BEM Reference Signal Estimation for an Airborne Passive Radar Antenna Array

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In this paper, we consider an airborne passive radar using digital video broadcasting—terrestrial OFDM signals. In such an application, the channel composite nature and the mobility of the receiver deeply alter the signal and, therefore, degrade the decoding processing. Classic methods experience difficulties in dealing with such channel impacts. In order to reconstruct a proper reference signal, we propose here a channel estimation method that exploits both antenna diversity and a basis expansion model modeling the channel time variations. The application of this method to experimental in-flight recorded data demonstrates the performance improvement provided by the proposed method over the classic demodulation and other methods developed to cope with intercarrier interferences.

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I. INTRODUCTION

Passive radars have attracted a lot of attention and concern during the past decade. Signal processing methods have been developed for all available transmitters of opportunity such as FM [1], digital audio broadcasting (DAB) [2], digital video broadcasting-terrestrial (DVB-T) [3]-[5], and even global system for mobile or Wi-Fi [6]. This topic has nowadays reached a certain maturity. Systems have been designed and tested for different applications: At first air surveillance [7], but also ground moving target identification and sea surveillance [8], [9]. Besides passive radar characteristics, such as low weight, low power consumption, and stealthyness, they are good candidates to airborne applications. As a consequence, analyses have been recently drawn so as to study the applicability of passive airborne radars. Kulpa et al. [10] describe the overall concept, whereas Tan et al. [11] provide an insight of the necessary system performance, and Tan et al. [12] simulate the received signal and analyses the ambiguity function. The first experiments have allowed to examine ground clutter [13], and to consider the detection capabilities of such a system [14]-[16] based on FM signal. Moreover, Ulander et al. [17], [18] show the benefits of a suitable bistatic geometry to face ground clutter echoes for DVB-T SAR imagery.

In this context, the SONDRA laboratory, the Research Center of the French Air force, and the French Aerospace Laboratory (ONERA) have developed a complete passive receiver, including a specific antenna array. The system flew on the ONERA motorglider, named BUSARD, so as to detect airborne targets using the OFDM modulated DVB-T signals [19].

In addition, the detection step needs a relevant reference signal. This reference signal can be obtained either using a specific dedicated channel pointed toward the transmitter direction, or estimated from the received signal, possibly by exploiting the specific properties of the transmitted signal. The latter is especially true in case of digital modulations, DAB and DVB, for instance. The signal is demodulated and decoded in order to remove any multipath effect, before it is re-encoded and modulated again to create the proper multipath-free reference signal. This procedure presents the huge advantage to suppress the most of the multipath effects and target echoes. However, it cannot reproduce the transmitter and receiver behavior (filter, noise, etc.).

All previously cited articles take for granted that a relevant reference signal is available. However, the moving aeronautical environment introduces heavy constraints. Aeronautical channel models reasonably assume that the transmitter and the receiver are line-of-sight (LOS). So, to describe the en route scenario, Haas [20] proposes to divide the aeronautical channel into two contributions: an LOS, and a more stochastic diffuse backscattered component. A ground reflection component may be associated with this two-ray model [21]. This is all the more relevant that, in the current case, transmitters are generally tilted to the ground. This multipath environment implies that the aeronautical channel is strongly composite. Moreover, the aeronautical environment involves large distances and, as a consequence, potentially large multipath delays compared to the symbol period. Stronger ground echoes may, for instance, present a delay corresponding to the propagation time necessary to travel the distance from transmitter to ground and then from ground to receiver, that is about twice the receiver's altitude. Thus, the channel coherence bandwidth may be far less than the almost 8 MHz bandwidth of DVB-T signal. The aeronautical channel is hence frequency selective, which implies that deep fades may corrupt the channel frequency response (CFR).

In addition, it is also characterized by the mobility of the receiver. As a first consequence, the channel impulse response (CIR) may fluctuate over with time. Each component of the multipath aeronautical channel may show a different Doppler frequency. This differential Doppler shift degrades the orthogonality among the subcarriers and may introduce ICI. In other words, the received power does not only come from the desired subcarrier but also from its neighbors. The elementary symbol to be read is, therefore, corrupted by these interferences.

Usually, the OFDM channel estimation uses pilots, that is to say, symbols that are known at the receiver to get the least squares (LS) estimate of the CFR at pilot subcarrier indexes. The whole CFR is then inferred thanks to frequency domain interpolation. But it does not take into account time variations or potential ICI. Therefore, the literature proposes a large solution panel to estimate a time-varying channel in case of OFDM signals. First, some articles propose to expand interpolation to the time and frequency domains [22], [23]. As the estimation methods need to catch the channel fluctuations, Mostofi and Cox [24] derive a linear model based either on consecutive symbols, or only one symbol including its guard interval. And Hijazi and Ros [25] expand it to higher order polynomial.

