A Fast-Convergent Modulation Integral Observer for Online Detection of the Fundamental and Harmonics in Grid-Connected Power Electronics Systems

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Abstract— Harmonics detection is a critical element of active power filters. A previous review has shown that the Recursive Discrete Fourier Transform and the Instantaneous p-q Theory are effective solutions to extracting power harmonics in single-phase and three-phase power systems, respectively. This paper presents the operating principle of a new modulation function integral observer algorithm that offers a fast solution for the extraction of the fundamental current and the total harmonic current when compared with existing methods. The proposed method can be applied to both single- and three-phase systems. The observer-based algorithm has an advantageous feature of being able to be tuned offline for a specific application, having fast convergence and producing estimated fundamental component with high circularity. It has been tested with both simulations and practical measurements for extracting the total harmonic current in a highly efficient manner. The results have confirmed that the proposed tool offers a new and highly effective alternative to the smart grid industry.

Index Terms— fundamental extraction, harmonics detection, active power filters

I. INTRODUCTION

Nonlinear electric loads draw non-sinusoidal currents from the power supplies. Such harmonics have been identified as key factor for poor power quality and a range of adverse effects, including but not limited to, electromagnetic interference (EMI), overheating of cables, and low power factors [1]-[4]. Since the 1980’s, the power electronics research community has responded with the solutions of active power filters (APF) to deal with the power harmonics problems [5]-[10]. In a typical setup of an active filter as shown in Fig.1, one critical element is the real-time detection of the harmonic current. The control block involves a fast harmonic current detection so that a reference of the harmonic current can be derived for the power inverter to inject the harmonic current into the power line.

Despite such APF technologies have been commercialized [11], active research is still on-going in searching for effective real-time harmonics detection methods. In a recent survey [12], it is stated that “There exist many implementations supported by different theories (either in time- or frequency-domain), which continuously debate their performances proposing ever better solutions.” The survey [12] provides a comprehensive comparison on six different methods covering (i) Discrete Fourier Transform (DFT), (ii) Recursive DFT, (iii) Synchronous Fundamental dq-Frame, (iv) 5th Harmonic dq-Frame, (v) p-q Theory and (vi) 5th Generalized Integrator. Methods (i), (ii) and (vi) are suitable for single-phase systems, and hence are suitable for three-phase ones as well. Methods (iii), (iv) and (v) can only be applied for three-phase systems and require filtering. In terms of settling time and phase error, the Recursive DFT and the p-q Theory methods offer better performance than the others in the comparison.

Fig.1 A typical setup of an Active Power Filter
The methods described in [12] are familiar ones in the power electronics and power systems communities. However, significant progress has been achieved in signal processing and some of the emerging signal-processing methods are also applicable to harmonics detection. The frequency domain detection methods, such as discrete-time Fourier transform (DFT) and the sliding DFT [14]-[17] represent the most commonly used techniques for their fast convergence speed and higher accuracy comparing to the time-domain methods. Despite their successful applications, the existing methods based on DFT also suffer from two main drawbacks: (1) a synchronous sampling is needed, (2) problems in processing the inter-harmonics may occur. On the other hand, in the time-domain, methods like the Kalman filter, adaptive Notch filter (ANF), Phase-locked-loop (PLL) and second-order generalized integrators (SOGI) are proposed. These methods usually ensure enhanced noisy immunity at the expense of longer transient time than frequency-domain algorithms [18]-[25]. Moreover, the implementation of time-domain methods is more instrumental in the context of real-time harmonic compensation. In [26], PLL, recursive DFT (RDFT), and discrete Kalman filtering (DKF) from aforementioned two categories are compared analytically in terms of real-time implementation.

Very recently, a novel modulation-function integral observer (MF observer) has been developed for general single-input single-output linear systems. It offers practically instantaneous convergence without high gain injection and complexity [27]. The novel theory reported in [27] is supported by a rigorous and rather complex mathematical analysis. Thus, it may not be obvious to power electronics engineers without the relevant theoretical background. It is the aim of this paper to explore the use of this new modulation integral observer for extracting the fundamental current and the total harmonic current, the knowledge of the latter being required in active power filtering applications. The practical relevance of this MF observer with active power filter applications is explained in this paper for the first time. It is discovered that this new method, although unknown to the power electronics and power system communities, can perform fundamental current and total harmonic current extraction quickly even in presence of unknown dc offsets. The fast-convergent MF observer is implemented in non-selective form to achieve a low computational burden for on-line total harmonics extraction. As confirmed by the simulation and experimental results, this new method is favorably compared with some of the best harmonics extraction methods commonly used in the active power filtering applications.

