

Statistical Investigation of the Effects of Fetal Heart Rate (FHR) and Uterine Contractions (UC) Signals on Apgar

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ABSTRACT

Apgar scoring is a method which is a fast and an effective way of understanding whether newborn babies are healthy or not. Any intervention to the baby such as resuscitation or intubation is done, if necessary, after evaluating the Apgar score and examining various physiological data. Shortening the response time for medical support is vitally important for babies. In this study, the physiological data gathered from mother and fetus along with the features extracted from FHR (fetal heart rate) and UC (uterus contraction) signals were examined in order to determine whether the newborn will have any immediate problems that will need medical support before the baby was actually born and a study towards Apgar scoring was made. There were two classes (intervention, non-intervention) used in this study. Data was analyzed by using Mann-Whitney U test. The classes were compared in terms of statistical data, FHR and UC signals and according to the obtained the results, intervention and non-intervention classes have been shown to have a significant difference for 5 extracted features. However, no significant changes have been detected in other features.

Keywords

Apgar Score; Cardiocography; Fetal Heart Rate (FHR); Uterine Contraction (UC); Mann-Whitney U Test.

1. INTRODUCTION

The initial assessment of the newborn right after birth is quite important. Whether the baby needs any intervention should be determined within seconds. For this purpose, it would be a great advantage to know whether the baby needs any post-natal intervention (resuscitation or intubation) before birth. Assessing the patients by scoring over general findings is a widely used method. The method of scoring used for the newborn is called the Apgar Score [1].

1.1 Apgar Score

The Apgar score is a simple and repeatable method used to assess the health of the newborn quickly and briefly right after birth. The Apgar scoring was developed in order to detect the effect of obstetric anaesthesia on babies, and is quite important for the neurologic development of the baby. The newborn are assessed within the scale between 0 and 2 for 5 different criteria, and the sum of the relevant values determines the apgar score of the baby between 0 and 10. The baby is applied the apgar scoring twice right after birth (at the first and fifth minutes). While a score of 8 and above is normal according to apgar scoring; for a score of 7 and below, interventions such as resuscitation (basic life support, bringing back to life) or intubation (opening of the artificial airway, providing the first breathing using tube) are applied on the babies [1] [2].

While there are studies related to the use of the FHR signal in the prediction of the diseases such as foetal acidosis (acid

poisoning), asphyxia (lack of oxygen) and hypoxia (decreased oxygen levels in the body) requiring resuscitation, using the FHR and UC signals for the determination of Apgar score by examining the physiological data as well is a new approach [2] [3] [4] [5].

1.2 Factors of Apgar Score

5 criteria taken into consideration in Apgar scoring are as follows.

1.2.1 Appearance/Complexion (Skin Colour)

The newborn are generally cyanotic (the appearance of a blue coloration of the skin or mucous membranes) due to low oxygen intake despite their capacity to carry high oxygen. Many children show cyanotic hand and foot symptoms for a few minutes despite excellent ventilation and oxygen supplementation. At the stage of scoring, 2 points are given when the hands, feet and body of the baby are pink; 1 point is given when there is acrocyanosis (cyanotic state of the hands and feet) while the body is pink; and 0 is given when the whole body of the baby is of bluish or purple colour.

1.2.2 Pulse (Heart Apex Beat)

2 points are given when the baby's heart beat rate is 100 and above; 1 point is given when it is below 100; and 0 is given when there is no heart beat at all.

1.2.3 Grimace (Response to Stimulus)

When oropharynx, or aspiration using a soft rubber catheter on the nostrils is applied (the mechanical cleansing of the mouth and nose of the baby), 2 points are given if the baby exhibits one of the responses of crying, sneezing, coughing or pulling itself back; 1 point is given if it makes a grimace or cries weakly; and 0 is given if it does not respond at all.

1.2.4 Activity (Muscular Tonus)

2 points are given to the babies whose body strength is good and flex their arms and legs on their own; 1 point is given to those who can bend them a bit; and 0 is given to those who fail to exhibit any resistance.

