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ENHANCED SINGLE ANTENNA INTERFERENCE CANCELLATION FROM MMSE THIRD-ORDER COMPLEX VOLTERRA FILTERS

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ABSTRACT

Widely linear (WL) filters have the capability to perform single antenna interference cancellation (SAIC) of one rectilinear or quasi-rectilinear co-channel interference (CCI). The SAIC technology for quasi-rectilinear signals is operational in GSM handsets but requires enhancements for both VAMOS standard, an evolution of GSM/EDGE standard, and FBMC-OQAM networks, which are candidate for 5G mobile networks, in particular. For this reason, we propose in this paper a SAIC enhancement based on the use of third-order complex Volterra (CV) filtering, exploiting both the non-Gaussianity and the non-circularity of the signals up to the 6th-order. Limiting the analysis to rectilinear signals for space limitations, the performance of the proposed receiver are proved to outperform those of WL receivers for SAIC of one CCI.

Index Terms— Widely linear, SAIC, non-circular, non-Gaussian, complex Volterra filter, CCI, MMSE.

1. INTRODUCTION

It is now well-known that one of the most important properties of WL filtering [1] is its capability to perform SAIC of one rectilinear (BPSK, ASK) or quasi-rectilinear (MSK, GMSK, OQAM) multi-user CCI, allowing the separation of two users from only one receive antenna [2–4]. The effectiveness of this concept jointly with its low complexity explain why it is operational in most of GSM handsets, allowing significant network’s capacity gains for the GSM system [4, 5]. However, as explained in [6–8], SAIC enhancements are required for new technologies such as voice services over adaptive multi-user channels on one slot (VAMOS) technology, an evolution of GSM standardized recently [9]. A similar need is also required to mitigate both inter-carrier interference (ICI) and CCI for networks which will use filter bank multi-carrier (FBMC) waveforms coupled with OQAM modulation, which are considered as promising candidates for the 5G mobile networks in particular [10].

In this context, the purpose of this paper is to propose and to analyze the performance of an enhanced SAIC receiver

based on third-order CV filtering instead of WL filtering. While WL receivers exploit the second-order non-circularity of the signals, third-order CV receivers exploit both the non-Gaussianity and the non-circularity of the signals up to the 6th-order [11], hence a performance gain. To introduce this new receiver, we adopt an MMSE approach and due to space limitations, we restrict the analysis to rectilinear signals and channels with no delay spread. Under these assumptions, it is shown in particular that the performance gain obtained in using the third-order CV receiver instead of the WL one depends on several parameters such as the constellation of the symbols and differential phases and delays between the signal of interest (SOI) and the CCI.

Note that the scarce papers dealing with CV filters for CCI mitigation correspond to [12–15]. While [12] deals with general M th-order CV filters for detection and estimation in non-Gaussian total noise with an application to CCI mitigation for $M = 2$, [13–15] concern third-order CV filters for CCI mitigation. Nevertheless [13] is limited to CDMA communications networks, whereas [14] and [15] consider a beamforming application requiring at least two receive antennas.

2. MODEL, STATISTICS AND PROBLEM FORMULATION

2.1. Signal model and statistics

We consider a narrow-band antenna receiving the contribution of a rectilinear SOI corrupted by a rectilinear CCI and a background noise. Assuming propagation channels with no delay spread and perfect time and frequency synchronization of the SOI, the complex envelope of the signals at the output of the antenna can then be written as

$$x(t) = \mu_s e^{i\phi_s} \sum_k a_k v(t-kT) + \mu_j e^{i\phi_j} \sum_k b_k v(t-kT-\tau_j) + n(t), \quad (1)$$

where a_k and b_k are real-valued zero-mean i.i.d. random variables corresponding to the symbols of the SOI and CCI respectively, independent to each other and with the zero-mean Gaussian background noise contribution $n(t)$. T is the symbol period, $v(t)$ is the impulse response of the pulse shaping

filter of both the SOI and CCI, μ_s and μ_j control the amplitude of the SOI and the CCI respectively, ϕ_s and ϕ_j are the phase of the SOI and CCI respectively and $\tau_j \in [0, T]$ is the delay of the CCI w.r.t. the SOI. Assuming that $v(t)$ is a raised cosine 1/2 Nyquist filter and denoting by x_k the sampled observation, at the symbol rate, at time kT at the output of a matched filtering operation to the pulse shaping filter, we obtain:

$$x_k = \mu_s e^{i\phi_s} a_k r(0) + \mu_j e^{i\phi_j} \sum_{\ell} b_{\ell} r((k-\ell)T - \tau_j) + n_k, \quad (2)$$

where $r(t) \stackrel{\text{def}}{=} v(t) \otimes v^*(-t)$ is the real-valued impulse response of a Nyquist filter where \otimes is the convolution operation, whereas n_k is the background noise contribution at the output of the matched filter, which is zero-mean, stationary, Gaussian and circular with power η_2 .