II. BRIEF SUMMARY OF FREQUENCY-DOMAIN AND TIME-DOMAIN HARMONIC DETECTION METHODS

Although conducted 10 years ago, the review in [12] still provides some important insights into the operating principles and the algorithm structures of different harmonic detection methods. In this section, the schematics of the main methods are used to highlight the differences of the methodologies in terms of their algorithm structures. The understanding of such differences and the schematic structures of the harmonic detection algorithms will form a basis for comparing various approaches, including the fast-convergent integral observer under investigation in this paper.

A. Frequency-Domain Discrete Fourier Transform and Recursive DFT

Fig. 2 shows the schematic of the DFT method. The Recursive DFT (RDFT) is similar to the DFT, except that the samples are obtained within a moving window of the sampled data. Note that the DFT method and its variants require the measurements of the current only. There is no need to measure the mains voltage. Its ability to handle 1-phase measurements naturally indicates that it can be applied to a 3-phase system. There are two ways to work out the total harmonic current.

- The first one is to obtain the sum of the dc component and fundamental component. By subtracting this sum from the input current waveform, the total harmonic current function can be obtained.
- The second way is to add several of the low-order current harmonic components together to form an approximate harmonic current.

B. Time-Domain dq-Frame methods

Unlike the DFT method, the time-domain dq-Frame approach requires the measurements of both of the mains voltages and currents for the 3 phases. Therefore, this method requires more sensors and is not suitable for single-phase applications. The phase currents are measured and transformed into the direct (d) and quadrature (q) components. The time-domain dq-Frame method can also be used in two ways as described for the DFT method.

(i) Synchronous fundamental dq-Frame method

Fig.3 shows the schematic of this method for extracting the harmonic components. The rotating frame of this method rotates at the fundamental frequency. This makes the fundamental component appearing as the dc component after the transformation in the rotating frame. With the use of a high-pass filter with a typical bandwidth of 25Hz to 120Hz, the fundamental d- and q- current components can be eliminated. The high-pass filter enables the d- and q-components of the harmonic currents (iₐ₋ and iₜ₋) to go through. By doing a dq-abc transformation, the harmonic currents of the three phases can be re-constructed as iₐ*, i₉* and iₜ* in Fig.3.
(ii) Synchronous harmonic dq-Frame method

The schematic of the harmonic dq-Frame is shown in Fig. 4. Again, the three phase currents and voltages have to be measured. The procedures for extracting the harmonic currents are similar to that of the fundamental dc frame method. The major difference is that the rotating frames for the harmonics will be according to their respectively harmonic frequencies. Once the harmonic currents are obtained, they can be added together to form the total harmonic current. In practice, only several of the low-order harmonic currents are usually used to reconstruct a good approximation of the total harmonic currents for the three phases.

It is important to note that the Fundamental dq-Frame and Harmonic dq-Frame methods require the use of high-pass or low-pass filters in the process of deriving the required current components. Consequently, such approach unavoidably introduces phase error, as previously discussed in [12].

(iii) Instantaneous p-q power theory

Another common approach for harmonic extraction is based on the instantaneous p-q power theory pioneered by Prof. Akagi [28]. The schematic is shown in Fig. 5. Like the dq-Frame methods, this approach needs the measurements of the phase currents and voltages. Thus, it is suitable for 3-phase systems, but not for single-phase ones. The phase voltage and current measurements are transformed into their phase systems, but not for single-phase. The major difference is that the rotating frames for the harmonics will be according to their respective harmonic frequencies. Once the harmonic currents are obtained, they can be added together to form the total harmonic current. In practice, only several of the low-order harmonic currents are usually used to reconstruct a good approximation of the total harmonic currents for the three phases.

