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- Likelihood-Based Automatic Modulation Classification in OFDM With Index Modulation
- Joint Power Control and Beamforming for Uplink Non-Orthogonal Multiple Access in 5G Millimeter-Wave Communications
- Subchannel Allocation and Hybrid Precoding in Millimeter-Wave OFDMA Systems

Likelihood-Based Automatic Modulation Classification(AMC) in OFDM With Index Modulation (J.Zheng and Y.Lv)

- Contributions- ALRT-based classifier, HLRT-based classifier, low-complexity HLRT-LLR, HLRT-Energy classifier and a blind AMC
- N_c Subcarriers in one OFDM symbols are divided into G blocks with N subcarriers, i.e., N_c=NG
- $I_g^t = \{ i_{g,1}^t, \dots, i_{g,K}^t \}$ denote the set of indices of K active subcarriers
- $S_g^t = \{ s_{g,1}^t, \dots, s_{g,K}^t \}$ corresponds to the set of symbols transmitted over the active subcarriers

- x^t_g = x^t_g(I^t_g, S^t_g) = ∑^K_{k=1} s^t_{g,k} e^t_{ig,k} denote the frequency domain(FD) signal in gth block
 y^t_g = N√(G/K)X^t_gh_g + w^t_g, g = 1, 2, ..., G ,received FD signal ,where h^t_{g,j}, w^t_{g,j} ~ CN(0, σ²_f), j = 1, 2, ..., N
- ALRT-bsed AMC: hypothesis $\mathcal{H}_m = (K, M)$ is accepted if it achieves largest posterior probability

$$p(K, M|y, \theta) = \frac{p(y|K, M, \theta)p(K, M)}{p(y|\theta)}$$
$$\tilde{\mathcal{H}} = \underset{\mathcal{H} = (K, M) \in K \times M}{\operatorname{argmax}} \ln p(y|M, K, \theta)$$

• **Theorem 1**: In ALRT-based AMC, the classifier accepts modulation parameter hypothesis according to

$$\hat{\mathcal{H}} = \underset{\mathcal{H} = (K,M) \in K \times M}{\operatorname{argmax}} \Gamma_{ALRT}(K,M)$$

where
$$\Gamma_{ALRT}(K, M) = -TGIn(C(N, K)M^K) +$$

$$\sum_{t=1}^{T} \sum_{g=1}^{G} \ln \sum_{I_g^t} (exp(-\frac{\sum_{\alpha \in \tilde{I}_g^t} |y_{g,\alpha}^t|^2}{\sigma_f^2}) x$$

$$\prod_{\alpha \in I_g^t} \sum_{X_{g,\alpha}^t \in \mathcal{X}_M} \exp(-\frac{|y_{g,\alpha}^t - N\sqrt{G/K} x_{g,\alpha}^t h_{g,\alpha}|^2}{\sigma_f^2}))$$

- Complexity of processing the hypothesis is O(TGC(N,K)(N-K+KM)) in ALRT-based AMC
- HLRT-based AMC: To reduce the complexity, first I^t_g is estimated and then S^t_g is averaged out from x^t_g

• Corollary 1: HLRT-based AMC, the classifier accepts the modulation parameter as

$$\hat{\mathcal{H}} = \underset{\mathcal{H} = (K,M) \in K \times M}{\operatorname{argmax}} \Gamma_{HLRT}(K,M)$$

where
$$\Gamma_{HLRT}(K, M) = -TGln(M^{K}) +$$

 $\sum_{t=1}^{T} \sum_{g=1}^{G} \left(\left(-\frac{\sum_{\alpha \in \tilde{l}_{g}^{t}} |y_{g,\alpha}^{t}|^{2}}{\sigma_{f}^{2}} \right) + \sum_{\alpha \in \hat{l}_{g}^{t}} ln \sum_{x_{g,\alpha}^{t} \in \mathcal{X}_{M}} exp\left(-\frac{|y_{g,\alpha}^{t} - N\sqrt{G/K}x_{g,\alpha}^{t}h_{g,\alpha}|^{2}}{\sigma_{f}^{2}} \right) \right)$

 Complexity of processing the hypothesis is O(TG(N-K+KM+N(M+1))) in HLRT-based AMC Joint Power Control and Beamforming for Uplink Non-Orthogonal Multiple Access in 5G Millimeter-Wave Communications (L.Zhu , J.Zhang , Z.Xiao)

- Contributions- (Suboptimal solution) Decomposes the original problem into two sub-problems: one is a power control and beam gain allocation problem, and the other is a beamforming problem
- Uplink Scenario is used where BS is equipped with a N-element antenna array
- $y = h_1^H w \sqrt{p_1} s_1 + h_2^H w \sqrt{p_2} s_2 + n^H w$, 2-user NOMA model where w denotes a beamforming vector with $|[w]_k| = \frac{1}{\sqrt{N}}$ for $k = 1, 2, \dots, N$ and $h_1 \ge h_2$

- h_i = λ_ia(N, Ω_i), channel between user i and BS is an mm-wave channel, a(.) is a steering vector
- **Decoding** Case 1: s₁ is decoded first

$$R_1^{(1)} = \log_2(1 + \frac{|h_1^H w|^2 p_1}{|h_2^H w|^2 p_2 + \sigma^2}), R_2^{(1)} = \log_2(1 + \frac{|h_2^H w|^2 p_2}{\sigma^2})$$

Case 2: s_2 is decoded first

$$R_1^{(2)} = \log_2(1 + rac{|h_1^Hw|^2 p_1}{\sigma^2}), \ R_2^{(2)} = \log_2(1 + rac{|h_2^Hw|^2 p_2}{|h_1^Hw|^2 p_1 + \sigma^2})$$

