

Truncation effects on muscular fatigue indexes based on M waves analysis

Maxime Yochum, Toufik Bakir, Romuald Lepers and Stéphane Binczak

Abstract—In this paper, we investigate muscular fatigue. We propose a new fatigue index based on the continuous wavelet transform (CWT) and compare it with the standard fatigue indexes from literature. Fatigue indexes are all based on the electrical activity of muscles (electromyogram) acquired during an electrically stimulated contraction (ES). The stimulator and electromyogram system, which were presented in a previous work, allows real-time analysis. The extracted fatigue parameters are compared between each other and their sensitivity to noise is studied. The effect of truncation of M waves is then investigated, enlightening the robustness of the index obtained using CWT.

I. INTRODUCTION

Since last decades, researchers are trying to characterize muscles behavior during contractions. These analysis permit to obtain for instance useful information leading to diagnose muscle diseases [1] [2]. These methods are specially adapted for paralyzed patients as acquired signal analysis could replace usual muscular feeling which is not perceived [3]. The most common signal to extract muscular data is the electromyogram (EMG) [4]. It represents electrical muscular activity of muscular fibers under electrodes. These electrical activities can be acquired during voluntary contractions or during electrical stimulations. Our study is focused on the determination of the muscular fatigue through EMG. Some fatigue indexes are already widely used to estimate the muscular fatigue during contractions. It is the case for the peak to peak (PTP), the root mean square (RMS), the mean (F_{mean}) and the median frequency (F_{med}) [5]–[7]. Those indexes are suitable for electrical contractions obtained with electrostimulators [8]. Our previous studies [9] [10] proposed a new fatigue index based on wavelet computation with continuous wavelet transform (CWT) named I_{CWT} . Those indexes are performed in real time during an electrical stimulated contraction thanks to an electrical stimulator delivering constant current pulses. The purpose of this paper is to analyze and compare these fatigue indexes between each other and to check their robustness to noise. Furthermore, this study includes also the influence of truncation which appears in case of high frequency stimulations or when the muscle fatigue becomes large enough during moderate frequency stimulations. Indeed, it is proved that the truncations affect the fatigue algorithms results [11]. Therefore, it is worthwhile to

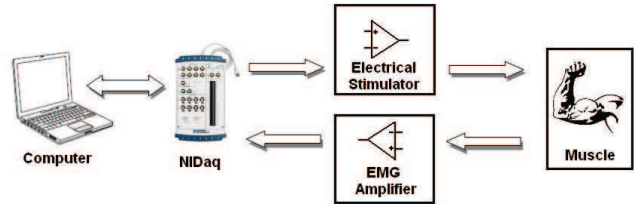


Fig. 1. General diagram of the system

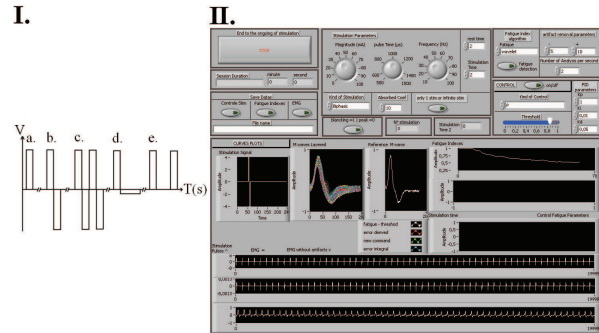


Fig. 2. I : Waveforms that can be sent to the stimulation board. a. Monophasic. b. Biphasic. c. Dual Biphasic. d. Absorbed Biphasic. e. doublet Nlet [12]. II : Software interface

study the influence of truncations on fatigue indexes, which is the main contribution of this paper.