1.2.5 Respiration

2 points are given to the babies exhibiting normal respiration and crying heavily in the 60 seconds following birth; 1 point is given to those who breathe weekly and irregularly; and 0 is given to those who are apnoeic (cessation of the baby's breathing).

Looking at the apgar scoring, it can be decided whether the newborn baby needs any intervention. In this study, it is aimed to determine whether the baby would need any intervention using certain physiological data, FHR and UC signals prenatally. By this means, it will be possible to determine the situations that may require intervention before making apgar scoring, whereby the intervention time will be minimized and any physical damage to the baby can be

prevented. As a result of the study, the efficiency of the system will be checked by comparing the result of whether intervention is required or not required obtained prenatally with the post-natal apgar scoring.

2. MATERIAL AND METHOD

2.1 Data Collection

The database used in the study contains 132 carefully selected cardiocography (CTG) records of 9164 records collected in Czech Technical University (CTU) and University Hospital Brno (UHB) between 2010 and 2012; and all records were taken from Physionet [3]. The CTG data consist of records that were started to be recorded 90 minutes before birth, which are no longer than 90 minutes in length. Each CTG record contains foetal heart beat rates and uterine contraction signal exemplified in 4 Hz. Only the patient records that fulfil below criteria were used in order to create as homogenous a series as possible.

2.1.1 Factors Considerations in Patient Records

The assessment of the CTG data were made by 9 obstetricians looking at the variability and reliability signals for each signal, and sticking to the explanations of the signals (following the FIGO instructions used in Czech Republic).

Only patient records providing the following criteria are used to create a homogeneous array as possible.

- Singleton pregnancy
- Gestation period > 36 weeks
- Developmental defects previously unknown
- Second level labour time ≤ 30 minutes
- Quality of the FHR signal (percentage of record, when FHR data are appropriate) $> 50\%$ of each 30 minute period
- Having a blood gas analysis taken from umbilical artery
- Vaginal (normal) birth majority (only includes 46 caesarean deliveries)

Additional parameters collected from all records;

- Maternal data; the age of the mother, number of deliveries and pregnancies
- Birth data; type of birth (vaginal delivery and caesarean delivery), duration of delivery, fluid meconium, measurement type (ultrasound or direct cranial ultrasound)
- Foetal data; gender and birth weight
- Data of the newborn; analysis of the arterial blood sample (ph, pCO₂, pO₂, base deficit and calculated BDecf), apgar scoring, neonatologic assessment (O₂ need, seizures, entry into the newborn intensive care unit (NICU))

2.2 Properties of FHR and UC Signals

Data used in the study were started to be recorded 90 minutes before the birth. 10 properties were extracted from FHR and UC signals exemplified in 4 Hz in time space. These properties were started to be tested using Mann-Whitney U test, and it was endeavoured to determine whether there is a significant difference between them. The properties extracted from FHR and UC signals in time space are as follows:

1. Standard deviation [4] [5] [6]
2. Maximum value [7]
3. Coefficient of skewness [8] [9] [10] [11]
4. Coefficient of kurtosis [4] [10] [11]
5. Shape factor [12]
6. Variance [13] [14] [5] [6]
7. Hjorth activity parameter [15] [16]
8. Hjorth complexity parameter [15] [16]
9. Average curve length [17]
10. Average teager energy [17]

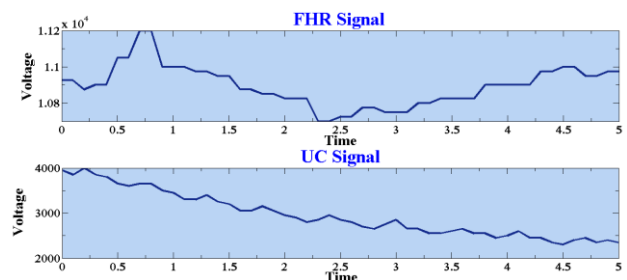


Fig. 1: Exemplary FHR and UC signals

2.3 Properties of Statistical Data

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2.3.1 Foetal Information

11. Week of gestation: Indicates how many weeks old the baby is.
12. Weight of the baby (g): Indicates the prenatal weight of the foetus.
13. Gender: Indicates the sex of the baby.

