



**INSTITUTE**

**OF SCIENTIFIC INSTRUMENTS**

Academy of Sciences of the Czech Republic, v. v. i.



# **False Alarms in ICU Monitors: Detection of Life-threatening Arrhythmias**

F. Plesinger, P. Klimes, J. Halamek, P. Jurak

## What is an ICU?

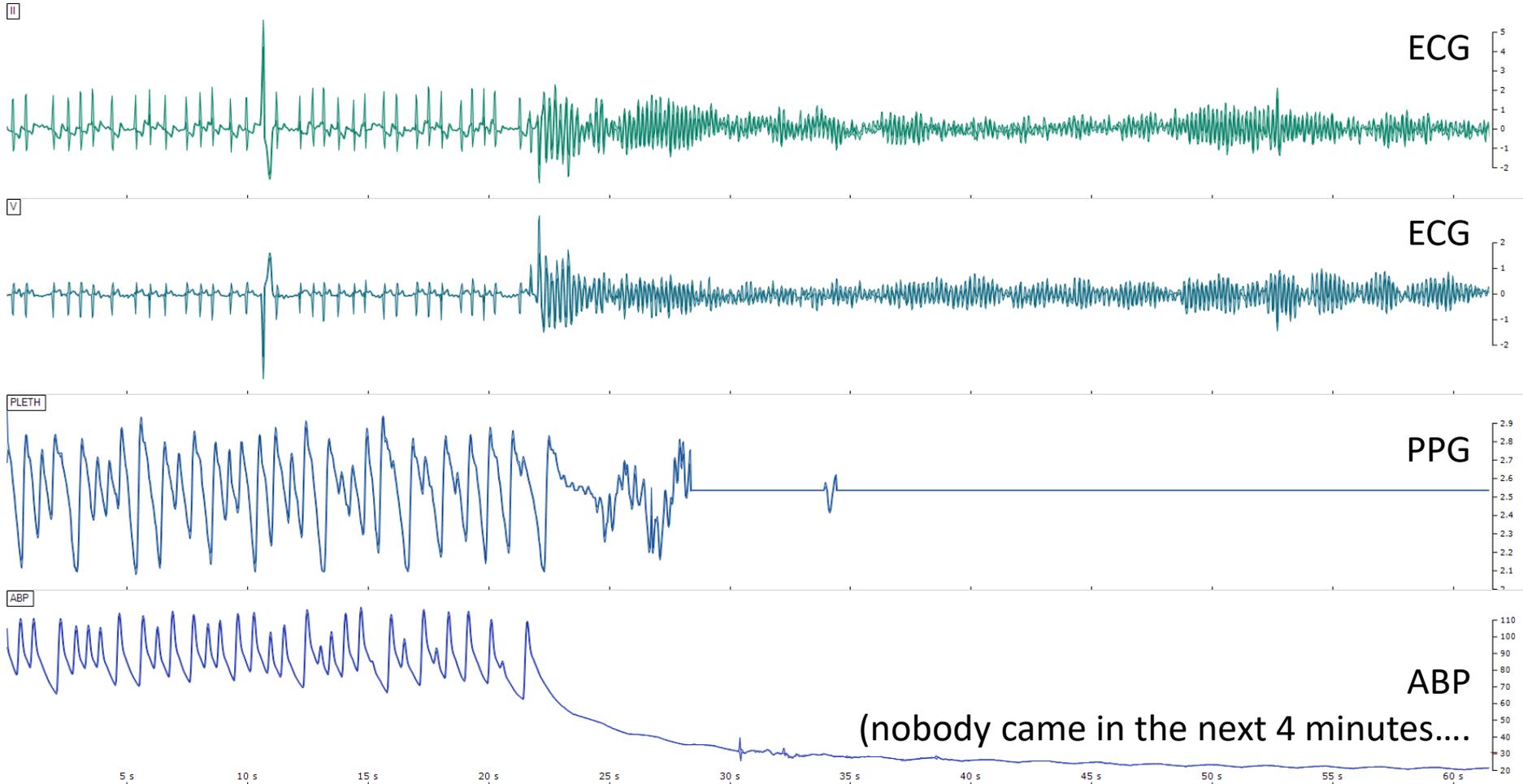
ICU – Intensive Care Unit. When the **life-threatening arrhythmia** occurs, the ICU monitor raises an alarm.

