

A selective control information detection scheme for OFDM receivers

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Abstract In wireless communications, both control information and payload (user-data) are concurrently transmitted and required to be successfully recovered. This paper focuses on block-level detection, which is applicable for detecting transmitted control information, particularly when this information is selected or chosen from a finite set of information that are known at both transmitting and receiving devices. Using an orthogonal frequency division multiplexing architecture, this paper investigates and evaluates the performance of a time-domain decision criterion in comparison with a form of Maximum Likelihood (ML) estimation method. Unlike the ML method, the proposed time-domain detection technique requires no channel estimation as it uses the correlation (in the time-domain) that exists between the received and the transmitted selective information as a means of detection. In comparison with the ML method, results show that the proposed method offers improved detection performance, particularly when the control information consists of at least 16. However, the implementation of the proposed method requires a slightly increased number of mathematical computations.

Keywords Channel estimation · Sequence correlation · Maximum Likelihood (ML) sequence estimation ·

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Orthogonal frequency division multiplexing (OFDM) · Time-domain decision criterion

1 Introduction

Over the last few years, increased demand for high-speed downloads and the rapid growth of mobile services have led to an exponential growth in wireless data traffic. To meet mobile subscriber and internet user demands for high-speed data, wireless telecommunication bodies have proposed some high speed wireless communication standards including the 3rd Generation Partnership Project (3GPP) Long Term Evolution (LTE) standard for 4G communication systems, and IEEE 802.11 a/g within wireless local area networks (WLANs) for Wi-Fi. In these wireless systems, orthogonal frequency division multiplexing (OFDM) is the adopted physical layer technology because it is spectrally efficient, and offers some immunity to multipath fading [1–3].

In wireless communications systems, the transmitted information usually consists of two components: control information, which may be chosen from a finite set of known candidate information; and the payload (user-data), randomly generated at the transmitter. However, detection of control-data requires different decoding procedures compared with user-data. OFDM decoding schemes, which exist in the literature can involve either one-tap equalization or block-level detection. The latter is most useful for the detection of selective control information while the former is widely applied for the recovery of payload data information. In a one-tap equalization (symbol-by-symbol) decoding scheme, each OFDM subcarrier symbol is independently decoded as described in [4–7] while in a block-level decoding procedure, a group of subcarriers are collectively considered at

the receiver [8]. In the case of a block-level detection, it is usually assumed that a group of subcarriers (at the receiver-end) is chosen from a fixed number of possible set (known at both transmitting and receiving terminals), as described in [8–11]. This paper studies block-level detection of selective control information in an OFDM system.

A practical application of block-level detection can be found in the determination of the modulation scheme for a given transmission scenario [12]. For instance, since the chosen modulation scheme could be 4-QAM, 16-QAM, 64-QAM or even 256-QAM, the receiver must first determine the modulation type before the subsequent user-data data decoding procedure. Hence, the control information that specifies the modulation scheme is considered as a selective control information because in this example, it can be chosen from four possible values. Therefore, in practical sense, the control information for each modulation type can be encoded and transmitted on a group of subcarriers, which are known at both transmitting and receiving terminals. Then, during decoding at the receiver, the received group of subcarriers that represent the control information is compared with all the four possible candidate control information that represent each modulation type, in order to determine the modulation type that nearly correlates with or corresponds to the received control information. Another practical example of selective control information is the control format indicator (CFI), which carries key LTE system information that enables each user equipment (UE) to correctly decode the main LTE control information within the LTE physical downlink control channel. In the LTE standard, the possible CFI value is 1, 2 or 3 [8]. A detailed description of LTE control information can be found in [13]. At the receiver, an appropriate detection scheme is required to recover the transmitted selective control information. In practical systems, successful detection of the control information is usually required in order to perform subsequent recovery or detection of the transmitted payload [14].

In the literature, a Maximum Likelihood (ML) criterion is considered as the standard block-level detection approach because it is more practical and computationally efficient (in terms of hardware implementation) compared to other methods such as successive interference cancellation (SIC) and K-best list sphere detector (K-LSD) [8]. An FPGA implementation of the ML estimation method for the detection of a critical LTE control information is described in [15]. Unfortunately, the ML estimation scheme requires channel estimation to mitigate channel fading effects. Hence, the detection performance of the ML scheme is largely dependent on the channel estimation. In practical systems, channel estimation is often achieved through the transmission of additional system resources in the form of pilot or training signals [4]. However, transmission of pilots reduces the overall spectral efficiency and a large number of pilots is often required

to improve channel estimation [16]. Several forms of channel estimation exist and in practical systems, an adaptive channel estimation is usually implemented at the receiver in order to select the best channel estimation approach for different channel fading environments. In addition, channel estimation methods such as the linear minimum mean square error (L-MMSE) have very high computational complexity compared with the least squares (LS) method of channel estimation [4, 17–19]. Moreover, the L-MMSE method requires a priori knowledge of both the channel correlation and noise statistics, which may either be unavailable or further introduce practical implementation and design constraints [19, 20].

