

A NEW METHOD FOR THE ANALYSIS OF
SIGNALS: THE SQUARE WAVE
TRANSFORM (SWT)

UN NUEVO MÉTODO PARA EL ANÁLISIS DE
SEÑALES: LA TRANSFORMADA DE
LAS ONDAS CUADRADAS

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Abstract

The results obtained by analyzing signals with the Square Wave Method (SWM) introduced previously can be presented in the frequency domain clearly and precisely by using the Square Wave Transform (SWT) described here. As an example, the SWT is used to analyze a sequence of samples (that is, of measured values) taken from an electroencephalographic recording. A computational tool, available at www.appliedmathgroup.org/, has been developed and may be used to obtain automatically the SWTs of sequences of samples taken from registers of interest for biomedical purposes, such as those of an EEG or an ECG.

Keywords: signal analysis; square wave method; square wave transform.

Resumen

Los resultados obtenidos al analizar señales con el Método de las Ondas Cuadradas (Square Wave Method, SWM) —previamente introducido— pueden ser presentados en el dominio de la frecuencia de manera clara, precisa y concisa mediante el uso de la Transformada de las Ondas Cuadradas (Square Wave Transform, SWT). Se caracteriza la SWT y, como ejemplo, se la utiliza para analizar una secuencia de muestras (es decir, de valores medidos) tomadas de un registro electroencefalográfico. En www.appliedmathgroup.org, se encuentra disponible un recurso computacional que posibilita obtener, de manera automatizada, las SWT de secuencias de muestras tomadas de registros de interés biomédico como el EEG y el ECG, entre otros.

Palabras clave: análisis de señales; método de las ondas cuadradas; transformada de las ondas cuadradas.

Mathematics Subject Classification: 94A12, 65F99.

1 Introduction

Consideration was previously given to the analysis of functions of one variable using the Square Wave Method (SWM) [4]. This method, which will be reviewed briefly in the following section, was generalized for functions of two variables and applied to the analysis of images [5].

The objective of this article is to specify how the results obtained by analyzing signals with the SWM can be presented in the frequency domain clearly and concisely, using the mathematical process described below: the Square Wave Transform (SWT).

A preliminary version of this article was made available on arXiv [6].

2 Brief review of the application of the SWM to the analysis of functions of one variable

Given that this article is devoted to the analysis of signals, it will be considered that the independent variable is time (t).

Let $f(t)$ be a function of a variable t , which in a given interval Δt , satisfies the conditions of Dirichlet [2]: (1) In the interval Δt , the function $f(t)$ to be analyzed must have a finite number of relative maximums and minimums; (2) in that interval it also must have a finite number of points of discontinuity; and (3) for any instant of Δt , $f(t)$ must have a finite value. That function can then be approximated in that interval by means of a particular sum of trains of square waves. The use of the SWM makes it possible to specify these trains of square waves unambiguously.

Consider, for example, the function $f(t)$, as indicated below:

$$f(t) = 3 \sin(2\pi \cdot 5 \cdot t) + 4 \sin(2\pi \cdot 7 \cdot t); \quad 0 \leq t \leq 1 \text{ s.} \quad (1)$$

In figure 1, $f(t)$ is shown for the interval specified in (1).

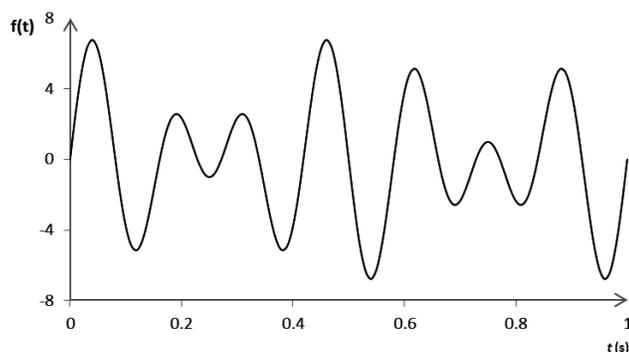


Figure 1: $f(t) = 3 \sin(2\pi \cdot 5 \cdot t) + 4 \sin(2\pi \cdot 7 \cdot t); \quad 0 \leq t \leq 1 \text{ s.}$

Note that the interval of t (Δt), in which $f(t)$ will be analyzed, has a length of 1 second (1 s): $\Delta t = (1 - 0) \text{ s} = 1 \text{ s}$.

First, an explanation will be given about how to proceed if one wants to obtain an approximation to $f(t)$, in the interval Δt specified in (1), composed of the sum of 10 trains of square waves. The interval Δt is then divided into a number of sub-intervals – of equal length – which is the same as the number of trains of square waves. In this case, there will be 10 sub-intervals. The approximation to $f(t)$ to be obtained in interval Δt will be the sum of 10 trains of square

waves: $S_1, S_2, S_3, \dots, S_9$, and S_{10} . The first of the trains of square waves will be referred to by S_1 , the second by S_2 , and so on.

Each of these trains of waves S_i , for $i = 1, 2, 3, \dots, 9$ and 10, will be characterized by a certain frequency f_i (that is, the number of waves in the train of square waves considered which is contained in the unit of time), and a certain coefficient C_i whose absolute value is the amplitude of the corresponding train.

For the case considered here, a description will be provided below of how the amplitudes corresponding to the different trains of square waves are determined (see figure 2).

Δt									
C_1	C_1	C_1	C_1	C_1	C_1	C_1	C_1	C_1	C_1
C_2	C_2	C_2	C_2	C_2	C_2	C_2	C_2	C_2	$-C_2$
C_3	C_3	C_3	C_3	C_3	C_3	C_3	C_3	$-C_3$	$-C_3$
C_4	C_4	C_4	C_4	C_4	C_4	C_4	$-C_4$	$-C_4$	$-C_4$
C_5	C_5	C_5	C_5	C_5	C_5	$-C_5$	$-C_5$	$-C_5$	$-C_5$
C_6	C_6	C_6	C_6	C_6	$-C_6$	$-C_6$	$-C_6$	$-C_6$	$-C_6$
C_7	C_7	C_7	C_7	$-C_7$	$-C_7$	$-C_7$	$-C_7$	C_7	C_7
C_8	C_8	C_8	$-C_8$	$-C_8$	$-C_8$	C_8	C_8	C_8	$-C_8$
C_9	C_9	$-C_9$	$-C_9$	C_9	C_9	$-C_9$	$-C_9$	C_9	C_9
C_{10}	$-C_{10}$	C_{10}	$-C_{10}$	C_{10}	$-C_{10}$	C_{10}	$-C_{10}$	C_{10}	$-C_{10}$



Figure 2: How to apply the SWM to the analysis of the function represented in figure 1. (See indications in text.)

The vertical arrow pointing down at the right of figure 2 indicates how to add the terms corresponding to each of the 10 sub-intervals of Δt . This procedure will make it possible to compute the values of the coefficients $C_1, C_2, C_3, \dots, C_9$, and C_{10} shown in figure 2. First, however, indications will be given about how to compute the frequencies $f_1, f_2, f_3, \dots, f_9$, and f_{10} corresponding respectively to the square wave trains $S_1, S_2, S_3, \dots, S_9$, and S_{10} .

Each row in figure 2 corresponds to the part of each of the trains of square waves $S_1, S_2, S_3, \dots, S_9$, and S_{10} in interval Δt . In the first place, the structure

of the last row in that figure corresponds to the part of the train of square waves S_{10} in interval Δt . Each pair of consecutive coefficients $C_{10} \mid -C_{10}$ corresponds to a square wave in S_{10} . Observe that in Δt there are 5 of these pairs of elements (that is, there are 5 square waves in Δt). The frequency f_{10} , corresponding to S_{10} , is obtained by dividing, by Δt , the number of square waves occurring in Δt ; thus, $f_{10} = \frac{5}{\Delta t}$. In the case discussed, $\Delta t = 1$ s, so the value of f_{10} is as follows: $f_{10} = 5 \text{ s}^{-1}$. Of course, Δt can be different from 1 s. Suppose that we had taken $\Delta t = 5$ s. The following value would have been obtained for f_{10} : $f_{10} = \frac{5}{5 \text{ s}} = 1 \text{ s}^{-1}$; and there would have been only one square wave in each time unit 1 s.

The next to the last row in figure 2 corresponds to the part of the train of square waves S_9 in the interval Δt . Here the structure of each square wave is as follows: $C_9 \mid C_9 \mid -C_9 \mid -C_9$. The length of the wave of each square wave corresponding to S_9 is double that of the wave corresponding to S_{10} . Note that each square wave corresponding to S_{10} is included in 2 sub-intervals of Δt , whereas each square wave in S_9 is encompassed by 4 intervals of Δt . The value of S_9 is obtained by dividing, by Δt , the number of square waves S_9 in Δt : $f_9 = \frac{2 \cdot 5}{\Delta t} = 2.5 \text{ s}^{-1}$; in other words, in each unit of time 1 s, there are two and a half waves of S_9 .) Observe that because the length of each square wave corresponding to S_9 is twice the length of each wave corresponding to S_{10} , the following result is to be expected: $f_9 = \frac{1}{2} f_{10}$.

