Concurrent Constant Modulus Algorithm and Soft Decision Directed Scheme for Fractionally-Spaced Blind Equalization

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Constant modulus algorithm aided soft decision directed scheme for low complexity blind equalization of high-order QAM channels

 \bigcirc Existing works:

Constant modulus algorithm, concurrent CMA and decision directed scheme (De Castro et al, ICC2001)

○ Concurrent CMA and soft decision directed scheme:

Simpler operational requirements, faster convergence and similar steady-state equalization performance, compared with CMA+DD scheme

 \bigcirc Comparative simulation results



Constant Modulus Algorithm

CMA is a popular choice for blind equalization of high-order QAM channels

 \Uparrow Very simple, need to know little e.g. robust to carrier recovery error, capable of opening initially closed eye

 \Downarrow Steady-state performance may not be sufficiently accurate to achieve adequate BER

○ A solution is to switch to (hard) decision directed adaptation after convergence

 \bigcirc When to switch ? and can it be switched to ?



De Castro *et al* (2001) split equalizer into a CMA sub-equalizer and a DD sub-equalizer: $\mathbf{w} = \mathbf{w}_c + \mathbf{w}_d$

1. CMA adaptation: with $y(k) = \mathbf{w}_c^T(k)\mathbf{r}(k) + \mathbf{w}_d^T(k)\mathbf{r}(k)$

$$\left. \begin{array}{l} \epsilon(k) = y(k)(\Delta_2 - |y(k)|^2) \\ \mathbf{w}_c(k+1) = \mathbf{w}_c(k) + \mu_c \epsilon(k) \mathbf{r}^*(k) \end{array} \right\}$$

2. DD adaptation: with $\tilde{y}(k) = \mathbf{w}_c^T(k+1)\mathbf{r}(k) + \mathbf{w}_d^T(k)\mathbf{r}(k)$

$$\mathbf{w}_d(k+1) = \mathbf{w}_d(k) + \mu_d \delta(\mathcal{Q}[\tilde{y}(k)] - \mathcal{Q}[y(k)])(\mathcal{Q}[y(k)] - y(k))\mathbf{r}^*(k)$$

where $\mathcal{Q}[\bullet]$ denotes quantization operation or hard decision, $\delta(x) = 1$ if x = 0 + j0 and $\delta(x) = 0$ if $x \neq 0 + j0$

 \bigcirc \mathbf{w}_d is updated only if hard decisions before and after CMA adaptation are the same, to reduce error propagation due to incorrect decisions



Motivation for Soft DD

After equalization is accomplished, the *a posteriori* PDF of y(k) is approximately:

$$p(\mathbf{w}, y(k)) \approx \sum_{q=1}^{Q} \sum_{l=1}^{Q} \frac{p_{ql}}{2\pi\rho} \exp\left(-\frac{|y(k) - s_{ql}|^2}{2\rho}\right)$$

where s_{ql} are constellation points of Q^2 -QAM, p_{ql} are the *a priori* probabilities of s_{ql} , and ρ is variance of y(k)

 \bigcirc Divide complex plane into $Q^2/4$ regular regions, each region $S_{i,l}$ contains four symbol points

If y(k) is within $S_{i,l}$, a local approximation to the *a posteriori* PDF of y(k) is





Motivation (continue)

 \bigcirc SDD equalizer is designed to maximize log of the local *a posteriori* PDF

 $\bar{J}_{\text{LMAP}}(\mathbf{w}) = \mathsf{E}[J_{\text{LMAP}}(\mathbf{w}, y(k))]$

by adjusting \mathbf{w}_d using a stochastic gradient algorithm, where

$$J_{\text{LMAP}}(\mathbf{w}, y(k)) = \rho \log \left(\hat{p}(\mathbf{w}, y(k)) \right)$$

 \bigcirc Stochastic gradient of $J_{ ext{LMAP}}(\mathbf{w},y(k))$ is

$$\frac{\partial J_{\text{LMAP}}(\mathbf{w}, y(k))}{\partial \mathbf{w}_d} = \frac{\sum_{p=2i-1}^{2i} \sum_{q=2l-1}^{2l} \exp\left(-\frac{|y(k)-s_{pq}|^2}{2\rho}\right) (s_{pq} - y(k))}{\sum_{p=2i-1}^{2i} \sum_{q=2l-1}^{2l} \exp\left(-\frac{|y(k)-s_{pq}|^2}{2\rho}\right)} \mathbf{r}^*(k)$$

