Multiple-Symbol Detection Scheme for IEEE 802.15.4c MPSK Receivers over Slow Rayleigh Fading Channels

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Although the full multiple-symbol detection (MSD) for IEEE 802.15.4c multiple phase shift keying (MPSK) receivers gives much better performance than the symbol-by-symbol detection (SBSD), its implementation complexity is extremely heavy. We propose a simple MSD scheme based on two implementation-friendly but powerful strategies. First, we find the best and second-best decisions in each symbol position with the standard SBSD procedure, and the global best decision is frozen. Second, for the remaining symbol positions, only the best and second-best symbol decisions, not all the candidates, are jointly searched by the standard MSD procedure. The simulation results indicate that the packet error rate (PER) performance of the simplified MSD scheme is almost the same as that of the full scheme. In particular, at PER of $1 \times 10^{-3}$, no more than 0.2 dB performance gap is observed if we just increase the observation window length $N$ to 2. However, the number of decision metrics needed to be calculated is reduced from 256 to 2. Thus, much balance gain between implementation complexity and detection performance is achieved.

1. Introduction

With the widespread application of new information and communication technologies such as the Internet of Things (IoT), cloud computing, and big data, smart cities have developed rapidly in recent years. They have penetrated into all aspects of people’s lives and greatly meet the modern people’s pursuit of convenient, fast, and high-quality life [1–7]. Reliable and effective transmission of the sensing data is obviously important for the construction of the new smart city [8]. The IEEE 802.15.4c protocol provides the physical layer specification of the low-power short-distance IoT for China [9, 10]. The multiple phase shift keying (MPSK) is provided in IEEE 802.15.4c. This mainly follows from the fact that MPSK modulation is the most able to provide high reliability as well as data rate for sensing data transmission.

Therefore, it is important to study robust detection technology of MPSK signal in line with the characteristics of wireless IoT. This paper focuses on the multiple-symbol detection (MSD) of IEEE 802.15.4c MPSK receiver.

Although the MSD scheme has excellent detection performance, its implementation complexity increases exponentially with the increase of the observation window length [11, 12]. In recent years, many concentrations have been achieved on complexity reduction of MSD. Stephen et al. studied the maximum likelihood detection (MLD) based on information symbol blocks. The corresponding block signal is used to limit only a part of the possible signal decisions, which will reduce the complexity of the receiver. However, there is a partial performance loss [13]. LoRici proposed a suboptimal receiver based on Viterbi algorithm. The complexity of the receiver increases in polynomial form...
of $M$. However, the performance of the algorithm is related to the memory length $L$ of continuous phase modulation signal. For continuous-phase frequency-shift keying (CPFSK) signal ($L = 1$), its detection performance is seriously degraded [14, 15]. Several low-complexity MSD algorithms are also proposed by Fischer and Wang Xin, but their performance is far behind that of the traditional MSD algorithm [16, 17].

In this work, we propose a simple MSD scheme for IEEE 802.15.4c MPSK receivers. Unlike the traditional receivers that were equipped with full MSD scheme with high complexity to achieve the best possible reliability, we pay our full attention to the simple design to balance the complexity and reliability. We summarize our main contributions as follows:

(i) The optimal MSD scheme for IEEE 802.15.4c MPSK receivers based on the maximum likelihood criterion can give excellent results in the case of both slow fading and pure additive white Gaussian noise (AWGN) channels. However, the implementation is relatively complex and unachievable for IEEE 802.15.4c MPSK receivers. As an implementation achievable benchmark, a full MSD scheme based on compensation is proposed.

(ii) As for the proposed full MSD scheme, more than two-hundred-decision statistic should be calculated before making final decision even if we set the observation window length $N$ to 2. Thus, we propose a new MSD algorithm, which greatly simplifies the full scheme.

(iii) In order to verify the desirable properties we obtained from this simple scheme, the characteristics of the receiver are studied from many aspects with extensive simulations.

The rest of this paper is organized as follows: Section 2 focuses on the signal model under the slow fading Rayleigh channel. Section 3 describes the full MSD scheme, and Section 4 introduces the proposed simplified MSD scheme. Section 5 concentrates on frequency offset estimation. The simulation results are discussed in Section 6. Finally, some conclusions and future work are provided in Section 7.