The third row from the bottom in figure 2 corresponds to the part of the train of square waves S_8 in the interval Δt . In this case, the structure of each square wave is: $C_8 \mid C_8 \mid C_8 \mid -C_8 \mid -C_8 \mid -C_8$. The length of each square wave in S_8 is three times that of each square wave in S_{10} . That is, each square wave in S_{10} is encompassed by 2 sub-intervals of Δt , whereas each square wave of C_8 is encompassed by 6 sub-intervals of Δt . Of course, the value of f_8 is obtained by dividing, by Δt , the number of square waves corresponding to S_8 in the interval Δt : $f_8 = \frac{\frac{10}{6}}{\Delta t} = \frac{1}{3} \cdot 5 \text{ s}^{-1}$.

Since the length of each square wave corresponding to S_8 is triple the length of each square wave corresponding to S_{10} , the validity of the following equality was foreseeable: $f_8 = \frac{1}{3} f_{10}$.

In figure 2, it can be seen that the lengths of the square waves $S_7, S_6, S_5, S_4, S_3, S_2, S_1$ are respectively 4, 5, 6, 7, 8, 9, and 10 times longer than the square wave S_{10} .

Therefore, the following values are obtained for frequencies $f_1, f_2, f_3, \dots, f_{10}$, corresponding respectively to $S_1, S_2, S_3, \dots, S_9$, and S_{10} :

$$\begin{aligned} f_1 &= \frac{1}{10} \cdot f_{10} = \frac{1}{10} \cdot 5 \text{ s}^{-1} = 0.5000000 \text{ s}^{-1} \\ f_2 &= \frac{1}{9} \cdot f_{10} = \frac{1}{9} \cdot 5 \text{ s}^{-1} = 0.5555556 \text{ s}^{-1} \\ f_3 &= \frac{1}{8} \cdot f_{10} = \frac{1}{8} \cdot 5 \text{ s}^{-1} = 0.6250000 \text{ s}^{-1} \\ f_4 &= \frac{1}{7} \cdot f_{10} = \frac{1}{7} \cdot 5 \text{ s}^{-1} = 0.7142857 \text{ s}^{-1} \\ f_5 &= \frac{1}{6} \cdot f_{10} = \frac{1}{6} \cdot 5 \text{ s}^{-1} = 0.8333333 \text{ s}^{-1} \\ f_6 &= \frac{1}{5} \cdot f_{10} = \frac{1}{5} \cdot 5 \text{ s}^{-1} = 1.0000000 \text{ s}^{-1} \\ f_7 &= \frac{1}{4} \cdot f_{10} = \frac{1}{4} \cdot 5 \text{ s}^{-1} = 1.2500000 \text{ s}^{-1} \\ f_8 &= \frac{1}{3} \cdot f_{10} = \frac{1}{3} \cdot 5 \text{ s}^{-1} = 1.6666667 \text{ s}^{-1} \\ f_9 &= \frac{1}{2} \cdot f_{10} = \frac{1}{2} \cdot 5 \text{ s}^{-1} = 2.5000000 \text{ s}^{-1} \\ f_{10} &= \frac{1}{1} \cdot f_{10} = \frac{1}{1} \cdot 5 \text{ s}^{-1} = 5.0000000 \text{ s}^{-1}. \end{aligned}$$

More concisely, these ten frequencies can be expressed as:

$$f_i = \frac{1}{10 - i + 1} \cdot f_{10} = \frac{1}{10 - i + 1} \cdot 5 \text{ s}^{-1}; \quad i = 1, 2, 3, \dots, n.$$

If the same approach is used for any Δt expressed in seconds and any natural number n of sub-intervals into which the interval Δt is divided, for the frequencies $f_1, f_2, f_3, \dots, f_n$ corresponding respectively to the different trains of square waves $S_1, S_2, S_3, \dots, S_9$, and S_n , the following equation is obtained:

$$f_i = \frac{1}{n - i + 1} \cdot f_n = \frac{1}{n - i + 1} \cdot \frac{\frac{n}{2}}{\Delta t} = \frac{1}{n - i + 1} \cdot \frac{n}{2\Delta t}; \quad i = 1, 2, 3, \dots, n.$$

How to compute the values of the coefficients $C_1, C_2, C_3, \dots, C_9$, and C_{10} shown in figure 2 will be specified below.

First, the sum of all the coefficients in the first column of figure 2 is made equal to the value of the function which one wants to approximate – that is, (1) – at the midpoint of the first of the ten sub-intervals into which the interval Δt was divided. This value will be called V_1 . Hence the following equation is obtained: $C_1 + C_2 + C_3 + C_4 + C_5 + C_6 + C_7 + C_8 + C_9 + C_{10} = V_1$.

Second, the sum of all the coefficients in the second column of figure 2 is made equal to the value of the function which one wants to approximate – that is (1) – at the midpoint of the second of the ten sub-intervals into which the interval Δt was divided. This value will be called V_2 , and the following equation is obtained: $C_1 + C_2 + C_3 + C_4 + C_5 + C_6 + C_7 + C_8 + C_9 - C_{10} = V_2$.

Third, the sum of all the coefficients in the third column of figure 2 is made equal to the value of the function which one wants to approximate – that is (1) – at the midpoint of the third of the ten sub-intervals into which the interval Δt was divided. This value will be called V_3 , and the following equation is obtained: $C_1 + C_2 + C_3 + C_4 + C_5 + C_6 + C_7 + C_8 - C_9 + C_{10} = V_3$.

The same is done for each of the remaining columns of coefficients in figure 2. Thus it is possible to obtain another seven equations which, together with the first three, constitute the following system of linear algebraic equations:

$$\left. \begin{aligned} C_1 + C_2 + C_3 + C_4 + C_5 + C_6 + C_7 + C_8 + C_9 + C_{10} &= V_1 \\ C_1 + C_2 + C_3 + C_4 + C_5 + C_6 + C_7 + C_8 + C_9 - C_{10} &= V_2 \\ C_1 + C_2 + C_3 + C_4 + C_5 + C_6 + C_7 + C_8 - C_9 + C_{10} &= V_3 \\ C_1 + C_2 + C_3 + C_4 + C_5 + C_6 + C_7 - C_8 - C_9 - C_{10} &= V_4 \\ C_1 + C_2 + C_3 + C_4 + C_5 + C_6 - C_7 - C_8 + C_9 + C_{10} &= V_5 \\ C_1 + C_2 + C_3 + C_4 + C_5 - C_6 - C_7 - C_8 + C_9 - C_{10} &= V_6 \\ C_1 + C_2 + C_3 + C_4 + C_5 - C_6 - C_7 + C_8 - C_9 + C_{10} &= V_7 \\ C_1 + C_2 + C_3 - C_4 - C_5 - C_6 - C_7 + C_8 - C_9 - C_{10} &= V_8 \\ C_1 + C_2 - C_3 - C_4 - C_5 - C_6 + C_7 + C_8 + C_9 + C_{10} &= V_9 \\ C_1 - C_2 - C_3 - C_4 - C_5 - C_6 + C_7 - C_8 + C_9 - C_{10} &= V_{10}. \end{aligned} \right\} (2)$$

In the preceding system of linear algebraic equations (2), $V_1, V_2, V_3, \dots, V_9$, and V_{10} are the values for $f(t)$ as specified in (1) at the midpoints of the first, second, third, \dots , ninth, and tenth sub-intervals, respectively, of the interval Δt in which $f(t)$ is analyzed. It follows that the values V_i , for $i = 1, 2, 3, \dots, 9$ and 10, can be computed given that $f(t)$ has been specified in (1). These values are:

$$\begin{aligned} V_1 &= 6.2360680 & V_6 &= -6.2360680 \\ V_2 &= -1.7639320 & V_7 &= 1.7639320 \\ V_3 &= -1.0000000 & V_8 &= 1.0000000 \\ V_4 &= -1.7639320 & V_9 &= 1.7639320 \\ V_5 &= 6.2360680 & V_{10} &= -6.2360680. \end{aligned}$$

