x(t). We want to predict the value of x(r) for arbitrary r, given a known, T-length segment of $\tilde{x}(t)$, τ seconds in the past, using a linear predictor

$$\hat{x}(r) = \int_{r-\tau-T}^{r-\tau} \tilde{x}(t)h(r-t)dt.$$
 (2)

We seek a function h(t) that minimizes the mean squared prediction error,

$$J = \mathcal{E}\{[x(r) - \hat{x}(r)]^2\}.$$
 (3)

Because of the stationarity of $\tilde{x}(t)$, J does not depend on r. Thus, making use of (1), (2) and (3), and letting $r = \tau + \frac{T}{2}$ we have

$$J = \mathcal{E}\left\{ \left[x(\tau + \frac{T}{2}) - \int_{-\frac{T}{2}}^{\frac{T}{2}} \tilde{x}(t)h(\tau + \frac{T}{2} - t)dt \right]^{2} \right\}$$

$$= \mathcal{E}\left\{ \left[x(\tau + \frac{T}{2}) - \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t)h(\tau + \frac{T}{2} - t)dt - \int_{-\frac{T}{2}}^{\frac{T}{2}} v(t)h(\tau + \frac{T}{2} - t)dt \right]^{2} \right\}$$

$$= \mathcal{E}\left\{ \left[x(\tau + \frac{T}{2}) - \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t)h(\tau + \frac{T}{2} - t)dt \right]^{2} \right\}$$

$$+ \mathcal{E}\left\{ \left[\int_{-\frac{T}{2}}^{\frac{T}{2}} v(t)h(\tau + \frac{T}{2} - t)dt \right]^{2} \right\}$$
(4)

because x(t) and v(t) are uncorrelated. Note that we have chosen r such that the limits of integration in (4) are symmetric. This is advantageous since it leads to a symmetric eigenvalue problem in what follows.

3. PROBLEM TRANSFORMATION

We start by expressing h(t) as an expansion,

$$h(t) = \sum_{n=0}^{\infty} a_n \phi_n(\tau + \frac{T}{2} - t) \quad t \in [\tau, \tau + T].$$
 (5)

with $\{\phi_n(t)\}\$ the countable orthogonal solutions of

$$\int_{-\frac{T}{2}}^{\frac{T}{2}} \phi(s) R_{xx}(t-s) ds = \lambda \phi(t), \tag{6}$$

where $R_{xx}(t)$ is the autocorrelation of x(t). In keeping with the convention of Slepian, et al. [5], we scale each $\phi_n(t)$ such that

$$\int_{-\frac{T}{2}}^{\frac{T}{2}} \phi_n^2(t) dt = \lambda_n.$$
(7)

The expansion (5) is valid if $\{\phi_n(t)\}$ is complete on $t \in [-\frac{T}{2}, \frac{T}{2}]$. This is assured if $R_{xx}(t)$ is positive definite [3, p. 374]; that is, if

$$\int_{-\frac{T}{2}}^{\frac{T}{2}} \int_{-\frac{T}{2}}^{\frac{T}{2}} R_{xx}(t-s)f(s)f^*(t)\,ds\,dt > 0 \qquad (8)$$

for any f(t) with

$$\int_{-\frac{T}{2}}^{\frac{T}{2}} |f(t)|^2 dt > 0.$$
(9)

To determine a sufficient condition to satisfy (8), let f(t) = 0 for $t \notin [-\frac{T}{2}, \frac{T}{2}]$. Then, writing $R_{xx}(t)$ as the inverse Fourier transform of the power spectral density, $S_{xx}(\omega)$, and changing the order of integration, it is not difficult to show that

$$\int_{-\frac{T}{2}}^{\frac{T}{2}} \int_{-\frac{T}{2}}^{\frac{T}{2}} R_{xx}(t-s)f(s)f^{*}(t) \, ds \, dt$$

= $\frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xx}(\omega)|F(\omega)|^{2} d\omega,$ (10)

where $F(\omega)$ is the Fourier transform of f(t). Now f(t) it time limited and, from (9), not identically zero. Thus we can have $F(\omega) = 0$ only on a set of zero measure. Hence, if $S_{xx}(\omega) > 0$ on a set of positive measure, (10) shows that (8) is satisfied. This condition on $S_{xx}(\omega)$ is met whenever the fading envelope is modeled as arising from a diffuse set of local scatterers, as it often is when the number of such scatterers is considered to be large, and thus not individually resolvable (see e.g. Clarke [4]). Bearing this restriction in mind, and letting P_{xx} represent the power of x(t), we have