2. System Model

According to the IEEE 802.15.4c protocol [18], the specific data modulation process for the MPSK physical layer is shown in Figure 1. From the binary data of the physical layer protocol data unit (PPDU), in each symbol period, four chips in the spreading sequence are MPSK-modulated onto the carrier. For more of 16 orthogonal spreading sequences. The chips in the spreading sequence form a symbol, which is used to select one protocol data unit (PPDU), in each symbol period, four chips in the spreading sequence are MPSK-modulated onto the carrier. For more of 16 orthogonal spreading sequences. Moreover, several low-complexity MSD algorithms are also proposed by Fischer and Wang Xin, but their performance is far behind that of the traditional MSD algorithm [16, 17].

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2. System Model

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Ideal carrier synchronization is assumed at the receiver. Specifically, for the $x$th symbol $E[x]$, the received complex baseband chip sequence can be expressed as

$$r_{x,m} = h_{x,m}^* s_{y,m} e^{i(\omega_{x,m} m T_c + \theta_{x,m})} + \eta_{x,m}, \quad 1 \leq m \leq M,$$  \hspace{1cm} (1)

where $h_{x,m}$ represents multiplicative fading, $s_{y,m}$ is the $m$th chip of the $y$th pseudorandom (PN) sequence $s_y$, and Table 1 shows the detailed correspondence. $\omega_{x,m} = 2\pi f_{x,m}$ represents the carrier frequency offset (CFO) in radians, and $f_{x,m}$ represents the residual CFO in Hz. $\theta_{x,m}$ represents the carrier phase offset (CPO) in radians, and $T_c$ represents the spreading chip period. $\eta_{x,m}$ is a discrete, cyclic symmetric, complex Gaussian random variable with zero mean and variance $\sigma^2_{x,m}$, and $M = 16$ represents the length of the PN sequence [18].

We assume that a piecewise constant approximation is made to the multiplicative fading, CFO, and CPO [20]. That is, $h_{x,m} = h$, $\omega_{x,m} = \omega$, and $\theta_{x,m} = \theta$. In addition, the receiver does not have any prior information about the CFO; that is to say, the uniform distribution in the interval $(-\pi, \pi)$ is assigned to $\theta$. The normalized complex Gaussian process $h$ follows Rayleigh distribution; that is, the mean $\bar{h} = 0$. The CFO $f$ follows a symmetrical triangular distribution.

3. The Full MSD Scheme

Following the idea in [21], we can easily develop the optimal MSD scheme for IEEE 802.15.4c MPSK receivers based on MLD. However, the implementation complexity is extremely heavy as shown in [22], which limits its application in smart cities. Here, we consider a heuristic configuration. The specific detection process is as follows.

First, the baseband chip sample after carrier frequency offset effect (CFOE) compensation can be expressed as

$$r'_{x,m} = r_{x,m} e^{-j m \bar{\theta}},$$  \hspace{1cm} (2)

where $\bar{\theta} = \bar{\omega} T_c$ denotes the estimated CFOE. The estimation of CFOE should be carefully developed and will be described in detail in Section 5. Please note that we assume that the effect of redundant parameter $\omega T_c$ on $r_{x,m}$ is completely eliminated after compensation. In addition, the information is embedded in the carrier phase but not in the carrier amplitude. Therefore, there is no need to estimate and compensate for the multiplicative fading $h$ even if serious fading of the received signal strength may be exhibited.

Secondly, we divide the whole compensated chip sequence into block, and each block contains $N$ symbols. The detection metric for the $i$th block can be then expressed as [23]

$$U_x = \left| \sum_{i=1}^{N} \sum_{m=1}^{16} r'_{x,m} s_{y,m}^* \right|^2, \quad 1 \leq i, 1 \leq y \leq 16,$$  \hspace{1cm} (3)

where

$$U_x = \left\{ U_{(i-1)N+1,1}, U_{(i-1)N+1,2}, \ldots, U_{(i-1)N+1,16}, U_{(i-1)N+2,1}, U_{(i-1)N+2,2}, \ldots, U_{(i-1)N+2,16}, U_{iN+1,1}, U_{iN+1,2}, \ldots, U_{iN+1,16} \right\}$$  \hspace{1cm} (4)

and $*$ represents complex conjugate operation. Note that, for $N = 1$, (3) reduces to the symbol-by-symbol detection (SBSD) scheme. From (3), we can also see that the multiplicative fading $h$ has no effect on the final decision, and there is no need to estimate and compensate for the fading coefficient $h$.