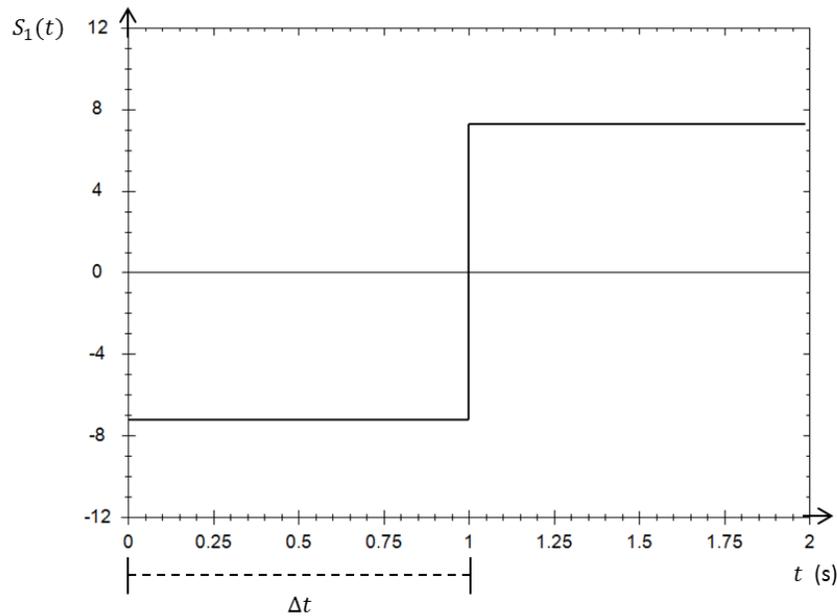
The ten unknowns of the system of equations specified in (2) are $C_1, C_2, C_3, \dots, C_9$, and C_{10} . $|C_i|$ refers to the amplitude of the train of square waves S_i ,

for $i = 1, 2, 3, \dots, 10$. The (constant) value of each positive square semi-wave of the train of square waves S_i is $|C_i|$, and the (constant) value of each negative square semi-wave of that S_i is $-|C_i|$.

The system of equations (2) was solved by using LAPACK [1], and the following results were obtained for the unknowns:

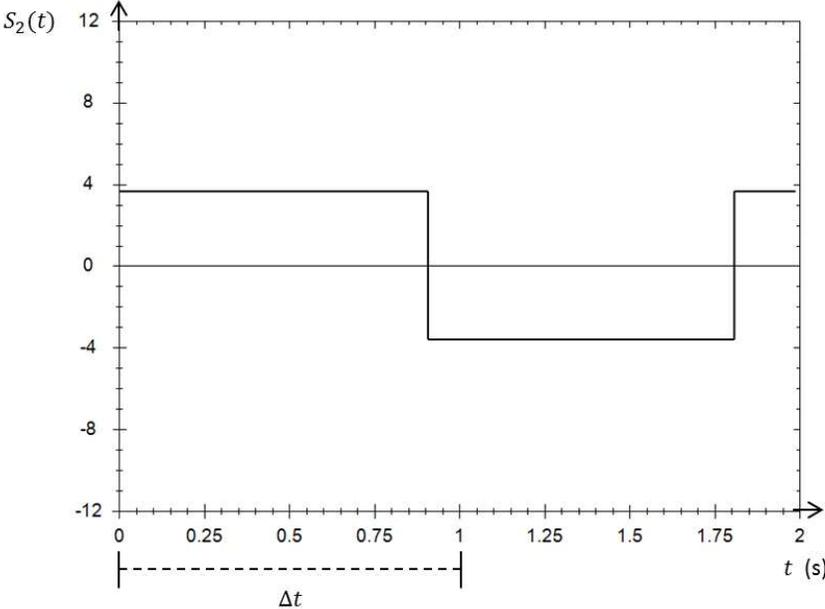
$$\begin{array}{ll} C_1 = -7.23607 & C_6 = 2.23607 \\ C_2 = 3.61803 & C_7 = 3.61803 \\ C_3 = 10.85410 & C_8 = -3.61803 \\ C_4 = -3.61803 & C_9 = 3.61803 \\ C_5 = -7.23607 & C_{10} = 4.00000. \end{array}$$

The trains of square waves $S_1, S_2, S_3, \dots, S_9$, and S_{10} have been shown for interval Δt , in figures 3a, 3b, 3c, \dots , 3i, and 3j, respectively.

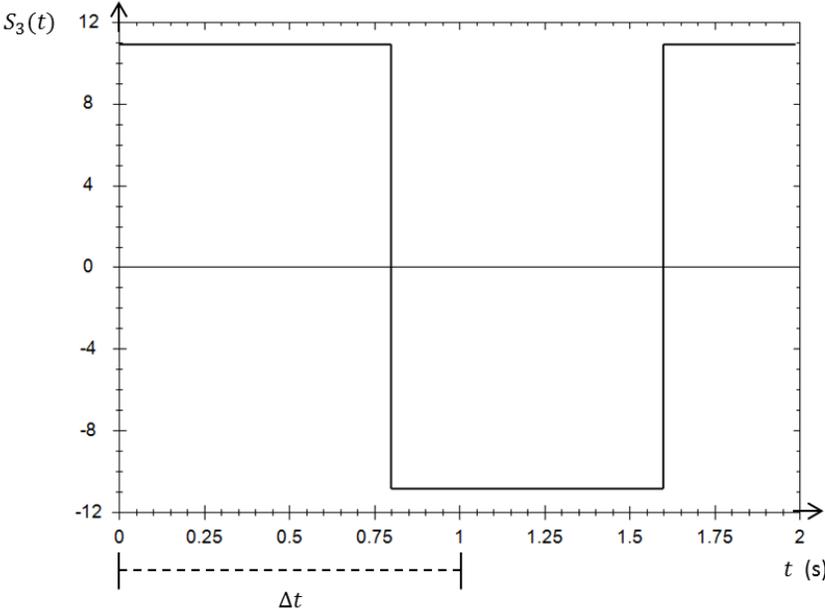


(a) $S_1(t)$.

Figure 3

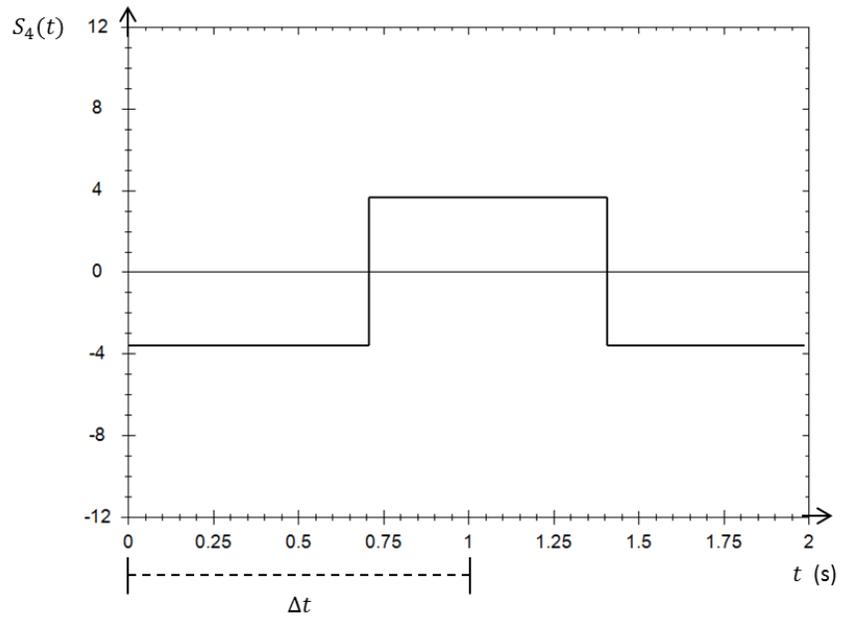
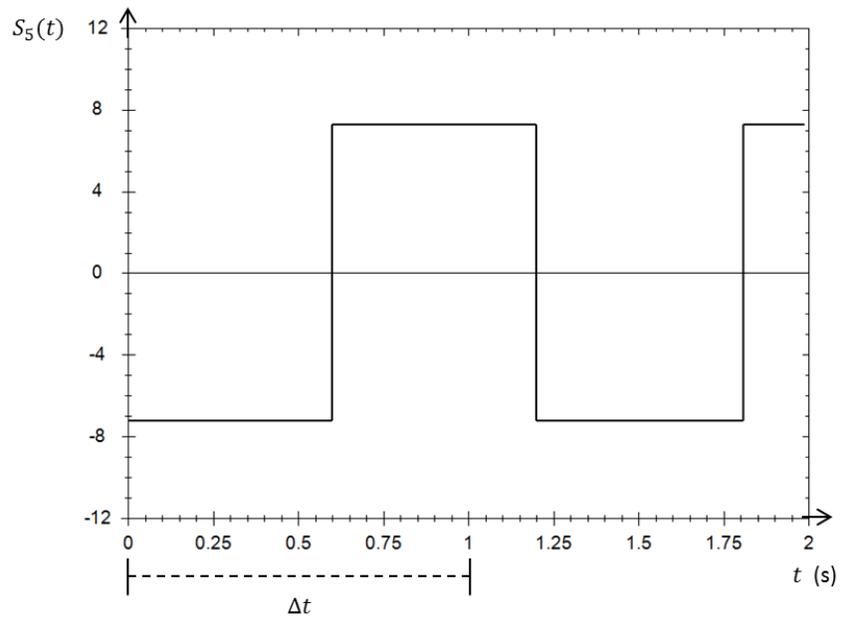


