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Assessment of Sedation-Analgesia by means of Poincaré Analysis of the Electroencephalogram *

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Abstract— Monitoring the levels of sedation-analgesia may be helpful for managing patient stress on minimally invasive medical procedures. Monitors based on EEG analysis and designed to assess general anesthesia cannot distinguish reliably between a light and deep sedation. In this work, the Poincaré plot is used as a nonlinear technique applied to EEG signals in order to characterize the levels of sedation-analgesia, according to observed categorical responses that were evaluated by means of Ramsay Sedation Scale (RSS). To study the effect of high frequencies due to EMG activity, three different frequency ranges (FR1=0.5-110 Hz, FR2=0.5-30 Hz and FR3=30-110 Hz) were considered. Indexes from power spectral analysis and plasma concentration of propofol and remifentanil were also compared with the bispectral index BIS. An adaptive Neurofuzzy Inference System was applied to model the interaction of the best indexes with respect to RSS score for each analysis, and leave-one-out cross validation method was used. The ability of the indexes to describe the level of sedation-analgesia, according with the RSS score, was evaluated using the prediction probability (Pk). The results showed that the ratio $SD1/SD2_{FR3}$ contains useful information about the sedation level, and $SD1_{FR2}$ and $SD2_{FR2}$ had the best performance classifying response to noxious stimuli. Models including parameters from Poincaré plot emerge as a good estimator of sedation-analgesia levels.

I. INTRODUCTION

Adequate sedation and analgesia may be helpful for managing patient stress on minimally invasive medical procedures as endoscopies. Monitoring the levels of sedation-analgesia may provide appropriate requirements to control the administration of sedative drugs, minimizing the impact on aggression and on outcome process in the patient. Various anesthesia monitors have been introduced into clinical practice [1]-[3], providing different

*Research supported by COLCIENCIAS (1217-669-45428 / 297-2015) from Colombia government; MINECO (TEC2013-44666-R and TEC2014-60337-R) from Spanish Government. CIBER of Bioengineering, Biomaterials and Nanomedicine is an initiative of ISCIII. The End of Residency Award of Hospital CLINIC de Barcelona and the FIS (Fondo de Investigaciones Sanitarias, Health Department, Government of Spain) grants nº PI/050072 and PS09/01209.

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electroencephalogram (EEG) indexes to predict loss of consciousness, such as bispectral (BIS) index, auditory-evoked potential (AEP) index and cerebral state index (CSI). Nevertheless, these monitoring systems may offer inadequate anesthesia detection due to different reasons [4]-[6], such as due to the time varying dynamics involved in the EEG cerebral function that is very sensitive to the state of the patient. Therefore, monitors designed to assess deep levels of sedation as in general anesthesia cannot distinguish reliably between a light and deep sedation [7].

Although recently studies based on nonlinear signal processing dynamics in the brain could determine certain associated changes between EEG complexities [8] and the prediction of the painful stimulation in the conscious and unconscious state induced by anesthesia, these EEG changes are not fully understood. This suggests to extend the analysis of EEG signals to more methods in the time-domain and spectral-domain analysis, in order to find robust predictors of the level of sedation-analgesia. An interesting and simple nonlinear method is the Poincaré plot, which can analyze the variability of time series by describing the behavior of the signal in a phase-space trajectory as a function of a constant time delay. Some published works [9],[10] state that it is a valuable method due to its ability to exhibit nonlinear features in the time domain series, and it can be used in the analysis of nonfiltered and also nonstationary data. Therefore, it can be a useful method to explain complex brain dynamic regulatory processes during anesthesia, sleep and consciousness.

In this work, the Poincaré plot is used to characterize the levels of sedation-analgesia by analyzing EEG signals. Because the slow response to painful stimulation still remains an open problem, the prediction of these responses using the proposed indexes is going to be quantified. Indexes from power spectral analysis and plasma concentration of propofol and remifentanil are also compared with the bispectral index BIS.