$$\begin{split} \mathcal{E}\left\{ \left[x(\tau + \frac{T}{2}) - \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t)h(\tau + \frac{T}{2} - t)dt \right]^2 \right\} \\ &= \mathcal{E}\left\{ x^2(\tau + \frac{T}{2}) \\ &- 2\int_{-\frac{T}{2}}^{\frac{T}{2}} x(\tau + \frac{T}{2})x(t)h(\tau + \frac{T}{2} - t)dt \\ &+ \int_{-\frac{T}{2}}^{\frac{T}{2}} \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t)x(s)h(\tau + \frac{T}{2} - t)dt \\ &+ \int_{-\frac{T}{2}}^{\frac{T}{2}} \int_{-\frac{T}{2}}^{\frac{T}{2}} R_{xx}(\tau + \frac{T}{2} - t)h(\tau + \frac{T}{2} - t)dt \\ &+ \int_{-\frac{T}{2}}^{\frac{T}{2}} \int_{-\frac{T}{2}}^{\frac{T}{2}} R_{xx}(t - s)h(\tau + \frac{T}{2} - t)dt \\ &+ \int_{-\frac{T}{2}}^{\frac{T}{2}} \int_{-\frac{T}{2}}^{\frac{T}{2}} R_{xx}(t - s)h(\tau + \frac{T}{2} - t)dt \\ &+ \int_{-\frac{T}{2}}^{\frac{T}{2}} \int_{-\frac{T}{2}}^{\frac{T}{2}} R_{xx}(t - s)h(\tau + \frac{T}{2} - t) \\ &\cdot h(\tau + \frac{T}{2} - s)dtds \\ &= P_{xx} \\ &- 2\int_{-\frac{T}{2}}^{\frac{T}{2}} R_{xx}(t - s) \\ &\cdot \sum_{n=0}^{\infty} \sum_{p=0}^{\infty} a_n a_p \phi_n(t)\phi_p(s)dtds \\ &= P_{xx} \\ &- 2\sum_{n=0}^{\infty} a_n \int_{-\frac{T}{2}}^{\frac{T}{2}} \phi_n(t)R_{xx}(\tau + \frac{T}{2} - t)dt \\ &+ \sum_{n=0}^{\infty} \sum_{p=0}^{\infty} a_n a_p \int_{-\frac{T}{2}}^{\frac{T}{2}} \phi_n(t) \\ &\cdot \left[\int_{-\frac{T}{2}}^{\frac{T}{2}} \phi_p(s)R_{xx}(t - s)ds \right]dt \\ &= P_{xx} - 2\sum_{n=0}^{\infty} a_n a_p \lambda_p \int_{-\frac{T}{2}}^{\frac{T}{2}} \phi_n(t)\phi_p(t)dt \\ &= P_{xx} - 2\sum_{n=0}^{\infty} a_n \lambda_n \phi_n(\tau + \frac{T}{2}) \\ &+ \sum_{n=0}^{\infty} \sum_{p=0}^{\infty} a_n \lambda_n \phi_n(\tau + \frac{T}{2}) + a_n^2 \lambda_n^2, \end{split}$$

where (5), (7) and the orthogonality of $\{\phi_n(t)\}\$ were

used, and where each $\phi_n(t)$ is defined for all t by (6). Considering now the noise term of (4) we have

$$\mathcal{E}\left\{ \left[\int_{-\frac{T}{2}}^{\frac{T}{2}} v(t)h(\tau + \frac{T}{2} - t)dt \right]^2 \right\}$$

$$= \sigma_v^2 \int_{-\frac{T}{2}}^{\frac{T}{2}} h^2(\tau + \frac{T}{2} - t)dt$$

$$= \sigma_v^2 \int_{-\frac{T}{2}}^{\frac{T}{2}} \left[\sum_{n=0}^{\infty} a_n \phi_n(t) \right]^2 dt$$

$$= \sigma_v^2 \sum_{n=0}^{\infty} a_n^2 \int_{-\frac{T}{2}}^{\frac{T}{2}} \phi_n^2(t)dt$$

$$= \sigma_v^2 \sum_{n=0}^{\infty} a_n^2 \lambda_n.$$
(12)

4. SOLUTION OF THE PREDICTOR

Combining (4), (11), and (12) we have

$$J = P_{xx} + \sum_{n=0}^{\infty} -2a_n\lambda_n\phi_n(\tau + \frac{T}{2}) + a_n^2\lambda_n(\lambda_n + \sigma_v^2).$$
(13)

Setting the partial derivatives to zero we have

$$\frac{\partial J}{\partial a_k} = -2\lambda_k \phi_k(\tau + \frac{T}{2}) + 2a_k \lambda_k(\lambda_k + \sigma_v^2) = 0. \quad (14)$$

Solving for a_k yields

$$a_k = \frac{\phi_k(\tau + \frac{T}{2})}{\lambda_k + \sigma_v^2}.$$
(15)

This may be substituted into (5) to obtain h(t). Thus, the predictor (2) is determined.