To address these practical issues (associated with the ML based detection scheme), this paper introduces an alternative scheme that eliminates the need for channel estimation during block-level detection in an OFDM receiver. The proposed detection technique is based on a time-domain decision rule that uses the correlation that exists between the transmitted and received selective information as the basis for detection. Simulations compare the performance of the conventional ML estimation approach and the proposed method in terms of: block error rate (BLER) for different block sizes; and computational complexity, with regards to the number of real multiplications (RMs) and real additions (RAs).

The paper is structured as follows. Section 2 outlines a single antenna uncoded OFDM architecture and describes an OFDM receiver that uses an ML based block-level detection approach. Section 3 describes the implementation of the proposed time-domain detection technique. Section 4 presents the simulation results and related discussions on the BLER performance of the two considered methods. It also presents a comparison of the level of mathematical computations (RM and RA) required by each method. Finally, Sect. 5 highlights the main contributions of the paper.

2 OFDM architecture and ML based detection

This section describes the OFDM baseband architecture used for the investigations within this paper. It also gives an overview of the ML based estimation method for block-level detection as studied in, for example, [8].

Figure 1 shows a block diagram representation of a baseband OFDM architecture. It can be noted that this architecture inserts some pilots to enable channel estimation as required by the ML estimation method. However, as will be shown later, these pilots are not used and remain passive in the proposed detection method.

2.1 Transmitter

Let X be an OFDM sequence of length N_s , which is the raw input for the inverse fast Fourier transform (IFFT) stage

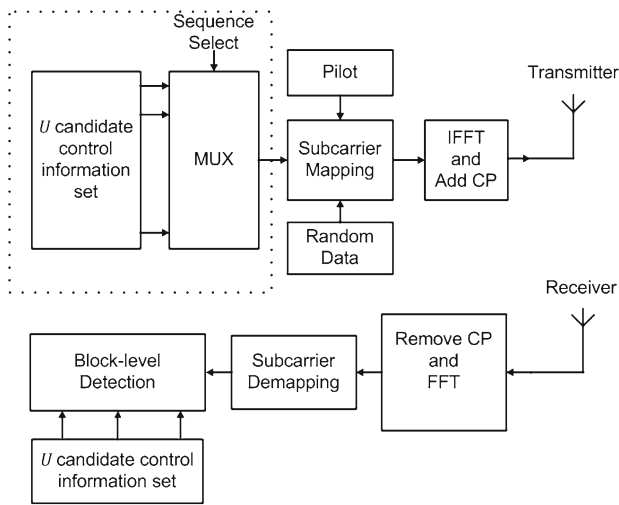


Fig. 1 System model

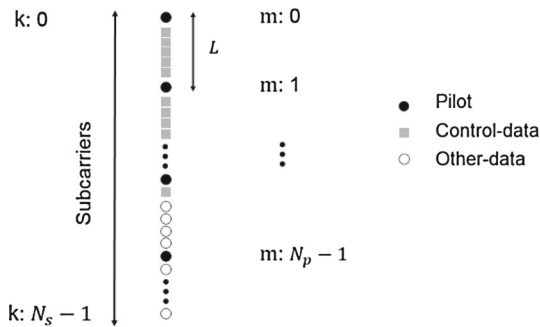


Fig. 2 An example of subcarrier allocation

before zero padding. Then, for $0 \leq k \leq N_s - 1$ where k represent a subcarrier index, \mathbf{X} is written as

$$\mathbf{X} = [X[0], X[1] \dots X[k] \dots X[N_s - 1]]. \tag{1}$$

For simplicity, it is assumed that \mathbf{X} consists of: a pilot sequence, \mathbf{X}_p of size N_p , a randomly generated data, \mathbf{X}_d of size N_d , and a selective sequence vector, \mathbf{X}_c of size N_c , such that $N_s = N_p + N_c + N_d$. Each element of \mathbf{X} is mapped to a subcarrier symbol $X[k]$. Figure 2 shows an example of the considered subcarrier allocation of pilot and non-pilot subcarriers achieved through, for example, a subcarrier mapping scheme described as follows:

1. First, N_p pilot subcarrier symbols are inserted at every L subcarrier indices where $L > 0$ and defines the pilot spacing. In this paper, L is set to 6, so that the pilots are mapped to $k = \{0, 6, 12, 18, \dots\}$.
2. Given that the selective sequence, \mathbf{X}_c is chosen from a finite set \mathcal{S} , which consists of U candidate sequence vectors where \mathcal{S} is considered to be deterministic and is

known at both transmitter and the receiver. For $1 \leq u \leq U$, each sequence vector in \mathcal{S} is denoted by S_u . Hence,

$$\mathcal{S} = \{S_1, S_2, S_u \dots S_U\}. \tag{2}$$

Let the u -index, \bar{u} define the index of the selected sequence vector from the set \mathcal{S} , then $\mathbf{X}_c \in \mathcal{S}$ i.e.,

$$\mathbf{X}_c = S_{\bar{u}} \text{ where } S_{\bar{u}} \in \mathcal{S}. \tag{3}$$

Each element of \mathbf{X}_c is then mapped to the next available N_c subcarriers in \mathbf{X} . As an example, S_1, S_2 , and S_3 may represent the control information for 4-QAM, 16-QAM and 64-QAM respectively. Hence, a block-level detection is performed at the receiver in order to determine an estimate of \bar{u} . Thus, in this example, the modulation scheme (decoded at the receiver) will be 4-QAM, 16-QAM and 64-QAM respectively when the value of \bar{u} is 1, 2 or 3.