In the next section, the proposed fast fundamental and harmonics estimator will be illustrated.
\[ \sum_{k=2}^{N} i_k(t) = i(t) - y(t) \]

Therefore, \( y(t) \) is chosen to be the single output of the integral observer introduced in the following lines. Differentiating (3) twice leads to:

\[ \ddot{y}(t) = -\omega^2 i_1(t) \]

(5)

where \( \ddot{y}(t) \) is the second-order time-derivative of \( y(t) \). Based on (3) and (5), one can form the following expression:

\[ \begin{bmatrix} y(t) \\ \dot{y}(t) \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 0 & -\omega^2 \end{bmatrix} \begin{bmatrix} c \\ i_1(t) \end{bmatrix} \]

(6)

Now, assume that the single output \( y(t) \) is generated by a general 3rd order state-space observable system,

\[ \begin{align*}
\dot{z}(t) &= A_z z(t) \\
y(t) &= c_z^T z(t)
\end{align*} \]

(7)

where we let

\[ z(t) = \begin{bmatrix} z_1(t) \\ z_2(t) \\ z_3(t) \end{bmatrix} \]

\[ A_z = \begin{bmatrix} 0 & 1 & 0 \\ -\omega^2 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}, \quad c_z = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \]

(8)

It can be easily seen that

\[ z_1(t) = y(t) \]

\[ z_3(t) = \ddot{y}(t) + \omega^2 y(t) \]

(9)

(10)

Clearly, the total harmonic current can be derived as long as \( z_1(t) \) is available. However, in the following lines, we will show that the dc offset and the fundamental signal can be individually recovered by using the state vector \( z(t) \). Since only \( z_1(t) \) and \( z_3(t) \) are relevant in the calculation, the following matrix equations can be derived:

\[ z_e(t) = A y(t) \]

(11)

where

\[ z_e(t) = \begin{bmatrix} z_1(t) \\ z_3(t) \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 0 \\ \omega^2 & 1 \end{bmatrix} \]

and

\[ y(t) = \begin{bmatrix} y(t) \\ \ddot{y}(t) \end{bmatrix} \]

thus leading to

\[ y(t) = A^{-1} z_e(t) \]

(12)

The 2×1 state column vector \( z_e(t) \) can be related to the 3×1 state column vector \( z(t) \) as:

\[ z_e(t) = I_e z(t) \]

(13)

with

\[ I_e = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \]

In view of (13), (12) and (6), the fundamental and the dc component modelled in the right hand side of (6) can be retrieved by simple algebra (see A.3-A.5 in Appendix I).

### B The Modulation Function and the Modulation Matrix

The modulation integral observer uses a set of the time-monomial modulating functions (see [27]). In the case addressed in this paper, the specific functions take on the form

\[ \phi_h(t) = \frac{w_n t^{2n-h}}{(2n-h)!} \quad \text{for} \quad h \in \{1,2,...,n\} \]

(14)

where \( n \) is the order of the system (i.e., \( n = 3 \) for the 3rd order observer in this case) and \( w_n > 0 \) is a suitable positive weighting factor. According to [27], the chosen modulation functions must satisfy:

\[ \frac{d^n \phi_h(0)}{dt^n} = 0 \]

(15)

The reason for selecting the modulation function (14) that is characterized by condition (15) is provided in the Appendix II. With the modulation function selected, the modulation function matrix is defined as:

\[ \Gamma(t) = \begin{bmatrix} \phi_1(t) & -\phi_1(t) & \phi_1(t) \\ \phi_2(t) & -\phi_2(t) & \phi_2(t) \\ \phi_3(t) & -\phi_3(t) & \phi_3(t) \end{bmatrix} \]

(16)

where \( \phi_h(t) \) and \( \dot{\phi}_h(t) \) are the first and second derivatives of \( \phi_h(t) \) for \( h \in \{1,2,3\} \). For any choice of \( w_n \), it holds that

\[ \phi_1(t) \neq \phi_2(t) \neq \phi_3(t), \forall t > 0 \]

\[ \phi_1(t) \neq \phi_2(t) \neq \phi_3(t), \forall t > 0 \]

\[ \phi_1(t) \neq \dot{\phi}_2(t) \neq \dot{\phi}_3(t), \forall t > 0 \]

thus \( \Gamma(t) \) is invertible for any \( t > 0 \).

