

## 1 Signal Model and problem formulation

Assume that an array of  $M$  sensors receives  $K$  targets impinging from unknown time-varying directions. Over the observation interval, the Rayleigh fading amplitude of the target is assumed to vary in time according to Jakes' or first order autoregressive (AR1) correlation models. The  $M \times 1$  array snapshot complex vectors at the output can be expressed as

$$\mathbf{y}(n) = \sum_{k=1}^K \mathbf{a}(\theta_k(n)) h_k(n) + \mathbf{n}(n) = \mathbf{A}(\boldsymbol{\theta}_n) \mathbf{h}(n) + \mathbf{n}(n), \quad n = 0, \dots, N-1 \quad (1.1)$$

where  $\mathbf{A}(\boldsymbol{\theta}_n)$  is an  $M \times K$  ( $K < M$ ) matrix formed by the steering vector  $\mathbf{a}(\theta_k(n))$ . We suppose  $\|\mathbf{a}(\theta_k(n))\|^2 = M$ . In (1.1),  $\boldsymbol{\theta}_n \stackrel{\text{def}}{=} (\theta_1(n), \dots, \theta_K(n))^T$  is the unknown time-varying DOA vector of  $K$  targets,  $\mathbf{h}(n) = (h_1(n), \dots, h_K(n))^T$  is the time-varying vector of complex amplitudes of  $K$  target returns. The process  $h_k(n)$  is the sample of the fading amplitude of the  $k$ th target assumed to be zero-mean circular complex Gaussian with unknown variance  $\sigma_k^2$  and correlation function given by:

$$R_h^J(m) \stackrel{\text{def}}{=} \sigma_h^2 \mathbb{E}(h_k(n) h_k^*(n-m)) = \sigma_k^2 J_0(2\pi f_d T m),$$

where  $J_0(\cdot)$  is the first kind 0th-order Bessel function,  $T$  is the symbol period and  $f_d$  denotes the maximum Doppler shift. This is frequently referred to as the Jakes' model [3].

**AR1 model of fading** Among various channel models, the information theoretic results in [4] show that the first-order AR model provides a sufficiently accurate model for time fading channels

$$h_k(n) = \gamma h_k(n-1) + e_k(n) \quad (1.2)$$

where  $e_k(n) \sim \mathcal{N}(0, \sigma_k^2(1-\gamma^2))$  is the additive driving noise and where  $\gamma \stackrel{\text{def}}{=} J_0(2\pi f_d T)$  is assumed to be unknown.

For simplicity, the time-varying  $\boldsymbol{\theta}(n)$  is modeled as

$$\boldsymbol{\theta}(n) = \boldsymbol{\theta}_0 + n\boldsymbol{\theta}_1, \quad (1.3)$$

where  $\boldsymbol{\theta}_0 = (\theta_{01}, \dots, \theta_{0K})^T$  and  $\boldsymbol{\theta}_1 = (\theta_{11}, \dots, \theta_{1K})^T$ . The linear polynomial (1.3) can be seen as a truncated Taylor expansion which gives a good description for the source motion in a small observation [6, 7].

Having the model for the variation of the channel (1.2) and the DOA (1.3), and from (1.1), we can obtain the following state space representation of the problem

$$\begin{cases} \mathbf{y}(n) = \mathbf{A}(\boldsymbol{\theta}(n)) \mathbf{h}(n) + \mathbf{n}(n) \\ \mathbf{h}(n) = \gamma \mathbf{h}(n-1) + \mathbf{e}(n) \\ \boldsymbol{\theta}(n) = \boldsymbol{\theta}_0 + n\boldsymbol{\theta}_1 \end{cases} \quad (1.4)$$

where  $\mathbf{e}_n \stackrel{\text{def}}{=} (e_1(n), \dots, e_K(n))^T \sim \mathcal{N}(0, (1-\gamma^2)\mathbf{Q})$  and where  $\mathbf{Q} \stackrel{\text{def}}{=} \text{Diag}(\sigma_1^2, \dots, \sigma_K^2)$ .

The estimation problem can now be formulated as follows: Given the received signal  $\mathbf{y} = (\mathbf{y}(0), \dots, \mathbf{y}(N-1))^T$  an unknown parameter vector  $\boldsymbol{\alpha} \stackrel{\text{def}}{=} (\mathbf{h}^T, \gamma, \boldsymbol{\sigma}, \boldsymbol{\theta}_0^T, \boldsymbol{\theta}_1^T)^T$ , estimate  $\boldsymbol{\theta} \stackrel{\text{def}}{=} (\boldsymbol{\theta}_0^T, \boldsymbol{\theta}_1^T)^T$ . In this problem,  $\boldsymbol{\theta}$  is the parameter of interest and the other parameters are nuisance parameters.

## 2 Objective

The main objective of this project is to extend the algorithm proposed in [1, 2] by combining the expectation-maximization (EM) algorithm with Kalman smoother algorithm to yield time-varying DOA estimation and ML estimate of channel parameters.

To evaluate the performance of this estimator it should be interesting to calculate the hybrid Cramér Rao-bound basing on the work presented in [5].

## References

- [1] T. Y. Al-Naffouri, “An EM-based forward-backward Kalman filter for the estimation of time-variant channels in OFDM,” *IEEE Transactions on Signal Processing*, vol. 57, no. 7, pp. 3924-3930, July 2007.
- [2] H. Abeida, J-M Brossier, L. Ros, J. Vilà Valls, “An EM algorithm for path delay and complex gain estimation of slowly varying fading channel for CPM signals,” *Proceedings - IEEE Globecom'09*, Dec. 2009.
- [3] W. C. Jakes, *Microwave Mobile Communications*, New York: Wiley, 1974.
- [4] H. Wang and P. Chang, “On verifying the first-order Markovian assumption for a Rayleigh fading channel model,” *IEEE Trans. Veh. Technol.*, vol. 45, pp. 353-357, May 1996.
- [5] S. Bay, C. Herzet, J. M. Brossier, J. P. Barbot and B. Geller, “Analytic and asymptotic analysis of Bayesian Cramer-Rao bound for dynamical phase offset estimation” *IEEE Trans. Signal Processing*, vol. 56, pp. 61-70, Jan. 2008.
- [6] V. Katkovnik, “A new concept of adaptive beamforming for moving sources and impulse noise environment,” *Signal processing*, vol. 80, pp. 1863-1882, 2000.
- [7] P. J. Chung and J. F. Bohme, “DOA Estimation of Multiple Moving Sources Using Recursive EM Algorithms,” *Proc. Sensor Array and Multi-channel Signal Processing Workshop* 323.327, Washington DC, USA, August, 2002.