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- ▶ **Sparse Channel Estimation and Equalization for OFDM-Based Underwater Cooperative Systems With Amplify-and-Forward Relaying**

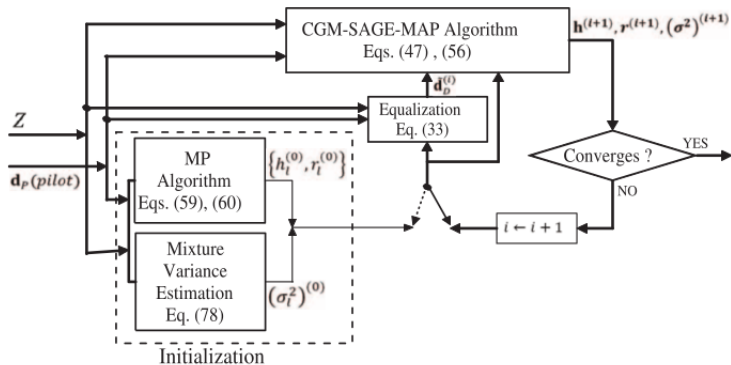
Authors: E. Panayirci, H. Senol, M. Uysal, and H. V. Poor

Cascaded system model: $z = \mathbf{A}h + \mathbf{V}$

Goal: Estimate unknown parameter $\Phi = \{\mathbf{h}, \mathbf{r}, \mathbf{v}\}$

Algo:

- ▶ update $\Phi_l = \{h_l, r_l\}$ for $l = 1, 2, \dots, L$ using CGM-MP-SAGE, keep $\Phi_{L+1} = \{\mathbf{v}\}$ fixed
- ▶ update Φ_{L+1} using SAGE, keep Φ_l fixed



CGM-MP-SAGE:

E-step:

$$\begin{aligned} Q_l(\Phi_l | \Phi^{(i)}) &= \mathbb{E}\{\log p(\chi_l, \Phi_l, \bar{\Phi}_l^{(i)}) | \mathbf{Z}, \Phi^{(i)}\} \\ &\sim \mathbb{E}\{\log p(\mathbf{X}_l | h_l, r_l, \mathbf{v}, \bar{\Phi}_l^{(i)}) | \mathbf{Z}, \bar{\Phi}^{(i)}\} \\ &\quad + \mathbb{E}\{\log p(h_l | v_l) | \mathbf{Z}, \bar{\Phi}^{(i)}\} \end{aligned}$$

M-step:

$$\Phi_l^{(i+1)} = \arg \max_{\Phi_l} Q_l(\Phi_l | \Phi^{(i)})$$

▶ **A Minorization-Maximization Method for Optimizing Sum Rate in the Downlink of Non-Orthogonal Multiple Access Systems**

Authors: M. F. Hanif, Z. Ding, T. Ratnarajah, and G. K. Karagiannidis

Optimization problem:

$$R_{sum} = \sum_{k=1}^{N-1} \log_2 \left(1 + \min(\text{SINR}_k^k, \dots, \text{SINR}_N^k) \right) + \log_2 \left(1 + \frac{|\mathbf{h}_N^H \mathbf{w}_N|^2}{\sigma^2} \right)$$

maximise R_{sum}
 \mathbf{w}_i

$$\text{s.t. } |\mathbf{h}_k^H \mathbf{w}_1|^2 \geq \dots \geq |\mathbf{h}_k^H \mathbf{w}_{k-1}|^2 \geq |\mathbf{h}_k^H \mathbf{w}_k|^2 \geq$$

$$|\mathbf{h}_k^H \mathbf{w}_{k+1}|^2 \dots \geq |\mathbf{h}_k^H \mathbf{w}_N|^2,$$

$$1 \leq k \leq N$$

$$\sum_{i=1}^N \|\mathbf{w}_i\|_2^2 \leq P_{th}$$

Proposed iterative solution:

given randomly generated Λ^0 feasible to original opt-prob.,

$t := 0$

repeat: 1. Solve equivalent form of original opt-prob.

2. set $\Lambda^{t+1} = \Lambda^t$

3. update $t := t+1$

until converges

where, $\Lambda^t = [w_k^t, r_k^t, v_j^t, \theta_{k,k}^t, \theta_{j,k}^t, \theta_{N,N}^t, \Phi_{k,m}^t]$

▶ **Parallel Two-Way Relaying in Cooperative OFDMA Cellular Systems**

Authors: S. Jin and L. Ping

Goal: Maximize the throughput

$$\bar{R}_{PTWR-OFDMA} = \beta \bar{R}_{direct} + (1 - \beta) \bar{R}_{indirect}$$

Under power constraints

$$\begin{aligned} \max : \bar{R}_{indirect} &= \frac{1}{2} \cdot (\log_2(1 + \gamma_{BS}) + \log_2(1 + \gamma_{FN})) \\ \text{s.t.} : P_{BS} + P_{FN} + \sum_{k=1}^{K'} P_{NN k} &\leq P_{sum} \end{aligned}$$

Maximum sum rate is achieved when: $\gamma_{BS} = \gamma_{FN}$

and for the optimal solution $P_{sum, NNs} = \sum_{k=1}^{K'} P_{NN k} = \frac{1}{2} P_{sum}$

Asymptotically Optimal β and Average Achievable Rate:

$$\beta^{opt} \approx \beta^{app} \triangleq \frac{1}{\ln K}, \text{ and}$$

$$\bar{R}_{PTWR-OFDMA} = (\ln K - 1) / \ln 2 + o(\ln K)$$

Average assisting power per NN, $\bar{P}_{assisting} = (\ln K - 1)P$

Conclusion: Throughput increases in the order of $\ln K$ compared to $\ln \ln K$ in conventional OFDMA

Some other papers:

- ▶ Learning in an Uncertain World: MIMO Covariance Matrix Optimization With Imperfect Feedback
-P. Mertikopoulos and A. L. Moustakas
- ▶ Outlier-Robust Greedy Pursuit Algorithms in l_p -Space for Sparse Approximation
-W.-J. Zeng, H. C. So, and X. Jiang
- ▶ An Improved Algebraic Solution for Moving Target Localization in Noncoherent MIMO Radar Systems
-H. Yang and J. Chun