**Remarks:**

Although \( \phi_h(t), h = 1,2,3 \) are monotonically increasing functions with respect to \( t \), they are only computed over \( 0 \leq t \leq T_\Delta + T_r \) due to the action of reset (which is described later on), thus evading the issue of data overflow. In this respect, excessively large \( w_n \) is not advisable. The same rule is applicable to \( \phi_h(t) \) and \( \dot{\phi}_h(t) \). Moreover, by observing the pattern of \( \Gamma(t), w_n, h = 1,2,3 \) should be chosen such that the resultant \( \phi_h(t), h = 1,2,3 \) are enough separated to facilitate the computation of matrix inverse. It is also important to note that the elements in the matrix of (17) can be pre-determined in an offline manner in the interval: \( 0 \leq t \leq T_\Delta + T_r \). Therefore, the matrix inversion is done only once. These pre-determined values can be stored in memory for real-time implementation.

With the availability of the modulation function matrix, the following matrix equation can be formed:
\[
\begin{bmatrix}
  v_{\phi_1}(t) \\
v_{\phi_2}(t) \\
v_{\phi_3}(t)
\end{bmatrix} = \begin{bmatrix}
  \dot{\phi}_1(t) & -\phi_1(t) & \phi_3(t) \\
  \dot{\phi}_2(t) & -\phi_2(t) & \phi_3(t) \\
  \dot{\phi}_3(t) & -\phi_3(t) & \phi_3(t)
\end{bmatrix} \begin{bmatrix}
  z_1(t) \\
z_2(t) \\
z_3(t)
\end{bmatrix}
\]

Hence,
\[
\begin{bmatrix}
z_1(t) \\
z_2(t) \\
z_3(t)
\end{bmatrix} = \begin{bmatrix}
  \dot{\phi}_1(t) & -\phi_1(t) & \phi_3(t) \\
  \dot{\phi}_2(t) & -\phi_2(t) & \phi_3(t) \\
  \dot{\phi}_3(t) & -\phi_3(t) & \phi_3(t)
\end{bmatrix}^{-1} \begin{bmatrix}
v_{\phi_1}(t) \\
v_{\phi_2}(t) \\
v_{\phi_3}(t)
\end{bmatrix}
\]

(17a)

where the elements of column vector
\[
v_{\phi}(t) = \begin{bmatrix}
v_{\phi_1}(t) \\
v_{\phi_2}(t) \\
v_{\phi_3}(t)
\end{bmatrix}^T
\]
can be obtained from the practical measurements of the line current \(i(t)\). It follows that
\[
v_{\phi}(t) = \int_0^t \begin{bmatrix}
  \phi_1(\tau) \\
  \phi_2(\tau) \\
  \phi_3(\tau)
\end{bmatrix} d\tau
\]
(18)

Thus,
\[
\dot{v}_{\phi}(t) = \begin{bmatrix}
  \phi_1(\tau) + \omega^2 \phi(t) \\
  \phi_2(\tau) + \omega^2 \phi(t) \\
  \phi_3(\tau) + \omega^2 \phi(t)
\end{bmatrix} \cdot i(t)
\]
(19)

By observation (19), in order to avoid error-accumulation due to the integration in case the measurement of \(i(t)\) is noisy, we adopt a discrete-time deflation strategy which is described in the following lines. Let us define the time variable
\[
t_k = T + (k+1)T_r
\]
(20)

where \(T_r\) is the initial time, \(T_r\) is a repetitive period (note: within each repetitive period are many small sampling periods), and \(k\) is a natural number, the first derivative of \(\dot{v}_{\phi}(t)\) can be obtained from the practical measurement of the line current \(i(t)\). When \(k = 0, t = t_0 = T_\Delta + T_r\), this forms the first repetitive period. Beyond this time, the time variable lies within the next repetitive period \(T_r\) (i.e. \(t_k < t < t_{k+1}\)).

\[
\dot{v}_{\phi}(t) = \begin{bmatrix}
  \phi_1(t) + \omega^2 \phi(t) \\
  \phi_2(t) + \omega^2 \phi(t) \\
  \phi_3(t) + \omega^2 \phi(t)
\end{bmatrix} \cdot i(t)
\]
for \(0 \leq t < t_0\)

\[
\left(\phi_1(t - t_k + T_\Delta) + \omega^2 \phi(t - t_k + T_\Delta)\right) \cdot i(t)
\]
for \(t_k < t < t_{k+1}\)
(21)

The time integrals of \(\dot{v}_{\phi}(t)\) in (21) will lead to the values of \(v_{\phi}(t)\) for \(h = (1,2,3)\). At the alternation of each repetitive period, the values of \(v_{\phi}(t)\) should be reset as:
\[
v_{\phi}(t) = E_h^T \Gamma(T_\Delta + T_r) v_{\phi}(t)
\]
for \(t = t_k\)
(22)

where \(E_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, E_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}\) and \(E_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}\).