We denote by a'_k and $b'_k(\tau_j)$ the normalized symbol a_k and quantity $\sum_{\ell} b_{\ell} r((k-\ell)T - \tau_j)$, respectively such that $E(a'_k{}^2) = E(b'_k{}^2(\tau_j)) = 1$. Under these assumptions, (2) takes the form

$$x_k = \sqrt{\pi_s} e^{i\phi_s} a'_k + \sqrt{\pi_j} e^{i\phi_j} b'_k(\tau_j) + n_k, \quad (3)$$

where $\pi_s \stackrel{\text{def}}{=} \mu_s^2 \pi_a r^2(0)$ and $\pi_j \stackrel{\text{def}}{=} \mu_j^2 \pi_b (\sum_n r^2((k-n)T - \tau_j))$ (with $\pi_a \stackrel{\text{def}}{=} E(a_k^2)$ and $\pi_b \stackrel{\text{def}}{=} E(b_k^2)$) are the power of the SOI and CCI symbols, respectively. We will need in Section 4, the fourth and sixth-order moments of $b'_k(\tau_j)$ given by

$$\kappa_j \stackrel{\text{def}}{=} E(b'_k{}^4(\tau_j)) = E(b_k^4) \sum_{\ell} r_{\ell}^4 + 6[E(b_k^2)]^2 \sum_{k < \ell} r_k^2 r_{\ell}^2 \quad (4)$$

$$\begin{aligned} \chi_j \stackrel{\text{def}}{=} E(b'_k{}^6(\tau_j)) &= E(b_k^6) \sum_{\ell} r_{\ell}^6 + 15E(b_k^2)E(b_k^4) \sum_{k \neq \ell} r_k^4 r_{\ell}^2 \\ &+ 90[E(b_k^2)]^3 \sum_{i < k < \ell} r_i^2 r_k^2 r_{\ell}^2. \end{aligned} \quad (5)$$

where $r_k \stackrel{\text{def}}{=} r(kT - \tau_j) / \sqrt{\pi_b \sum_{\ell} r^2(\ell T - \tau_j)}$.

2.2. Problem formulation

The problem addressed in this paper is to detect the symbols a_k from the observations x_k through an MMSE approach. Naturally, the best estimate y_k of a_k according to the MMSE criterion is the conditional expectation $y_k = E(a_k | x_k)$. Note that for respectively circular or non-circular mutually Gaussian distributions of (a_k, x_k) , this conditional expectation becomes linear or widely linear [1] respectively. But for non-Gaussian distribution of (a_k, x_k) , the derivation of this conditional expectation becomes generally non-linear in x_k and needs this distribution, which is unknown in practice.

For this reason, we consider in this paper an approximation of this conditional expectation through the analysis of a particular class of non-linear filters corresponding to the CV filters, introduced for the first time in [12] and [16] in the context of detection and estimation. The general model of a memoryless full M th-order time invariant CV filter is defined by

$$y_k = \sum_{m=0}^M \sum_{q=0}^m w_{m,q}^* x_k^{m-q} x_k^{*q}. \quad (6)$$

For $w_{0,0} = 0$, rel. (6) defines a widely linear filter [1] for $M = 1$ and a full complex linear-quadratic filter [17] for $M = 2$. Rel. (6) can be compactly written in the form

$$y_k = \tilde{\mathbf{w}}^H \tilde{\mathbf{x}}_k, \quad (7)$$

where $\tilde{\mathbf{w}} \stackrel{\text{def}}{=} (w_{0,0}, w_{1,0}, w_{1,1}, w_{2,0}, w_{2,1}, w_{2,2}, \dots, w_{M,M})^T$ and $\tilde{\mathbf{x}}_k \stackrel{\text{def}}{=} (1, x_k, x_k^*, x_k^2, x_k x_k^*, x_k^{*2}, \dots, x_k^{*M})^T$ is the augmented observation. The problem of the optimal M th-order CV filter is then to find $\tilde{\mathbf{w}}$ minimizing the MSE between y_k and a_k .