• $R_1 + R_2 = log_2(1 + \frac{|h_1^H w|^2 p_1 + |h_2^H w|^2 p_2}{\sigma^2})$, sum rate of different decoding orders are identical

• Problem:

 $\begin{array}{l} \underset{p_1,p_2,w}{\textit{Maximize }} R_1 + R_2 \\ \text{Subject to } R_1 \geq r_1, R_2 \geq r_2 \\ 0 \leq p_1, p_2 \leq P \\ |[w]_k| = \frac{1}{\sqrt{(N)}} \end{array}$

- **Solution**: Problem is non-convex, proposed solution is sub-optimal(decomposes the original problem into two sub-problems)
- Solution of Power Control and Beam gain allocation Sub-Problem:

Lemma 1: With the ideal beamforming, the beam gains satisfy

$$\frac{c_1}{|\lambda_1|^2}+\frac{c_2}{|\lambda_2|^2}=N$$
 where $c_1=|h_1^Hw|^2$ and $c_2=|h_2^Hw|^2$

Lemma 2: With the ideal beamforming, the optimal transmission power is

$$p_1^* = P, p_2^* = P$$

• From Lemma 1 and Lemma 2, the optimal values of $|h_2^H w|^2$ and $|h_1^H w|^2$ are

$$c_2^* = rac{(2'^2-1)\sigma^2}{P}, \ c_1^* = |\lambda_1|^2 (N - rac{c_2^*}{|\lambda_2|^2})$$

Decoding s_1 first is optimal

• Solution of the Beamforming Sub-Problem:

$$\begin{array}{l} & \underset{w}{\textit{Maximize}} \ |h_1^H w|^2 \\ & \text{Subject to } \ |h_2^H w|^2 \geq c_2^* \\ & |[w]_k| = \frac{1}{\sqrt{(N)}}, k = 1, 2, \dots, N \end{array}$$

 Above problem is non-convex, after relaxing contraints we get, Maximize $a_1^H w$ Subject to $|a_2^H w| > g$ $|[w]_k| \leq \frac{1}{\sqrt{(N)}}, k = 1, 2, \dots, N$ where $g = \sqrt{\frac{c_2^*}{|\lambda_2|^2}}$, $a_i \equiv a(N, \Omega_i)$ Still it is not convex, split it into a serial of convex optimization problems , i.e., assume different phases for $a_2^H w$ Maximize $a_1^H w$ Subject to $Re(a_2^H w e^{2\pi j \frac{m}{M}}) \geq g$ $|[w]_k| \leq \frac{1}{\sqrt{(N)}}, k = 1, 2, \dots, N$ where m = 1, 2, ..., M and M is the number of candidate phases

Subchannel Allocation and Hybrid Precoding in Millimeter-Wave OFDMA Systems (V.Ha, D.Nguyen, J.Frigon)

- Contributions- Maximizes the sum rate under a constraint on the number of data streams
- System model- downlink mm-wave MU-OFDMA HB(Hybrid beamforming) system with a BS with N_T antennas and N_{RF} RF chains serves K remote single-antenna UE over S subcarriers
- SINR at user k over subcarrier s ,

$$SINR_{k,s} = \frac{a_{k,s} |h_{k,s}^{H} A w_{k,s}|^{2}}{\sum_{j \neq k} a_{j,s} |h_{k,s}^{H} A w_{j,s}|^{2} + \sigma^{2}}$$

where

$$a_{k,s} = 1$$
 if user k is assigned subcarrier s
0 ,else

 $w_{k,s}$, digital precoding vector applied to data symbols $x_{k,s}$ $\bf A$ denotes analog precoding matrix to map RF signals to RF chains

• Problem-

$$\max_{a_{k,s},w_{k,s},A} \sum_{k,s} \log(1 + SINR_{k,s})$$
(1)



$$A\epsilon A_R$$
 (2)

$$\sum_{k,s} |a_{k,s}|^2 w_{k,s}^H A^H A w_{k,s} \le P_T$$
(3)

$$\sum_{k,s} a_{k,s} \le \bar{D} \tag{4}$$

where \bar{D} denotes highest number of concurrent data streams which system can support in one transmission slot

• Solution- let

$$\widetilde{w}_{k,s} = a_{k,s} w_{k,s}$$

 $\widetilde{p}_{k,s} = \widetilde{w}_{k,s}^H \widetilde{w}_{k,s} = |a_{k,s}|^2 w_{k,s}^H w_{k,s}$

• Inequality (4) can be transformed as

$$||\tilde{\rho}||_0 \le \bar{D} \tag{5}$$

• Problem-

$$\max_{\tilde{w}_{k,s},A} \sum_{k,s} \log(1 + SINR_{k,s})$$

s.t. (2),(3),(5)

• Sparse HP Design:

Stage one- max-sum-rate Optimization:

$$\begin{split} \max_{u_{k,s}} \sum_{k,s} \log(1 + \frac{|h_{k,s}^{H}u_{k,s}|^{2}}{\sum_{j \neq s} |h_{k,s}^{H}u_{j,s}|^{2} + \sigma^{2}}) \\ \text{s.t. } \sum_{k,s} u_{k,s}^{H}u_{k,s} \leq P_{T} \end{split}$$

where $u_{k,s} = A\tilde{w}_{k,s}$ **Stage two**- Sparse HP Design: Let $u_{k,s}^{opt}$ be the outcome of above problem

$$\min_{\tilde{w}_{k,s},A} \sum_{k,s} ||u_{k,s}^{opt} - A\tilde{w}_{k,s}||_{2}^{2}$$

$$s.t.(2), (3), (5)$$

 Results show that the joint SA and HP design achieves better sum-rate while requiring the lowest computational complexity