Note that the line current is sampled at a relatively high speed within each repetitive period, and the integral values of (21) can provide the time-domain vector \(v_{\phi}(t)\) continuously at each sampling instant, then the state vector \(z(t)\) can be obtained from (17).
\[
z(t) = \begin{bmatrix}
  \Gamma^{-1}(t) \cdot v_{\phi}(t) \\
  \Gamma^{-1}(t - t_k + T_\Lambda) \cdot v_{\phi}(t)
\end{bmatrix}
\]
(23)

where \(T_\Delta\) is a small time. It is needed because the modulation function matrix is not invertible at \(t=0\). Thanks to (9), \(z_1(t)\) represents an estimate of \(y(t)\), then the total harmonic current can be obtained from (4). In the practical implementation, the proposed integral observer algorithms can be discretized as shown in the Appendix I.

IV. SIMULATION STUDY OF CONVERGENT RATES AND TRACKING ERRORS

In order to evaluate the performance of the methods under consideration, a simulation study based on the conditions similar to those set in [12] has been conducted. The characteristic of the line currents are tabulated in Table I and such reference current waveform is shown in Fig.7. We compare the MF observer with two well-known methods that are successfully used in the active power filter context: the recursive DFT, p-q theory, and a variant of the adaptive notch filtering (ANF) [24] recently proposed for this typical application. The realizable algorithms of each method are described in the Appendix. For the MF observer, the weighting factors are tuned offline to be \(w_1 = 1, w_2 = 2\) and \(w_3 = 3, T_r = T_\Lambda = 0.1\). For the Instantaneous p-q power theory, the high-pass filter has a cut off frequency of 200 rad/s. A pair of RDFT filters are employed with order 0 and 1 for the offset and fundamental respectively. The ANF is tuned with \(\zeta_0 = \zeta = 0.8\) and \(\zeta = 30\).

TABLE I  CHARACTERISTICS OF LINE CURRENT

<table>
<thead>
<tr>
<th>Indices</th>
<th>Amplitude</th>
<th>Phase</th>
</tr>
</thead>
<tbody>
<tr>
<td>Offset</td>
<td>3</td>
<td>0</td>
</tr>
<tr>
<td>Fundamental</td>
<td>30</td>
<td>0, 120, 240</td>
</tr>
<tr>
<td>5th harmonic</td>
<td>10</td>
<td>5 × (0, 120, 240)</td>
</tr>
<tr>
<td>7th, 11th, 13th, 17th</td>
<td>1.5</td>
<td>h × (0, 120, 240)</td>
</tr>
</tbody>
</table>

Fig.7 A reference current waveform constructed from the components tabulated in Table I the for the simulation study.
method and the MF method have the fastest convergence performance.

The MF observer takes about 10ms to track the harmonic waveform and is faster than RDFT and ANF, which require at least 20ms. Although RDFT offers minimal steady-state error in a steady-state situation, it is important to note that fast convergence offers advantage to signal detection applications in which fast transients and frequency variations are common. Based on these simulation results, a comparative table with respect to the steady state error is also constructed as shown in Table III. The following points can be noted:

1. The results shown in the Table II confirm that the MF performs the best among all the approaches
2. RDFT is very smooth in steady state; therefore the relative estimation error decreases as more cycles are considered.
3. ANF has the worst steady state behavior; therefore the relative error increases as more cycles are considered.

V. EXPERIMENTAL VERIFICATION

A third-order MF observer has been implemented to derive the fundamental and the dc bias current only. The total harmonic current is then obtained by subtracting dc and fundamental current components from the line current. Based on the platform suggested in [12], the total harmonic currents, settling times, phase errors and overshoots of the following methods are compared with the practical measurements.