3. THIRD-ORDER COMPLEX VOLTERRA MMSE RECEIVER

3.1. M th-order complex Volterra MMSE filter

The full M th-order CV MMSE filter corresponds to the filter $\tilde{\mathbf{w}}$ which minimizes the criterion $\text{MSE}(\tilde{\mathbf{w}}) = E|a_k - \tilde{\mathbf{w}}^H \tilde{\mathbf{x}}_k|^2$. This filter is classically given by

$$\tilde{\mathbf{w}}_{\text{CV-MMSE}} = \mathbf{R}_{\tilde{\mathbf{x}}}^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a}, \quad (8)$$

with $\mathbf{R}_{\tilde{\mathbf{x}}} \stackrel{\text{def}}{=} E(\tilde{\mathbf{x}}_k \tilde{\mathbf{x}}_k^H)$ and $\mathbf{r}_{\tilde{\mathbf{x}},a} \stackrel{\text{def}}{=} E(\tilde{\mathbf{x}}_k a_k^*)$. The MSE obtained with the full M th-order CV MMSE filter (8) is given by

$$\text{MMSE} \stackrel{\text{def}}{=} \text{MSE}[\tilde{\mathbf{w}}_{\text{CV-MMSE}}] = \pi_a - \mathbf{r}_{\tilde{\mathbf{x}},a}^H \mathbf{R}_{\tilde{\mathbf{x}}}^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a}. \quad (9)$$

If some components of the full M th-order CV MMSE filter (6) are withdrawn, we obtain a partial M th-order CV MMSE filter. The increase $\Delta_{\text{CV-MMSE}}$ of MMSE obtained by such a partial M th-order CV filter can be derived by partitioning $\tilde{\mathbf{x}}_k$ into the retained, $\tilde{\mathbf{x}}_{1,k}$, and the discarded, $\tilde{\mathbf{x}}_{2,k}$, part. Applying the matrix inversion lemma to the partitioned augmented covariance matrix $\mathbf{R}_{\tilde{\mathbf{x}}}$ written as

$$\mathbf{R}_{\tilde{\mathbf{x}}} = \begin{bmatrix} \mathbf{R}_{\tilde{\mathbf{x}}_{11}} & \mathbf{R}_{\tilde{\mathbf{x}}_{12}} \\ \mathbf{R}_{\tilde{\mathbf{x}}_{12}}^H & \mathbf{R}_{\tilde{\mathbf{x}}_{22}} \end{bmatrix},$$

the increase of MMSE given by the partial M th-order CV MMSE filter that only uses $\tilde{\mathbf{x}}_{1,k}$ is given by

$$\begin{aligned} \Delta_{\text{CV-MMSE}} &= (\mathbf{r}_{\tilde{\mathbf{x}}_{2,a}}^H - \mathbf{r}_{\tilde{\mathbf{x}}_{1,a}}^H \mathbf{R}_{\tilde{\mathbf{x}}_{11}}^{-1} \mathbf{R}_{\tilde{\mathbf{x}}_{12}}) \\ &(\mathbf{R}_{\tilde{\mathbf{x}}_{22}} - \mathbf{R}_{\tilde{\mathbf{x}}_{12}}^H \mathbf{R}_{\tilde{\mathbf{x}}_{11}}^{-1} \mathbf{R}_{\tilde{\mathbf{x}}_{12}})^{-1} (\mathbf{r}_{\tilde{\mathbf{x}}_{2,a}} - \mathbf{R}_{\tilde{\mathbf{x}}_{12}}^H \mathbf{R}_{\tilde{\mathbf{x}}_{11}}^{-1} \mathbf{r}_{\tilde{\mathbf{x}}_{1,a}}) \geq 0. \end{aligned} \quad (10)$$

Consequently, the term $\tilde{\mathbf{x}}_{2,k}$ does not bring any information ($\Delta_{\text{CV-MMSE}} = 0$) if it is not correlated with both a_k and $\tilde{\mathbf{x}}_{1,k}$. An example of such a situation, in the presence of zero-mean signals with symmetric distributions, is the case where $\tilde{\mathbf{x}}_{1,k}$ and $\tilde{\mathbf{x}}_{2,k}$ gather the odd and even terms m of (6), respectively. Consequently, only M th-order CV MMSE filters such that M is odd containing only polynomial terms of odd order m ought to be used. For such filters, $\tilde{\mathbf{w}}$ and $\tilde{\mathbf{x}}_k$ are reduced to $\tilde{\mathbf{w}} = (w_{1,0}, w_{1,1}, w_{3,0}, w_{3,1}, w_{3,2}, w_{3,3}, \dots, w_{M,M})^T$ and