2.3.2 Maternal Risk Factors

14. Age: Indicates the age of the mother.
15. Gravidity: Indicates the number of pregnancy.
16. Parity: Indicates the number of birth.
17. Diabetes: A metabolic disorder occurring due to an increase in glucose levels in the blood under the influence of genetic and environmental factors.
18. Hypertension: A chronic condition occurring as a result of the rise of blood pressure in the arteries.
19. Preeclampsia: A complication that occurs in the presence of hypertension, high amounts of protein in the urine and toxic effects in the second half of pregnancy.
20. Liquid Precocious: Discharge of the amniotic fluid in the form of vaginal discharge as a result of the premature rupture of the amniotic membrane.
21. Hyperpyrexia: Elevation of body temperature.

22. Meconium: First stool of the foetus.

2.3.3 Birth Information

- 23. Presentation: Indicates the position of the baby's arrival during childbirth.
- 24. Induced labour: Indicates starting of the birth using stimuli such as artificial birth pain or drug.
- 25. 2nd Stage: Indicates the period from the full opening of the cervix to the birth of the baby.
- 26. Birth type: Indicates whether the birth is normal birth or caesarean birth.

2.4 Mann-Whitney U Test

The data set used in the study consists of a single group and two classes. First, the normality test was applied on the data set. Shapiro-Wilk table was examined as the number of the data used is above 50. As a result, it was seen that the distribution for 26 data used in the study is not normal distribution. Data used are ordered and consist of independent groups, however they are not parametric. This is the reason why the study was conducted using the SPSS (Statistical Package for the Social Sciences) program. As the Apgar score is an ordering data type, Mann-Whitney U test was chosen as the suitable analysis method. Mann-Whitney U test is a non-parametric statistical test used to examine whether two samples come from the same distribution. According to this test, data with a p-value that is lower than 0.05 are statistically significant, while those greater than 0.05 are statistically insignificant [4].

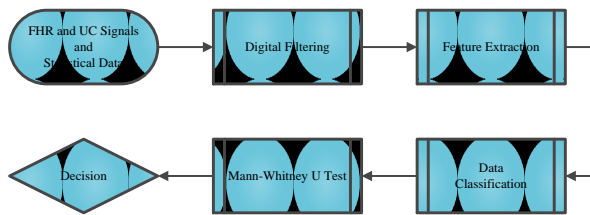


Fig 2. Exemplary FHR and UC signals

3. RESULTS

The steps of the process were fulfilled in line with the flow chart in figure 2, and the results obtained in accordance with Mann-Whitney U test were given in table 1, table 2 and table 3.

Means, standard deviations and p values of the data pertaining to FHR signals are shown in table 1.

Means, standard deviations and p values of the data pertaining to UC signals are shown in table 2.

While the properties in table 1 do not constitute a significant difference in terms of p-values, only the 9th property constitutes a significant difference in table 2. That the p value is lower than 0.004 means that this property is statistically significant and can be used in classification.

The mean, standard deviation and p values of the data related to foetus information, maternal risk factors and birth information are shown in table 3. In the light of the results in Table 3, it is seen that the p values of the properties no. 12, 13, 16 and 25 are lower than 0.05. Thus, the two classes called “requires intervention” and “does not require intervention” can be differentiated as two different classes in terms of these properties.

