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| n_R | | | | | |
| mcd | | | μ | μ | μ |
| D | | | wavelet | μ | |
| A | | | wavelet | | |
| R | | μ | | | |
| d_k^j | Detail | | | j | k |
| a_L^j | Approximation | | | | j |
| $\hat{\Sigma}_\varepsilon$ | μ | | | μ | |
| \hat{D}_k | | μ | | wavelet | μ |
| X | | | $\mu\mu$ | | |
| λ_i | μ | | | | |
| n | | | | | |
| I | | | | | |
| e_i | | μ | | | |
| $s(t)$ | | | μ | μ | |
| $\hat{s}(t)$ | μ | μ | | μ | μ |
| w_{ij} | | μ | | | |
| $E\{s^2\}$ | | | | | |
| p_j | RMS | | | | s_j |
| P | | μ | | | |
| y_i | | μ | | | |
| $p(y_1, \dots, y_n)$ | | | | | |

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|----------------------------------|-----------------|-------|---------|-------|---------|
| μ | | μ | | μ | |
| σ^2 | | | | μ | |
| $kurt(x)$ | | | μ | | x |
| $H(x)$ | | | μ | | x |
| $J(X)$ | | | | μ | |
| X_N | | | μ | | |
| C_x | | | μ | | |
| c_{xx}^T | | | | μ | |
| \hat{s}_f | | μ | | | μ |
| \hat{x}_f | | μ | μ | | |
| $qSNR$ | | | | μ | |
| $W_x(s,t)$ | | μ | | | wavelet |
| $y(k)$ | μ | | | | |
| $s(i)$ | μ | | | | |
| b | | | FIR | | |
| Ts | | | μ | | |
| $\xi(w,b)$ | | | | | |
| | | μ | | μ | |
| $lead_i(t)$ | μ | | i | | μt |
| | | μ | μ | | |
| $SPWVD_x$ | μ | μ | | -WVD | μ |
| ω | | | | | |
| $g(\cdot), h(\cdot)$ | | | | | |
| $E(t)$ | | | μ | | |
| $t_{Q_i}^{end}$ | μ | | | μ | Q |
| $t_{S_i}^{start}, t_{S_i}^{end}$ | μ | | | | μS |
| $\psi(x)$ | -spline wavelet | μ | | | |
| f_c | | | | | wavelet |
| f_b | μ | | | | |
| $mCCWT$ | E | | wavelet | | |

| | | | |
|--------------------|--------|-------|-------------|
| <i>smCCWT</i> | $O\mu$ | μ | wavelet |
| $f(i)$ | | μ | |
| $V(a,b)$ | | μ | |
| w_i | | μ | |
| f | | μ | μ |
| Δf | | | μ μ |
| $\Delta^2 f$ | | | μ μ |
| λ | | | |
| σ | | | |
| H | | | μ |
| Λ | | μ | |
| R | | | |
| W | | | μ |
| $eSNR$ | , | μ | |
| $ecSNR$ | , | μ | μ |
| p_1, p_2 | | μ | |
| cor | | μ | |
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| <i>isoelectric</i> | | | $\mu\mu$ |

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« μ » « ».

μ , World Federation of Neurology Group

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μ (μμ -12 -16 mmol/L)¹³,
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μ μ [14,15].

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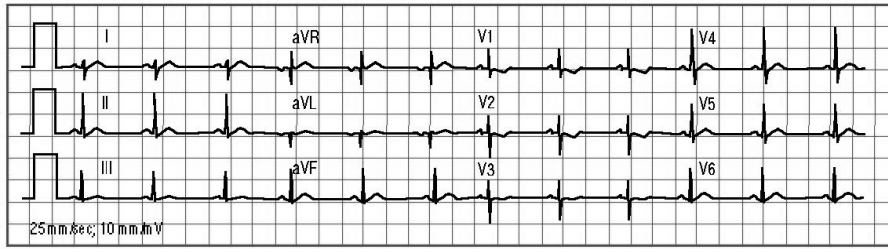
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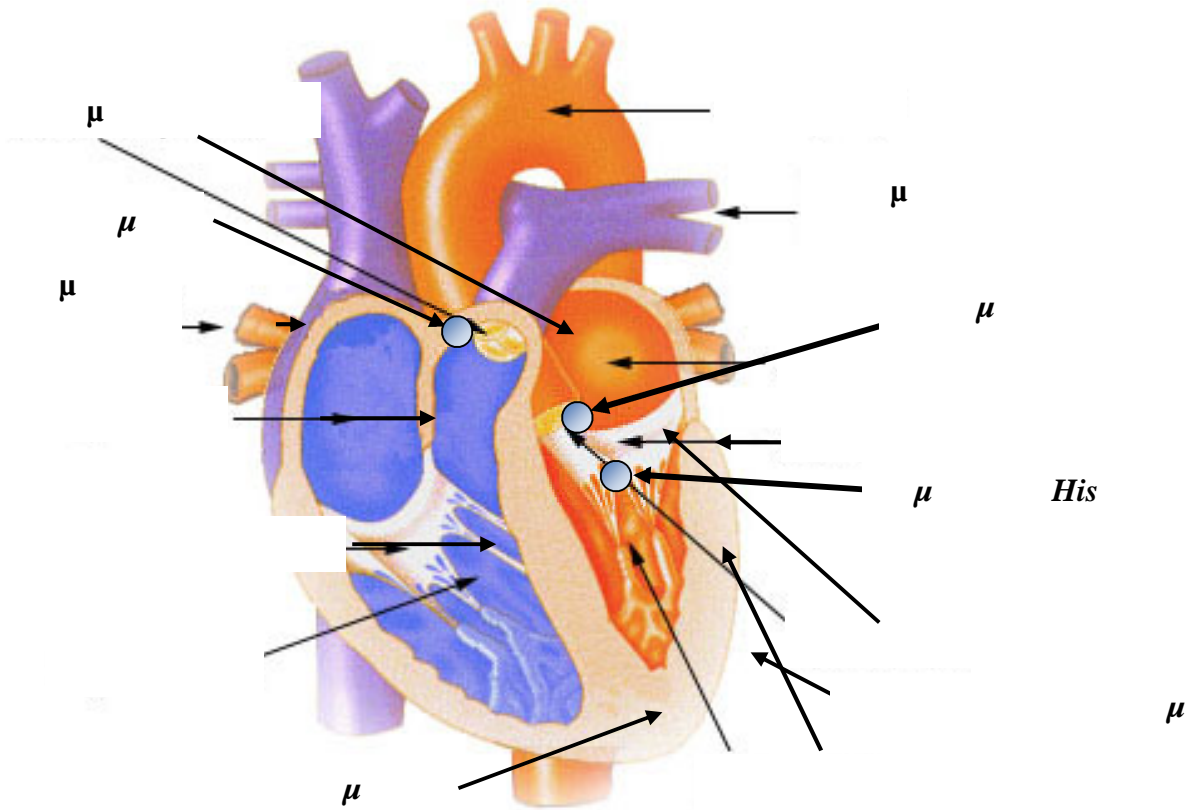
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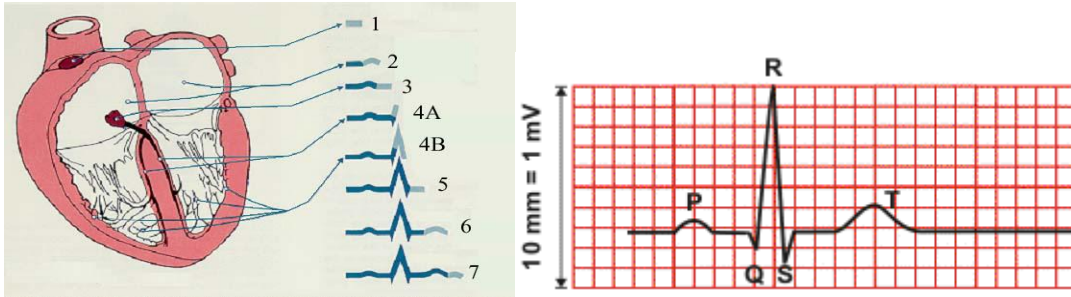
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μ : μ μ QRS, 4) μ ()

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: μ , 6) : μ , 7) μ

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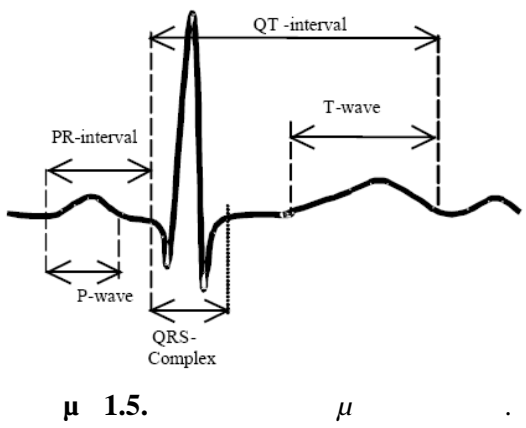
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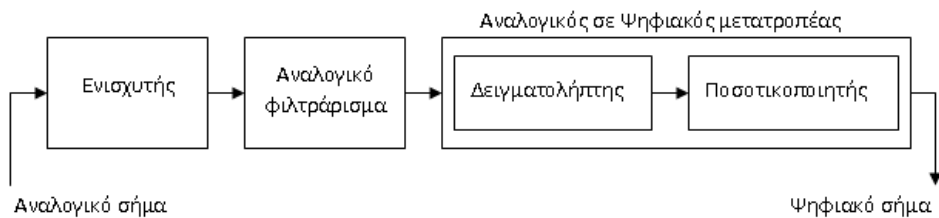
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2:

The figure shows a signal waveform with a QRS complex. The signal is plotted against time, and the QRS complex is labeled. The signal is noisy, and the QRS complex is the most prominent feature. The signal is labeled with μ at various points, indicating the mean value of the signal. The QRS complex is labeled with μ at its peak. The signal is labeled with μ at its trough. The signal is labeled with μ at its zero-crossing. The signal is labeled with μ at its maximum value. The signal is labeled with μ at its minimum value. The signal is labeled with μ at its average value. The signal is labeled with μ at its standard deviation. The signal is labeled with μ at its variance. The signal is labeled with μ at its correlation coefficient. The signal is labeled with μ at its SNR. The signal is labeled with μ at its SNR. The signal is labeled with μ at its SNR.

2.1.

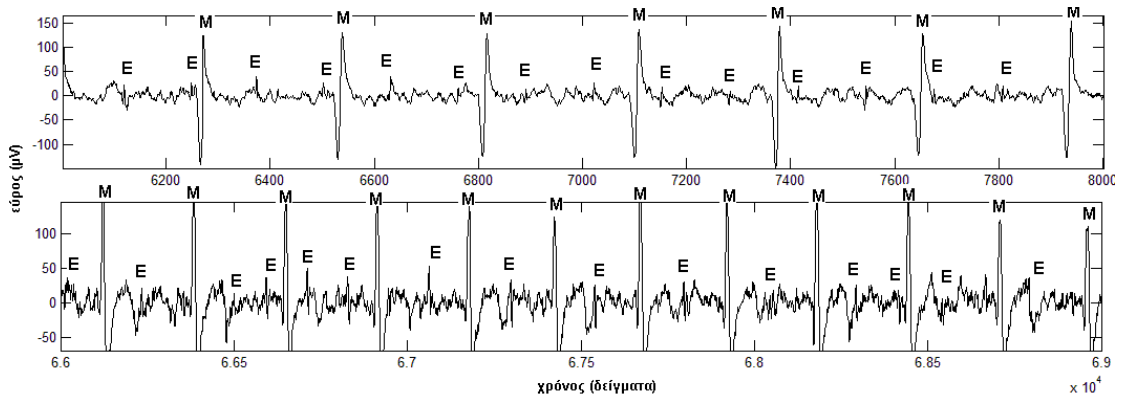
The figure shows a signal waveform with a QRS complex. The signal is plotted against time, and the QRS complex is labeled. The signal is noisy, and the QRS complex is the most prominent feature. The signal is labeled with μ at various points, indicating the mean value of the signal. The QRS complex is labeled with μ at its peak. The signal is labeled with μ at its trough. The signal is labeled with μ at its zero-crossing. The signal is labeled with μ at its maximum value. The signal is labeled with μ at its minimum value. The signal is labeled with μ at its average value. The signal is labeled with μ at its standard deviation. The signal is labeled with μ at its variance. The signal is labeled with μ at its correlation coefficient. The signal is labeled with μ at its SNR. The signal is labeled with μ at its SNR. The signal is labeled with μ at its SNR.

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μ μ μ μ
 () . 5 μ 2.1.



μ 2.1. μ : : μ , :

2.2. μ
 Η μ μ μ [5, 10],
 μ μ .

- , μ μ
- μ - μ μ μ
- μ μ μ (beat-to-beat).

- μ - μ μ
- μ , μ .
- μ μ .
- μ () ,
- μ , μ , μ .
- μ μ μ , μ .
- μ μ .
- μ , μ .

μ , μ μ 2
 μ μ :
 (μ 2.2).



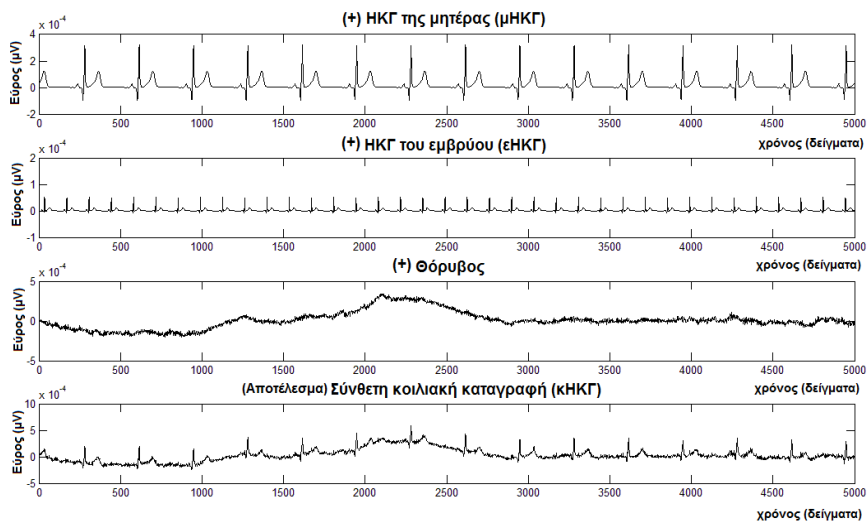
μ 2.2. μ 5 μ , 3 μ .

μ μ , μ μ .
 μ μ μ .
 μ , μ μ .
 μ μ ,

[12,13].

μ Myles *et al.* [14] Sato *et al.* [15]
 14 μ μ
 Martens *et al.* [16] 13 μ .
 μ ; Pieri *et al.* [5] μ μ μ
 3
 Assaleh *et al.* [17,18]
 μ μ . μ
 μ μ μ
 μ Assaleh *et al.* [17,18] Badee *et al.*
 [19], μ μ .
 μ Lathauwer *et al.* [20] Zarzoso *et al.* [21] μ
 μ μ (3).
 $\mu\mu$ μ
 μ μ ; , μ
 μ , μ μ μ
 (μ), . ,
 μ
 μ (. μ), μ μ μ ,
 μ μ μ (μ
 2.3). μ μ μ
 μ (μ μ), μ μ μ
 . μ
 μ μ μ
 , μ
 μ [22]. μ
 μ μ μ μ
 μ μ μ .
 μ μ μ μ , μ μ
 μ 2.3 3 ; μ ,

μ artifacts. μ μ .
 μ , μ μμ
 μ μ μ μ .
 μ
 μ

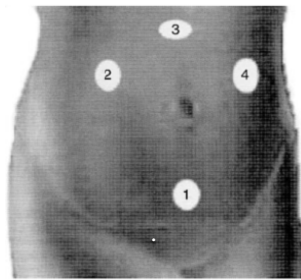


μ 2.3. : () μ , () , ()
 () .

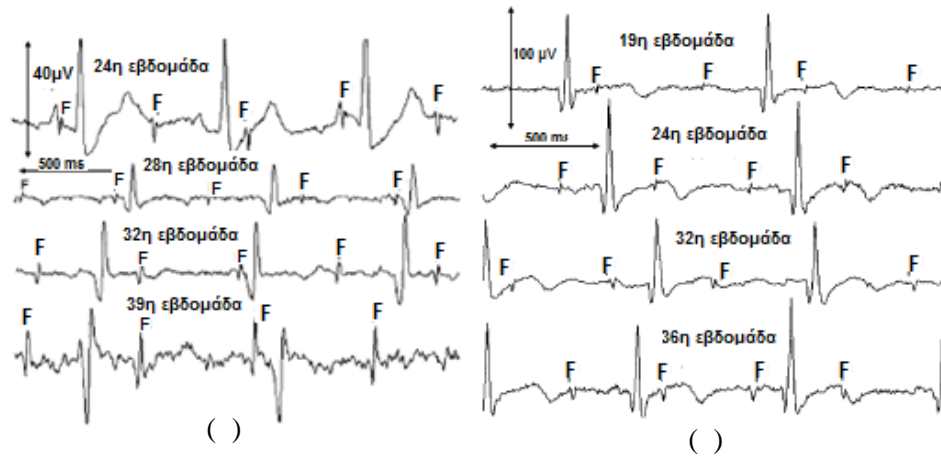
μ ,
 μ μ μ
 μ , μ μ [23].
 ()
 [24], μ μ μ
 μ μ [22].
 ,
 μ , μ
 μ [25]. , μ

2001 Pieri *et al.* [5] μ μ
 μ μ μ μ

μ 0.3 pV
 μ 1 pV
 μ 2.4)
 (μ 2.4)
 μ 2.4.
 μ 4 Hz 100 Hz,
 μ 12-bit. μ 300 Hz μ
 μ 7800.
 μ 20 41
 μ

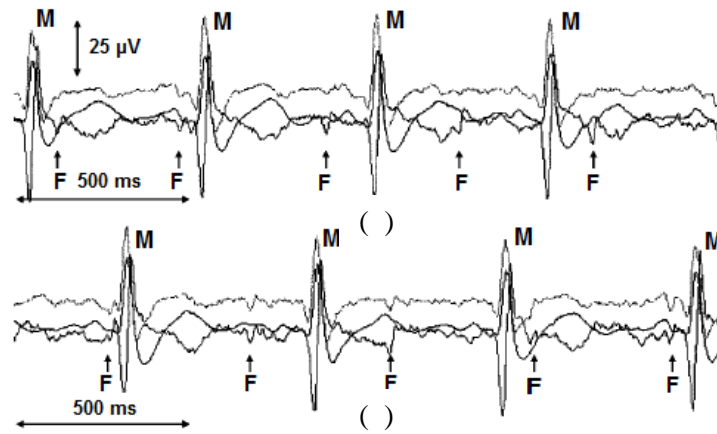


μ 2.4. μ
 μ 2.5 μ
 μ (Nottingham Database). μ 2.5() 2.5()
 2 μ
 19 μ 2.5() μ
 μ μ μ
 μ



μ 2.5. () μ 2

Nottingham Database μ μ μ
 4 , μ
 (μ 2.6).



μ 2.6. 3 μμ μ
 5 . () μ

() μ .

2.3.

μ : ,
 μ HK μ . μ
 μ μ 0.3-150 Hz. μ
 μ μ , μ μ
 μ μ 100 Hz. μ
 μ μ 10-400 Hz. μ μ

μ μ μ μ .
 μ μ , μ
 μ ,
 . μ μ
 μ μ μ .
 μ μ μ μ
 μ , μ μ
 μ μ [26]. μ
 “ μ ”. μ μ .
 μ 0.01-0.5 Hz [27]. μ μ
 μ μ μ .
 , μ μ μ
 μ (50 60 Hz) μ
 μμ ().

: μ
 μ μ (P, QRS, T . . .)
 μ . μ
 μ .
 μ μ μ μ
 (μ) μ μ
 . μ μ (μ
 μ) ,
 artifacts μ . μ
 μ

- μ :
- [28,29].
- μ 5-1000 μ
(μ) [30].
- :
- μ 25 μ , μ μ
μ 32 μ , [16,31,32].

• μ

μ [33].

μ (P, QRS T) μ

μ

μ μ μ μ

μ .

: μ μ

μ μ μ μ

. μ

μ μ μ . μ ,

Golbach *et al.* [25]

μ μ μ - μ - μ .

μ μ μ μ μ

μ , μ

μ

QRS μ P .

μ , ()

μ μ . μ

(9) , μ

() . μ

μ ,

μ RR- μ ,

μ P, μ μ QRS, μ T,

μ PR μ QT (μ 2.7).

μ

μ μ μ μ μ (QRS).

μ μ μ μ

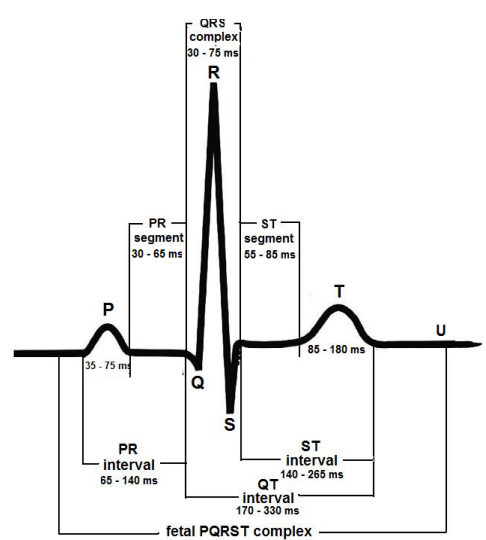
μ μ , μ μ 4 μ 5
 μ . μ 466
 48 765 μ .
 μ μ , P- μ

μ μ
 Sato *et al.* 2007 [15]. μ μ μ 35
 μ μ

μ μ . ,
 μ 2 μ , μ
 μ μ ().

: μ
 . μ
 μ μ . μ 2.8() 2.8()
 μ ,
 [15] [25], .

μ 2.7 μ μ μ
 μ μ μ



μ 2.7. μ .

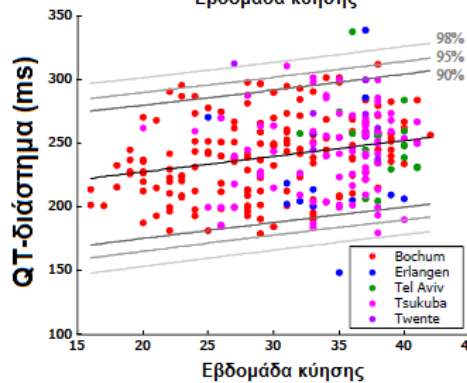
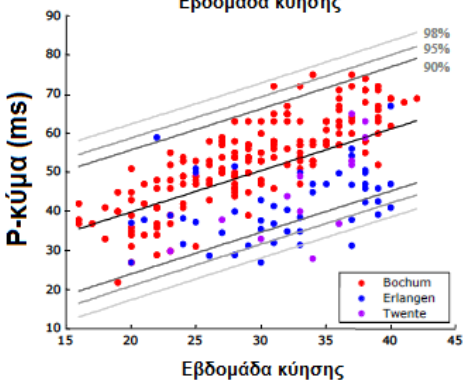
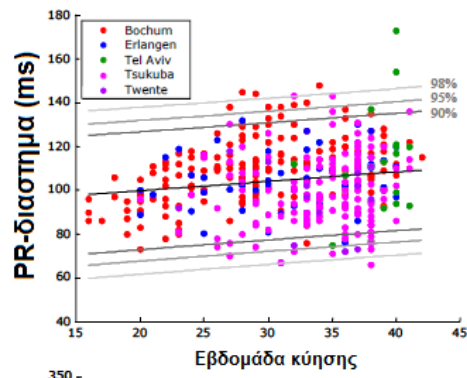
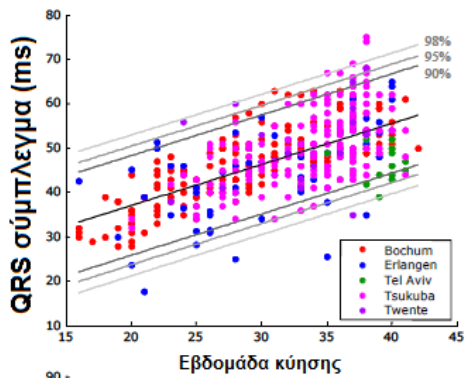
μ μ :
 μ μ μ μ , μ μ
 μ 4 5 μ . μ
QRS , μ μ μ (N =
1386) μ μ .
 μ 2.9 $\mu\mu$ μ μ μ
QRS μ μ μ .

 μ μ P- μ QRS μ
 μ , μ T .
 μ P μ , μ μ

 μ P, QRS, PR QT

 μ μ μ μ

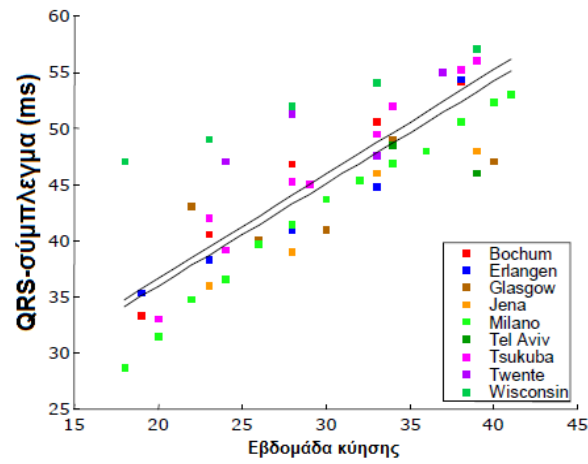
 μ .



μ 2.8. $\mu\mu$
 μ μ

μ μ
[15] [25].

μ , μ
 μ , μ
 μ μ



μ 2.9. μμ μ QRS μ μ
 μ μ [23].

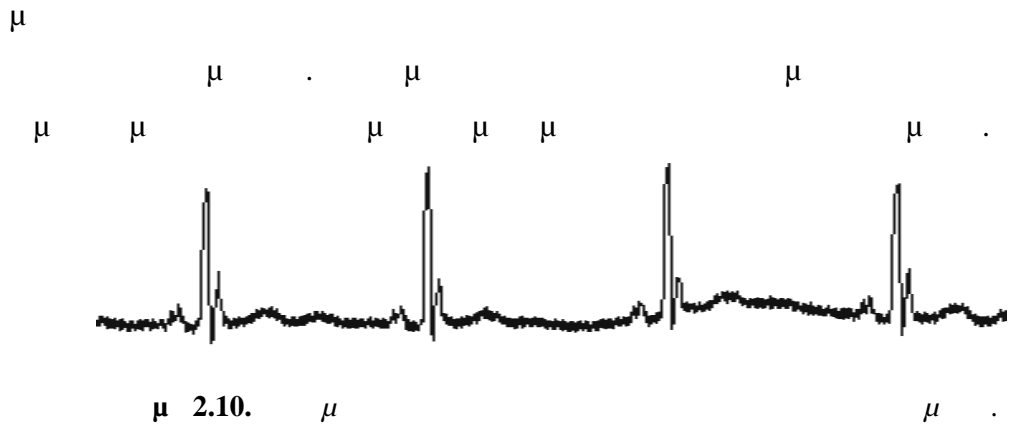
μ μ , μ μ
 μ μ , μ
 μ μ [34].

μ
 , μ μ μ μ
 μ , μ μ μ μ 0.5 sec,
 120 μ (beats/m).
 μ μ μ QRS 0,42 sec. μ
 , ' : 60/0,42 beats/m. 142
 μ

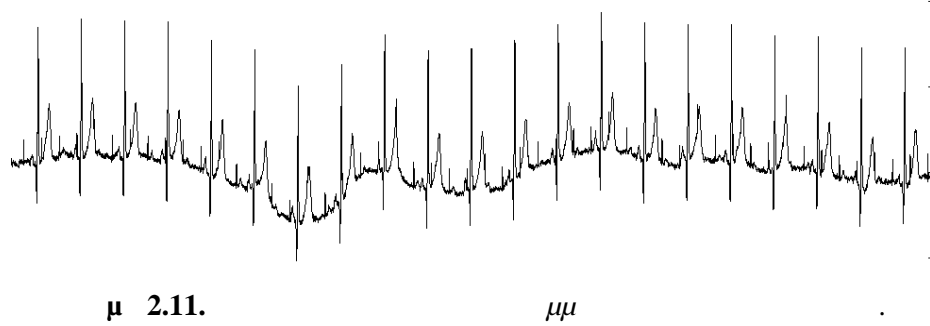
2.4.

μ μ μ μ
 μ μ μ
 μ , μ , , μ
 μ μ μ [35]. μ
 μ :

• μ $50\ 60\ \text{Hz}$ (/).
 μ $(\ \mu\ 2.10).$ μ
 μ μ μ μ
 μ μ μ μ μ μ [36].
 μ $50\ (\ 60)\ \text{Hz},$

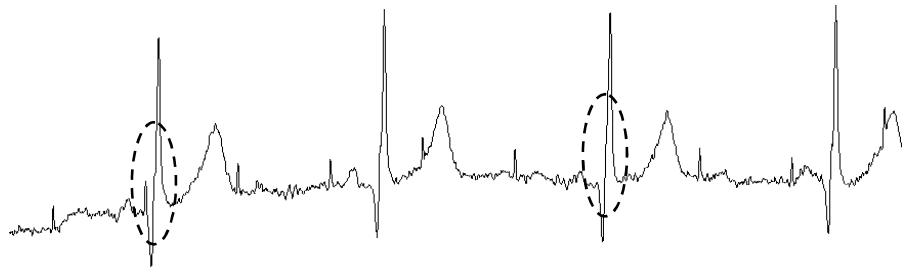


• μ μ (μ 2.11),
 $\mu\mu$ (). μ μ .
 / , μ μ .
 μ . μ , μ μ
 μ μ . μ , μ μ μ
 , μ μ μ μ μ
 [37].



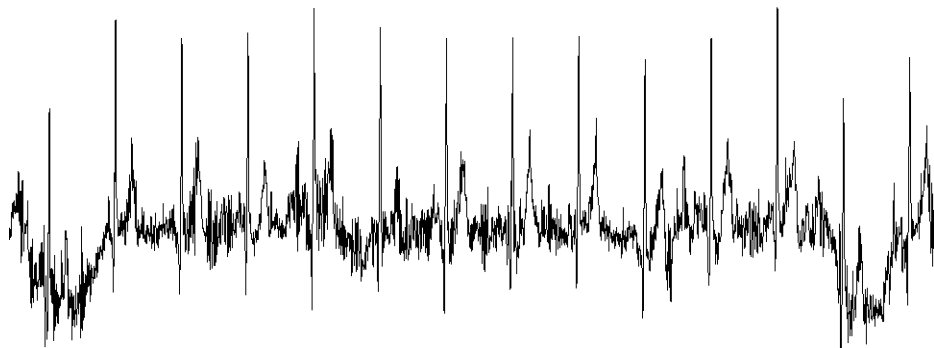
μ μ ($1\ \text{Hz}$), μ μ μ μ

μ .
 μ :
 • μ
 μ (μ 2.13). , QRS μ μ μ
 (μ QRSs) μ (overlap) QRS
 μ μ μ (QRSs) μ
 μ [38]. μ μ μ ,
 μ QRS .



μ 2.13. μ QRSs μ .

• μ ,
 μ
 (μ 2.14). ,
 μ , μ ,
 μ μ , [39].

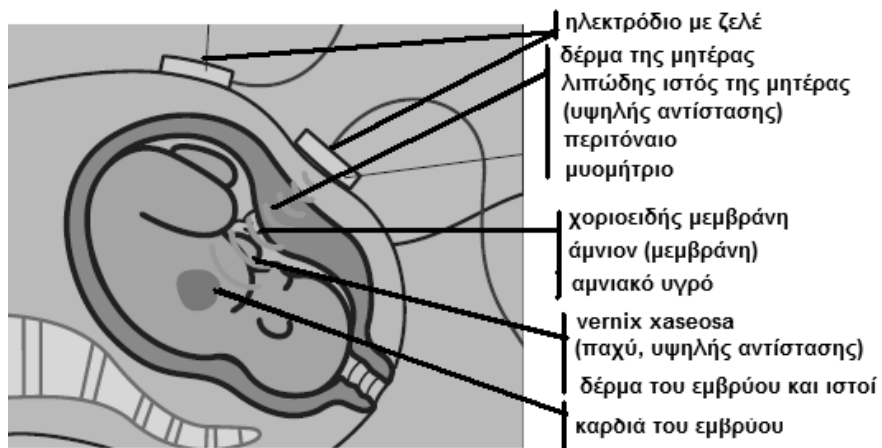


μ 2.14. μ μ .

μ μ μ ,
 μ μ μ . μ

μ μ μ
 μ μ μ
 μ μ μ
 μ μ μ (uterine electromyogram-EMG) μ
 (electrohysterogram-EHG), μ μ μ [40].
 μ μ μ
 μ μ μ
 μ μ μ
 μ μ μ
 SNR

• μ μ
 μ μ μ
 μ μ μ
 (μ 2.15) [41].

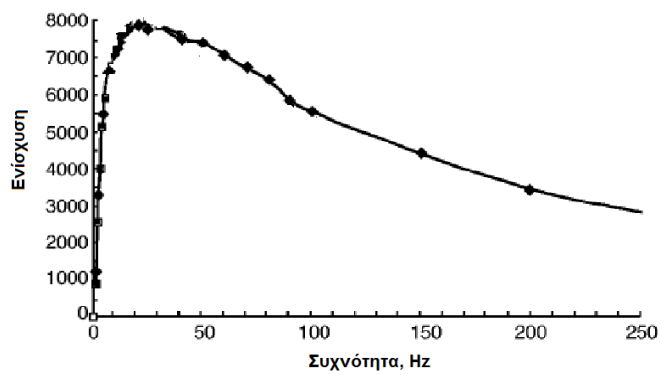


μ 2.15.

2.1 μ

, μ μ μ . 2 μ
(μ μ) μ μ μ μ
 μ μ . μ μ μ
. μ μ
 μ μ μ μ μ
 μ . (μ) μ
, μ .
 μ μ (/): μ μ
[42] μ
 μ ,
50 60 Hz. μ [43]
, μ μ μ
 μ μ μ ,
 μ . μ
 μ μ μ [44],
 μ μ μ
 μ ,
[45], μ . , μ
 μ μ , $\mu\mu$.
 μ μ μ μ
 μ .
 $\mu\mu$ (): μ ()
 $\mu\mu$ μ μ ,
 μ μ μ .
 $\mu\mu$ μ Daskalov [46]
 μ . μ μ
 μ
 μ [47], μ μ ,
. μ , μ μ [48]
 μ [49,50]

μ
 Hz, μ ST, μ μ
 μ μ
 : μ
 μ μ
 μ
 μ
 μ
 μ μ μ
 μ
 μ
 μ 4-100Hz,
 μ 2.17.



μ 2.17.

Al-Zaden *et al.* [51] μ μ
 μ SVD - μ μ μ
 (neuro-fuzzy inference systems-ANFIS). SVD μ
 μ μ μ

- :
1. μ (1).
 2. R- μ μ .
 3. μ SVD μ .
 4. μ ANFIS.