- MF observer
- Recursive DFT,
- The Instantaneous p-q Theory method
- Adaptive Notch Filter (ANF) method [24]
The algorithms of the 3rd order MF observer are implemented in a discretized form which is included in Appendix I. The algorithms of the RDFT and ANF methods also can be referred to the Appendix I. The algorithms of the instantaneous p-q theory are well known and simply follow the block diagrams as shown in Fig. 5. The algorithms under consideration are implemented in a dSpace DS1006 system. Fig. 10 shows the hardware setup, which includes a programmable power supply (AMETEK CSW11100), three inductors of slightly different inductance values, and three resistive loads. The line currents of the three phase system are displayed in Fig. 11.

A Extraction of the fundamental current component

Since the Instantaneous p-q power theory method uses a high-pass filter to remove the fundamental component in order to derive the total harmonic current (Fig. 5), it is not included in this comparison because it does not generate the fundamental current. Only the experimental results of the MF observer, RDFT and ANF methods are included in this section. Under a fixed sampling frequency of 10 kHz, the same dSpace system is used to calculate the fundamental current in real time. The analogue of this estimated fundamental current is outputted to a digital storage oscilloscope (DSO X3034A). The estimated fundamental currents of the MF observer, RDFT and ANF methods are captured and displayed in Figs. 12(a), 12(b) and 12(c), respectively.

In order to observe the deviation of these waveforms from a pure sinusoidal waveform, let us transform the estimated sinewave into a unit “circle” by picking arbitrarily consecutive 200 samples in one cycle. It is worth noting that a perfect time-domain sinusoidal function can be represented as an ideal circle on the 2-D plane. By comparing the trajectory of the measured current with a perfect circuit, any deviation from the perfect circle can be observed. One method of measuring the quality of the extracted fundamental component is the concept of Circularity [29], which defines the formula for Circularity (C)

\[ C = \frac{4\pi A}{P^2} \]  

with A and P the area and the perimeter enclosed by the circular figure, respectively. Meanwhile, the isoperimetric theorem indicates that \( C = 1 \) only holds for an ideal circle. To this end, the circularity of each figure is evaluated in the following way:

1. Divide the shape into circular sectors with radius \( r_i \), \( i = 1, 2, \cdots, 200 \) based on the samples.
2. Approximation of the area \( A = \sum_{i=1}^{200} \frac{1}{2} \pi r_i^2 \)
3. The perimeter \( P \) of the curve is approximated by the sum of the distances between every two adjacent samples.

As such, the vector trajectories of one period of three estimated fundamental current waveforms are plotted and compared with a perfect circle in Figs. 13(a), 13(b) and 13(c). It can be observed that the MF observer method generate the best fundamental current with the least amount of noise and distortion. For the sake of completeness, the circularity values of the MF observer, the RDFT and the ANF have been obtained and are recorded in Table IV. It can be seen that the curve generated by MF has the circularity closest to 1 (i.e with the least absolute deviation from 1.0).

**Table IV Comparison of Circularity of the Extracted Fundamental Components**

|             | Circularity C | Absolute deviation \( |C - 1.0| \) | Figure   |
|-------------|---------------|----------------------------|----------|
| MF observer | 1.0017        | 0.0017                     | 13(a)    |
| RDFT       | 0.8925        | 0.1075                     | 13(b)    |
| ANF        | 0.8883        | 0.1117                     | 13(c)    |
B Detection of the total harmonic current

The second set of evaluation involves all of the four methods under consideration. This time, the total harmonic currents estimated by the four methods are captured. Their harmonic spectra are then plotted and compared with the harmonic spectrum of the mains current, which is included in Fig.14 as a reference. Figs. 15(a), 15(b), 15 (c) and 15(d) show the frequency spectra of the estimated total harmonic currents obtained from the (a) MF observer, (b) RDFT, (c) Instantaneous p-q power Theory and (d) the ANF methods, respectively. Comparing with the spectrum of the measured line current in Fig.14 and ignoring the part of the spectrum up to about 200 Hz (to remove the effect of the fundamental current), the spectra of MF observer, RDFT method and the ANF method are close to that of the line current.

C Overall Comparison

With the results obtained, the performance of the four methods can be summarized as tabulated in Table V. The RDFT has generally good performance. The only limitation is its
requirement for synchronous sampling that may lead to accuracy issues when the frequency is changing dynamically.

The Instantaneous p-q power theory requires the measurements of voltage and current. It is a fast method to generate the total harmonic content. But the introduction of filter in the process of harmonic extraction inevitably introduces phase error. The proposed MF observer has fast convergence for tracking the total harmonic content. Among the three methods that can extract the fundamental current, the MF observer offers the highest quality in terms of circularity. Because of the high quality of the fundamental current, its resultant total harmonic current (which is obtained by subtracting the fundamental current from the line current) is also of high quality.

