

Robust and Efficient OFDM Synchronization for FPGA-Based Radios

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Abstract Orthogonal frequency division multiplexing (OFDM) is a popular modulation technique that can combat impulsive noise, is robust to multipath fading, is spectrally efficient, and can allow flexible allocation of spectrum. It has become a key standard in cognitive radio systems as well as an enabling technology for mobile data access systems. An OFDM receiver's performance is heavily impacted by the accuracy of its symbol timing offset (STO) and carrier frequency offset (CFO) estimation. This paper proposes a novel OFDM synchronization method that combines robust performance with computational efficiency. FPGA prototyping is used to explore the trade-off between the number of computations to be performed and computation word length with respect to both synchronization performance and power consumption. Through simulation, the proposed method is shown to provide accurate fractional CFO estimation as well as STO estimation in a range of channels. In particular, it can yield excellent synchronization performance in the face of a CFO that is larger than many state-of-the-art synchronization implementations can handle. The system implementation demonstrates efficient resource usage and reduced power consumption compared with existing methods, and this is explored as a fine-grained trade-off between performance and power consumption. The result is a robust method suitable for use in low-power radios, enabling less precise analog front ends to be used.

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1 Introduction

Orthogonal frequency division multiplexing (OFDM) is an effective modulation technique that has been used in both wired and wireless communication systems. Its advantages of combating impulsive noise, its robustness to multipath effects, and its spectral efficiency have led to OFDM being chosen for various high-bit-rate wireless transmission systems such as wireless local area networks (WLAN) defined in IEEE 802.11 and the metropolitan area networks (MAN) in IEEE 802.16. OFDM is also a key modulation scheme for cognitive radios systems [15] where it provides robustness in frequency selective channels and the ability to employ spectrum pooling, where unlicensed users may temporarily access spectrum resources during licensed users' idle periods, thus facilitating dynamic spectrum access (DSA) networks. However, OFDM performance is sensitive to receiver synchronization; carrier frequency offset (CFO) causes inter-carrier interference, and errors in timing synchronization can lead to inter-symbol interference [6]. Therefore, synchronization is critical to the performance of OFDM systems.

Many techniques have been proposed for effective OFDM synchronization. The two different transmission modes, continuous mode and burst packet mode, may also require different synchronization schemes [1]. In this paper, we investigate synchronization for burst packet mode, in which all OFDM frames begin with preamble symbols, and both frequency offset estimation and timing synchronization have to be completed within the duration of this preamble. Exploiting the characteristic preamble symbol of two identical halves, Schmidl and Cox [18] introduced metrics for autocorrelation-based synchronization, which can be computed iteratively at low cost and is robust to frequency offset. However, the resulting metric exhibits a plateau which leads to some uncertainty in determining the start of a frame. It is also limited in terms of CFO estimation.

Kim and Park [9] proposed a method to improve the robustness of fine STO estimation in the face of large CFO. The method applies multiple cross-correlations with pre-rotation by all possible integer CFOs to calculate fine STO, assuming integer CFO is less than ± 6 sub-carrier spacings. However, the method is unsuited to large CFO offsets, and the computational cost is significant since all possible integer CFOs must be calculated.

Kishore and Reddy [10] introduced new metrics based on cross-correlation between the known and received preamble symbols. These can accurately determine start of frame even at low signal-to-noise ratios (SNRs). Moreover, the method is robust to large CFO. However, the cross-correlation operation used in this method is computationally expensive, limiting suitability for implementation, especially in low-power systems.

This paper proposes a revised synchronization approach which supports the IEEE 802.16-2009 preamble [8]. It is robust to large frequency offset, has accurate STO estimation, and is computationally efficient. The novel method is presented along

with simulations to demonstrate performance. As cognitive radios are increasingly being built using FPGAs [12, 20], we then present results for an FPGA implementation with an analysis of computational complexity compared against other state-of-the-art methods.

In summary, the novel method involves a reformation of the synchronization sequence, a widened synchronization search space, and a computation engine that can formulate the timing metrics with a significant efficiency improvement over existing methods.

This paper is organized as follows: Sect. 2 discusses OFDM synchronization and outlines the proposed method, which is then simulated and evaluated in Sect. 3. The hardware implementation is presented and discussed in Sect. 4 and Sect. 5 concludes the paper.

2 OFDM Synchronization

We first investigate two state-of-the-art OFDM synchronization methods in terms of performance and computational complexity, before presenting our proposed method and evaluating it.

2.1 Existing Methods

Autocorrelation-based approaches are commonly preferred for implementing synchronization in practical systems due to low computational complexity [5, 11, 13, 21]. The state-of-the-art scheme in [7, 17, 19] uses autocorrelation-based timing metrics initially presented in [18] for frame detection, coarse STO, and fractional CFO estimation, followed by an additional cross-correlation to compute fine STO and integer CFO.

These timing metrics are defined as:

$$\begin{aligned}
 P(d) &= \sum_{m=0}^{L-1} (r_{d+m}^* r_{d+m+L}) \\
 R(d) &= \sum_{m=0}^{L-1} |r_{d+m+L}|^2 \\
 M(d) &= \frac{|P(d)|^2}{(R(d))^2}, \tag{1}
 \end{aligned}$$

where d denotes a time index corresponding to the first sample in a window of $2L$ samples of received signal r , and $*$ denotes complex conjugation. The $M(d)$ metric forms a plateau when the preamble is present in the received window [18]. In implementation, these metrics are generally most efficient to compute in a recursive manner as follows:

$$\begin{aligned}
 P(d+1) &= P(d) + (r_{d+L}^* r_{d+2L}) - (r_d^* r_{d+L}) \\
 R(d+1) &= R(d) + |r_{d+2L}|^2 - |r_{d+L}|^2 \tag{2}
 \end{aligned}$$

The presence of the preamble can be detected by thresholding the magnitude of $M(d)$. After detecting the frame, the fractional CFO is estimated based on the angle of the $P(d)$ metric [18]. Before the fine STO estimation is performed using cross-correlation—which is sensitive to CFO—the estimated fractional CFO must be compensated by time-domain phase de-rotation of the received samples.

