

Chapter 4

Event Detection-1

Dr. Bülent Yılmaz

Contents

- Introduction
- Problem statement
- Detection of events and waves
- QRS detection
 - Derivative-based methods
 - Pan-Tompkins algorithm

Introduction

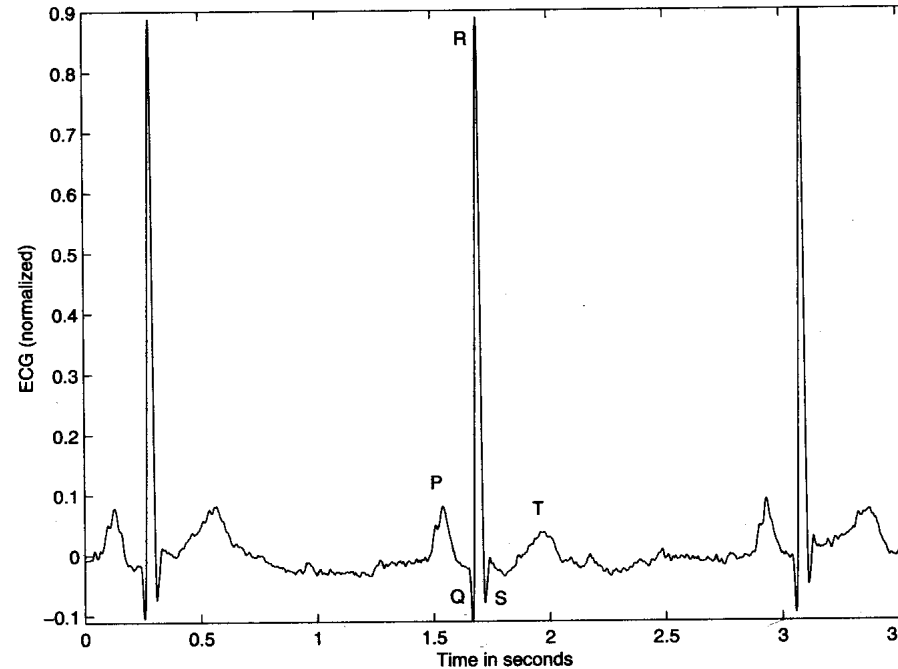
- Biomedical signals carry signatures of physiological events
- Part of a signal related to a specific event of interest is referred to as an “*epoch*”
- Analysis requires identification of epochs
 - For monitoring and diagnosis
- The corresponding waveform may be segmented and analyzed in terms of its
 - Amplitude, waveform, time-duration, intervals between events, energy distribution, frequency content

Problem statement

- Given a biomedical signal, identify discrete signal epochs and correlate them with events in the related physiological process