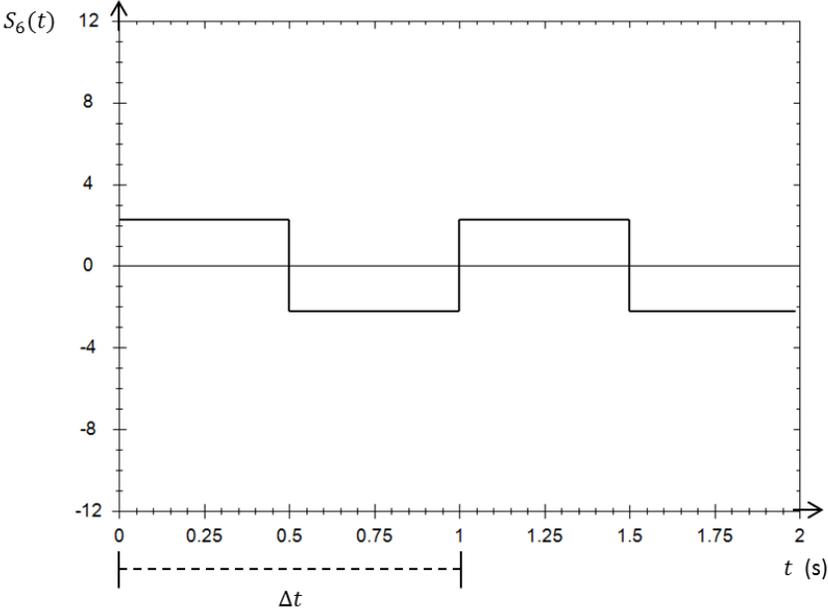
(b) $S_2(t)$.



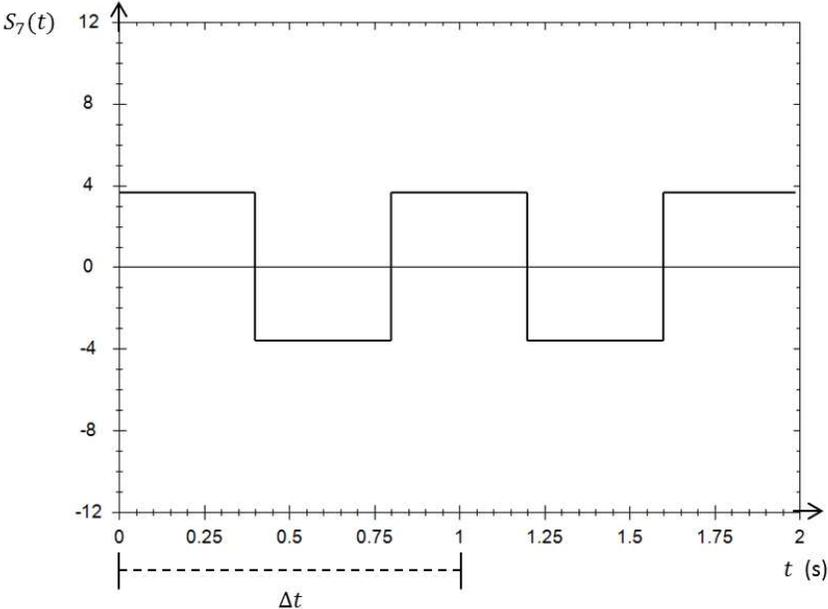
(c) $S_3(t)$.

Figure 3

(d) $S_4(t)$.(e) $S_5(t)$.**Figure 3**

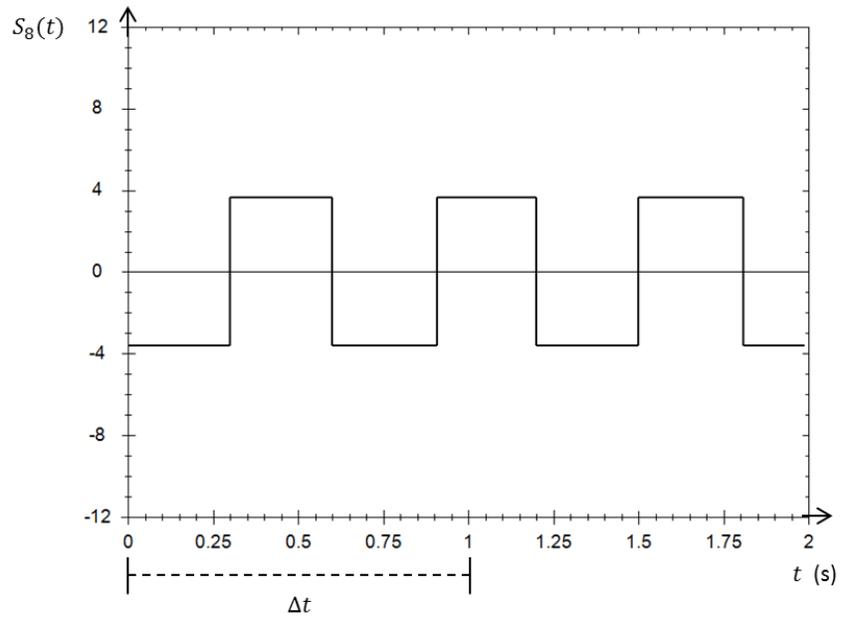
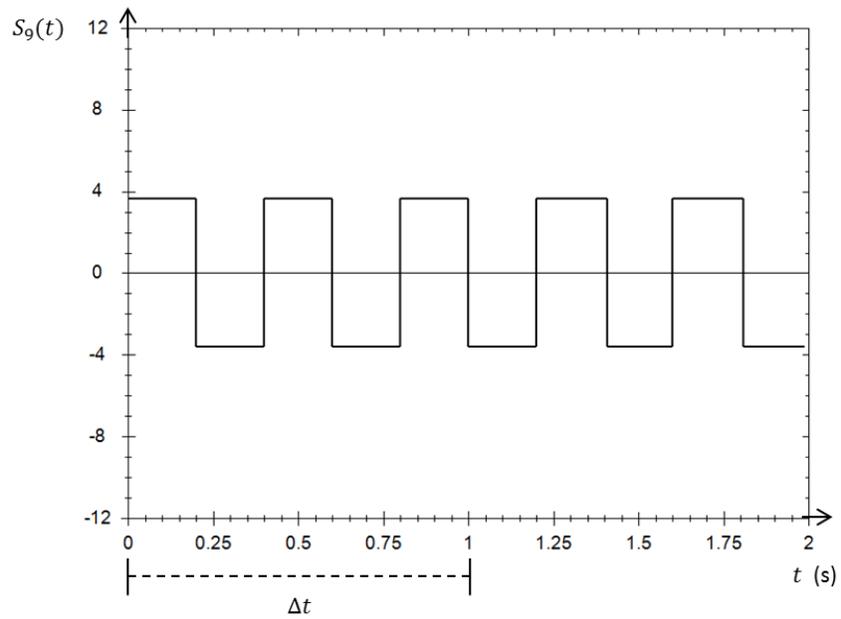


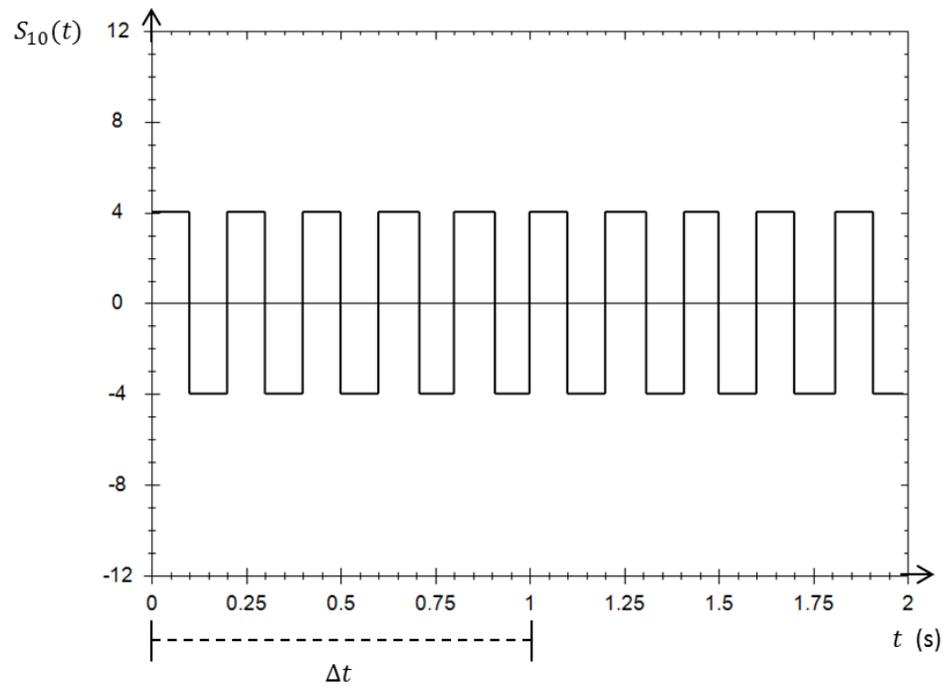
(f) $S_6(t)$.



