FRACTIONAL CALCULUS DEFINITIONS, APPROXIMATIONS, AND ENGINEERING APPLICATIONS

O. ELWY, A. M. ABDELATY, L. A. SAID, AND A. G. RADWAN

ABSTRACT

The basic idea behind fractional calculus is that it considers derivatives and integrals of non-integer orders giving extra degrees of freedom and tuning knobs for modeling complex and memory dependent systems with compact descriptions. This paper reviews fractional calculus history, theory, and its applications in electrical engineering. The basic definitions of fractional calculus are presented together with some examples. Integer order transfer function approximations and constant phase elements (CPEs) emulators are overviewed due to their importance in implementing fractional-order circuits and controllers. The stability theory of fractional-order linear systems is outlined and discussed. Four common electrical engineering applications are surveyed. Fractional-order oscillators allow controlling the phase difference, as well as achieving high oscillation frequency independently. Fractional order electronic filters are used to provide non-integer order slopes eliminate the need to round up the filter order and achieve the exact required time and frequency domain specifications. Studying fractional-order bioimpedance models provides better fitting to the measured data from fruits and vegetables. Fractional order DC-DC converter models provide a better estimation of the power conversion efficiency by incorporating frequency-dependent losses.

KEYWORDS: Fractional-order Circuits, Caputo, Fractional Calculus, Cole-Impedance Model, DC-DC Converters, Filters, Oscillators.

1. INTRODUCTION

Gottfried Leibniz and Guillaume L’Hôpital are the first ones to wonder about the existence of the fractional order derivative in the end of the seventeenth century [1]. Since then, many great mathematicians had developed a strong theoretical foundations
of the topic and, unfortunately, it remained within the boundaries of pure mathematics for a long time. The early contributors list include but not limited to: Leibniz, L’Hôpital, Bernoulli, Laplace, Euler, and Fourier [1]. Euler introduced the gamma function as a generalization of the factorial function and consequently generalized the formula of the $n$-th order derivative to the intermediate fractional orders. The first appearance of Riemann-Liouville fractional integrals and Caputo fractional derivative were presented in Abel’s approach to solve the tautochrone problem. The systematic formulation of fractional calculus theory began with Liouville during the mid 19$^{th}$ century followed by the contributions of Grunwald and Letnikov for the arbitrary order difference [1].

Fractional calculus and fractional-order modeling are areas of mathematics concerned with the treatment of the processes having non-integer differentiation and integration orders [2]. Despite the early start in developing the theoretical foundations of fractional calculus, it is only recently when researchers uncovered its strength in modeling many natural and complex phenomena. The application areas of fractional calculus and fractional-order modeling include but not limited to: control [3–6], chaotic systems [7–12], encryption [13–16], super-capacitor modeling [17–19], filters [20–26], differentiators and integrators [27], and bioengineering [28–30].

The remaining of this review is organized as follows: Section 2 introduces the basic definitions of fractional calculus with some examples. Section 3 discusses various integer order approximation techniques for the fractional Laplacian operator while in Section 4, different circuit topologies of passive fractance emulators are discussed. The stability analysis concepts in the fractional domain are summarized in Section 5. Four electrical engineering applications of fractional calculus are summarized in Section 6 including oscillators, filters, bio-impedance, and DC-DC converters.

2. FRACTIONAL CALCULUS DEFINITIONS

Fractional calculus is the branch of mathematics that deals with non-integer order differentiation or integration. This gives the generality feature to the ordinary calculus. The $n$-fold integration of $f(t)$ can be calculated according to Cauchy’s formula:
\[ D^{-n} f(t) = \frac{1}{(n-1)!} \int_0^t (t - \tau)^{n-1} f(\tau) d\tau \]  
\[ D^+ f(t) = f'(t) \]

Where \( D^- \) denotes the integration process and \( D^+ \) is the differentiation. This formula was generalized to a continuous one by Riemann and Liouville [2] using the continuous gamma function \( \Gamma(\alpha) \) instead of the discrete factorial as follows:

\[ D^{-\alpha} f(t) = \frac{1}{\Gamma(\alpha)} \int_0^t (t - \tau)^{\alpha-1} f(\tau) d\tau, \alpha > 0 \]  

which is known as the Riemann-Liouville (RL) fractional integral. There are several definitions of the fractional differentiation such as RL, Caputo, and Grünwald-Letnikov (GL) [2, 31]. The fractional differentiation calculation in terms of an integer derivative and an RL fractional integral. The RL definition of the fractional derivative uses the ceil function to achieve that \[ D^{\alpha}_{RL} f(x) = D^{\lceil \alpha \rceil}_{\alpha} D^{\lceil \alpha \rceil - \alpha} f(x). \]

Then, The overall operation includes two sub-operations. The first one is fractional-order integration, and the second one is integer order differentiation. The general formula of RL definition of fractional derivative can be written as follows:

\[ RLD^\alpha_{a} D_t f(t) = \frac{1}{\Gamma(m - \alpha)} \left( \frac{d}{dt} \right)^m \int_a^t (t - \tau)^{m-\alpha-1} f(\tau) d\tau, \quad (m - 1 \leq \alpha < m) \]  

This procedure results in non-integer order initial conditions which cannot be interpreted physically. The solution to this problem is the Caputo’s definition of the fractional derivative, which depends on the same previous idea but with interchanging the positions of the integer differentiation and fractional integration. Caputo’s definition is written as: \[ D^\alpha_C f(x) = D^{\lceil \alpha \rceil - \alpha} D^{\lceil \alpha \rceil} f(x), \]

where the integer-order differentiation occurs first. The definition of Caputo is as follows:

\[ C_a D^\alpha_t f(t) = \frac{1}{\Gamma(m - \alpha)} \int_a^t (t - \tau)^{m-\alpha-1} f^{(m)}(\tau) d\tau, \quad (m - 1 < \alpha < m) \]  

Caputo’s definition uses initial conditions of the integer derivatives which can be physically measured. While, the RL definition uses the initial conditions of the fractional derivatives which have no physical meaning and cannot be measured. There is another
difference with RL’s definition, the constant differentiation. Caputo gives the ordinary answer by zero, while RL gives a solution as follows:

\[ RL_0 D_0^\alpha C = \frac{C t^{-\alpha}}{\Gamma(1 - \alpha)}, \quad RL_0 D_0^\alpha 0 = 0, \]  

(4)

where \( C \) is the constant [2]. The RL differ-integral of a constant \( C \) and a sinusoid \( f(t) = \sin(2t) \) for \(-1 < \alpha < 1\) are illustrated in Fig. 1.