5 properties of the 26 properties extracted as a result of this study are statistically significant. These properties are the properties no. 12, 13, 16 and 25 related to foetus information, maternal risk factors and birth information; and property no. 9 extracted for the UC signal. There is a significant difference between the two classes in terms of these properties. When the properties no. 12, 13, 16, 25 used in the study and the property no. 9 extracted for the UC signal are used for the determination of the apgar score, it will provide a great ease in terms of establishing whether the babies require any post-natal intervention (resuscitation or incubation) before birth. This will reduce the intervention period, whereby minimising the foetal and neonatal deaths.

In future studies, new analyses can be conducted using different features to be obtained from FHR and UC signals.

Table 1. Means, standard deviations and p values of FHR signals

Properties	Requires Intervention		Does not Require Intervention		p
	Mean	Standard Deviation	Mean	Standard Deviation	
1	5.66E+03	± 1.41E+03	5.77E+03	± 1.25E+03	0.665
2	19662.50	± 2500.720	19532.95	± 2797.341	0.652
3	-	± 2.57E+00	-	± 2.48E+00	0.665
4	2.42E+01	± 4.12E+01	2.13E+01	± 3.69E+01	0.633
5	1.35E+02	± 1.40E+01	1.35E+02	± 1.35E+01	0.598
6	3.39E+07	± 1.56E+07	3.48E+07	± 1.33E+07	0.665
7	3.39E+07	± 1.56E+07	3.48E+07	± 1.33E+07	0.665
8	3.24E+00	± 1.17E+00	2.90E+00	± 9.73E-01	0.085
9	-	± 3.90E-01	-	± 4.61E-01	0.311
10	1.32E+06	± 1.10E+06	1.39E+06	± 9.55E+05	0.328

Table 2. Means, standard deviations and p values of UC signals

Properties	Requires Intervention		Does not Require Intervention		p
	Mean	Standard Deviation	Mean	Standard Deviation	
1	2.47E+03	± 7.77E+02	2.44E+03	± 6.03E+02	0.877
2	9899.242	± 2794.988	10020.45	± 2698.018	0.822
3	-	± 5.12E+00	-	± 5.19E+00	0.80

	6.33E+00	6.61E+00	6
4	1.21E+02 ± 1.05E+02	1.22E+02 ± 9.98E+01	0.799
5	9.38E+01 ± 2.24E+01	1.01E+02 ± 6.31E+01	0.917
6	6.72E+06 ± 5.09E+06	6.33E+06 ± 3.09E+06	0.877
7	6.72E+06 ± 5.09E+06	6.33E+06 ± 3.09E+06	0.877
8	6.63E+00 ± 2.29E+00	6.24E+00 ± 1.95E+00	0.518
9	- ± 2.00E-01	- ± 2.23E-01	0.001
10	1.25E+05 ± 2.25E+05	8.58E+04 ± 2.71E+04	0.193

Table 3. Means, standard deviations and p values of the statistical data

Properties	Requires Intervention		Does not Require Intervention		p
	Mean	Standard Deviation	Mean	Standard Deviation	
11	39.788	± 1.1302	84.015	± 358.2072	0.702
12	3441.364	± 443.0589	3243.591	± 626.4837	0.044
13	1.348	± 0.4801	2.000	± 3.6585	0.013
14	28.758	± 5.5666	30.091	± 5.9271	0.077
15	1.242	± 0.7455	1.530	± 1.0985	0.060
16	0.227	± 0.4573	0.591	± 0.7437	0.002
17	0.045	± 0.2099	0.091	± 0.2897	0.302
18	0.091	± 0.2897	0.045	± 0.2099	0.302
19	0.076	± 0.2666	0.030	± 0.1727	0.246
20	0.273	± 0.4488	0.288	± 0.4562	0.847
21	0.000	± 0.0000	0.000	± 0.0000	1.000
22	0.106	± 0.3103	0.030	± 0.1727	0.085
23	1.212	± 0.5411	1.167	± 0.5430	0.339
24	0.333	± 0.4750	0.379	± 0.4888	0.587
25	13.727	± 8.5459	8.545	± 7.5165	0.000
26	1.152	± 0.3613	1.197	± 0.4008	0.493

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