II. MATERIAL

A. Experimental Setup

The experimental setup is composed of an electrical stimulation circuit and a real time EMG acquisition part [9] [10]. A general diagram of the global system is shown in Fig. 1. The electrical stimulator generates controlled current pulses to have a constant stimulation over time independently of various impedance skin values. It can supply a maximum of 100 mA for a body impedance of 1.5 KΩ between the both stimulation electrodes, a frequency pulse train from 10 Hz to 100 Hz and a pulse time from 0.5 ms to 2 ms for five different kind of pulse shapes shown in Fig. 2.I. which are the most common ones in literature. The EMG amplifier acquires EMG during the stimulation from bipolar electrodes laid on muscle and a reference electrode. Amplifiers used for the system are INA128 from Texas instruments with a 120 dB of Common Mode Rejection Ratio (CMRR) yielding a good removal of common voltage of bipolar electrodes. The EMG is used to find M waves over the stimulation. A software allows to manage stimulation parameters and to estimate

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the fatigue indexes which are obtained using the algorithms presented in the next section. A screenshot of the software interface is shown in Fig. 2.II. A processing is also included in order to remove stimulation artifacts [9] [10] which are replaced by null values.

B. Fatigue Indexes extraction

Each fatigue index is computed for each M waves of a stimulation train. The voltage of M waves is called $V_{EMG}(t)$. The fatigue index based on CWT (I_{CWT}) has been detailed in [9] [10]. Briefly, it gives an estimation of the temporal dilation undergone by M waves during an electrical stimulation. I_{CWT} is computed thanks to the CWT equation

$$C_{a,b}(V_{EMG}, \hat{\psi}(t)) = \int_{\mathbb{R}} V_{EMG} \frac{1}{\sqrt{a}} \hat{\psi}\left(\frac{t-b}{a}\right), \quad (1)$$

where $\hat{\psi}$ is an estimated wavelet obtained from the first M wave of the stimulation, a is a scale factor (the temporal dilation) and b indicates the temporal location. Then, the a scale parameter is taken such that the corresponding $C_{a,b}(V_{EMG}, \hat{\psi}(t))$ reaches its maximum value. Therefore, the a scale parameter represents the temporal dilation of M wave. The other fatigue indexes from literature tend to decrease over time while the a scale parameter increases. So, in order to keep a common tendency, the fatigue index constructed with parameter a is defined such as

$$I_{CWT} = \left[\underset{a}{\operatorname{argmax}} \{C_{a,b}(s(t), \hat{\psi}(t))\} \right]^{-1}. \quad (2)$$

Four fatigue indexes from literature [5]–[8] [13] [14] are also implemented. The PTP parameter is related to the amplitude of M waves and is computed as

$$PTP = V_{EMG}(max) - V_{EMG}(min). \quad (3)$$

The RMS parameter is computed from the total surface area of M waves, such as

$$RMS = \sqrt{\sum_{i=0}^n V_{EMG}^2}. \quad (4)$$

The two last fatigue indexes are obtained in the frequency domain. The first one is the mean frequency which is an average of the total power spectrum, computed such as

$$F_{mean} = \frac{\sum_{i=0}^n PSD_i \cdot f_i}{\sum_{i=0}^n PSD_i}, \quad (5)$$

where PSD is the power spectrum density of V_{EMG} and f is the frequency vector. The second one is the median frequency, which is the frequency separating the total spectrum density in two identical parts :

$$\sum_{i=0}^{F_{med}} PSD_i = \sum_{i=F_{med}}^n PSD_i = \frac{1}{2} \cdot TSD, \quad (6)$$

where TSD is the total spectrum density.

All previous fatigue indexes are normalized to vary between 0 and 1. The fatigue indexes are normalized as :

$$nF_{index} = \frac{F_{index}}{\max(F_{index})}, \quad (7)$$

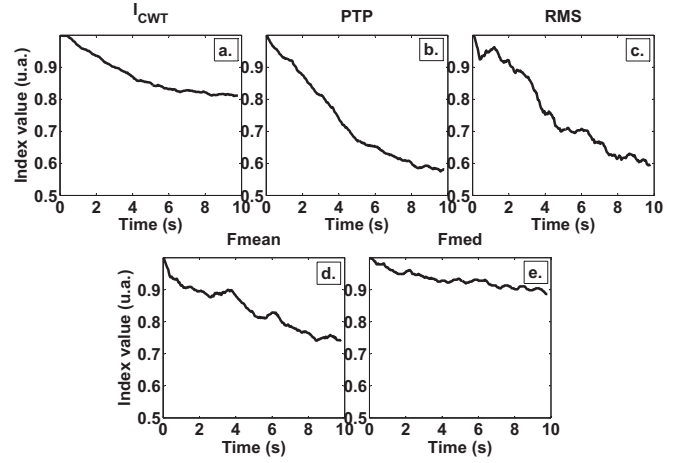


Fig. 3. Ewemple of fatigue indexes : a. CWT based (I_{CWT}), b. Peak to Peak (PTP), c. Root Mean Square (RMS), d. Mean Frequency (F_{mean}), e. Median Frequency (F_{med}).