Grünwald-Letnikov (GL) definition is much similar to the definition of the classical \( n \)-th order derivative. There is a proven correspondence between GL and RL definitions [2]. For every \((0 < \alpha < n)\), RL definition exists and coincides with GL definition, if \(0 \leq m - 1 \leq \alpha < m \leq n\). This equivalence occurs under the conditions:

1. The function \( f(t) \) is \((n - 1)\) times continuously differentiable in the targeted interval,

2. \( f^{(n)}(t) \) is integrable in the targeted interval.

The GL derivative, in the interval \([a, t]\) is defined as follows:

\[ GL_n D_n^\alpha f(t) = \lim_{h \to 0} \frac{1}{h^\alpha} \sum_{j=0}^{[(t-a)/h]} \omega_j^{(\alpha)} f(t - jh), \]  

(5)

where \( \omega_j^{(\alpha)} \) are the binomial coefficients, which can be calculated as: \( \omega_j^{(\alpha)} = (1 - \frac{\alpha+1}{j}) \omega_{j-1}^{(\alpha)}, \quad \omega_0^{(\alpha)} = 1, \quad j = 1,2,3,\ldots \). This approach gives the opportunity to use RL derivative during the problem formulation, and going back and forth to GL derivative to obtain numerical solutions. This definition is not interval bounded. Getting an accurate
result requires the number of terms to be approaching to $\infty$, or at least to be a very large number.

3. INTEGER ORDER APPROXIMATION OF THE FRACTIONAL ORDER LAPLACIAN OPERATOR

The behavior of fractional linear systems is mimicked using either traditional integer transfer functions or digital filters. These approximations are important because [32]: commercial off-the-shelf constant phase elements are not available, reliable and well known simulation software are based on integer order calculus. So, having integer approximation of fractional order systems facilitates the use of these software tools, and synthesis of fractional order circuit is, in fact, based on integer transfer functions.

Discrete approximations of $s^\alpha$ can be converted into continuous approximations and continuous ones into discrete ones. Since digital approximations normally perform worse than continuous ones, the first alternative is seldom used.

Laplace transform is a mathematical operation that converts a function of a real variable $(t)$ into a function of a complex variable $(s)$. The Laplace transform of the any classical fractional derivative ($F_1$ class) under zero initial condition is [2, 31]:

$$L\{0D_t^\alpha f(t)\} = s^\alpha F(s),$$

which can be translated to a one-to-one correspondence between derivative term and the Laplacian operator ($\frac{d^\alpha}{dt^\alpha} \longleftrightarrow s^\alpha$), where $s = j\omega$. For the case of $s^\alpha$, the magnitude is a linear line plotted with the logarithmic frequency axis as shown in Fig.2a. The phase angle has a single value forming a horizontal line as shown in Fig.2b. Both of the magnitude and phase obey the following equations:

$$Magnitude(s^\alpha) = 20\alpha \log(\omega), \quad Phase(s^\alpha) = \frac{\alpha\pi}{2}.$$  \hspace{1cm} (7)

This property allows representing Laplacian operator-based transfer function using Bode diagrams [33]. They have two versions: one is the magnitude and the other is the phase angle. Both of these plots have logarithmic frequency x-axis. The magnitude and phase
Fig. 2. Bode diagrams of $s^\alpha$ (a) Magnitude of $s^\alpha$ for any $\alpha$ and (b) Phase of $s^\alpha$ for any $\alpha$. are calculated for a given transfer function $H(j\omega)$ as follows:

$$H(j\omega) = \frac{z(j\omega)}{p(j\omega)},$$  \hspace{1cm} (8a)

$$\text{Magnitude}(H(j\omega)) = 20\log|H(j\omega)|,$$  \hspace{1cm} (8b)

$$\text{phase}(H(j\omega)) = \angle H(j\omega) = \tan^{-1}\left(\frac{\text{Im}\{H(j\omega)\}}{\text{Re}\{H(j\omega)\}}\right) = \angle z(j\omega) - \angle p(j\omega),$$  \hspace{1cm} (8c)

where $z(j\omega)$, and $p(j\omega)$ are the zeros and poles of the transfer function respectively. Figure 2a and 2b show the effects of zeros and poles on the phase of a transfer function. Zero would take the phase angle higher, while pole would pull it down. The overall phase angle is the summation of these ups and downs as indicated in Eq. (8c). Because of the advances made on integer-order analysis, researchers tended to approximate $s^\alpha$ in terms of integer order $s$. The trick of integer order approximation techniques is to find a way to distribute/interlace the poles and zeros to obtain the desired response by zigzagging about the ideal fractional order response. The desired output should follow Eq. (7). Therefore, the approximation technique should try to find a representation that oscillates around $\alpha\pi/2$ using the angles’ properties of poles and zeros. Also, the magnitude should preserve the presented relation in Eq. (7) and Fig.2a.

Many techniques were introduced to obtain distribution of zeros and poles. There are two categories of approximations according to [34]. The first category is named continued fraction expansions and interpolation techniques such as the approximations of Matsuda and Carlson. The second category is characterized by curve fitting and identification techniques such as Oustaloup. A summary of the design flow charts of
four approximation techniques is provided in Table 1. The flowcharts show that all these approximation techniques provide control of the approximation order $N$ and the fractional order $\alpha$. However, only Oustaloup and Matsuda have the operating frequency range as their input parameters. Additionally, the effects of different parameters on each approximation are depicted in Table 2. For Carlson and CFE, the operating frequency range is enlarged by increasing $N$. While for Oustaloup and Matsuda, at a constant operating frequency range, the phase and magnitude errors decrease by increasing $N$. By varying $\alpha$, it is evident that Carlson approximation is acceptable only at $\alpha = 0.5$.