II. METHODOLOGY

A. Data Base

Data were recorded, after receiving approval from the Ethics Committee of Hospital Clinic de Barcelona and signed informed consent, from 110 patients scheduled to undergo ultrasonographic endoscopy (USE) under sedation-analgesia. USE is a relatively long procedure with periods of stability of effect, allowing study of the repercussion of painful stimulus on the level of sedation. Every patient was routinely monitored, including: effect-site predicted concentrations of propofol (Cepró) and remifentanil (Ceremi) from a target

controlled infusion system (FreseniusVial, Chemin de Fer, Béziers, France). The raw EEG signal was recorded with a sampling frequency of 900 Hz, resolution of 16 bits and a recording time of about 60 min (acquired with the AEP monitor/2 - Danmeter, Odense, Denmark); and BIS of the EEG (continuously measured with an A2000 monitor - Aspect Medical Systems, Newton, MA). Observed categorical responses were evaluated by means of Ramsay Sedation Scale (RSS) [11]. RSS score (Table I) was estimated at random times during the procedure in order to avoid factors correlated with time, which could confound the results of the RSS measurements. In this study, the whole database contains annotated RSS scores from 2 to 6.

TABLE I. THE RAMSAY SEDATION SCALE

Score	Description
1	Patient awake, anxious, agitated or restless
2	Patient awake, cooperative, orientated and tranquil
3	Patient drowsy with response to commands
4	Patient asleep, brisk response to glabella tap or loud auditory stimulus
5	Patient asleep, sluggish response to stimulus
6	No response to firm nail-bed pressure or other noxious stimuli.

B. Preprocessing

EEG signals were resampled at 256 Hz after applying a Chebyshev low pass filter of 6th order with cut-off frequency of 127Hz. Then, the EEG signals were segmented into windows of 1 minute length between 90 s and 30 s before the response annotation of RSS. Assuming that, in an ideal situation, the sedation level should be constant if plasma concentration of the anesthetic and analgesic agents remains without changes, the annotated RSS was assigned to the previous 1-minute length window if the differences in the predicted concentrations of remifentanyl ($\Delta CeRemi$) and propofol ($\Delta CeProp$), calculated between the first and the last second of the window, were $\Delta CeRemi < 0.1$ ng/ml and $\Delta CeProp < 0.1$ μ g/ml. Otherwise, the window was cut at the sample where the conditions were satisfied. EEG amplitudes out of the range of ± 200 μ V were considered as artifact.

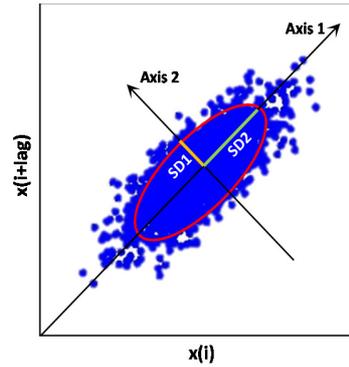
C. Poincare Plot

Poincare plot or return map is a phase space where a time series is represented in a Cartesian plane [12]. A phase space allows studying variations in a signal only with respect to itself. Given a time series $x(i)$, $i = 1, \dots, N$, the Poincaré plot is constructed plotting $x(i)$ against $x(i+lag)$, where the parameter lag is the time delay between x and y coordinates of the phase space (Fig. 1). Two standard descriptors used in quantifying the plot are $SD1$ and $SD2$, which can be obtained by fitting an ellipse to the plot shape as shown in Fig. 1. In terms of linear statistics, $SD1$ measures the standard deviation of the points perpendicular to the line of identity and determines the width of the ellipse (short-term variability), whereas $SD2$ measures the standard deviation along the line of identity and determines the length of the ellipse (long-term variability) [10]. The $SD1/SD2$ ratio characterizes the sharpness of the scatter pattern. According with [10] a lag of 8 ms was fixed in this work.

Epochs of 10 seconds were considered in the Poincaré analysis, with an overlapping of 90% between consecutive

epochs, allowing $SD1$, $SD2$ and $SD1/SD2$ ratio to be obtained each second. Also, to study the effect of high frequencies due to EMG signal content, the Poincaré analysis was applied to the time series in three different frequency ranges: $FR1=0.5-110$ Hz, $FR2=0.5-30$ Hz (low frequency range) and $FR3=30-110$ Hz (high frequency range). Although EEG and EMG overlap their frequency range, $FR2$ emphasizes the EEG components and $FR3$ the EMG components.