where F_{index} is the fatigue index among I_{CWT} , PTP, RMS, F_{mean} and F_{med} . In Fig. 3, an illustration of these fatigue index results are given, which correspond to the experiments presented in the following section. Initially, they all are set to one due to normalization. Then, they all decrease in different ways as the fatigue is increasing during the stimulation. Those changes are linked to physiological changes of muscular fibers [13] [8].

C. Experimental results

For our experiments, the pulses parameters have been set in order to deliver a biphasic symmetric stimulation with a 30,40,50,60 and 70 mA intensity, a 1 ms pulse duration, a 20,30,40,50 and 60 Hz pulse train frequency and a total duration of stimulation of 10 s. All stimulations have been applied on the biceps brachia isometrically, the position of arm and forearm were fixed. An example of experimental results is shown in Fig. 3. It is interesting to observe the

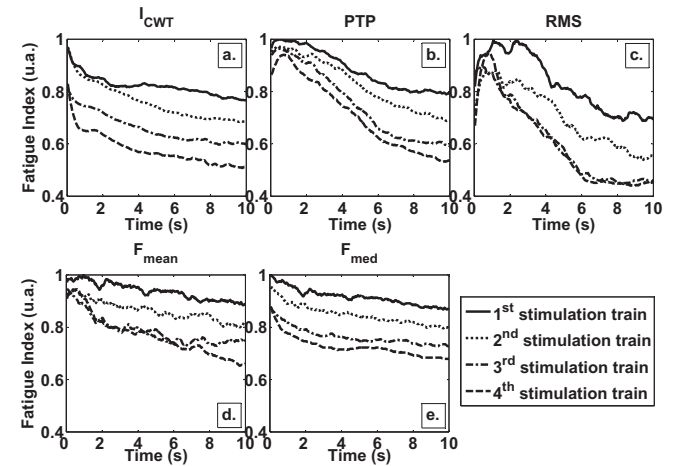


Fig. 4. Example of fatigue indexes for 4 consecutive contractions at 10 s of stimulation and 10 s of rest time : a. I_{CWT} , b. PTP, c. RMS, d. F_{mean} , e. F_{med} .

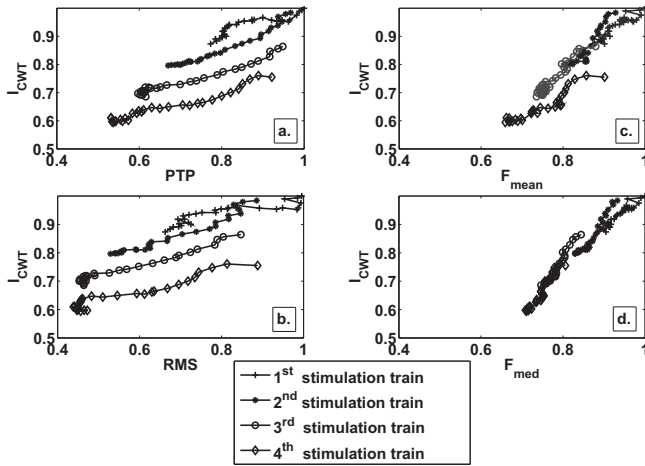


Fig. 5. Example of XY curves with : a. I_{CWT} and PTP, b. I_{CWT} and RMS, c. I_{CWT} and F_{mean} , d. I_{CWT} and F_{med}

differences of fatigue indexes during several consecutive contractions exercises. In the Fig. 4, fatigue indexes have been plotted for 4 consecutive 10 s stimulations with a rest time between two stimulations of 10 s. For all five fatigue index algorithms, the fatigue becomes more and more important over the stimulation trains. The fatigue index based on CWT follows the same tendency. This decrease to low values could be explained by the fact that the muscle does not have the time to recover its initial condition during the 10 s of rest, therefore the muscle becomes increasingly tired.