 $\bigcirc
ho$ is typically less than half the distance between neighbouring symbol points

 \bigcirc Rather than committed to a single hard decision $\mathcal{Q}[y(k)]$, tentative decisions are considered in a local region $S_{i,l}$ that includes $\mathcal{Q}[y(k)]$

CMA and SDD

○ Operations of CMA and SDD sub-equalizers are truly concurrent:

• With
$$y(k) = \mathbf{w}_c^T(k)\mathbf{r}(k) + \mathbf{w}_d^T(k)\mathbf{r}(k)$$

CMA:
$$\begin{cases} \epsilon(k) = y(k)(\Delta_2 - |y(k)|^2) \\ \mathbf{w}_c(k+1) = \mathbf{w}_c(k) + \mu_c \epsilon(k) \mathbf{r}^*(k) \end{cases}$$

SDD:
$$\mathbf{w}_d(k+1) = \mathbf{w}_d(k) + \mu_d \frac{\partial J_{\text{LMAP}}(\mathbf{w}(k), y(k))}{\partial \mathbf{w}_d}$$

 \bigcirc Computational complexity is less than CMA+DD:

equalizer	multiplications	additions	$\exp(ullet)$
CMA	$8 \times m_L + 6$	$8 imes m_L$	—
CMA+DD	$16 \times m_L + 8$	$20 imes m_L$	—
CMA+SDD	$12 \times m_L + 29$	$14 \times m_L + 21$	4

where m_L is equalizer length



Simulation (Fixed SISO Channel)

 $\bigcirc T_s/2$ -spaced 22-tap channel and 26 tap equalizer, where T_s is symbol period, 256-QAM \bigcirc Let $\{f_i\}_{i=0}^{N_f-1}$ be combined impulse response of channel and equalizer. Maximum distortion is defined by

$$MD = \frac{\sum_{i=0}^{N_f - 1} |f_i| - |f_{i_{\max}}|}{|f_{i_{\max}}|}$$





Signal Constellation (Fixed SISO Channel)

Equalizer output signal constellations after convergence (a) CMA, (b) CMA+DD, and (c) CMA+SDD





64-QAM Fading SISO Channel (CMA)

CMA equalizer output signal constellations: (a) after adaptation of 20000 symbols, (b) after adaptation of 25000 symbols, and (c) after adaptation of 30000 symbols. 6000 T_s -spaced samples were used in showing signal constellation with continuous adaptation





64-QAM Fading SISO Channel (CMA+DD)

CMA+DD equalizer output signal constellations: (a) after adaptation of 15000 symbols, (b) after adaptation of 20000 symbols, and (c) after adaptation of 25000 symbols. 6000 T_s -spaced samples were used in showing signal constellation with continuous adaptation





64-QAM Fading SISO Channel (CMA+SDD)

CMA+SDD equalizer output signal constellations: (a) after adaptation of 15000 symbols, (b) after adaptation of 20000 symbols, and (c) after adaptation of 25000 symbols. 6000 T_s -spaced samples were used in showing signal constellation with continuous adaptation





256-QAM Fixed SIMO Channel

 \bigcirc Single transmit antenna and four receive antennas

 $\bigcirc T_s/2$ -spaced channels and $T_s/2$ -spaced space-time equalizer





Space-time equalizer output signal constellations after convergence: (a) CMA, (b) CMA+DD, and (c) CMA+SDD for fixed SIMO channel with 256-QAM and SNR of 40 dB





Conclusions

- A constant modulus algorithm aided soft decision directed scheme has been derived for low complexity blind equalization of high-order QAM channels
- Compared with an existing CMA and decision directed scheme, the proposed blind equalization scheme has simpler operational requirements and faster convergence, and achieves similar steady-state equalization performance
- Original derivation is for SISO systems, but the scheme can be extended to blind space-time equalization of SIMO and MIMO systems