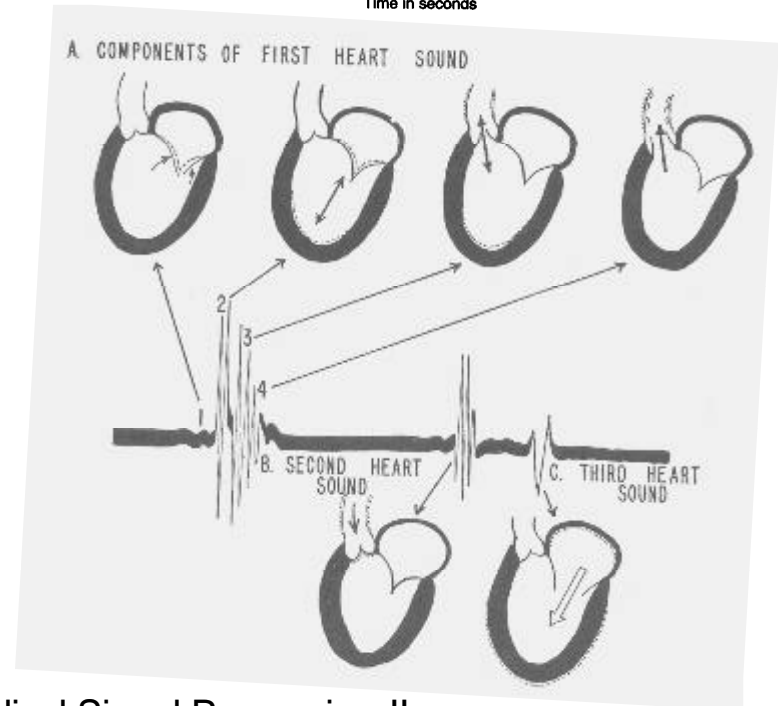
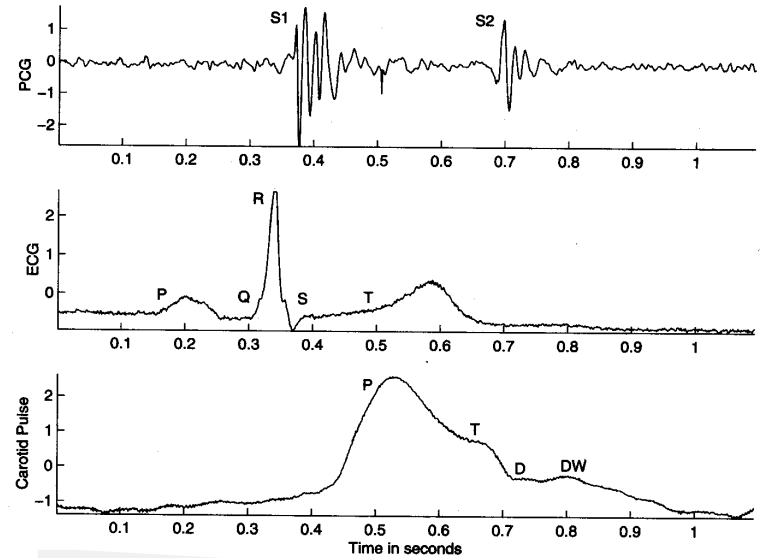
Normal ECG

- Slow P wave: 0.1-0.2 mV
60-80 ms
- PQ segment: AV delay
60-80 ms
 - isoelectric
- QRS complex: sharp
biphasic or triphasic wave
of about 1 mV amplitude
and 80 ms duration
- ST segment: 100-120 ms
 - Isoelectric
- Slow T wave: 0.1-0.3 mV
and duration 120-160 ms



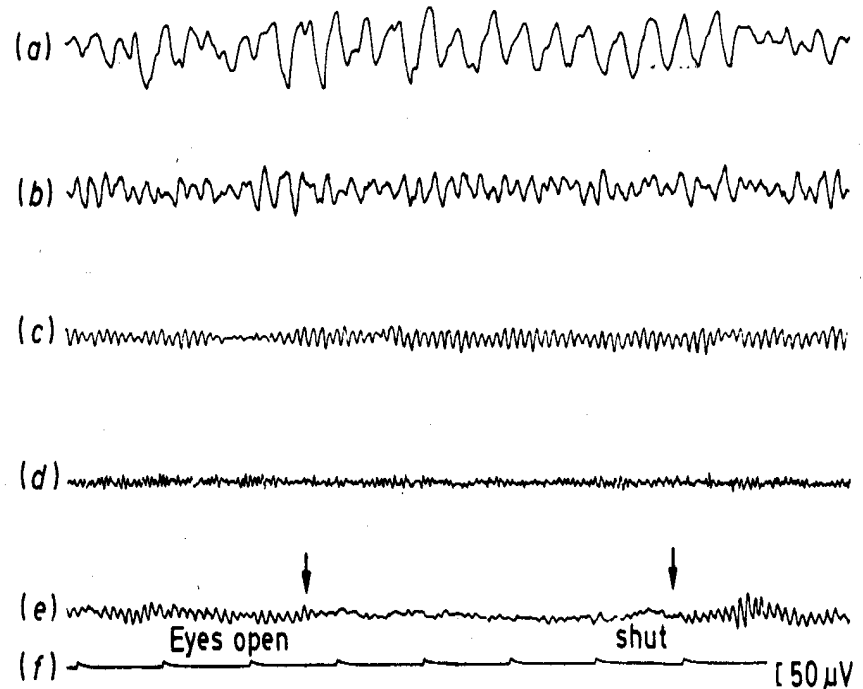
PCG signal

- S1 occurs at the onset of ventricular contraction
 - Corresponds in timing to the QRS complex in the ECG signal
- S2 is caused by the closure of the semilunar valves (aortic and pulmonary valves)