The purpose of timing synchronization is to find a correct starting point for the discrete Fourier transform (DFT) window used by the receiver for demodulation. It should be noted that coarse time synchronization using autocorrelation (mentioned above) does not consistently detect the correct starting point; therefore, fine STO estimation is necessary for accurate timing synchronization. This can be found accurately by cross-correlating the received samples with the known preamble to detect the peak which occurs when the received samples match the known preamble. Unfortunately, this requires a significant amount of high-speed computation. One way of reducing the overhead of cross-correlation is to use a sign-bit multiplier [19] in place of a full word multiplier. However, this is highly sensitive to CFO.

In practice, fractional CFO can be effectively estimated using autocorrelation only within a limited range; in the case of the IEEE 802.16 preamble, up to ± 2 sub-carrier spacings [9]. If the CFO is outside this range, integer CFO has been introduced as a multiple of the coarse CFO's range. This integer CFO is most commonly estimated by frequency-domain cross-correlation [2, 9]. However, a critical problem arises if the CFO is larger than the range of the coarse CFO. In this case, fractional CFO estimation does not correctly estimate the CFO, and hence it is not suitably compensated, leading to a degradation in fine STO performance.

Previous implementations of such systems tend to assume that CFO is less than the range of the coarse CFO [3, 5, 7, 13, 17, 19]. However, this necessitates highly accurate timing at the RF interface, which translates to increased system cost. While the method generally works well and copes with low SNR, it can only operate when the CFO is within a limited range.

Kishore and Reddy [10] presented an alternative method using cross-correlation-based timing metrics between the known transmitted and received preamble symbols. These metrics are defined similarly to those in [18] but the $P(d)$ metric is reformulated to rely on cross-correlation; cross-correlation presents distinct peaks corresponding to when the received samples match the known transmitted preamble samples. A frame is detected when M crosses a threshold and the start of frame is estimated by searching for peaks in M . This method can accurately determine the start of a frame even at low SNR. Moreover, this method is robust to large CFO for time synchronization. Simulations in [10] show that time synchronization can be performed with a CFO of up to 10.5 carrier spacings. The price for this performance is that cross-correlation requires complex computation and a large number of multiplications are involved. This can be alleviated to some extent by using a multiplierless correlator [22] (as mentioned above), at the expense of a degraded P metric which is used to estimate frequency offset. So, although this method presents a useful solution, it is hampered in practice by computational complexity issues. As a result, no practical implementation of that scheme has been reported.

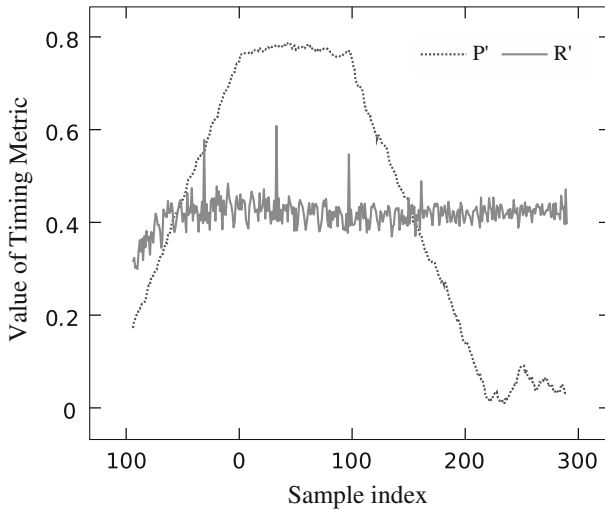


Fig. 1 Proposed timing metrics applied to the IEEE 802.16 preamble in AWGN (SNR = 10 dB, CFO = 10.5)

2.2 Proposed Metric

Considering the limitations of previous approaches, we propose a new timing metric that takes advantage of preamble characteristics such as period and energy distribution, again based upon the IEEE 802.16-2009 preamble. Firstly, we define an autocorrelation between the two halves of the receiver window for normalization purposes:

$$P'(d) = \sum_{m=0}^{C-1} (r_{d+m}^* r_{d+m+L}), \tag{3}$$

where L denotes the length of the preamble period (64 for the IEEE 802.16 preamble), and C is the length of received samples r for estimation.

Next, a power measure R' is proposed which takes account of both received and known transmitted preamble symbol power;

$$R'(d) = \sum_{m=0}^{C-1} |r_{d+m+L}|^2 |a_m|^2, \tag{4}$$

when $|a_m|$ denotes the normalized amplitude of the preamble at the transmitter.

In common with the metric in [18], our $P'(d)$ detects the periodic characteristic of the preamble and can thus be used for estimating fractional CFO. In fact, it forms a plateau when the preamble is presented within the receiver window. Figure 1 plots the values of both P' and R' when the preamble is received over a 10 dB SNR AWGN channel. It can also be seen that R' peaks at the points where the energy distribution of the received samples matches that of the transmitted preamble. Since it is the R' metric that is used to estimate time synchronization, and this is computed on the real square