5. μ μ ANFIS

6. μ μ 1-4

μ μ , μ μ μ μ

μ μ : $x=[x(0)x(1)...x(N-1)]$, μ

μ μ μ μ μ R ,

$R_0, R_1, R_2, \dots, R_m$. $n \times m$ μ

:

$$A = \begin{bmatrix} x(R_0 + n_R) & \dots & x(R_1 - 1) & x(R_1) & x(R_1 + 1) & \dots & x(R_1 + n_R) \\ x(R_1 + n_R) & \dots & x(R_2 - 1) & x(R_2) & x(R_2 + 1) & \dots & x(R_2 + n_R) \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x(R_{m-1} + n_R) & \dots & x(R_m - 1) & x(R_m) & x(R_m + n_R) & \dots & x(R_m + n_R) \end{bmatrix} \quad (1)$$

n_R μ μ μ μ

μ μ μ

μ μ μ $[x(R_{i-1} + n_R), \dots, x(R_i + n_R)]$

μ - - (R-R μ)

μ μ μ μ

μ $\mu\mu$ $[x(R_{i-1} + n_R), \dots, x(R_i - 10)]$ ($i = 1, \dots, m$)

μ μ μ $n_R - 10$ μ .

μ SVD $B=AT$, μ $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r \geq 0$ μ r

$= \min(n, m)$ $\sigma_{r+1} = \sigma_{r+2} = \dots = \sigma_m = 0$.

[52,53], μ

$$x_I = u_I \cdot v_I^T$$

μ , μ x_I μ .

,

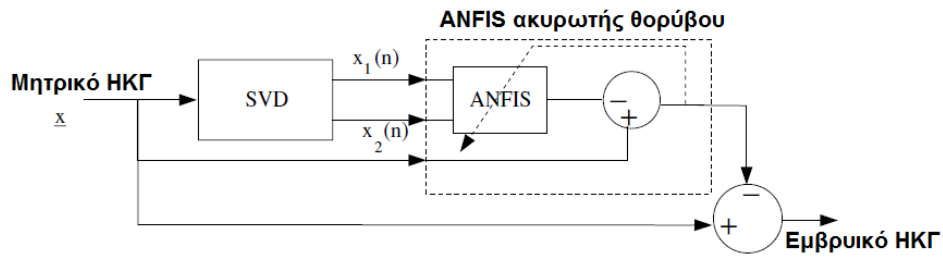
μ μ μ .

μ μ μ μ .

μ , μ $x_2 = u_I \cdot v_I^T +$

$u_2 \cdot v_2^T$. μ x_I x_2 μ μ

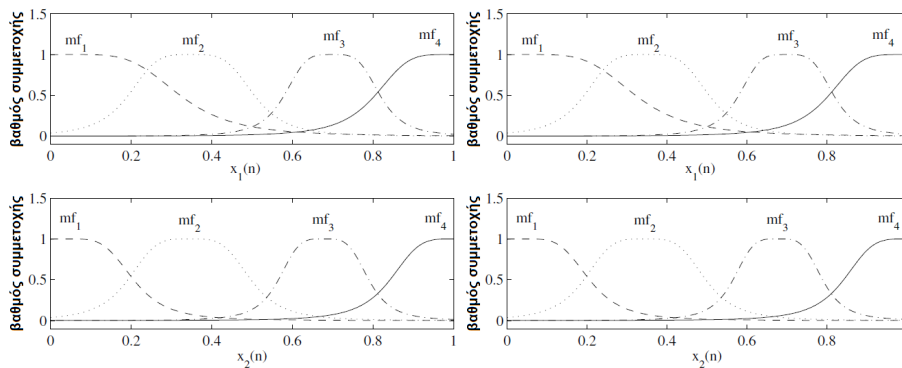
μ ANFIS μ 2.18.



μ 2.18. μμ μ .

ANFIS μ μ μ
 . n x1 x2,
 μ μ μ μ
 μ , xM,
 μ x , μ μ
 μ ; μ $\tilde{x}_M(n) = f(x_1(n), x_2(n))$.

ANFIS μ μ
 μ $\tilde{x}_M(n)$ $e^2(n)$ μ $x_M(n)$
 $\tilde{x}_M(n)$;
 ANFIS $\{e^2(n)\}$,
 μ . μ ANFIS μ
 μ 2.19.



μ 2.19. μ ANFIS.

μ μ μ
 μ (e).
 μ , , e μ

(6).

Ibrahimy *et al.* [54] 2

10 40 Hz)

Hamming.

QRS

(30 40 Hz)

QRSs.

2.6.

QRS artifacts

[55].

(μ SNR).

1

(fiducial points),

QRSs (μQRSs)

4-20 Hz [26].

FIR

P

μQRSs.

2 μQRSs

(μ μ e) 2

QRSs

4-80 Hz [4,5,26],

FIR

e

μ artifacts μ 4-100 Hz. μ QRSs μ
 μ - μ μ
 (Multiscale Principal Component Analysis-MSPCA) [56-58], μ μ
 μ μ (multivariate denoising) [57]. MSPCA
 μ PCA μ μ
 μ $\mu\mu$, μ wavelets μ μ
 μ μ μ μ μ .
 MSPCA PCA wavelet μ μ μ
 μ μ μ μ μ .
 μ - μ μ , MSPCA
 μ μ - μ ,
 [56].

multivariate denoising [57] μ μ
 MSPCA

- μ μ :
1. wavelet μ μ μ L
 () e : $\{d_k^j, a_L^j\} = WT(x_j)$, $j=1, \dots, M$, $k=1, \dots, L$, x_j
 j^{th} () e μ μ
 μ , d_k^j detail j k
 a_L^j approximation j .
 2. μ robust PCA (RPCA) [58] D_1 : μ
 $\hat{\Sigma}_\varepsilon$, μ μ
 $\hat{\Sigma}_\varepsilon = mcd(D_1)$, mcd μ μ μ
 [58] D_1 details 1
 ($D_1 = [d_1^j]$, $j=1, \dots, M$). μ V $\hat{\Sigma}_\varepsilon = V\Lambda V^T$,
 $\Lambda = diag(\lambda_j)$, $j=1, \dots, M$. μ

wavelet ‘ μ ’ (wavelet details) μ : $D_k V$,
 $k=1,\dots,L$, $D_k = [d_k^j]$, $j=1,\dots,M$. μ

$$\mu \sqrt{2\lambda_j \ln(N)}, \quad j=1,\dots,M, \quad (N \mu \mu)$$

$$\mu \quad M \quad D_k V . \quad \mu \quad \mu$$

μ wavelet ‘ μ ’ \hat{D}_k , $k=1,\dots,L$.

3. PCA wavelet ‘ μ ’ (wavelet approximations) A_L ($A_L = [a_L^j]$, $j=1,\dots,M$, A_L $N \times M$)
 μ μ ,

$$\mu \quad \text{Kaiser [57]}, \quad \hat{A}_L .$$

Kaiser μ μ

$$\mu \quad \mu .$$

4. μ ‘ μ ’ e ,

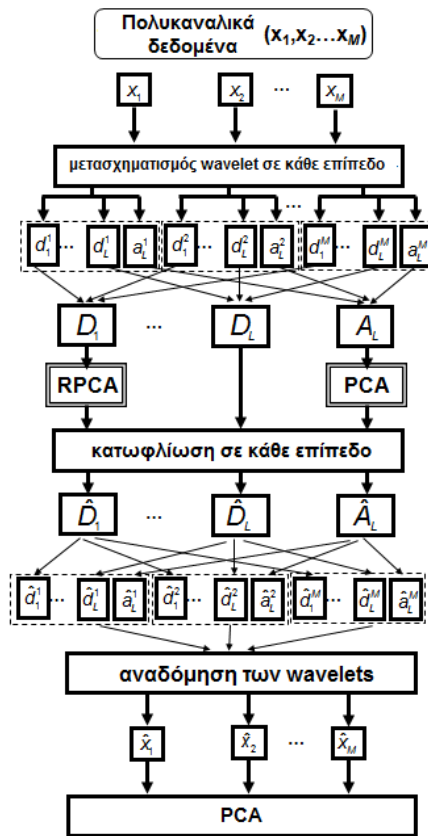
wavelet ‘ μ ’ (\hat{D}_k , $k=1,\dots,L$) wavelet ‘ μ ’
 (\hat{A}_L) , μ μ V^T wavelet

$$\mu \quad \mu \quad \mu .$$

5. μ PCA e μ
 μ , μ Kaiser.

μ μ μ μ μ
 μ μ MSPCA , μ μ
 μ [57]. multivariate denoising

μ 2.20.



μ 2.20. μμ multivariate denoising .

μ μ , μ multivariate denoising :

1 μ , wavelet Daubechies 4

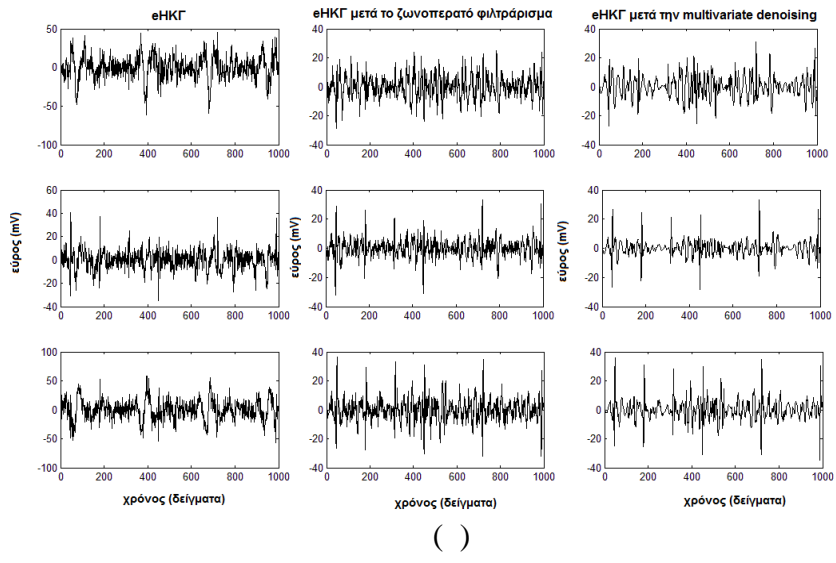
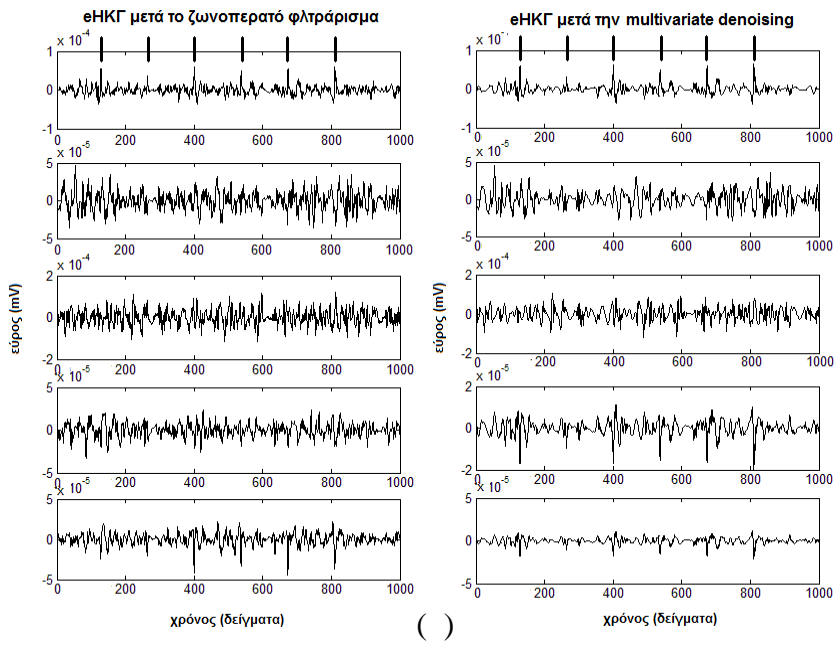
μ 6 (L = 6). μ

μ μ μ D₁, μ

μ 75% μ μ .

μ 3 (wavelet ‘ ’) 5 (PCA μ

μ wavelet) μ Kaiser.



μ 2.21. () 5 μ μ μ (μ SNR = $-5dB$) μ μ μ μ
multivariate denoising . () μ μ μ
;a to e μ , μ μ
 μ μ *multivariate denoising* .

μ μ μ μ μ μ μ μ
 (μ R-) [57]. μ 2.21
 μ μ μ μ
 () 5 μ μ () 3
 μ e μ . $\mu\mu$

The cardiac cycle consists of a series of electrical events that produce the ECG trace. The ECG trace is a recording of the electrical activity of the heart over time. The ECG trace is divided into several segments: the P wave, the QRS complex, and the T wave. The P wave represents the atrial depolarization, the QRS complex represents the ventricular depolarization, and the T wave represents the ventricular repolarization. The ECG trace is used to diagnose various heart conditions, such as arrhythmias, heart block, and myocardial infarction.

2.7. The ECG trace is a recording of the electrical activity of the heart over time. The ECG trace is divided into several segments: the P wave, the QRS complex, and the T wave. The P wave represents the atrial depolarization, the QRS complex represents the ventricular depolarization, and the T wave represents the ventricular repolarization. The ECG trace is used to diagnose various heart conditions, such as arrhythmias, heart block, and myocardial infarction.

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μ , μ μ μ QRS μ .
 μ μ μ μ QRS μ
 μ (ventricular arrhythmias),
 μ (cardiac rhythm) μ , μ
 μ .
 μ μ μ μ μ μ QRS, μ
 μ μ μ (μ Q), μ
 μ His (μ μ QRS
), (μ
 μ μ QRS) μ
 .
 μ μ μ μ ST μ μ
 μ μ (μ μ ST,
 μ , ,
 μ μ QT), μ μ ,
 (ST ,
 aVR, μ μ μ μ),
 μ μ (μ μ ST μ μ
 μ ST-T) , , μ
 (μ μ
 ST μ μ
 μ). μ
 μ ST-T μ
 μ .

, μ μ μ
 . μ
 μ μ μ μ , μ μ μ
 μ , μ μ μ
 ST μ , μ μ
 μ μ , μ μ μ
 μ .

, μ , μ μ μ
 μ μ .

2.8.

μ μ μ
 μ . , μ μ μ
 μ . μ
 μ μ QRS
 μ R-R. μ , μ
 μ QRS (μ R μ S),
 μ μ μ . μ
 QRS μ μ
 μ μ μ (μ μ μ). μ μ
 , QRS μ μ μ
 . μ
 μ μ μ μ
 μ μ μ μ
 μ . QRS μ
 μ μ . μ μ QRS
 . , μ
 QRS μ , μ μ
 μ μ μ μ
 μ μ μ .

2.9.

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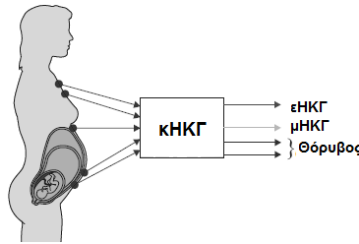
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(Blind Source Separation problem - BSS) (3.1).
 [1]. BSS
 Value Decomposition-SVD) [2-4] (Singular Value Decomposition - SVD) (PCA) [2,5]
 Component Analysis-ICA) [2]. (Independent Component Analysis - ICA)



3.1. BSS [1].

ICA wavelets [6]. BSS
 PCA [7], (temporal structure) [8], [3], [9], [5, 10], ο (fractals) [11], [12] [13].
 ANFIS [14] wavelet [23].
 [15], BSS [16], [17] [18].

3.2. Blind Source Separation (BSS)

Consider a mixture of m independent sources $s_1(t), s_2(t), \dots, s_m(t)$ (1) recorded by n sensors. The observed signal $x(t)$ is a linear combination of the sources:

$$x_i(t) = a_{i1} \cdot s_1(t) + a_{i2} \cdot s_2(t) + \dots + a_{in} \cdot s_n(t) \quad i = 1, 2, \dots, m$$

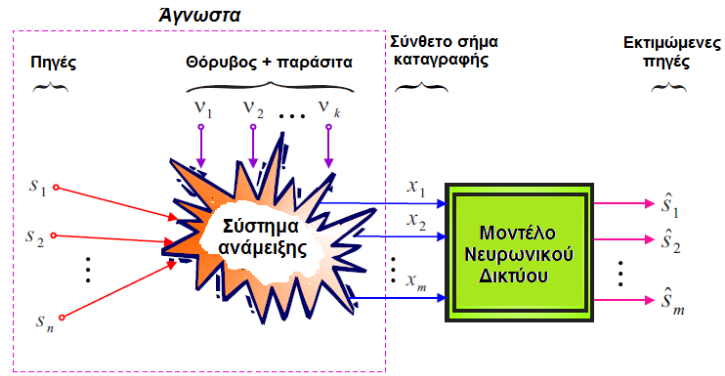
where $a_{i1}, a_{i2}, \dots, a_{in}$ are the mixing coefficients. The source vector $s(t) = [s_1(t), s_2(t), \dots, s_n(t)]^T$ and the observation vector $x(t) = [x_1(t), x_2(t), \dots, x_m(t)]^T$ are related by the mixing matrix A :

$$x(t) = A s(t)$$

The matrix A is of size $m \times n$. The i -th row of A is $a_i = [a_{i1}, a_{i2}, \dots, a_{in}]^T$. The goal of BSS is to estimate the source signals $s_j(t)$ from the observations $x(t)$.

The BSS problem is to find the source signals $s_j(t)$ from the observations $x(t)$. This is achieved by finding the inverse of the mixing matrix A , which is possible if A is invertible. The source signals can then be estimated as $s(t) = A^{-1} x(t)$.

(tracking capability)

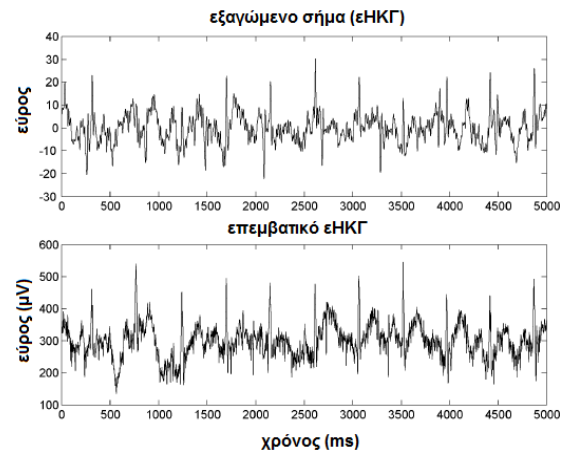


μ 3.2. μμ BSS μ .

μ , μ
 μ μ μ
 μ (. . μ μ μ
 μ μ μ) μ μ
 μ , μ μ
 μ μ μ
 . , μ
 μ μμ . μ
 μ μ . μ μ μ
 μ μ ,
 μ μ : μ μ
 (Independent Components Analysis-ICA), μ μ (Blind
 Source Separation-BSS), μ (Blind Signal Extraction-BSE)
 (Multichannel Blind Deconvolution-MBD) [19].

, μ μ μ
 μ μ μ
 μ
 “μ ” μ :
 μ μ μ /
 μ . μ
 . μ ,

[22]. , μ , , μ , μ (μ μ). , μ μ μ (μ μ μ) , μ μ μ (μ μ μ) . μ μ μ μ BSS μ μ μ μ ; μ () μ μ μ . , [23] μ , μ μ μ μ μ μ (μ 3.3).



μ 3.3. μ μ μ μ ICA () μ (μ). μ μ .

3.4. -

- i. $m \times n$ matrix X (SVD): $X = U \Sigma V^T$, SVD factorization
- U ($U^T U = I_{n \times n}$) ($m \times n$), left singular vectors
 - Σ ($m \times n$), singular values
 - V ($V^T V = I_{n \times n}$) ($n \times n$), right singular vectors

$$X = U \Sigma V^T$$

$[m \times n] = U_{[m \times n]} L_{[n \times n]} (V_{[n \times n]})^T$

(mode amplitudes) Σ V^T (right singular vectors).

$$\hat{X} = U \Sigma U^T X$$

- ii. (PCA): PCA
- (target data) X

$$C = \frac{1}{n} X^T X$$

$$|C - \lambda_i I| = 0$$

$i \in [1, 2, \dots, n]$, n

$$(C - \lambda_i I)e_i = 0.$$

$$P = T \cdot D, \quad D$$

$$T = \begin{bmatrix} e_{11} & \dots & e_{1n} \\ \vdots & \ddots & \vdots \\ e_{m1} & \dots & e_{mn} \end{bmatrix}$$

PCA

$$\% \text{ Variance} = \lambda_i \cdot 100 / \sum_{k=1}^n \lambda_k.$$

3.5.

3.5.1.

- ICA

ICA BSS μ . μ , μ , μ . μ μ μ μ μ μ μ μ μ . μ μ μ μ μ μ μ μ μ .

3.5.2. μ

ICA μ μ $s(t)$ $n \times m$ W μ

$y(t) = \hat{s}(t) = Wx(t).$ W μ μ

A μ $\hat{s}(t)$ μ μ

$s(t).$ $\hat{s}(t)$

μ , μ μ information-theoretic
Kullback-Leibler

μ , μ $\mu\mu$ μ . ,

μ w_{ij} W $y(t) = \hat{s}(t) = Wx(t)$

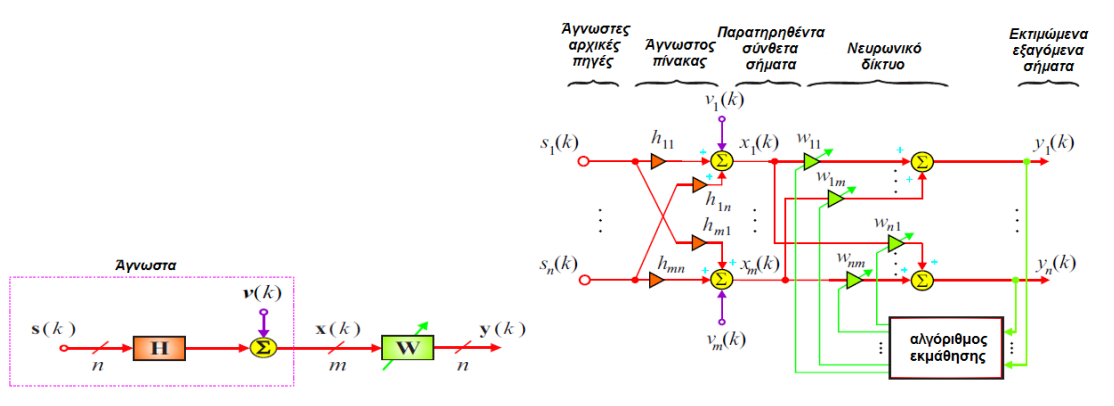
(μ feed-forward) μ

$x_i(t)$ μ μ

$y(t) = \hat{s}_j(t) = \sum_{i=1}^m w_{ji}x_i(t), j = 1, 2, \dots, n.$

μ μ $\hat{s}(t).$

μ 3.4.



μ 3.4. () μ $ICA.$ μ , $x(t)$

$\mu\mu$ μ $s(t).$ ICA

$W,$ μ μ $s(t).$

() μ .

3.5.3. μ

μ μ μ $x(t) = As(t)$ ICA

μ . μ μ μ

μ . ,
 μ μ . μ μ , μ μ
 μ μ μ μ μ μ
 μ μ μ [28].

μ : μμ $x(t)=As(t)$,
 μ . μ μ

μ : $x(t) = As(t) + n(t)$.

ICA μ $x(t)=As(t)$, μ μ
 μ . μ

μ : $m \times n$ μ
 (m=n), , μ
 μ ICA, ,

μ . μ μ n
 μ μ μ m. -
 μ μ 8
 , (n<4).

$n < m$ μ
 PCA, μ μ μ $x(t)$
 A μ ICA [25].

μ ICA μ μ n .
 μ $\hat{s}(t)$ μ .

3.5.4. μ
 - μ μ μ ICA
 μ
 μ $s(t)$ $x(t)=As(t)$.

(gaussian) μ μ μ 2
 μ , μ μ μ μ
 μ . μ
 μ μ μ μ μ μ
 μ μ μ μ

- μ μ μ μ μ μ ()
 μ . A $s(t)$ $x(t)=As(t)$
 μ , μ μ $x(t) = \sum_{i=1}^n a_i s_i(t)$, μ μ μ :

$$x(t) = \sum_{i=1}^n \left(\frac{1}{a_i} \right) (a_i s_i(t)) .$$

μ , μ

μ μ . μ $s_i(t)$ μ
 μ μ .

- μ , μ μ μ μ
 $s_i(t)$ μ s_i
 μ : $E\{s^2\}=1$. , μ μ
 μ A μ μ μ
 s_j : $p_j = \sqrt{\frac{1}{m} \sum_{i=1}^n (a_i^j)^2}$, a_i^j : i
 j p_j : RMS s_j ($1 <$
 $j < n$).

- μ μ μ μ .
 μ
 $x(t) = \sum_{i=1}^n a_i s_i(t)$, μ μ μ
 $s(t)$ μ μ ICA. μ
 μ μ P P^{-1}

$$x(t) = As(t) = APP^{-1}s(t) \quad AP$$

3.5.5.

$$\hat{s}(t) = Wx(t) \quad \mu$$

$$x(t) = W^{-1}\hat{s}(t) \quad \mu$$

$$x(t) = \sum_{i=1}^n a_i s_i(t) \quad \mu$$

$$: x(t) = \sum_{i=1}^n w_i^{-1} \hat{s}(t), \quad w^{-1} \quad \mu \quad -$$

$$W^{-1} \quad \mu \quad \mu \quad w_i^{-1} \hat{s}(t)$$

$$\mu \quad \mu \quad , \quad \mu \quad \mu \quad \mu$$

$$, \quad \mu \quad \mu$$

μ

$$\mu \quad , \quad \mu \quad -$$

$$\mu \quad \mu \quad \mu \quad \mu \quad \mu$$

, μ μ ICA,

$$\mu \quad . \quad \mu \quad \mu \quad \mu$$

$$x(t) = \sum_{i=1}^n a_i s_i(t) \quad \mu \quad \mu$$

$$\mu \quad \mu \quad x(t) \quad \mu \quad \mu \quad \mu \quad s(t)$$

$$\mu \quad \mu \quad \mu \quad \mu \quad \mu \quad \mu$$

$$\mu \quad \mu \quad x(t) \quad \mu \quad \mu \quad \mu \quad s_i(t)$$

$$x(t) = \sum_{i=1}^n a_i s_i(t) \cdot$$

3.5.6.

ICA

i

y_j $i=j$

(y_1, y_2, \dots, y_n)

$p(y_1, y_2, \dots, y_n)$

$p_i(y_i)$

$$p(y_1, y_2, \dots, y_n) = p_1(y_1)p_2(y_2) \dots p_n(y_n) = \prod_{i=1}^n p_i(y_i)$$

4.5.4,

$[s_1, s_2, \dots, s_m]$

$$x = \sum_{i=1}^m a_i s_i$$

$$\sigma^2 = \sum_{i=1}^m \sigma_i^2$$

m ($m = 10$),

s_i

ICA

m

μ
 μ μ μ μ - .

μ *Fast ICA* (Hyvarinen-Oja, 1997).

μ , μ : μ
 μ μ - μ ,

μ
 [25,29]. μ μ μ μ μ μ :

kurt(x) = E{x⁴} - 3(E{x²})².

μ 3(E{x²})² μ

μ , μ

μ
 μ μ .

μ *Infomax ICA* (Bell-Sejnowski, 1995).

μ μ :

μ μ - μ μ μ μ :

μ μ μ

μ , μ μ μ .

μ μ , μ

μ μ x μ p_x :

$$H(x) = \int_{-\infty}^{+\infty} p_x \ln p_x(x) dx .$$

μ x. μ x

μ μ μ , " "

μ , μ

μ 0 1, μ x

μ μ .

μ μ x μ μ μ μ

μ ² μ x

μ μ

μ μ - : J(x) = H(X_N) - H(x), μ .

μ μ
 μ [30].

μ **JADE** (Hyvarinen, 2001). μ μ μ
 : μ JADE
 (cumulant tensors) .
 μ μ (covariance matrix)
 , μ
 (covariance).

. μ μ
 , $C_x = \langle xx^T \rangle$, μ , ,
 μ - μ μ , μ , μ
 , μ . μ
 μ μ μ
 . μ μ μ μ -
 gaussian μ .
 , μ JADE,
 μ , μ μ μ μ
 , μ μ μ ,
 μ [31].

ii.

μ μ μ μ
 μ ICA, μ s_i
 μ μ .
 μ μ μ -
 μ μ .
 μ μ μ μ
 μ $x(t)$ μ μ
 μ μ μ μ
 : $c_{xx}^T \stackrel{def}{=} E\{x(t)x(t+\tau)\}$,

$$= 0, 1, 2, 3, \dots, \mu \quad \mu \quad \mu \quad \mu \quad \mu \quad \mu \quad N,$$

$$c_{xx}^T \stackrel{\text{def}}{=} \frac{1}{N} \sum_{t=0}^{N-1} x(t)x(t+\tau), \quad = 0, 1, 2, 3, \dots, \mu \quad \mu \quad \mu$$

$$c_{x_1 x_2}^T \cdot \quad \mu \quad x_1(t) \quad x_2(t) \quad n$$

$$x = [x_1, x_2, \dots, x_n], \quad \mu \quad n \times n \quad \mu$$

$$c_x^T \stackrel{\text{def}}{=} E \{ x(t)x(t+\tau)^T \}, \quad c_{ij} = c_{x_i x_j}^T \cdot \quad i > 1, j < n.$$

$$c_{x_i x_j}^T = 0, \forall i \neq j \quad \mu \quad \mu \quad \mu \quad \mu \quad \mu$$

$$c_x^T = \text{diag} [c_{x_1 x_1}^T, c_{x_2 x_2}^T, \dots, c_{x_n x_n}^T], \quad = 0, 1, 2, 3, \dots, \mu \mu \quad \mu$$

$$\hat{s}(t) = Wx(t), \quad \mu :$$

$$C_s^0 = E \left\{ \hat{s} \hat{s}^T \right\} = E \{ Wx(Wx)^T \} = E \{ Wxx^T W^T \} = WE \{ xx^T \} W^T = WC_x^0 W^T.$$

$$\mu \quad s_i(t) \quad \mu \quad \mu \quad \mu \quad \mu \quad C_s^0 \quad \mu$$

$$\mu \quad , \quad W \quad \mu$$

$$\mu \quad C_s^0 \quad \mu \quad , \quad \mu \quad \mu$$

$$\mu \quad \mu \quad A^{-1} \quad \mu \quad x(t) = As(t).$$

$$\mu \quad \text{ICA} \quad \mu \quad s$$

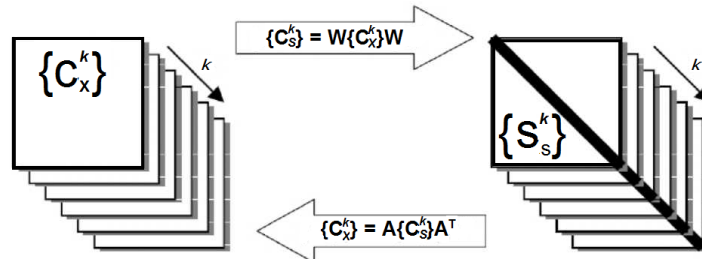
$$\mu \quad \mu \quad \mu \quad \mu \quad W$$

$$\mu \quad \mu \quad \mu \quad \mu \quad \mu$$

$$C_s^0 = WC_x^0 W^T \quad \mu \quad \mu$$

$$(c_x^T = \text{diag}[c_{x_1x_1}^T, c_{x_2x_2}^T, \dots, c_{x_nx_n}^T]), \quad W$$

$$C_s^0 = WC_x^0W^T, \quad = 0, 1, 2, 3, \dots, k .$$



μ 3.5. μ x s
 μ μ μ μ μ μ
 $k+1$ k μ
 [34]).

μ ICA μ μ μ μ μ μ
 (joint diagonalization) (μ 3.5).
 μ
 μ μ μ μ ,
 ICA μ μ μ [32,33].
 μ $\hat{s}(t) = Wx(t)$ μ
 μ , μ

3.5.7.

ICA, μ μ μ μ :
 (centering) (whitening),
 : μ μ
 μ x μ μ μ
 $\hat{x} = x - E\{x\}$ μ μ μ μ

$$\hat{s} = W \hat{x} \quad : \quad E\{\hat{s}\} = E\{W \hat{x}\} = WE\{x\} = 0.$$

$$\begin{aligned} & \text{ICA} \\ & \mu \\ & \mu \\ & \mu \end{aligned} \quad W \quad \mu \quad \mu \quad \mu$$

$$s = \hat{s} + WE\{x\}$$

$$\hat{s} = W \hat{x} = W(x - E\{x\}) = s - WE\{x\}.$$

x is an $n \times 1$ vector, C_x is an $n \times n$ matrix, V is an $n \times n$ matrix, D is an $n \times n$ matrix, P is an $n \times n$ matrix.

$$C_x = \begin{pmatrix} c_{x_1 x_1} & c_{x_1 x_2} & \dots & c_{x_1 x_n} \\ c_{x_2 x_1} & c_{x_2 x_2} & \dots & c_{x_2 x_n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{x_n x_1} & c_{x_n x_2} & \dots & c_{x_n x_n} \end{pmatrix} = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{pmatrix} = I$$

$$z = Vx$$

$$V = PD^{-1/2}P^T, \quad P$$

$$D$$

$$C_x^0 = PDP^T, \quad P, D, \mu$$

$$C_x^0 = PDP^T, \quad \mu, \mu, \mu, \mu, V$$

$$x, \mu, \mu, \mu, \mu$$

$$\begin{aligned} C_z^0 &= E\{Vx(Vx)^T\} = VE\{xx^T\}V^T = VC_x^0V^T = (PD^{-1/2}P^T)(PDP^T)(PD^{-1/2}P^T)^T = \\ & PD^{-1/2}(P^T P)D(P^T P)D^{-1/2}P^T = P(D^{-1/2}DD^{-1/2})P^T = PIP^T = PP^T = I \end{aligned}$$

ICA $x = As, \quad \mu : z = Vx = VAs = As.$

$$\hat{A} = VA$$

$$C_z^0 = E\{VA_s(VA_s)^T\} = (VA)C_z^0(VA)^T = (VA)(VA)^T = I$$

3.6. BSS

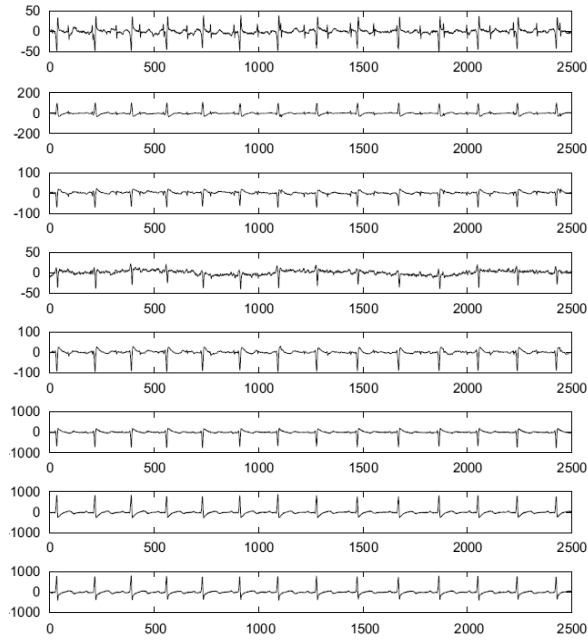
Lathauwer *et al.* [2] BSS

$$X = AS + N \quad (1)$$

$X \in R^I$, $S \in R^J$, $N \in R^I$, $A \in R^{I \times J}$

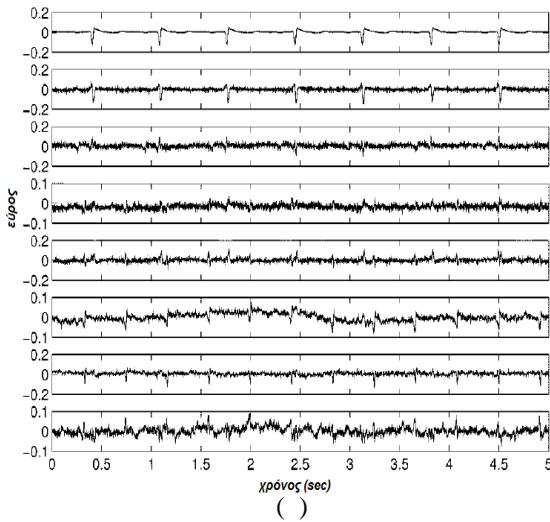
ICA

(maximum-likelihood) [35] ([36] [37])

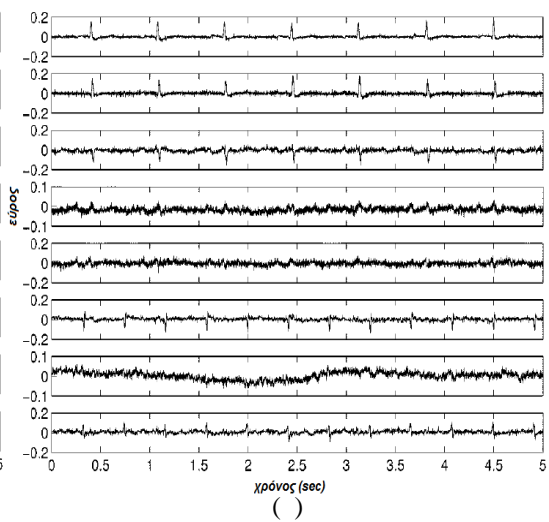


μ 3.6. δ-

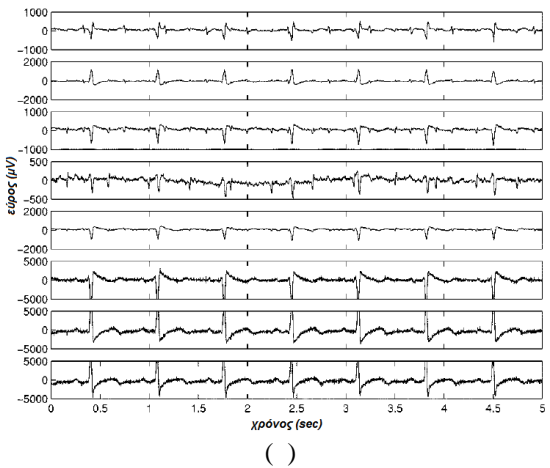
Daisy [38].