4. FRACTIONAL-ORDER CAPACITOR EMULATORS

Resistors, capacitors, and inductors resemble the three basic circuit elements, and they are related through the formula $Z(s) = ks^\alpha$, where $k$ is a coefficient that contributes to the impedance magnitude value and $\alpha$ is the order. For $\alpha = -1, 0, 1$, the relation represents the traditional capacitor, resistor, and inductor, respectively. Moreover, Frequency Dependent Negative Resistor (FDNR) [35] follow this generalized equation as shown in Fig. 3. Intermediate circuit elements appear in Fig.3 for non-integer values of $\alpha$. These devices are called fractances or Constant Phase Elements (CPE), due to their nature of constant phase at $\alpha \pi /2$. The absence of off-the-shelf fractional order capacitors (FOC) encourages the researchers to find some alternatives. Many trials were introduced to emulate their behavior through either active or passive circuits.

Examples of passive CPE emulators are found in [36–38] while examples of active CPE emulators are found in [39–41]. Although active emulators allow more tunability, passive

![Fig. 3. Conventional elements and their relation to the fractional ones.](image-url)
circuits based on RC networks are much cheaper. The passive emulators are obtained through two approaches, the first is a direct circuit approximation such as [36,38,42–44]. The second approach is through using integer order approximations of $s^\alpha$, and converting them to a network of Lumped elements. The conversion can be done to yield circuits in the form of Foster-I, Foster-II, Cauer-I, or Cauer-II [45].

4.1 Emulators of $\alpha = 0.5$

The work of Roy [36] is one of the earliest approaches to build a circuit realization for FOC. Three different realizations were introduced, which are shown in Table 3. These designs are for approximating the order $\alpha = 0.5$ only. Later on, Nakagawa proposed another network [46]. Nakagawa employed the idea of the geometric mean to obtain a FOC’s emulator in the shape of a self similar RC tree as shown in Table 3. Increasing the number of stages enlarges the operating frequency range, but it’s not cost effective.

4.2 Generic Passive Emulators

Sugi proposed a network that emulates the behavior of FOC based on Distributed Relaxation Time models [37]. The current of a FOC can be expressed in terms of a superposition decay processes. The proposed two topologies are shown in Table 3. Valsa proposed another methodology [38,44], and the networks are shown in Table 3. This method is analogous to the work of [42,47]. The advantage of Valsa is the ability of controlling the phase variation.
Table 1. Flowcharts depicting the process of four approximation techniques.

<table>
<thead>
<tr>
<th>Carlson</th>
<th>Matsuda</th>
<th>Oustaloup</th>
<th>Krishna (CFE)</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\alpha$: fractional-order</td>
<td>$\omega_{\text{min}}$: minimum frequency</td>
<td>$\omega_{\text{max}}$: maximum frequency</td>
<td>$\alpha$: fractional-order</td>
</tr>
<tr>
<td>$H_0(s) = 1$</td>
<td>$H(s) = s^\alpha$</td>
<td>$H(s) = \frac{\omega_{\text{min}}}{\omega_{\text{max}}} (2i-1-\alpha)/2N$</td>
<td>$H(s) = \frac{1}{1 - \frac{\alpha(s-1)}{\omega_{\text{max}} \omega_{\text{min}}}}$</td>
</tr>
<tr>
<td>$\alpha = \frac{m}{p}$, $k = \frac{p-m}{p+m}$</td>
<td>$d_0(\omega) =</td>
<td>H(\omega)</td>
<td>$</td>
</tr>
<tr>
<td>$H_i = H_{i-1}(s) s + k</td>
<td>H_{i-1}(s)</td>
<td>^2$</td>
<td>$d_i(\omega) = \frac{d_{i} - d_{i-1} \omega_{i-1}}{\omega_{i} - \omega_{i-1}}$, $i = 1, 2, ..., 2N$</td>
</tr>
<tr>
<td>$H_{\infty} = H_0(s) + \sum_{i=1}^{2N} \frac{S}{\omega_{z,i}} + \sum_{i=1}^{2N} \frac{S}{\omega_{p,i}}$</td>
<td>$H(s) = 1 - \frac{\alpha(s-1)}{\omega_{\text{max}} \omega_{\text{min}}}$</td>
<td>$\omega_{p,i} = \omega_{\text{min}} \omega_{\text{max}} (2i-1+\alpha)/2N$</td>
<td>$H(s) = \frac{1}{1 + \frac{(1 + \alpha)(s-1)}{2 + (1 - \alpha)(s-1)}}$</td>
</tr>
<tr>
<td>End</td>
<td>End</td>
<td>End</td>
<td>End</td>
</tr>
</tbody>
</table>

**Notes:**
- $\omega_{\text{min}}$: minimum frequency
- $\omega_{\text{max}}$: maximum frequency
- $N$: approximation order
- $\alpha$: fractional-order
- $N$: No. of iterations
Table 2. The magnitude and phase responses of four approximation techniques.