Finally, the decision rule can be expressed as follows:

$$\hat{U}[x] = \arg \max \{U_x\}.$$  \hspace{1cm} (5)
After demapping, we can obtain the final detection result. This detection scheme is based on [23] but is different from [23]. The signal model in [23] only considers phase offset. In this work, we further considered CPO, spread spectrum, and slow Rayleigh channel. Therefore, we summarize the detailed process of the complete MSD program.

As shown in (3), based on an exhaustive search, 256 detection metrics need to be calculated for the full MSD even if we set the observation window length $N$ to 2. This is clearly complexity-heavy. In order to make MSD easy for hardware implementation, we consider two simple strategies, which parallels Wilson’s approach in [15]. First, we find the best and second-best decisions in each symbol position with the standard SBSD procedure characterized by (3) and freeze the decision result corresponding to the most reliable symbol position, which is achieved by searching all the local best metrics. For the remaining symbol position, the number of symbols to be searched is truncated. That is to say, only the symbols corresponding to the local best and second-best metrics are considered as the candidates. In this context, for observation window length $N = 2$, we have reduced the number of the metrics given in (3) to be calculated from 256 to 2. However, the simulation results in Section 6 show that the performance loss is very small. The specific implementation process is detailed as follows.

For the $i$th block, the decision metric for each symbol position is first calculated as

$$V_{x,y} = |w_{x,y}|^2, \quad 1 \leq y \leq 16.$$  \hspace{1cm} (6)

Here, $w_{x,y} = \sum_{m=1}^{16} r_{x,m}^* s_{y,m}$, which is the complex cross-correlation function.

Secondly, the best and the second-best metrics for the $n$th symbol in the $i$th block can be given as follows:

$$V_{N(i-1)+n,y_1} = \arg \max_{1 \leq y_1 \leq 16} \left\{ V_{N(i-1)+n,y_1} \right\}, \quad i \geq 1, n = 1, 2, \ldots, N,$$

$$V_{N(i-1)+n,y_2} = \arg \max_{1 \leq y_2 \leq 16, y_2 \neq y_1} \left\{ V_{N(i-1)+n,y_2} \right\}, \quad i \geq 1, n = 1, 2, \ldots, N,$$  \hspace{1cm} (7)

where $y_1$ and $y_2$, respectively, represent the estimated value of the index for the PN sequence corresponding to the best and second-best metrics of the $n$th symbol. For example, we can see that, as shown in Figure 2, a compensated baseband chip sequence $r_{i,m}$ passes through decision block 1 to generate decision set $\{V_{1,1}, V_{1,2}, \ldots, V_{1,16}\}$, and the best and second-best metrics in the decision set are recorded as $V_{1,y_1}$ and $V_{1,y_2}$, respectively.

Furthermore, find the global best metric, and freeze the detection result:

$$\text{find } \tilde{n} \text{ and } s_{\tilde{n}} \text{ if } V_{N(i-1)+n,\tilde{n}} \text{ is maximum,}$$  \hspace{1cm} (8)

that is, let the detection result of the $\tilde{n}$th symbol be $s_{\tilde{n}}$.

Figure 2 gives the implementation structure.

Finally, the data in the remaining $N - 1$ symbol periods are jointly determined as follows:

$$\text{find } \left\{ s_{\tilde{n}} \right\} \text{ for } n \neq \tilde{n} \text{ if } \left| w_{N(i-1)+n,\tilde{n}} \sum_{n \neq \tilde{n}} w_{N(i-1)+n,\tilde{n}} \right|^2, \quad k \in \{1, 2\} \text{ is maximum.}$$  \hspace{1cm} (9)
### Table 1: Symbol-to-chip mapping rule for IEEE 802.15.4c MPSK physical layer.