(g) $S_7(t)$.

Figure 3

(h) $S_8(t)$.(i) $S_9(t)$.**Figure 3**



(j) $S_{10}(t)$.

Figure 3: Trains of square waves $S_1, S_2, S_3, \dots, S_9$, and S_{10} .

The approximation obtained for $f(t)$ (as specified in (1), in interval Δt , by adding the 10 trains of square waves) has been displayed in figure 4.

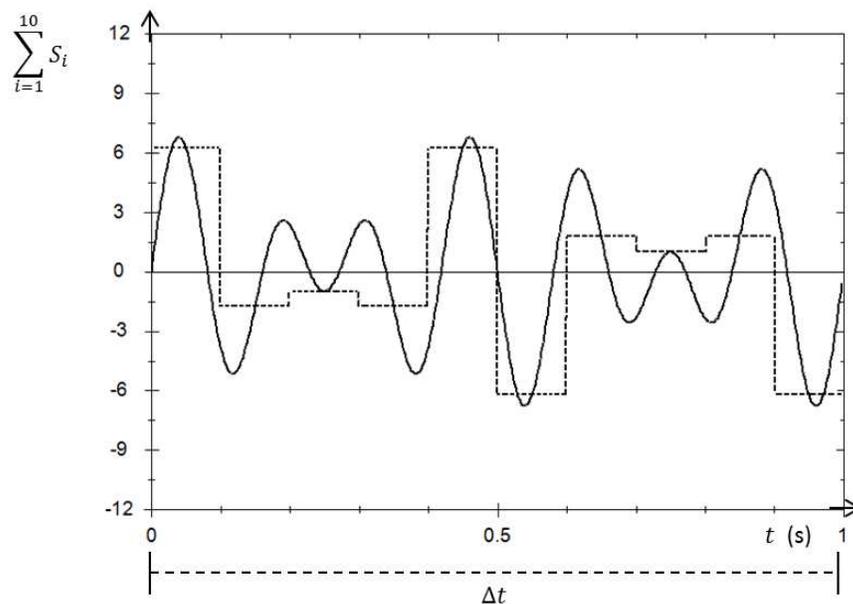


Figure 4: The dashed line indicates the approximation to $f(t)$, specified in (1), by

$$\sum_{i=1}^{10} S_i.$$

If one wants to achieve a better approximation to $f(t)$, in interval Δt , by adding the trains of square waves, then Δt should be divided into a larger number of equal sub-intervals. The larger the number of these sub-intervals, the better the approximation. Thus, for example, the approximation to $f(t)$ that can be achieved if Δt is divided into 100 sub-intervals of equal length, is shown in figure 5. In this case, it is clear that 100 trains of square waves were added together.

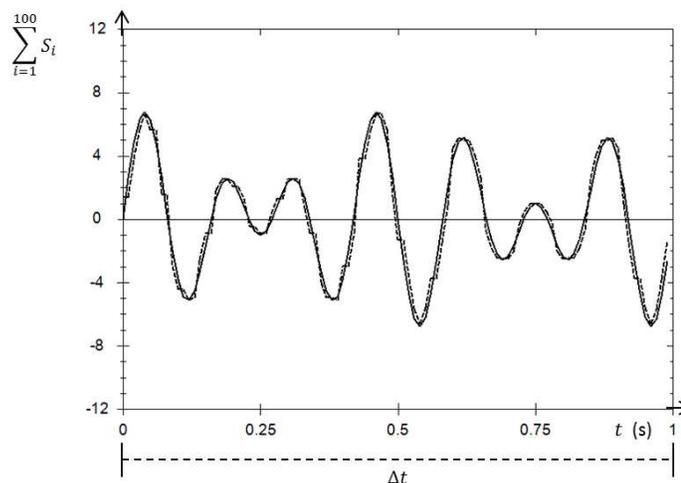


Figure 5: The dashed line indicates the approximation to $f(t)$, specified in (1), by

$$\sum_{i=1}^{100} S_i.$$

The SWM cannot be considered a branch of Fourier analysis; that is, the trains of square waves S_i , for $i = 1, 2, 3, \dots, n$, do not make up a system of orthogonal functions.

3 The Square Wave Transform (SWT) as a way of presenting the results of the analysis of $f(t)$ specified in (1)

First, let us examine the results obtained when using the SWT to analyze the function $f(t)$ specified in (1), for the case specified above, in which the interval Δt was divided into 10 equal sub-intervals. These results can be presented by a sequence of 10 dyads (ordered pairs) such that the first element of the first dyad is the frequency f_1 corresponding to S_1 , and the second element of that first dyad is the coefficient C_1 ; the first element of the second dyad is the frequency f_2 corresponding to S_2 , and the second element of that second dyad is the coefficient C_2 ; and so on successively, such that the first element of the tenth dyad is the frequency f_{10} corresponding to S_{10} , and the second element of that tenth

dyad is the coefficient C_{10} :

$$\begin{aligned} (f_1; C_1) &= (0.5000000; -7.23607) & (f_2; C_2) &= (0.5555556; 3.61803) \\ (f_3; C_3) &= (0.6250000; 10.85410) & (f_4; C_4) &= (0.7142857; -3.61803) \\ (f_5; C_5) &= (0.8333333; -7.23607) & (f_6; C_6) &= (1.0000000; 2.23607) \\ (f_7; C_7) &= (1.2500000; 3.61803) & (f_8; C_8) &= (1.6666667; -3.61803) \\ (f_9; C_9) &= (2.5000000; 3.61803) & (f_{10}; C_{10}) &= (5.0000000; 4.00000). \end{aligned}$$

This approximation of the function $f(t)$ specified in (1) can be expressed in the frequency domain. To achieve this objective, for each of the frequencies considered, $f_1, f_2, f_3, \dots, f_{10}$, the corresponding coefficients $C_1, C_2, C_3, \dots, C_{10}$ must be indicated. The expression in the frequency domain of this approximation to $f(t)$ will be called the Square Wave Transform (SWT) of the approximation to that $f(t)$. This SWT is displayed in figure 6.

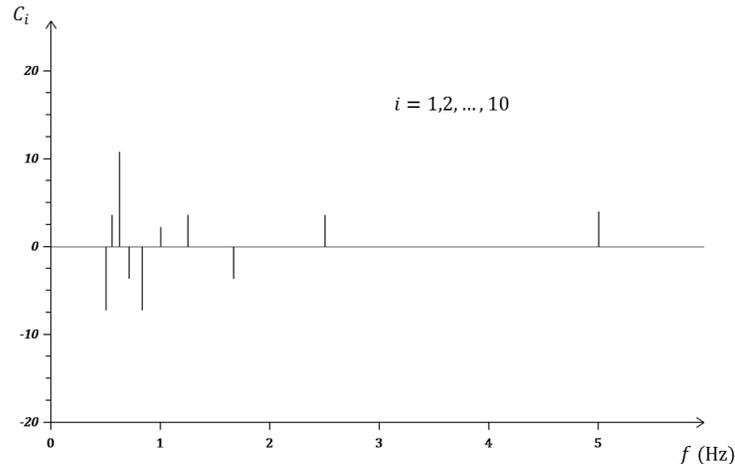
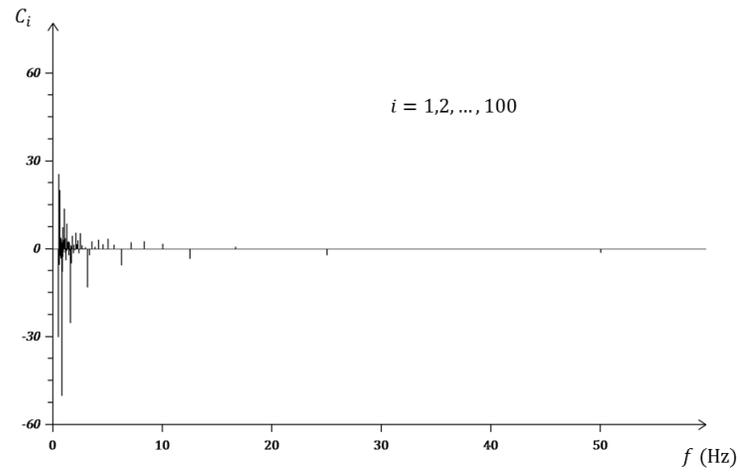