3. Finally, the remaining N_d un-allocated subcarriers in \mathbf{X} are assigned to each element of randomly generated modulated data \mathbf{X}_d .

For an equi-spaced pilot arrangement (as in Fig. 2), studies in [4,5,21] showed that for $0 \leq m \leq N_p - 1$ and $0 \leq l \leq L - 1$ where m and l are arbitrary indices, each subcarrier symbol $X[k]$ can be represented as

$$\begin{aligned} X[k] &= X[mL + l], \quad 0 \leq l \leq L - 1 \\ &= \begin{cases} X_p[m], & l = 0, \text{ for pilots} \\ X_d[k], & l > 0, \text{ otherwise.} \end{cases} \end{aligned} \tag{4}$$

After subcarrier mapping as in the standard OFDM transmitter, the OFDM sequence \mathbf{X} is transformed into a time-domain OFDM signal using an IFFT to produce a time-domain OFDM signal \mathbf{x} of length N where $N > N_s$. For $0 \leq n \leq N - 1$, each individual sample $x[n]$ within \mathbf{x} is represented as

$$\begin{aligned} x[n] &= \frac{1}{\sqrt{N}} \sum_{k=0}^{N_s-1} X[k] \exp(j2\pi nk/N) \\ &= \text{IFFT}_{N\text{-point}} \{ \mathbf{X} \}. \end{aligned} \tag{5}$$

Similarly, as in standard OFDM, the length of OFDM signal \mathbf{x} is further extended by a cyclic prefix (CP) as illustrated in [22]. The use of CP provides some guard interval (GI) between consecutive OFDM transmissions, and involves copying the last N_{CP} samples in \mathbf{x} and placing them before the first sample in \mathbf{x} . CP also serves to reduce channel induced

inter-symbol interference (ISI). In addition, since a multipath fading channel has a finite impulse response, the use of CP also transforms the linear convolution between the transmitted signal and the channel impulse response into a circular convolution. However, to completely mitigate channel induced ISI, the GI duration should be larger than the maximum delay spread, τ_{\max} of the multipath fading channel.

2.2 Receiver

At the receiver, CP samples are first removed. In the presence of a fading channel with impulse response $\mathbf{h}[n]$ and additive white Gaussian noise (AWGN) $\mathbf{v}[n]$, each received time-domain OFDM signal sample, $y[n]$, is represented by

$$y[n] = \mathbf{h}[n] \otimes \mathbf{x}[n] + \mathbf{v}[n] \quad (6)$$

where \otimes represents a linear convolution. In addition, for a fading channel that has P taps with complex-valued gains, the channel impulse response $\mathbf{h}[n]$ may be expressed as

$$\mathbf{h}[n] = \begin{cases} 0, & P \leq n \leq N - 1 \\ \neq 0, & \text{otherwise.} \end{cases}$$

This implies all the channel energy is concentrated within the first P components of $\mathbf{h}[n]$. The next stage involves use of the fast Fourier transform (FFT), which transforms the time-domain signal into the frequency domain to produce \mathbf{Y} where

$$\mathbf{Y} = \mathbf{H}\mathbf{X} + \mathbf{V}. \quad (7)$$

i.e.,

$$\mathbf{Y}[k] = \mathbf{H}[k]\mathbf{X}[k] + \mathbf{V}[k]. \quad (8)$$

The terms $\mathbf{H}[k]$ and $\mathbf{V}[k]$ respectively represent the frequency domain representation of the channel and noise.

2.2.1 ML based detection

Let \mathbf{H}_c and \mathbf{V}_c respectively represent the sub-channel gains and AWGN components associated with the received selective sequence \mathbf{Y}_c of similar size. Similarly, \mathbf{H}_p and \mathbf{V}_p respectively represent the sub-channel gain and AWGN components within the received pilot sequence \mathbf{Y}_p . Then, similar to (7), \mathbf{Y}_c and \mathbf{Y}_p are respectively expressed as

$$\begin{aligned} \mathbf{Y}_c &= \mathbf{H}_c\mathbf{X}_c + \mathbf{V}_c \\ \mathbf{Y}_p &= \mathbf{H}_p\mathbf{X}_p + \mathbf{V}_p. \end{aligned} \quad (9)$$

Let \hat{u} denote an estimate of u . Then, using the ML estimation criterion, \hat{u} is obtained from [8]

$$\hat{u} = \arg \min_u \left\| \mathbf{Y}_c - \hat{\mathbf{H}}_c \mathbf{S}_u \right\|^2. \quad (10)$$

