Fading Performance Evaluation of an Adaptive MSER Beamforming Receiver for QAM Systems

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Motivation



System Model



$$\mathbf{b}(k) = \mathbf{P}(k) \xrightarrow{\mathbf{x}(k)} \mathbf{w}^{H}(k) = \mathbf{w}^{H}(k)$$
$$\mathbf{n}(k)$$

- $L-number \ of \ antenna \ array \ elements$
- $S-number \ of \ users$
- $\mathbf{b}(k)$ symbols transmitted by each user at time instant k
- $\mathbf{P}(k)$ describes the propagation channel and physical arrangement of antenna elements
- $\mathbf{n}(k)$ additive white Gaussian noise
- $\mathbf{x}(k)$ vector of received signal samples
- $\mathbf{w}(k)$ beamformer weights
- y(k) beamformer soft output

MMSE Beamforming

MMSE-Minimum Mean-Squared-Error

Minimise the error between estimated and transmitted signal waveforms

If $\varepsilon(k)$ is the instantaneous error, then MSE $\xi = E[|\epsilon(k)|^2]$

The MMSE problem can be defined $\mathbf{w}_{MMSE} = \arg\min_{\mathbf{w}} \xi(\mathbf{w})$

Its solution is the minimum of the MSE cost function (right), the Wiener solution

$$\mathbf{w}_{MMSE}(k) = \left(\mathbf{P}(k)\mathbf{P}^{H}(k) + \frac{2\sigma_{n}^{2}}{\sigma_{b}^{2}}\mathbf{I}_{L}\right)^{-1}\mathbf{p}_{1}(k)$$



LMS Algorithm



LMS – Least Mean Squares

LMS is a stochastic gradient adaptive algorithm that converges towards the solution on each update

$$\mathbf{w}(k+1) = \mathbf{w}(k) + \mu \{b_1(k) - y(k)\}^* \mathbf{x}(k)$$

 μ – the adaption constant

MSER Beamforming

- MSER Minimum Symbol-Error-Rate
- Minimise the probability of incorrectly decoding a signal
- The probability of an error occurring at a particular time instant *k* is

 $P_{E_B}(\mathbf{w}) = P\{\hat{b}_1(k) \neq b_1(k)\}$

The MSER problem is therefore defined as

 $\mathbf{w}_{MSER} = \arg\min_{\mathbf{w}} P_{E_B}(\mathbf{w})$

- Its solution is the minimum of the SER cost function (right).
- There is no closed form solution

An iterative conjugate-gradient algorithm can be use to find the solution



Adaptive Beamforming: LSER



Simulation Study



- Time-varying channel (correlated Rayleigh fading over 250 fades)
- Frame structure



Results: SER Performance



- Min. angular separation: $\theta = 27^{\circ}$
- Normalised Doppler frequency

 $f_D = 10^{-4}$ and 10^{-3}

- Modulation: 64-QAM
- LMS: *μ* = 0.0002

• LSER:
$$\mu = 0.00005$$
, $\rho_n = 4\sigma_n$

Results: Parameter Tuning



• Min. angular separation: $\theta = 44^{\circ}$

- Modulation: 64-QAM
- Normalised Doppler frequency $f_D = 10^{-4}$

• LMS: $\mu = 0.0002$; LSER: $\mu = 0.00005$, $\rho_n = 4\sigma_n$

Results: Averaged SER Performance



- Min. angular separation: θ
 averaged over [20°, 50°]
- Normalised Doppler frequency

 $f_D = 10^{-3}$

- Modulation: 64-QAM
- LMS: *μ* = 0.0002

• LSER:
$$\mu = 0.00005$$
, $\rho_n = 4\sigma_n$

Conclusions

- LSER is an adaptive implementation of the MSER beamforming solution
- LSER algorithm can operate successfully in
 - Fast fading conditions
 - With bandwidth-efficient M-QAM modulation
 - An SDMA environment with more users than antenna elements
- LSER algorithm consistently outperforms the adaptive LMS algorithm bench marker
- Benefits
 - Higher network capacity
 - Higher data rates
 - Longer range
 - Lower transmit power