| $S_{y,0}$ | $S_{y,1}$ | $S_{y,2}$ | $S_{y,3}$ | $S_{y,4}$ | $S_{y,5}$ | $S_{y,6}$ | $S_{y,7}$ | $S_{y,8}$ | $S_{y,9}$ | $S_{y,10}$ | $S_{y,11}$ | $S_{y,12}$ | $S_{y,13}$ | $S_{y,14}$ | $S_{y,15}$ | $S_{y,16}$ |
|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|
| $e^{0}$   | $e^{0}$   | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ | $e^{0}$ |
| $e^{0+}$  | $e^{0+}$  | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ | $e^{0+}$ |
| $e^{0-}$  | $e^{0-}$  | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ | $e^{0-}$ |

Note: The symbols $S_{y,0}$ to $S_{y,16}$ represent different symbols mapping to the chip. The mapping is designed for compatibility with the IEEE 802.15.4c standard, ensuring robust transmission and reception in the physical layer of the wireless communication system.
where $\tilde{n}$ and $\tilde{y}_1$ are given by (8). Figure 3 is a structural diagram of this joint decision.

Algorithm 1 introduces the detailed implementation step of proposed MSD scheme. For simple implementation, we only selected the most and second-most reliable symbols here. More metrics can also be involved, which, however, are complexity-intensive and not suitable for our purposes. In essence, when 16 metrics are selected, we arrive at the full MSD. Furthermore, the simulation results in Section 6.2 show that excellent performance has been exhibited even if we only equip the MSD scheme with the most and second-most reliable metrics.

5. Estimation Scheme

Clearly, the chip sample $r_{x,m}$ in (1) is dependent on the transmitted chip symbol $s_{x,m}$, but this dependence can be eliminated if we follow the property $s_{x,m}^* s_{x,m} = 1$:

$$Gr_{x,m} \triangleq r_{x,m}^* s_{x,m} = he^{i(mwT_c+\phi)} + \eta_{x,m} s_{x,m}, \quad 1 \leq x \leq P_1, 1 \leq m \leq M,$$

where $P_1$ is the length of the preamble, $1 \leq P_1 \leq P$, and $P = 8$ is the maximum length. $\eta_{x,m} s_{x,m}$ is statistically equivalent to $\eta_{x,m}$. In this context, our purpose is to estimate $\omega T_c$ based on the sample observations given in (10).

Within $N$ symbol intervals, the normalized autocorrelation function of samples is as follows:

$$Z(n) = \frac{1}{(P_1-n)L_1} \sum_{x=1}^{P_1} \sum_{m=2}^{L_1} (Gr_{x,m} Gr_{x,m-n}^*) = |h|^2 e^{i\mu T_c} + \eta_n,$$

where $L_1$ is the sample number of the preamble, and $2 \leq L_1 \leq M$. $\eta_n$ represents the integrated noise. $n$ denotes the
number of chip delays, and 1 ≤ n ≤ K. K represents the maximum chip-delay number.

Following the idea in [24], a simple estimation scheme without phase unwrapping can be expressed as follows:

\[ \hat{\varphi} = \hat{\omega}T_c = g(Q), \]  

(12)

where the quantization function \( g(Q) \) is

\[ g(Q) = \frac{2}{K+1} \arg(Q), \]  

(13)

where \( Q = \sum_{m=1}^{K} Z(n) \). The structure of this estimator is shown in Figure 4.
Table 2: Parameters used in simulations.

| Parameter                                      | Detailed description                                      |
|-----------------------------------------------|-----------------------------------------------------------|
| Channel condition                             | Slow fading or pure AWGN                                  |
| Power of the complex AWGN                     | 1/SNR                                                     |
| Power of Rayleigh fading channel              | Normalized                                                |
| Detection scheme                              | MSD                                                       |
| Compensation scheme                           | Precompensation                                           |
| Timing synchronization                        | Perfect                                                   |
| Data modulation                               | MPSK                                                      |
| Symbols                                       | 16-ary orthogonal                                         |
| PPDU payload length (bits)                    | 176                                                       |
| Spreading factor                              | 16                                                        |
| Chip rate (Mchip/s)                           | 1                                                         |
| Binary data rate (kb/s)                       | 250                                                       |
| Carrier frequency (MHz)                       | 786                                                       |
| CFO $f$ (ppm)                                 | Symmetrical triangular distribution in $(−80, 80)$        |
| CPO $\theta$ (rads)                          | Uniform distribution in $(−\pi, \pi)$                    |
| PN length $M$                                 | 16                                                        |
| Preamble length $P_1$                         | 8                                                         |
| Sample number $L_1$                           | 16                                                        |

Figure 4: The structure diagram of frequency offset estimator.
Figure 5: Continued.
Figure 5: In a pure AWGN channel, the effect of parameter $K$ on the performance of the full estimator in (16), $N = 2$. (a) BER performance; (b) SER performance; (c) PER performance.