Besides, methods were derived so as to benefit from all possible channel information. Especially, Tomasin and Butussi [23] propose a way to evaluate the precise channel length.Hijazi [26] considers a solution based on the combination of a minimum description length method that provides the number of active taps, and the ESPRIT algorithm that estimates their respective delays as detailed in [27]. It participates to decrease the number of coefficients to estimate [28].

When intercarrier interference (ICI) is preponderent, it must be estimated but also canceled. Several iterative solutions have been presented. Poggioni *et al.* [29] propose to decode and then re-encode the data to make use of virtual pilots, whereas Hijazi and Ros [30] derive a QR-factorization-based equalizer to remove ICI.

To represent the time-varying channel, basis expansion model (BEM) has also been introduced [31], [32]. The general time-domain model has then been adjusted to OFDM signals [33]–[35]. The BEM estimation principle consists in decomposing the CIR on an orthonormal basis of timevarying functions restricted over a certain span, to model the taps time variations. It includes polynomial BEM. However, many BEM types exist such as the complex-exponential BEM (CE-BEM), the generalized CE-BEM, the Karhunen-Loeve BEM [36] or the prolate spheroidal BEM [37]. The BEM type may be chosen so as to fit a possible statistical channel *a priori*. Thanks to the BEM, the number of coefficients to estimate deeply reduces.

Tang *et al.* [38] point the benefits of using the BEM on several consecutive OFDM symbols. In fact, this paper shows that this solution implies to increase the basis size so as to model the intrasymbol channel variations along this extended time window. A larger basis means more unknown parameters to estimate. Indeed, the CIR needs to be evaluated for each signal sample. Yet the number of CFR samples, i.e., the number of pilots, remains limited. Thus, only a few symbols may be taken into account.

These solutions focus on rapidly time-varying channel that are mainly characterized by the ICI impact. Yet the channel faced during our airborne passive radar experimentation, appears to be slow-time varying, and shows a wide diffuse backscattered component. Simple study on real data proves indeed that most of classic intrasymbol BEM coefficients are negligible. So we developed an original solution that is able to catch the composite nature of the channel and its intersymbol variations instead of intrasymbol fluctuations. Consequently, we derive in this paper a CE-BEM estimator tuned to OFDM signal based on a large block of consecutive symbols. Contrary to intrasymbol channel estimation that demands to estimate the CIR at every sample time, our method needs to evaluate the CIR only once per symbol. Thus, the complexity is severely decreased. However, this method needs some adjustment, as it appears that taking into account several consecutive symbols generates a certain phase rotation of the constellation due to the shift of the pilot position from one symbol to the next one. We describe and explain this phenomenon in this paper and propose solutions that correct this phase rotation. The proposed method has been implemented and tested on real data recorded during a flight test that happened in October 2015. The results confirm its ability to improve the channel estimation, and consequently the reference signal reconstruction.

Section II introduces the signal model and the environment constraints. Section III details the proposed method. Finally, Section IV presents the experimental results and Section V concludes this paper.

II. SIGNAL MODEL

A. DVB-T Signal

The DVB-T signal is based on an OFDM modulation [19]. Table I summarizes the 8K-mode parameters.

According to OFDM principles, first binary data are mapped into elementary symbols of a complex constellation. N elementary symbols constitute an OFDM symbol X. These elementary symbols are placed on N orthogonal subcarriers. L pilots are inserted among these N subcarriers. These pilots are encoded using a BPSK modulation and transmitted at boosted power. They are periodically distributed every $D_f = 12$ subcarriers and shifted by three

TABLE I DVB-T Mode 8K Signal Configuration

Constellation	64-QAM
Subcarrier number, N	8192
Useful Subcarrier number	6817
Pilot number, L	568
Guard Interval Samples, G	1024
Symbol useful duration, T_u	896 μs
Subcarrier spacing, $1/T_u$	1116 Hz
Bandwidth	7.61 MHz



Fig. 1. DVB-T scattered pilot pattern–black dots represent the pilot location.



Fig. 2. OFDM signal modulation.

subcarriers from one OFDM symbol to the next. As a consequence, the location of these scattered pilots is periodic in every four OFDM symbols (see Fig. 1). Let $P_n(m)$, for m = 0, ..., L - 1, denote their index for the *n*th OFDM symbol as

$$P_n(m) = mD_f + d_n \tag{1}$$

where $d_n = 3 \mod_4(n)$ and mod denotes the modulo operator. d_n represents the fact that pilot indexes are translated by three subcarriers from one OFDM symbol to the next. But, as it is based on the modulo operator, this structure is periodically repeated every four OFDM symbols, as illustrated by the black dots in Fig. 1. For instance, $d_0 = 3 \times \mod_4(0) = 0$, $d_1 = 3 \times \mod_4(1) = 3$, and $d_1 = d_5$.