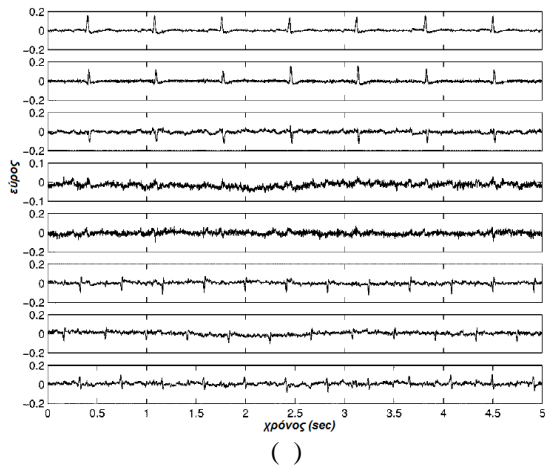
μ 3.7. μ μ



() PCA () ICA [8].



μ 3.8. δ- μ



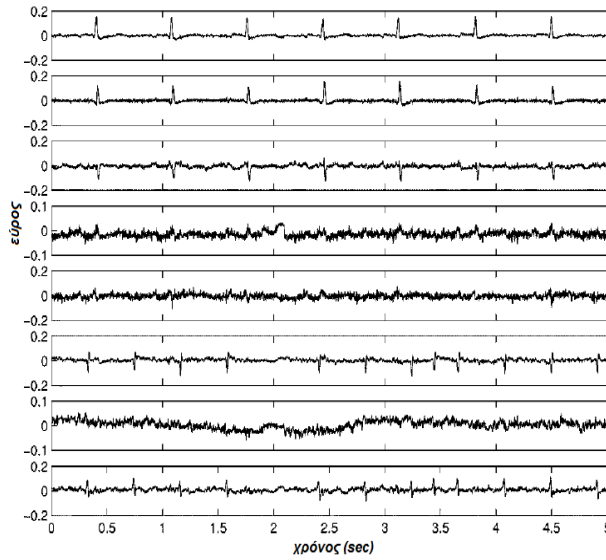
() μ () μ μ

ICA.

[38]. μ μ μ μ Daisy
 μ μ μ μ μ ,
 μ μ μ (μ μ).
 μ 10 sec μ μ 250 Hz
 μ 5 μ 3 (μ 3.6).
 μ PCA ICA μ 3.7.
 μ PCA μ - (3 μ
 μ 5 μ μ), ICA
 μ (μ
 1-3 μ 3.7).

μ , 7 PCA- μ 8 ICA- μ
 μ , 6 μ ICA μ - -
 μ , 6 μ PCA. μ μ 6 μ
 PCA μ μ (7 μ ICA;
 μ μ
 μ - . ..). μ 4 8
 PCA μ 5 μ
 ICA. μ 3.8() μ μ μ
 μ 2 μ , μ ICA μ μ 3.8().
 μ μ μ 6 7.

μ BSS μ μ
 μ , μ μ ICA μ μ μ
 Daisy μ . μ 3.9 μ
 μ μ , μ

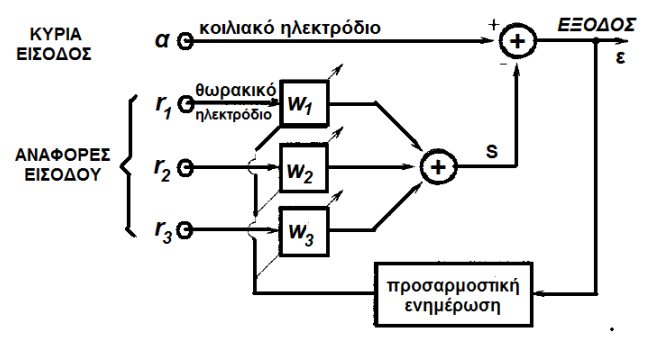


μ 3.9. μ ICA, μ μ 3.5
sec 2 sec.

μ μ , PCA μ
μ μ , μ (μ
μ μ , μ
... μ .
- (ICA)
μ μ .

Zarzoso et al. [5] , μ
μ μ Widrow's (MRANC) [39].
MRANC $a(k)$

μ , μ μ
(μ 3.10).



μ 3.10. MRANC μ .

$(r_i(k), i = 1, \dots, n)$
 $(w_i(k), i = 1, \dots, n)$ (taps)- μ N , (FIR)

$$\varepsilon(k) = \alpha(k) - \varepsilon(k), \quad s(k) = \sum_{i=1}^n w_i(k) * r_i(k),$$

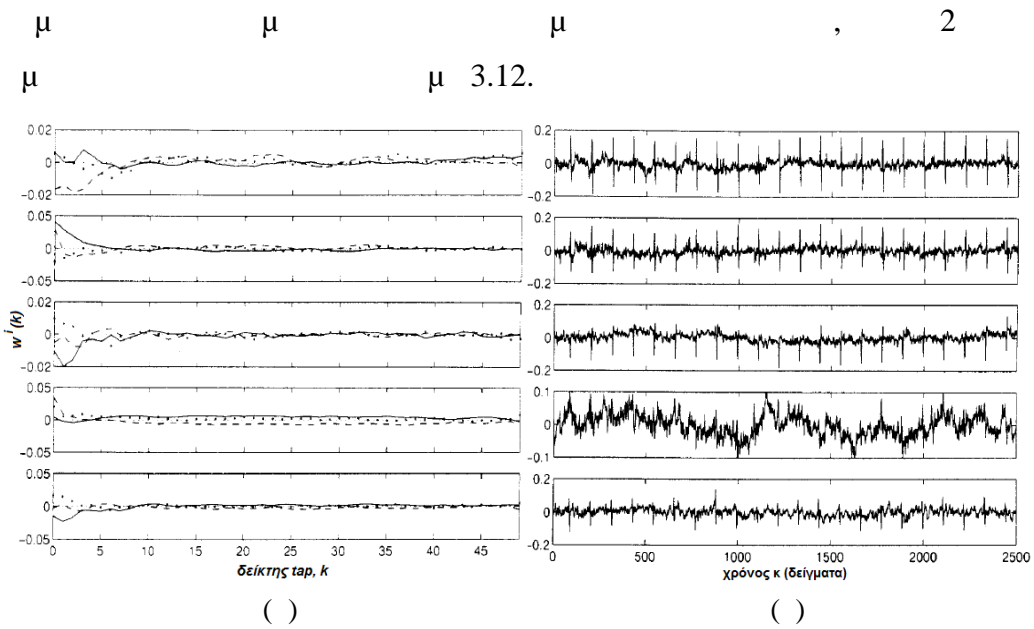
$*$
 (MSE) .
 (LMS)

[40].

MSE
 MSE (MMSE),
 misadjustment. MMSE
 Wiener-Hopf (WH) [40]

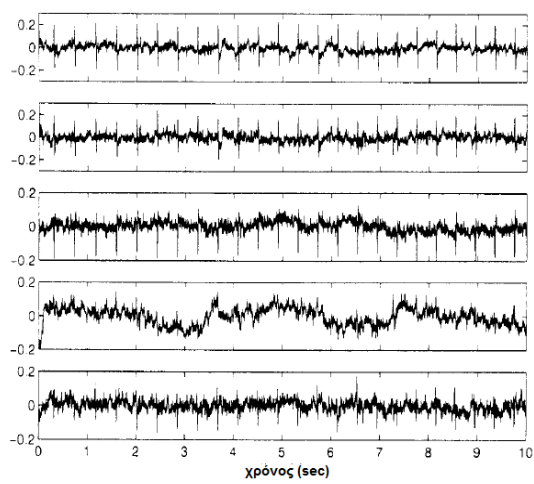
2 (ICA
 2μ
 L. De Lathauwer (5000 μ)
 Daisy (2500 μ) [39]. Widrow MRANC, 3

$(\dots, n=3, \mu \dots 3.10)$,
 $(\mu \dots)$,
 WH 50
 $\mu \dots 3.11(\dots)$,
 $3.11(\dots)$. $\mu \dots$,
 $\mu \dots$ MRANC,



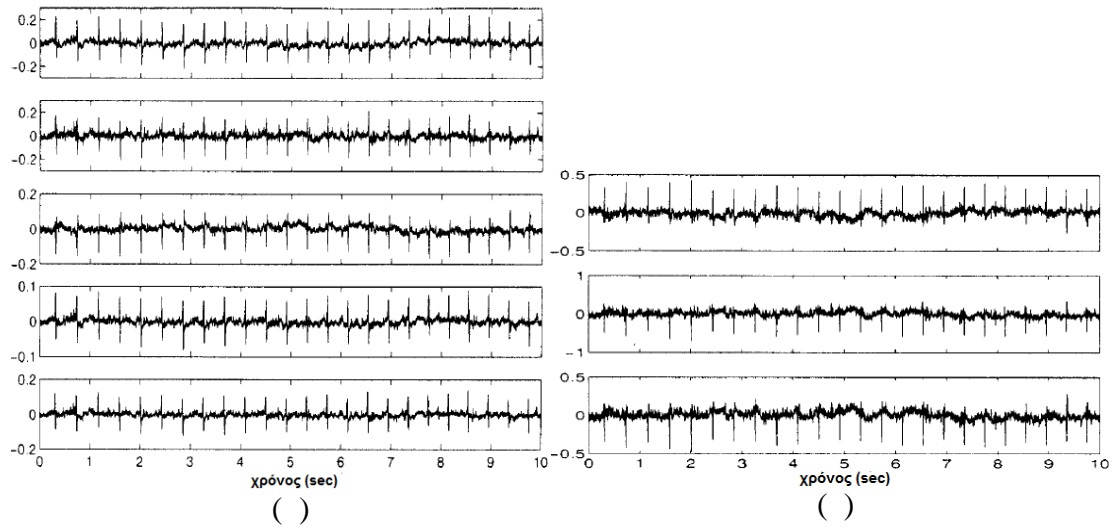
μ 3.11. () MRANC 1 μ . ()
 μ (5) μ 1
 μ μ MRANC μ .

BSS-ICA (μ 3.7()), μ _____
 μ ()
 μ 2) . μ
 μ $\hat{s}_f = [\hat{s}_{f_1}, \hat{s}_{f_2}]$, μ
 μ $\hat{a}_f = [\hat{a}_{f_1}, \hat{a}_{f_2}]$. μ
 μ $\hat{x}_f = \hat{a}_f \hat{s}_f$.

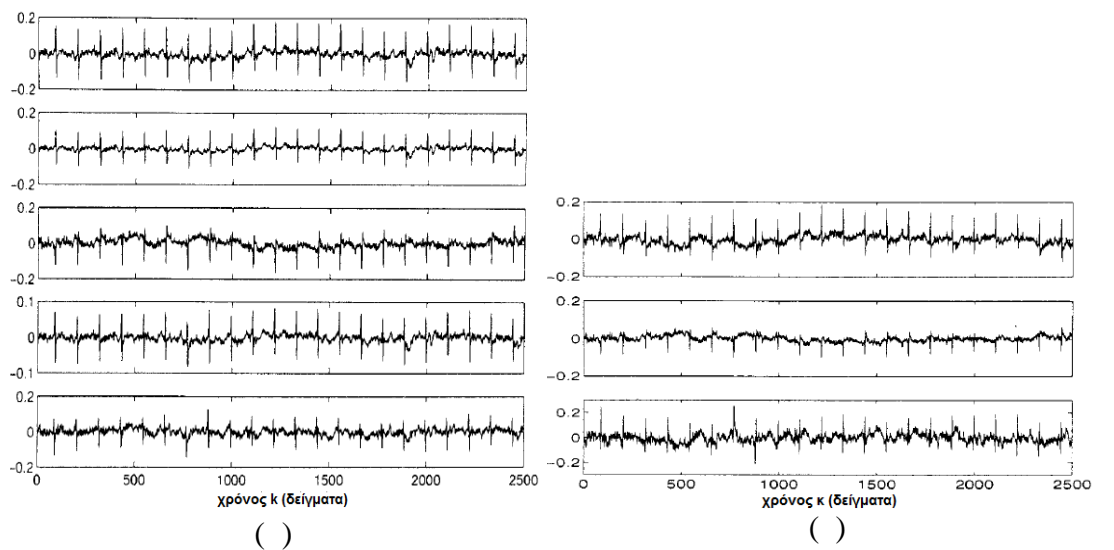


μ 3.12. μ (μ)
 μ 2 Daisy μ μ MRANC.

μ
 μ 3.13() μ
 μ MRANC. μ
 μ 3.13().

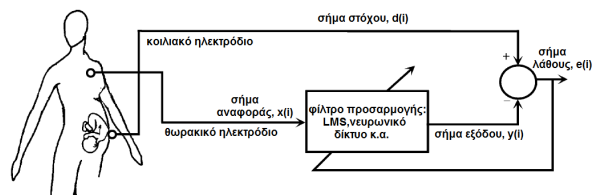


μ 3.13. () μ (μ)
 I μ μ μ BSS μ . () μ
 (3 μ) I μ μ μ
 BSS μ .



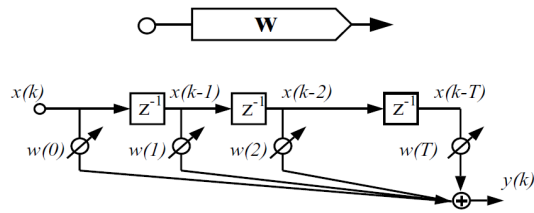
μ 3.14. () μ (μ)
 2 μ μ μ BSS μ . () μ
 (3 μ) 2 μ μ μ
 BSS μ .

μ 2 μ ,
 μ $3.14()$ $3.14()$.
 μ , , .
 μ BSS-ICA
 μ MRANC,
 μ .
 Camps-Valls *et al.* [13] μ μ FIR
 μ gamma ANC μ (LMS
 μ μ μ μ , NLMS)
 μ μ μ μ μ μ μ , μ
 μ [41], μ .
 μ ANC μ , μ μ μ (μ 3.15).
 $x(i)$ μ μ
 μ μ , μ μ $d(i)$
 μ . μ (μ)
 μ μ $e(i)$.



μ 3.15. μ ANC.

FIR μ
 (multilayer feedforward neural network-MFNN)
 μ μ FIR [42]. μ μ μ
 μ μ μ 3.16. FIR ,
 $y(k)$ μ μ
 μ $x(k): y(k) = \sum_{n=0}^T w(n)x(k-n)$,
 $w(n)$ FIR .

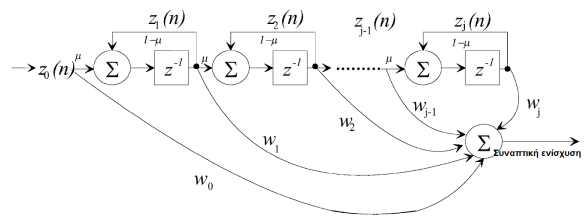


μ 3.16. FIR .

gamma μ MFNN
 μ μ . gamma IIR
 μ
 - μ : $G(z) = \frac{\mu}{z - (1 - \mu)}$, μ

$G(z)$. gamma μ μ

μ 3.17.



μ 3.17. gamma.

back-propagation μ Wan [42]
 μ FIR . μ
 μ μ μ μ μ :
 μ /μ μ μ μ (SNRfm), μ /gaussian
 (SNRfn) μ / μ μ (SNRfe) μ
 μ μ μ μ μ (BW), μ
 μ μ μ (HRV_m HRV_f,)
 μ .
 μ μ μ μ μ
 SNRfm. , μ μ μ μ μ
 μ μ LMS μ . μ
 LMS ANOVA μ , μ μ
 μ , SNRfm -35 dB.

μ , μ μ NLMS
 μ μ
 (SNR_{fm} SNR_{fn}) , μ ,
 $\mu\mu$, μ μ QRS
 μ , μ , μ FIR μ ,
 μ LMS μ ,
 μ μ μ . gamma,
 μ μ μ R- ,
 μ μ μ .
 μ μ μ . FIR
 μ μ μ
 μ . μ ' ' μ
 μ , μ
 μ . μ μ
 μ . μ μ
 μ . μ μ
 3.1. μ
 FIR , μ μ
 3-18% μ μ (Positive Predictive Value-PPV) 6-
 18% (Sensitivity-Se) μ LMS- μ
 μ .

3.1. Se(%) PPV(%) () μ μ

| | <i>LMS</i> | <i>NLMS</i> | <i>FIR</i> | <i>Gamma</i> |
|-----------------|--------------|-------------|--------------------|--------------------|
| <i>H1</i> | 75.0 (33.00) | 87.5 (30.0) | 88.9 (20.0) | 87.5 (22.2) |
| <i>M1</i> | 60.0 (40.0) | 66.7 (40.0) | 70.0 (22.2) | 70.0 (30.0) |
| <i>Catift22</i> | 66.7 (33.3) | 66.7 (33.3) | 72.7 (33.3) | 70.0 (30.0) |
| <i>Catijg24</i> | 60.0 (25.0) | 60.0 (33.3) | 77.7 (22.2) | 77.7 (30.0) |
| <i>Reg25</i> | - | - | - | - |
| <i>Reg1000</i> | - | - | - | - |
| <i>MexIigus</i> | 60.0 (33.3) | 60.0 (33.3) | 77.7 (30.0) | 70.0 (30.0) |
| <i>Mex2</i> | - | - | - | - |

μ μ μ 3.18, μ μ μ
 μ r_i , μ μ μ y_i μ
 μ μ μ

μ μ μ μ μ (MSE)
 μ 3.18
 [43] μ [44] μ

μ μ μ w_i μ
 μ y_i μ r_i y_i

μ μ μ μ μ ,
 μ μ μ μ μ ,
 μ μ , μ μ μ

μ μ μ μ μ μ μ μ μ μ
 μ μ μ μ μ μ μ μ μ μ
 μ μ μ μ μ μ μ μ μ μ

ii. μ μ μ μ μ μ μ μ μ μ
 μ μ , μ μ μ μ μ μ μ μ μ μ

μ ICA μ μ μ μ μ μ μ μ μ
 . 4.5.6(ii), μ μ μ μ μ μ μ μ μ μ
 μ μ , μ μ μ μ μ μ μ μ μ μ

, μ

[45,46].

iii.

μ

. 4.5.5

μ

μ

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, μ

μ

μ A.

μ

μ

μ

μ

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μ

.

μ

A

$\mu \mu$

μ

ICA, μ

μ

μ

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μ

μ

Sato *et al.* [23]

$\mu \mu$

μ

BSS

μ

(BSSR).

μ

μ

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μ

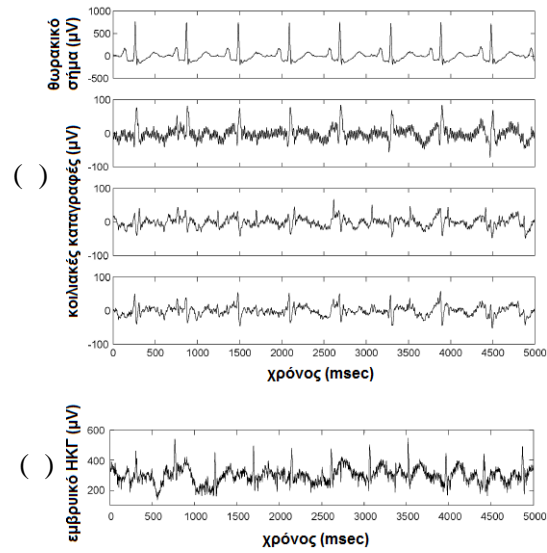
μ

μ

μ

$\mu\mu$

μ (μ 3.19).



μ 3.19. ()

() μ

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μ μ

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μ 3.20

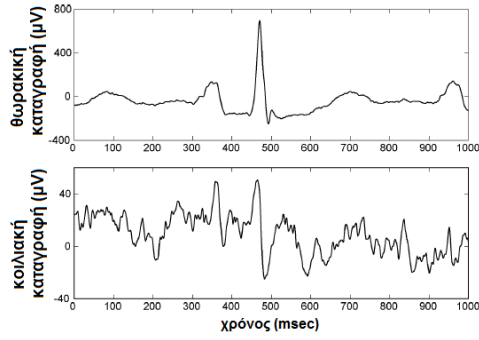
μ

μ

,

μ

.

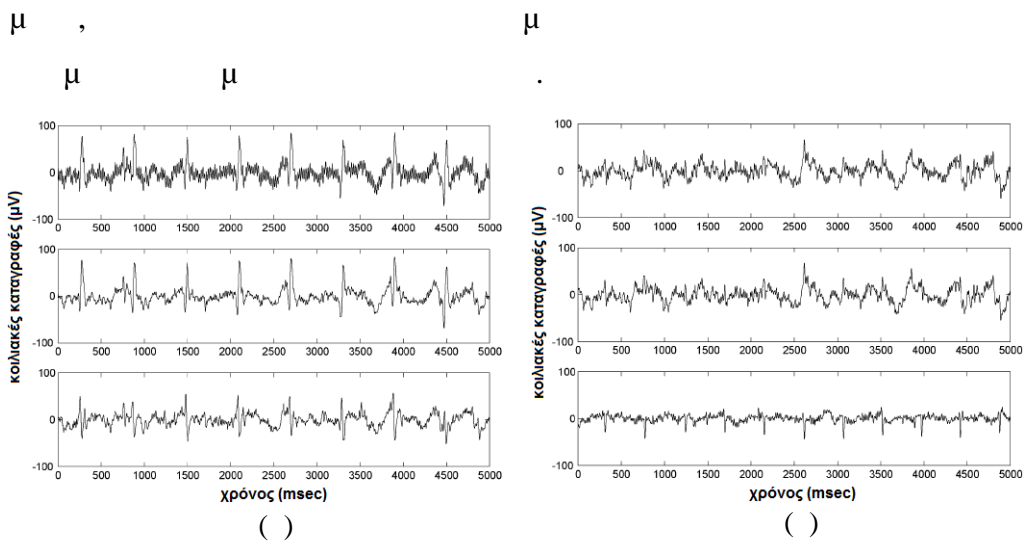


μ 3.20. μ μ μ μ μ ,
 μμ . μ
 , μ μ .

μ 3.21 μ μ .

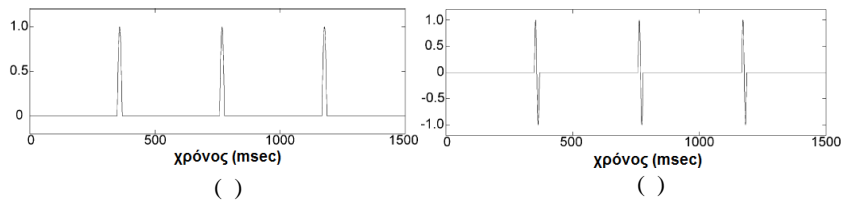
μ μ ICA/BSS.

BSSR μ P
 μ R μ μ μ
 μ . μ BSS $x=As \Rightarrow y=w^T x$,
 y w μ μ μ $^{-1}$. μ



μ 3.21. μ μ μ μ
 . () μ μ
 . () μ .

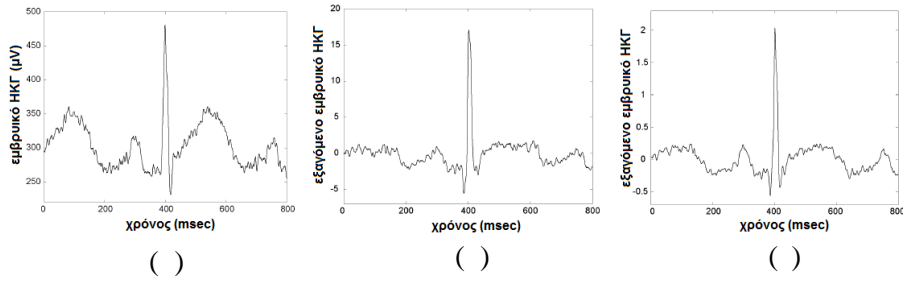
μ , μ μ μ 3.22
 μ , μ μ .
 μ μ Doppler
 μ μ



μ 3.22. μ . () μ . () μ

μ BSSR,
 μ μ μ μ μ , μ , μ
 μ μ μ μ , EMG
 μ . 5 autoregressive μ (AR) μ
 μ μ μ μ μ
 μ μ μ . μ μ μ
 μ . ICA μ ,
 μ 3.23(). , BSSR
 μ , μ 3.23().

μ 3.24 μ , ,
 μ ICA BSSR μ μ ,
 μ BSSR, μ
 μ μ μ μ
 μ .



μ 3.25. μ μ μ μ $\mu\mu$ μ
 μ μ R μ μ $\mu\mu$
 $()$ μ $ICA.$ $()$ μ $BSSR$.
 , μ μ
 μ μ μ μ μ :
 P $\mu = +.05ms/ \mu.$, QRS $\mu = -.17 ms/ \mu.$, PR $\mu = +.03 ms/ \mu.$,
 PQ $\mu = +0.03 ms/ \mu.$

3.8. (BSE)

μ BSS
 μ μ μ . BSS μ ,
 μ [47]. , BSS μ
 $\mu\mu$
 μ μ . , μ μ
 μ μ μ μ (μ), μ
 μ μ μ μ
 μ μ .
 μ (BSE) [1,16,25,47].
 μ μ 'deflation' , BSE μ μ
 μ μ
 [47]. BSE μ μ μ BSS
 [1]. μ , μ μ μ

“”; μ μ μ μ ; μ
 μ .

μ , μ μ μ . μ
 μ , , μ
 , μ , μ , μ .
 , μ $\mu\mu$ μ . . , *Lu et al.*

[48,49] μ ICA-R ,
 μ μ μ μ μ .
 μ - μ ICA
 μ (trace) μ
 μ ,

μ , , μ μ ,
- . μ μ ,
 μ μ . O Barros *et al.*

[8] batch μ μ (BCBSE) μ -
 μ μ , μ μ
, , μ . ,
 μ μ μ μ ,
 μ μ
 μ (innovation) μ .
 μ μ

μ [50]. Shi *et al.* [50] μ μ - μ
(SemiBSE), μ -gaussianity
 μ μ . μ μ

[50].

[51] Shi *et al.*

EBS (GABSE). GABSE

SemiBSE [50].

Zhang *et al.* [52] fixed-point

GABSE (BCBSE, SemiBSE GABSE)

[38]

[53-55],

ICA

μ - μμ μ

μ . , - μ ,

μ μ μ μ , μ , μ

μ μ ,

. μ , μ

μ μ μ [1]. μ ,

, μ μ ICA μ ,

μ μ μ μ

; μ μ .

, μ μ μ μ μ

μ μ , μ μ

μ μ μ μ

R-R μ , “ ” μμ

μμ μ .

μ μ Sameni *et al.* [56] μ μ

μ μ (CA) [57],

μ μ [58].

, μ μ ICA

μ .

μ μ μ μ

μ μ R-

μ μ PCA ICA .

, -

, μ μ μ

μ .

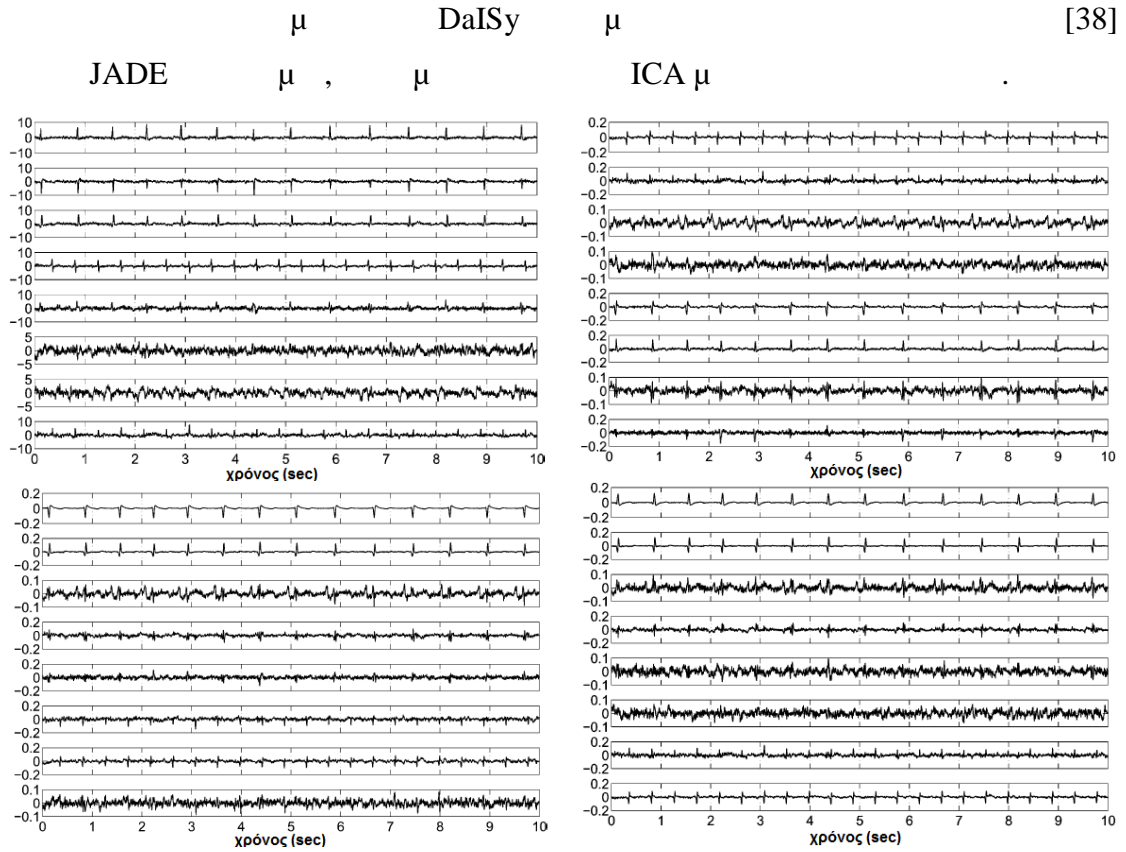


Figure 3.26: Comparison of JADE and ICA methods. The figure shows two columns of plots. The left column, labeled 'JADE', shows eight rows of plots. Each row contains a clean signal (top) and a noisy version (bottom). The y-axis ranges from -10 to 10. The right column, labeled 'ICA', shows eight rows of plots. Each row contains a clean signal (top) and a noisy version (bottom). The y-axis ranges from -0.2 to 0.2. The x-axis for all plots is 'χρόνος (sec)' from 0 to 10.

Figure 3.26 shows the results of JADE and ICA methods. The figure is divided into two columns. The left column shows the results of JADE, and the right column shows the results of ICA. Each column contains eight rows of plots. Each row contains a clean signal (top) and a noisy version (bottom). The x-axis for all plots is 'χρόνος (sec)' from 0 to 10. The y-axis for the JADE plots ranges from -10 to 10, and for the ICA plots, it ranges from -0.2 to 0.2.

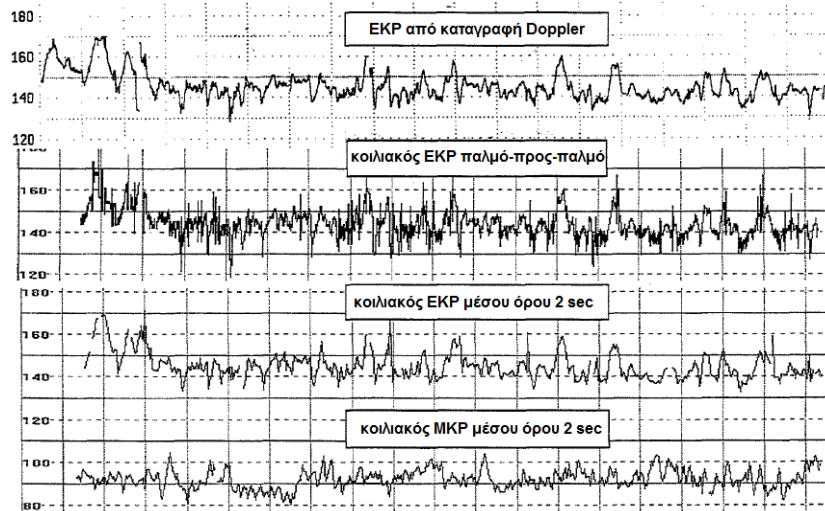
3.9. Comparison of JADE and ICA methods. The figure shows two columns of plots. The left column, labeled 'JADE', shows eight rows of plots. Each row contains a clean signal (top) and a noisy version (bottom). The y-axis ranges from -10 to 10. The right column, labeled 'ICA', shows eight rows of plots. Each row contains a clean signal (top) and a noisy version (bottom). The y-axis ranges from -0.2 to 0.2. The x-axis for all plots is 'χρόνος (sec)' from 0 to 10.

Pieri *et al.* [59]. : *i.* (μ QRSs μ) *ii.* $\mu\mu$ $\mu\mu$ 'blanking' μ (μ). , μ QRS $\mu\mu$. μ (phase space plot), μ $\mu\mu$ $\mu -$. μ μ μ « » μ . μ , QRSs μ μ $\mu -$. μ μ μ . μ $\mu\mu$ μ μ μ . μ 3.27 μ , μ

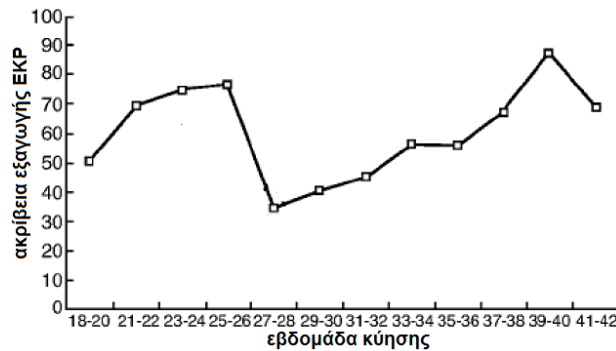
Doppler.