III. EXPERIMENTAL FATIGUE INDEXES COMPARISON

A. Qualitative comparison

In order to compare I_{CWT} to the others fatigue indexes (PTP, RMS, F_{mean} and F_{med}), different XY displays have been plotted as in Fig. 5. The Y axes are the I_{CWT} indexes and the X axes are the common fatigue indexes. This representation leads to observe the differences between I_{CWT} and the other indexes, yielding a qualitative relationship between them. For example, the curves in Fig. 5.a (I_{CWT} with PTP) shows that the results for each stimulation train are different. There is clearly a drift between the two indexes during the experiments. Therefore, there is no trivial correlation between I_{CWT} and PTP. We observe the same behavior between I_{CWT} and RMS (see Fig. 5.b). As these fatigue indexes are not equivalent, their combination could bring more physiological information when measuring amplitudes, shapes and frequencies aspects of M waves voltages. We can note that the plots c. and d. in the Fig. 5 show that there is a linear relationship between I_{CWT} and F_{mean} . It is more marked with F_{med} . Indeed, the shape of M waves is directly related to their frequencies.

B. Noise sensitivity

V_{EMG} are very little signals that can engender a significant noise to signal ratio (NSR). Therefore, it may be useful to study the influence of noise on the process. A noise has been added on V_{EMG} with an arbitrary amplitude of 10 %

TABLE I
MEAN ABSOLUTE ERROR FOR 5 INDEXES (%)

Fatigue index	PTP	RMS	F_{mean}	F_{med}	I_{CWT}
Noisy Signal	34.29	68.61	204.48	173.18	6.69

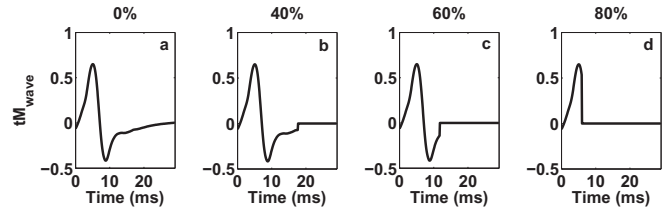


Fig. 6. Example of truncation of M wave for several percentage : a. 0 % b. 40 % c. 60 % d. 80 %

of the maximum of V_{EMG} values. Then, fatigue algorithms are applied and mean square errors have been computed between the fatigue results without noise and the fatigue results with noise. On table I, the mean square error is shown for the five fatigue parameters. The error for I_{CWT} is rather low compared to the others fatigue indexes. Therefore, I_{CWT} is less noise dependent than the standard fatigue indexes, reinforcing the interest to use the waveform of M waves rather than only the amplitudes or the frequencies.

C. Truncation influence

During electrical stimulation, M waves are acquired between two stimulation pulses. If the frequencies of the stimulation pulse train are too high, the M waves are not complete; The end of each M waves is superimposed with the following stimulation pulse. Due to the fact that the stimulation amplitude is much larger than the EMG amplitude and that the electrodes acquire the total signal, it yields the impossibility to extract this EMG part. This phenomenon is called the truncation. Due to the dilation of M waves during the electrical stimulation, this phenomenon tends to increase over the stimulation. Therefore, it can also occur for moderate stimulation frequencies. In order to check this truncation sensitivity, artificial truncations have been included in V_{EMG} signals. A constant truncation value has been applied for the whole exercise, such as

$$tM_{wave}(x) = H(l(1 - \frac{i}{100}) - x) \cdot M_{wave}, \quad (8)$$

where tM_{wave} is the truncated M wave, H is the Heaviside function, l is the size on the M wave and i is the applied percentage of truncation. This task is performed for a wide range of truncation values. An example is shown on the Fig. 6 where the evolution of truncation is depicted related to 0%, 40%, 60% and 80% of truncation. The five fatigue indexes computations are then applied on the tM_{waves} signal. The Fig. 7 shows the results of fatigue estimation for five truncation percentages (0%, 20%, 40%, 60% and 80%). We can note that more the truncation is important, worse the results are. It is not surprising because the signals include less and less information if the truncation is higher. However, it is difficult to observe which index is less sensitive to truncations. To