$\tilde{\mathbf{x}}_k = (x_k, x_k^*, x_k^3, x_k^2 x_k^*, x_k x_k^{*2}, x_k^{*3}, \dots, x_k^{*M})^T$. In this case, the components of $\tilde{\mathbf{x}}_k$ can be rearranged in order as $\tilde{\mathbf{x}}_k = [\mathbf{x}'_k, \mathbf{x}''_k]^T$ where $\mathbf{x}'_k = (x_k, x_k^3, x_k^2 x_k^*, \dots, x_k^{(M+1)/2})^T$ and $\mathbf{x}''_k = (x_k^*, x_k^{*3}, x_k x_k^{*2}, \dots, x_k^{*(M-1)/2})^T$. Then, the partial M th-order CV MMSE estimate y_k of a_k can be interpreted as the WL-MMSE estimate of a_k given \mathbf{x}'_k and thus the partial M th-order CV MMSE estimate inherits the properties of the WL-MMSE estimator [1]. In particular, for real-valued SOI symbols a_k , the estimate y_k given by (7) is real-valued.

In contrast, if $\tilde{\mathbf{x}}_{1,k}$ gathers the terms (x_k, x_k^*) and $\tilde{\mathbf{x}}_{2,k}$ the odd higher order terms, all the terms $\tilde{\mathbf{x}}_{2,k}$ are correlated with a_k and $\tilde{\mathbf{x}}_{1,k}$ and thus contribute to decrease the MMSE with respect to that of the WL-MMSE filter. This proves the better performance, in terms of MMSE, of the partial M th-order CV MMSE filter with only odd order terms with respect to the WL-MMSE filter.

Finally, note that in practice, $\mathbf{R}_{\tilde{\mathbf{x}}}$ and $\mathbf{r}_{\tilde{\mathbf{x}},a}$ are not known a priori and have to be estimated from a training sequence correlated with the SOI symbols using a least square approach.

3.2. Third-order complex Volterra MMSE filter

We consider in this paper M th-order CV MMSE filters with $M = 3$ and odd order terms only (i.e. $m = 1, 3$), whose input/output relation is given by

$$y_k = w_{1,0}^* x_k + w_{1,1}^* x_k^* + w_{3,0}^* x_k^3 + w_{3,1}^* x_k^2 x_k^* + w_{3,2}^* x_k x_k^{*2} + w_{3,3}^* x_k^{*3} \stackrel{\text{def}}{=} \tilde{\mathbf{w}}^H \tilde{\mathbf{x}}_k. \quad (11)$$

where $\tilde{\mathbf{w}}$ is defined by (8), but where $\tilde{\mathbf{w}}$ and $\tilde{\mathbf{x}}_k$ are now restricted to $\tilde{\mathbf{w}} = [w_{1,0}, w_{1,1}, w_{3,0}, w_{3,1}, w_{3,2}, w_{3,3}]^T$ and $\tilde{\mathbf{x}}_k = [x_k, x_k^*, x_k^3, x_k^2 x_k^*, x_k x_k^{*2}, x_k^{*3}]^T$, respectively. A filter defined by (11) is called a full WL Cubic filter or a WL-C(0, 1, 2, 3) filter, i.e., a WL-Cubic filter taking into account the cubic terms 0, 1, 2 and 3. We will see in Section 4 that partial WL-Cubic MMSE filters with a single third-order term 0, 1, 2 or 3 in (11), called WL-C(i) ($i = 0, 1, 2$ or 3) allow us to obtain an MSE close to the MMSE (9) of the full WL-Cubic structure.