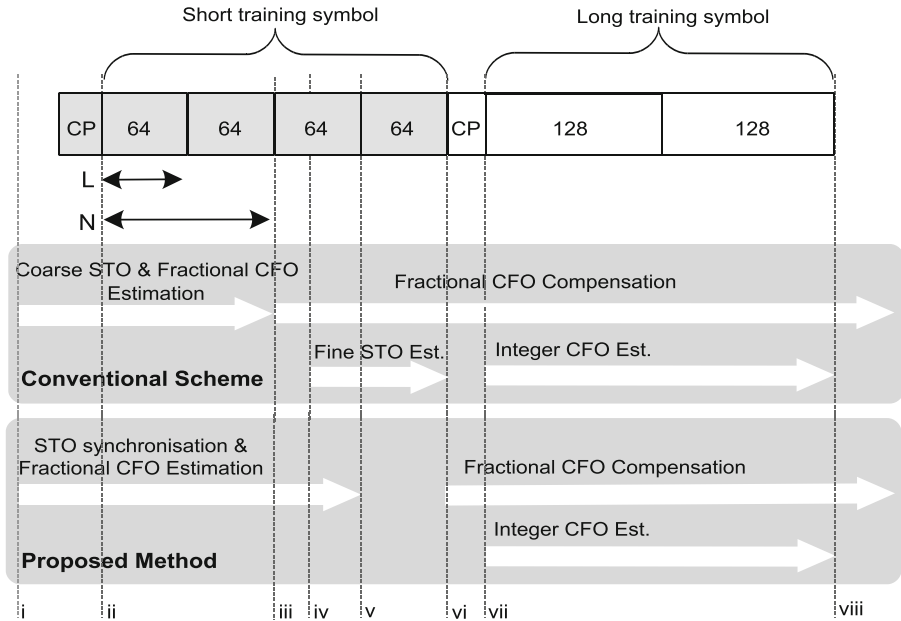


Fig. 2 The IEEE 802.16 preamble format (*top*), the conventional synchronization sequence (*middle*), and proposed reformation of the sequence (*bottom*)

of the amplitude of received samples, it uses fewer resources than approaches such as [10] that rely on complex correlation. In addition, the peak value of R' is used for time synchronization. To further reduce complexity compared with existing approaches, the computation of R' will also be evaluated when implemented using a multiplierless correlator. Performance will be evaluated with respect to both computation word size and correlation window length.

2.3 Proposed Synchronization Flow

Apart from a new metric, we also propose altering the conventional sequence and scope of STO and CFO operations in general. This is illustrated in Fig. 2 where the IEEE 802.16 preamble structure is drawn along the top. In the middle is the conventional computation flow, with the proposed method shown below. As can be seen, in the conventional scheme, the coarse STO and fractional CFO estimation are performed during the first two periods of the short training symbol from (i) to (iii), and then the estimated fractional CFO is used for compensation. The fine STO is estimated during the last period of the short training symbol from (iv) to (vi). By contrast, the proposed method performs STO synchronization and fractional CFO estimation during the first three periods of the short training symbol from (i) to (v), with the start of frame being detected at (ii). Then, the fractional CFO compensation is performed for subsequent symbols.

The method used for synchronization based upon these metrics is as follows: First, frame detection is performed by comparing P' to R' with a threshold ζ :

$$|P'(d)| > \zeta \times R'(d). \quad (5)$$

In order to increase stability, a frame is detected when condition (5) is met for n consecutive samples (with $n = 4$ used throughout this paper). The threshold is different for each channel and needs to be determined empirically by simulation, in common with many other systems, such as [10]. If the threshold is too small, noise may cause a failure of detection. Otherwise, if it is too large, the frame detection may be missed. Assuming the channel characteristics are known, the threshold can be selected from a lookup table in a practical system. After a frame is detected, the starting point of the short preamble is found by searching for the peaks in $R'(d)$. As can be seen in Fig. 1, two peaks in $R'(d)$ will bracket the transition to the plateau in $P'(d)$ shown at sample indices of -31 and 33 , respectively. The second peak is used to determine the starting point for the DFT window. This is accomplished for the maximum values of $R'(d) + R'(d - L)$ over the next L samples after frame detection. Secondly, fractional CFO estimation and correction is based on the $P'(d)$ metric and phase rotation as in other work [5, 7, 13, 17–19].

In this proposed method, the fractional CFO is estimated within the preamble to ensure that the correct angle of $P'(d)$ is then available for use in the estimation process. Thus, the estimated fractional CFO will tend to be more accurate as a result. Moreover, this method separates the length of the preamble period, L , from the length of received samples for estimation, C , and as a result, L can be shortened to extend the range of fractional CFO, while C can be lengthened to increase robustness to channel noise. For the IEEE 802.16 preamble, C can be in the range of L to $3L$ to improve overall synchronization precision, but clearly entails more computation when extended. In Sect. 3, which investigates the performance of the proposed method, a suitable value of C is chosen in order to obtain a good trade-off between accuracy and hardware cost.

3 Simulation

We evaluate the performance of the proposed synchronization method, applied to the IEEE 802.16-2009 downlink preamble, in MATLAB under both AWGN channels and more realistic SUI channels that are widely used in the literature [9, 10].

The Stanford University Interim (SUI) channel model [4] is used to simulate a frequency selective channel and takes into account many wireless channel effects including delay spread, Doppler spread, phase noise, and channel interference. In addition, the value of metrics computed in MATLAB is later verified with corresponding values from the FPGA simulation, to ensure practical functional equivalence.