Figure 1. Poincaré plot and descriptors $SD1$ and $SD2$



D. Power Spectral Analysis

The same epochs considered in the Poincaré analysis were also used for power spectral analysis. The power spectral density (PSD) for each EEG epoch was calculated using the FFT after the application of a Hamming window. Then, spectral power in each band (P_δ , 0.1–4 Hz; P_θ , 4–8 Hz; P_α , 8–12 Hz; P_β , 12–30 Hz) was computed as the area under the normalized PSD curve for the given frequency range. Spectral edge frequency 95% ($SEF95$) and spectral entropy ($SpEn$) were also calculated. $SEF95$ represents the frequency below which 95% of the power in the spectrum resides. This value **decreases** during induction of anesthesia as the power in high frequencies shifts toward lower frequencies [13]. $SpEn$ quantifies the irregularity or complexity in the EEG. This value is low when the signal frequency components are concentrated on a small set of the spectrum, and increases the more uniformly the frequency components spread to all possible values of the spectrum [13].

E. Statistics

The ability of the indexes to describe the level of sedation-analgesia, according with the RSS score, was evaluated using the prediction probability (P_k), which compares the performance of indicators [14]. The P_k coefficient is a statistic commonly used to measure how well an index predicts the state of the patient. A P_k of 1 represents a perfect prediction and 0.5 is not better than tossing a fair coin. Kruskal-Wallis test was applied in order to find statistically significant differences between RSS groups and the significance level was set at p-value < 0.01 . Two analysis were considered: i) Trial 1, taking into account all the different RSS groups (RSS2, RSS3, RSS4, RSS5 and RSS6); ii) Trial 2, only considering groups RSS5 and RSS6, which correspond to the observed categorical responses after nail bed compression (noxious stimuli). Adaptive Neurofuzzy Inference System (ANFIS) was applied to model the interaction of the best

indexes with respect to RSS score for each analysis, and leave-one-out cross validation method was used. The model was fit to the data by iteratively minimizing the root mean square errors using a backpropagation gradient descent method. In Trail 2, accuracy (*Acc*) was also computed.

III. RESULTS

A. Assessment the level of sedation

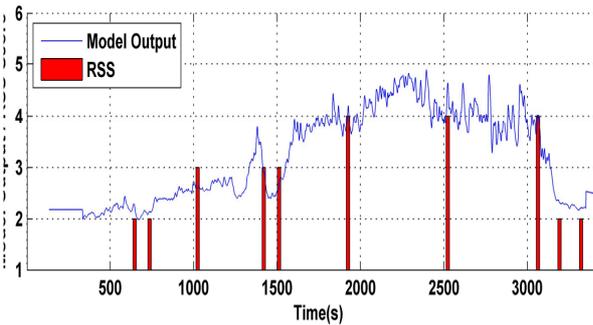
Table II shows the mean and the standard deviation, for each RSS of the indexes proposed in this work for Trial 1.