3.3. Orthogonal decomposition

To give an enlightening interpretation of a full or partial WL-Cubic MMSE filter allowing one to understand its better behavior w.r.t. to the WL-MMSE filter, we extend the interpretation of the latter introduced in [18] using the orthogonal projection theorem. To this aim, we note that all the terms of $\tilde{\mathbf{x}}_k$ contain an SOI component through the orthogonal decomposition of $\tilde{\mathbf{x}}_k$

$$\tilde{\mathbf{x}}_k = a_k \begin{pmatrix} \mathbf{r}_{\tilde{\mathbf{x}},a} \\ \pi_a \end{pmatrix} + \tilde{\mathbf{i}}_k, \quad (12)$$

From (12), the ratio of the powers of the SOI component and the associated global noise component at the output of an arbitrary WL-Cubic filter $\tilde{\mathbf{w}}$, defines an SINR at its output:

$$\text{SINR}(\tilde{\mathbf{w}}) = \frac{|\tilde{\mathbf{w}}^H \mathbf{r}_{\tilde{\mathbf{x}},a}|^2}{\pi_a \tilde{\mathbf{w}}^H \mathbf{R}_{\tilde{\mathbf{x}}} \tilde{\mathbf{w}}}, \quad (13)$$

where $\mathbf{R}_{\tilde{\mathbf{x}}} \stackrel{\text{def}}{=} E(\tilde{\mathbf{i}}_k \tilde{\mathbf{i}}_k^H)$ is the covariance matrix of the component of $\tilde{\mathbf{x}}_k$ (12) which gathers all its terms uncorrelated with the SOI symbol a_k . It is easy to prove that the WL-Cubic filters $\tilde{\mathbf{w}}$ which maximize this SINR (13) are collinear to $\mathbf{R}_{\tilde{\mathbf{x}}}^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a}$. Applying the matrix inversion lemma to $\mathbf{R}_{\tilde{\mathbf{x}}} = \mathbf{R}_{\tilde{\mathbf{x}}} - \pi_a^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a} \mathbf{r}_{\tilde{\mathbf{x}},a}^H$ derived from the orthogonal decomposition (12), it is easy to verify that $\mathbf{R}_{\tilde{\mathbf{x}}}^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a}$ and $\mathbf{R}_{\tilde{\mathbf{x}}}^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a}$ are collinear. Consequently the WL-Cubic filters $\tilde{\mathbf{w}}$ which maximize the SINR (13) are collinear to $\tilde{\mathbf{w}}_{\text{CV-MMSE}}$ (8) taking into account the new definitions of $\tilde{\mathbf{w}}_{\text{CV-MMSE}}$ and $\tilde{\mathbf{x}}_k$. The maximum of the SINR (13), denoted MSINR is thus given by:

$$\text{MSINR} = \frac{1}{\pi_a} \mathbf{r}_{\tilde{\mathbf{x}},a}^H \mathbf{R}_{\tilde{\mathbf{x}}}^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a} = \frac{\pi_a^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a}^H \mathbf{R}_{\tilde{\mathbf{x}}}^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a}}{1 - \pi_a^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a}^H \mathbf{R}_{\tilde{\mathbf{x}}}^{-1} \mathbf{r}_{\tilde{\mathbf{x}},a}}. \quad (14)$$

From (12) and (13), it is straightforward to deduce the following general relation linking the MSE and the SINR at the output y_k of an arbitrary WL-Cubic filter $\tilde{\mathbf{w}}$:

$$\text{MSE}(\tilde{\mathbf{w}}) = \pi_a \left| 1 - \frac{\tilde{\mathbf{w}}^H \mathbf{r}_{\tilde{\mathbf{x}},a}}{\pi_a} \right|^2 + \frac{|\tilde{\mathbf{w}}^H \mathbf{r}_{\tilde{\mathbf{x}},a}|^2}{\pi_a \text{SINR}(\tilde{\mathbf{w}})}. \quad (15)$$

We deduce from (15) that the WL-Cubic filter $\tilde{\mathbf{w}}$, which minimizes $\text{MSE}(\tilde{\mathbf{w}})$ under the constraint $\tilde{\mathbf{w}}^H \mathbf{r}_{\tilde{\mathbf{x}},a} = \pi_a$ is also the WL-cubic filter which maximizes $\text{SINR}(\tilde{\mathbf{w}})$ under the same constraint. This shows that under the constraint $\tilde{\mathbf{w}}^H \mathbf{r}_{\tilde{\mathbf{x}},a} = \pi_a$, MSE minimization and SINR maximization are equivalent criteria, which gives a physical interpretation of the SINR criterion (13) in term of MSE minimization and we obtain

$$\text{MMSE} = \frac{\pi_a}{1 + \text{MSINR}}. \quad (16)$$

Without this constraint $\tilde{\mathbf{w}}^H \mathbf{r}_{\tilde{\mathbf{x}},a} = \pi_a$, (15) shows that the MSE minimization is no longer equivalent to SINR maximization, but $\tilde{\mathbf{w}}_{\text{CV-MMSE}}$, which minimizes $\text{MSE}(\tilde{\mathbf{w}})$ also maximizes $\text{SINR}(\tilde{\mathbf{w}})$, but is not the only one.