<table>
<thead>
<tr>
<th></th>
<th>Carlson</th>
<th>Matsuda</th>
<th>Oustaloup</th>
<th>Krishna (CFE)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Magnitude (N)</strong></td>
<td><img src="image1" alt="Graph" /></td>
<td><img src="image2" alt="Graph" /></td>
<td><img src="image3" alt="Graph" /></td>
<td><img src="image4" alt="Graph" /></td>
</tr>
<tr>
<td><strong>Phase (N)</strong></td>
<td><img src="image5" alt="Graph" /></td>
<td><img src="image6" alt="Graph" /></td>
<td><img src="image7" alt="Graph" /></td>
<td><img src="image8" alt="Graph" /></td>
</tr>
<tr>
<td><strong>Magnitude (α)</strong></td>
<td><img src="image9" alt="Graph" /></td>
<td><img src="image10" alt="Graph" /></td>
<td><img src="image11" alt="Graph" /></td>
<td><img src="image12" alt="Graph" /></td>
</tr>
<tr>
<td><strong>Phase (α)</strong></td>
<td><img src="image13" alt="Graph" /></td>
<td><img src="image14" alt="Graph" /></td>
<td><img src="image15" alt="Graph" /></td>
<td><img src="image16" alt="Graph" /></td>
</tr>
</tbody>
</table>
Table 3. Examples of passive emulator circuits for different $\alpha$ values.

<table>
<thead>
<tr>
<th>Emulators of $\alpha = 0.5$</th>
<th>Emulators of any $\alpha$.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Equations</strong></td>
<td><strong>Diagram</strong></td>
</tr>
<tr>
<td>$R_n = \frac{R}{2^n}$, $C_n = \frac{C}{2^n}$</td>
<td><img src="image1" alt="Diagram" /></td>
</tr>
<tr>
<td>$R_n = (4n - 5)R$, $C_n = (4n - 3)C$</td>
<td><img src="image2" alt="Diagram" /></td>
</tr>
<tr>
<td>$R_n = \frac{R}{4n-1}$, $C_n = \frac{C}{4n-3}$</td>
<td><img src="image3" alt="Diagram" /></td>
</tr>
<tr>
<td>$Z = \left( \frac{R}{\alpha} \right)^{0.5} \omega^{-0.5} e^{-\frac{\alpha}{4}}$</td>
<td><img src="image4" alt="Diagram" /></td>
</tr>
<tr>
<td>Nakagawa [46]</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
</tr>
</tbody>
</table>
5. STABILITY

Systems can be modeled using differential equations which can be transformed to its counterpart in \( s\)-domain using Laplace transform. Checking the system stability is one of the advantages of this transformation. The characteristic equation of a commensurate order fractional order system is \( \sum_{k=0}^{N} a_k s^{k \alpha} = 0 \). Stability analysis is a discussion about the location of the poles of the characteristic equation and how they affect the time domain response of the system. The response is required to be bounded for any bounded excitation signal.

A new domain was introduced in [48] known as the \( W\)-plane. This method is for the rational powers, which can be written in the form \( \alpha = k/m \), where \( k \) and \( m \) are positive integers. Mapping the roots in \( s\)-domain to \( W\)-plane, requires introducing \( W = s^{1/m} \), which makes the mapping process independent on \( k \). The unstable region is at \(|\theta_W| < \pi/2m\). The stable region is when \(|\theta_W| > \pi/2m\), which includes physical and non-physical roots. Physical roots have correspondence to the \( s\)-plane while non-physical ones don’t. The stability regions of the \( W\)-plane are illustrated in Fig.4.

A general procedure for analyzing the stability of a linear fractional-order differential equation was introduced in [48] for equations on the following form:

\[
\sum_{k=0}^{N} a_k s^{k/m} = \sum_{k=0}^{N} a_k W^k = 0,
\]

where \( W = s^{k/m} \), and it is a polynomial of order \( N \). The first step is to calculate the roots of

\[
\text{Re}(j\omega) > 0 \quad \text{stable}
\]

\[
\text{Im}(j\omega) > 0 \quad \text{oscillatory}
\]

\[
\text{Im}(j\omega) < 0 \quad \text{unstable}
\]

\[
\text{Re}(j\omega) < 0 \quad \text{unstable}
\]

\[
\frac{\pi}{2m} < |\theta_W| < \frac{\pi}{2m} \quad \text{stable}
\]

\[
\frac{\pi}{2m} < |\theta_W| < \frac{\pi}{2m} \quad \text{unstable}
\]

Fig. 4. The stability regions for \( s^\alpha \) (a) conventional \( s\)-plane at \( \alpha = 1 \) and (b) the \( W\)-plane.
Table 4. Summary for the cases of quadratic equations.

<table>
<thead>
<tr>
<th>Relation</th>
<th>condition</th>
<th>physical roots</th>
</tr>
</thead>
<tbody>
<tr>
<td>$b &lt; 0$ or $(a^2 \geq b$ and $a &lt; 0)$</td>
<td>unstable independent on $\alpha$</td>
<td>$\alpha &lt; 1$</td>
</tr>
<tr>
<td>$a^2 \geq b$ and $a &gt; 0$ and $b &gt; 0$</td>
<td>stable if $\alpha &lt; 2$</td>
<td>$\alpha &lt; \frac{\pi}{2}$</td>
</tr>
<tr>
<td>$a^2 &lt; b$ and $a &gt; 0$ and $b &gt; 0$</td>
<td>stable if and only if $\alpha &lt; \frac{2\pi}{\pi}$, $\alpha &gt; \frac{\pi}{\pi}$</td>
<td>$\alpha &lt; \frac{\pi}{2}$</td>
</tr>
<tr>
<td>$\delta = \cos^{-1}(-\frac{a}{\sqrt{b}}) &gt; \frac{\pi}{2}$</td>
<td></td>
<td></td>
</tr>
<tr>
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</table>

Eq. (9) for given $a_k$. Then, finding the minimum absolute phase for all roots $|\theta_{W_{\text{min}}}|$. The stable region is when $(|\theta_{W_{\text{min}}}| > \pi/2m)$, the system is oscillatory at $(|\theta_{W_{\text{min}}}| = \pi/2m)$, and it is unstable otherwise. Table 4 shows a summary for the conditions of stability, as well as the physical roots existence condition for the cases as introduced in [48] for system having a characteristic equation of the form $s^{2\alpha} + as^{\alpha} + b = 0$.