The modulated signal is built in this way in the frequency domain. The time-domain equivalent symbol is inferred from **X** by applying an inverse discrete Fourier transform: $\mathbf{x} = \mathbf{F}^H \mathbf{X}$, where **F** is the unitary discrete Fourier matrix, and \cdot^H symbolizes the Hermitian transpose. A guard interval is concatenated to the front of **x**, which is composed of its last *G* samples. Fig. 2 represents the modulation of an OFDM symbol. In case of a stationary receiver, terrestrial for instance, the guard interval ensures the orthogonality among the subcarriers if the maximum delay does not exceed the length of this interval. However, the aeronautical mobile environment undermines this property. Namely, all the received paths coming from different direction show different Doppler shifts. As a consequence, the subcarriers are no longer orthogonal.

Let T_s and F_s denote, respectively, the sample period and rate. The pilots provide *L* samples of the CFR. As a consequence, it gives access to the CIR over only LT_s delays. More precisely, the DVB-T norm includes a guard band that prevents from interferences with neighboring transmitted channel. The CFR *L* samples are located in the useful band Bw_u as

$$Bw_u = LD_f \frac{F_s}{N}.$$
 (2)

It implies that the CIR may be observed at sample period

$$T_s^{(\text{CIR})} = \frac{N}{LD_f} T_s.$$
 (3)

Yet, as $D_f \neq N/L$, $T_s^{(\text{CIR})} \neq T_s$. So the CIR sample period differs from the signal one. As a consequence, virtual pilots have to be considered so as to cover the whole band. This equates to zero padding. It is as if there were $L' = \lfloor \frac{N}{D_f} \rfloor$ pilots. Thus, the CIR is sampled as so

$$T_s^{(\text{CIR})'} = \frac{N}{L'D_f} T_s \simeq T_s.$$
(4)

It can be consequently inferred that the maximum channel delay that may be observed is $L'T_s$, where L = 682. That represents a maximum delay of 75 μ s or 22.4 km. In the following, we assume, this is the channel length. For sake of simplicity, L will refer to L'.

B. Aeronautical Channel

As our interest is focused on low-altitude targets, the receiver altitude is limited up to $z_{RX} = 1500$ m. We assume that the transmitter is located at the same altitude– see Fig. 4(b): $z_{TX} = z_{RX}$. Therefore, the main echoes are mostly concentrated during the first 10 μ s, that is at most $100 \times T_s$. First, it points out the fact that the delay spread do not exceed the $1024 \times T_s$ guard interval. So we neglect the intersymbol interferences. Moreover, it implies that the channel coherence bandwidth is about 100 kHz, far less than the almost 8MHz. As previously stated, it confirms that the aeronautical channel is frequency selective.

The channel taps $h_{(n,l)}^{(t)}$ denote the complex path gain for the *n*th symbol and path delay lT_s . The received signal for symbol *n* can be written in a matrix way as

$$\mathbf{y}_n = \mathbf{H}_n^{(t)} \mathbf{x}_n + \mathbf{u}_n \tag{5}$$



Fig. 3. Channel matrix (a) without and (b) with ICI.

where \mathbf{u}_n is an additive white Gaussian noise, and where the *l*th diagonal of $\mathbf{H}_n^{(t)}$ corresponds to the *l*th delay $h_{(n,l)}^{(t)}$

In the frequency domain, this can be written as

$$\mathbf{Y}_n = \mathbf{F} \mathbf{y}_n = \mathbf{H}_n \mathbf{X}_n + \mathbf{U}_n \tag{7}$$

where $\mathbf{H}_n = \mathbf{F}\mathbf{H}_n^{(t)}\mathbf{F}^H$ is the channel matrix, and \mathbf{U}_n remains white Gaussian.

In case the channel is time invariant, $h_{(n,l)}^{(t)} = h_{(m,l)}^{(t)}$ for l = 0, ..., L - 1, and all n, m = 0, ..., N - 1. So, in that particular case, $\mathbf{H}_{n}^{(t)}$ is circulant according to (6) and \mathbf{H}_{n} is diagonal. On the contrary, the multipath aeronautical channel introduces interferences from each subcarrier to its neighbors. The channel matrix off-diagonal terms are, thus, not null anymore. The larger the product $f_{dT} = \Delta f_{\text{Dmax}} \times$ T_u , the heavier the consequences, where T_u is the useful symbol duration (see Table I), Δf_{Dmax} is the maximum Doppler shift relatively to the direct path Doppler shift, and f_{dT} is called the normalized Doppler spread–(see [26], ch. 1.2.4). So f_{dT} is the ratio of the maximum Doppler shift and the intercarrier spacing. It measures the loss of orthogonality. At least, only a few neighbor subcarriers can be assumed predominant. \mathbf{H}_n may then be banded, or tailored so as to be considered as in [33] in [39]. Fig. 3 illustrates the channel matrix.