μ $\mu -$ μ μ μ μ μ μ (SNR). $\mu -$ μ μ , μ μ μ μ $\mu -$ μ $\mu -$ - μ μ . μ

Doppler μ 3.27.



μ 3.27. μ μ Doppler
μ μ μ .



μ 3.28. μ μ .

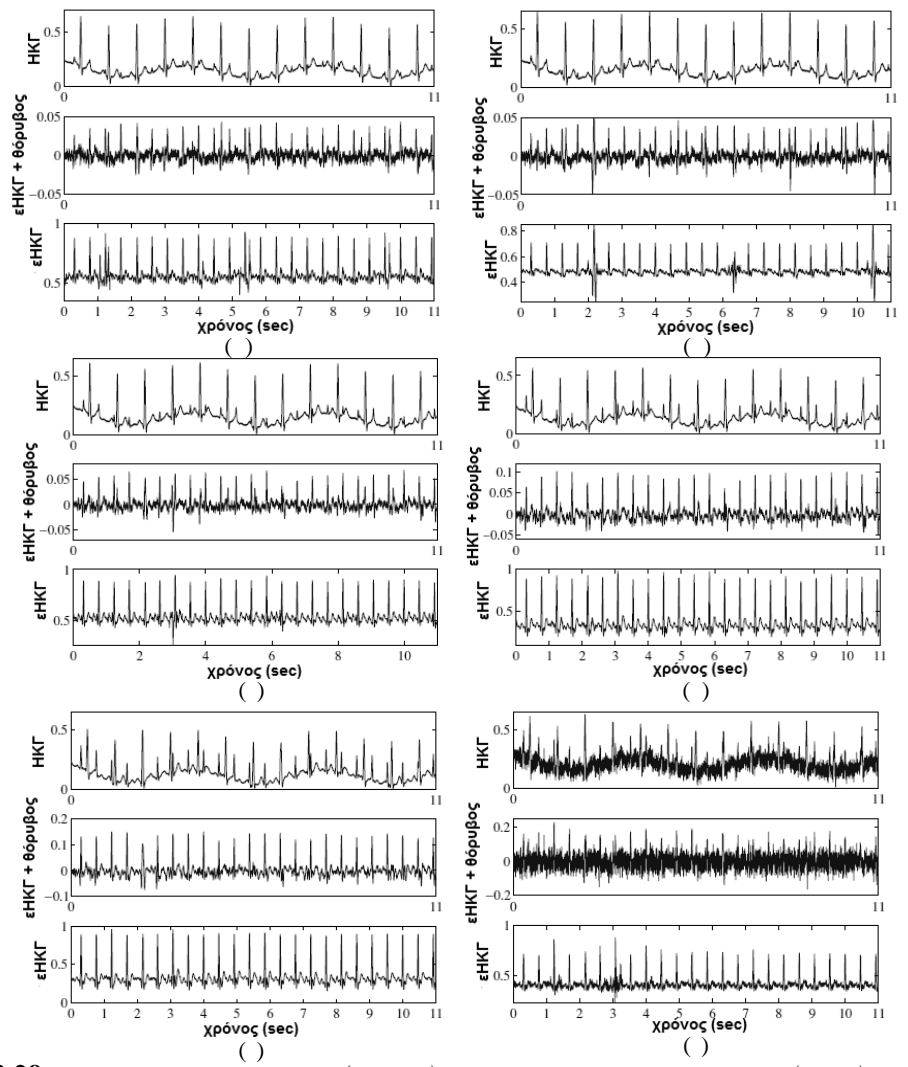
μ , 400 μ
(5-10). μ 3.28
, μ μ μ
(. > 80%) , μ
60% 18-26 μ 33-41
μ μ . μ μ ,
65%.

Al-Zaden *et al.* [3] . 3.5,

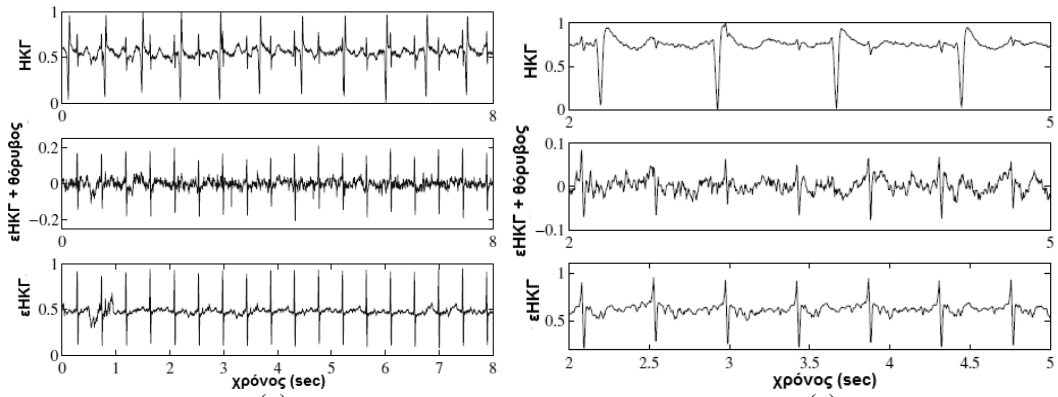
μ μ ()
μ μ
2 μ : *i.* μ

ii.

SVD ANFIS



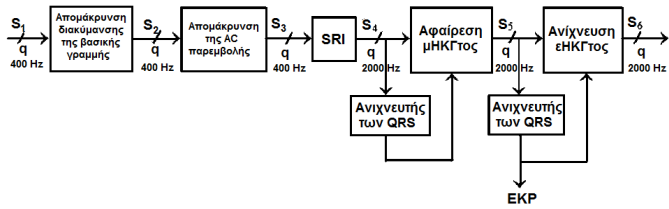
μ 3.29. () () MFSNR=25 dB () MFSNR=20 dB () MFSNR=15 dB () MFSNR=10 dB () MFSNR=5 dB () MFSNR=5 dB μ white Gaussian μ SNR=15.



μ 3.30. (a) $signal$ (), $DaISy$ (), $μ$ (), $μ$ ().

μ (ecgsyn) [60] μ .
 μ μ μ SNRs. μ μ μ 3.29
 μ μ μ (), μ μ
 1 μ μ μ (μ)
 μ μ μ ()
 μ Maternal-Fetal SNR (MFSNR).

μ , μ μ
 μ DaISy [38]. μ 3.30() μ
 μ μ , μ 3.30() μ .
 μ Martens *et al.* [7] μ μ - (non-blind) μ
 . μ
 μ (sequential analysis approach),
 μ μ
 . μ 3.31 block μ μ μ



μ 3.31.

Η διαδικασία επεξεργασίας του σήματος ECG περιλαμβάνει τα ακόλουθα βήματα:

- Απομάκρυνση της διακομμένης της διαρκούς γραμμής (S1 → S2): Απομάκρυνση της διακομμένης της διαρκούς γραμμής (S1) με συχνότητα 400 Hz.
- Απομάκρυνση της AC παρεμβολής (S2 → S3): Απομάκρυνση της AC παρεμβολής (S2) με συχνότητα 400 Hz.
- Σύμφωνο (SRI) (S3 → S4): Σύμφωνο (SRI) με συχνότητα 400 Hz.
- Αφαίρεση μΗΚΓΤος (S4 → S5): Αφαίρεση μΗΚΓΤος (S4) με συχνότητα 2000 Hz.
- Ανίχνευση εΗΚΓΤος (S5 → S6): Ανίχνευση εΗΚΓΤος (S5) με συχνότητα 2000 Hz.

Επιπλέον, ο ανιχνευτής των QRS (QRS) παρέχει πληροφορίες για την εμφάνιση του QRS, οι οποίες χρησιμοποιούνται για την αφαίρεση του μΗΚΓΤος και την ανίχνευση του εΗΚΓΤος.

Η συχνότητα του σήματος είναι 400 Hz. Η συχνότητα του φίλτρου είναι 2000 Hz. Η συχνότητα του φίλτρου είναι 150 Hz. Η συχνότητα του φίλτρου είναι 3 Hz. Η συχνότητα του φίλτρου είναι 400 Hz. Η συχνότητα του φίλτρου είναι 2000 Hz (upsampling).

Η συχνότητα του φίλτρου είναι 150 Hz. Η συχνότητα του φίλτρου είναι 3 Hz. Η συχνότητα του φίλτρου είναι 400 Hz. Η συχνότητα του φίλτρου είναι 2000 Hz (upsampling).

Η συχνότητα του φίλτρου είναι 150 Hz. Η συχνότητα του φίλτρου είναι 3 Hz. Η συχνότητα του φίλτρου είναι 400 Hz. Η συχνότητα του φίλτρου είναι 2000 Hz (upsampling).

Η συχνότητα του φίλτρου είναι 150 Hz. Η συχνότητα του φίλτρου είναι 3 Hz. Η συχνότητα του φίλτρου είναι 400 Hz. Η συχνότητα του φίλτρου είναι 2000 Hz (upsampling).

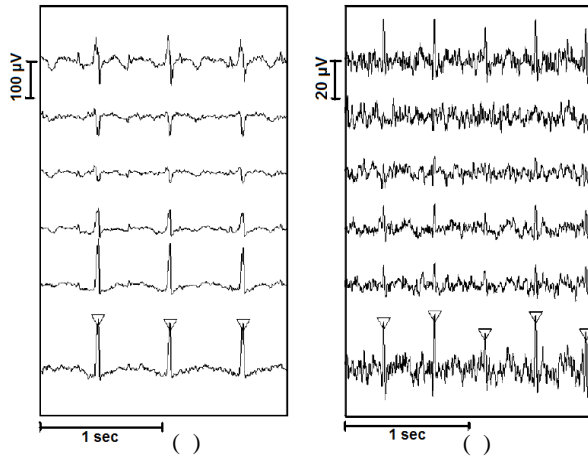
Η συχνότητα του φίλτρου είναι 150 Hz. Η συχνότητα του φίλτρου είναι 3 Hz. Η συχνότητα του φίλτρου είναι 400 Hz. Η συχνότητα του φίλτρου είναι 2000 Hz (upsampling).

Η συχνότητα του φίλτρου είναι 150 Hz. Η συχνότητα του φίλτρου είναι 3 Hz. Η συχνότητα του φίλτρου είναι 400 Hz. Η συχνότητα του φίλτρου είναι 2000 Hz (upsampling).

PCA

μ μ μ
Martens *et al.* [61].

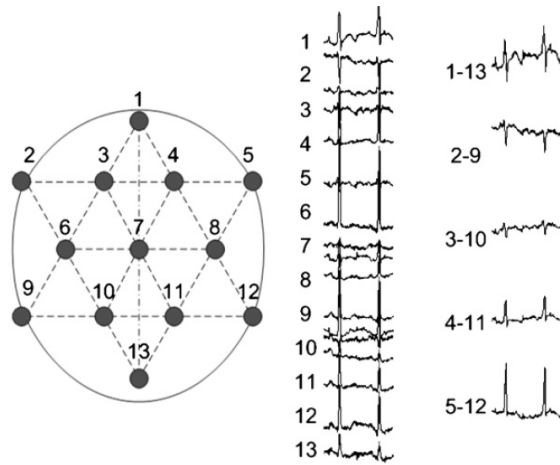
μ μ μ μ μ μ
 μ μ μ μ μ μ
 μ SNR.
QRS. QRS μ μ
 μ QRS. QRSs μ
 μ QRS [62].
 μ 3.32 μ QRS μ
 μ



μ 3.32. QRS μ () μ S4 (μ) μ QRS
(μ). () μ S5 (μ) μ μ QRS
 μ QRS (μ). S5
() μ μ μ QRS
(μ). μ QRSs μ .

μ μ S4 .
 μ Cerutti *et al.* [63]. μ
 μ μ μ μ μ μ μ
 μ μ μ .
 μ P, QRS [64],
 μ μ μ μ μ ,

μ P, QRS μ .
 μ , $-\mu$ μ μ μ μ ,
 μ S5 μ μ
 μ QRS μ μ , ' μ QRS
 μ 20 μ , $\mu\mu$ 20
 Maxima Medical Center Veldhoven () .
 30 .
 μ , μ 19 .
 μ μ Porti 5-24/ASD (TMS
 International), μ 24 . μ μ
 13 μ μ
 μ 3.33. 13 μ

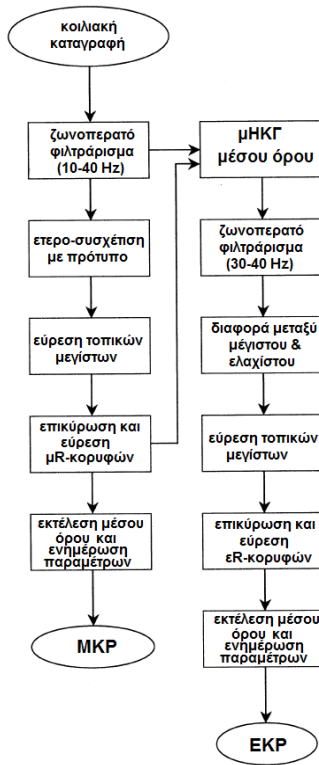


μ 3.33. μ () ,
 μ 13 μ (μ) 5- ()
 μ , μ
 μ $\mu\mu$ μ (1-13, 2-9,
 3-10, 4-11 5-12).
 $\mu\mu$ S1. μ μ μ
 μ μ μ μ .

, μ μ , SNR
 , μ μ μ μ .
 μ μ JADE μ (ICA) ,
 (85% 60%) μ JADE. μ JADE
 μ μ . 28-
 32 μ μ μ
 μ . μ μ
 μ ICA ,
 μ SNR.

Ibahimy *et al.* [17] μ
 μ , μ ,
 μ , μ μ
 μ μ μ μ μ
 μ μ μ μ μ μ μ . To μ
 μ μ μ μ , (μ)
 μ μ μ μ 500 Hz
 μ 13-bit. μ 5 μ
 μ μ μ μ 35 40
 μ . μ
 (μ),

μ μ μ μ μ
 μ [65-67]. μ 3.34, μ
 μ μ QRS μ μ , .



μ 3.34. μμ μ .

μ - μ μ

μ QRS μ μ μ , μ

μ , μ μQRS.

μ μ μ QRS,

μ QRS . μ (μ)

μ μ μ μ R

μ .

μ μ R μ μ

[66] μ QRS.

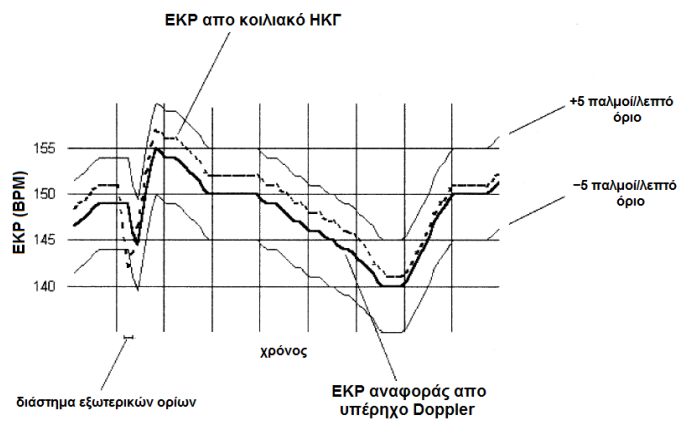
μQRSs, μ

μ μ μ μ μ μ

μ . [67]

μ μ μ

μ .
 μ (μ).
 μ 3.35 μ μ
 μ μ Doppler, μ μ ± 5 bpm
 μ 4164
 (μ μ 37.6) μ 5
 μ μ μ
 μ (PRD) μ 3.31% 6.63% , μ
 μ μ μ 5.32%.



μ 3.35. μ (μ μ) μ
 μ (μ μ μ μ) (μ μ)
 μ (μ μ μ) μ μ
 5 BPM μ ()
 84.1% . μ μ
 0.84 0.93 , μ μ μ μ 0.89.

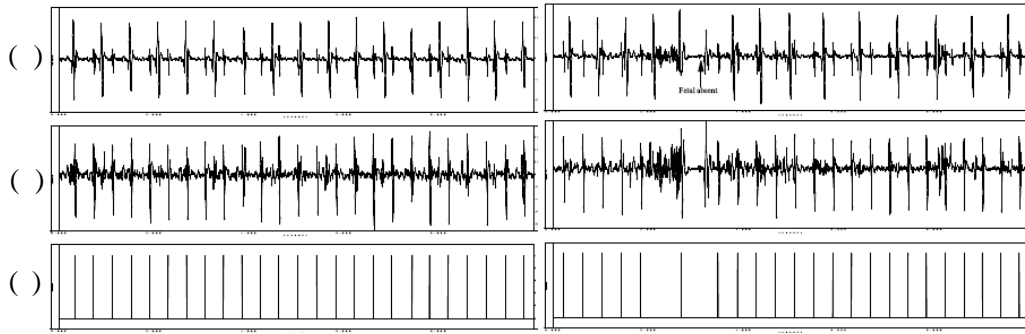
μ Azad [18] μ .
 μ , μ
 μ .
 μ μ 3.36.

μ μ μ μ .
 μ μ μ
 μ μ μ μ .
 μ μ μ QRSs μ μ
 μ μ μ μ μ
 . μ μ μ μ μ
 μ μ μ μ
 - μ μ , μ
 RS μ
 - - RS μ .
 μ μ 5 μμ
 [68] μ .
 μ μ μ 500
 Hz 31 41 μ μ 3.37
 μ μ μ μ μ
 μ 39 μ μ μ
 . μ μ

[69]:

$$performance = \frac{\# fetal_R_wave - (\# misses + \# false)}{\# fetal_R_wave}$$

μ μ μ μ μ μ μ 89% (μ
 73%) μ μ R- μ .



μ 3.37. μ μ μ , () μ μ () μ QRS () μ
 $QRSs$.

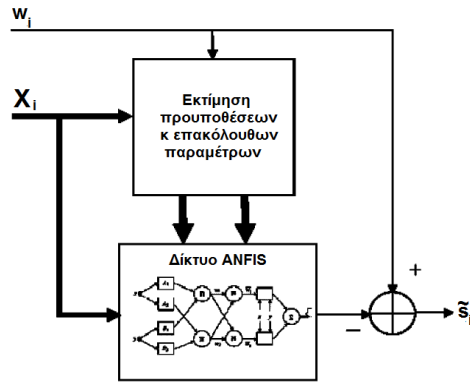
Assaleh [14] ANFIS
 μ , μ 2 μ , μ μ μ - $\mu\mu$
 μ μ μ μ . μ ANFIS
 μ μ μ - $\mu\mu$, $\mu\mu$ μ
 μ μ .

ANFIS Sugeno μ
 μ μ μ
 μ [70]. μ
 μ μ ANFIS μ (pattern learning)
 μ $\mu\mu$ μ .
 ANFIS
 μ μ μ $\mu\mu$
 μ μ μ
 μ μ . ,
 ANFIS (initializations)
 μ
 [71].

μ , μ μ μ μ μ T -
 based). , $(x(n))$ μ $(w(n))$
 μ N - μ μ μ μ μ
 μ i - μ : $x_i(m) = x_i(N-P+m)$
 $w_i(m) = w_i(N-P+m)$, $0 \leq m \leq N-1, i \geq 0$.

ANFIS μ $\mu\mu$ $x(n)$
 μ μ $w(n)$. μ , ANFIS
 μ μ (μ)
 $x(n)$, μ ANFIS $w(n)$.
 ANFIS μ (mapping) μ
 $x(n)$ $w(n)$ μ $x(n)$ $\hat{x}(n)$. μ
 μ , i - μ , ANFIS μ
 μ (μ) X_i μ w_i μ
 μ
 μ μ :
 $X_i = [x_i(0), x_i(1), \dots, x_i(N-1)]^T$ μ $x_i(m) = [x_i(m), x_i(m-1), \dots, x_i(m-j)]$
 $w_i = w_i(m) = [w_i(0), w_i(1), \dots, w_i(N-1)]^T$.

μ
 μ μ $\mu\mu$ μ μ μ
 $x(n)$ μ $\hat{x}(n)$ $\mu\mu$ μ μ
 $w(n)$. , μ μ , μ μ
 μ $s(n)$ μ $\hat{s}(n)$ μ μ $\hat{x}(n)$
 $w(n)$. μ μ 3.38.



μ 3.38.

ANFIS

i.

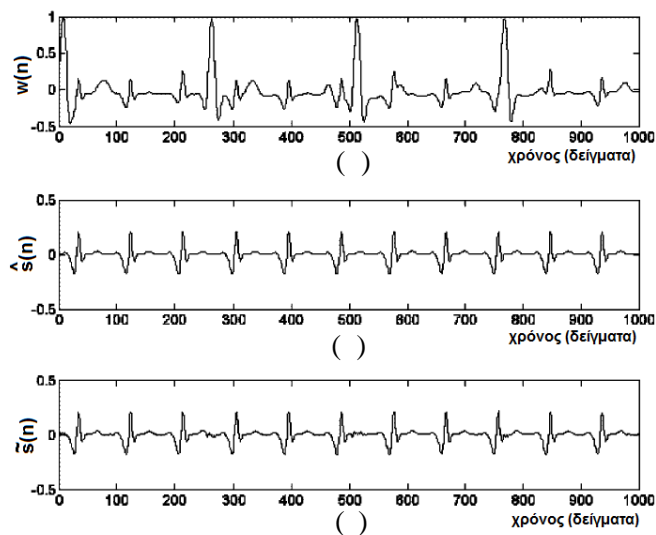
μ ANFIS μ μ
 (.. μ μ) ,
 , μ ANFIS μ μ
 , w_i . μ μ
 μ ANFIS μ μ
 x_i , μ w_i - x_i .
 μ μ μ μ μ μ .
 μ μ μ μ μ μ
 McSharry et al. [60]. μ (s(n)) (x(n))
 μ μ μ μ μ μ
 μ , μ μ μ
 (w(n)) μ
 μ μ . (fmSNR = 10 log₁₀ (Σ_n (ŝ(n))² / Σ_n (x̂(n))²)), ŝ(n)
 x̂(n) μ s(n) x(n) .
 μ μ μ μ μ μ
 μ μ μ μ μ μ
 3.39 μ μ μ μ μ μ
 fmSNR=-10dB μ μ μ
 s(n).

, μ μ μ μ μ μ
 μ (qSNR).

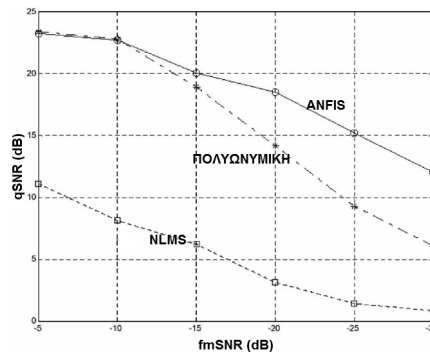
$$qSNR = 10 \log_{10} \left(\frac{\sum_n (\hat{s}(n))^2}{\sum_n (s(n) - \hat{s}(n))^2} \right).$$

qSNR μ 22.7 dB.

fmSNR μ μ -5 dB -30 dB μ μ 5 dB,
 qSNR μ . ANFIS μ
 (μ μ μ
 μ - [60]). μ qSNR
 μ 3.40 μ ANFIS
 μ .

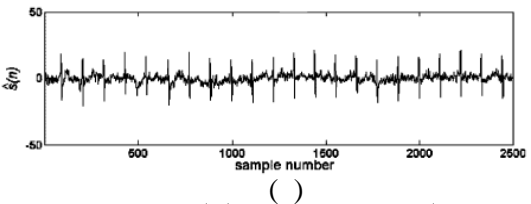
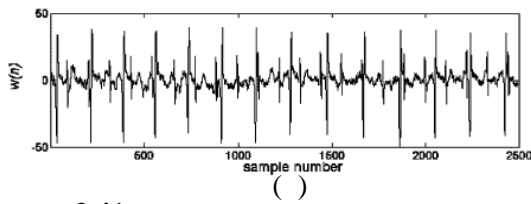


μ 3.39. . () μ μ fmSNR = -10
 dB, () , () μ (qSNR) = 22.7 dB.



μ 3.40. μ μ μ (fmSNR)
 qSNR μ

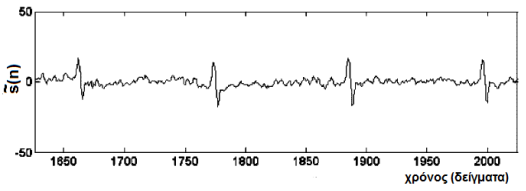
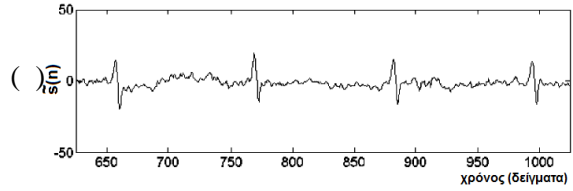
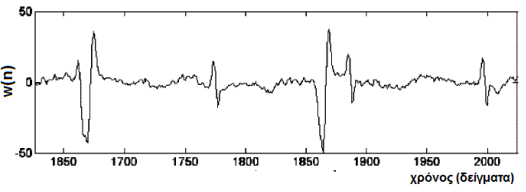
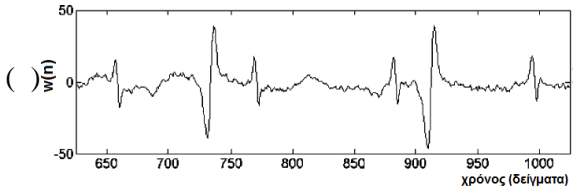
Daisy [38].
 fmSNR,
 SNR.



μ 3.41. () μ () ()

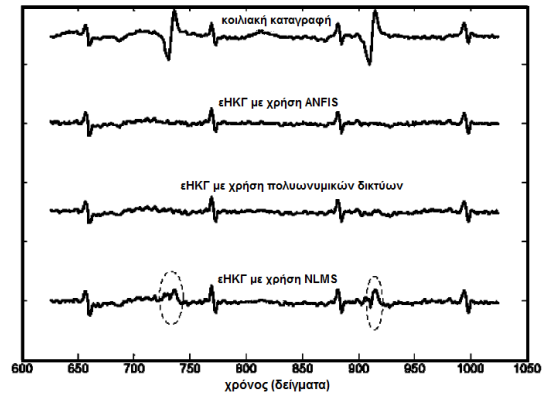
μ 3.42 (400 μ) μ

2



μ 3.42. () μ () μ

μ 3.43 μ μ 2 μ μ μ ANFIS μ μ μ μ ANFIS.



μ 3.43. μ μ μ μ μ μ , μ μ NLMS.

Khamene Negahdaripour 2000 [72] μ μ μ wavelet μ μ μ μ .

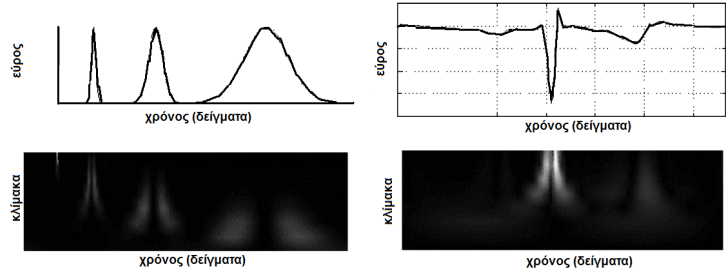
μ μ μ wavelet μ μ μ (WT) μ $x(t)$

$$W_x(s,t) = x(t) * \psi(t) = \frac{1}{s} \int_{-\infty}^{+\infty} x(t) \psi\left(\frac{t-\tau}{s}\right) d\tau$$

$\psi(t)$ μ wavelet , s μ . $W_x(s,t)$ μ wavelet . (μ) μ wavelet .

μ μ , bi-orthogonal quadratic spline wavelet. wavelet μ μ (μ 3.44). μ wavelet μ .

μ μ μ (singular points)
 [73]. μ μ μ μ μ μ
 μ μ (., P, QRS) μ μ μ .
 , μ , μ μ
 μ μ , μ , ,
 μ μ .



μ 3.44. () μ μ μ . ()
wavelet μ ,
 μ μ . () μ . () *wavelet*
 μ . *P, QRS,*
 μ μ μ μ μ .

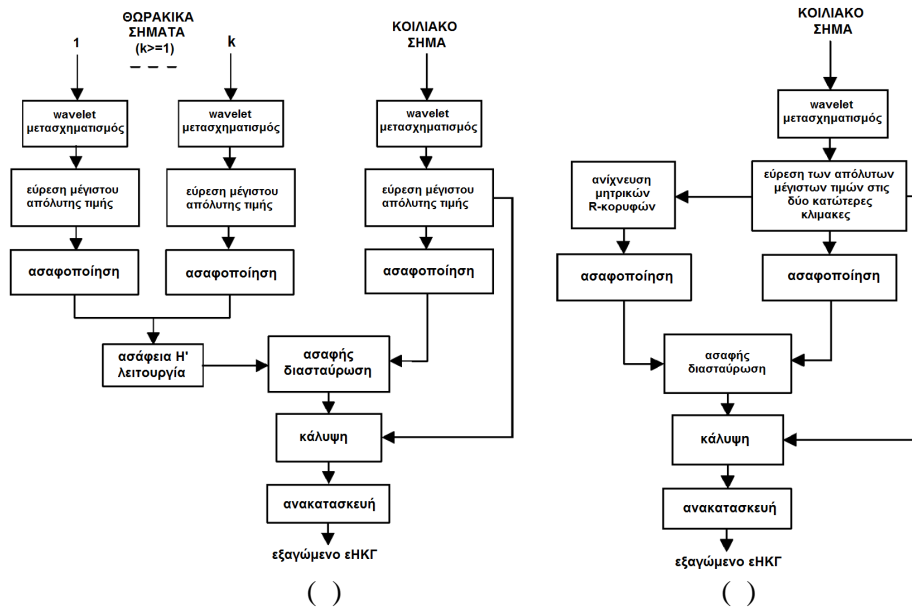
μ μ , μ μ
 μ . (μ) μ $x(t)$ μ
 μ μ Lipschitz [74]. μ μ
 μ *wavelet*
 μ .

μ μ μ *wavelet* μ μ -
 . μ μ μ
 μ , μ
 μ μ . μ μ μ *wavelet*
 μ

μ μ μ [74].
 μ , μ μ *wavelet*
 μ μ μ
 μ [73]. μ μ μ

wavelet [75].
 (,)
 [75]. $x(n)$ N ,
 wavelet ,
 .
 ,
 .
 N , wavelet $2^j, 1 \leq j \leq \log_2(N)$,
 .
 ,
 ,
 .
 .
 3.45()
 .
 [73],
 wavelet \mathbb{R}
 .
 $2^j, j=1,2$. ,
 wavelet $W_\alpha(2^j, m_{2^j}^l), \{m_{2^j}^l, l=1 \dots N_{2^j}\}$,
 \mathbb{R}
 , -
 .
 \mathbb{R}
 ,
 .
 3.45().

2 , μ ,
 . μ 2 μ
 μ μ MIT/ΒΙΗ, μ μ
 μ μ , μ . μ
 μ
 μ μ - μ (inverse-tangent
 operator) μ μ μ μ μ .



μ 3.45. μ μ () 1 () 2 .

1 SNR μ μ μ
 μ μ 41 dB. μ μ P, QRS,
 , μ μ R-
 μ . μ μ μ μ ,
 μ μ , μ μ
 μ μ SNR (μ μ
 μ). μ SNR 35-40 dB.
 2 SNR 20-25 dB.
 μ μ μ . μ

, μ μ μ 2 μ ; Daisy
 [38] μ 37 μ [76]

(μ). μ , 1
 μ μ , μ
 μ μ , μ
 μ , μ μ
 μ ‘ , μ . ,
 μ (μ) μ 15
 μ - μ . , 15
 μ , μ
 μ . μ μ μ
 μ , 3.2, μ
 μ 0.89
 μ μ , μ μ
 μ .
 μ μ μ μ
 μ . 2
 μ μ μ
 μ . μ μ μ
 μ μ μ μ
 μ , μ μ
 μ μ
($\mu\mu$ 3.2).

3.2. μ μ μ , μ
 μ (1) (2).

| μ | 1 | 2 | 3 | 4 | 5 |
|-------|------|------|------|------|------|
| | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 |
| 1 | 0.92 | 0.93 | 0.91 | 0.93 | 0.92 |
| 2 | 0.92 | 0.89 | 0.93 | 0.91 | 0.90 |
| 3 | 0.93 | 0.91 | 0.93 | 0.94 | 0.89 |
| | 0.89 | 0.82 | 0.79 | 0.83 | 0.83 |

Barros Cichori [8] batch
 μ μ μ μ μ
 μ $\mu\mu$ μ μ . μ μ

$x(k) = As(k)$. μ μ μ
 $(s_i(k))$ μ μ μ
 batch μ
 s_i μ
 i $:$
 $E[s_i(k)s_i(k-\tau_i)] \neq 0$ $E[s_i(k)s_j(k-\tau_i)] = 0, i \neq j$.

$y(k) = w^T x(k)$, $y(k)$ μ μ μ , μ μ
 s_i, k μ μ , w μ .
 μ ,
 $\varepsilon(k) = y(k) - by(k-p)$, b FIR μ
 z^{-pTs} , Ts μ (
 1).

$(w, b) = [\quad]^T$ μ μ
 μ μ , $(w, b) = w^T E[xx^T]w - 2bE[y_p w^T x] + b^2 E[y_p^2]$.
 μ w b , μ $y = w^T x$,

$$\frac{\partial \xi(w, b)}{\partial w} = 2E[xx^T]w - 2bE[y_p x] + b^2 E[x_p x_p] = 0 \quad \frac{\partial \xi(w, b)}{\partial w} = 2E[y_b y] - 2bE[y_b^2] = 0.$$

$w = E[xx^T]^{-1} E[y_b x] \frac{b}{1+2b^2}$. μ $\mu\mu$ $w=0$,
 $w = w/||w||$, $b/1+2b^2$ μ μ .
 μ
 $\mu\mu$ μ $[xx^T] = 1$,
 μ $: w = [xy_p]$.

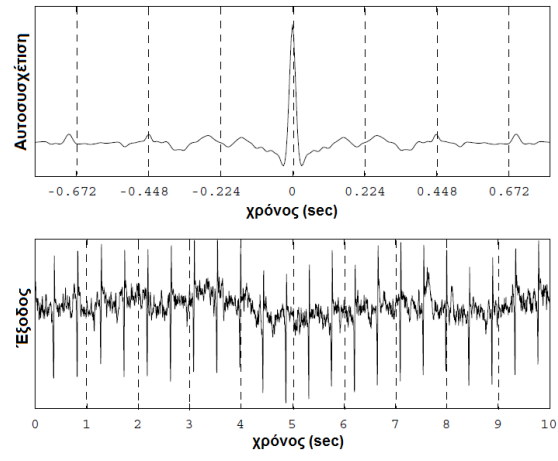
μ μ μ
 i
 $s(p) = E[x_j(t)x_j(t-p)]$ μ
 μ $s(p)$ μ
 $y = p^T s(p)$ μ
 $y = Wx = WAs, y = DPs,$ p
 μ μ μ , μ μ

μ 4 μ . 1-
 -1μ (μ μ):
 $1=113, 2=15, 3=6, 4=11,$

μ μ μ , μ μ
 μ μ 100 μ μ
 μ μ $i (i=1,2,3,4)$
 μ μ μ μ
 μ μ $i-$ μ
 i μ μ
 μ μ μ . μ
 100μ : $= 0.05 \pm 0.02, \mu$
 $= 6.09 \pm 2.19.$

μ μ μ Daisy [38].
 μ μ $p.$ μ
 μ μ $s(p).$ μ 3.46,
 μ μ μ ,
 μ μ μ .
 μ μ $s(p)$ μ .
 $p,$ μ
 120 bpm (0.5 .)
 $RR \mu$. μ

μ 0.448, $p=112$ μ .
 μ
 μ , μ .