For the case of fractional-order characteristic equation, the procedure of [48] is applied and validated for the following cases and summarized in Table 5(1) $s^{2\alpha} + 4s^{\alpha} + 1$ and (2) $s^{2\alpha} - 4s^{\alpha} + 1$, for $0 < \alpha \leq 2$. This representation employs the same $1/m$ to show how the stability is affected by the sign of parameters. The theory of fractional-order stability using W-plane concept [48] is a milestone in this research area. It is one of the grounding concepts and many researchers make use of it in many applications.

6. APPLICATIONS

6.1 Oscillators

Oscillators are widely used in many applications [49] such as: communications (modulation and demodulation), generating clock pulses for microprocessors and microcontrollers, testing and measurements, times and clocks, signal generators, alarms and buzzers. Fractional-order oscillator circuits have two main types according to the nature of the generated signal: sinusoidal and relaxation. Introducing the fractional-order permits an extra design degree of freedom.

The theory of fractional-order sinusoidal oscillators was presented in [50]. It included the design procedure of oscillator of any number of fractance devices. Many fractional-
order oscillators were reported in literature based on the introduced theory such as [51–56]. All possible fractional order oscillators depending on the two-port network concept was proposed in [51,54,57]. Nine possible oscillators were investigated with their mathematical formulas in [58]. Another work showed the design procedure of sinusoidal oscillator using differential voltage current conveyors (DVCC) [59]. A general procedure for designing an oscillator with a specific phase and frequency was also included in [53].

The general state-space formula of the Wien-bridge oscillator shown in Fig. 5a was proposed in [50] and can be written as follows:

\[
\begin{pmatrix}
D^\alpha V_{C1} \\
D^\beta V_{C2}
\end{pmatrix} = \begin{pmatrix}
\frac{a-1}{R_1 C_1} & \frac{-1}{R_1 C_1} \\
\frac{a-1}{R_2 C_2} & \frac{-1}{R_2 C_2}
\end{pmatrix}\begin{pmatrix}
V_{C1} \\
V_{C2}
\end{pmatrix} + \begin{pmatrix}
\frac{h}{R_1 C_1} \\
\frac{h}{R_2 C_2}
\end{pmatrix},
\]

(10)

where \(\alpha\) and \(\beta\) are the fractional-order differentiation. The values of \((a,b)\) are characterized as: \((a,b) = (0, V_{sat})\) when \(KV_{C1} \geq V_{sat}\), \((K,0)\) when \(-V_{sat} < KV_{C1} < V_{sat}\), and \((0,-V_{sat})\) when \(-V_{sat} \geq KV_{C1}\) where \(V_{sat}\) is the saturation voltage of the employed operational amplifier and \(K\) is the gain factor, and its value is \(1 + \frac{R_3}{R_4}\). Subsequently, the
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The characteristic equation can be written as:

\[ s^{\alpha+\beta} + \left(\frac{K - 1}{R_2C_1} - \frac{1}{R_1C_1}\right)s^\beta + \left(\frac{1}{R_2C_2}\right)s^\alpha + \left(\frac{1}{R_1R_2C_1C_2}\right) = 0. \]  
\( \text{(11)} \)

The following procedure is substituting each \( s \) by \( j\omega \), then separating the real and imaginary equations. Thus, Eq. (11) turns into:

\[ \omega^{\alpha+\beta}\cos\left(\frac{(\alpha + \beta)\pi}{2}\right) - \left(\frac{K - 1}{R_2C_1} - \frac{1}{R_1C_1}\right)\omega^\beta\cos\left(\frac{\beta\pi}{2}\right) + \left(\frac{1}{R_2C_2}\right)\omega^\alpha\cos\left(\frac{\alpha\pi}{2}\right) + \left(\frac{1}{R_1R_2C_1C_2}\right) = 0, \]  
\( \text{(12)} \)

\[ \omega^\beta\sin\left(\frac{(\alpha + \beta)\pi}{2}\right) - \left(\frac{K - 1}{R_2C_1} - \frac{1}{R_1C_1}\right)\omega^\beta\sin\left(\frac{\beta\pi}{2}\right) + \left(\frac{1}{R_2C_2}\right)\omega^\alpha\sin\left(\frac{\alpha\pi}{2}\right) + \left(\frac{1}{R_1R_2C_1C_2}\right) = 0, \]  
\( \text{(13)} \)

Sustained sinusoidal oscillation is achieved if solving Eqs. (12, 13) results in a real value for oscillation frequency. Moreover, the oscillation condition \( K \) is the value at

<table>
<thead>
<tr>
<th>Table 6. Design parameters of Wien oscillator</th>
</tr>
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<tbody>
<tr>
<td>case</td>
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<td>( \alpha = \beta \neq 1 )</td>
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<td>( \alpha = \beta \neq 1 )</td>
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<td>( R_1 = R_2 = R )</td>
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<td>( \alpha = \beta = 1 )</td>
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Fig. 5. (a) Wien-bridge oscillator, and (b) Simulation results.
which this real solution exists. This solution scheme starts by excluding the term $K$ from the two formulas of Eqs. (12,13). Consequently, the resultant equation contains only one unknown, the oscillation frequency ($\omega_{osc}$). Due to the difficulty of having closed formulas for the oscillation condition and frequency, some special cases are summarized in Table 6.

According to Table 6, the oscillation frequency gets higher values for $\alpha < 1$. For example, the case of equal $R$, $C$, and $\alpha$, the integer order Wien oscillator whose values ($R = 1 \text{k}\Omega$, and $C = 100 \text{nF}$), has oscillation frequency $\omega = 10 \text{krad/sec}$. In the case of employing fractional-order capacitor, utilizing the same components’ values despite of the order $\alpha = 0.8$, the resultant frequency is $100 \text{krad/sec}$, which is 10 times the integer order case. The simulation of this fractional-order case is depicted in Fig. 5b, where the dotted and solid lines are $V_{c1}$ and $V_{c2}$, respectively. Moreover, the phase difference is more controllable in case of fractional-order oscillator as shown in Table 6.