The term $\hat{\mathbf{H}}_c$ represents sub-channel estimates (associated with the received selective sequence). For a given non-pilot subcarrier index k , the sub-channel estimate at index k is determined by linear interpolation between two sub-channel pilot estimates $\hat{\mathbf{H}}_p[m]$ and $\hat{\mathbf{H}}_p[m+1]$, as described in [4–6] through

$$\begin{aligned} \hat{\mathbf{H}}_c[k] &= \hat{\mathbf{H}}[mL + l] \quad l > 0 \\ &= \hat{\mathbf{H}}_p[m] + \left(\hat{\mathbf{H}}_p[m+1] - \hat{\mathbf{H}}_p[m] \right) \frac{l}{L} \end{aligned} \quad (11)$$

where

$$\begin{aligned} \hat{\mathbf{H}}_p[m] &= \mathbf{Y}_p[m] / \mathbf{X}_p[m] \\ &= \left(\mathbf{H}_p[m]\mathbf{X}_p[m] + \mathbf{V}_p[m] \right) / \mathbf{X}_p[m] \\ &= \mathbf{H}_p[m] + \left(\mathbf{V}_p[m] / \mathbf{X}_p[m] \right). \end{aligned} \quad (12)$$

In terms of computational complexity and from the expression in (10), the ML method requires $UN_c -$ complex multiplications (CM), complex additions (CA) and $|\cdot|^2$ operations. These operations can also be expressed in terms of RM and RA since each CM and CA operation relates to RM and RA operations through [21]

$$1 \text{ CM} \cong 4 \text{ RM} + 2 \text{ RA} \quad \text{and} \quad 1 \text{ CA} \cong 2 \text{ RA}. \quad (13)$$

Based on the Taylor series expansion of trigonometric functions, each $|\cdot|^2$ will require ≈ 20 RMs plus 8 RAs, as implied in [23]. In summary, the ML detection approach requires a total of $24UN_c$ RMs and $12UN_c$ RAs. It can be noted that similar to studies in [21, 24], these evaluations exclude both the computational and implementation complexities of channel estimation and channel interpolation procedures associated with the ML method.

3 Proposed method

Correlation is one of the most frequently applied signal processing techniques for detection, as studied in, for example, [17, 25], for channel estimation. In addition, in terms of eliminating channel fading, studies have shown that partly

due to the partial elimination of the channel tap with the most energy in the time-domain, the use of a time-domain approach can produce improved performance i.e. is more robust against severe ISI compared with frequency-domain approaches [26]. The use of correlation and time-domain detection form the basis of the proposed detection scheme as it computes some form of time-domain correlation function between the received sequence Y_c and each candidate selective control information, S_u . The rationale for the use of a time-domain correlation also follows a general understanding that in the presence of limited noise level, there exists an inherent correlation between Y_c and X_c .

This paper produces an appropriate decision rule, which will facilitate correct detection or reduce the probability of erroneous detection, particularly in the case of an OFDM system. The proposed time-domain decision rule is now described.

Time-domain decision criterion

From the discrete correlation theorem (DCT), the correlation function (CORR) of two arbitrary time-domain signals a_1 and a_2 (of the same size) is obtained from [27]

$$\text{CORR}\{a_1, a_2\} = \text{IFFT}\{A_1 \times A_2^*\} \tag{14}$$

where * represents the complex conjugation, and A_1 and A_2 are respectively the frequency domain representations of a_1 and a_2 i.e.,

$$\begin{aligned} A_1 &= \text{FFT}\{a_1\} \\ A_2 &= \text{FFT}\{a_2\}. \end{aligned} \tag{15}$$

Using the DCT, the first step in the proposed method computes a term Z_u from direct multiplication of Y_c with the complex conjugate of S_u i.e.,

$$\begin{aligned} Z_u &= Y_c \times S_u^* \\ &= (H_c X_c + V_c) \times S_u^* \\ &= (H_c X_c S_u^*) + (V_c S_u^*) \\ &= \left(\frac{H_c X_c}{S_u}\right) + \left(\frac{V_c}{S_u}\right) \\ &= \left(\frac{H_c X_c}{S_u}\right) + V'_{c,u}. \end{aligned} \tag{16}$$

For $0 \leq c \leq N_c - 1$, Z_u may be represented as

$$Z_u = [Z_u[0], Z_u[1], Z_u[c] \dots Z_u[N_c - 1]]. \tag{17}$$

Alternatively, from the expression in (16), Z_u can also be written as

$$Z_u = \begin{cases} H_c + V'_{c,u}, & u = \bar{u} \ (X_c = S_u) \\ H_c X_c S_u^* + V'_{c,u}, & \text{otherwise.} \end{cases} \tag{18}$$

By omitting the noise terms in (18) for simplicity, the expression for Z_u can be re-written as

$$Z_u = \begin{cases} H_c, & u = \bar{u} \\ H_c X_c S_u^*, & \text{otherwise.} \end{cases} \tag{19}$$

From the expression in (19), the main difference between $Z_{\bar{u}}$ (when $u = \bar{u}$) and Z_u , (when $u \neq \bar{u}$) is described by

$$Z_u[c] = \begin{cases} 1, & u = \bar{u} \\ X_c S_u^*, & \text{otherwise} \end{cases} \tag{20}$$

where $X_c S_u^*$ is a complex-valued number with unity magnitude since each S_u is assumed to have the same magnitude. It can be noted that the expression in (20) is derived from (19) by setting H_c to 1 since from (19), the same channel term H_c is common to both $Z_{\bar{u}}$ (when $u = \bar{u}$) and Z_u , (when $u \neq \bar{u}$). In addition, the condition when $H_c = 1$ shows the scenario where there is no channel fading i.e. presence of only additive noise, as will be shown later in the next section.