TABLE II. STATISTICAL PARAMETERS FOR EACH RAMSAY SEDATION SCALE: FR1=0.5-110 Hz, FR2=0.5-30 Hz AND FR3=30-110 Hz

Index	RSS2 mean±std	RSS3 mean±std	RSS4 mean±std	RSS5 mean±std	RSS6 mean±std	Pk
<i>BIS</i>	93.0±7.6	82.0±10.8	74.0±12.0	65.6±14.2	64.4±14.6	0.778
<i>CeRemi</i>	0.557±0.880	1.10±0.827	1.25±0.824	0.967±0.876	1.45±0.728	0.614
<i>CePropo</i>	0.527±0.792	1.90±0.712	2.16±0.670	2.39±0.648	2.19±0.628	0.716
P_{δ}	0.542±0.224	0.295±0.210	0.245±0.170	0.230±0.171	0.281±0.168	0.625
P_{α}	0.084±0.070	0.177±0.109	0.260±0.136	0.317±0.141	0.317±0.144	0.755
P_{β}	0.201±0.212	0.390±0.207	0.342±0.173	0.288±0.154	0.216±0.128	0.509
<i>SEF95</i>	18.2±7.32	22.5±4.34	20.5±3.70	18.9±3.12	18.2±2.92	0.588
<i>SpEn</i>	0.776±0.082	0.846±0.062	0.839±0.049	0.826±0.048	0.818±0.040	0.501
$SD1_{FR1}$	10.8±7.25	9.60±6.41	8.93±5.13	9.02±5.51	8.14±5.55	0.553
$SD2_{FR1}$	22.4±12.0	15.6±7.27	16.0±6.67	17.2±7.05	14.6±7.32	0.576
$SD1/SD2_{FR1}$	0.526±0.282	0.605±0.233	0.550±0.195	0.504±0.169	0.531±0.169	0.505
$SD1_{FR2}$	3.22±1.34	3.22±1.34	3.27±1.36	3.48±1.34	2.55±1.32	0.550
$SD2_{FR2}$	20.1±12.2	12.1±6.06	12.4±6.10	13.6±5.90	10.6±5.53	0.606
$SD1/SD2_{FR2}$	0.193±0.091	0.283±0.079	0.273±0.057	0.259±0.047	0.241±0.043	0.532
$SD1_{FR3}$	9.88±7.26	8.60±6.61	7.85±5.48	7.82±5.98	7.42±5.77	0.553
$SD2_{FR3}$	7.21±5.69	8.00±6.51	8.34±6.04	8.74±7.02	8.47±7.04	0.520
$SD1/SD2_{FR3}$	1.45±0.35	1.18±0.34	1.01±0.26	0.981±0.244	0.966±0.252	0.711

Indexes *CePropo*, P_{α} and $SD1/SD2_{FR3}$ were used as inputs of an ANFIS system (implemented in MATLAB) in order to model the RSS score, achieving a $Pk = 0.791$. Figure 2 shows the course of the ANFIS's output and the observed categorical responses of one EEG record.

Figure 2. Course of the ANFIS's output (blue line) and the observed categorical responses according to the RSS (red line).



B. Assessment the level of analgesia (nociception)

Table III includes the indexes with $Pk \geq 0.60$ for Trial 2, when the observed categorical responses were classified in RSS5 and RSS6 after the application of nail bed painful stimulation, which is a kind of noxious stimuli. Although the indexes *BIS*, *SEF95* and *SpEn* have $Pk < 0.55$, they were also

Also, the *Pk* computed considering all the scales is presented. Almost all the indexes showed a significant p-value between consecutive RSS groups, with the exception of: P_{α} (RSS5 vs RSS6), $SD1_{FR2}$ (RSS2 vs RSS3), and the indexes $SD1_{FR1}$, $SD1_{FR3}$ and $SD2_{FR3}$ in (RSS4 vs RSS5). Only three indexes (*BIS*, $SD1_{FR3}$ and $SD1/SD2_{FR3}$) have a monotonic increase or decrease of the mean value as function of the RSS, though the standard deviation of $SD1_{FR3}$ was relatively very high, which is reflected in a low *Pk* value. The indexes with a $Pk \geq 0.70$ were: *BIS*, *CePropo*, P_{α} and $SD1/SD2_{FR3}$.

included as reference. The individual index with the highest *Pk* (0.714) was $SD1_{FR2}$, while the ANFIS model showed the best *Pk* (0.780). The ANFIS system was trained using 4 inputs (*CeRemi*, P_{β} , $SD1_{FR2}$ and $SD1_{FR2}$) in order to estimate the scales RSS5 and RSS6, as a model to predict the response to noxious stimuli. The highest *Acc* (70.9 %) was also obtained with the output of the ANFIS model.