### 6.2 Filters

The first systematic analysis of fractional order filters was introduced in [60,61], then many studies followed. The fractional-step Tow-Thomas filter was studied in [62] where the filter topology was generalized to realize band-pass and low-pass filters based on CFE approximation. The detailed analysis of the fractional order Butterworth filter was carried out in [63]. Also, the least square optimization algorithm was used in [64] to approximate the stop-band behavior of fractional order inverse Chebyshev filter. In [21], steps for the realization of fractional-order complex Chebyshev filter were introduced. In [65], the fractional order low-pass, band-pass and high-pass inverse filters were investigated using different approximation techniques. A design procedure for the fractional order Chebyhshev filter having the same poles as the integer order ones was presented in [66]. Fractional-order filters have an extra degree of freedom compared to the conventional ones.

There are some critical frequencies to be calculated for fractional order filters:

- Cutoff(half power) frequency ($\omega_c$): at which the power drops to half the pass band
Table 7. Important frequencies of FO BP filter.

| ω  | |T(jω)| | ∠T(jω) |
|----|---------------------------------|
| 0  | 0                               | 2π     |
| ω₀ | \(\frac{b^{\beta/\alpha}}{2a} \left(\frac{\pi}{2}\right)\) | \(\frac{\beta\pi}{2} - \frac{\alpha\pi}{4}\) |
| at \(\infty\) | \(b\omega^{(\beta-\alpha)}\) | \(\frac{(\beta-\alpha)\pi}{2}\) |
| ωₘ | \(\frac{d}{\sin(\frac{\pi}{2})}\) | \(\frac{(1-\alpha)\pi}{2}\) |
| ωₗ | \(\frac{d}{\alpha\sqrt{2}}\) \(\tan^{-1}\left(\frac{\sin(\frac{\pi}{2})}{2\cos(\frac{\pi}{2})+\sqrt{1+\cos^2(\frac{\pi}{2})}}\right)\) | \(\frac{\pi}{2}\) |

\[|T(j\omega)| = \frac{|T(j\omega_{\text{pass band, } a, b, \alpha})|}{\sqrt{2}}\]

- Maximum frequency (\(\omega_m\)): at which the magnitude has a maximum and it is obtained by solving \(\frac{d}{d\omega}|T(j\omega)|_{\omega=\omega_l} = 0\), and it can be solved for a certain maximum value.

- Right phase frequency (\(\omega_{rp}\)): at which the phase \(\angle T(j\omega_{rp}) = \pm \frac{\pi}{2}\), and the transfer function is pure imaginary.

An example of simple fractional-order Band Pass Filter (BPF) of the following transfer function has the magnitude response shown in Fig. 6: \(T_1(s) = \frac{b^{\beta}}{s^\alpha + a}\). Some important frequency values are summarized on Table 7. In case of \(\beta < \alpha\), \(\lim_{\omega \to \infty} |T(j\omega)| = 0\). Therefore, the filter acts as a Band Pass Filter. The other case of \(\beta = \alpha\) leads to \(\lim_{\omega \to \infty} |T(j\omega)| = b\). Thus, the filter develops into a High Pass Filter. In case of \(\beta = \alpha = 2\), the center frequency \(\omega_o\) equals to the maxima frequency. Figure 6 depicts this idea for an example of BPF of equal \(a\) and \(b\). The three aforementioned cases are characterized for constant \(\alpha = 0.9\) and three different \(\beta\) values at \((B = 0, \alpha/2, \alpha)\).

![Graph showing magnitude response of various BPFs with different values of \(\beta\).](image-url)
6.3 Bio-impedance

Bio-impedance is the biological tissue resistance to an applied electrical excitation as shown in Fig. 7 [30]. It is mainly affected by the cell shape and the structure of its membrane [67]. Measuring bio-impedance is a vital indication for any change in cells’ structure. It has many applications in different areas such as in medicine [68]. Moreover, it is employed in food industries, to evaluate the state of the tested item. Measuring the maturity or estimating lifespan for storage purposes of fruits and vegetables using the electrical impedance characteristics as achieved in [69]. Moreover, it is employed in monitoring the effects of drying and freezing/thawing treatments on eggplants [70].

The application of bio-impedance rely on the electrical modeling of the tissue cell. There are many introduced models in the literature trying to emulate the most real response. The circuit elements in each topology mimics a component of the tissue cell. The fractional-order based models showed more accurate results compared to its integer counterparts [30,71]. Table 8 shows a summary of some fractional-order circuit models the tissue cell.

The single dispersion model was employed in many applications such as assessing the quality of stored red blood suspension [68]. Measuring the maturity or estimating lifespan for storage purposes of fruits and vegetables using the electrical impedance characteristics is achieved in [69,72]. The double dispersion Cole model is found in many applications such as the characterization of intestinal tissue excised from sheep [73]. The age-related

Fig. 7. (a) Bio-impedance measurement schematic. (b) SP150 impedance analyzer experimental setup. (c) Portable impedance analyzer.
Table 8. Examples of fractional-order bio-impedance models.

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<tr>
<td>Single dispersion [75]</td>
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<tr>
<td></td>
<td>$R_0 = 3.89510^4 \Omega$</td>
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<tr>
<td></td>
<td>$R_\infty = 544.4\Omega$</td>
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<tr>
<td></td>
<td>$C = 1.005710^{-7} F \cdot \sec^{\alpha-1}$</td>
<td>$\alpha = 0.629$</td>
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<tr>
<td>Simplified Hayden [71]</td>
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<tr>
<td></td>
<td>$R_1 = 3.89510^4 \Omega$</td>
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</tr>
<tr>
<td></td>
<td>$R_3 = 552.1\Omega$</td>
<td></td>
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<tr>
<td></td>
<td>$C = 97.7810^{-7} F \cdot \sec^{\alpha-1}$</td>
<td>$\alpha = 0.629$</td>
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<tr>
<td>Double shell [71]</td>
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<tr>
<td></td>
<td>$R_1 = 71.55k\Omega$</td>
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<td></td>
<td>$R_3 = 181k\Omega$</td>
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<td></td>
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<tr>
<td></td>
<td>$C_1 = 1.465510^{-4} F \cdot \sec^{\alpha-1}$</td>
<td>$\alpha = 0.656$</td>
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<tr>
<td></td>
<td>$C_2 = 1.8810^{-4} F \cdot \sec^{\beta-1}$</td>
<td>$\beta = 0.74$</td>
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Changes of dentine was investigated for potential nondestructive dental tests in [74]. The fractional-order simplified Hayden and double shell models were proposed recently [71].