EEG signals

- **Delta waves**
 - $0.5 \leq f < 4$ Hz, appear at deep-sleep stages
- **Theta waves**
 - $4 \leq f < 8$ Hz, appear at the beginning stages of sleep
- **Alpha waves**
 - $8 \leq f < 13$ Hz, principal resting rhythm
 - Auditory and mental arithmetic tasks with eyes closed
- **Beta waves**
 - $f > 13$ Hz, background activity in tense and anxious subjects



Detection of Events and Waves-1

- QRS detection
 - Derivative-based methods
 - Pan-Tompkins algorithm
- Correlation analysis of EEG channels
 - Detection of EEG rhythms
 - Template matching for EEG spike-and-wave detection

Detection of Events and Waves-2

- Matched filter
- P-wave detection

Applications

- ECG rhythm analysis

QRS Detection

- Derivative-based methods
- Pan-Tompkins algorithm

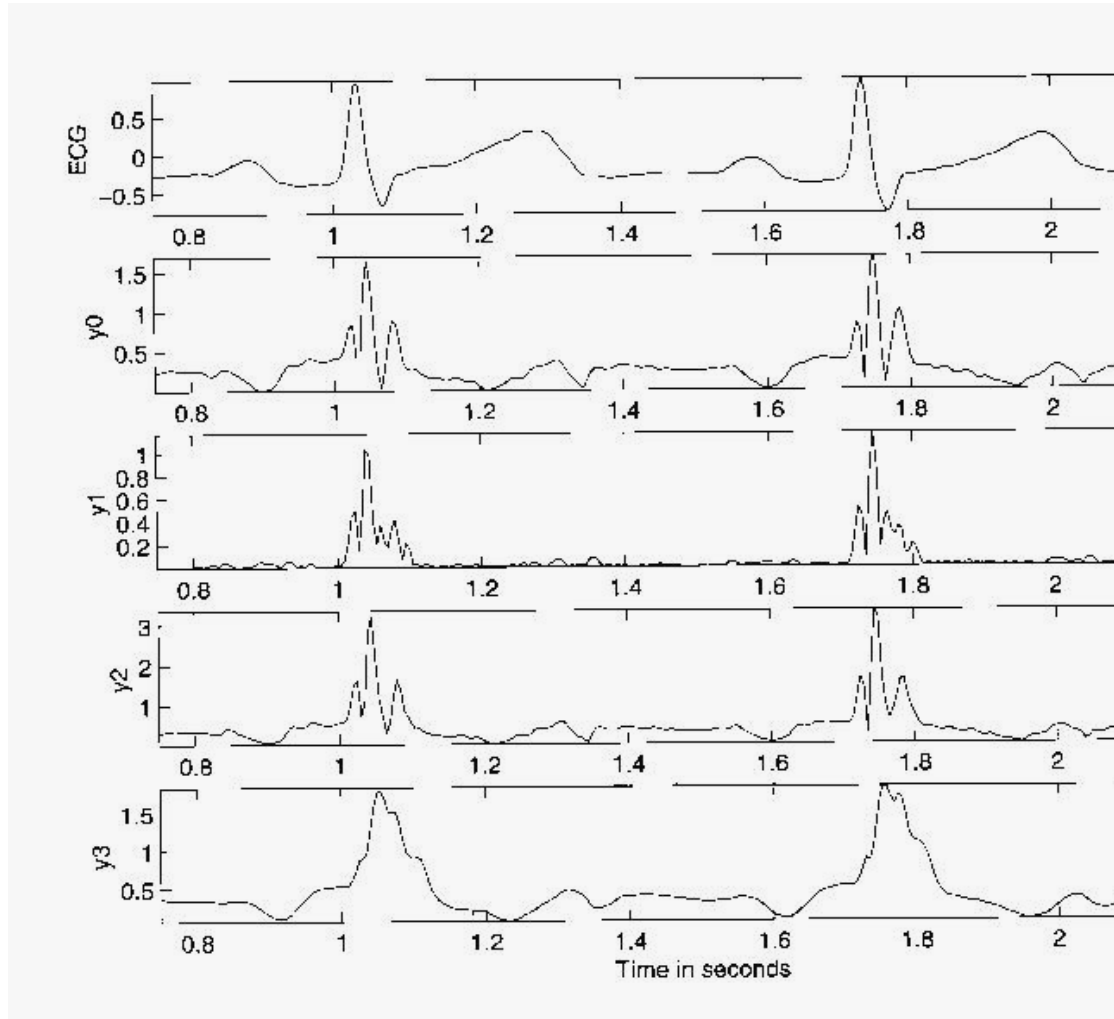
Derivative-based methods

- QRS might not always be the highest wave in a cardiac cycle
 - artifacts may upset the peak search algorithm
- QRS complex has the largest slope (rate of change of voltage)
- Rate of change = derivative operator (d/dt)
- Derivative operator:
 - P and T waves will be suppressed
 - Output is the highest at the QRS

Derivative-based algorithm

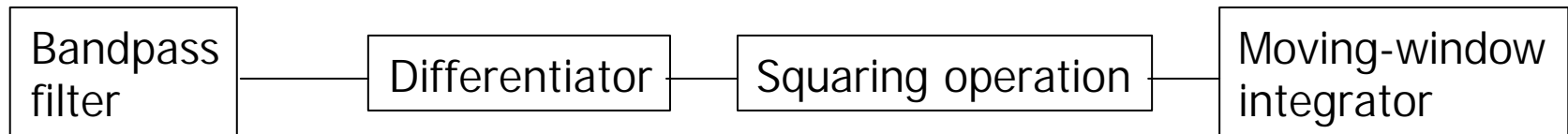
- Balda et al proposed an algorithm
 - Three-point first derivative
 - $y_0[n] = |x[n] - x[n-2]|$
 - Second derivative
 - $y_1[n] = |x[n] - 2x[n-2] + x[n-4]|$