In total, 100,000 OFDM frames, preceded by noise with randomly seeded AWGN and followed by preamble and data symbols, are used to evaluate synchronization performance for each method. The proposed method is compared to the state-of-the-art methods, in terms of accuracy of both time synchronization and fractional CFO estimation. The performance of STO estimation is measured in terms of failure

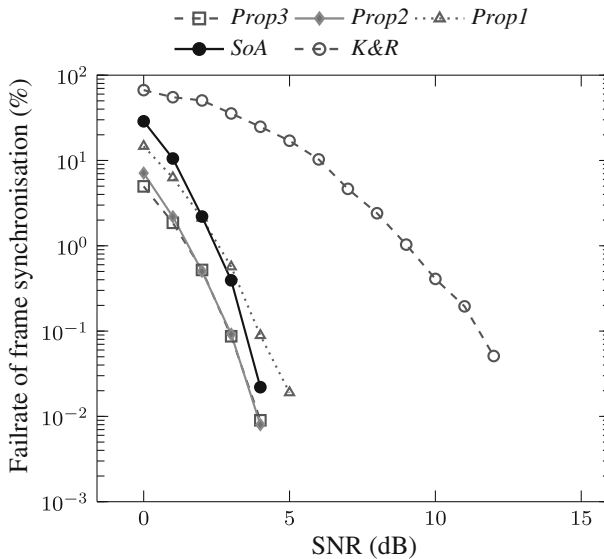


Fig. 3 Performance of time synchronization in AWGN channels with a frequency offset of 0.5 time subcarrier spacing

rate (%), and the accuracy of CFO estimation is evaluated in terms of mean square error (MSE). We separately evaluate the robustness of time synchronization against large CFO for each method. Three versions of the proposed approach are constructed by varying the length, C , of received samples for estimation based on (3). These are investigated to determine the trade-off between accuracy and computational cost, with C defined as follows in each case:

- Prop 1: $C = L$
- Prop 2: $C = 2L$
- Prop 3: $C = 3L$

These are compared to the state-of-the-art method (denoted as *SoA*) [11, 17] and the method of Kishore and Reddy [10] (denoted as *K&R*) for a number of evaluation scenarios. First, the performance of each method is found for AWGN channels beginning with CFO = 0.5, then AWGN channels with CFO varying from -10 to $+10$ times carrier spacing, then SUI1, and finally SUI2 channels.

3.1 Performance in AWGN

Figures 3 and 4 plot the performance results of STO and CFO estimation in AWGN with a frequency offset of 0.5 subcarrier spacings, respectively. *SoA* and the proposed methods have much better performance than *K&R* in these tests, achieving perfect synchronization when SNR exceeds 5 dB. *Prop1*'s STO estimation has slightly better accuracy with SNR below 3 dB but is worse at higher SNR than *SoA*. Increasing the length of received samples for estimation, i.e., setting $C = 2L$, allows *Prop2* to

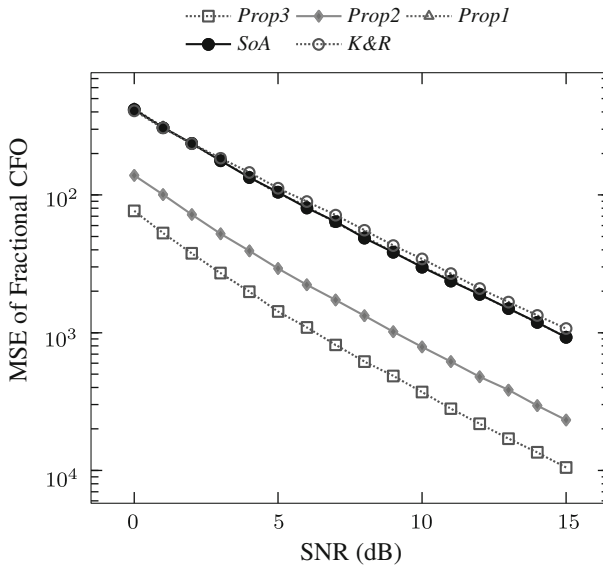


Fig. 4 Performance of fractional frequency offset estimation in AWGN channels

obtain a remarkable improvement in STO estimation, clearly better than the estimation achievable by *SoA*. *Prop3*, with $C = 3L$, demonstrates decreasing gains: it is not able to enhance accuracy as much as *Prop2*, despite a considerable hardware cost incurred when increasing C . In addition, the CFO of *Prop3* and *Prop2* achieves significant improvement compared with the other methods in Fig. 4, while the accuracy of *Prop1* and *SoA* is identical (the curve for *Prop1* is hidden behind the curve for *SoA* as a result).

The gap between *Prop1* and *Prop2* is much larger than that between *Prop2* and *Prop3*, again demonstrating decreasing gains as C is extended. Thus, increasing C from L to $2L$ is a more effective improvement than extending C from $2L$ to $3L$. The accuracy of CFO estimation in *Prop2* is improved by about 5 dB in comparison with *SoA* and *K&R*. This improvement is as a result of the increased length of received samples for estimation, C . These results show the proposed method to be competitive with state-of-the-art methods.

3.2 Performance in Fading Channels

Figures 5 and 6 present the performance results of STO estimation in SUI1 and SUI2 channels, respectively. The proposed methods are seen to achieve much better accuracy than the *K&R* method. Compared with *SoA*, Fig. 5 reveals that the estimation of *Prop1*, *Prop2*, and *Prop3* is more accurate when SNR is below 3 dB. However, for higher SNRs, the accuracy of *SoA* is slightly better than that of the proposed method. Increasing the length of received samples for estimation achieves an improvement when SNR is below about 5 dB, although the results for *Prop1*, *Prop2*, and *Prop3* saturate and become almost identical at higher SNRs, as does *K&R*.