Electro-chemical Impedance Spectroscopy (EIS) is the most widely known measuring methodology. The basic setup and connections of EIS are depicted in Fig. 7. It involves exposing the targeted tissue as a black box to a wide range of frequencies. Then, optimization algorithms are employed to fit the results to the best model [76]. There are many measuring techniques such as transient time measurements, where a step function of voltage is applied to the targeted impedance, then applying Fast Fourier Transform (FFT) on the the resultant current [77].

Most commercial impedance analyzers such as the SP150 (see Fig. 7) and their associated programs have built-in function that allows the user to fit the measurement either to well-known impedance models or even a user-defined one. The common fitting technique is the complex nonlinear least squares technique (CNLS). Although it might be satisfactory in some cases, when the model gets more complicated it gives un-acceptable fits. This is due to the fact that CNLS is a gradient based optimization technique,
also known as deterministic technique, which is prone to local minima and measurement outliers. Recently, researchers used meta-heuristic optimization algorithms to identify the parameters of many bio-impedance models. They mimic the searching/hunting behavior of animals in nature in order to find the global optimal solution and are less likely to fall into local minima of the optimization problem due to a balance between intensification and diversification of their search agents. Examples of these algorithms include flower pollination algorithm (FPA), grey wolf optimizer (GWO), moth flame optimizer (MFO), grasshopper optimization algorithm (GOA), whale optimization algorithm (WOA).

Three items must be defined in any optimization problem. These items are the search vector, the objective function, and the feasible region. In parameter estimation of bio-impedance models, the search vector has a dimension equal to the number parameters to be identified. For example, in case of single dispersion model, the parameters are $R_0$, $C_\alpha$, $R_\infty$, and $\alpha$, so the dimension of the search vector is 4 and is written as $X = [R_0, C_\alpha, R_\infty, \alpha]$.

For the objective function, the common one in literature is the sum of absolute error given as $SAE = \sum_{i=1}^{M} |Z_i - \tilde{Z}_i|$, where $M$ is the number of data-points, $Z_i$ is the measured complex impedance at frequency $\omega_i$, $\tilde{Z}_i$ is the estimated impedance at frequency $\omega_i$ and is calculated from the model formula as $Z(s) = R_\infty + \frac{R_0 - R_\infty}{1 + s^\alpha (R_0 - R_\infty) C}$.

The feasible region is described through implying constraints on the search vector variables. For example, in case of identifying the model parameters of Lemon, the lower limits are defined as $lb = [10k, 1n, 1, 0.4]$, and the upper limits are defined as $ub = [500k, 1\mu, 10k, 1.0]$. The same formulation can be made for identifying the fractional order Hayden model and the fractional order double-shell model.

When comparing the performance of different meta-heuristic optimization techniques at solving this problem, three main aspects are defined which are accuracy and consistency, objective function final value, and run-time. The accuracy and consistency means the algorithm achieves almost the same results in each independent run. This is measured from the standard deviation of the results of all independent runs made on the same PC. Lower standard deviation means more accurate and consistent algorithm. It was found that FPA is the most consistent in solving this problem. The run-time needed
to get the results depends on two main factors, the convergence speed of the algorithm, and the number of calculations per iteration of the algorithm. Unfortunately, usually meta-heuristic optimization techniques takes longer run-times than gradient based algorithms like CNLS and nonlinear least squares (NLS). This means that there is a trade-off between speed and accuracy in this case. The fit results of identifying the bioimpedance parameters of lemon are shown in Table 8.

### 6.4 DC-DC Converters

In the ideal case, the energy storage elements are lossless and the efficiency of the system is 100%. However, in practicality, the coils have many types of losses. One of these types is the hysteresis loss which results from the use of ferromagnetic cores that retain magnetization and increases both the inductance and losses which were found to be frequency dependent [78]. This is due to the fact that the coils in power converter operate near their magnetic saturation levels where the magnetic losses and skin effect can not be modeled linearly. Fractional order models can be used here to provide an accurate emulation of such losses associated with inductors inside DC-DC converter operating conditions.

The circuit schematic of the Buck DC-DC converter topology is illustrated in Fig. 8. The converter has two switches, one inductor and one capacitor. Switch $S_1$ is usually is the active bidirectional switch while switch $S_2$ is the passive uni-directional switch. In continuous conduction mode (CCM), the circuit has only two states: when $S_1$ is on and $S_2$ is off, and when $S_1$ is off and $S_2$ is on. In discontinuous conduction mode (DCM), another state is added which is both switches are off. Analysis of fractional order DC-DC converter in literature is either concerned with the steady state behavior [79] or the transient behavior [80].

The generalized procedure for steady state analysis of fractional order DC-DC converter in CCM mode can be summarized as follows. Assume that, at steady state, the voltage across the fractional-order inductor is constant during each switching state and changes accordingly between $v_1$ in the interval $[nT, (n + D)T]$ to $v_2$ in $[(n + D)T, (n + 1)T]$.
Fig. 8. (a) Buck converter. (b) Gain vs duty cycle. (c) The efficiency.

where \( n \in \mathbb{Z}^+ \). Where \( v_1 \) is the voltage across the inductor when the active switch is ON and the passive switch is OFF (diode is reverse biased), while \( v_2 \) is the inductor voltage when the active switch is OFF and the passive switch is ON (diode is forward biased).