4. SINR AND BER PERFORMANCE ANALYSIS

4.1. SINR performance analysis

For arbitrary rectilinear SOI and CCI signals, it is easy to derive the SINR at the output of the WL-MMSE receiver for strong CCI (i.e., for $\pi_j \gg \eta_2$) [3]. It is given by

$$\text{SINR}_{\text{WL}} \approx \frac{2\pi_s}{\eta_2} (1 - \cos^2(\Delta\phi)), \quad (17)$$

where $\Delta\phi \stackrel{\text{def}}{=} \phi_j - \phi_s$. Rel. (17) shows that the WL-MMSE receiver performs the SAIC for $\Delta\phi \neq 0$, with decreasing performance as $\Delta\phi$ decreases to zero. In contrast, the derivation of the SINR at the output of the WL Cubic MMSE receiver is much more intricate. But using MATLAB symbolic algebra and calculus tools, we have proved that this output SINR follows the rational fraction form

$$\text{SINR}_{\text{WL-C}} = \frac{a_P \pi_j^P + \dots + a_1 \pi_j + a_0}{b_P \pi_j^P + \dots + b_1 \pi_j + b_0}, \quad (18)$$

where P depends on the considered partial WL-Cubic MMSE structure and the coefficients $a_0, \dots, a_P, b_0, \dots, b_P$, are functions of $\pi_s, \eta_2, \kappa_j, \chi_j, \kappa_s \stackrel{\text{def}}{=} E(a'_k{}^4), \chi_s \stackrel{\text{def}}{=} E(a'_k{}^6)$ and $\Delta\phi$.

In particular, the maximum value of P is 4 for WL-C(i) ($i = 0, 1, 2$ or 3) MMSE receivers. In this case, we have proved the following two asymptotic results according to the values of $\chi_j - \kappa_j^2 \geq 0$:

For $\chi_j - \kappa_j^2 = 0$ which occurs using (4) and (5) i.i.f. $\chi_j = \kappa_j = 1 \Leftrightarrow b'_k(\tau_j) = \pm 1$, i.e., for synchronized BPSK CCI,

$$\lim_{\pi_j \rightarrow \infty} \text{SINR}_{\text{WL-C}(i)} = \frac{a_3}{b_3} = \frac{2\pi_s}{\eta_2}, \quad \forall \Delta\phi. \quad (19)$$

For $\chi_j - \kappa_j^2 > 0$, i.e., for arbitrary rectilinear unsynchronized CCI:

$$\lim_{\pi_j \rightarrow \infty} \text{SINR}_{\text{WL-C}(i)} = \frac{a_4}{b_4} = \frac{2\pi_s}{\eta_2} (1 - \cos^2(\Delta\phi)). \quad (20)$$

Rel. (19) shows that for synchronized BPSK sources, the CCI is completely removed whatever $\Delta\phi$ thanks to an amplitude discrimination between the sources. In this case, the performance gain with respect to the WL MMSE receiver increases with $\Delta\phi$. This amplitude discrimination is generally strong in full-duplex systems [19]. For non-synchronized or non-BPSK sources, despite the absence of gain brought by WL-C(i) MMSE receivers with respect to WL MMSE receivers for infinitely strong CCI, it is possible to show still significant gains in SINR for strong but not infinitely strong CCI. This is illustrated in Figs.1 and 2 which show the SINR of the WL-Cubic(0) and WL receivers, as a function of $\Delta\phi$ for different values of τ_j for a roll-off of 0.3, and as a function of τ_j/T for different values of the roll-off α for $\Delta\phi = 0$, respectively. For both figures, the SOI and CCI have BPSK symbols with $\text{SNR} = \pi_s/\eta_2 = 10\text{dB}$ and $\text{INR} = \pi_j/\eta_2 = 30\text{dB}$. These two figures show that this gain in SINR decreases when the roll off α of $v(t)$ decreases and $\tau_j \in [0, T/2]$ increases. This is explained by the presence of increasing inter-symbol interference due to the pulse shaping filter, which Gaussianizes the CCI component $b'_k(\tau_j)$ due to the central limit theorem and for which the gain decreases toward zero.