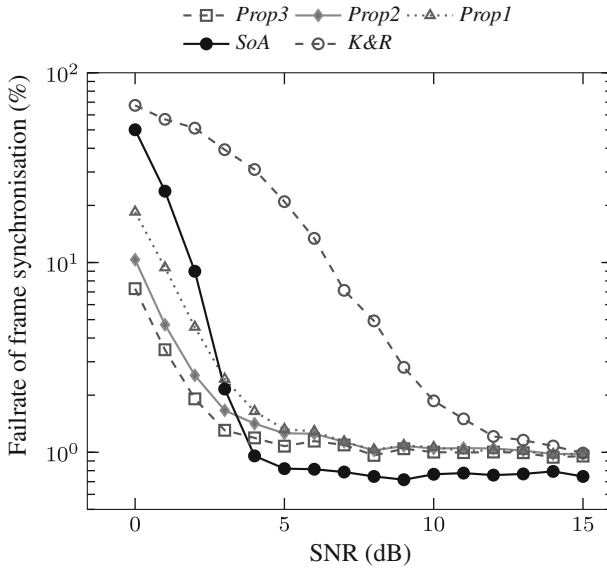


Fig. 5 Frame synchronization performance of various methods in an SUI1 channel with respect to SNR

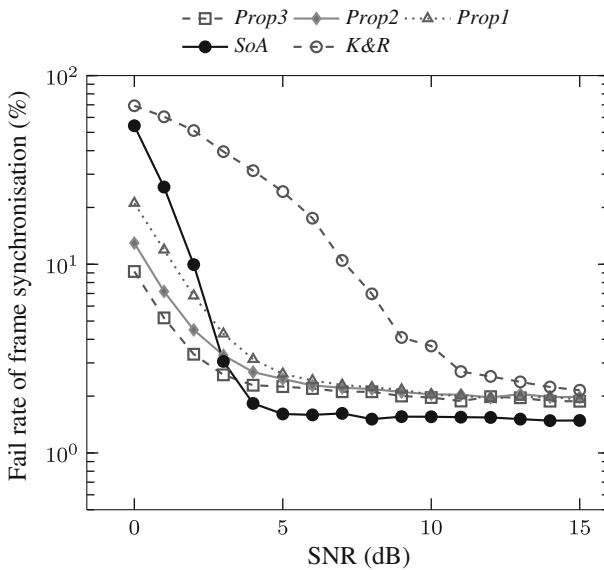


Fig. 6 Frame synchronization performance of various methods in an SUI2 channel with respect to SNR

3.3 Performance with Large Frequency Offset

The proposed method is designed to work even with large frequency offsets. Figure 7 explores performance over 100,000 tests where the frequency offset is chosen randomly (with uniform distribution) from -10 to $+10$ times the subcarrier spacing for

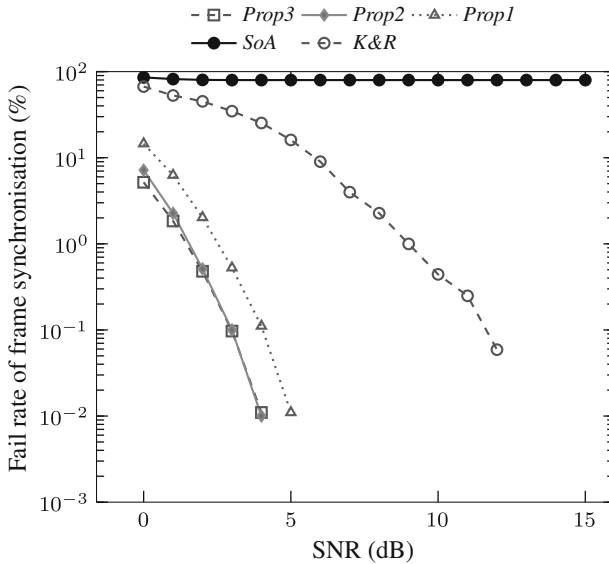


Fig. 7 Performance of frame synchronization in an AWGN channel with uniform random frequency offset varying from -10 to 10 times carrier spacing, with respect to SNR

each test, in an AWGN channel. This experiment is specifically designed to investigate the robustness of STO estimation in large CFO conditions and shows that the proposed method still maintains good performance. *K&R* exhibits some accuracy degradation, however, all outperform *SoA*. The proposed method is therefore seen to offer robustness of STO estimation against large CFO. This robustness against large CFO is because the proposed metric is computed on magnitude values that are insensitive to phase errors. CFO estimation is not evaluated here because large CFO estimation requires an integer CFO estimator that is not within the scope of this paper.

In summary, simulation results show that *Prop3* and *Prop2* have better STO and CFO estimation accuracy compared with other methods. *Prop3* enjoys just a small improvement in terms of STO estimation compared with *Prop2* but this improvement incurs a significant hardware cost because the length of received samples for estimation must increase from $2L$ to $3L$.

Given the results described in this section, *Prop2* is selected as an implementation candidate in Sect. 4, where the trade-off between accuracy and hardware cost is explored in more detail.

4 Hardware Implementation

This section discusses the implementation of and investigates the hardware optimization of the proposed method in terms of word size. The target FPGA is a low-power Xilinx Spartan-6 XC6SLX45 device, with ISE 13.2 used to evaluate both hardware resource and power consumption. The results illustrate the trade-off between hardware consumption and the accuracy of the proposed method. Section 3 already revealed

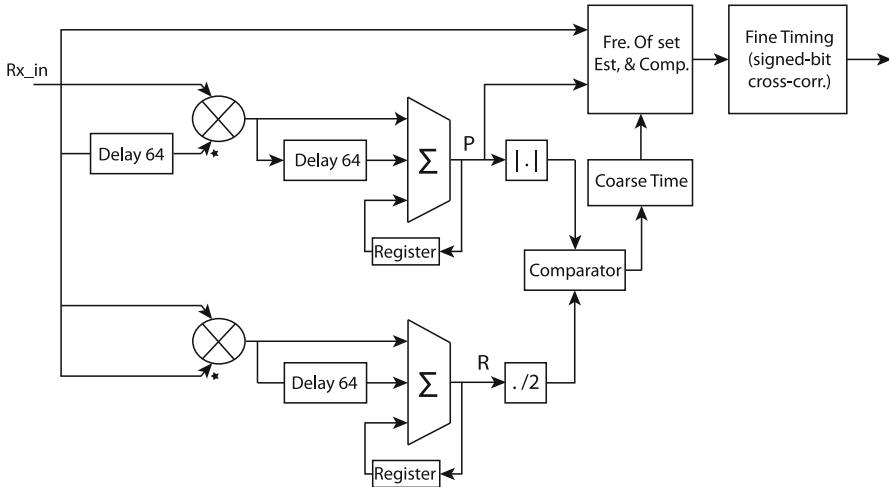


Fig. 8 Architecture of the conventional synchronization FPGA implementation

that the performance of *K&R* is generally worse than the other methods; moreover, it requires many complex multiply operations to compute cross-correlation for the *P* Metric. For this reason, *K&R* is not implemented, only the proposed method and *SoA* are compared here, in terms of hardware resources, power consumption, and accuracy. As mentioned in Sect. 3, we set $C = 2L$ since *Prop2* consistently showed performance close to *Prop3* and significantly better than *Prop1*. It therefore offers a good balance between increased hardware cost and performance.