Another assumption is that the capacitor ripple voltage is negligible. Also, assume that the inductor current has stabilized which means no rising or falling in its average value, so the inductor current value at \( t = nT \) will be the same and is equal to \( i_{L2}(T, \alpha) \). For simplicity, we will write the equations of the current waveforms in the first period \((n = 0)\).

Therefore, the current \( i_L(t, \alpha) \) is given by: in the interval \([0, DT]\)

\[
i_{L1}(t, \alpha) = i_{L2}(T, \alpha) + \frac{1}{L \Gamma(\alpha)} \int_0^t (t-\tau)^{\alpha-1} v_1 d\tau = i_{L2}(T, \alpha) + \frac{v_1 t^\alpha}{L \Gamma(\alpha + 1)}. \tag{14}
\]

In the interval \([DT, T]\)

\[
i_{L2}(t, \alpha) = i_{L2}(T, \alpha) + \frac{1}{L \Gamma(\alpha + 1)} (v_1 t^\alpha + (v_2 - v_1)(t - DT)^\alpha). \tag{15}
\]

By setting \( t = T \) in equation (15), the relation between \( v_1 \) and \( v_2 \) becomes:

\[
v_1 T^\alpha + (v_2 - v_1)(T - DT)^\alpha = 0 \quad \Rightarrow \quad \frac{v_1}{v_2} = \frac{(1 - D)^\alpha}{(1 - D)^\alpha - 1} \quad \Rightarrow \quad D = 1 - \left( \frac{v_1}{v_2} \right)^\frac{1}{\alpha}. \tag{16}
\]

which is the fractional equivalent of the inductor volt-second balance.

In case of the fractional order buck converter, by substituting \( v_1 \) and \( v_2 \) with \((v_{\text{in}} - v_{\text{out}})\) and \((-v_{\text{out}})\), respectively in the previous equation, the duty cycle \( D \) and voltage gain \( G \)
Fractional Calculus Definitions, Approximations, ...

can be written as:

\[ D = 1 - \left( \frac{v_{in} - v_{out}}{v_{in}} \right)^{\frac{1}{\alpha}}, \quad G = \frac{v_{ao}}{v_{in}} = 1 - (1 - D)^{\alpha}. \]  

(17)

Note that when \( D = 1 \), then \( v_{out} = v_{in} \) for any value of \( \alpha \). It is evident that \( D \) increases for smaller values of \( \alpha \) to compensate for the inductor fractional losses. The relation between the gain \( G \) and the duty cycle \( D \) at different values of the fractional order \( \alpha \) is shown in Fig. 8b.

The efficiency is calculated as the ratio between average energy dispatched to the load and average energy taken out of the source and assuming the steady state operation condition which means that the input and output voltages variations are negligible. Therefore, the average power efficiency in this case is given as [79]:

\[ \eta = \frac{E_{out}}{E_{in}} \left[ \frac{i_{load}}{i_{load} - v_{in}T^{\alpha}} \frac{L\Gamma(\alpha + 2)}{(1 - (1 - D)^{\alpha}) (D^{\alpha} - D)} \right], \]  

(18)

where \( i_{load} \) is the average load current. Figure 8c shows the efficiency of the buck converter as a function of \( \alpha \) and \( i_{load} \). It is evident that the efficiency degrades with decreasing \( \alpha \) and \( i_{load} \).

7. CONCLUSION

A literature survey was presented about the fractional calculus fundamentals, integer-order approximations of fractional-order transfer functions, fractional-order element realization, fractional-order circuits and applications, and stability analyses of fractional-order linear systems. The investigated applications included fractional-order oscillators, filters, passive emulators, bio-impedance modeling and DC-DC converters. In most cases, the fractional-order models are more accurate than the integer-order ones. All fractional operators consider the entire history of the process being considered, thus being able to model the non-local and distributed effects often encountered in natural and technical phenomena. The extra degrees of freedom give different design alternatives due to the fractional-order parameters. Fractional-order models provide an improved descrip-
tion of observed bio-impedance behavior. Additionally, the fractional-order enhanced the model of DC-DC converters by incorporating the effect of frequency-dependent inductor loss.

DECLARATION OF CONFLICT OF INTERESTS

The authors have declared no conflict of interests.

ACKNOWLEDGMENTS

Authors would like to thank Science and Technology Development Fund (STDF) for funding the project # 25977 and Nile University for facilitating all procedures required to complete this study.

REFERENCES


FRACTIONAL CALCULUS DEFINITIONS, APPROXIMATIONS, ...


التفاضل الكسري: تعريفات وتقيمات وتطبيقات هندسية

الفكرة الأساسية في التفاضل الكسري هي إمكانية حساب مشتقات وتكاملات من درجات بكسيرية مما أتاح مزيداً من القدرة على وصف أنظمة معقدة وصفاً محكماً. هذا البحث يقدم وصفاً موجزاً عن التفاضل الكسري و بعض تطبيقاته. يسرد البحث بعض من التعريفات الأساسية مصحوبة ببعض الأمثلة التوضيحية. يتبع ذلك ثلاث تطبيقات و مقدمات المحاكاة. كذلك تحتوي على تلخيص لنظرية إنزكان الأنظمة الكسرية. يتبع ذلك وصفاً موجزاً لأربعة تطبيقات هندسية.

المذبذبات إلكترونية الدرجة التي تتيح مزيداً من التحكم في أطوار الموجات المتولدة كما تسهل الحصول على ترددات عالية. المروحة طبقية الكسرية الدرجة التي أتاحت مزيداً من الدقة في حسابات خصائص المرشح. دراسة نماذج المقاومة الحيوية للأنسجة، التي تمكننا من إجراء العديد من القياسات الدقيقة على أنظمة المشتقات. تطورات الجهد الثابت التي تتيح تقيمات أفضل في حسابات توزيع القوى والاستيعاب بحساب قواعد الطاقة المعتمدة على الترددات.