Figure 6: Continued.
Figure 6: In a pure AWGN channel, the effect of parameter $K$ on the performance of the simplified estimator in (15), $N = 2$. (a) BER performance; (b) SER performance; (c) PER performance.

Figure 7: Continued.
Figure 7: Performance comparison of different receivers in pure AWGN channel. (a) BER performance; (b) SER performance; (c) PER performance.

Figure 8: Continued.
Figure 8: Performance comparison of different receivers in slow fading Rayleigh channel. (a) BER performance; (b) SER performance; (c) PER performance.

Figure 9: Continued.
Figure 9: Detection performance comparisons of the proposed scheme under various estimators versus CFO over pure AWGN channel, $N = 2$. (a) BER performance; (b) SER performance; (c) PER performance.
Figure 10: In the pure AWGN channel, the proposed scheme is compared with the detection performance of dynamic CPO under the full estimator in (16); \( N = 2 \). (a) BER performance; (b) SER performance; (c) PER performance.

Figure 11: Continued.
It is apparent from (13) and Figure 4 that the estimation process involves complex inverse tangent operation. According to our previous work [21, 25–27], two simplified estimation schemes can be obtained:

\[
\phi \approx \begin{cases} 
\frac{2}{K+1} \frac{\text{Im}(Q)}{\text{Re}(Q)}, & \text{if } \text{Re}(Q) > 0 \text{ and } |\text{Re}(Q)| \geq |\text{Im}(Q)|, \\
\frac{2}{K+1} \left( \frac{\pi}{2} - \frac{\text{Re}(Q)}{\text{Im}(Q)} \right), & \text{if } \text{Im}(Q) > 0 \text{ and } |\text{Re}(Q)| < |\text{Im}(Q)|, \\
\frac{2}{K+1} \left( -\pi + \frac{\text{Im}(Q)}{\text{Re}(Q)} \right), & \text{if } \text{Re}(Q) < 0 \text{ and } |\text{Re}(Q)| \geq |\text{Im}(Q)|, \\
\frac{2}{K+1} \left( -\frac{\pi}{2} - \frac{\text{Re}(Q)}{\text{Im}(Q)} \right), & \text{if } \text{Im}(Q) < 0 \text{ and } |\text{Re}(Q)| < |\text{Im}(Q)|,
\end{cases}
\]

(14)

\[
\phi \approx \begin{cases} 
\frac{2}{K+1} \frac{\text{Im}(Q)}{\sqrt{\text{Re}^2(Q) + \text{Im}^2(Q)}}, & \text{if } \text{Re}(Q) > 0 \text{ and } |\text{Re}(Q)| \geq |\text{Im}(Q)|, \\
\frac{2}{K+1} \left( \frac{\pi}{2} - \frac{\text{Re}(Q)}{\sqrt{\text{Re}^2(Q) + \text{Im}^2(Q)}} \right), & \text{if } \text{Im}(Q) > 0 \text{ and } |\text{Re}(Q)| < |\text{Im}(Q)|, \\
\frac{2}{K+1} \left( -\pi - \frac{\text{Im}(Q)}{\sqrt{\text{Re}^2(Q) + \text{Im}^2(Q)}} \right), & \text{if } \text{Re}(Q) < 0 \text{ and } |\text{Re}(Q)| \geq |\text{Im}(Q)|, \\
\frac{2}{K+1} \left( -\frac{\pi}{2} + \frac{\text{Re}(Q)}{\sqrt{\text{Re}^2(Q) + \text{Im}^2(Q)}} \right), & \text{if } \text{Im}(Q) < 0 \text{ and } |\text{Re}(Q)| < |\text{Im}(Q)|.
\end{cases}
\]

(15)
For integrity of this work, we also give the full estimation scheme here.