From the DCT, a correlation function, W_u can be computed from the \mathcal{W} -point IFFT of Z_u where \mathcal{W} is a power of 2 and $\mathcal{W} \geq N_c$ i.e. $\mathcal{W} = N_c$ if N_c is a power of 2, otherwise, $\mathcal{W} > N_c$. For $0 \leq w \leq \mathcal{W} - 1$, W_u is given as

$$W_u = [W_u[0], W_u[1], W_u[w] \dots W_u[\mathcal{W} - 1]] \tag{21}$$

where

$$\begin{aligned} W_u[w] &= \sum_{c=0}^{N_c-1} Z_u[c] \exp(j2\pi cw/\mathcal{W}) \\ &= \text{IFFT}_{\mathcal{W}\text{-point}}\{Z_u\}. \end{aligned} \tag{22}$$

From the definition of Z_u in (20), the magnitude of $W_{\bar{u}}[w]$ (when $u = \bar{u}$) gives

$$|W_{\bar{u}}[w]| = \begin{cases} 1, & w = 0 \\ 0, & 1 \leq w \leq N_c - 1. \end{cases} \tag{23}$$

Table 1 Power-delay profile of the ETU multipath fading channel

Parameters	Tap 1	Tap 2	Tap 3	Tap 4	Tap 5	Tap 6	Tap 7	Tap 8	Tap 9
Path delay (μs)	0	0.05	0.12	0.20	0.23	0.50	1.60	2.30	5.00
Power (dB)	-1.0	-1.0	-1.0	0	0	0	-3.0	-5.0	-7.0

Table 2 Power-delay profile of the 3GPP-TU multipath fading channel

Tap 1–10			Tap 11–20		
Tap #	Path delay (μs)	Power (dB)	Tap #	Path delay (μs)	Power (dB)
Tap 1	0	-5.7	Tap 11	1.349	-17.4
Tap 2	0.217	-7.6	Tap 12	1.533	-19.0
Tap 3	0.512	-10.1	Tap 13	1.535	-19.0
Tap 4	0.514	-10.2	Tap 14	1.622	-19.8
Tap 5	0.517	-10.2	Tap 15	1.818	-21.5
Tap 6	0.674	-11.5	Tap 16	1.836	-21.6
Tap 7	0.882	-13.4	Tap 17	1.884	-22.1
Tap 8	1.230	-16.3	Tap 18	1.943	-22.6
Tap 9	1.287	-16.9	Tap 19	2.048	-23.5
Tap 10	1.311	-17.1	Tap 20	2.140	-24.3

Otherwise, $0 < |W_u[w]| < 1$ when $u \neq \bar{u}$. Now, using the statistical mean, the expression in (23) simply implies that

$$\frac{1}{\mathcal{W}} \sum_{w=0}^{\mathcal{W}-1} |W_{\bar{u}}[w]| \ll \frac{1}{\mathcal{W}} \sum_{w=0}^{\mathcal{W}-1} |W_{u \neq \bar{u}}[w]|. \quad (24)$$

Since $|W_u[w]|$ is positive-valued, then for a large value of N_c , the magnitude of the correction function can be described by Rayleigh distribution, as highlighted in the Appendix.

From the expression in (24), the proposed decision criterion can be defined by

$$\hat{u} = \arg \min_u \sum_{w=0}^{\mathcal{W}-1} |W_u[w]|. \quad (25)$$

It can be seen that in the proposed method, no channel estimation is required because the time-domain correlation inherently facilitates detection even in the presence of a fading channel. However, as will be shown in the next section, one of the main limiting factors for the proposed detection method is the inaccuracy of the related decision metric for small values of N_c .

4 Simulation results

This section presents comparison of both BLER performance and computational requirements between the two considered detection methods.

Table 3 Simulation Parameters

Parameters	Values
OFDM subcarrier spacing	15 KHz
Sampling frequency	3.84 MHz
Pilot spacing, L	6
IFFT/FFT size, N	256
OFDM sequence size, N_{sub}	180
Guard interval (GI)	5.21 μs

4.1 BLER performance

Simulations consider transmission over two well-known frequency-selective Rayleigh fading channels, namely: the extended Typical urban (ETU), with a root mean square (RMS) delay spread, τ_{rms} of 1 μs as defined in the LTE standard [28]; and the 3GPP Typical urban (3GPP-TU) channel, with τ_{rms} of 0.5 μs [29]. Tables 1 and 2 respectively show the power-delay profile of ETU and 3GPP-TU. To further understand the detection performance of each considered detection schemes, simulations also evaluate the BLER performance in the absence of a fading channel i.e. the presence of AWGN only.