Note that the expression (15) for the coefficients $\{a_k\}$ has the same form as the expression we previously found for bandlimited processes with flat spectral densities [6, eq. 14]. In that case, σ_v^2 was interpreted as a Lagrange multiplier arising out of an energy constraint. The basis functions $\{\phi_n(t)\}$ were the prolate spheroidal wave functions [5], which are the solutions of (6) when

$$R_{xx}(t) = \frac{\sin \Omega t}{\pi t},\tag{16}$$

with Ω the band limits of the process x(t). The solution presented in this paper generalizes those results, by accommodating processes with a wider class of spectral densities, including non-bandlimited processes. On the other hand, since mobile-radio fading envelopes are often modeled as bandlimited [4,7,8], it is important to note that for such processes, the modeling of estimation errors v(t) in (4) is critical, since the optimization problem otherwise fails to have a solution [9].

When noise is modeled, the predictor problem can be solved by factoring the spectrum [3, Sec. 11.6], as is often done in Wiener problems. But for bandlimited and narrowband fading envelopes, this approach is difficult, because the signal-plus-noise spectrum is nearly discontinuous, and is not easy to approximate using a rational function.

We may also substitute (15) into (13) to get the minimum mean square prediction error,

$$J_{\min} = P_{xx} - \sum_{n=0}^{\infty} \lambda_n \frac{\phi_n^2 (\tau + \frac{T}{2})}{\lambda_n + \sigma_v^2}.$$
 (17)

5. COMPUTING BASIS FUNCTION VALUES

The basis functions $\{\phi_n(t)\}\$ and eigenvalues $\{\lambda_n\}\$ are solutions of the integral eigenvalue problem (6). We compute these solutions using the Nystrom method [10, p. 791]. First, we approximate the integral in (6) with a quadrature rule

$$\sum_{j=1}^{N} w_j \phi(t_j) R(t-t_j) = \lambda \phi(t), \qquad (18)$$

where $\{w_j\}$ are the weights of the quadrature rule and $\{t_j\}$ are the quadrature points. Now we evaluate this equation at the quadrature points,

$$\sum_{j=1}^{N} w_j \phi(t_j) R(t_i - t_j) = \lambda \phi(t_i) \quad i = 1, 2, \dots, N.$$
 (19)

Next, we let K be an $N \times N$ matrix with

$$[K]_{i,j} = w_j R(t_i - t_j), \tag{20}$$

and we let ϕ be an $N \times 1$ vector with

$$\phi = [\phi(t_1) \phi(t_2) \dots \phi(t_N)]^T.$$
(21)

Then (19) may be written as a matrix eigenvalue problem

$$K\phi = \lambda\phi. \tag{22}$$

This problem has N eigenvectors $\{\phi_n\}$ and N eigenvalues $\{\lambda_n\}$, which may be determined numerically and substituted into (18). The resulting equation may be solved for the eigenfunction

$$\phi_n(t) = \frac{1}{\lambda_n} \sum_{j=1}^N w_j \phi_n(t_j) R(t - t_j).$$
(23)



Fig. 1. Minimum mean squared prediction error of a Clarke-type fading envelope.

This allows us to evaluate $\phi_n(t)$ for t that are not quadrature points. Note that, in (23), only terms with significant eigenvalues need be used.

Also, from (20), note that K is not a symmetric matrix. Symmetry can be restored to the eigenvalue problem (22), though, using a straightforward technique [10, p. 794].

6. OBSERVATIONS AND CONCLUSIONS

Using Gaussian quadrature with N = 96 in (19) [11, p. 887], and making use of (17) and (23), we obtain the curves shown in Figure 1. These show the minimum mean squared prediction error of a unity-power Clarketype mobile-radio fading envelope with $R_{xx}(t) =$ $J_0(\Omega t)$ [4], where again, x(t) is an Ω -bandlimited process. The prediction is based on past estimates over an interval of length T = 0.2 sec. The level of whitenoise errors on this interval is expressed as a signal-tonoise ratio on the horizontal axis. The prediction is for $\tau = .04$ sec. in the future.

Note that, as the estimation SNR improves, the prediction error approaches zero. This is a general property of bandlimited processes [9]. Also, it is clear that the prediction error depends strongly on the maximum Doppler frequency $f_m = \Omega/2\pi$, but the curves in Figure 1 hardly vary from analogous ones plotted using (16). This suggests that the predictability of the mobile-radio fading envelope may be more profoundly affected by the maximum Doppler frequency than by the exact shape of the Doppler spectrum.

7. REFERENCES

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