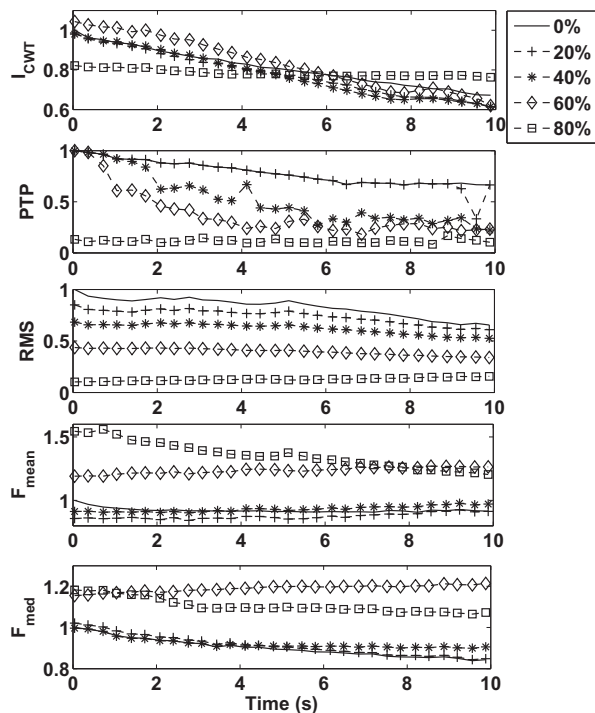


Fig. 7. Example of fatigue algorithms results (I_{CWT} , PTP, RMS, F_{mean} and F_{med}) for different level of truncation

quantify this truncation sensitivity, a quadratic error $QE(t)$ is calculated between the results with truncation and the reference results (which is for 0% of truncation), as

$$QE(t) = \sqrt{\sum_{i=1}^N (R_{ref}(i) - R_t(i))^2}, \quad (9)$$

where t is the percentage of truncation, N is the number of fatigue indexes, R_{ref} is the reference indexes and R_t is the indexes for truncated signal by $t\%$. An example of this quantification is shown on the Fig. 8. We can see that the error for I_{CWT} (continuous + line) is the lowest one. Therefore, this fatigue index is the less truncation dependent with respect to the others. It is more interesting to use waveform dilation quantification (I_{CWT}) for M waves corrupted by truncation.

IV. CONCLUSION & DISCUSSION

This work uses five fatigue indexes computed from experimental M waves obtained from electromyograms acquired during electrical stimulations. This study discusses about a comparison of several fatigue indexes and their influences of noise and truncation. This study shows that the I_{CWT} fatigue index based on wavelet is less noise and truncation dependent than the standard fatigue indexes from literature. This index could be therefore relevant in case of a high noise or truncation level, which can exist in EMG as for high frequency pulse train. In addition we have seen that the dilation of M waves was directly related to the frequencies of M waves and not directly related to the M waves amplitudes. It should be interesting to verify that the new fatigue index keeps its

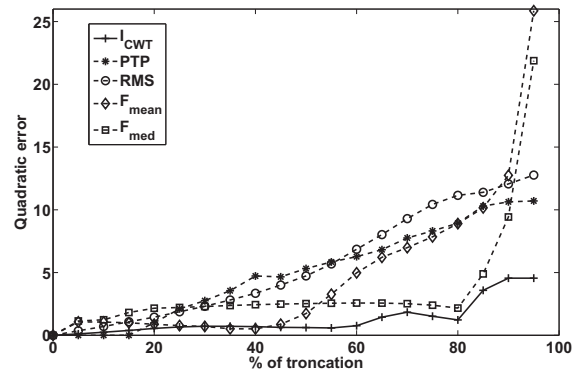


Fig. 8. Error of fatigue algorithms results (I_{CWT} , PTP, RMS, F_{mean} and F_{med}) for different level of truncation

valuable properties when computed on EMG obtained from different kind of exercises and different muscles.

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