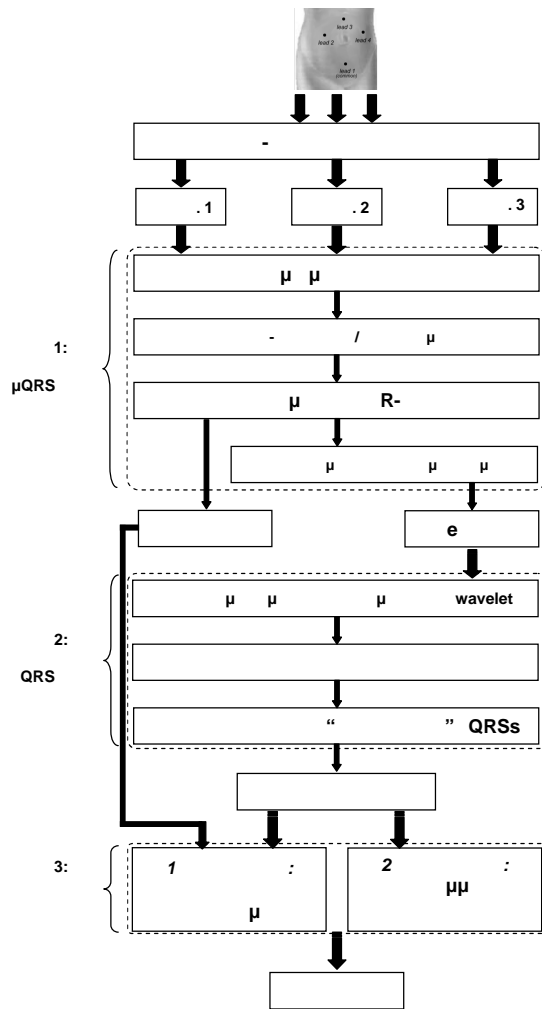


μ 3.46. () 1 μ *Daisy* .
 0.448. () μ μ

3.10. μ μ - μ wavelet μ μ μ μ

3.10.1.

μ μ μ [77] μ μ μ
 μ μ . μ
 μ μ ,
 μ R- μ μ (QRS),
 μ - (t-f) μ μ
 μ . , μ QRS
 μ μ . “ ”
 μ R- μ μ wavelets μ
 μ , μ R-
 μ μ QRS μ μ (μ μ
 μ QRSs) μ μ 2
 μ : μ μ μ
 μ μ . μ R- , μ
 μ μ . μ μ μ
 μ (8) μ (10), μ
20 41 μ . μ
97.47%. μ μ μ μ
 μ , μ μ
 μ μ μ μ .
 μ μ μ , μ ,
 μ 3.47, μ .



μ 3.47. 1 μ μ 3 .

3.10.2.

μ

- 1: μQRSs

μ μ μ μ μ , t-f
 μ μ , μ R-
 μ μ . μ
 () e μ (μ
 μQRSs) μ μ ().

$\kappa_{HK\Gamma}(t) = \sum_{i=1}^M lead_i(t) - \frac{1}{N} \sum_{t=1}^N \sum_{i=1}^M lead_i(t), \quad (1)$

$lead_i(t)$

(t-f): Wigner-Ville μ (WVD) [78-80]

μ - μ

$\mu\mu$ μ , μ μ

t-f. WVD

μ μ -WVD (Smoothed Pseudo WVD-SPWVD), :

$$SPWVD_x(t, \omega) = \int_{-\infty}^{+\infty} h(s) \left(\int_{-\infty}^{+\infty} g(\tau) x\left(t + \frac{\tau}{2}\right) x^*\left(t - \frac{\tau}{2}\right) e^{-j2\pi\omega\tau} d\tau \right) ds, \quad (2)$$

$x(\cdot)$ μ , t , ω $h(\cdot)$ $g(\cdot)$

μ μ τ

s , . SPWVD μ

(cross terms), μ . μ μ

SPWVD μ . μ 3.48() μ

μ 3.48() PSD .

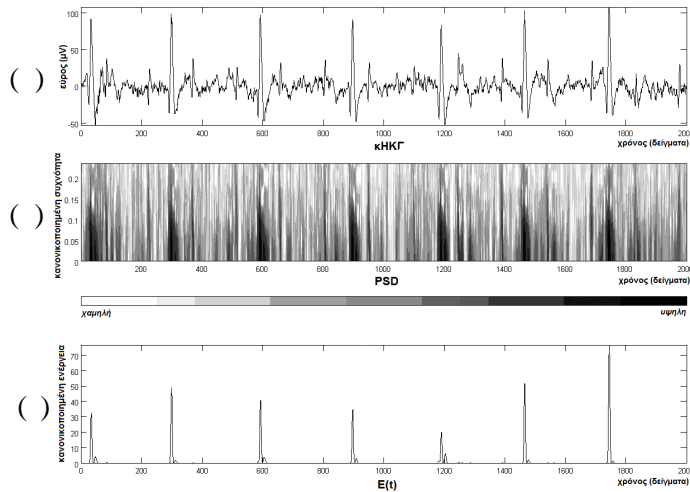
Hamming 64- μ , Hamming

32- μ . PSD μ

$E(t)$:

$$E(t) = \left(\int_{-\infty}^{+\infty} SPWVD_x(t, \omega) d\omega \right)^2. \quad (3)$$

$E(t)$ μ . 3.48().



μ 3.48. - : () μ , () PSD , ()
μ .

μ R- : μ R- μ
 $E(t)$: μ μ t_i , $E(t_i) > \theta$
 μ R- μ t_i . θ
 μ μ : $\theta = 10 \frac{1}{N} \sum_{t=1}^N E(t)$.

μ μ
 μ R- [62]. μ R- ,
 μ t_i t_j :
 1. $t_j - t_i < 0.2 \delta\epsilon\upsilon\tau$. μ R-
 μ $E(\cdot)$ (R- μ μ
 μ 0.2 sec μ).
 2. $t_j - t_i > 2 \delta\epsilon\upsilon\tau$. μ μ $E(\cdot)$
 μ R- , R- (μ μ
 μ μ 2 sec R-
).

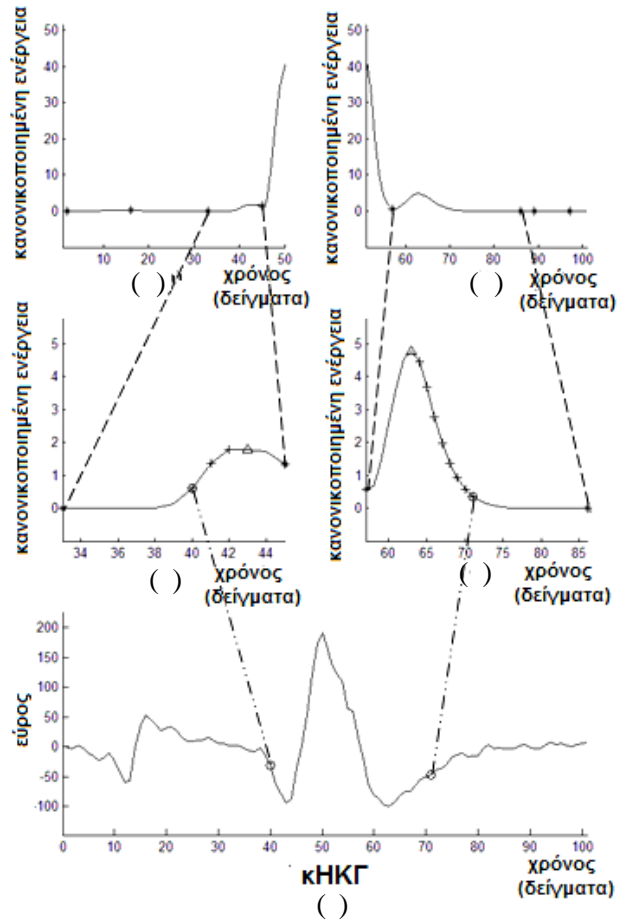
μ ,
 μ R- (1) μ μ
 R- (2). R-

μ R- , μ μ ,
 μ μ μ : μ Q μ (μ QRS
onset) S μ (μ QRS offset) , μ
 $E(t)$ R- , :
101 μ i^{th} μ R- (t_i) - $[t_{i-50}, t_i)$
 $(t_i, t_{i+50}]$ μ Q S μ ,
 $E(t)$ -
(μ 3.49() 3.49()). t_i μ
Q μ ($t_{Q_i}^{end}$),
 t_i μ (t_{Q_i}) μ $[t_{Q_i}, t_{Q_i}^{end}]$ (μ 3.49()),
Q μ ($t_{Q_i}^{start}$).
Q μ ($t_{Q_i}^{start}$) :

- | | |
|----|--|
| 1. | $t = \arg \max_{t \in [t_{Q_i}, t_{Q_i}^{end}]} E(t)$ |
| 2. | $\left(\left(\int_t^{t_{Q_i}^{end}} E(t) < 0.99 \int_{t_{Q_i}}^{t_{Q_i}^{end}} E(t) \right) \kappa \alpha t (t > t_{Q_i}) \right) t = t - 1$ |
| 3. | $t_{Q_i}^{start} = t$ |

μ μ μ $\left[t_{Q_i}, \arg \max_{t \in [t_{Q_i}, t_{Q_i}^{end}]} E(t) \right]$.
 μ μ $t_{S_i}^{start}$ $t_{S_i}^{end}$.
 μ t_i μ S μ
($t_{S_i}^{start}$), $t_{S_i}^{end}$ μ μ $[t_{S_i}^{start}, t_{S_i}^{end}]$ (μ 3.49()),
 t_{S_i} μ t_i μ .

1. $t = \arg \max_{t \in [t_{S_i}^{start}, t_{S_i}]} E(t)$
2. $\left(\left(\int_{t_{S_i}^{start}}^t E(t) < 0.99 \int_{t_{S_i}^{start}}^{t_{S_i}} E(t) \right) \text{ και } (t < t_{S_i}) \right) \quad t = t + 1$
3. $t_{S_i}^{end} = t$



μ 3.49. μ μ μ μ : $() E(t)$ -
 $[t_i - 50, t_i)$ μ $(\mu \mu)$. $()$
- $(t_i, t_i + 50]$ μ $(\mu \mu$
 $)$. $()$ $t_{Q_i}^{start}$ $(\mu ,$
 μ μ μ
 μ $t_{Q_i}^{start}$) . $()$ $t_{S_i}^{end}$ $(\mu ,$
 μ μ μ μ $(\mu \mu \mu$) .
 μ $t_{S_i}^{end}$) . $()$ $\mu \mu$ $\mu \mu$ $(\mu \mu \mu \mu$) .

$$\mu \left[\arg \max_{t \in [t_{S_i}^{start}, t_{S_i}^{end}]} E(t), t_{S_i} \right]. \quad \mu \quad t_{Q_i}^{start}$$

$$t_{S_i}^{end} \quad \mu \quad \mu \quad \mu \text{QRS} \quad (\quad \mu \quad 3.49()).$$

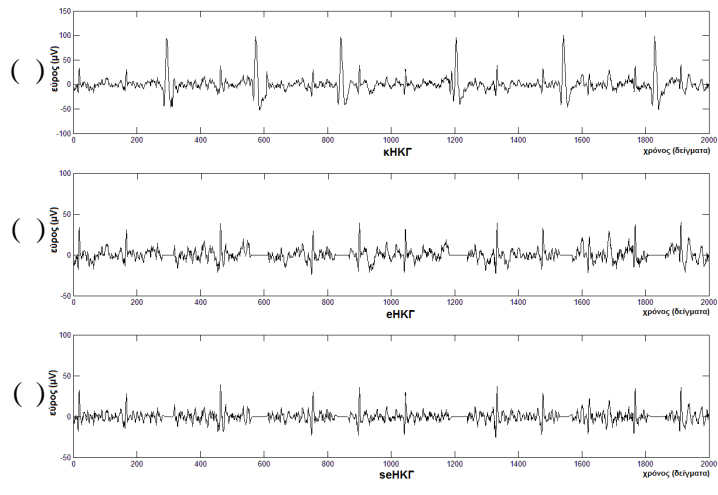
$$\mu \quad \mu \quad \mu \quad \mu, \quad \mu \text{QRSs} \quad :$$

$$\forall \mu \text{QRS}_i, \kappa \text{HK}\Gamma(t) = 0, \quad \forall t \in [t_{Q_i}^{start}, t_{S_i}^{end}]. \quad , \quad \mu \quad \mu \quad \mu \text{QRSs}$$

$$(\quad \mu \quad \mu \quad e \quad) \quad \mu \quad \mu \quad \mu \quad :$$

$$se \text{HK}\Gamma(t) = e \text{HK}\Gamma(t) - \frac{1}{11} \sum_{k=-5}^5 e \text{HK}\Gamma(t+k), \quad (4)$$

se $\mu \quad \mu \quad e \quad . \quad \mu \quad 3.50$
 $\mu \quad , e \quad se \quad .$



$\mu \quad 3.50. \quad \mu \text{QRS} \quad \mu \quad : () \quad \mu \quad \mu \quad \mu$
 $\mu \quad . () \quad \mu \quad \mu \quad \text{QRS} (e \quad \mu). () \quad \mu \quad \mu \quad e$
 $\mu (se \quad).$

- 2:** $\mu \quad \text{QRS}$
 $\mu \quad , \quad , \quad se \quad \mu \quad \mu$
 wavelet $\mu \quad \mu \quad (\text{CCWT}) \quad , \quad ,$
 wavelet $\mu \quad \mu \quad \text{R-}$
 $\mu \quad .$
 $\mu \quad \text{R-} \quad , \quad \mu \text{QRSs}$
 $(\quad 1).$

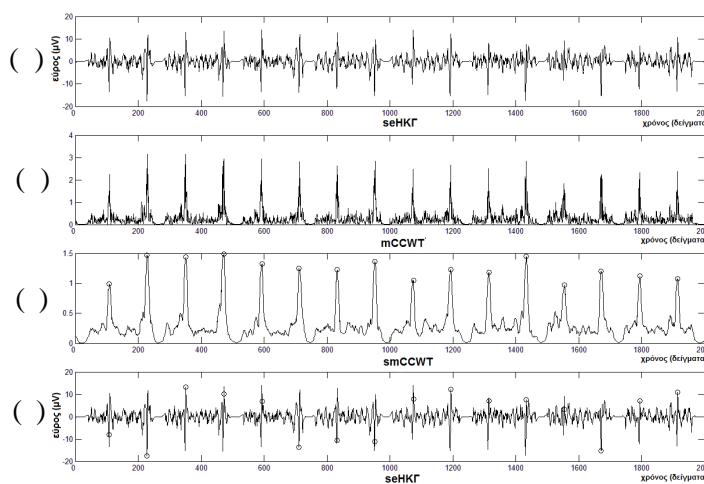
wavelet μ μ : wavelet μ μ μ ,
 (CWT) μ μ μ ,
 μ . μ wavelet ,
 μ μ μ μ μ
 wavelets μ wavelet
 μ μ μ (CCWT) [81,82] wavelet μ
 μ μ μ wavelets. μ wavelets
 μ , μ
 wavelet μ μ μ [83]. CCWT μ μ
 μ μ μ μ
 μ μ μ μ μ μ
 μ μ μ . μ CCWT
 μ μ , μ , μ μ .
 μ μ , μ μ B-spline
 wavelet [84], μ .
 μ B-spline wavelet :

$$\psi(x) = \sqrt{f_b} \left(\sin c \left(\frac{f_b x}{m} \right) \right)^m e^{2\pi i f_c x}, \quad (5)$$

m μ μ , f_b μ
 f_c wavelet. μ
 wavelet ($m=1$), μ μ
 $f_b=1$ $f_c=0.5$. μ R-
 μ CCWT . μ
 σ HK Γ μ μ μ μ
 . μ μ seHK Γ
 , μ μ μ
 [83]. μ μ R-
 μ , wavelet μ μ b-spline
 wavelet μ . μ μ μ
 μ wavelet (m CCWT) μ μ
 μ :

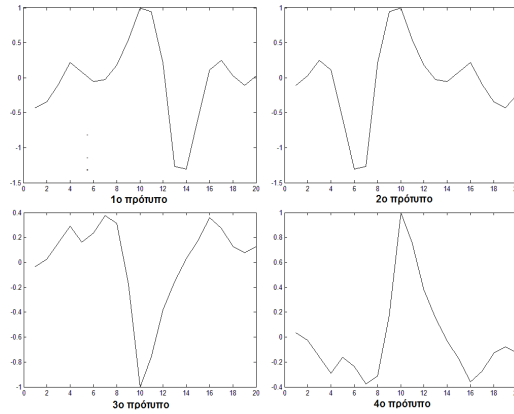
$$smCCWT(i) = \frac{1}{11} \sum_{k=-5}^5 mCCWT(i+k). \quad (6).$$

μ μ $mCCWT$ ($smCCWT$) μ
 μ μ μ μ μ
 R- μ μ μ μ μ μ μ 0.3 sec
 μ μ R- μ 3.51 μ $seHK\Gamma$,
 $mCCWT$, $smCCWT$ μ μ R- .



μ 3.51. *wavelet* μ R- : () μ
 se , () $CCWT$ ($mCCWT$), () μ μ $CCWT$
 ($smCCWT$) μ () μ se μ
 μ R- () .
 : μ R- ,
 μ μ , μ μ μ
 , QRS :
 μ μ μ
 QRS. ,
 QRS
 μ QRS (QRS) μ μ μ

3.52.



μ 3.52. 4 μ QRS (QRS).

μ 20- μ μ R-
 μ seHKΓ μ ,
 QRS: $\varepsilon QRS_i(l) = seHK\Gamma(t)$, $l = 1, \dots, 20$, $t = t_i - 10, \dots, t_i + 9$, t_i
 μ μ R- $i = 1, \dots, K$, K
 μ μ R- i ,
 μ : $mgn_i = \max_l(\varepsilon QRS_i(l)) - \min_l(\varepsilon QRS_i(l))$.
 μ μ R-
 μ : μ
 - μ εQRS_i QRS μ 0.6
 mgn_i 60% 200% μ mgn
 ($mean_mgn$), μ R- ,
 . μ 60% 200% .
 / μ R- , :

$$\left(\max_{j=1, \dots, 4} (C(\varepsilon QRS_i, \pi \varepsilon QRS_j)) > 0.6 \right) \quad \left(\frac{mgn_i}{mean_mgn} \in [0.6, 2] \right)$$

μ R-
 μ R- .

μ , :
 $C(x, y) = \frac{\langle x \cdot y \rangle}{\sqrt{\langle x \cdot x \rangle \langle y \cdot y \rangle}}$, μ μ μ
 μ μ μ $(x(t) \quad y(t))$. μ
 20 μ (. 66 msec μ 300 Hz)
 176

μ ; μ QRS μ
 μ 34μ , μ QRS
 μ 65 msec [85].

• 3:
 μ $R-$
 μ QRSs. : i.
 μ , ii. μ
 μ $\mu\mu$. μ $R-$
 $(\mu R-$), 1, μ $R-$
 $(R-$), 2. μ
 $(R-$ μ QRSs), μ μ
 $R-$ 2, μ μ

μ : μ μ μ
 μ $R-$. , μ μ RR μ
 (εRR) , $R-$. , μ RR μ (εRR_m) .
 μ RR μ (εRR_i) μ εRR_m . $\varepsilon RR_i > 1.5\varepsilon RR_m$
 μ $\mu\mu$ $R-$. ,
 150 msec μ εRR_i μ ,
 μ $\mu R-$, $\mu R-$
 μ $R-$.

μ $\mu\mu$: $R-$
 2 μ μ μ
 RR μ $\mu\mu$ (μ $3.53()$). $\mu\mu$,
 $(\mu \mu$
 μ $3.53()$ “ ” εRR μ (t_{RR}) . ,
 $2t_{RR}$ (μ μ μ $3.53() \mu$).

εRR μ μ μ $t \in \left[2t_{RR} - \frac{t_{RR}}{2}, 2t_{RR} + \frac{t_{RR}}{2} \right]$ “ ”

μ , . μ μ μ μ R- ,

εRR μ μ μ $t > 2t_{RR} + \frac{t_{RR}}{2}$ “ ” μ ,

. μ μ μ R- .

“ ” εRR μ εRR μ , μ

R- « » μ , μ μR-

. “ ” εRR μ

‘ , μ

. μ μ 3.53.

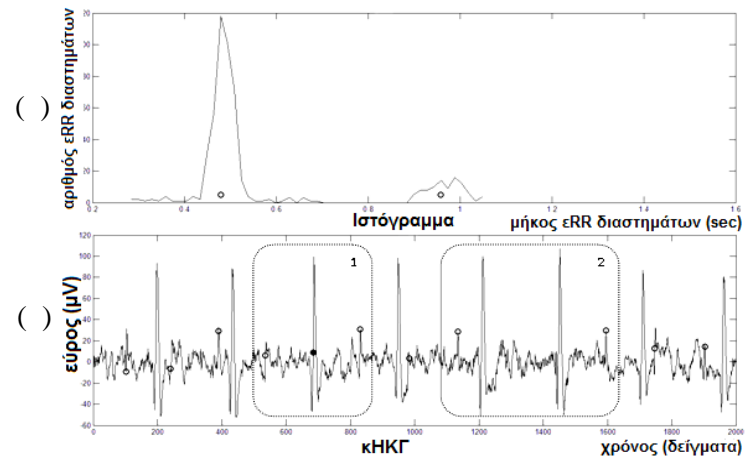
μ 3.53() μ μ μ , μ μ

μ QRSs (μ μ μ), μ 3.53()

μμ (. μ μ εRR μ

μ μ εRR μ μ μ

QRS μ 3.53()), .



μ 3.53. μ μ μμ : () μμ μ

RR μ . () 1. « » RR μ . 2. « » RR

μ . μ R- μ μ μ

R- μμ μ μ .

3.10.3. μ

μ μ μ μ

μ μ μ μ μ

Nottingham [15]. μ 8 μ (60 sec

(15 μ), 8 μ , 10 μ ,
 μ), 5 μ ,
 μ 20 41 μ .

3.10.4.

μ fR-
 μ : true positive
 (TP) μ fR- μ μ ,
 false negative (FN) μ μ fR- false positive
 (FP) artifact fR- . μ
 (sensitivity-Se),
 : $Se = TP / (TP + FN)$, (positive diagnostic value-PDV),
 : $PDV = TP / (TP + FP)$ (accuracy-Acc),
 : $Acc = TP / (TP + FP + FN)$. Se μ fR-
 μ , PDV μ μ fR-
 μ fR- μ μ , Acc
 μ μ μ μ μ
 μ μ [86].

3.3 3.4 μ
 μ μ , μ
 μ μ , μ 8 μ 10
 μ , .

3.3. μ μ δ μ - .

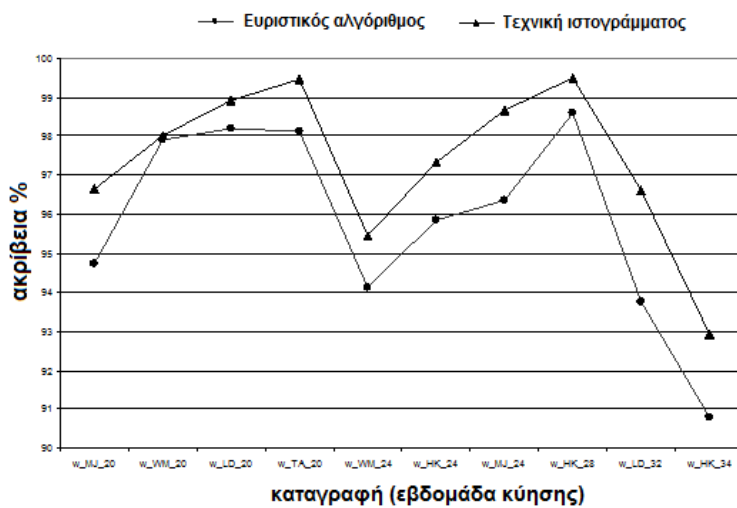
| | | | μ | | | | | | $\mu\mu$ | | | | | |
|-------|--------------|-------|-------------|----------|-----------|--------------|--------------|--------------|-------------|----------|----------|--------------|--------------|--------------|
| | | | TP | FP | FN | Se(%) | PDV(%) | Acc(%) | TP | FP | FN | Se(%) | PDV(%) | Acc(%) |
| μ | 24 | 1 min | 137 | 0 | 5 | 96.48 | 100 | 96.48 | 142 | 0 | 0 | 100 | 100 | 100 |
| μ | 26 | 1 min | 129 | 0 | 6 | 95.55 | 100 | 95.56 | 135 | 0 | 0 | 100 | 100 | 100 |
| μ | 29 | 1 min | 137 | 0 | 1 | 99.27 | 100 | 99.28 | 138 | 0 | 0 | 100 | 100 | 100 |
| μ | 35 | 1 min | 128 | 0 | 1 | 99.22 | 100 | 99.22 | 129 | 0 | 0 | 100 | 100 | 100 |
| μ | 37 | 1 min | 124 | 1 | 13 | 90.51 | 99.20 | 89.86 | 134 | 1 | 3 | 97.81 | 99.26 | 97.10 |
| μ | 39 | 1 min | 135 | 1 | 10 | 93.10 | 99.26 | 92.47 | 144 | 1 | 1 | 99.31 | 99.31 | 98.63 |
| μ | 40 | 1 min | 130 | 0 | 7 | 94.89 | 100 | 94.89 | 134 | 0 | 3 | 97.81 | 100 | 97.81 |
| μ | 41 | 1 min | 112 | 0 | 6 | 94.91 | 100 | 94.92 | 118 | 0 | 0 | 100 | 100 | 100 |
| | 8 min | | 1032 | 2 | 49 | 95.49 | 99.80 | 95.33 | 1074 | 2 | 7 | 99.37 | 99.82 | 99.19 |

3.4. μ μ 10 μ - .

| | | μ | | | | | | $\mu\mu$ | | | | | |
|----------------------|---------|-------|-----|-----|-------|--------|--------|----------|-----|-----|-------|--------|--------|
| | | TP | FP | FN | Se(%) | PDV(%) | Acc(%) | TP | FP | FN | Se(%) | PDV(%) | Acc(%) |
| w_HK_24 | 15 min | 1913 | 6 | 77 | 96.13 | 99.69 | 95.84 | 1938 | 1 | 52 | 97.39 | 99.95 | 97.34 |
| w_HK_28 | 15 min | 1903 | 5 | 22 | 98.86 | 99.74 | 98.60 | 1915 | 0 | 10 | 99.48 | 100 | 99.48 |
| w_HK_34 | 15 min | 1580 | 99 | 61 | 96.28 | 94.10 | 90.80 | 1624 | 107 | 17 | 98.96 | 93.82 | 92.91 |
| w_LD_20 ¹ | 10 min | 1254 | 4 | 19 | 98.51 | 99.68 | 98.20 | 1262 | 3 | 11 | 99.14 | 99.76 | 98.90 |
| w_TA_20 | 15 min | 1967 | 11 | 27 | 98.65 | 99.44 | 98.10 | 1987 | 4 | 7 | 99.65 | 99.80 | 99.45 |
| w_WM_20 | 15 min | 3055 | 10 | 55 | 98.23 | 99.67 | 97.92 | 3059 | 11 | 51 | 98.36 | 99.64 | 98.01 |
| w_WM_24 | 15 min | 1908 | 10 | 109 | 94.60 | 99.48 | 94.13 | 1935 | 10 | 82 | 95.93 | 99.49 | 95.46 |
| w_MJ_20 | 15 min | 2050 | 12 | 102 | 95.26 | 99.42 | 94.73 | 2094 | 15 | 58 | 97.30 | 99.29 | 96.63 |
| w_MJ_24 | 15 min | 2111 | 14 | 66 | 96.97 | 99.34 | 96.35 | 2162 | 14 | 15 | 99.31 | 99.36 | 98.68 |
| w_LD_32 | 15 min | 1730 | 26 | 89 | 95.11 | 98.52 | 93.77 | 1789 | 33 | 30 | 98.35 | 98.19 | 96.60 |
| | 145 min | 19471 | 197 | 627 | 96.86 | 98.91 | 95.84 | 19765 | 198 | 333 | 98.39 | 98.93 | 97.35 |

¹ μ , μ , μ μ TP, FN FP μ R- μ .

μ , μ μ Se 95.49%
 μ 96.86% μ
 . μ PDV 99.80% μ 98.91%
 μ , μ Acc 95.33%,
 μ 95.84% μ ,
 μ μ μ Se 99.37%
 μ 98.39% μ
 μ PDV 99.82% μ 98.93% μ
 . , μ Acc 99.19%, μ
 97.35% μ ,
 . μ μ 1.081 R-
 49 (4.53%) μ
 μ 7 (0.64%) μ μ ,
 2 artifacts (0.18%) R-
 μ 20.098 R-
 627 (3.12%) μ μ 333
 (1.66%) μ μ , 197 (0.98%) 198 (0.99%)
 artifacts R- , μ
 μ μ , μ
 μ 97.47%. Acc μ
 μ , μ μ , μ
 3.54.



μ 3.54. μ μ μ μ

3.11. $\mu \mu \mu \mu$ (3D phase space)

3.11.1.

[87], $\mu \mu \mu \mu$ $\mu \mu$

, , $\mu \mu$

μ . $\mu R-$ $\mu \mu$ μ

μ [88]. $\mu \mu \mu$ (QRS)

(3D phase space analysis) [89-91]. μ

μ (multivariate analysis) [92-94] μ

(μ) e .

, μQRS $\mu \mu$ R- . R-

$\mu \mu QRSs$, μ

. , , μ $\mu \mu$ R- .

μ , μ μ $\mu \mu$ [87],

μ 4.10, $\mu \mu$ μ

$\mu \mu$ R- , μ R-

R- ,

. $\mu \mu$ б $\mu \mu$

, μ μ μ

μ [95,96], SNRs.

, μ 13 μ

, μ μ , μ

μ .

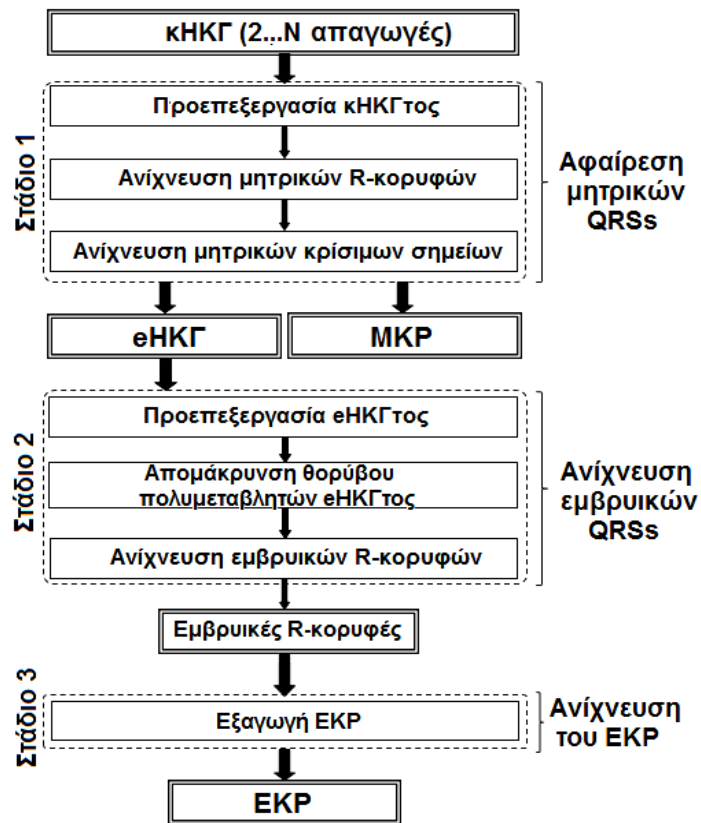
$\mu \mu \mu \mu$ $\mu \mu$ $\mu \mu$

BSS- μ , μ $\mu \mu$ $\mu \mu$

μ [15],

3.11.2.

μ $N \times M$, N μ
 μ M μ μμ ,
 μ () μ . μ
 , μμ μ μ
 μ 3.55.



μ 3.55. μμ 2 μ μ 3-

- I : $mQRSs$
 ()
 μ

, μ μ [88]
 . μ ()
 μ μ ()
); μ μ
 μ μ μ μ
 . μ
 μ μ FIR $\mu\mu$ - μ
 4-20 Hz [97], μ μ f .
 μ , μ μ P ,
 μ μ , μ QRS μ μ
 .
 μ R- : μ μ
 μ μ μ f , μ μ R-
 , μ μ [98].
 X μ μ 2 μ μ f μ μ
 μ : $f(i) \approx -a^2(l-k)+b$, k μ
 μ . μ μ a b , k ,
 μ μ :

$$V(a,b) = \sum_{l=k-[0.02*sf]}^{k+[0.02*sf]} w_l \left(f(i) - (a^2(l-k)+b) \right)^2$$
, sf
 μ w_l 's μ .
 k $(a(k), b(k))$. R
 μ : $ind(k) = a(k) * b(k)$,
 μ f μ k .
 μ μ μ : μ μ
 μ μ QRS, μ μ ,
 μ Poincare (3D $\mu\mu$).
 , (Δf) $(\Delta^2 f)$ f

μ . , $(f(i), \Delta f(i), \Delta^2 f(i))$ μ , μ
 3D $\mu\mu$. μ μ
 μ μ μ QRS μ ,
 μ , μ 3D $\mu\mu$
 , μ (μ 3.56)
 μ P , μ
 , μ
 μ QRS μ μ μ
 μ [95]. , μ
 [89]. μ 3D $\mu\mu$
 μ μ QRS .
 μ N , μ μ μ μ μ μ
 μ , : $\lambda\sigma$, λ
 $\lambda = \sqrt{2 \ln N}$ σ . μ
 Goring Nikola [89] Wahl
 [90].

μ
 μ μ μ
 μ μ . ,
 μ , μ μ μ
 gaussian μ μ μ .
 , μ μ (a_1, b_1, c_1)
 μ , . 3
 μ QRS
 - :

1. μ μ f μ μ μ : $f = f - \frac{1}{N} \sum_{i=1}^N f(i)$.

2. μ Δf $\Delta^2 f$: $\Delta f(i) = (f(i+1) - f(i-1))/2$
 $\Delta^2 f(i) = (f(i+2) + f(i-2) - 2f(i))/4$ i μ .