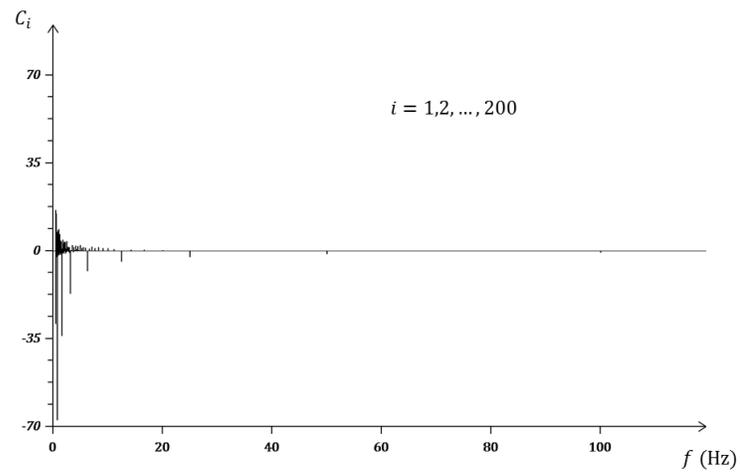
Figure 6: SWT of the approximation to $f(t)$, specified in (1), obtained by dividing the interval Δt into 10 sub-intervals.

Of course, the SWTs corresponding to numbers as large as desired of equal sub-intervals into which Δt is divided can be obtained for the $f(t)$ specified in (1), or for any other function of the time which, in a particular interval Δt , satisfies the conditions of Dirichlet.

Here, the symbol N_s will be used to refer to the number of equal sub-intervals into which Δt is divided. In figure 7, the SWTs of the approximations (to the $f(t)$ specified in (1)) obtained are shown for the following values of N_s : 100, 200, and 400.

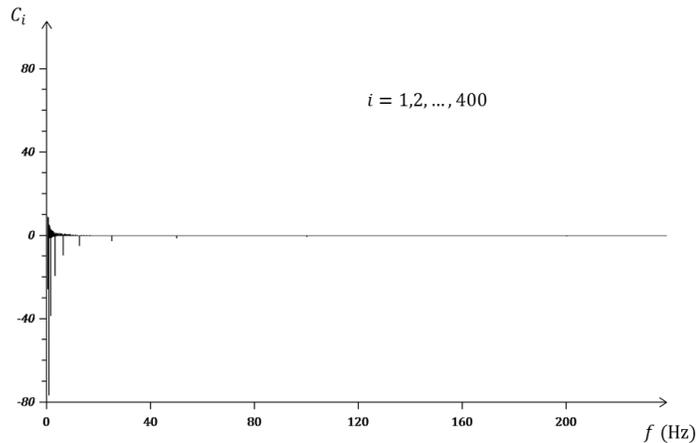


(a) SWT corresponding to $N_s = 100$.



(b) SWT corresponding to $N_s = 200$.

Figure 7



(c) SWT corresponding to $N_s = 400$.

Figure 7: The SWTs obtained of the approximations to the function $f(t)$ (specified in (1)), have been displayed in 7a, 7b, and 7c, for $N_s = 100$, $N_s = 200$, and $N_s = 400$, respectively.

Note that in the three cases displayed in figure 7, different scales were used for the axes of the abscissas. The same scale will be used for the axes in figure 11.

4 The SWT as a tool for the analysis of an electroencephalographic signal

The SWT can be used for the analysis of signals of biomedical interest, such as those of electrocardiograms (ECG), electroencephalograms (EEG), electromyograms (EMG), etc.

Suppose that one has a sequence of 10 values of an electrophysiological signal, such as an electroencephalographic recording. To obtain the SWT corresponding to that sequence, the sequence of values is treated the same as was treated, with the same objective (that of obtaining the SWT) the sequence of values $V_1, V_2, V_3, \dots, V_{10}$, in the system of algebraic equations (2). Generally, if one wants to obtain the SWT corresponding to a sequence $V_1, V_2, V_3, \dots, V_N$ of measured values from that recording, that sequence of values is treated the same as the sequence of values $V_1, V_2, V_3, \dots, V_{N_s}$ is treated, with $N_s = N$. Let us recall that $V_1, V_2, V_3, \dots, V_{N_s}$ is a sequence of values of a function which are computed at the midpoints of the N_s equal sub-intervals of interval Δt for which the function is characterized analytically.

A sequence of 160 “samples” (i.e., measured values) from an electroencephalographic recording is displayed in figure 8. The data were taken from the EEG Motor Movement/ Imagery Dataset (tagged MMIDB) in PhysioBank [3]. That recording corresponds to FC5 of run 01 of Subject S001, with a “sampling” frequency of 160 Hz.

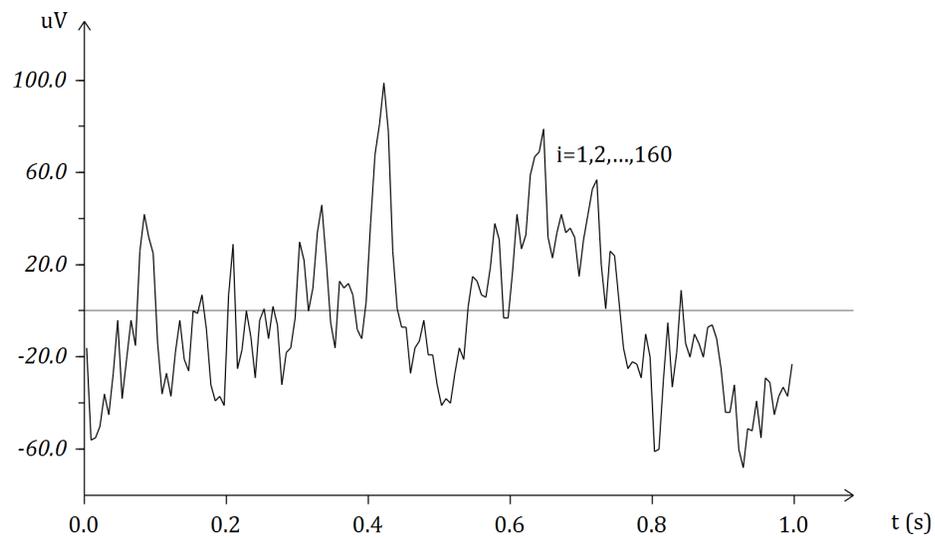


Figure 8: Excerpt from an EEG, 160 Hz.

The sequence of voltage measurements specified in microvolts (μV), shown in figure 8, is as follows:

$V_1 = -16$	$V_{33} = 7$	$V_{65} = 38$	$V_{97} = 17$	$V_{129} = -61$
$V_2 = -56$	$V_{34} = 29$	$V_{66} = 68$	$V_{98} = 42$	$V_{130} = -60$
$V_3 = -55$	$V_{35} = -25$	$V_{67} = 81$	$V_{99} = 27$	$V_{131} = -30$
$V_4 = -50$	$V_{36} = -17$	$V_{68} = 99$	$V_{100} = 33$	$V_{132} = -5$
$V_5 = -36$	$V_{37} = 0$	$V_{69} = 78$	$V_{101} = 59$	$V_{133} = -33$
$V_6 = -45$	$V_{38} = -11$	$V_{70} = 26$	$V_{102} = 67$	$V_{134} = -18$
$V_7 = -27$	$V_{39} = -29$	$V_{71} = 1$	$V_{103} = 69$	$V_{135} = 9$
$V_8 = -4$	$V_{40} = -4$	$V_{72} = -7$	$V_{104} = 79$	$V_{136} = -14$
$V_9 = -38$	$V_{41} = 1$	$V_{73} = -7$	$V_{105} = 32$	$V_{137} = -20$
$V_{10} = -21$	$V_{42} = -12$	$V_{74} = -27$	$V_{106} = 23$	$V_{138} = -10$
$V_{11} = -4$	$V_{43} = 2$	$V_{75} = -16$	$V_{107} = 34$	$V_{139} = -14$
$V_{12} = -15$	$V_{44} = -6$	$V_{76} = -13$	$V_{108} = 42$	$V_{140} = -20$
$V_{13} = 26$	$V_{45} = -32$	$V_{77} = -4$	$V_{109} = 34$	$V_{141} = -7$
$V_{14} = 42$	$V_{46} = -18$	$V_{78} = -19$	$V_{110} = 36$	$V_{142} = -6$
$V_{15} = 32$	$V_{47} = -16$	$V_{79} = -19$	$V_{111} = 32$	$V_{143} = -12$
$V_{16} = 25$	$V_{48} = -3$	$V_{80} = -32$	$V_{112} = 15$	$V_{144} = -25$
$V_{17} = -14$	$V_{49} = 30$	$V_{81} = -41$	$V_{113} = 31$	$V_{145} = -44$
$V_{18} = -36$	$V_{50} = 22$	$V_{82} = -38$	$V_{114} = 42$	$V_{146} = -44$
$V_{19} = -27$	$V_{51} = 0$	$V_{83} = -40$	$V_{115} = 53$	$V_{147} = -32$
$V_{20} = -37$	$V_{52} = 10$	$V_{84} = -27$	$V_{116} = 57$	$V_{148} = -60$
$V_{21} = -18$	$V_{53} = 34$	$V_{85} = -16$	$V_{117} = 20$	$V_{149} = -68$
$V_{22} = -4$	$V_{54} = 46$	$V_{86} = -21$	$V_{118} = 1$	$V_{150} = -51$
$V_{23} = -21$	$V_{55} = 22$	$V_{87} = 2$	$V_{119} = 26$	$V_{151} = -52$
$V_{24} = -26$	$V_{56} = -5$	$V_{88} = 15$	$V_{120} = 24$	$V_{152} = -39$
$V_{25} = 0$	$V_{57} = -16$	$V_{89} = 13$	$V_{121} = 4$	$V_{153} = -55$
$V_{26} = -1$	$V_{58} = 13$	$V_{90} = 7$	$V_{122} = -16$	$V_{154} = -29$
$V_{27} = 7$	$V_{59} = 10$	$V_{91} = 6$	$V_{123} = -25$	$V_{155} = -31$
$V_{28} = -8$	$V_{60} = 12$	$V_{92} = 19$	$V_{124} = -22$	$V_{156} = -45$
$V_{29} = -32$	$V_{61} = 7$	$V_{93} = 38$	$V_{125} = -23$	$V_{157} = -37$
$V_{30} = -39$	$V_{62} = -8$	$V_{94} = 31$	$V_{126} = -29$	$V_{158} = -33$
$V_{31} = -37$	$V_{63} = -12$	$V_{95} = -3$	$V_{127} = -10$	$V_{159} = -37$
$V_{32} = -41$	$V_{64} = 4$	$V_{96} = -3$	$V_{128} = -20$	$V_{160} = -23$

In figure 9, the SWT is shown for the sequence of 160 “samples” (i.e., measured values) of that electroencephalographic recording.

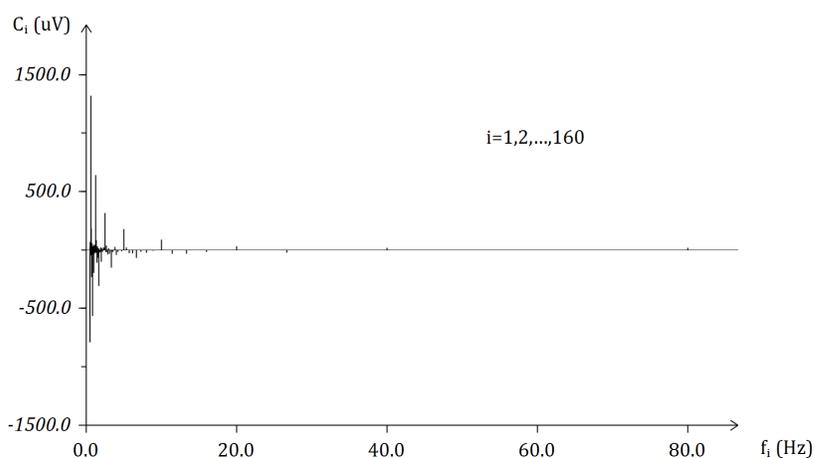


Figure 9: The SWT corresponding to the sequence of “samples” represented in figure 8, from an electroencephalographic recording.

In figure 10, a close-up of the most notable portion of the SWT is shown.

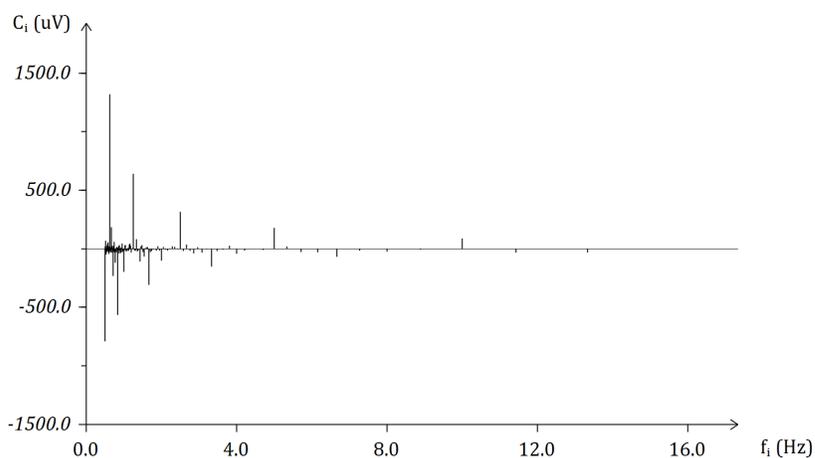


Figure 10: Close-up of the SWT corresponding to the sequence of “samples” represented in figure 9, from an electroencephalographic recording, where $f_i < 16$.

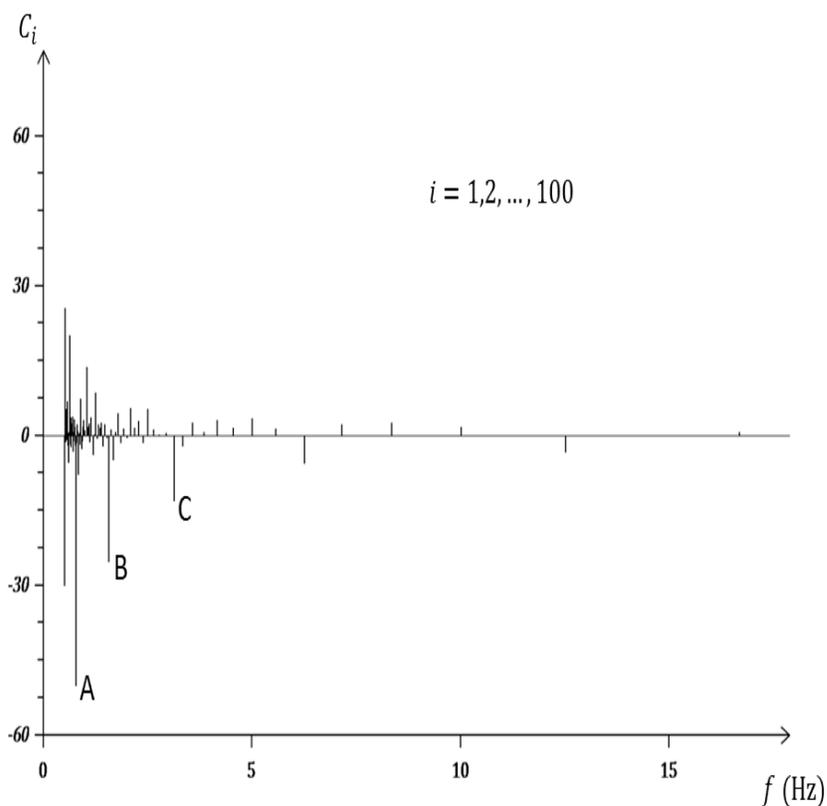
The sequence of 160 dyads such that the first element of the i^{th} dyad ($i = 1, 2, 3, \dots, 160$) is f_i (i.e., the frequency corresponding to S_i), and the second element of that dyad is C_i , is given below. (The 160 dyads are in increasing order according to the corresponding frequencies. These frequencies have been specified with a 4-digit accuracy.)