Simulations are based on standard LTE parameters defined in Table 3 and all the transmitted information (payload and control) is QPSK modulated with U set to 8. The BLER performance of the two considered methods is investigated as a function of N_c by considering 50,000 OFDM symbol blocks for each SNR level where N_c is set to 8, 16, 32 and 64. Hence, $N_c = \mathcal{W}$. For the ML scheme, simulations implement

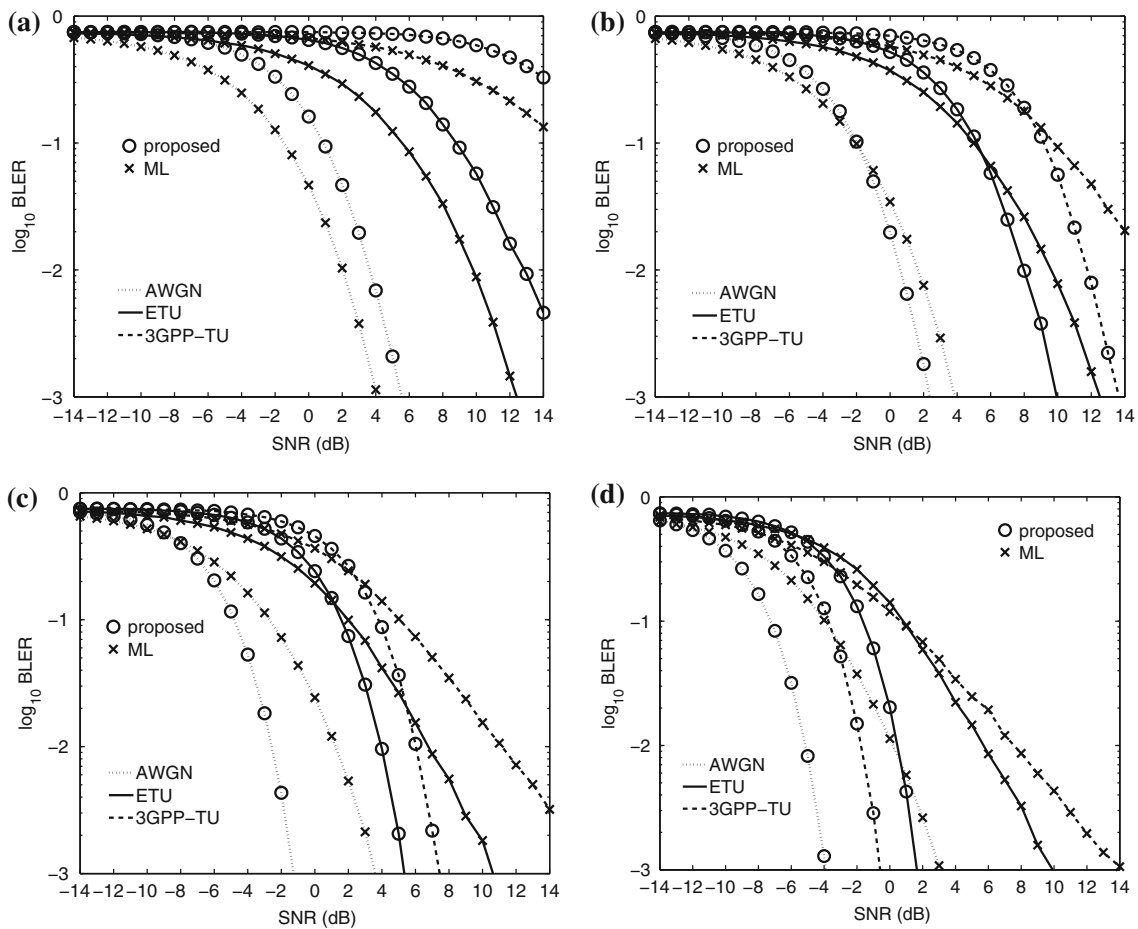


Fig. 3 BLER comparisons between the ML and the proposed estimation methods as a function of N_c . **a** $N_c = 8$, **b** $N_c = 16$, **c** $N_c = 32$ and **d** $N_c = 64$

Table 4 Estimated SNR (dB) requirements of the ML scheme and the proposed method at BLER = 1 %

Channel	Detection schemes	SNR (dB)			
		$N_c = 8$	$N_c = 16$	$N_c = 32$	$N_c = 64$
With no fading	ML	2.0	1.7	1.3	0.2
	Proposed	3.7	0.7	-2.4	-5.1
ETU	ML	9.9	9.7	6.8	5.8
	Proposed	12.9	8.0	4.0	0.6
3GPP-TU	ML	$\gg 12$	> 12	11.2	7.6
	Proposed	$\gg 12$	11.8	6.1	-1.6

the pilot-assisted channel estimation procedures previously outlined in Sect. 2.