4.1 Implementation of Conventional Synchronizer

First, let us consider a conventional synchronizer. We have implemented this as shown in Fig. 8, following the efficient implementation methods presented in [5, 11, 13, 21]. The *P* and *R* timing metrics in (2) are computed using delay and summation, while a very efficient signed bit multiplier [19] is used to significantly reduce the computational overhead of the cross-correlation for fine timing synchronization.

We assume a 16-bit two's complement fixed point representation with 15 fractional bits (i.e., Q1.15 format in Q-notation [14]). The autocorrelation and squared amplitude of received samples are computed with a complex multiplier IP core that uses DSP slices. Results are scaled to Q1.15 to reduce resource usage in subsequent pipeline stages. Since this synchronizer is for the IEEE 802.16 preamble, the length of the delay is 64 samples. Each element is scaled to be less than 1, guaranteeing that the final summation result is less than 64, needing just 7 bits for representation in two's complement. Hence, the *P* and *R* values are represented in Q7.15 format. For the $|P|$ metric, the autocorrelation uses a complex multiplier IP core to multiply the current received sample and the 64th delayed sample. The magnitude of the *P* metric is approximated to reduce hardware complexity as per [11]. For the *R* metric, another complex multiplier is used to compute the squared magnitude of the received sample.

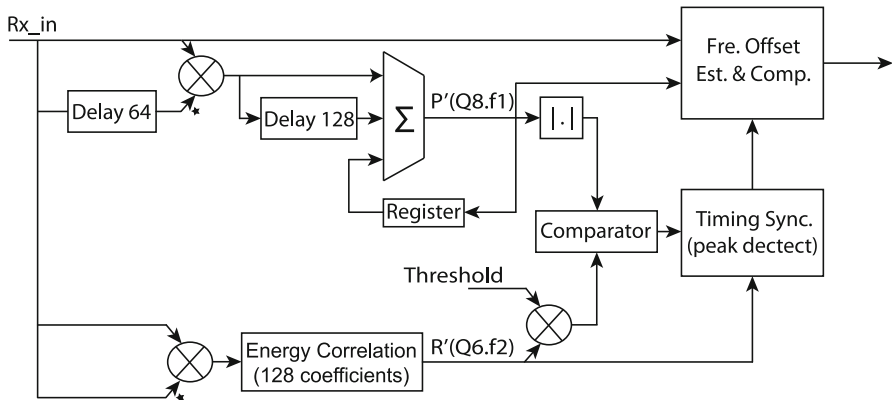


Fig. 9 Architecture for the proposed synchronization method implemented on FPGA

The threshold is commonly chosen to be 0.5, which can be implemented using a right shift by 1 bit [9] instead of using a multiplier. After the frame is detected, the P metric is used to estimate and correct the fractional CFO. A CORDIC IP core is used to determine the phase of the P metric and to derive the estimated fractional CFO. The received samples are then compensated using phase accumulation and phase rotation. These compensated samples are now used to determine fine timing synchronization.

4.2 Implementation of Proposed Synchronizer

The architecture for our proposed method is shown in Fig. 9.

The format of received samples is similar to the conventional design. The R' metric is determined using an energy correlator as illustrated in Fig. 10. The number of samples used to compute it is 128 and since the 128 samples of the short preamble are arranged in two identical spans of 64 samples each, the correlator only needs 64 taps. Equation (6) shows the derived equations of this optimization:

$$\begin{aligned}
 R'(z) &= I(z)A_{127} + I(z)z^{-1}A_{126} + \dots + I(z)z^{-63}A_{64} \\
 &\quad + I(z)z^{-64}A_{63} + \dots + I(z)z^{-127}A_0, \\
 &= I(z)(1 + z^{-64})A_{63} + z^{-1}(I(z)(1 + z^{-64})A_{62} \\
 &\quad + z^{-1}(\dots + z^{-1}I(z)(1 + z^{-64})A_0)),
 \end{aligned}
 \tag{6}$$

where I is the squared amplitude of the received sample, and A_n denotes the normalized squared amplitude of known preambles. Following this, a multiplierless correlator, as described in detail in [16], is used to compute the output, shown in block diagram form in Fig. 10.

The values of A_n are quantized to 0.5, and a shift/multiplex operation replaces the multiplication. R' is always positive and smaller than twice the sum of all A_n elements, i.e., 63. So, the R' metric requires just 6 bits to represent its integer part.