In our case, the maximum velocity of the receiver is about $v_{\rm RX} = 50 \text{ m} \cdot \text{s}^{-1}$. Let us consider a 600 MHz carrier frequency, and a received path at angle α from the normal of the receiver. The Doppler shift relatively to the direct path can be written as

$$\Delta f_D(\alpha) = \frac{v_{\text{RX}}}{\lambda} \left[\cos(\pi/2 - \alpha) - \cos(\pi/2 - \alpha_{\text{DP}}) \right]$$
(8)



Fig. 4. Geometrical analyses to reach (a) Doppler and (b) delay estimations.

where α_{DP} denotes the direct path direction relatively to the receiver normal, and λ denotes the wavelength. If we consider that main echoes concentrate into a $\beta = 60^{\circ}$ wide beam–see Fig. 4(a)–the maximum Doppler shift is

$$\Delta f_{\text{Dmax}} = \max\left[\left| \Delta f_D(\alpha_{\text{DP}} - \beta/2) \right|, \left| \Delta f_D(\alpha_{\text{DP}} + \beta/2) \right| \right]. \tag{9}$$

Whatever α_{DP} , it leads to $\Delta f_{Dmax} \simeq 50$ Hz. Therefore, $f_{dT} \simeq 4.5\%$. Consequently, thanks to the slow velocity of the considered aircraft, the channel may be roughly supposed slow time varying. ICI impact may be then considered at most as limited.

C. Reference Signal Misestimation Impact on System Detection Capacity

Before considering any detection capacity, the passive systems have to retrieve the reference signal. To quantify the impact of reference signal misestimation is in fact complex. It mainly depends on the clutter nature and on the clutter cancelation processings. However, Cardinali *et al.* [40] highlight the detrimental influence of multipath residues in the reference signal on the cancelation capability of the algorithm proposed in [41]. Moreover, Tang *et al.* [42] points out that the quality of constellation estimation, that is to say of the reference signal, has a direct influence on the direct path suppression, even with different signal-to-noise ratio (SNR).

To the same extent, we propose here to measure the benefits of a proper reference signal on detection probability, thanks to a simple but representative simulation. In this simulation, we generate a DVB-T reference, which is used to generate the direct path at zero delay as well as one single target delayed in time (for simplicity but without lack of representativity, no Doppler shift is considered here). We then add random errors on a few elementary symbols of the DVB-T signal in order to mimic estimation errors on the reference signal. These errors are randomly drawn in the neighborhood of the original symbol as described in Fig. 5. After direct path cancelation according to [41] using the erroneous reference signal, a constant false alarm rate test is applied to perform the target detection, and enables us to evaluate the detection probability.

Table II summarizes the parameters used for the simula-



Fig. 5. Potential erroneous candidate symbols.

TABLE II Simulation Parameters		
Signal	8 K mode DVB-T	
Direct Path SNR False alarm probability Target delay Monte-Carlo simulations	$\begin{array}{l} {\rm SNR_{\rm DP}} = 100dB\\ P_{fa} = 10^{-4}\\ 200 \; {\rm samples}\\ 2000 \end{array}$	



Fig. 6. Detection probability versus number of erroneous elementary symbols per OFDM symbol.

tions. Fig. 6 shows the detection probability versus the number of erroneous elementary symbols per OFDM symbol for different target SNR: $20 dB \leq \text{SNR}_{TG} \leq 40 dB$. These results highlight the fact that even a few number of erroneous symbols may introduce a substantial decrease on detection probability. Let us consider a 30 dB target SNR for instance. Only about 125 erroneous symbols over the 6817 subcarriers lead to a detection probability lower than 0.5. The misestimation of the reference signal does not allow us to properly cancel the direct path, and therefore, involves higher spurious residual. Fig. 7 illustrates it by providing matched filter examples for two numbers of erroneous symbols in case of $SNR_{TG} = 30 dB$. The secondary lobes are more than 10 dB stronger in case of $n_{\rm err} = 250$ erroneous symbols than in the almost error free case where $n_{\rm err} = 10$. As illustrated by Fig. 7, the target may as a consequence be hidden under the residual interferences.

III. REFERENCE SIGNAL ESTIMATION

The description of the effects that the channel may have on the signal, and the impact of the reference signal misestimation on detection probability demonstrates the necessity of a relevant channel estimation method. We propose in the



Fig. 7. Matched Filter examples after direct cancelation considering two different numbers of erroneous elementary symbols per OFDM symbol–SNR_{TG} = 30 dB.

following a method to improve the quality of the reference signal estimate.

A. Antenna Diversity

We assume that the antenna array is composed of N_{RX} sensors. Let $\mathbf{y}_a = [\mathbf{y}_1, \mathbf{y}_2, \dots, \mathbf{y}_{N_{\text{RX}}}]^T$ denotes the received array vector.