## So what is the problem?

**Up to 86% of ICU alarms are false =>**

- Sleep deprivation of patients
- Psychically demanding for personell => less attention => „Crying wolf“ effect

**Which may results in this...**  
(f545l.mat)



**A solution to this problem was the task of Physionet Challenge 2015.**

## **What is the PhysioNet /CinC Challenge?**

- Guys from **MIT** started this challenge in the year 2000. It is tightly connected to the conference Computing in Cardiology (CinC)
- Each year different topic aimed to processing of physiological signals
- Teams from all over the world (each year 20-110 teams). Most from academic field, some from private companies
- Usual scenario is to develop algorithm and train it using public training set. Then the algorithm is sent to PhysioNet servers and tested on subset of hidden test set (max 10 trials for each team).
- <https://physionet.org/>

## Which signals were available?

(All signals were downsampled to 250Hz, length 300 or 330 seconds)

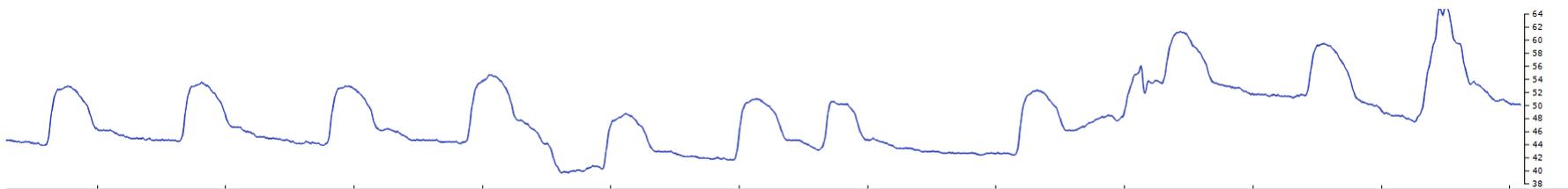
2x ECG (Electrocardiogram)



PPG (Photoplethysmogram)



ABP (arterial blood pressure)



(respiratory curve – not used)

# Detection of Life-threatening Arrhythmias

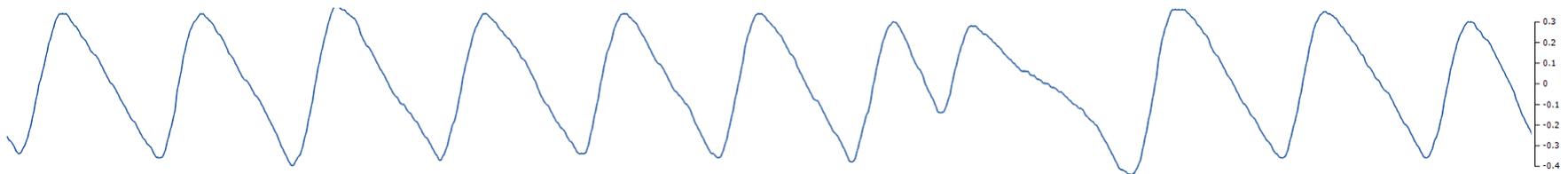
## Which signals were available?

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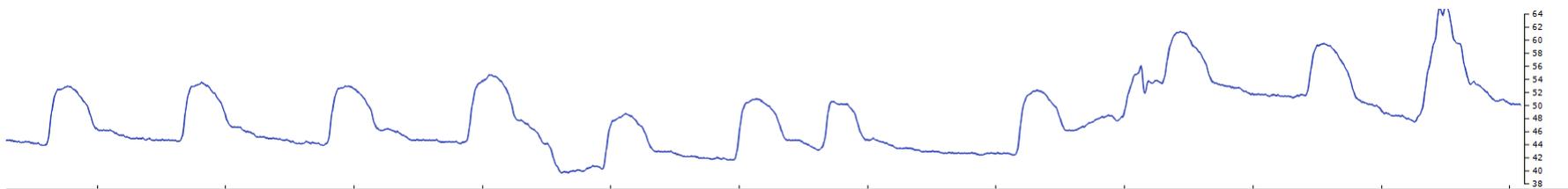
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PPG (Photoplethysmogram)