\[
\hat{\varphi} = \begin{cases} 
\frac{2}{K+1} \tan^{-1} \left( \frac{\text{Im}(Q)}{\text{Re}(Q)} \right), & \text{if } \text{Re}(Q) > 0 \text{ and } |\text{Re}(Q)| \geq |\text{Im}(Q)|, \\
\frac{2}{K+1} \left( \frac{\pi}{2} - \tan^{-1} \left( \frac{\text{Re}(Q)}{\text{Im}(Q)} \right) \right), & \text{if } \text{Im}(Q) > 0 \text{ and } |\text{Re}(Q)| < |\text{Im}(Q)|, \\
\frac{2}{K+1} \left( -\pi + \tan^{-1} \left( \frac{\text{Im}(Q)}{\text{Re}(Q)} \right) \right), & \text{if } \text{Re}(Q) < 0 \text{ and } |\text{Re}(Q)| \geq |\text{Im}(Q)|, \\
\frac{2}{K+1} \left( -\frac{\pi}{2} - \tan^{-1} \left( \frac{\text{Re}(Q)}{\text{Im}(Q)} \right) \right), & \text{if } \text{Im}(Q) < 0 \text{ and } |\text{Re}(Q)| < |\text{Im}(Q)|.
\end{cases}
\]

A data-aided detection scheme is considered for simple implementation. However, the non-data-aided detection method can also be applied. This follows from the fact that the modulated data within the sample chip can be easily wiped out with the aid of any PN code using cross-correlation operation. Then, the CFOE can be easily estimated and compensated. At present, there are many other CFO estimation schemes [28–31]. These schemes involve complex mathematical operations, such as the logarithmic operation, the exponential operation, and the trigonometric operation. Our scheme further simplifies these complex operations, as shown in (14) and (15).

6. Simulation Result

In this section, we evaluate the bit error rate (BER), symbol error rate (SER), and packet error rate (PER) performance of various detection schemes. Note that, in the simulation, the PPDU payload length is set to 22 bytes. We choose the maximum in 780 MHz frequency band as the carrier frequency, that is, 786 MHz. The detailed simulation parameters are shown in Table 2.

6.1. Influence of Maximum Chip-Delay Number K on Detection Performance. The performance of IEEE 802.15.4c MPSK receiver can be improved by introducing a compatible maximum chip-delay number K. In slow fading Rayleigh channel and the pure AWGN channel with different K, we compared the BER, SER, and PER of our proposed MSD scheme with the full estimator in (16) and the simplified estimator in (15).

It can be seen from Figures 5 and 6 that when the maximum chip-delay number K increases from 1 to 5, the BER, SER, and PER performance can improved under the pure AWGN channel. In particular, as depicted in Figure 5(c), when PER = 1 × 10⁻³, as K increases from 1 to 2, the SNR gain is approximately 2.2 dB; when K increases from 2 to 3, the SNR gain is about 0.5 dB; when K increases from 3 to 4, the SNR gain is about 0.1 dB. Furthermore, K = 3 is sufficient to meet the performance requirements of the receiver in pure AWGN channel [9]. Also, the improvement is so small when the maximum chip-delay number ranges from 4 to 5, so we set the maximum chip-delay number to be 4, that is, K = 4.

6.2. Detection Performance Comparison. The BER, SER, and PER results for various detection schemes under pure AWGN channel and slow fading Rayleigh channel are, respectively, shown in Figures 7 and 8. In theory, the full MSD scheme is extremely close to the optimal coherent detection with the increase of the observation window length N. The implementation of the full MSD scheme is too complex. For the convenience of comparison, we use the optimal coherent detection to replace the simulation results of the full MSD scheme. We take the optimal coherent detection as the lowest bound.

As shown in Figure 7, when N = 2, the simplified estimation in (14) would lead to serious error floor. This is
caused by the continuous accumulation of larger estimation errors in (3). However, when the full estimation in (16) and the simplified estimation in (15) are used, the detection performance is excellent. There is a little gap between those and the optimal coherent detection, especially at high SNR. Furthermore, when we have $1 \times 10^{-3}$, compared with the SBSD method, the proposed scheme can achieve gain about 1.6 dB.