Antenna diversity can help to reduce the impact of channel fluctuations due to multipath fading. The probability that each receiver channel faces simultaneously a deep fade lowers as N_{RX} increases (see [43], ch. 13.4). As a consequence, antenna diversity techniques contribute to face ICI [44]. By weighting each array channel properly, these methods aim at optimizing different output parameters such as SNR. It is the case for maximum ratio combining [44], [45]. Here, we propose to focus the antenna beam in the direction provided by the maximum eigenvalue of the received signal covariance matrix $\hat{R}_{y_{a}y_{a}}$

$$\lambda_{\max} = \max_{1,\dots,N_{\text{RX}}} \Lambda \tag{10}$$

where $\Lambda = [\lambda_1, \lambda_2, \dots, \lambda_{N_{RX}}]$ is the eigenvalue decomposition of $\hat{R}_{\mathbf{y}_a \mathbf{y}_a}$. Thus, the unitary eigenvector associated with λ_{max} , denoted as \mathbf{v}_{max} , may correspond to the direct path direction as LOS is assumed between the receiver and the transmitter. As the influence of secondary paths decreases, the proportion of useful signal increases. Therefore, it also contributes to improve SNR.

The diffuse backscattered component contributions may in that way be reduced. As ICI has been previously assumed limited and as antenna diversity helps reducing its impact, ICI is neglected in the following. The validity of this assumption was verified on the experimental data. It implies in particular that the CIR may be assumed constant during a symbol period.

From then on, y refers to the resulting received vector

$$\mathbf{y} = \mathbf{v}_{\max}^T \mathbf{y}_a \tag{11}$$

where \mathbf{y}_n represents the *n*th OFDM symbol extracted from **y**. \mathbf{Y}_n denotes its frequency domain representation as expressed by (7).

B. BEM Channel Model

The aim of the BEM modeling is to represent the channel fluctuations along N_a consecutive OFDM symbols contrary to their usual utilization to catch rapid intrasymbol variations. Let us consider a basis **B** containing Q + 1 functions. So Q corresponds to the number of functions that are used to model the channel. This parameter is generally set. The larger the basis, the better the channel modeling is supposed to be, but the larger the number of coefficients to estimate. Thus, Q is limited by the number of available equations.

The *l*th channel taps $h_{(.,l)}^{(t)}$ over the N_a symbols are described by the BEM coefficients $h_{(.,l)}^{(b)}$. Its modeling by the BEM is written as

$$h_{(n,l)}^{(t)} = \sum_{q=0}^{Q} \mathbf{B}(n,q) h_{(q,l)}^{(b)} + \epsilon_{n,l}$$
(12)

with **B** the $N_a \times (Q + 1)$ basis matrix, and $\epsilon_{n,l}$ the modeling error. For instance, the CE-BEM is defined as in [34]

$$\mathbf{B}(n,q) = \exp\left(j\frac{2\pi n}{N}\left(q - \frac{Q}{2}\right)\right). \tag{13}$$

We introduce $\mathbf{h}_q = [h_{(q,0)}^{(b)}, \dots, h_{(q,L-1)}^{(b)}]^T$ the decomposition coefficients for the *q*th function, $q = 0, \dots, Q$. So it comes that

$$\mathbf{H}_{n}^{(t)} = \sum_{q=0}^{Q} \mathbf{B}(n,q) \mathbf{H}^{(q)}$$
(14)

where $\mathbf{H}^{(q)}$ is the circulant matrix defined from \mathbf{h}_q

 ${\bf H}^{(q)} =$

$$\begin{bmatrix} h_{(q,0)}^{(b)} & 0 & \dots & 0 & h_{(q,L-1)}^{(b)} & \dots & h_{(q,1)}^{(b)} \\ \vdots & & & \ddots & \vdots \\ & \ddots & & & & h_{(q,L-1)}^{(b)} \\ h_{(q,L-1)}^{(b)} & \dots & h_{(q,0)}^{(b)} & \ddots & \ddots & 0 \\ 0 & & & & \vdots \\ \vdots & \ddots & & & & \\ \vdots & \ddots & & & & \\ 0 & \dots & 0 & h_{(q,L-1)}^{(b)} & \dots & h_{(q,0)}^{(b)} \end{bmatrix}$$

$$(15)$$

Therefore,

$$\mathbf{H}_{n} = \sum_{q=0}^{Q} \mathbf{B}(n,q) \mathbf{F} \mathbf{H}^{(q)} \mathbf{F}^{H}.$$
 (16)

Let us denote $\Delta_q = \mathbf{F}\mathbf{H}^{(q)}\mathbf{F}^H$. As used in [34], $\Delta_q = \text{diag}\left(\sqrt{N}\mathbf{F}_L\mathbf{h}_q\right)$, where \mathbf{F}_L is the restriction of \mathbf{F} to the *L* first columns. Finally,

$$\mathbf{Y}_n = \sum_{q=0}^{Q} \mathbf{B}(n,q) \Delta_q \mathbf{X}_n + \mathbf{V}_n.$$
(17)

 V_n expresses at once the channel additive noise and the modeling error. The *L* pilot rows of the previous equation can be extracted so that

$$\mathbf{Y}_{n}^{(p)} = \sum_{q=0}^{Q} \mathbf{B}(n,q) \Delta_{q}^{(p)} \mathbf{X}_{n}^{(p)} + \mathbf{V}_{n}^{(p)}.$$
 (18)

From the definition of Δ_q , it can be further shown that $\Delta_q^{(p)}$ can be inferred from the *L*-sized Fourier transform of \mathbf{h}_q .