 - The two results are weighted and combined
 - $y_2[n] = 1.3 y_0[n] - 1.1 y_1[n]$
 - The result $y_2[n]$ is scanned with a threshold of 1.0
 - Whenever threshold is crossed
 - Subsequent 8 samples also tested against the same threshold
 - If at least 6 pass the threshold test
 - The segment of eight samples is taken to be a part of a QRS complex

Figure 4.2



The Pan-Tompkins algorithm

- Pan and Tompkins proposed a real-time QRS detection algorithm based on
 - Slope, amplitude, and width of QRS complexes



Algorithm details

- Recursive LPF

- $H(z) = (1/32) \left(\frac{(1-z^{-6})^2}{(1-z^{-1})^2} \right)$

- $y[n] = 2 y[n-1] - y[n-2] + (1/32) [x[n] - 2x[n-6] + x[n-12]]$

- Sampling rate = 200 Hz, $f_c = 11$ Hz
- Filter introduces 5 samples of delay (25 ms)

Algorithm details

- HPF
- Allpass filter minus a LPF
 - $H_{lp}(z) = (1-z^{-32})/(1-z^{-1})$
 - $y[n] = y[n-1] + x[n] - x[n-32]$
- $H_{hp}(z) = z^{-16} - (1/32)H_{lp}(z)$
 - $p[n] = x[n-16] - (1/32)[y[n-1] + x[n] - x[n-32]]$
- $f_c = 5$ Hz
- Filter introduces 80 ms of delay

Derivative operator

- $y[n] = (1/8) [2x[n] + x[n-1] - x[n-3] - 2x[n-4]]$
 - Approximates the ideal d/dt operator up to 30 Hz
- Suppresses P and T waves

Squaring

- Makes the results positive and emphasizes large differences arising from QRS complexes
- Small differences arising from P and T waves are suppressed