Figure 3a–d shows the BLER comparisons between the two considered detection methods with N_c set to 8, 16, 32 and 64 respectively. Results in Fig. 3a shows that for a small block size ($N_c = 8$), the ML scheme produces improved BLER performance compared with the proposed method. Results in Fig. 3 also show that with no fading channel effects, the BLER performance of the ML scheme is relatively the same even when N_c is increased. This is because as N_c is further

increased, evaluation of the applied decision metric within the proposed method becomes more accurate and as a result, detection performance of the proposed method is improved compared with the ML scheme. In general, as N_c is increased, the proposed method produces improved detection performance in the form of reduced BLER. Results further show that in the presence of frequency-selective channel fading effects and due to poor channel estimation in these conditions, the ML scheme only produces minimal improvement in detection performance as N_c is increased.

Table 5 Computational requirements of the ML scheme and the proposed method

Complexity	ML	Proposed
RM	$24UN_c$	$(4UN_c) + (2UW \log_2 W) + (19UW)$
RA	$12UN_c$	$(2UN_c) + (3UW \log_2 W) + (8UW)$

For instance, at a BLER level of 1%, results further show that with N_c set to 16, 32 and 64, the proposed method requires relatively smaller SNR levels compared with the ML scheme. Table 4 shows the approximate SNR level required by each considered detection method in order to achieve, for example, a target BLER level of 1%. Results in Table 4 show that for each of the considered channel conditions and when $N_c = 8$, the ML approach requires a relatively smaller SNR level to achieve the target BLER of 1% compared with the proposed method. However, compared with the ML method and as N_c is increased, the proposed method requires significantly smaller SNR level to achieve the same BLER level of 1%. It can be noted that in the ETU fading channel, the proposed method has ≈ 2.8 and 5.2 dB SNR gain (with respect to the ML scheme) for N_c set to 32 and 64 respectively. Similarly, in the 3GPP-TU channel and with N_c set to 32 and 64 respectively, the proposed method has ≈ 5.1 and 9.2 dB gain in SNR. For practical purposes, low SNR requirement of the proposed method makes

it a viable and an attractive block-level detection scheme in OFDM receivers because the need for higher SNR levels for the case of the ML approach increases power consumption.

4.2 Computational requirements

From the expression in (25), it can be noted that the proposed method requires approximately UN_c CMs, $UW |\cdot|$ computations and U IFFTs (W -point) where each IFFT requires approximately $W/2 \log_2 W$ CMs and $W \log_2 W$ CAs, as indicated in [30]. As before, each complex-valued operation (i.e. CAs and CMs) can also be expressed in terms of real-valued (RM and RA) operations. Table 5 shows the approximate level of computational requirements of the ML and the proposed time-domain sequence level decision methods. However, these evaluations exclude the computational complexity of channel estimation and added system resources associated with the ML method.

Results in Table 5 show that due to the use of multiple IFFTs in the proposed method, it requires higher levels of computations compared with the ML method, particularly when the value of N_c or W is large. With $U = 8$, Fig. 4a–d shows the graphical comparison of computational complexity (in terms of RAs and RMs) between the ML scheme and the proposed method with N_c set to 8, 16, 32 and 64 respectively. For instance, using the expressions in Table 5 and

Fig. 4 Comparisons of computational complexity (number of RAs and RMs) between the ML scheme and the proposed method. **a** $N_c = 8$, **b** $N_c = 16$, **c** $N_c = 32$ and **d** $N_c = 64$