VI. CONCLUSIONS
This paper presents the first practical implementation of a new modulation function integral observer for the extraction of the fundamental current and the total harmonic current of a three-phase nonlinear load. The observer algorithm has been adopted specifically for fundamental current and total harmonic current extraction. It has been explained step by step so that power electronics researchers and engineers can implement it without understanding the full details of the highly mathematical observer theory given in [27] in a more general setting. Based on the concept of circularity, it has been practically confirmed that this method has excellent performance in obtaining a high-quality representation of the fundamental current in real time from a highly nonlinear current waveform. Consequently, the total harmonic current, which is obtained by subtracting the fundamental current from the total line current, is also of high quality. Such high-quality signal extraction capability has been favorably compared with the RDFT, the Instantaneous p-q power theory (that are commonly used in the power electronics community) and the ANF methods. The fast convergent feature of this observer has also been demonstrated in a simulation study. Based on the comparison provided, it is concluded that the modulation function integral observer method offers a fast and practical online solution to obtaining the fundamental component and the total harmonic content of a highly nonlinear signal. Such functions can find many power electronics applications (including active power filters) in which highly nonlinear currents are involved.

<table>
<thead>
<tr>
<th>TABLE V COMPARISON OF FOUR METHODS FOR HARMONICS EXTRACTION</th>
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<tbody>
<tr>
<td></td>
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<tr>
<td>Required measurements</td>
</tr>
<tr>
<td>Response time</td>
</tr>
<tr>
<td>Steady state accuracy</td>
</tr>
<tr>
<td>Selective harmonic compensation</td>
</tr>
<tr>
<td>1-ph/3-ph applications</td>
</tr>
<tr>
<td>Need for synchronous sampling</td>
</tr>
<tr>
<td>Computational load (p-u)²</td>
</tr>
<tr>
<td>Estimation of the dc offset</td>
</tr>
<tr>
<td>Circularity of fundamental</td>
</tr>
</tbody>
</table>

APPENDIX I

1. MF observer
   Discretized algorithm of (21)-(23) by a trapezoidal integration rule (see A1 at the bottom of this page):

   Hence,

   \[
   z(k) = \begin{cases} 
   0,0 \leq k \leq N_e \\
   (\Gamma(k))^{-1}v(k), N_e \leq k \leq N_0 \\
   (\Gamma(k - N_k + N_0))^{-1}u(k), N_k \leq k \leq N_{k+1} 
   \end{cases}
   \]

   (A.2)

1 Multiple harmonics can be dealt with by a higher order MF observer (e.g. 2n + 1-th order for n harmonics and the dc offset), which omitted here for sake of unnecessary complexity

² The computation time is evaluated by Matlab under a fixed simulation example

where \( T_s \) is the fixed sampling period, \( N_k = \frac{t_k}{T_s}, k \in \mathbb{N}, N_\Delta = \frac{T_\Delta}{T_s}, N_r = \frac{T_r}{T_s}, N_\epsilon = \frac{T_\epsilon}{T_s} \).

Particularly, in the discrete-time case \( T_r, T_\Delta, T_\epsilon \) are chosen as multiples of the sampling time \( T_s \).

Finally

\[
\begin{align*}
\mathbf{z}_e(k) &= I_z \mathbf{z}(k) \\
\mathbf{y}(k) &= A^{-1} \mathbf{z}_e(k) \\
\left[ \begin{array}{c} c \\
0 \\
i_3(k) \end{array} \right] &= \begin{bmatrix} 1 & 1 \\
0 & -\omega^2 \end{bmatrix} \mathbf{y}(k)
\end{align*}
\]

(1.3) (A.4) (A.5)

2. RDFT
It has been show that the RDFT is very convenient in extracting a specific harmonic component from a given current \( i(t) \) by the following recursive algorithm (\( j \) is the order of the desired harmonic)

\[
h_j(k) = \frac{2}{N} (i(k) - i(k - N)) + W^j h_j(-1) \quad (A.6)
\]

which is driven by its equivalent transfer function \( G_j(z) \) between the input \( i(z) \) and the extracted harmonic signal \( h_j(z) \):

\[
G_j(z) = \frac{h_j(z)}{i(z)} = \frac{1}{N} \frac{1 - z^{-N}}{1 - W^j z^{-1}} \quad (A.7)
\]

in order to circumvent this problem, we design the following modulation operator by choosing a differentiable modulation function \( \phi(t) \), such that

\[
\frac{d^n}{dt^n} \phi(0) = 0 \quad (A9)
\]