When $N$ is increased from 2 to 3, the performance of the proposed detection schemes decreases. This is because when $N = 3$, the error caused by the estimation scheme introduces a large accumulation in (3), and a mismatch is then observed between the estimator and the detector. In conclusion, the estimation scheme in this paper is especially suitable for the detection scheme, wherein the observation window length $N$ is set to be 2. Moreover, when $N = 2$, the performance gap between our proposed scheme and the optimal coherent detection is so small, and there is almost no more room for improvement. Therefore, we choose $N$ as 2 in the subsequent simulations. In addition, as shown in Figure 8, we can draw similar conclusions under slow fading Rayleigh channel, which, however, is not illustrated here.

6.3. Frequency Offset Robustness of the Proposed Scheme in Pure AWGN Channel. In the pure AWGN channel, we show the BER, SER, and PER performance results with different estimation schemes in Figure 9. CFO $f$ obeys a symmetry triangular distribution at ($-80$, $+80$) ppm. The results of the full estimation in (16) are used as a benchmark. As shown in Figure 9, for the simplified estimation in (15), the detection performance is good for CFO between +60 and −60 ppm. However, the performance fluctuates when the SNR is greater than +60 ppm or less than −60 ppm. In addition, the performance fluctuation increases with the increase of SNR. However, according to the CFO probability distribution characteristic, the probability that the absolute value of CFO exceeds 60 is 0.0625, which is very small. Thus, the proposed detection scheme is not sensitive to frequency offset.

6.4. CPO Robustness under Pure AWGN Channel. In this part, we study the detection performance of the proposed receiver in pure AWGN channel with changing carrier phase, where $N = 2$. In Figures 10 and 11, the proposed scheme is robust to dynamic phase jitter. The phase $\theta$ is modeled as a Wiener process, wherein its initial value is uniformly chosen from ($-\pi, \pi$). As shown in Figures 10 and 11, the proposed scheme is robust to dynamic phase jitter. The performance of the proposed receiver does not significantly degrade if we increase the standard deviation of jitter from 0° to 3°. In addition, an irreducible error floor is observed for the estimators given in (15) and (16) with the increase of SNR.

6.5. Complexity Analysis. We compare the implementation complexity of various detection schemes in pure AWGN channels. It is assumed that the full MSD and the proposed detection scheme are equipped with the same estimator. The discrepancy in receiver implementation complexity is determined by the metrics given in (3), (5), and (9). Note that we set $J$, $L_1$, and $P_1$ to the maximum, that is, $J = 44$, $L_1 = 16$, and $P_1 = 8$. The structure block diagram of multiplication operation is shown in Figure 3 of [32]. Complex addition is the addition of two complex numbers. It is assumed that a comparison operation is equivalent to an addition operation. As shown in Table 3, our proposed detection scheme only requires 576 complex multiplications, 573 complex additions, and 34 modular squaring operations. The full MSD given in Section 3 requires 8192 complex multiplications, 8064 complex additions, and 256 modular squaring operations. Obviously, compared with the full MSD, the complexity of our scheme is extremely reduced.

In addition, the average running time can also partly reflect the implementation complexity. Specifically, for different detection schemes, we develop various simulations by running enough number of transmission frames. In fact, $10^5$ frames of data are implemented, and the average running time is achieved. Surprisingly, as shown in Figure 12, when the SNR is −4 dB, the average running time for the traditional SBSD developed from [23] is 4 times as much as that of our proposed MSD scheme. Furthermore, as for the SBSD developed from [33], the average running time is 5.5 times as much as that of our MSD scheme.

7. Conclusions

In this paper, a simple but reliable MSD scheme for IEEE 802.15.4c MPSK receiver has been proposed, wherein the CFO has been estimated and compensated by using preamble assisted method. Experimental results showed that our detection performance can meet the requirements of WSN with only four maximum chip delays. In addition, when the standard deviation of the phase jitter is as high as
3”, the performance does not significantly decrease. Finally, compared with the full MSD scheme, our improved scheme is more attractive in terms of complexity. Therefore, the research results of this paper have a positive role in promoting the engineering application of the IoT in the field of new smart city.

In order to avoid the channel and CFO estimation, double-differential modulation is famously used as shown in [34–37]. This idea can be directly borrowed and implemented in our detection scheme for further complexity reduction. Note, however, that more performance loss is exhibited in this case.

**Data Availability**

All of the underlying data in this manuscript are available from the corresponding author upon request.

**Conflicts of Interest**

The authors declare that there are no conflicts of interest.

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