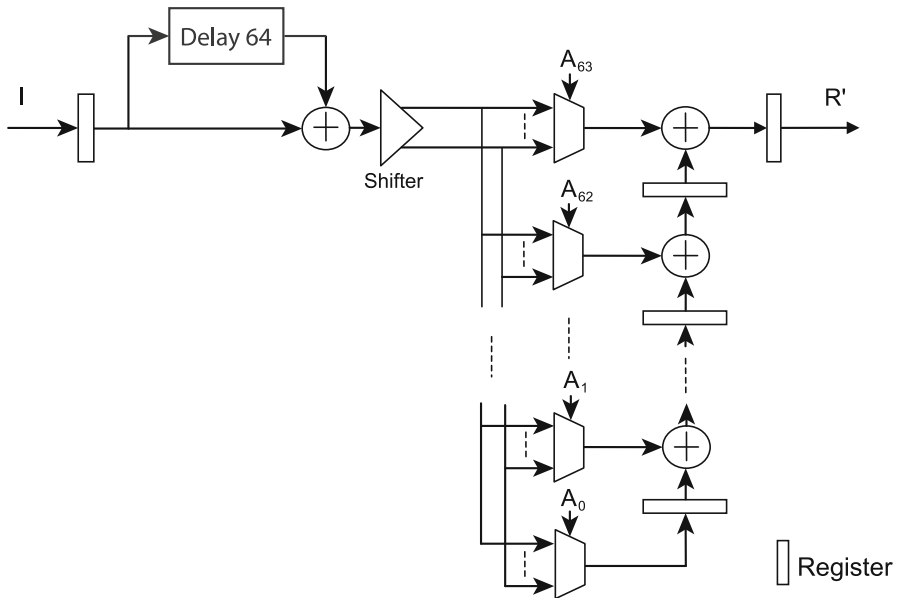


Fig. 10 Implementation of energy correlator on FPGA

The P' metric is computed in an identical way to the conventional one, but the number of samples used in the computation is 128 instead of 64 since $C = 2L$. So, the moving summation uses a delay buffer of 128 samples, and the result value now requires 8 bits to represent the integer part.

4.3 Effect of Reduced Precision

Let f_1 and f_2 be the number of bits representing the fractional part for computing P' and R' , respectively. Thus, P' has fixed point format $Q8.f_1$ and R' has fixed point format $Q6.f_2$. The effects of reducing the number of bits used to represent these fractional components of P' and R' will now be investigated, with the aim of optimizing the reduced precision against hardware savings.

Recall the results of CFO estimation in Sect. 3 where *Prop2* exhibited a significant improvement in CFO estimation accuracy compared with *SoA* thanks to an increase in the evaluation window size obtained by setting $C = 2L$. This requires more multiplications, leading to increased hardware cost to compute the P' metric. Reducing f_1 allows a reduction in this extra hardware cost by making each individual computation simpler. The question is to determine how much reduction in f_1 can be sustained without losing the performance advantage enjoyed by *Prop2*.

To understand precisely how f_1 reduction can save hardware, Table 1 details the hardware resource required for implementing five representative word sizes. Meanwhile, Fig. 11 plots CFO performance curves for the corresponding sizes of f_1 . It is clear from the graph that all degraded precision computations perform well—even the

Table 1 Resources required for computing P' on FPGA with different word lengths, $Q1.f1$

$f1$	FF	LUT	BRAM	DSP
<i>Prop-15b</i>	304	427	96	3
<i>Prop-7b</i>	240	323	64	3
<i>Prop-6b</i>	232	311	60	3
<i>Prop-5b</i>	224	297	56	3
<i>Prop-4b</i>	216	269	52	3

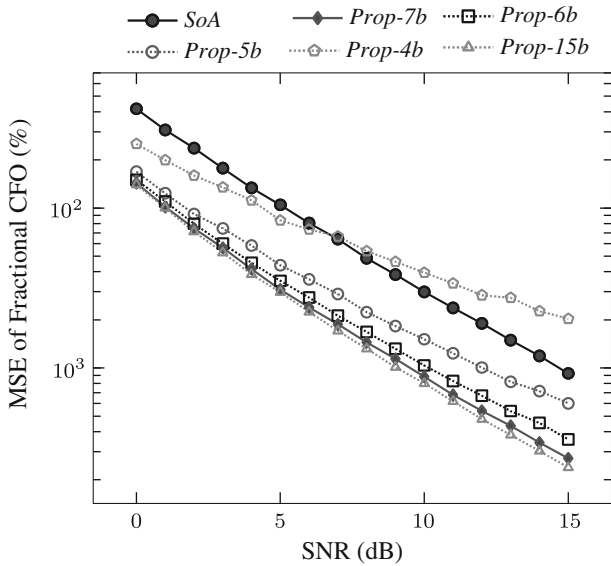


Fig. 11 Performance of CFO estimation in an AWGN channel against SNR, with different numbers of fractional bits used in the computation of P'

lowest Q1.4 precision computation (*Prop-4b*) can outperform *SoA* below about 7 dB. The optimal choice for this range of SNRs is probably $f1 = 7$ bits (*Prop-7b*) which suffers just a slight decrease in accuracy compared with the ‘full length’ 15-bit version (*Prop-15b*). This allows a reduction of 21, 24, and 33 % in the number of flipflops (FF), lookup tables (LUT), and BRAM blocks, respectively. Moreover, *Prop-7b* still achieves excellent performance when compared with the state-of-the-art method, *SoA*.

Similarly, the optimized tradeoff between the accuracy of STO estimation and hardware usage for computation of the R' metric is obtained based on reducing $f2$. In this case, Table 2 reveals the corresponding reduction achieved in computation resources and Fig. 12 plots the frame synchronization fail rate with SNR for several values of $f2$. The performance of the proposed method with $f2 = 6$ bit, *Prop-6b*, can be seen to be almost identical to the ‘full length’ computation using 15 fractional bits, *Prop-15b*. Overall, *Prop-6b* achieves much more accurate estimation compared with the state-of-the-art method, *SoA*. Reducing $f2$ to 6 bits allows a reduction of 41, 42, and 56 % in the number of FFs, LUTs, and BRAM blocks.