The ratio $\mathbf{Y}_n^{(p)}(m)/\mathbf{X}_n^{(p)}(m)$ gives a sampled estimation of the CFR at the *m*th pilot. Classic demodulation uses them to compute the whole CFR by interpolation, independently over each OFDM symbol.

Here it comes

$$\frac{\mathbf{Y}_{n}^{(p)}(m)}{\mathbf{X}_{n}^{(p)}(m)} = \begin{bmatrix} \mathbf{B}_{(n,0)}, \dots, \mathbf{B}_{(n,Q)} \end{bmatrix} \begin{bmatrix} \Delta_{0}^{(p)}(m,m) \\ \vdots \\ \Delta_{Q}^{(p)}(m,m) \end{bmatrix} + \mathbf{V}_{n}^{(p)}(m).$$
(19)

As a consequence,

$$\begin{bmatrix} \frac{\mathbf{Y}_{0}^{(p)}(m)}{\mathbf{X}_{0}^{(p)}(m)} \\ \vdots \\ \frac{\mathbf{Y}_{N_{a-1}}^{(p)}(m)}{\mathbf{X}_{N_{a-1}}^{(p)}(m)} \end{bmatrix} = \mathbf{B} \begin{bmatrix} \Delta_{0}^{(p)}(m,m) \\ \vdots \\ \Delta_{Q}^{(p)}(m,m) \end{bmatrix} + \begin{bmatrix} \mathbf{V}_{0}^{(p)}(m) \\ \vdots \\ \mathbf{V}_{N_{a-1}}^{(p)}(m) \end{bmatrix}.$$
(20)

Finally,

$$\mathbf{T} = \mathbf{B}\Delta^{(p)} + \mathbf{V} \tag{21}$$

where **T** is the matrix whose columns are composed of the ratio of the received sample over the transmitted pilot, for one pilot over N_a symbols, and $\Delta^{(p)}$ is the matrix whose *m*th column is composed of the *m*th diagonal value of $\Delta_q^{(p)}$ for $q = 0, \ldots, Q$. Equation (21) allows us to estimate $\Delta^{(p)}$ from **T** and, therefore, to get the BEM coefficients.

The channel matrix is then inferred. As ICI has been assumed negligible, \mathbf{H}_n may be considered as diagonal, for off-diagonal terms may be neglected. As a consequence, equalization is made simple. The estimated OFDM symbol is derived from the ratio of the received vector over the estimated CFR. The equalization process provides for the *k*th subcarrier a complex value $\tilde{\mathbf{X}}_n(k)$ corresponding to the in-phase and quadrature components. This estimated point is then mapped to the exact transmitted constellation point $\hat{\mathbf{X}}_n(k)$.

C. Previous Model Adaptation

At signal reception, time synchronization provides the sample index of the beginning of one OFDM symbol. However, it only detects the main path. But the DVB-T networks may be a single frequency network (SFN), implying that there may exist a re-emitter, closer than the main broadcaster from the receiver. Besides, artificial delays may be used for the purpose of network optimization. In any case, it is highly unlikely that the main path delay falls exactly on the sample grid. Thus, sidelobes are generated, which



Fig. 8. Algorithm overview.

spread apart from the main path sample. As a consequence, we use a delay guard. It means that we voluntarily set the symbol beginning a few samples before. Let τ_G denotes this sampled delay.

In other words, we aim at estimating the channel response $h_{(n,l-\tau_G)}^{(t)}$ for the *n*th symbol. So, according to Fourier transform properties, this τ_G time translation equates to a frequency domain modulation. At the *m*th subcarrier, it expresses as

$$\Phi(m) = 2\pi \frac{m\tau_G}{N}.$$
(22)

It is especially true at pilot index. However, in the DVB-T case, the pilots positions are shifted by three subcarriers from one symbol to the next, as previously detailed–see Fig. 1. So the experienced modulation at the *m*th pilot of the *n*th symbol can be precised as–see (1)

$$\Phi\left(P_n(m)\right) = 2\pi \left(\frac{mD_f\tau_G}{N} + \frac{d_n\tau_G}{N}\right).$$
(23)

So the combination of delay guard and pilot index shifts, provokes a phase shift, which is composed of a modulated and a constant part. This has no consequence if only one symbol is considered. This is, for instance, the case of terrestrial stationary systems using the classic OFDM demodulation. Indeed, the translated CFR samples at pilot index helps estimating the τ_G -delayed CIR, as (23) expresses.