Thanks to (A9), we design the following modulating function

\[
\phi_h = \frac{w_h t^{2n-h}}{(2n-h)!}, \quad h = \{1, 2, \cdots n\}
\]

APPENDIX II

A nominal state observer (i.e., Luenberger Observer) is applicable for the linear system (7). However, only asymptotic convergence is guaranteed (i.e., the estimated state \( \hat{z}(t) \rightarrow z(t) \), when \( t \rightarrow \infty \), because of the initial error driven by the unknown initial condition. In order to circumvent this restriction and to propose an observer that can converge within an arbitrary small finite time, we annihilate the influence of the unknown initial condition by exploiting an integral operator (see (18)) with respect to the designed modulation function. The next Lemma introduced in [27] is instrumental for the following description.

**Lemma 1 (Modulated signal’s derivative):** Consider a signal \( x(t) \), \( t \geq 0 \) that admits a \( i-th \) order derivative, and a \( i-th \) order differentiable modulation function \( \phi(t) \), \( t \geq 0 \). It holds that:

\[
[V_{\phi} x^{(i)}](t) = \sum_{j=0}^{i-1} (-1)^{i-j-1} x^{(j)}(t) \phi^{(i-j-1)}(t) + \sum_{j=0}^{i-1} (-1)^{j} x^{(j)}(0) \phi^{(i-j-1)}(0) + \phi(t)
\]

where we let the \( \frac{d^n}{dt^n} x(t) \) \( \phi(t) = x^{(i)}(t) \) for the sake of brevity and we define the modulation operator by

\[
[V_{\phi} x](t) \triangleq \int_0^t \phi(\tau) x(\tau) d\tau
\]

The above Lemma motivates us to eliminate the unknown initial conditions \( x(0), \frac{d}{dt} x(0), \frac{d^2}{dt^2} x(0), \cdots \) by choosing a modulation function \( \phi(t) \), such that

\[
\frac{d^n}{dt^n} \phi(0) = 0
\]

Differentiating both sides of (A1) with respect to time, we obtain

\[
\nu_{\phi_k}(k + 1) = \nu_{\phi_k}(k) + \frac{1}{2} T s \left( \left( \frac{d}{dt} \nu_{\phi_k}(k + 1) + 2 \phi_k(k + 1) \right) i(k + 1) \right), 0 \leq k < N_0
\]

\[
\nu_{\phi_k}(k + 1) = \nu_{\phi_k}(k) + \frac{1}{2} T s \left( \left( \frac{d}{dt} \nu_{\phi_k}(k + 1 - N_k + N_\Delta) \right) i(k + 1) \right), N_k < k < N_{k+1}
\]

\[
\nu_{\phi_k}(N_k^+) = E_h^T \Gamma(N_k) \Gamma(N_k + N_\Delta)^{-1} \nu(N_k), k = N_k.
\]

\[
\nu_{\phi_k}(k) = \nu_{\phi_k}(k) + 2 \phi_k(k) i(k), 0 \leq k < N_0
\]

\[
\nu_{\phi_k}(k + 1) = \nu_{\phi_k}(k) + \frac{1}{2} T s \left( \left( \frac{d}{dt} \nu_{\phi_k}(k + 1 - N_k + N_\Delta) \right) i(k + 1) \right), N_k < k < N_{k+1}
\]

\[
\nu_{\phi_k}(N_k^+) = E_h^T \Gamma(N_k) \Gamma(N_k + N_\Delta)^{-1} \nu(N_k), k = N_k.
\]
where the index $h$ allows a set of modulation functions in this form distinguished by different parameter $w_0$ and order $h$. Admittedly, the choice of the modulation function satisfying (A9) is not unique. For this application paper, we only show this typical function as a simple example. Although the modulation function plays an important role in this methodology, the selection does not vary from application to application; the modulation functions satisfying (A9) are applicable for the state estimation problem of any linear system in which the harmonic model can be embedded.

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