3. $\mu \quad \theta: \theta = \tan^{-1}((f \cdot \Delta^2 f^T)/(f \cdot f^T)).$

4. $\mu \quad :$

$$[f^T \quad \Delta f^T \quad \Delta^2 f^T] = [f^T \quad \Delta f^T \quad \Delta^2 f^T] \begin{bmatrix} \cos \theta & 0 & -\sin \theta \\ 0 & 1 & 0 \\ \sin \theta & 0 & \cos \theta \end{bmatrix}.$$

5. $\mu \quad \lambda: \lambda = \sqrt{2 \ln N}.$

6. $\mu \quad \sigma_f, \sigma_{\Delta f}, \sigma_{\Delta^2 f},$

$\mu \quad f, \Delta f, \Delta^2 f.$

7. $\mu \quad \mu \quad \mu \quad \mu \quad \lambda \sigma_f, \lambda \sigma_{\Delta f}, \lambda \sigma_{\Delta^2 f}.$

8. $\mu \quad a_1 \lambda \sigma_f, b_1 \lambda \sigma_{\Delta f}, c_1 \lambda \sigma_{\Delta^2 f},$

$\mu \quad (f(i), \Delta f(i), \Delta^2 f(i)) \quad (\quad .$

$$) : \frac{f(i)^2}{(a_1 \lambda \sigma_f)^2} + \frac{\Delta f(i)^2}{(b_1 \lambda \sigma_{\Delta f})^2} + \frac{\Delta^2 f(i)^2}{(c_1 \lambda \sigma_{\Delta^2 f})^2} > 1.$$

$\mu \quad \mu \quad \mu$

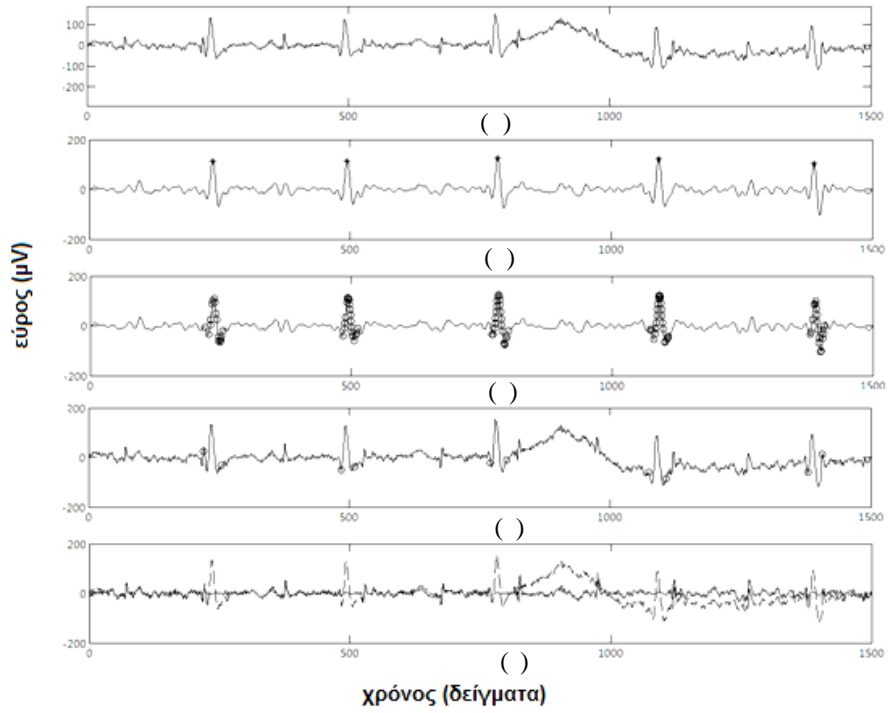
9. $\mu \quad \mu \quad \mu \quad \mu \quad 8.$

$\mu \mu \quad \mu \quad \mu \quad \mu$
 [91], $\mu \quad \mu \quad a_1, b_1, c_1 \quad \mu$
 $\mu \quad a_1 = b_1 = c_1 = 1 \quad \mu \quad (\lambda \sigma_f, \lambda \sigma_{\Delta f},$
 $\lambda \sigma_{\Delta^2 f}), \quad \mu, \mu$

$\mu \quad \mu \quad \mu \quad a_1 = b_1 = 1 \quad c_1 = 0.8. \quad 3D$
 $\mu \quad 3.56 (\quad .$

$\Delta^2 f \quad \mu \quad f \quad \Delta f).$

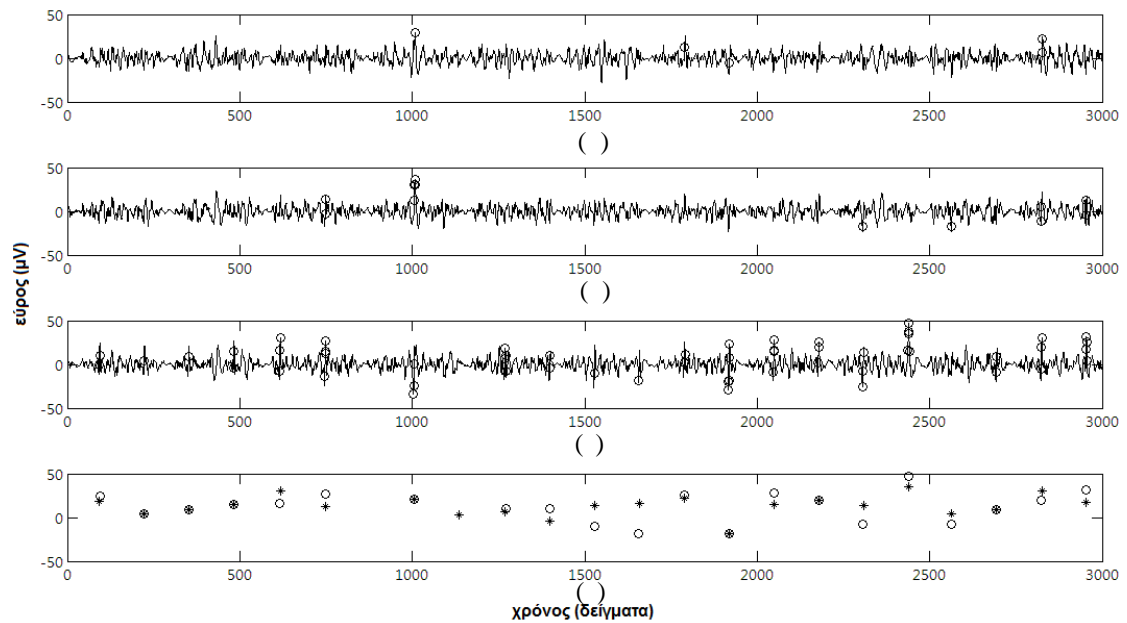
$\mu R-$
 $\mu QRS \quad \mu \quad (\quad \mu \quad \mu QRSs) \quad \mu$
 $\mu \quad \mu \quad \mu \quad \mu \quad \mu \quad .$



μ 3.57. μ 1 μ μ : ()
 μ , () μ μ , () μ μ_{QRS} μ , () μ
 μ μ μ () μ (μμ) e (μμ).

- 2: *QRSs*
 , , () e μ
 μ μ μ 2 μ :
 μ μ μ FIR μμ μ
 4-80 Hz [15,86,95,96,99] μ multivariate
 denoising μ . 2
 μ μ . 3.6
 μ R- μ μ .
 R-
 mQRSs, 1 .
 μ R- : ‘ ’ e μ
 R- , mQRSs.
 μ e

μ (μ) R- M μ
 : μ
 μ R- μ : ,
 20 μ , 2 μ
 μ μ μ R- μ
 20 μ (. 66 msec
 μ 300 Hz), μ μ QRS
 μ ;
 μ μ 34 μ ,
 μ μ QRS 65 msec [23,85]. μ
 μ R-μ
 μ 3.58.



μ 3.58. (, ,) QRS μ (μ μ)
 μ () μ μ
 μ μ (μ μ "ο")
 annotated μ R- (μ " +"), μ μ R-
 μQRSs.

QRS μ μ "+"

μ μ () , μ

μ R- (- μ μ μ $\mu\mu$

μ R-) $\mu\mu$ ()-().

μ "o") μ R- (μ "+"). μ

(a_2, b_2, c_2).

μ μ μ μ SNR

, μ μ μ .

- 3: μ μ QRSs
- μ R- μ μ QRSs.
- , μ R-
- μ QRSs) 3 (R- μ QRSs)

3.11.3. μ

μ μ μ μ μ μ

μ μ μ μ μ

[95], μ μ μ

μ Nottingham [15].

μ μ : μ μ μ

μ μ μ μ Sameni

et al. [95,96]. μ μ μ

μ , μ .

μ , μ

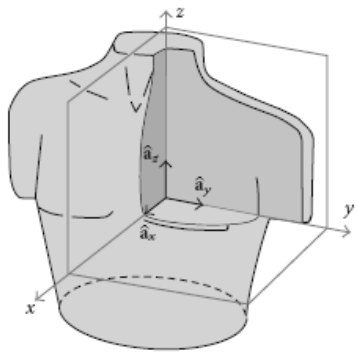
μ , μ -

[60],

$\hat{a}_x, \hat{a}_y, \hat{a}_z$

$d(t) = x(t)\hat{a}_x + y(t)\hat{a}_y + z(t)\hat{a}_z$,

3.59.



3.59.

(volume conductor)

[100,101],

$d(t)$

$v = a\hat{a}_x + b\hat{a}_y + c\hat{a}_z$

$\text{HKГ}(t) = \langle d(t), v \rangle = a \cdot x(t) + b \cdot y(t) + c \cdot z(t)$.

$$\hat{\theta} = \omega, \quad \dot{x} = -\sum_i \frac{a_i^x \omega}{(b_i^x)^2} \Delta \theta_i^x e^{\left[\frac{(\Delta \theta_i^x)^2}{2(b_i^x)^2} \right]}, \quad \dot{y} = -\sum_i \frac{a_i^y \omega}{(b_i^y)^2} \Delta \theta_i^y e^{\left[\frac{(\Delta \theta_i^y)^2}{2(b_i^y)^2} \right]}, \quad \dot{z} = -\sum_i \frac{a_i^z \omega}{(b_i^z)^2} \Delta \theta_i^z e^{\left[\frac{(\Delta \theta_i^z)^2}{2(b_i^z)^2} \right]} \quad [60]$$

о $d(t)$:

$$\hat{\theta} = \omega, \quad \dot{x} = -\sum_i \frac{a_i^x \omega}{(b_i^x)^2} \Delta \theta_i^x e^{\left[\frac{(\Delta \theta_i^x)^2}{2(b_i^x)^2} \right]}, \quad \dot{y} = -\sum_i \frac{a_i^y \omega}{(b_i^y)^2} \Delta \theta_i^y e^{\left[\frac{(\Delta \theta_i^y)^2}{2(b_i^y)^2} \right]}, \quad \dot{z} = -\sum_i \frac{a_i^z \omega}{(b_i^z)^2} \Delta \theta_i^z e^{\left[\frac{(\Delta \theta_i^z)^2}{2(b_i^z)^2} \right]}$$

$$, \quad \Delta \theta_i^x = (\theta - \theta_i^x) \bmod(2\pi), \quad \Delta \theta_i^y = (\theta - \theta_i^y) \bmod(2\pi), \quad \Delta \theta_i^z = (\theta - \theta_i^z) \bmod(2\pi),$$

$$\omega = 2\pi f, \quad f \text{ — частота,}$$

$$\hat{\theta}, \dot{x}, \dot{y}, \dot{z} \text{ — производные}$$

$$\mu \text{ — параметр}$$

$$\mu \text{ — параметр}$$

$$\mu \text{ — параметр}$$

$$\mu \text{ — параметр}$$

$$\mu \text{ — параметр}$$

$$\mu \text{ — параметр}$$

$$\mu \text{ — параметр}$$

$$\mu \text{ — параметр}$$

$$HK\Gamma(t) = H \cdot R \cdot \Lambda \cdot s(t) + W(t), \quad HK\Gamma(t)_{N \times 1}$$

$$s(t)_{3 \times 1} = [x(t), y(t), z(t)]^T$$

$$3 \quad \mu \quad d(t), \quad H_{N \times 3} \quad \mu$$

$$\mu, \quad \Lambda_{3 \times 3} = \text{diag}(\lambda_x, \lambda_y, \lambda_z)$$

$$\mu \quad \mu$$

$$x, y, z, \quad R_{3 \times 3}$$

$$\mu, \quad W(t)_{N \times 1} \quad N$$

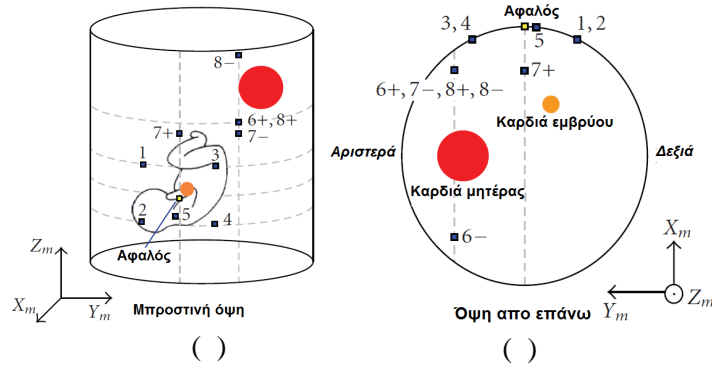
$$\mu \quad t. \quad H, R$$

μ μ μ ($d(t)$)
 μ , μ μ μ ,
 μ : $\kappa\text{HK}\Gamma(t) = H_m \cdot R_m \cdot \Lambda_m \cdot s_m(t) + H_f \cdot R_f \cdot \Lambda_f \cdot s_f(t) + W(t)$,
 $H_m, H_f, R_m, R_f, \Lambda_m, \Lambda_f$ μ μ
, μ m f
 μ μ , .
 μ μ μ μ μ μ
 μ μ μ ,
 μ μ . μ
MIT-BIH non-stress test μ (NSTDB) [105], μ
 μ μ - (μ)
 μ μ AR μ . μ μ μ
 μ μ μ μ μ , μ
 μ μ μ μ μ SNRs.
 μ μ - μ μ (μ , μ , μ ,
.) μ μ μ μ .
 μ μ , μ μ μ μ μ , μ
 μ () , μ 3.60() ,
 μ μ μ μ μ μ
 μ . μ μ μ μ μ
 μ $\theta_x, \theta_y, \theta_z$ R .
 μ μ μ ,
 μ ,
 μ 3.60() . μ μ
 μ μ μ μ μ [22,38,106]
 μ μ

μ

. μ , μ μ

μ μ .



μ 3.60.

μ μ , μ ()

μ

μ ()

μ μ μ μ .

μ μ μ , μ μ

, .

μ μ μ μ μ μ

μ 5 ,

5

. μ 5 ,

μ 300Hz. μ μ μ

μ μ - - (SNR): -5, -2, 0, 2, 5

10 dB.

, μ μ μ μ

μ .

μ : μ μ μ

μ μ Nottingham [15] μ μ 3

μ SNR μ μ μ μ

[77]. , μ μ μ μ 13

3 , 15 μ . μ

μ μ w_ (Patient_Code)_ (Gestation_Week). annotation

μ R- μ μ μ

3.11.4.

μ

, μ μ e e μ
 (multivariate denoising). X μ SNR μ :
 $eSNR = 10 \log \left(\frac{s^T s}{(x-s)^T (x-s)} \right)$, $eSNR$, μ SNR, x

μ s μ ($x-s$).

μ μ μ μ :

- μ , e 1 .

$$x = x_{initial} = eHKT \quad s$$

- μ μ , e μ μ
 $x = x_{initial}$ $s = x_{filtered}$

- μ , e μ

μ (multivariate denoising).

$$x = x_{filtered} \quad s = x_{filtered_denoised}$$

μ $eSNR$, μ μ

3.6 (μ),

μ $eSNR$ e

μ μ μ . μ μ

μ μ , μ μ μ ,

μ μ R- . μ R-

μ

μ 11 μ μ annotated R- .

, μ μ μ 6

μ μ μ μ

3.7. , μ μ μ

TP, FP, FN, Se, PDV $Acc.$ μ μ a_2, b_2 c_2

μ 1, μ

().

| SNR (dB) | μ | μ | μ |
|----------|--------|--------|-------|
| 1 (-5) | -31.08 | -17.88 | 7.854 |
| 2 (-2) | -28.08 | -17.49 | 9.81 |
| 3 (0) | -26.08 | -16.61 | 12.72 |
| 4 (2) | -24.08 | -16.14 | 7.53 |
| 5 (5) | -21.08 | -15.60 | 3.46 |
| 6 (10) | -16.28 | -12.34 | 4.92 |

μ μ Se, PDV Acc 97.29%, 71.28% 69.88%, .
 a_2, b_2, c_2
 $a_2 \in \{0.5, 0.6, \dots, 1.5\}, b_2 \in \{0.5, 0.6, \dots, 1.5\} c_2 \in \{1, 1.5, 2, 2.5\}.$
 $(a_2, b_2, c_2) = (1, 1.4, 1.5).$
 μ μ $Se,$
 PDV Acc 96.36%, 94.94% 92.02%, 3.9
 μ TP, FP, FN, Se, PDV Acc
 μ μ μ .
 μ μ μ $a_2, b_2, c_2,$
 μ (Acc).. μ
 μ μ [1.1, 1.5], [1.0, 1.5] [2.0, 2.5] $a_2, b_2, c_2,$
 μ μ a_2, b_2, c_2, μ
 μ annotation QRS μ μ
 μ μ μ μ
 μ $a_2, b_2, c_2,$
 μ 3.9. , μ
SNR μ μ μ μ $a_2, b_2, c_2,$
3.9, μ μ

μ μ a_2, b_2 c_2 , μ μ SNR

3.7. μ μ μ μ $(a_2, b_2, c_2) = (1, 1, 1)$ (μ μ).

| μ | | μ | | | | | |
|---------------|-------|-------------|-------------|------------|--------------|--------------|--------------|
| (SNR dB) | | TP | FP | FN | Se(%) | PDV(%) | Acc(%) |
| 1 (-5) | 5 min | 637 | 367 | 14 | 97.85 | 63.45 | 62.57 |
| 2 (-2) | 5 min | 633 | 284 | 18 | 97.24 | 69.03 | 67.70 |
| 3 (0) | 5 min | 614 | 229 | 37 | 94.32 | 72.84 | 69.77 |
| 4 (2) | 5 min | 637 | 248 | 14 | 97.85 | 71.98 | 70.86 |
| 5 (5) | 5 min | 637 | 211 | 14 | 97.85 | 75.12 | 73.90 |
| 6 (10) | 5 min | 642 | 211 | 9 | 98.62 | 75.26 | 74.48 |
| 30 min | | 3800 | 1550 | 106 | 97.29 | 71.28 | 69.88 |

3.8. μ μ μ μ $(a_2, b_2, c_2) = (1.0, 1.4, 1.5)$ (μ μ).

| μ | | μ | | | | | |
|---------------|-------|-------------|------------|------------|--------------|--------------|--------------|
| (SNR dB) | | TP | FP | FN | Se(%) | PDV(%) | Acc(%) |
| 1 (-5) | 5 min | 527 | 73 | 124 | 80.99 | 87.79 | 72.78 |
| 2 (-2) | 5 min | 646 | 50 | 5 | 99.3 | 92.76 | 92.16 |
| 3 (0) | 5 min | 642 | 41 | 9 | 98.59 | 93.96 | 92.72 |
| 4 (2) | 5 min | 646 | 14 | 5 | 99.3 | 97.92 | 97.24 |
| 5 (5) | 5 min | 651 | 9 | 0 | 100 | 98.61 | 98.61 |
| 6 (10) | 5 min | 651 | 9 | 0 | 100 | 98.61 | 98.61 |
| 30 min | | 3763 | 196 | 143 | 96.36 | 94.94 | 92.02 |

3.9. μ μ μ (μ μ).

| μ | | μ | | | | | | μ | | |
|---------------|-------|-------------|------------|------------|--------------|--------------|--------------|-------|-------|-------|
| (SNR dB) | | TP | FP | FN | Se(%) | PDV(%) | Acc(%) | a_2 | b_2 | c_2 |
| 1 (-5) | 5 min | 561 | 105 | 90 | 86.18 | 84.23 | 74.21 | 1.1 | 1.0 | 2.0 |
| 2 (-2) | 5 min | 643 | 11 | 8 | 98.77 | 98.32 | 97.13 | 1.1 | 1.2 | 2.0 |
| 3 (0) | 5 min | 648 | 5 | 3 | 99.54 | 99.23 | 98.78 | 1.2 | 1.3 | 2.5 |
| 4 (2) | 5 min | 648 | 4 | 3 | 99.54 | 99.39 | 98.93 | 1.3 | 1.3 | 2.5 |
| 5 (5) | 5 min | 650 | 1 | 1 | 99.85 | 99.85 | 99.69 | 1.4 | 1.4 | 2.5 |
| 6 (10) | 5 min | 651 | 0 | 0 | 100 | 100 | 100 | 1.5 | 1.5 | 2.5 |
| 30 min | | 3801 | 126 | 105 | 97.31 | 96.84 | 94.79 | | | |

3.10 $(a_2, b_2, c_2) = (1, 1, 1)$ Se, PDV

Acc 96.01%, 85.18% 82.3%,

$(a_2, b_2, c_2) = (1.1, 1.3, 1)$.

3.11.

25 32

3.10. $(a_2, b_2, c_2) = (1, 1, 1)$

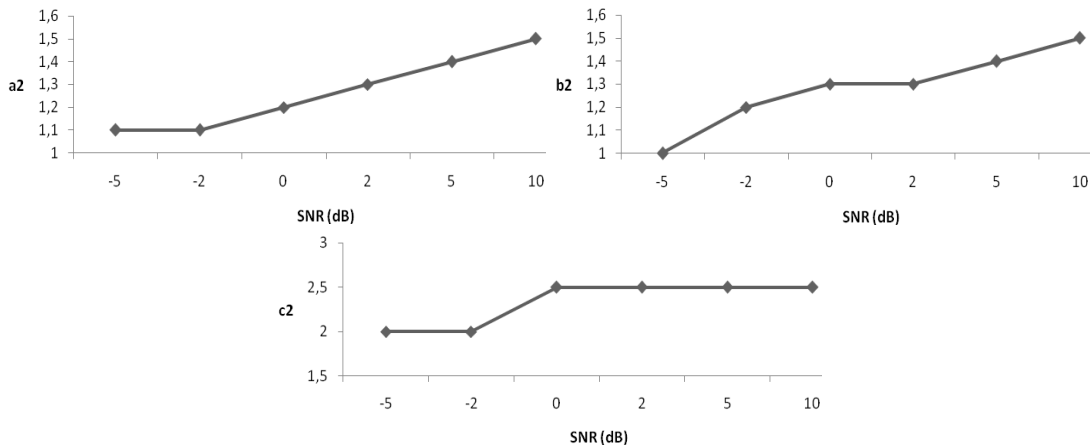
| (week) | μ | TP | FP | FN | Se(%) | PDV(%) | Acc(%) |
|----------------|--------|--------------|-------------|-------------|--------------|--------------|--------------|
| w_HK_24 | 15 min | 1975 | 202 | 31 | 98.45 | 90.72 | 89.45 |
| w_HK_28 | 15 min | 1860 | 394 | 70 | 96.37 | 82.52 | 80.03 |
| w_HK_34 | 15 min | 1595 | 280 | 46 | 97.20 | 85.07 | 83.03 |
| w_LD_20 | 15 min | 1526 | 336 | 61 | 96.16 | 81.95 | 79.36 |
| w_TA_20 | 15 min | 1902 | 747 | 95 | 95.24 | 71.80 | 69.31 |
| w_WM_20 | 15 min | 2833 | 553 | 312 | 90.08 | 83.67 | 76.61 |
| w_WM_24 | 15 min | 1968 | 397 | 80 | 96.09 | 83.21 | 80.49 |
| w_MJ_20 | 15 min | 2162 | 379 | 63 | 97.17 | 85.08 | 83.03 |
| w_MJ_24 | 15 min | 2122 | 567 | 78 | 96.45 | 78.91 | 76.69 |
| w_LD_32 | 15 min | 1802 | 216 | 49 | 97.35 | 89.30 | 87.18 |
| w_AW_20 | 15 min | 2043 | 135 | 70 | 96.69 | 93.80 | 90.88 |
| w_LG_24 | 15 min | 1892 | 217 | 148 | 92.75 | 89.71 | 83.83 |
| w_ME_20 | 15 min | 1947 | 178 | 37 | 98.14 | 91.62 | 90.06 |
| 180 min | | 25627 | 4601 | 1140 | 96.01 | 85.18 | 82.30 |

b_2 c_2 , μ μ a_2, b_2 c_2
 μ μ μ μ μ
 μ μ μ (3.9), μ
 μ μ SNR μ μ μ .
 μ μ 3.61.

3.11.

μ μ μ $(a_2, b_2, c_2) = (1.1, 1.3, 1.0)$ (μ μ).

| μ | μ | TP | FP | FN | Se(%) | PDV(%) | Acc(%) |
|----------------|--------|--------------|-------------|-------------|--------------|--------------|--------------|
| (week) | | | | | | | |
| w_HK_24 | 15 min | 1982 | 103 | 24 | 98.80 | 95.06 | 93.98 |
| w_HK_28 | 15 min | 1891 | 155 | 39 | 97.98 | 92.42 | 90.70 |
| w_HK_34 | 15 min | 1563 | 67 | 78 | 95.25 | 95.89 | 91.51 |
| w_LD_20 | 15 min | 1556 | 142 | 31 | 98.05 | 91.64 | 89.99 |
| w_TA_20 | 15 min | 1937 | 491 | 60 | 97.00 | 79.78 | 77.85 |
| w_WM_20 | 15 min | 3084 | 259 | 61 | 98.06 | 92.25 | 90.60 |
| w_WM_24 | 15 min | 2008 | 168 | 40 | 98.05 | 92.28 | 90.61 |
| w_MJ_20 | 15 min | 2119 | 91 | 106 | 95.24 | 95.88 | 91.49 |
| w_MJ_24 | 15 min | 2184 | 96 | 16 | 99.27 | 95.79 | 95.12 |
| w_LD_32 | 15 min | 1768 | 63 | 83 | 95.52 | 96.56 | 92.37 |
| w_AW_20 | 15 min | 1950 | 55 | 163 | 92.29 | 97.26 | 89.94 |
| w_LG_24 | 15 min | 1515 | 130 | 525 | 74.26 | 92.10 | 69.82 |
| w_ME_20 | 15 min | 1901 | 60 | 83 | 95.82 | 96.94 | 93.00 |
| 180 min | | 25458 | 1880 | 1309 | 95.04 | 93.37 | 89.00 |



μ 3.61. μ (a_2, b_2, c_2) SNR μ , μ
 μ μ μ μ μ .

$ecSNR = p_1 \cdot 10 \log \left(\frac{(s^T s)}{((x-s)^T (x-s))} \right) + p_2$, SNR :

p_1 SNR , x , s p_1 (4-80Hz)

p_2 (correction parameters).

$eSNR$ 4-80Hz.

p_1 p_2 SNR $eSNR$ s

(least mean squares). $ecSNR$

), SNR

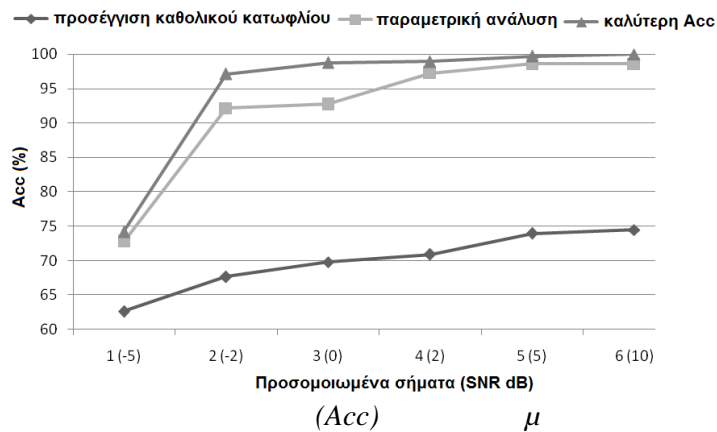
a_2, b_2, c_2 3.12 : SNR ,

$ecSNR$

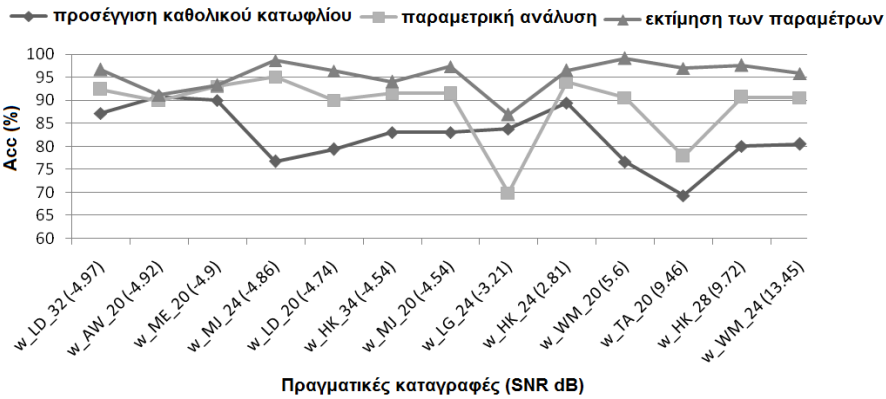
3.12. μ μ μ μ μ μ μ μ (μ μ).

| (week) | μ | μ | μ | | | μ | | | μ | | |
|----------------|--------|----------|--------------|------------|------------|--------------|--------------|--------------|-------|-------|-------|
| | | μ | TP | FP | FN | Se(%) | PDV(%) | Acc(%) | a_2 | b_2 | c_2 |
| | | SNR | | | | | | | | | |
| w_HK_24 | 15 min | 2.81 db | 1950 | 15 | 56 | 97.21 | 99.24 | 96.49 | 1.33 | 1.5 | 2.5 |
| w_HK_28 | 15 min | 9.72 db | 1920 | 36 | 10 | 99.48 | 98.16 | 97.66 | 1.49 | 1.5 | 2.5 |
| w_HK_34 | 15 min | -4.54 db | 1587 | 46 | 54 | 96.71 | 97.18 | 94.07 | 1.1 | 1.05 | 2.0 |
| w_LD_20 | 15 min | -4.74 db | 1570 | 41 | 17 | 98.93 | 97.45 | 96.44 | 1.1 | 1.03 | 2.0 |
| w_TA_20 | 15 min | 9.46 db | 1980 | 44 | 17 | 99.15 | 97.83 | 97.01 | 1.49 | 1.5 | 2.5 |
| w_WM_20 | 15 min | 5.6 db | 3136 | 18 | 9 | 99.71 | 99.43 | 99.15 | 1.41 | 1.5 | 2.5 |
| w_WM_24 | 15 min | 13.45 db | 2016 | 55 | 32 | 98.44 | 97.34 | 95.86 | 1.5 | 1.5 | 2.5 |
| w_MJ_20 | 15 min | -4.54 db | 2197 | 32 | 28 | 98.74 | 98.56 | 97.34 | 1.1 | 1.05 | 2.0 |
| w_MJ_24 | 15 min | -4.86 db | 2177 | 5 | 23 | 98.95 | 99.77 | 98.73 | 1.1 | 1.0 | 2.0 |
| w_LD_32 | 15 min | -4.97 db | 1823 | 34 | 28 | 98.49 | 98.17 | 96.71 | 1.1 | 1.0 | 2.0 |
| w_AW_20 | 15 min | -4.92 db | 2038 | 122 | 75 | 96.45 | 94.35 | 91.19 | 1.1 | 1.0 | 2.0 |
| w_LG_24 | 15 min | -3.21 db | 1970 | 227 | 70 | 96.57 | 89.67 | 86.90 | 1.1 | 1.18 | 2.0 |
| w_ME_20 | 15 min | -4.9 db | 1906 | 59 | 78 | 96.07 | 97.00 | 93.29 | 1.1 | 1.0 | 2.0 |
| 180 min | | | 26270 | 734 | 497 | 98.07 | 97.24 | 95.45 | | | |

μ
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μ 3.62.



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multivariate denoising)

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mQRSs), (ii) μ μ

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, (iii) μ 2 μ μ

(μ multivariate denoising)

(iv) μ 2 μ μ

μ μ

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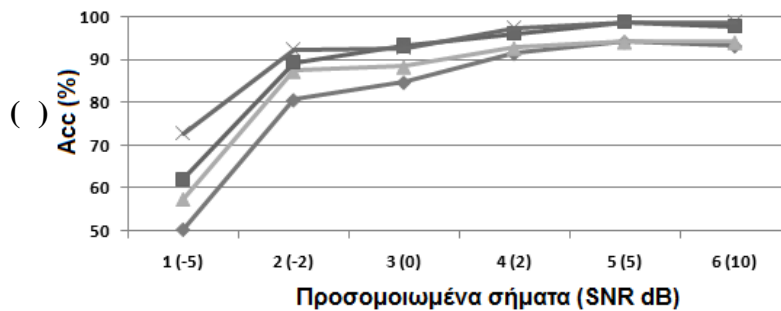
μ μ 3.64().

μ μ μ μ

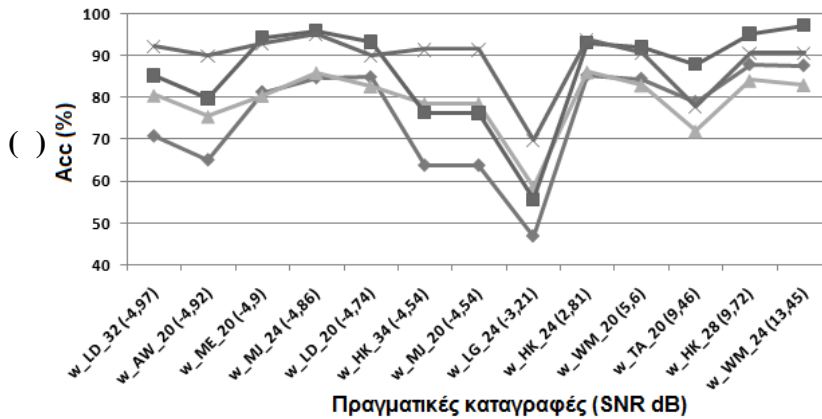
μ μ μ μ SNR. ,

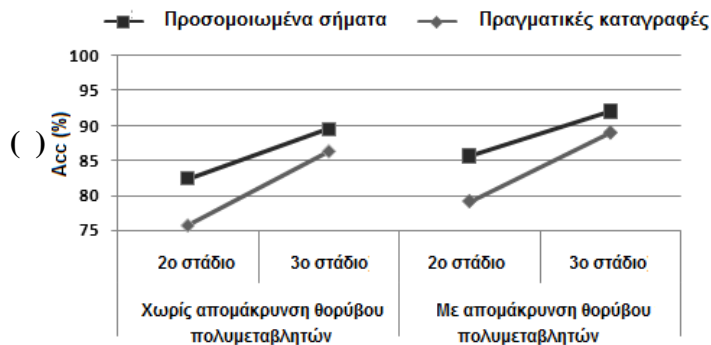
μ μ QRS

— Χωρίς απομάκρυνση θορύβου πολυμεταβλητών (2ο στάδιο) — Με απομάκρυνση θορύβου πολυμεταβλητών (2ο στάδιο)
 — Χωρίς απομάκρυνση θορύβου πολυμεταβλητών (3ο στάδιο) — Με απομάκρυνση θορύβου πολυμεταβλητών (3ο στάδιο)



— Χωρίς απομάκρυνση θορύβου πολυμεταβλητών (2ο στάδιο) — Με απομάκρυνση θορύβου πολυμεταβλητών (2ο στάδιο)
 — Χωρίς απομάκρυνση θορύβου πολυμεταβλητών (3ο στάδιο) — Με απομάκρυνση θορύβου πολυμεταβλητών (3ο στάδιο)





μ 3.64. Acc μ μ μ
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 μ μ ().