Yet in the present case, several consecutive symbols are used to catch channel time variations. Consequently from the *n*th symbol to the first, there exists a constant phase shift component $\Delta \Phi_n = \Phi(P_n(m)) - \Phi(P_0(m))$ for all m = 0, ..., L - 1

$$\Delta \Phi_n = 2\pi \left(\frac{d_n \tau_G}{N} - \frac{d_0 \tau_G}{N} \right)$$
$$\Delta \Phi_n = 2\pi \frac{d_n \tau_G}{N}$$
(24)

as $d_0 = 0$, as stated in Section II-A. This constant phase shift translates into a rotation of the estimated QAM constellation.

Two solutions can be considered to cope with this undesired effect. The first solution consists in creating a regular pilot grid by interpolation of the pilots every four subcarriers for every OFDM symbol. The second one imposes to transpose the pilot-based CFR into the time domain, in order to compensate the time modulation due to the pilot location shift and then to get back to the frequency domain. Here we resort to the first solution and initialize the BEM algorithm with a regular pilot pattern both in frequency and time dimension, derived from interpolation.

Fig. 8 provides an overview of the proposed algorithm.



Fig. 9. Busard Onera motorglider and Rivera antenna.

IV. EXPERIMENTAL RESULTS

A. Experimental Flight Tests

A flight test took place in South East of France. Data were recorded from the Mont Ventoux broadcaster:

- (1) effective radiated power: 50 kW;
- (2) transmitter altitude: 1900 m;
- (3) coarse transmitter to receiver distance: 47 km.

The DVB-T channels are transmitted over carrier frequencies from 570 to 618 MHz. The DVB-T mode 8 K is used, as previously described in Section II-A. A dedicated antenna, named Rivera, has been designed for the project. The main constraint comes from the necessity to fit the antenna array in the space available in the pod of the Busard motorglider (around 110 cm μ available), while the wavelength in the DVB-T dedicated frequency band may be up to 60 cm–Fig. 9. The $N_{RX} = 3$ receiving channels are disposed in a side-looking configuration. During the experimental flight, the aircraft speed was comprised between 45 and 50 m \cdot s⁻¹, and the altitude between 1000 and 1500 m above ground level.

B. Channel Estimation

To analyze the potential ICI, we use the linear intrasymbol model of the channel time variations as proposed by Mostofi and Cox [24] using the signal received on one of the three channels. Fig. 10 shows an extraction of the corresponding estimated channel matrix. The zoom view highlights that ICI spreads apart from the diagonal over a few subcarriers only. The channel matrix decrease from the diagonal to the outside reaches up to 30 dB over these maximum five subcarriers. It confirms our assumption that ICI may be neglected.

The BEM-based method has been built with $N_a = 32$ consecutive symbols. The basis **B** is provided by the exponential BEM, as defined by (13), as it aims at modeling Doppler-induced CIR phase evolution. *Q* has been empirically set to 32.

Fig. 11 displays the CFR for three different symbols.



Fig. 10. Estimated channel matrix-intrasymbol time variation model.



Fig. 11. Estimated CFR for three OFDM different symbols–every eight symbols.



Fig. 12. Estimated CIR-delay guard necessity.

There are eight OFDM symbols in between each of the considered symbols. It highlights the deep fades that characterize the channel we are facing. It can also be noticed that these fades are evolving along time. So the channel estimation has to cope with fading effects and to model their time fluctuations.

C. Model Initialization Necessity

Fig. 12 displays the estimated CIR for one symbol only. The direct path is obvious. It appears that it is unresolved. The fact that it falls between two samples, induces secondary sides lobes apart from the main taps toward the negative delaydomain in particular. Even if the present case



Fig. 13. Constellation rotation– $\tau_G = 30$.



Fig. 14. Estimated rotation angle for 32 consecutive symbols.



Fig. 15. Schematic definition of MER.



Fig. 16. Measured MER for four data records–classic demodulation (black), intrasymbol linear model (blue), intrasymbol BEM (green), antenna diversity (red), BEM intersymbol (orange), antenna diversity, and intersymbol BEM (yellow).



Fig. 17. Data record 2–demodulated constellation with four different methods. (a) Classic demodulation. (b) BEM. (c) Antenna diversity. (d) Antenna diversity + BEM.

does not correspond to an SFN environment, Fig. 12 justifies the necessity to introduce a delay guard.

Such a delay guard induces a rotation of the constellation according to Section III-C if no previous initialization of the BEM model is performed. The rotation angle depends on the symbol type, that is the symbol number modulo 4. As expressed by (24), in the present case, a delay guard of 30 samples implies a rotation angle that is a multiple of 3.95°. Fig. 13 shows the rotated constellations, for each symbol type. The angle may be estimated from the slope of the straight line joining the pilots. And Fig. 14 points the estimated angles for 32 consecutive symbols. It highlights the periodicity of their value, and confirms the theoretical analysis.