4.2. BER performance analysis

To complete the SINR performance analysis, let us present in this subsection the bit error rate (BER) obtained at the output of the WL-C(0) MMSE and WL MMSE filters, both using a simple sign detector for BPSK synchronized SOI and CCI. In Fig.3, this BER (blue curves) is compared to those of the MAP receiver which knows the parameters $\pi_s, \pi_j, \phi_s, \phi_j$ and η_2 , as a function of the INR. The SINR at the outputs of the third-order WL-C(0) MMSE and WL MMSE filters are also plotted (red curves) in this figure, where $\text{SNR} = 10\text{dB}$. This figure illustrates the power's discrimination allowed by the WL-C(0) MMSE receiver, for which its output SINR attains its minimum for $\text{SNR} \approx \text{INR}$ and its maximum $2\pi_s/\eta_2$ for strong CCI where the BER is minimum. So the BER and SINR obtained by the WL-C(0) MMSE receiver have consistent behaviors. This reinforces the meaning of the SINR

defined by the orthogonal decomposition (13). We note that although the MAP receiver outperforms the WL-C(0) MMSE receiver, their BERs are very close for strong CCI.

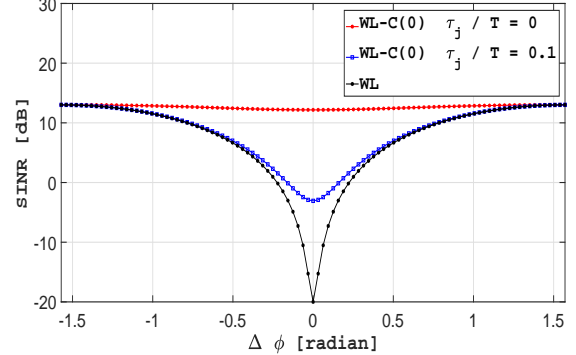


Fig.1 $\text{SINR}_{\text{WL-C}(0)}$ and SINR_{WL} as a function of $\Delta\phi$ for $\alpha = 0.3$.

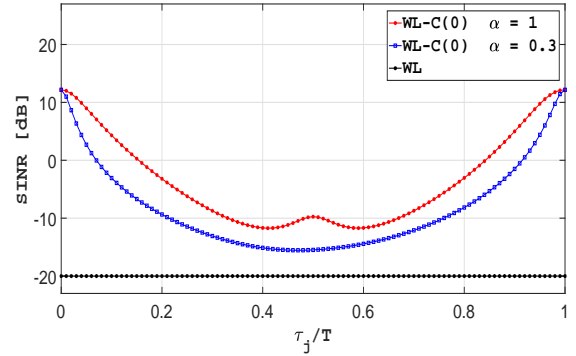


Fig.2 $\text{SINR}_{\text{WL-C}(0)}$ and SINR_{WL} as a function of τ_j/T for $\Delta\phi = 0$.

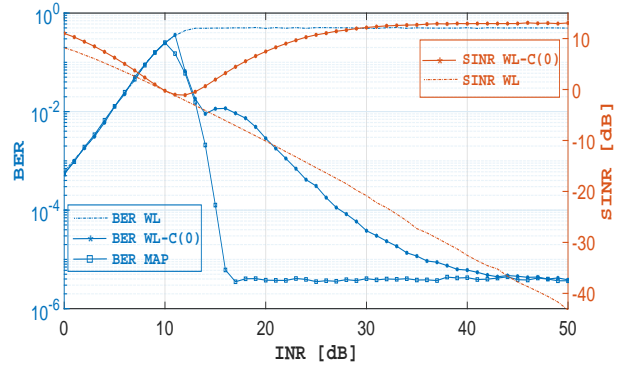


Fig.3 $\text{BER}_{\text{MAP}}, \text{BER}_{\text{WL-C}(0)}, \text{BER}_{\text{WL}}, \text{SINR}_{\text{WL-C}(0)}$ and SINR_{WL} as a function of INR for $\text{SNR} = 10\text{dB}$.

5. CONCLUSION

An enhancement of the SAIC concept based on WL-Cubic MMSE receivers, has been proposed in this paper for rectilinear signals. This new receiver allows to obtain high performance gain with respect to WL-MMSE receiver depending on the symbols constellation, the pulse shaping filter and relative phase and delays between the SOI and the CCI. More detailed performance results will be presented elsewhere jointly with extensions of this receiver to second-order circular constellations.

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