(0.5000; -789.5)	(0.6667; 186.0)	(1.0000; -195.5)	(2.0000; -100.0)
(0.5031; 21.5)	(0.6723; 21.5)	(1.0127; -13.5)	(2.0513; 18.0)
(0.5063; -11.0)	(0.6780; -33.0)	(1.0256; 30.5)	(2.1053; -5.5)
(0.5096; -22.0)	(0.6838; 28.0)	(1.0390; 33.5)	(2.1622; -14.5)
(0.5128; 70.0)	(0.6897; -5.5)	(1.0526; -18.0)	(2.2222; 4.5)
(0.5161; 20.5)	(0.6957; 11.0)	(1.0667; -1.0)	(2.2857; 23.0)
(0.5195; 20.5)	(0.7018; -3.0)	(1.0811; -19.5)	(2.3529; 18.5)
(0.5229; -2.0)	(0.7080; -25.5)	(1.0959; -10.0)	(2.4242; 6.0)
(0.5263; -47.5)	(0.7143; -231.0)	(1.1111; 10.0)	(2.5000; 317.5)
(0.5298; -26.5)	(0.7207; 25.5)	(1.1268; -16.0)	(2.5806; -18.0)
(0.5333; -19.0)	(0.7273; 25.0)	(1.1429; 38.5)	(2.6667; 37.5)
(0.5369; -28.5)	(0.7339; -21.0)	(1.1594; 46.0)	(2.7586; -16.5)
(0.5405; -23.0)	(0.7407; 61.0)	(1.1765; 27.0)	(2.8571; -38.5)
(0.5442; 5.5)	(0.7477; -24.0)	(1.1940; -29.0)	(2.9630; 16.0)
(0.5479; -15.5)	(0.7547; -31.0)	(1.2121; -0.5)	(3.0769; -31.5)
(0.5517; 12.0)	(0.7619; -17.5)	(1.2308; 8.5)	(3.2000; -4.0)
(0.5556; 33.0)	(0.7692; -117.5)	(1.2500; 642.0)	(3.3333; -150.5)
(0.5594; 29.0)	(0.7767; -25.0)	(1.2698; 5.0)	(3.4783; -17.5)
(0.5634; -12.5)	(0.7843; -13.0)	(1.2903; -15.5)	(3.6364; -5.5)
(0.5674; 28.5)	(0.7921; -24.0)	(1.3115; -12.5)	(3.8095; 27.0)
(0.5714; 45.5)	(0.8000; 8.0)	(1.3333; 84.5)	(4.0000; -41.5)
(0.5755; -17.0)	(0.8081; 14.0)	(1.3559; -21.0)	(4.2105; -15.0)
(0.5797; 57.5)	(0.8163; 18.0)	(1.3793; -14.5)	(4.4444; -1.5)
(0.5839; -25.0)	(0.8247; -32.5)	(1.4035; 3.0)	(4.7059; -9.0)
(0.5882; 22.5)	(0.8333; -564.0)	(1.4286; -107.5)	(5.0000; 180.5)
(0.5926; 2.5)	(0.8421; 10.5)	(1.4545; 23.5)	(5.3333; 21.5)
(0.5970; -42.0)	(0.8511; -9.0)	(1.4815; 32.5)	(5.7143; -26.0)
(0.6015; 19.0)	(0.8602; 24.0)	(1.5094; -26.0)	(6.1538; -28.0)
(0.6061; 20.0)	(0.8696; -38.0)	(1.5385; -64.5)	(6.6667; -67.0)
(0.6107; -32.5)	(0.8791; 33.0)	(1.5686; 6.5)	(7.2727; -14.5)
(0.6154; 5.5)	(0.8889; -7.5)	(1.6000; 15.5)	(8.0000; -23.5)
(0.6202; 18.0)	(0.8989; -17.0)	(1.6327; 15.5)	(8.8889; -6.0)
(0.6250; 1321.5)	(0.9091; -36.0)	(1.6667; -307.0)	(10.0000; 89.0)
(0.6299; -15.0)	(0.9195; 12.5)	(1.7021; -26.5)	(11.4857; -31.5)
(0.6349; -2.0)	(0.9302; -27.0)	(1.7391; -18.0)	(13.3333; -31.0)
(0.6400; 2.5)	(0.9412; 7.0)	(1.7778; -4.5)	(16.0000; -15.5)
(0.6452; -26.0)	(0.9524; 46.5)	(1.8182; 0.0)	(20.0000; 32.5)
(0.6504; 14.5)	(0.9639; -26.5)	(1.8605; -16.0)	(26.6667; -22.5)
(0.6557; -7.5)	(0.9756; -12.0)	(1.9048; 23.5)	(40.0000; 19.5)
(0.6612; 4.5)	(0.9877; -9.0)	(1.9512; -13.5)	(80.0000; 20.0)

5 Discussion and prospects

It must be emphasized that the SWTs of the corresponding approximations to a particular function have a pattern in common for high enough values of N_s . Although this topic will be addressed elsewhere, preliminary support for this will be given below.

Partial graphic representations of the SWTs displayed in figure 7 above are shown in figure 11. The SWTs in which N_s is equal to 100, 200 and 400, respectively, are partially presented in 11a, 11b, and 11c. In this case, the SWTs are described as “partial” because the axes of the abscissas extend only as far as the frequencies which are equal to or less than 15. To detect this pattern easily, the same scale has been used in the axes of the abscissas in 11a, 11b, and 11c.



(a) $N_s = 100$.

Figure 11

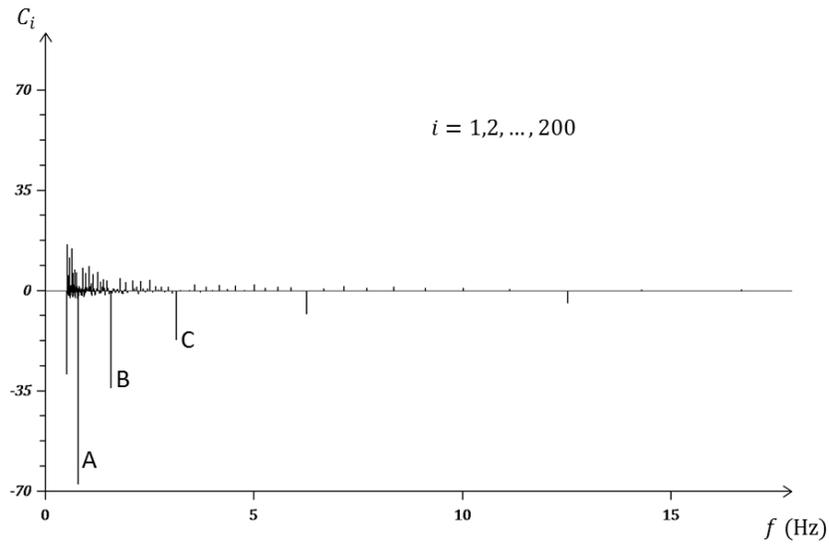
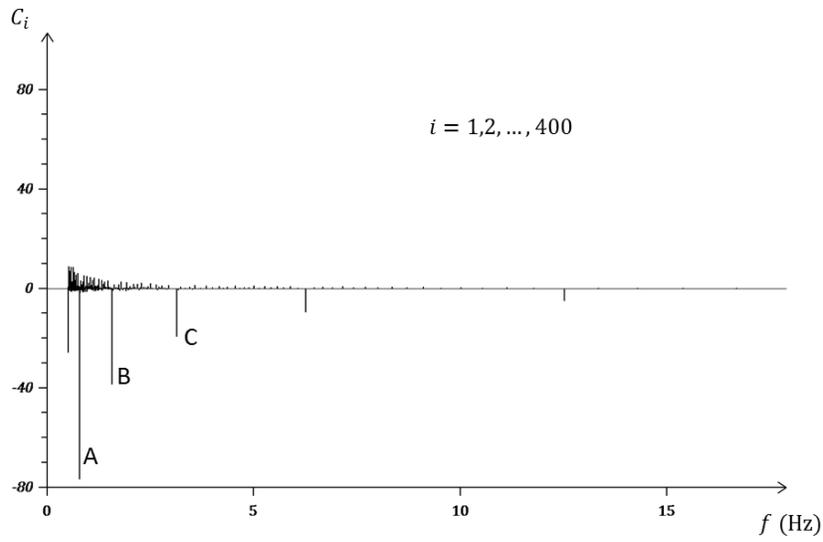
(b) $N_s = 200$.(c) $N_s = 400$.

Figure 11: The SWTs seen in figure 7 are displayed partially in 11a, 11b, and 11c. The same scale has been used in the axes of the abscissas.

Note, for example, the correspondence between the coefficients which have been indicated by the letter “A” in 11a, 11b and 11c. The correspondence between the coefficients indicated by “B” can also be seen, as can those indicated by the letter “C”. (Other interesting correspondences can also be observed, if desired.)

When comparing the SWTs corresponding to a given type of electroencephalographic recordings, care must be taken to use recordings made during the same Δt and with the same sampling frequency.

The first of several computational tools for the use of the SWT to be made available for interested users has been installed on the website of the Applied Mathematics and Computer Simulation Group (www.appliedmathgroup.org). This tool is that of the Square Wave Transform (SWT) and makes it possible to obtain the SWT of electroencephalographic recordings automatically [7].

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