TABLE III. STATISTICS USING RSS=5 AND RSS=6 AS REFERENCE

Index	<i>BIS</i>	<i>CeRemi</i>	P_{β}	<i>SEF95</i>	<i>SpEn</i>	$SD1_{FR2}$	$SD2_{FR2}$	ANFIS
<i>Pk</i>	0.528	0.675	0.647	0.544	0.570	0.714	0.699	0.780
<i>Acc (%)</i>	52.6	63.1	61.6	53.4	54.8	67.1	67.1	70.9

IV. DISCUSSION

In this work, Poincaré analysis was applied to EEG signals in three different frequency ranges: FR1=0.5-110 Hz, FR2=0.5-30 Hz and FR3=30-110 Hz, where FR2 emphasizes the EEG components and FR3 the EMG components. The assessment of the sedation, according with the RSS scores (Table II), showed that the ratio $SD1/SD2_{FR3}$, which was computed in the frequency range between 30-110 Hz, contains useful information about the sedation level. A significant decrease in the $SD1/SD2_{FR3}$ ratio was observed with each stepwise increase in RSS (p-value < 0.001 for each stepwise). Since in this study the patients were not under general

anesthesia but only under sedation, scalp and facial muscle activity are strongly present in the EEG records, especially in the groups with $RSS \leq 5$. This is a reason for the importance of the high frequency range in the estimation of the sedation level. This finding is according with [15], where it was established that quantitative facial surface electromyography enables discrimination of adequate vs. inadequate anesthesia, being also a useful measure of drug effect, vigilance levels and central nervous system integrity.

Hayashi et al. [10], in a research of Poincaré plot analysis in general anesthesia, indicated that the index $SD1/SD2$, computed in a frequency range of 0-30 Hz, may not be necessarily correlated with anesthesia depth in the near conscious level, and the index $SD2$ could estimate more accurately this light level of an anesthesia. In the present work, similarly $SD1/SD2_{FR2}$ was not able to estimate light sedation levels ($P_k=0.532$), and neither $SD2_{FR2}$ showed a good estimation ($P_k=0.606$). However, the index $SD1/SD2_{FR3}$ (in the range of 30-110 Hz) had a $P_k=0.711$, emerging as an index to be considered in the assessment of the sedation level. The individual indexes $SD1_{FR2}$ and $SD2_{FR2}$ achieve the best performance of classifying the responses in groups $RSS5$ and $RSS6$ with $Acc=67.1\%$. In this context, $SD1_{FR2}$ can include the information of the relative high frequency variations of the EEG in the frequency range 0.5-30 Hz (α and β activity), while $SD2$ may depend on the variations of the amplitude of the signals (shift from α to θ activity) [16]. It is important to note that δ activity is not strongly presented in this database where the mean BIS value was 64.4 in $RSS6$, the deepest sedation level.

Traditional spectral indexes as $SEF95$ and $SpEn$ have not shown a high P_k value neither in Trial 1 (0.588 and 0.501, respectively) nor Trial 2 (0.544 and 0.570, respectively). On the other hand, BIS and $CePropo$ had a similar behavior, being good estimators of RSS for Trial 1 but not for Trial 2, while $CeRemi$ was a better estimator in Trial 2 than in Trial 1. High levels of remifentanyl are associated with high level of analgesia, which are reflected in a reduction of the response to noxious stimuli, as it is observed in $RSS6$ ($CeRemi=1.45 \pm 0.728$ ng/mL). It is known that BIS is able to describe hypnotic effect as it was confirmed by results of Trial 1, but it cannot describe the analgesic effect and the response to noxious stimulation correctly. BIS might have the same value for different concentrations of drugs and it is also possible that in low doses of analgesia a response to noxious stimuli might be observed even at low BIS values [17].

It is important to point out that no advanced techniques were applied to reject artifacts in this work, suggesting that Poincaré analysis performed well with artifacts or can be improved with a good artifact rejection. Also, the Poincaré plot has the potential to be a good approach for short-duration dynamic signals and stable long EEG signals are not required for analysis [10].

V. CONCLUSION

In this work, Poincaré plot was used as a nonlinear technique applied to EEG signals in order to characterize the levels of sedation-analgesia, according to observed categorical responses that were evaluated by means of Ramsay Sedation Scale (RSS). The obtained results indicated that indexes obtained from filtered EEG signal from low to high frequency

range which contains EMG components improve the prediction of different stimuli responses in lightly sedated patients. Furthermore, the results showed that the ratio $SD1/SD2_{FR3}$ contains useful information about the sedation level, and $SD1_{FR2}$ and $SD2_{FR2}$ had the best performance classifying the noxious stimuli responses. Finally, it can be concluded that models including parameters from Poincaré plot **emerge** as a good estimator of sedation-analgesia levels.

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