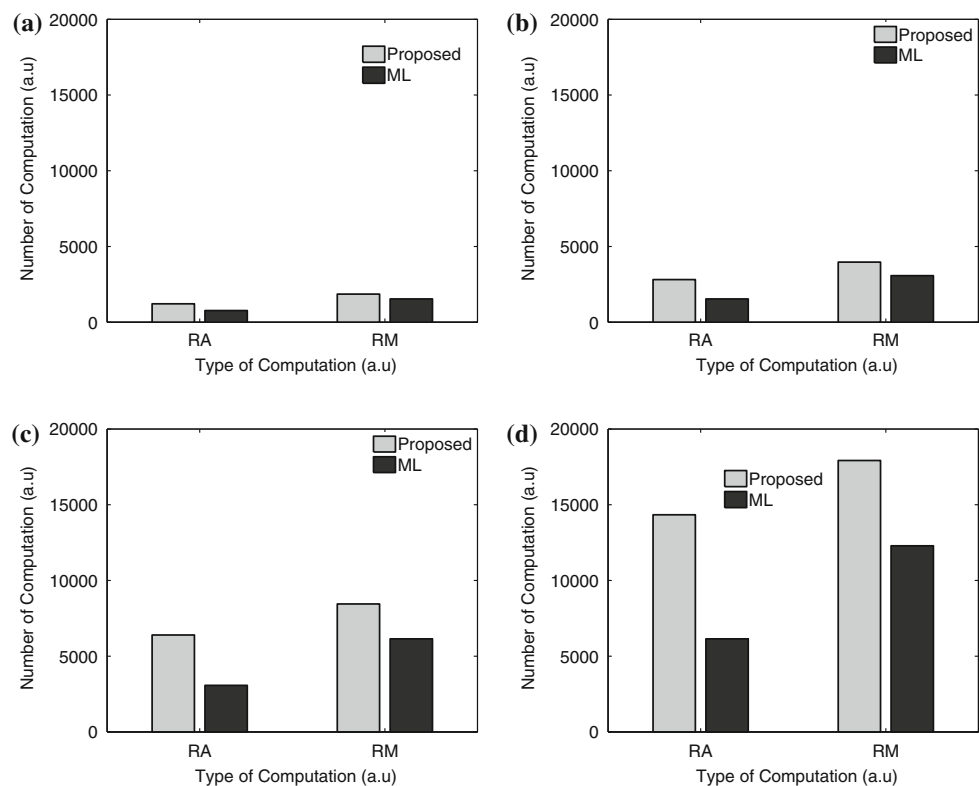
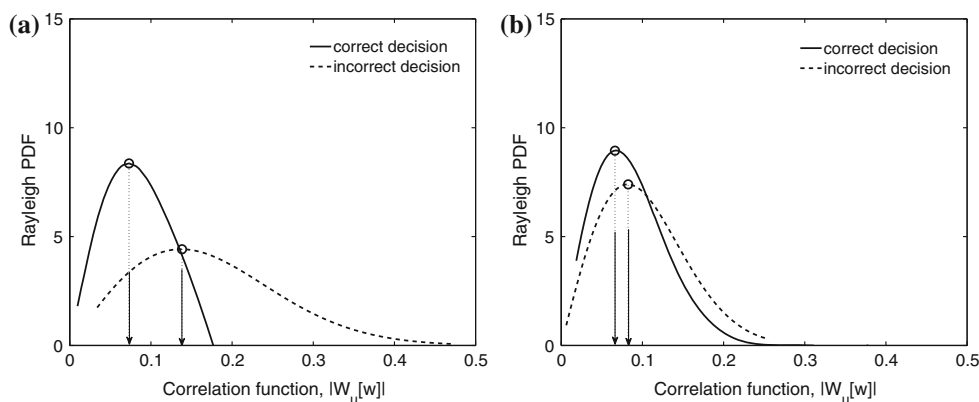


Fig. 5 Rayleigh distribution of $|W_u[w]|$. **a** No fading channel and **b** with a fading channel



in comparison with the ML approach, the proposed method requires approximately—17, 22, 27 and 31% extra RMs and 37, 45, 52 and 57% extra RAs for $N_c = 8, 16, 32$ and 64 respectively. Moreover, with advanced DSP devices, the computational complexity of IFFTs is no longer a critical issue. Hence, the proposed method is still an attractive choice, since it can produce improved detection performance compared with the ML approach and requires no channel estimation and no use of pilots.

5 Conclusions

This paper has presented and compared the detection performance of a time-domain decision technique against the conventional ML approach for block-level detection of selective control information in OFDM systems. An improved detection performance can be achieved with the proposed method compared with the ML approach. Another key benefit of the proposed method is that unlike the ML approach, it requires no channel estimation, which means no system data overhead in the form of training signals or preambles are required to be transmitted or used at the receiver. Hence, the proposed method reduces several practical implementation issues associated with channel estimation. However, without taking into account the added complexity of channel estimation associated with the ML method, the computational complexity of the proposed method is slightly higher than the ML estimation method due to the need for multiple IFFT computations. Results therefore suggest that the proposed method is a viable and an attractive detection scheme for OFDM systems.

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Appendix: Rayleigh PDF of $|W_u[w]|$

Using the Rayleigh distribution, this section describes the rationale behind the use of the proposed decision metric. The magnitude of each correlation function $|W_u[w]|$ can be represented by a Rayleigh probability density function (PDF) since it is positive valued and is obtained from the magnitudes of a Gaussian distributed complex numbers [31]. By letting $x = |W_u[w]|$, the PDF of x is represented as [32]

$$F(x) = \frac{x}{\Theta^2} \exp(-x^2/2\Theta^2), \quad x > 0$$

where Θ is the scale parameter of the Rayleigh distribution, which is a measure of the spread of the distribution. The value of Θ also indicates the mode of the distribution i.e. the point (the value of x) at which the PDF, $F(x)$ is maximum [32]. As a function of Θ , the mean of x , $E(x)$ is defined by [32]

$$E(x) = \Theta \sqrt{\frac{\pi}{2}}.$$

From the expression for $E(x)$, it seems there is a linear relationship between Θ and $E(x)$. Hence, as previously claimed and with regards to the proposed decision method, the corresponding value of Θ for the PDF of $|W_u[w]|$ is expected to be smaller in the case of correct decision compared with the case of incorrect decision. For instance, with $N_c = 32$ and SNR set to 6dB, Fig. 5a, b respectively shows examples of resulting Rayleigh PDF of $|W_u[w]|$ in the absence of channel fading and with ETU fading channel. These PDF curves clearly indicate that whenever there is a perfect or correct detection, the value of Θ is smaller than that when there is an incorrect decision. Therefore, amongst all the U correlation functions, the one that possesses the minimum or the lowest mean value gives correct decision.

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