D. Reference Signal Estimation

To quantify the performance of the different algorithms, we use the modulation error rate (MER) (see [46], ch. 9.18.2). MER corresponds to the ratio of the sum of the power of the ideal constellation symbol vectors over the sum of the power of error vectors, as illustrated by Fig. 15, and defined by

$$\text{MER} = 10 \times \log_{10} \left(\frac{\sum_{n,k} |\hat{\mathbf{X}}_n(k)|^2}{\sum_{n,k} |\tilde{\mathbf{X}}_n(k) - \hat{\mathbf{X}}_n(k)|^2} \right) \quad (25)$$

where $\tilde{\mathbf{X}}_n(k)$ represents the estimated point and $\hat{\mathbf{X}}_n(k)$ is the closest constellation point for each subcarrier *k*th. MER can be seen as a form of SNR that gives an indication of the estimation relevancy.

Fig. 16 presents the performance obtained for different demodulation methods. We compare our method to four different algorithms and use the classic demodulation as a reference, where the CFR is inferred from interpolation of its pilot-based LS estimate. We first evaluate the performances of the intrasymbol linear model proposed in [24], and an intrasymbol BEM algorithm with a polynomial basis and Q = 2 [34], [35]. It clearly points out the fact that intrasymbol channel variations modeling does not fit the considered aeronautical channel. The attempt to estimate the ICI may introduce estimation disturbances. It emphasizes the benefits of intersymbol model.

Then two other methods are compared: The classic demodulation that uses the antenna diversity combined signal, as explained in Section III-A, and the BEM only intersymbol model. A general trend can be derived from these examples. The antenna diversity and the BEM modeling provide each a substantial improvement. The first one underlines the benefits of antenna diversity against the diffuse backscattered components, whereas the second one points out the positive impact of the intersymbol model. However, our method, by combining both, enables even



Fig. 18. MER versus number of erroneous symbols per OFDM symbol.

better performance. The results are all the more interesting that the estimation is more homogeneous over the QAM constellation than in the classic case where the edges undergo a heavier channel impact as it can be seen in Fig. 17 that shows the estimated constellations obtained with the four methods for the data record number 2. Three OFDM symbols are superimposed for each subfigure.

For data record 3, classic demodulation performs equivalently to our method. It may be explained by the fact that some of the demodulated points $\tilde{\mathbf{X}}_n(k)$ lie farther from the edge of constellation with our method than with the classic demodulation. Referring to the MER definition (25), this may impair the MER denominator and decrease the global MER. However, the overall benefits of our method are clearly visible on Fig. 17.

The MER can also be estimated for the simulated case of Section II-C. The power of the ideal constellation vectors may be average by the average power of equiprobable QAM symbols, and the power of the error vectors may be approximated by $n_{\rm err} \times d_{\rm QAM}$, where $n_{\rm err}$ is the number of errors and d_{QAM} corresponds to the distance between two constellation points, as shown in Fig. 5. Fig. 18 displays the resulting MER. Only a few decibel MER improvement may involve a large decrease of the number of erroneous symbols. Let us consider data record 2: With our method, MER = 22.4 dB, whereas MER = 18.8 dB with the classic demodulation. It, respectively, corresponds to the following numbers of erroneous symbols: $n_{\rm err} = 110$ and $n_{\rm err} = 260$. These reference signal improvements may be translated into detection probability improvements in accordance with Fig. 6 in the simple simulated case of Section II-C. In case of $SNR_{TG} = 30 dB$, it means $P_D > 0.5$ with our method, but $P_D < 0.1$ with the classic demodulation. So the reference signal method we propose may have worthwhile impact on the detection capacity of the system.

V. CONCLUSION

Passive airborne radar is a recent topic of research. Theoretical studies have established its feasibility, and first experiments have confirmed it. Even if stationary bistatic system has produced a substantial literature, adapted signal processing need to be developed to ensure the detection from an airborne platform. Especially, we have pointed out the harsh impact of the aeronautical channel on the DVB-T OFDM system. The composite nature of the received signal makes it difficult to extract the direct path component. Moreover, the mobility of the receiver implies channel time fluctuations that may cause ICI. Nevertheless, the low velocity of the considered platform and the use of antenna diversity allow us to neglect this complex effect. Despite this, a suitable method had to be developed to catch channel intersymbol variations. Thus, we derived a BEM-based solution that enables us to model the channel tap variations over several consecutive symbols. The experimental results obtained by the application of the proposed method to real data measurements confirm the efficiency of combining antenna diversity and BEM to face the aeronautical channel induced disruptions.

Thanks to our method, we went through the first step of passive detection that consists in retrieving a proper reference signal. Yet we have now to proceed forward signal processing in order to cancel the direct path and suppress clutter echoes. Then it will be possible to quantify the influence of a well-estimated reference on these algorithms. This will be the object of a subsequent paper.

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