Table 2 Resources required for computing R' on FPGA with different word lengths, $Q1.f2$

$f2$	FF	LUT	BRAM	DSP
<i>Prop-15b</i>	1,404	1,017	16	2
<i>Prop-7 b</i>	884	633	8	2
<i>Prop-6 b</i>	818	589	7	2
<i>Prop-5 b</i>	753	537	6	2
<i>Prop-4 b</i>	689	504	5	2

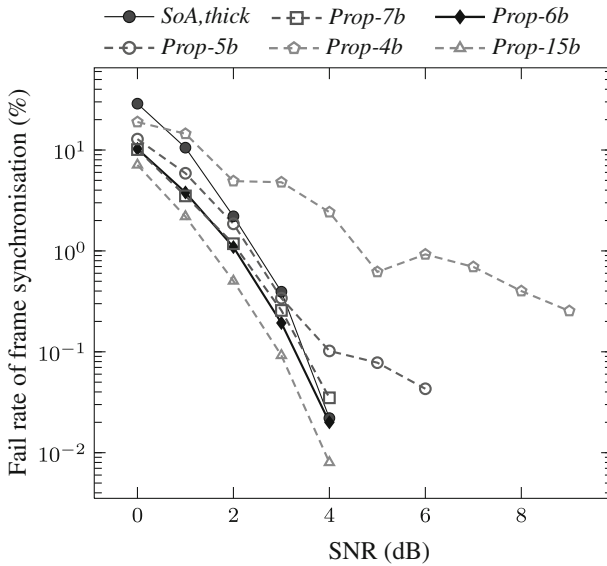


Fig. 12 Performance of frame synchronization in an AWGN channel against SNR, with different numbers of fractional bits used in the computation of R'

4.4 Optimized Alternatives

The preceding results are now used to define four alternative implementations of the proposed method to compare against the state-of-the-art method, *SoA* which uses full length $Q1.15$ arithmetic. These alternatives are namely:

- *Prop-A1*: a non-optimized instance of the proposed method with both $f1$ and $f2$ set to 15.
- *Prop-A2*: only P' is optimized with $f1 = 7$ while $f2$ remains set to 15.
- *Prop-A3*: only R' is optimized with $f2 = 6$ while $f1$ remains set to 15.
- *Prop-A4*: both P' and R' are optimized by setting $f1 = 7$ and $f2 = 6$.

Table 3 reports the overall hardware resources required for these instances of the synchronizer, as well as detailing the power consumption of each. It should be noted that the CFO estimation and frame synchronization performance of these instances can be seen by choosing the corresponding word length from plots of P' & R' in Figs. 11 and 12, respectively. In other words, all instances have been simulated and reported

Table 3 Total resources consumed by a full word length implementation of *SoA* and four reduced complexity instances of the proposed method

	Slices	BRAM	DSP	Qpwr	Dpwr	Fre.
<i>SoA</i>	930	112	13	37	41	121
<i>Prop-A1</i>	1000	118	14	37	43	133
<i>Prop-A2</i>	923	86	14	37	39	142
<i>Prop-A3</i>	869	109	14	37	38	137
<i>Prop-A4</i>	777	77	14	37	35	142

Dynamic (Dpwr) and quiescent power (Qpwr) consumption are reported in mA. Maximum frequency is reported in MHz

Table 4 Detailed resource comparison between two synchronization methods

Function	FF	LUT	BRAM	DSP
<i>SoA</i>				
$ P $ metric	303	427	64	3
R metric	168	186	16	2
CFO comp	1,478	1517	0	8
Coarse time	7	33	0	0
Fine time	942	1,009	32	0
Total	2,898	3,172	112	13
<i>Prop-A4</i>				
$ P' $ metric	240	323	64	3
R' metric	818	600	7	2
CFO comp	1,467	1,515	0	8
Time sync	66	98	6	1
Total	2,591	2,536	77	14

in the previous plots. From the table, it is evident that reducing word length can yield a significant reduction in both hardware requirement and power consumption. The fully optimized alternative, *Prop-A4*, achieves a reduction of 16.4, 31.2, and 14.6% in the number of occupied slices, BRAMs, and in dynamic power consumption, when compared with *SoA*. The maximum frequencies of the *SoA* and proposed method implementations are also reported. The required frequency for baseband processing in IEEE 802.16 ranges from 5.6 to 22.4 MHz, and this requirement is easily met by all tested implementations.

Table 4 details the comparison between *SoA* and *Prop-A4* in terms of their constituent building block resources (where the function names in this table correspond to the block diagrams of Figs. 8, 9). The function named ‘CFO comp’, which performs frequency offset estimation and compensation, clearly consumes the largest amount of hardware in both methods. The R' metric computation in *Prop-A4* uses more hardware than the computation of R in *SoA*. However, fine timing estimation, ‘Fine time,’ takes a large proportion of the total hardware cost in *SoA* while the alternative in the proposed method (timing synchronization known as ‘Time sync’), requires much less hardware.

5 Conclusion

This paper has presented and evaluated a novel algorithm for OFDM synchronization and discussed its implementation on FPGA, with particular relevance to low-power devices. The proposed method is designed to improve CFO estimation accuracy and to be much more robust to large CFO than common implementations, yet to be suitable for efficient hardware implementation. Simulations were performed in MATLAB to investigate the performance of three versions of the proposed method applied to the IEEE 802.16-2009 downlink preamble, which showed that the method compares very well to the two existing approaches. Specifically, the STO estimation results of the proposed method demonstrate an improved accuracy and robustness to large CFO compared with the conventional methods. Similarly, the CFO estimation ability of the proposed method achieves a significant improvement.

The proposed method and current state-of-the-art approaches were implemented on FPGA to evaluate and compare the hardware resources usage and power consumption. Several trade-offs were explored in terms of reduced computation word length and number of computations, with interesting results. As well as achieving excellent synchronization performance, the proposed method saves resources compared with the state-of-the-art-system and achieves an overall power saving of 14.6 %.

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