3.12. μ
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 μ μ 4
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i. μ μ [88] μ
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 20 ms, μ QRS μ μ μQRS
 [85]. μ μ μ
 μ RR μ [107].

ii. μ μ [107], μ
 μ μ . μ μ μ
 , μ , μ
 , QRS μ .

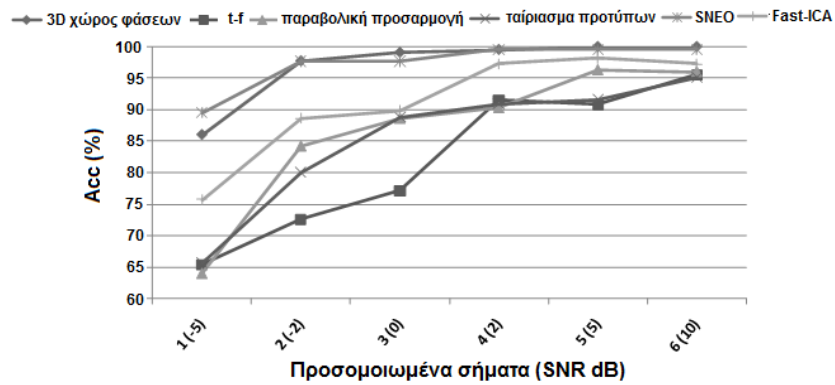
iii. QRS μ SNEO
 (Smoothed version Nonlinear Energy Operator) [108]. μ μ
 μ [108] μ μ μ
 FP FN μ .

iv. ICA μ . μ μ Fast-ICA μ
 [109], ICA μ .
 μ μ μ μ
 ICA μ μ μ
 . QRS μ μ μ

μ , μ μ 70%,
 μ μ QRS [77]. [108] μ
 μ .

t-f (1 μ) (2

μ) μ
 μ , μ , SNEO Fast-ICA
 e μ μ 3 μ . Fast-ICA μ
 e μ μ μ

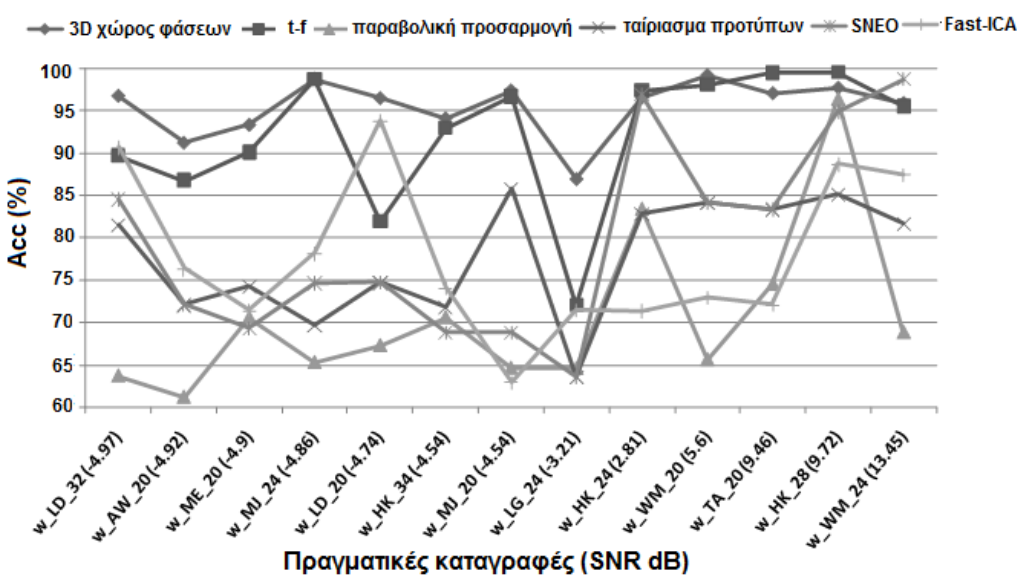


μ 3.65. μ μ (Acc)
 μ μ μ μ μ : t-f , μ
 μ , SNEO Fast-ICA.

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 μ μ μ μ 4 μ
 μ 3.65 3.66 μ μ μ μ , .

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 μ SNR. SNEO μ

2 μ μ μ (97.26% 97.06%,), t-f
 μ , μ , Fast-ICA
 μ μ . μ ,
 μ 2 μ μ (95.45%)
 μ . μ , μ 92.17%,
 70.47%, 80.23% 78.63% t-f μ , μ ,
 SNEO μ μ , .



μ 3.66. μ μ (Acc)
 μ μ μ μ : t-f , μ
 μ , SNEO Fast-ICA.

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3.13 μ) μ μ $\mu\mu$.
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 μ μ μ μ

Azad [18] μ μ 89% (μ
 $\alpha\pi \text{ } \delta\omicron\sigma\eta = 100 * (TP - FP - FN) / TP \%$), μ
 μ μ
 μ μ μ μ μ [87]
 90.1% μ μ μ 95.1% μ .
 Pieri *et al.* [15] μ μ μ μ μ
 μ 3.13 (400 5-10
 μ), μ μ (65%). μ Ibrahimy
et al. [17] μ 5
 20 , μ μ (89%)
 μ μ μ μ ;
 Doppler μ μ ;
 μ μ annotation . Martens *et*
al. [7] μ μ μ PCA μ (85%);
 μ μ μ μ 13
 μ , μ μ , μ 1
 [77] μ ,
 μ ; 2 μ μ [87]
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 μ μ . ,
 μ μ μ μ μ μ
 μ μ μ .

| 3.13. | 2 | μ | μ | μ | 5 | μ | μ | (Acc %) |
|-------------------------------------|-------|-------|-------|-------|----------|-------|------------|-----------------|
| Azad, 2000 [18] | | | | (3 | 5 |) | : | 89 ² |
| Pieri <i>et al.</i> , 2001 [15] | | | | (3 | 400 |) | : 5-10 min | 65 |
| Ibrahimy <i>et al.</i> , 2003 [17] | | | | (1 | 5 |) | : 20 min | 89 ¹ |
| Martens <i>et al.</i> , 2007 [7] | PCA & | | | (13 | 20 μ |) | : 30 min | 85 |
| Karvounis <i>et al.</i> , 2007 [77] | | | - | (3 | 8 μ |) | : 1 min | 99.19 |
| Karvounis <i>et al.</i> , 2008 [87] | | | | (3 | 10 μ |) | : 15 min | 97.35 |
| | | | | (3 | 13 μ |) | : 15 min | 95.45 |

1 μ μ $\mu\mu$ Doppler

2 μ : $\alpha\pi \delta\sigma\eta = \frac{TP - (FP + FN)}{TP} \times 100(\%)$.

3.13.

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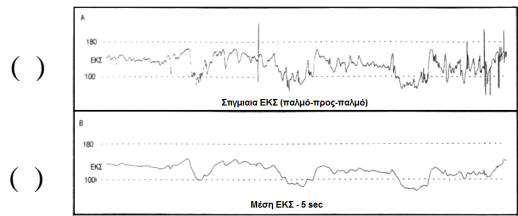
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4.1. μ

μ (μ) μ . μ , μ μ μ μ μ . μ μ , , μ μ μ . μ μ μ , μ μ μ . μ μ μ μ , μ μ μ . μ μ μ μ , μ μ μ μ . μ μ μ μ , μ μ μ μ . μ μ μ μ [1].

μ μ μ μ μ , μ μ μ μ μ , μ μ μ μ μ μ μ (msec), μ μ μ μ - μ μ - , μ μ μ μ (μ 4.1).

μ - μ (beat-to-beat), μ μ μ μ μ μ μ μ μ μ μ . , μ μ μ μ μ μ μ μ μ μ μ μ .



μ 4.1. μ μ () μ μ μ : () μ , μ μ μ μ (μ μ) . () μ μ μ 5 μ μ μ .

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4.2. μ μ μ μ
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 Roederer «Elementa artis obstetriciae» μ
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 1818, μ
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 μ (fetal distress).

μ μ (Pestalozza 1892, Seitz 1903, Hofbauer & Weiss 1908).
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Wohlgemuth 1961. Hon

1963, Hon

Hon Hammacher 1962

Doppler (Callagan 1964, Bishop, 1966,1968, Kratochwil & Eisenhut 1967, Barton 1968, Brown & Robertson 1968, Bernstine 1968). Mosler (1969). (Mosler 1970, Mosler 1971, 1972).

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- 4.3. μ ()
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μ [2].
1. ()
 2. μ μ μ ()
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 5. μ
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 7. Doppler μ μ
 8. 0 μ μ μ
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4.5.

4.5.1.1.

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μ [6-8]. , μ
μ μ μ μ μ [9,10].
, μ μ
« » [11], μ
μ [12].
, μ μ
μ ("conversion - sinusoidal patterns")
[9,12], μ μ [13].

4.6. μ
μ («fetal blood sampling» -
FBS) μ μ
μ μ μ («standard of care»)
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μ Saling 1962 [14]. μ
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μ [16,17]. μ μ

μ (2Hb) μ 2Hb μ Hb
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$$\text{Κορεσμός}(\%) = \frac{[O_2Hb]}{[O_2Hb][Hb]} * 100$$

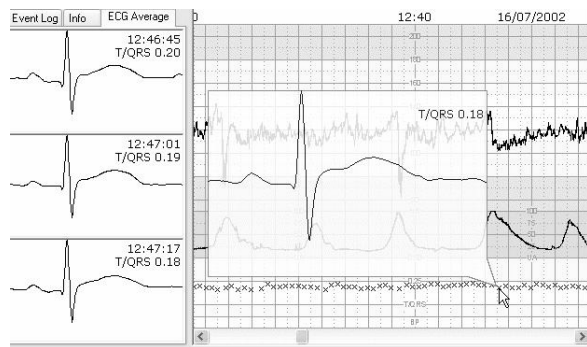
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 [19-22],
 «managing» [23] μ .

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 μ [25], μ μ μ
 μ « μ » [26]. ,
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 ACOG Committee Opinion μ 2001, μ μ
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 μ [28].

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 μ μ [29].

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(μ 4.5). μ , μ
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ST T/QRS [31].
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 μ T/QRS.

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[32]. μ μ μ
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4.8. $\mu \mu$

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2-6 , μ (variability).

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 μ (macrofluctuation - long term variability - frequency of variability)

$\mu \mu \mu$
 μ (amplitude of variability).

4.8.1.1. μ

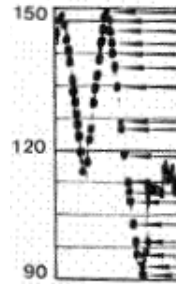
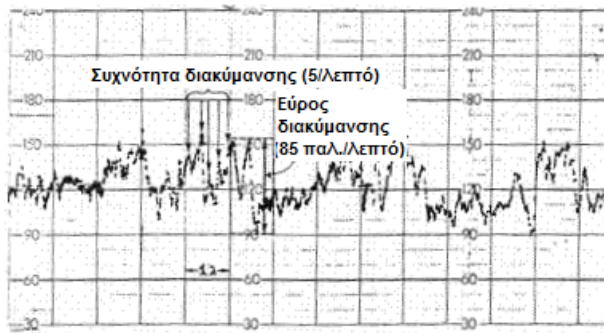
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(μ 4.8). $\mu \mu \mu$
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(microfluctuation-short term

variability).



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 μ (. 63):

- (silent) [μ < 5 μ]
- μ μ [μ 5-10 μ]
- μ [μ 10-25 μ]
- μ [μ > 25 μ]

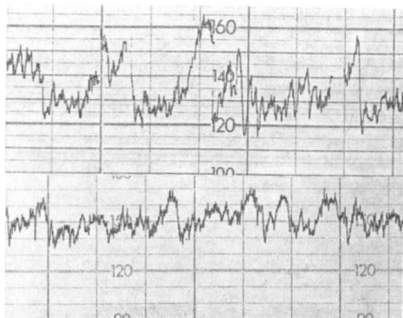
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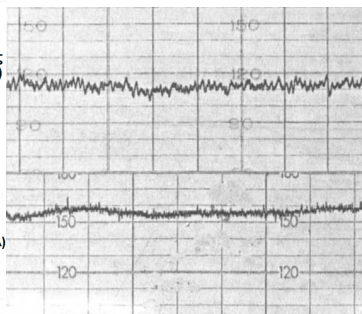
μ

μ



αλματωειδής τύπος
(εύρος > 25 παλ/λ)

κυματοειδής τύπος
(εύρος 10-25 παλ/λ)



εσπενωμένος
κυματοειδής τύπος
(εύρος 5-10 παλ/λ)

σιωπηρός τύπος
(εύρος < 5 παλ/λ)

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Hammacher [1].

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 (μ -amplitude of variability) μ
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4.9. μ

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- - μ μ ,
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4.10.

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 μ (floatingline)
(baseline),
(accelerations) (decelerations). μ
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 μ (Goeschen).

4.10.1.1.

(Accelerations)

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 , μ μ Goeschen, μ .

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4.10.1.2.

(Decelerations - Dips)

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 (Goeschen).
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American College of Obstetricians and Gynecologists) [55],

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- μ (< 100 bpm 3) (μ
μμ > 160 bpm).

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[58]. μ μ Low *et al.* [59]

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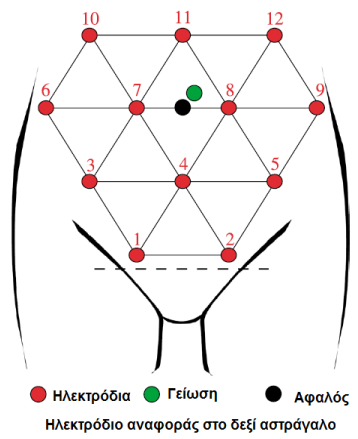
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4.11.

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Taylor *et al.* [63,64] μ
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 Hammersmith Hospitals NHS Trust, 241
 μ (15-41 μ), 58 μ (16-35 μ) 5 μ
 μ (20-33 μ). (12 μ 16) μ μ

(μ 4.10).



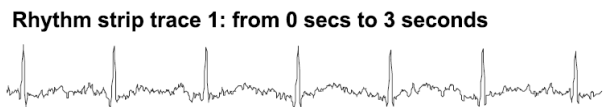
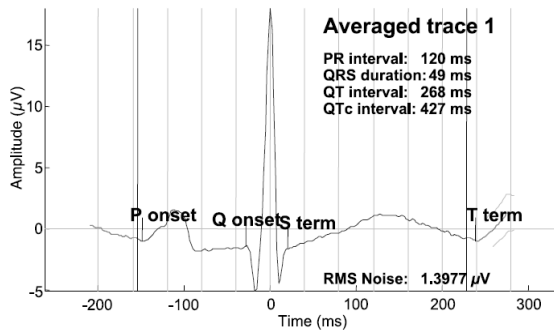
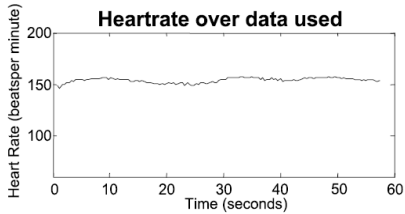
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16-bit
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 1-2 Hz
 150 Hz
 15-20
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 60 sec
 QinetiQ
 3-5
 4.11.

Fetal ECG Report

Name: QC50 **Date:** 18/7/2001
Singleton **Gestation:** 35 weeks
No. of sensors: 12 **Length of dataset:** 57.9 seconds
Data used: 57.9 seconds - From 0 to 57.9 seconds
Passband: from 1 to 150 Hz
Mean heartrate: 153 bpm
Max heartrate: 157 bpm
Min heartrate: 146 bpm

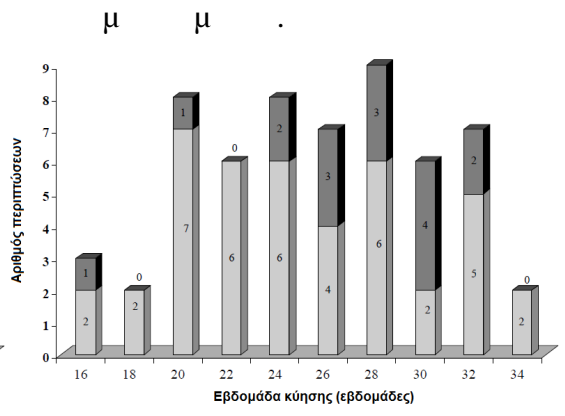
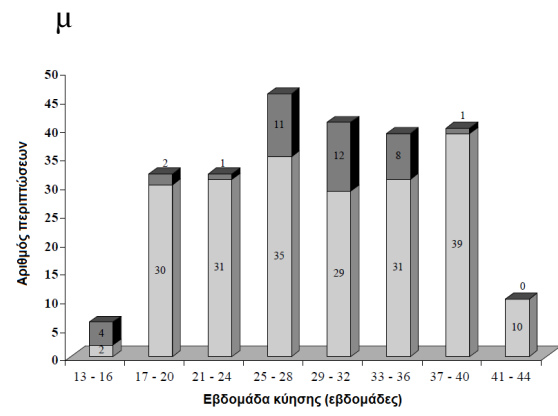


4.11. 35
 60 sec.

PR, QRS, QT μ μ μ μ μ QT (QTc)
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 μ μ μ
 μμ μ
 QRS μ μ . μ
 μ μ - - μ .
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 , μ μ , 120
 μ 150 μ . μ
 μ μ μ . ,
 μ , μ 10 μ
 μ , μ 40 .

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 , μ μ μ μ PR,
 QRS, QT and QTc μ . predicted
 μ (20 , 30 40 μ)

μ μ . 241 μ μ 250
 μ 4.13 μ



μ 4.13. μ μ () () () μ
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250 , μ μ
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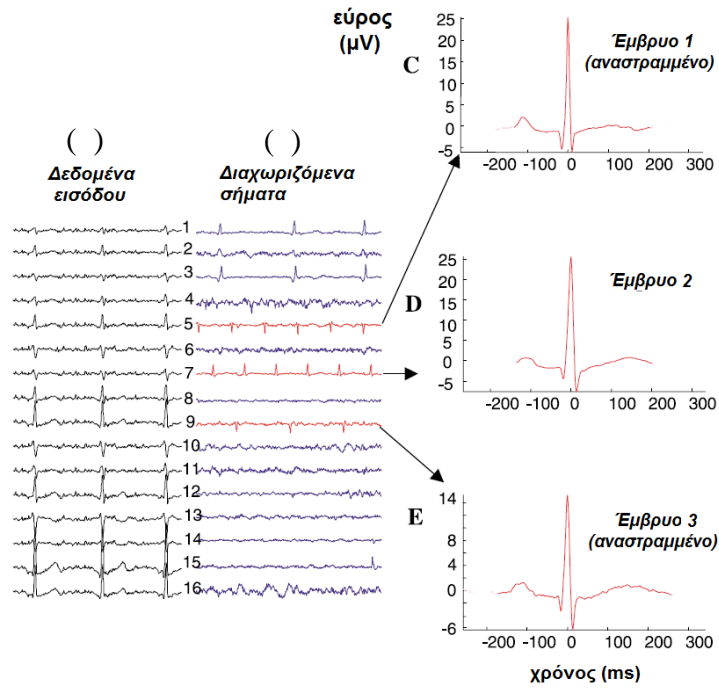
μ μ -
 - μ
 (37/250) , μ μ . 15%
 μ μ 27 36 μ , μ 84% (31/37)
 (5/250) μ μ .
 μ .

199

μ PR QRS 199
 μ QT/QTc 156. μ P, Q, R S
 μ μ . 22% (43/199)
 , μ , 63% (27/43)
 ≤ 24 μ .
 μ , μ 24 μ
 μ PR QRS μ .
 μ μ μ .
 μ , μ μ s_j μ
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 μ μ (μ 4.14).

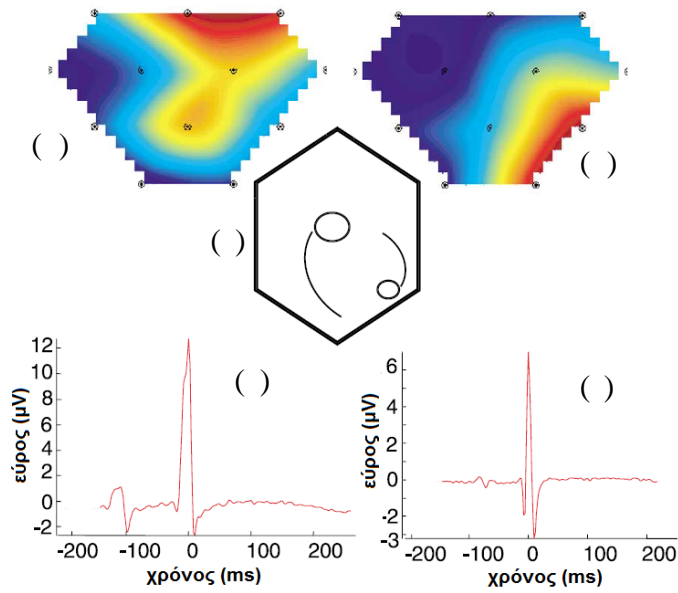
μ , μ , μ
 16% (9/58), 12% (7/58) 72% (42/58), , μ μ
 μ μ 24 32 μ .
 , μ μ 78% μ (91/116).
 μ P, Q, R S μ μ μ

59% (54/91). μ , μ μ 93% μ
 (14/15). μ P, Q, R S μ μ
 μ μ 57% (8/14).



μ 4.14. μ () μ μ () μ μ
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μ 4.15. μ

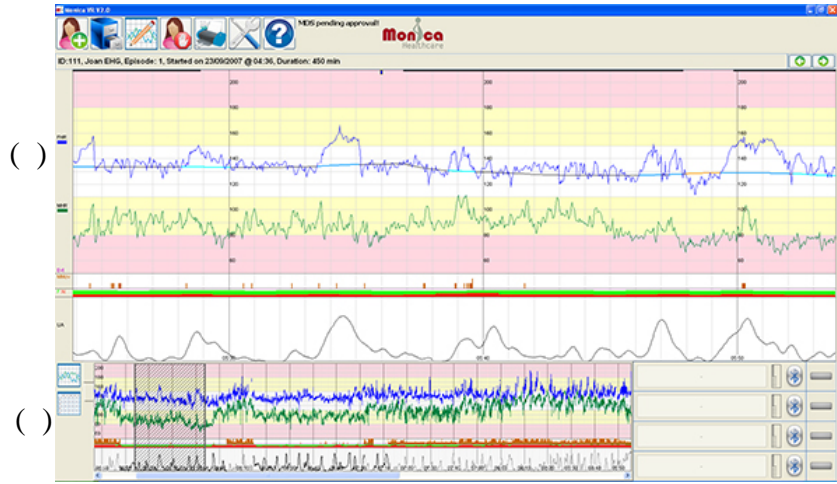
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Monica Healthcare Ltd. [65-67],

Pieri *et al.* [52],

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μ 4.16. () μ μ μ (μ), (), μ
 (), μ (μ) () .

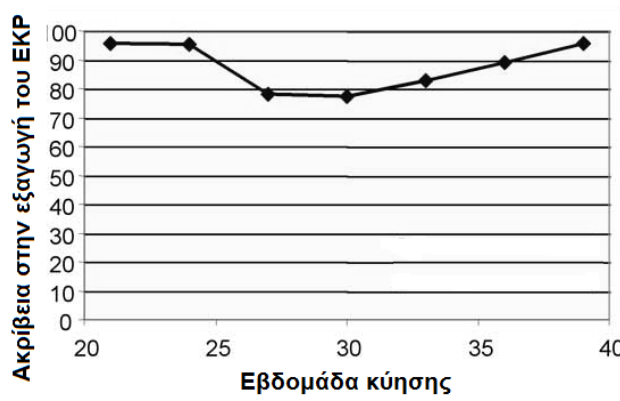
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 9-13 μ 2007 . Margo Graatsma [68]. Monica AN24
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 2- μ , μ μ (μ
 4.17). μ
 78%.



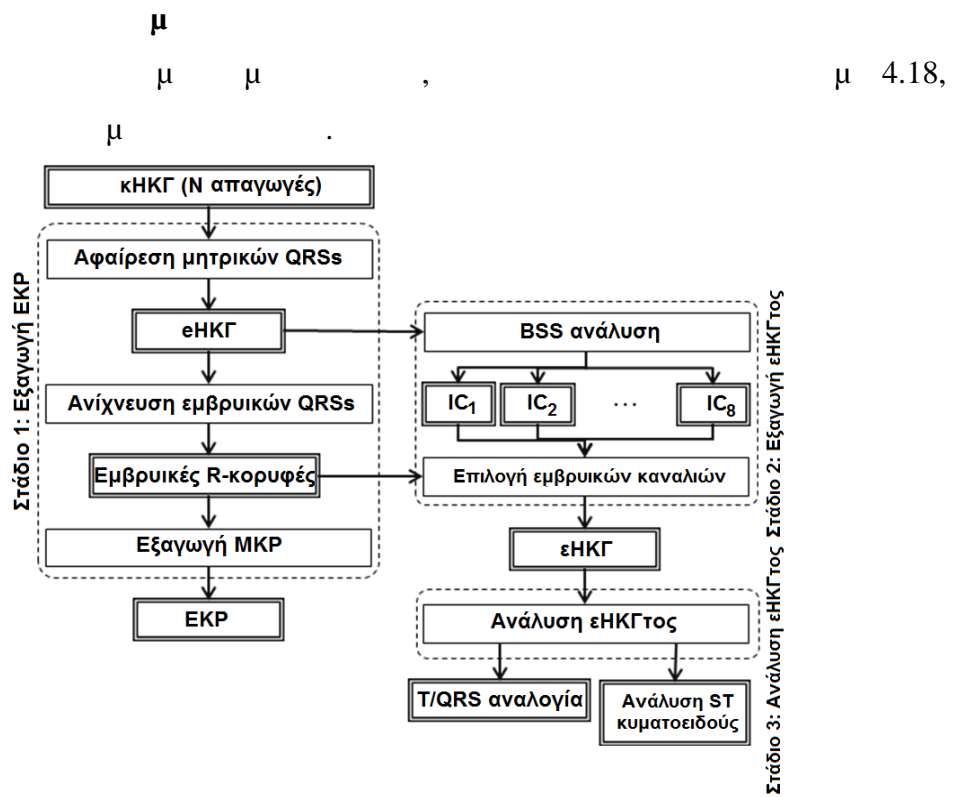
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4.12.2.



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 $\mu \quad R-$, $\mu \quad \mu$, $\mu QRSs$
 μ [77].
 μ
 e .

QRS: $\mu \quad R-$
 $e \quad (\mu \quad \mu) \quad \mu$: i)
 $\mu \quad 4-80 \text{ Hz}$ [78] ii) multivariate
 denoising [79]. μ
 [77] e , $\mu \quad R-$

$\mu QRSs.$, μ
 $\mu \quad \mu \quad a_2, b_2 \quad c_2 \quad a_2 = 1.1, b_2 = 1.3$ and
 $c_2 = 1.5$ ($\mu \quad \mu \quad a_2, b_2 \quad c_2$
 $2 \quad \mu \quad \mu$);
 $\mu \quad \mu \quad \mu \quad \mu$
 $\mu \quad \mu \quad \mu \quad \mu \quad R-$.

μ , $\mu \quad \mu \quad R-$,
 $\mu \quad \mu \quad \mu \quad \mu \mu$ [77]. μ

$\mu \quad \mu \quad \mu \quad \mu \quad R-$

- 2:**
 $e \quad e$,
 $\mu \quad \mu \quad \mu \quad \mu$ EFICA
 (Fast-ICA) [80], $\mu \quad \mu \quad \mu$ Fast-ICA
 μ , $\mu \quad 8 \quad (IC_1-IC_8)$. EFICA
 μ ,

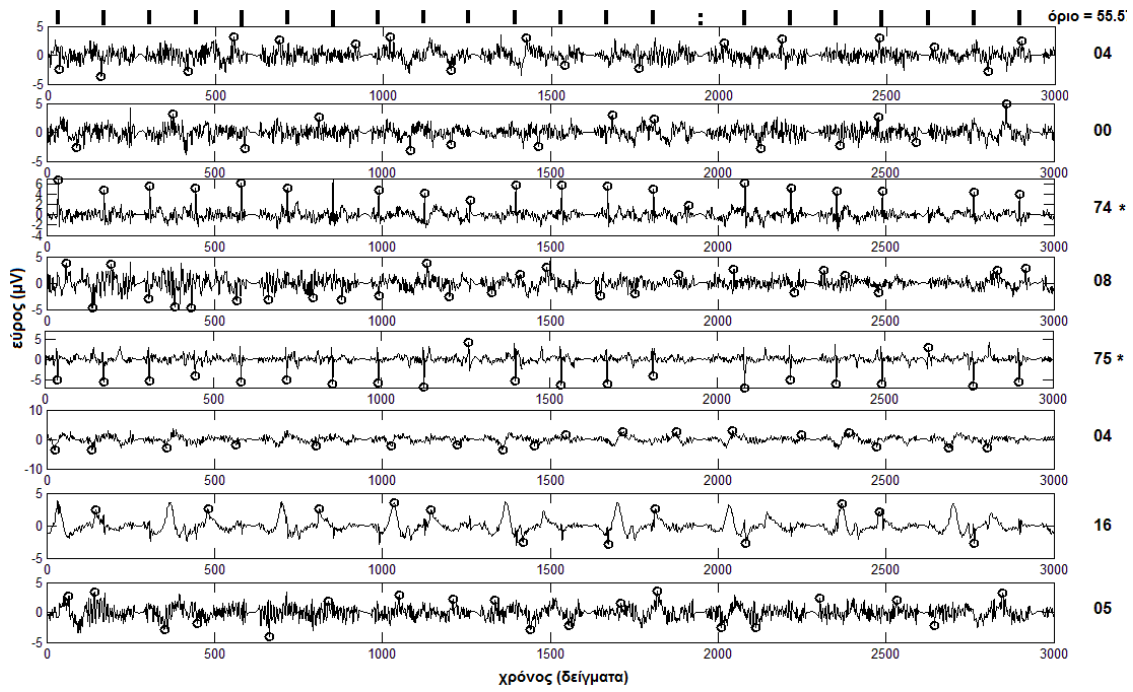
μ SNR (μ μ μ) μ 1 2
 μ (ICs).
 μ QRS [81] μ IC,
 μ (IC).
 μ QRSs μ μ μ .
 μ (1). μ μ μ
 μ IC, μ (cor) μ IC
 μ . μ μ , μ μ μ
 μ (standard deviation). ICs
 μ (fICs) :

$$\rho_{IO} = cor(IC_i, EKP) > \frac{1}{N} \sum_{i=1}^N cor(IC_i, EKP) + \frac{1}{N} \sum_{j=1}^N \left| cor(IC_j, EKP) - \frac{1}{N} \sum_{i=1}^N cor(IC_i, EKP) \right|$$

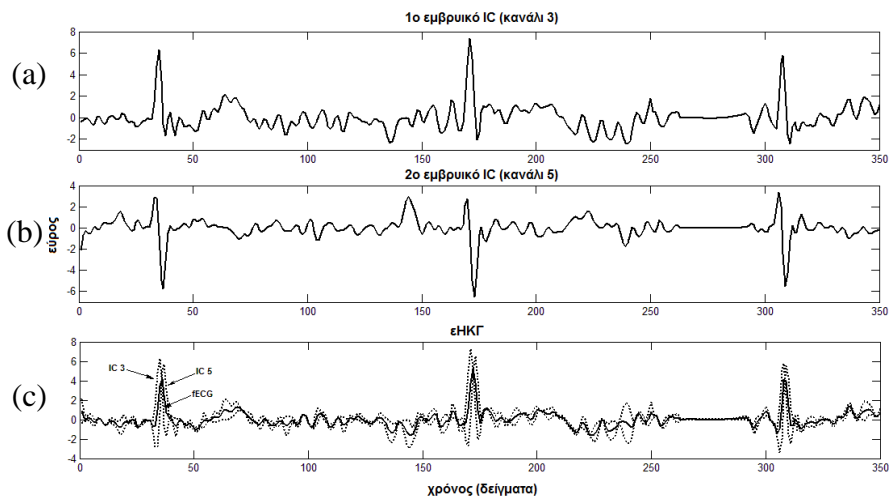
$$cor(IC_i, EKP) > \rho_{IO}$$

IC_i μ μ .

$cor(a, b)$ μ μ μ a b . μ
 μ TP R- , . μ IC
 μ R- . IC TP
 μ (20 μ) μ μ μ μ R- ;
 μ 20 μ μ μ
 μ QRS 65 msec, 20 μ μ
 μ 300 Hz [82,83]. μ μ
 μ (ICs EFICA μ
 μ fICs) μ 4.19.



μ 4.19. ICs μ μ EFICA e μ . μμ
 , “ο” ICs. μ “84”
 μ μ R- μ “04”, “00”... “05”
 μ TP R- . “*” fICs.



μ 4.20. μ μ : () μ
 IC (3), () μ IC (5)
 () μ .

μ fICs,
 (5 IC μ 4.19) (.

μ) μ μ
 fICs (μ μ):

$$\varepsilon_{HK\Gamma}(t) = \frac{1}{M} \sum_{i=1}^M fIC_i(t),$$

M μ μ fICs t μ .
 μ 4.20.

• 3:

μ
 T/QRS ST μ .
 μ 2.7 μ
 μ [82,83], μ

R- (fR), μ μ PQRST (fPQRST)
 μ μ 30 fPQRST μ . μ
 μ fPQRST μ 330 msec
 100 μ [(fR_i-100) msec - (fR_i +230) msec], fR_i ith fR.
 μ fPQRST T/QRS
 μ μ ST. μ μ
 μμ . μμ :

$$isoelectric = \frac{1}{21} \sum_{t=fR_i-45}^{fR_i-15} fPQRST(t)$$

μ fPQRST.

μ μ
 μ μ QRS
 (T/QRS). QRS :

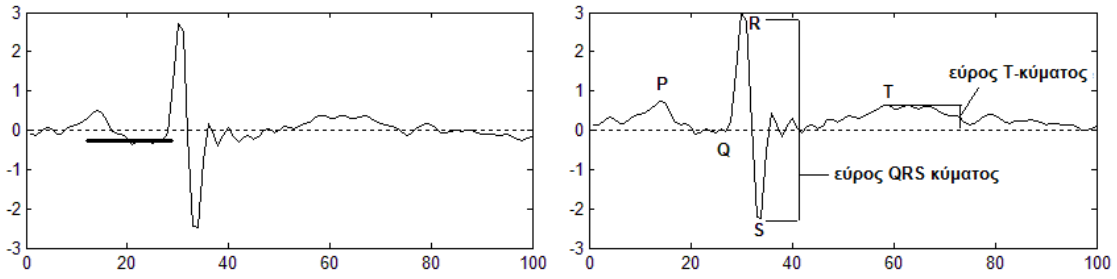
$$QRS_{amp} = \max_{t \in [fR_i-10, fR_i+10]} fTQRS(t) - \min_{t \in [fR_i-10, fR_i+10]} fTQRS(t),$$

μ

$$: T_{amp} = \max_{t \in [fR_i+35, fR_i+200]} fTQRS(t).$$

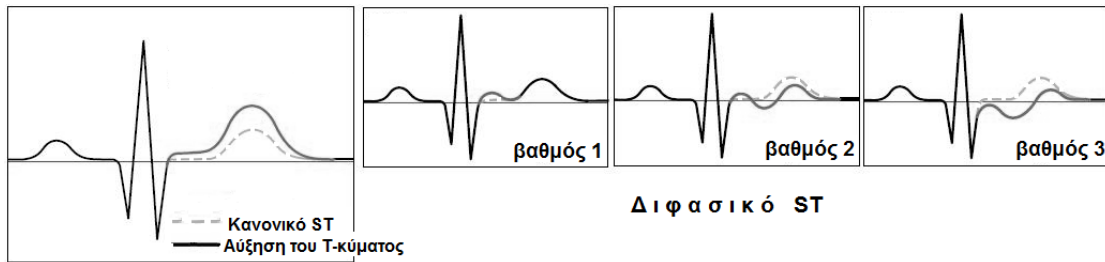
μ μ

μμ μ T/QRS
 μ 4.21.