Integration

- Multiple peaks within the duration of a single QRS complex
- Smoothing of the output of the preceding operations through a moving window integration filter
 - $y[n] = (1/N) [x[n - (N-1)] + x[n - (N-2)] + \dots + x[n]]$
 - N: window width (N=30 found to be suitable for $f_s=200$ Hz)

Adaptive thresholding

- A peak is said to be detected whenever the final output changes direction within a specified interval

Figure 4.6

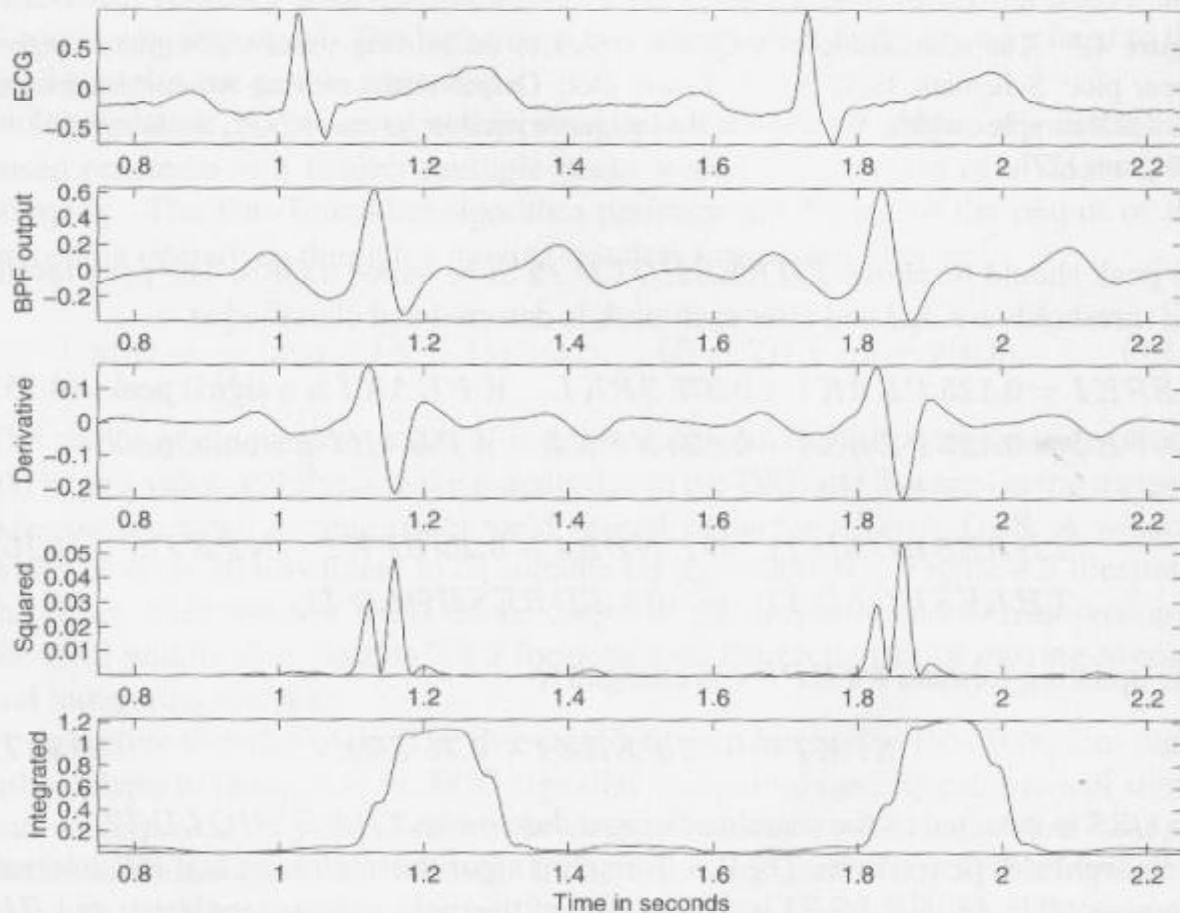


Figure 4.6. Results of the Derivative and Squared Derivative