μ 4.21. : (a) μ μμ (b)
μ T/QRS.

ST μ μ fPQRST
ST μ 1, 2 3. ST
μ μ 4.22 [84].



μ 4.22. ST μ [85]: () ST, () ST
μ 1, () ST μ 2 () ST μ 3.

, μ , ST μ μ
μ , ST.
μ μ μ
[75]. μ μ
μ μ μ
ST μ μ μ
μ μ μ
μ μ μ
STs : μ 1 ST
μ μ μ μμ , μ 2
ST μ μ μμ μ 3 μ μ ST
μμ . μ

μ

μ μ μ 1 μ 2 3.

$$\mu \quad ST \quad \mu \quad \mu \quad \mu$$

$$\mu \quad ST \quad : \quad ST_{\text{περιοχ}} = \sum_{t=fR_i+15}^{fR_i+80} fTQRST(t) \quad \mu$$

$$ST: ST_{|\text{περιοχ}|} = \sum_{t=fR_i+15}^{fR_i+80} |fTQRST(t)|, \quad \mu \quad :$$

$$ST_{\text{περιοχ}} \leq 0.2 \cdot ST_{|\text{περιοχ}|} \quad \mu \quad 3$$

$$- \quad ST_{\text{περιοχ}} < 0.8 \cdot ST_{|\text{περιοχ}|} \quad \mu \quad 2$$

μ ST, μ 1
(. μ μ [75])

μ μ ST μ μ 2 μ
3, ST μ , μ
ST μ ST μ μ 1.

4.12.3. μ

μ μ μ μ μ μ
, μ
μ μ [76].

μ μ μ μ μ μ ,
μ (μ
) μ μ

μ μ 8, μ μ μ
μ μ μ
BSS μ μ

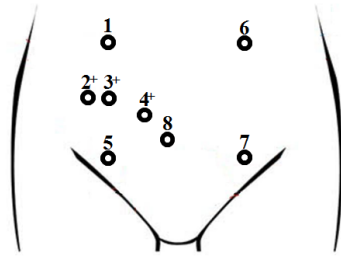
, μ μ [85].

μ 4.23, μ

μ μ

μ [63,86] μ 4.12 μ

μ μ [76].



μ 4.23. μ μ μ . 1 $5-7$
 μ μ , $2+ -4+$
 μ μ μ μ .

μ μ - - (SNR)
 μ μ μ : -5, -2, 0, 2, 5 10 dB. 5
 μ μ μ , μ μ μ μ
 μ (SNR) 8 (8

), μ 300 Hz μ :

1: ST μ (μ μ)

2: ST μ .

3: (μ 2) ST μ .

4: (μ 3) ST μ .

5: (μ 2 & 3) ST μ .

μ μ μ a, b
 c μ QRSs QRS

, μ . μ
 μ ST μ ,
 μ
 μ ST μ ,

3-5. μ , μ μ 2-5

, μ ST μ μ .

4.12.4.

μ μ , BSS
 . , BSS μ
 μ μ /
 μ , μ (μ QRSs).
 4.1 μ -
 μ μ μ
 : (i) μ BSS (μ
) (ii) μ BSS e (μ
). μ μ EFICA [88,89] BSS
 . - μ
 μ μ (8) μ
 μ μ BSS (8); μ μ
 4.1. μ μ μ μ μ
 μ μ 41.49%, μ
 μ e μ 75.25% (33.76%).

4.1. μ μ μ

| μ | μ (SNR dB) | | | | | | μ |
|-------|----------------|-------|-------|-------|-------|-------|-------------|
| | μ | μ | μ | μ | μ | μ | |
| | -5 | -2 | 0 | +2 | +5 | +10 | |
| | .252 | .234 | .261 | .326 | .639 | .777 | .415 |
| e | .634 | .682 | .683 | .742 | .871 | .903 | .753 |

EFICA BSS μ
 μ μ μ e
 μ μ μ , μ SNR μ 0 dB. μ
 μ μ μ ,
 μ μ μ . μ ,
 μ - μ
 μ (μ - μ
 μ

μ μ μ BSS), μ 16
 BSS [88]. , μ T/QRS
 μ . μ
 μ μ μ
 4.2.

4.2. μ μ T/QRS BSS
 . μ μ .

| T/QRS = 0.1523 | | | |
|----------------|--------------|--------------|-------------|
| BSS | T/QRS | | μ |
| AMUSE | .463 | .088 | .522 |
| SOBI | .491 | .093 | .551 |
| SOBI-PBF | .616 | .145 | .784 |
| WASOBI | .564 | .106 | .632 |
| EWASOBI | .589 | .108 | .649 |
| FJADE | .758 | .210 | .689 |
| JADEop | .763 | .212 | .686 |
| QJADE | .775 | .241 | .597 |
| FAJDC4 | .683 | .169 | .787 |
| FPICA | .734 | .204 | .699 |
| POWERICA | .662 | .166 | .788 |
| EFICA | .695 | .156 | .835 |
| COMBI | .688 | .172 | .781 |
| MULCOMBI | .677 | .165 | .797 |
| SANG | .685 | .161 | .815 |
| NG-FICA | .677 | .160 | .813 |
| Average | =.658 | =.160 | .714 |

, EFICA , μ
 μ μ μ BSS
 (), μ μ
 μ μ μ
 T/QRS (. T/QRS).

μ μ μ
 μ a, b and c, μQRSs μ

4.3. $(a_2, b_2, c_2) = (1.1, 1.3, 1.5)$

| SNR (dB) | TP | FP | FN | Se(%) | PDV(%) | Acc(%) |
|----------|-------------|------------|------------|--------------|--------------|--------------|
| -5 | 561 | 105 | 90 | 86.18 | 84.23 | 74.21 |
| -2 | 643 | 11 | 8 | 98.77 | 98.32 | 97.13 |
| 0 | 648 | 5 | 3 | 99.54 | 99.23 | 98.78 |
| 2 | 648 | 4 | 3 | 99.54 | 99.39 | 98.93 |
| 5 | 650 | 1 | 1 | 99.85 | 99.85 | 99.69 |
| 10 | 651 | 0 | 0 | 100 | 100 | 100 |
| | 3801 | 126 | 105 | 97.31 | 96.84 | 94.79 |

Figure 4.4 shows the performance of the proposed method in terms of True Positive Rate (TPR) and False Positive Rate (FPR) for different SNR values. The TPR and FPR are plotted against the SNR values from -5 dB to 10 dB. The TPR increases from 0.7421 at -5 dB to 1.0 at 10 dB, while the FPR decreases from 0.105 at -5 dB to 0 at 10 dB. The overall performance is summarized in Table 4.3.

4.4. μ , μ μ $SNRs, \mu$
 5 8 () 2 () 3 () 4 ()
 ()

| | 2 | | - | | ST | | μ | | |
|----------|-------|-------|-------|-------|-------|-------|-------|-------|--------------|
| μ | μ | | μ | | μ | | - | | () |
| SNR (dB) | .1 | .2 | .3 | .4 | .5 | .6 | .7 | .8 | μ |
| -5 | 0.753 | 0.694 | 0.617 | 0.684 | 0.276 | 0.378 | 0.713 | 0.715 | 0.604 |
| -2 | 0.786 | 0.708 | 0.66 | 0.685 | 0.317 | 0.436 | 0.775 | 0.758 | 0.641 |
| 0 | 0.845 | 0.777 | 0.721 | 0.724 | 0.356 | 0.48 | 0.827 | 0.807 | 0.692 |
| 2 | 0.902 | 0.834 | 0.753 | 0.8 | 0.34 | 0.474 | 0.862 | 0.856 | 0.728 |
| 5 | 0.906 | 0.824 | 0.771 | 0.78 | 0.375 | 0.513 | 0.893 | 0.87 | 0.742 |
| 10 | 0.979 | 0.91 | 0.826 | 0.863 | 0.383 | 0.527 | 0.938 | 0.928 | 0.794 |

()

| | 3 | | - | | (μ 2) | | ST | | μ |
|----------|-------|-------|-------|-------|------------|-------|-------|-------|--------------|
| μ | μ | | μ | | μ | | - | | () |
| SNR (dB) | .1 | .2 | .3 | .4 | .5 | .6 | .7 | .8 | μ |
| -5 | 0.599 | 0.57 | 0.538 | 0.524 | 0.283 | 0.375 | 0.609 | 0.57 | 0.509 |
| -2 | 0.703 | 0.652 | 0.61 | 0.593 | 0.285 | 0.397 | 0.702 | 0.661 | 0.575 |
| 0 | 0.81 | 0.765 | 0.69 | 0.706 | 0.291 | 0.42 | 0.783 | 0.751 | 0.652 |
| 2 | 0.871 | 0.832 | 0.762 | 0.743 | 0.346 | 0.482 | 0.853 | 0.807 | 0.712 |
| 5 | 0.882 | 0.844 | 0.772 | 0.756 | 0.352 | 0.49 | 0.862 | 0.818 | 0.722 |
| 10 | 0.936 | 0.902 | 0.819 | 0.812 | 0.371 | 0.515 | 0.907 | 0.867 | 0.766 |

()

| 4 - (μ 3) ST μ | | | | | | | | | |
|-------------------------|-------------------|-------|-------|-------|-------|-------|-------|-------|--------------|
| μ | μ (μ -) | | | | | | | | |
| SNR (dB) | .1 | .2 | .3 | .4 | .5 | .6 | .7 | .8 | μ |
| -5 | 0.516 | 0.63 | 0.621 | 0.421 | 0.429 | 0.493 | 0.521 | 0.444 | 0.509 |
| -2 | 0.572 | 0.671 | 0.67 | 0.467 | 0.447 | 0.526 | 0.586 | 0.504 | 0.555 |
| 0 | 0.594 | 0.717 | 0.718 | 0.482 | 0.484 | 0.568 | 0.607 | 0.513 | 0.585 |
| 2 | 0.624 | 0.759 | 0.734 | 0.53 | 0.48 | 0.56 | 0.615 | 0.536 | 0.605 |
| 5 | 0.681 | 0.82 | 0.806 | 0.571 | 0.528 | 0.622 | 0.683 | 0.589 | 0.663 |
| 10 | 0.719 | 0.844 | 0.85 | 0.581 | 0.568 | 0.67 | 0.742 | 0.632 | 0.701 |

()

| 5 - (μ 2 & 3) ST μ | | | | | | | | | |
|-----------------------------|-------------------|--------|--------|--------|--------|--------|--------|--------|--------------|
| μ | μ (μ -) | | | | | | | | |
| SNR (dB) | .1 | .2 | .3 | .4 | .5 | .6 | .7 | .8 | μ |
| -5 | 0.557 | 0.6223 | 0.5823 | 0.435 | 0.3318 | 0.3844 | 0.5343 | 0.523 | 0.496 |
| -2 | 0.569 | 0.6625 | 0.6696 | 0.6774 | 0.3948 | 0.3886 | 0.6434 | 0.6325 | 0.580 |
| 0 | 0.782 | 0.7516 | 0.6894 | 0.7308 | 0.4698 | 0.4037 | 0.6683 | 0.6512 | 0.643 |
| 2 | 0.858 | 0.8879 | 0.7877 | 0.753 | 0.4764 | 0.4567 | 0.6904 | 0.7174 | 0.703 |
| 5 | 0.896 | 0.8903 | 0.8237 | 0.7862 | 0.4948 | 0.5833 | 0.7892 | 0.7678 | 0.754 |
| 10 | 0.900 | 0.8922 | 0.8366 | 0.7962 | 0.499 | 0.6179 | 0.8878 | 0.7804 | 0.776 |

μ T/QRS : 4.5(-) μ
 μ T/QRS μ (T/QRS)
 μ μ (T/QRS). T/QRS
 :
 • μ T/QRS μ μ
 • μ μ T/QRS μ μ

• T/QRS μ μ μ μ
T/QRS μ (μ μ μ μ).

T/QRS μ μ μ T/QRS
 μ e . μ
 μ μ μ (SNR μ)
 μ μ .

4.5. μ μ μ T/QRS μ
(μ T/QRS μ), μ T/QRS μ
SNRs, μ 8
 μ ()2 ()3 ()4 ()5
()

2 – ST μ

| μ | T/QRS (T/QRS : 0.152) | | | | | | | | | | μ |
|--------------|------------------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|----------------|
| SNR (dB) | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | μ |
| -5 | 0.142 | 0.150 | 0.134 | 0.142 | 0.152 | 0.136 | 0.149 | 0.172 | 0.138 | 0.162 | 0.148 |
| -2 | 0.120 | 0.162 | 0.150 | 0.157 | 0.165 | 0.139 | 0.139 | 0.158 | 0.157 | 0.142 | 0.149 |
| 0 | 0.163 | 0.165 | 0.164 | 0.142 | 0.158 | 0.146 | 0.153 | 0.161 | 0.170 | 0.138 | 0.156 |
| 2 | 0.142 | 0.168 | 0.165 | 0.159 | 0.156 | 0.161 | 0.134 | 0.154 | 0.164 | 0.148 | 0.155 |
| 5 | 0.156 | 0.141 | 0.151 | 0.151 | 0.145 | 0.147 | 0.156 | 0.143 | 0.145 | 0.160 | 0.150 |
| 10 | 0.179 | 0.163 | 0.131 | 0.163 | 0.165 | 0.151 | 0.158 | 0.157 | 0.140 | 0.141 | 0.155 |
| T/QRS | | | | | | | | | | | : 0.152 |

()

| 3 - (μ 2) ST μ | | | | | | | | | | | |
|--------------------------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|------------------|
| μ | T/QRS | | | | | | | | | | (T/QRS : 0.113) |
| SNR (dB) | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | μ |
| -5 | 0.163 | 0.103 | 0.137 | 0.142 | 0.116 | 0.059 | 0.107 | 0.097 | 0.131 | 0.105 | 0.116 |
| -2 | 0.100 | 0.128 | 0.018 | 0.145 | 0.125 | 0.130 | 0.127 | 0.126 | 0.136 | 0.136 | 0.117 |
| 0 | 0.110 | 0.131 | 0.121 | 0.103 | 0.117 | 0.145 | 0.108 | 0.097 | 0.116 | 0.114 | 0.116 |
| 2 | 0.119 | 0.099 | 0.138 | 0.126 | 0.162 | 0.137 | 0.111 | 0.102 | 0.121 | 0.121 | 0.124 |
| 5 | 0.130 | 0.118 | 0.113 | 0.110 | 0.128 | 0.118 | 0.127 | 0.127 | 0.101 | 0.111 | 0.118 |
| 10 | 0.121 | 0.126 | 0.124 | 0.120 | 0.111 | 0.125 | 0.118 | 0.122 | 0.123 | 0.120 | 0.121 |
| T/QRS | | | | | | | | | | | : 0.119 |

()

| 4 - (μ 3) ST μ | | | | | | | | | | | |
|--------------------------------|-------|-------|------|------|------|------|------|------|------|------|------------------|
| μ | T/QRS | | | | | | | | | | (T/QRS : 0.088) |
| SNR (dB) | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | μ |
| -5 | 0.094 | 0.122 | .203 | .137 | .154 | .139 | .134 | .123 | .135 | .106 | .135 |
| -2 | 0.153 | 0.127 | .129 | .174 | .125 | .123 | .121 | .131 | .129 | .130 | .134 |
| 0 | 0.106 | 0.095 | .116 | .085 | .093 | .083 | .093 | .088 | .096 | .126 | .098 |
| 2 | 0.105 | 0.093 | .084 | .101 | .074 | .088 | .093 | .080 | .095 | .093 | .091 |
| 5 | 0.167 | 0.134 | .126 | .076 | .071 | .082 | .067 | .089 | .093 | .076 | .098 |
| 10 | 0.102 | 0.114 | .136 | .051 | .072 | .075 | .064 | .061 | .064 | .086 | .082 |
| T/QRS | | | | | | | | | | | : 0.106 |

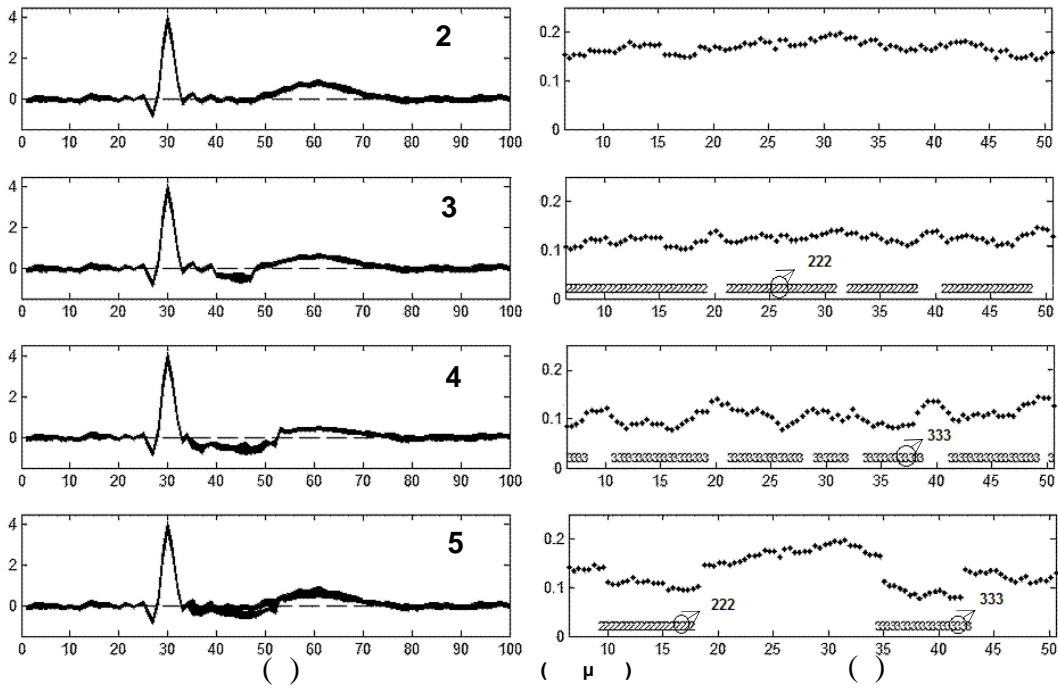
()

| 5 - (μ 2 & 3) ST μ | | | | | | | | | | | | |
|-----------------------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|--------|----------------|--|
| μ | T/QRS | | | | (| T/QRS | | | : | 0.115) | | |
| SNR (dB) | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | μ | |
| -5 | 0.143 | 0.104 | 0.112 | 0.155 | 0.148 | 0.150 | 0.111 | 0.135 | 0.089 | 0.170 | 0.132 | |
| -2 | 0.166 | 0.104 | 0.103 | 0.130 | 0.132 | 0.093 | 0.152 | 0.139 | 0.162 | 0.120 | 0.130 | |
| 0 | 0.148 | 0.130 | 0.154 | 0.136 | 0.163 | 0.088 | 0.132 | 0.109 | 0.142 | 0.095 | 0.129 | |
| 2 | 0.126 | 0.094 | 0.164 | 0.112 | 0.148 | 0.128 | 0.082 | 0.096 | 0.121 | 0.129 | 0.120 | |
| 5 | 0.160 | 0.156 | 0.099 | 0.133 | 0.115 | 0.164 | 0.095 | 0.104 | 0.101 | 0.088 | 0.122 | |
| 10 | 0.104 | 0.093 | 0.103 | 0.123 | 0.106 | 0.086 | 0.123 | 0.128 | 0.148 | 0.155 | 0.117 | |
| T/QRS | | | | | | | | | | | : 0.125 | |

ST μ : μ , μ μ ST μ
 : , μ 2 μ
 3. μ μ μ :
 μ μ 200 ST
 μ μ 2 200 ST μ μ 3.
 , μ 2 μ 3 ST
 μ , μ μ μ , 4.6.
 μ 4.24 μ μ μ μ
 μ (2 dB SNR) 1 ,
 ST μ (. , μ 2,
 μ 3, μ 2 3). ,
 μ μ PQRST (μ)
 μ T/QRS μ μ annotation
 ST μ . annotation μ μ 2 3
 ST μ μ 2 3, , μ
 annotation ST μ .

4.6. μ SNR. μ ST μ ,

| μ (SNR dB) | μ | | μ | | | |
|-------------------|-------|----------|-------------|------------|--------------|--------------|
| | ST | | TP | FN | Se(%) | Acc(%) |
| -5 | | | 155 | 80 | 65.96 | |
| | μ | 2 | 126 | 74 | 63.00 | 68.82 |
| | μ | 3 | 156 | 44 | 78.00 | |
| -2 | | | 169 | 66 | 71.91 | |
| | μ | 2 | 133 | 67 | 66.50 | 73.07 |
| | μ | 3 | 162 | 38 | 81.00 | |
| 0 | | | 179 | 56 | 76.17 | |
| | μ | 2 | 144 | 56 | 72.00 | 77.95 |
| | μ | 3 | 172 | 28 | 86.00 | |
| 2 | | | 183 | 52 | 77.87 | |
| | μ | 2 | 150 | 50 | 75.00 | 80.63 |
| | μ | 3 | 179 | 21 | 89.50 | |
| 5 | | | 201 | 34 | 85.53 | |
| | μ | 2 | 163 | 37 | 81.50 | 86.93 |
| | μ | 3 | 188 | 12 | 94.00 | |
| 10 | | | 208 | 27 | 88.51 | |
| | μ | 2 | 179 | 21 | 89.50 | 90.71 |
| | μ | 3 | 189 | 11 | 94.50 | |
| | | | 1095 | 315 | 77.66 | |
| | μ | 2 | 902 | 298 | 74.58 | 79.87 |
| | μ | 3 | 1046 | 154 | 87.17 | |



μ 4.24. μ μ μ μ μ μ μ (2 dB SNR). ()

μ μ μ PQRS T ()

ST μ μ T/QRS . annotation

μ μ μ μ 2 3 ST μ μ 2

3, .

μ 4.25 μ μ μ μ .

μ 4.25() μ , μ 4.25()

μ 4.25() T/QRS annotation ST

μ (μ) .

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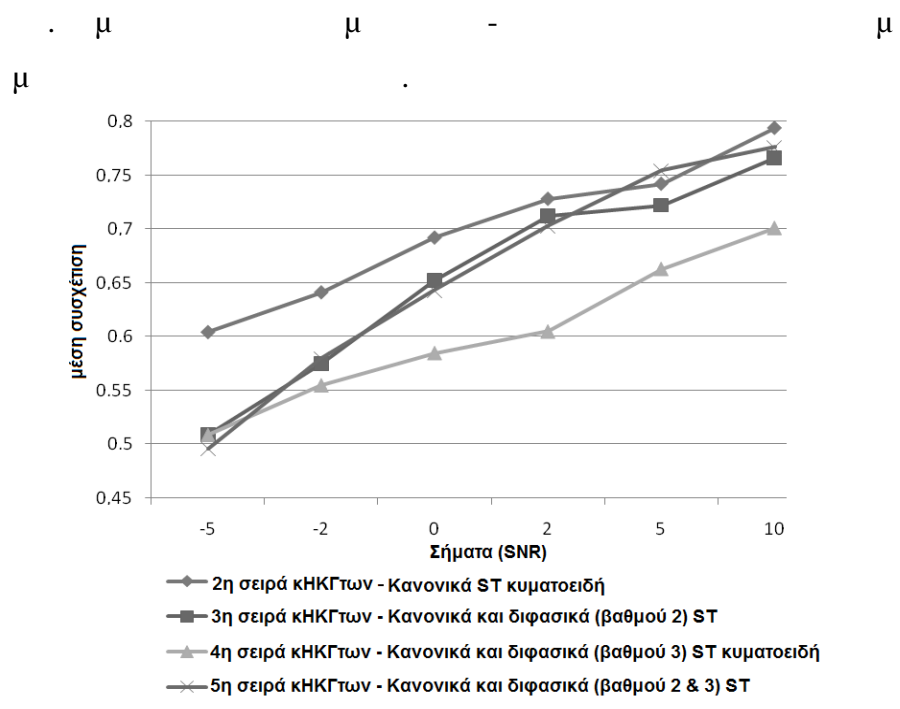
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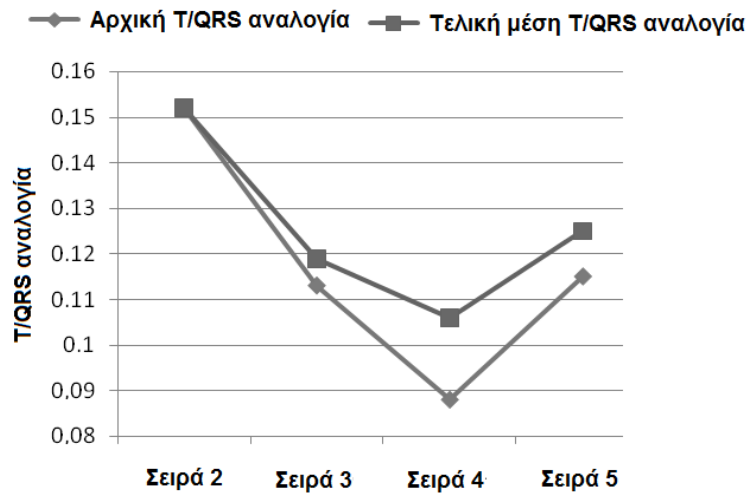


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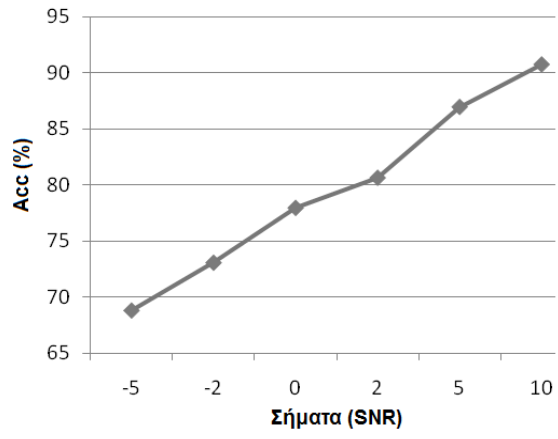
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SUMMARY IN ENGLISH

Karvounis C. Evaggelos, PhD, Department of Materials Science and Engineering, University of Ioannina, Greece, September 2009. Thesis: Automated diagnosis of fetal Electrocardiogram (ECG). Supervisor: Dimitrios I. Fotiadis.

During pregnancy, the flow of oxygen and nutrients to the fetus and the removal of carbon dioxide and other waste gases from it is achieved through the placenta. Adequate blood flow to and from the placenta, and in both the maternal and fetal circulations, is necessary in order for the baby to receive enough oxygen and for it to be able to expel carbon dioxide and other waste gases. Any alteration in placental function can lead to decreases in the delivery of oxygen to the baby, a condition known as fetal hypoxia. The motivation for monitoring the fetus during pregnancy is to recognize pathologic conditions, typically decreased oxygen saturation, accompanied with sufficient warning to enable intervention by the clinician before irreversible changes take place.

Scientists are working for the last decades to develop new technologies for the continuous intrapartum fetal monitoring. Early approaches are used for monitoring of the fetal heart rate (fHR) and the mother's uterine contractions (cardiotocography-CTG). In CTG, fHR is monitored using an ultrasound transducer strapped to the mother's abdomen, while uterine activity is recorded from an external toco sensor. Monitoring using CTG mainly identifies fetuses affected by intrapartum asphyxia, resulting in early intervention and a reduction in cerebral palsy. Unfortunately, a large number of fetuses affect fHR without being asphyxiated. Thus, electronic fHR monitoring based on CTG provides with poor specificity in detecting fetal hypoxia and cannot provide all information which is required. This has created an increased rate of intervention and uncertainty about the clinical value of CTG.

Doppler ultrasound is a widely used technique by medical doctors to monitor fHR. However, except some specific disadvantages (e.g. need for experienced personnel, specialized equipment and use in hospital environments), the major limitation of the Doppler ultrasound is its sensitivity to any movement. The movement of the mother can result in Doppler-shifted reflected waves, which are stronger than the cardiac signal. Thus this technique is not suitable for long-term monitoring of the fHR as it requires the patients to be bed-rested. In addition, there have been a number of publications linking diagnostic ultrasound to an increase in Intrauterine Growth Restriction (IUGR) and the stimulation of endothelial cell growth and the release of adenosine triphosphate (A-tp). However, the effect of ultrasound on the fetus is not completely clear.

Important clinical studies support the incorporation of ST waveform analysis into fHR analysis for intrapartum monitoring, with reduction in the rates of neonatal metabolic acidosis as well as neonatal encephalopathy. ST waveform analysis is mainly performed in the fetal ECG (fECG) signal, recorded using a fetal scalp electrode. Repolarisation of myocardial (heart muscle) cells is very sensitive to metabolic dysfunction, and might be reflected in changes of the ST waveform. The changes in fECG associated with fetal hypoxia are either an increase in T-wave, quantified by the ratio of the T-wave to the QRS amplitude (T/QRS ratio), or biphasic ST-pattern: the combination of these features with fHR pattern analysis and additional clinical information can lead to accurate identification of hypoxia cases and to the avoidance of unnecessary interventions. However, application of a fetal scalp electrode has the risk of maternal to fetal infection, which contraindicates any invasive monitoring technique. In addition, as an invasive technique, this type of fetal monitoring is less acceptable than external monitoring from pregnant women and midwives. Also, as the system responds primarily to changes in the ST segment, if it is applied when such changes have already occurred, there is the possibility of a false-negative result (inappropriate reassurance about a fetal condition). Thus, automated assessment of fetal cardiac health status based on non-invasive monitoring techniques is an important issue which must be investigated.

The abdominal ECG (abdECG) recordings is a very promising field, since it offers several advantages over Doppler ultrasound; lightweight electrodes are used and it is simple to operate, even by the mothers themselves, therefore, it can be used in home environment. The procedure is non-invasive and can be used for long duration recordings. However, the abdominal leads record a composite signal (abdECG), consisting of the contributions from both the maternal ECG (mECG), the fECG and sources of interference including intrinsic noise from the recorder, noise from electrode-skin contact and movement, baseline drift (DC shift), A/C interference noise, uterine contractions activity, etc. The main goal in this technique is the extraction of fECG and fHR from the abdECG. fECG occasionally overlaps with mECG while fECG signal depends on the gestational age, the position of the fetus and the positioning of the electrodes (there is no standard electrode positioning for optimal fECG acquisition [13]). All the above, make the problem of fECG and/or fHR extraction from the abdECG a very complex task. Two major approaches exist in the analysis of fetal electrical activity from abdECG signals: (a) extraction of the fECG from the abdECG and, subsequently, identification of the fHR from it, or (b) direct extraction of the fHR. The analysis of the abdECG depends on the scope of the study.

In this PhD, we proposed mainly: i. two novel methodologies for the automated detection of the fHR and mHR signals, based on the analysis of abdECG leads, and ii. a prenatal methodology for monitoring of the cardiac condition of both the mother and the fetus, by developing a transabdominal system for long-term monitoring of ECG leads, placed on the mother's abdomen.

The first proposed methodology introduces an automated methodology for the extraction of fHR from cutaneous potential abdominal ECG recordings. A three-stage methodology is proposed. Having the initial recording, which consists of a small number of abdominal ECG leads, in the first stage the maternal R-peaks and fiducial points (QRS onset and offset) are detected, using time-frequency analysis and medical knowledge. Then, the maternal QRS complexes are eliminated. In the second stage the positions of the candidate fetal R-peaks are located using complex wavelets and matching theory techniques. In the third stage, the fetal R-peaks which overlap with the maternal QRS complexes (eliminated in the first stage) are found using two

approaches: a heuristic algorithm and a histogram-based technique. The fetal R-peaks detected, are used to calculate the fHR. The methodology is validated using a dataset of 8 short and 10 long duration recordings, obtained between the 20th and the 41st week of gestation and the obtained accuracy is 97.47%. The proposed methodology is advantageous since it is based on the analysis of few abdominal leads, in contrast to other proposed methods which need a large number of leads.

The second proposed methodology contains three-stages in the same way for the detection of fHR from multivariate abdECG recordings. In the first stage, the maternal R-peaks and fiducial points (maternal QRS onset and offset) are detected, using band-pass filtering and phase space analysis. The maternal fiducial points are used to eliminate the maternal QRS complexes from the abdominal ECG recordings. In the second stage, two denoising procedures are applied to enhance the fetal QRS complexes. The phase space characteristics are employed to identify fetal heart beats not overlapping with the maternal QRSs which are eliminated in the first stage. The extraction of the fetal heart rate is accomplished in the third stage, using a histogram based technique in order to identify the location of the fetal heart beats which overlap with the maternal QRSs. The methodology is evaluated on simulated multichannel ECG signals, generated by a recently proposed model with various signal-to-noise ratios (SNR), and on real signals, recorded from pregnant women in various weeks during gestation. In both cases, the obtained results indicate high performance; in the simulated ECGs the accuracy ranges from 72.78 – 98.61%, depending on the employed SNR, while in the real recordings the average accuracy is 95.45%. The proposed methodology is advantageous since it copes with the existence of noise from various sources while it is applicable in multichannel abdominal recordings.

In the last work a methodology is proposed for monitoring the fetal cardiac health status during pregnancy, through the effective and non-invasive monitoring of the abdECG of the mother. For this purpose, a three-stage methodology has been developed. In the first stage, the fHR is extracted from the abdECG signals, using nonlinear analysis. Also, the eliminated ECG (eECG) is calculated, which is the abdECG after the maternal QRSs elimination. In the second stage, a blind source separation technique is applied to the eECG signals and the fECG is obtained. Finally,

monitoring of the fetus is implemented using features extracted from the fHR and fECG, such as the T/QRS ratio and the characterization of the fetal ST waveforms. The proposed methodology is evaluated using a dataset of simulated multichannel abdECG signals, exhibiting a high diagnostic accuracy: 94.79% accuracy for fHR extraction, 92.49% accuracy in T/QRS ratio calculation and 79.87% in ST waveform classification.

Feature work: Except the first 2 proposed methodologies related to fHR extraction, additional features, related to the fHR, have to be extracted. These features are the baseline fHR, the fHR variability, the reactivity and the appearance of accelerations and decelerations, while based on these parameters; fHR can be classified as reassuring, non-reassuring and abnormal. Also, ST events, which are heart episodes related to the T/QRS ratio and the ST waveform classification, can be characterized as normal, episodic, baseline or biphasic. All the above can easily be extracted from the results obtained from the last proposed methodology (fHR, fECG, T/QRS ratio, ST waveform classification). However, they have not been included in this work since the evaluation is made using only simulated signals, which present specific characteristics (such as fHR) and do not include abnormalities. This is a limitation of the last proposed methodology and evaluation using real abdECG recordings is a very important issue in order to fully exploit its potentials. In addition, the proposed methodology has the opportunity to extract mECG and mHR. In that way, monitoring of the mother can be also implemented, using sophisticated techniques in arrhythmia and/or ischemia detection. The employment of the above described (additional) features along with the application of the methodology to real abdECG recordings and maternal monitoring will be addressed in future communications. Another limitation of the last proposed procedure is the lack of a uterine contraction recording. Uterine contraction can affect the fHR, therefore, the incorporation of an external transducer for recordings of the uterine activity is essential for the detection of all types of periodic fHR changes, leading to a more accurate and precise monitoring procedure.

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