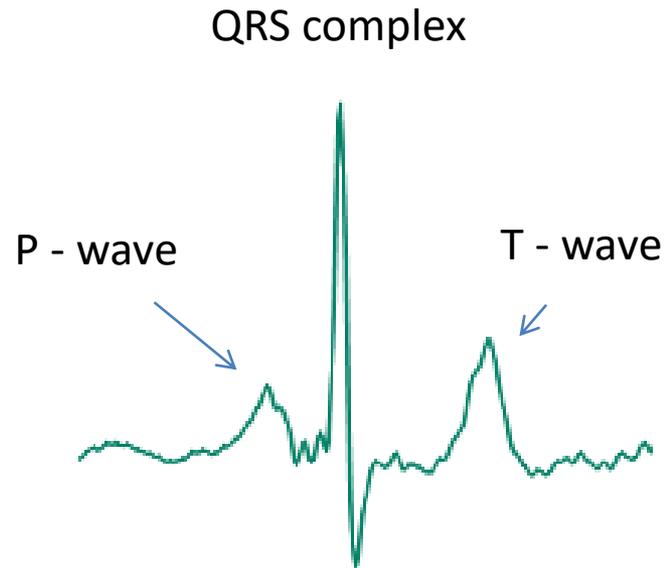


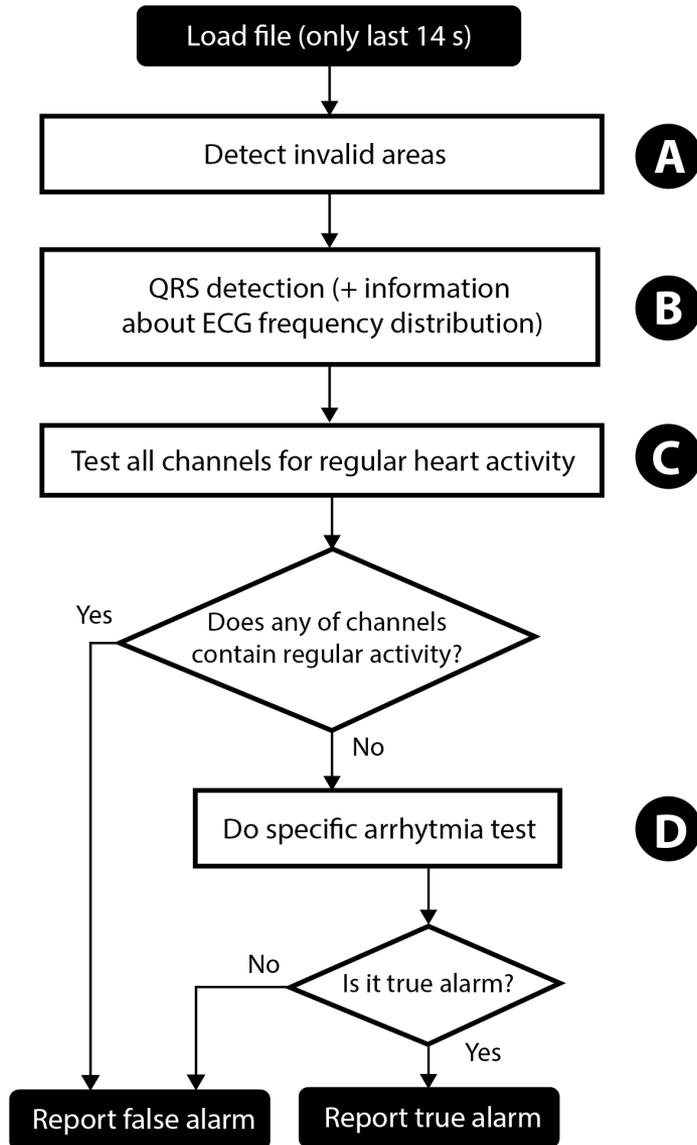
ABP (arterial blood pressure)



(respiratory curve – not used)

## ECG detail

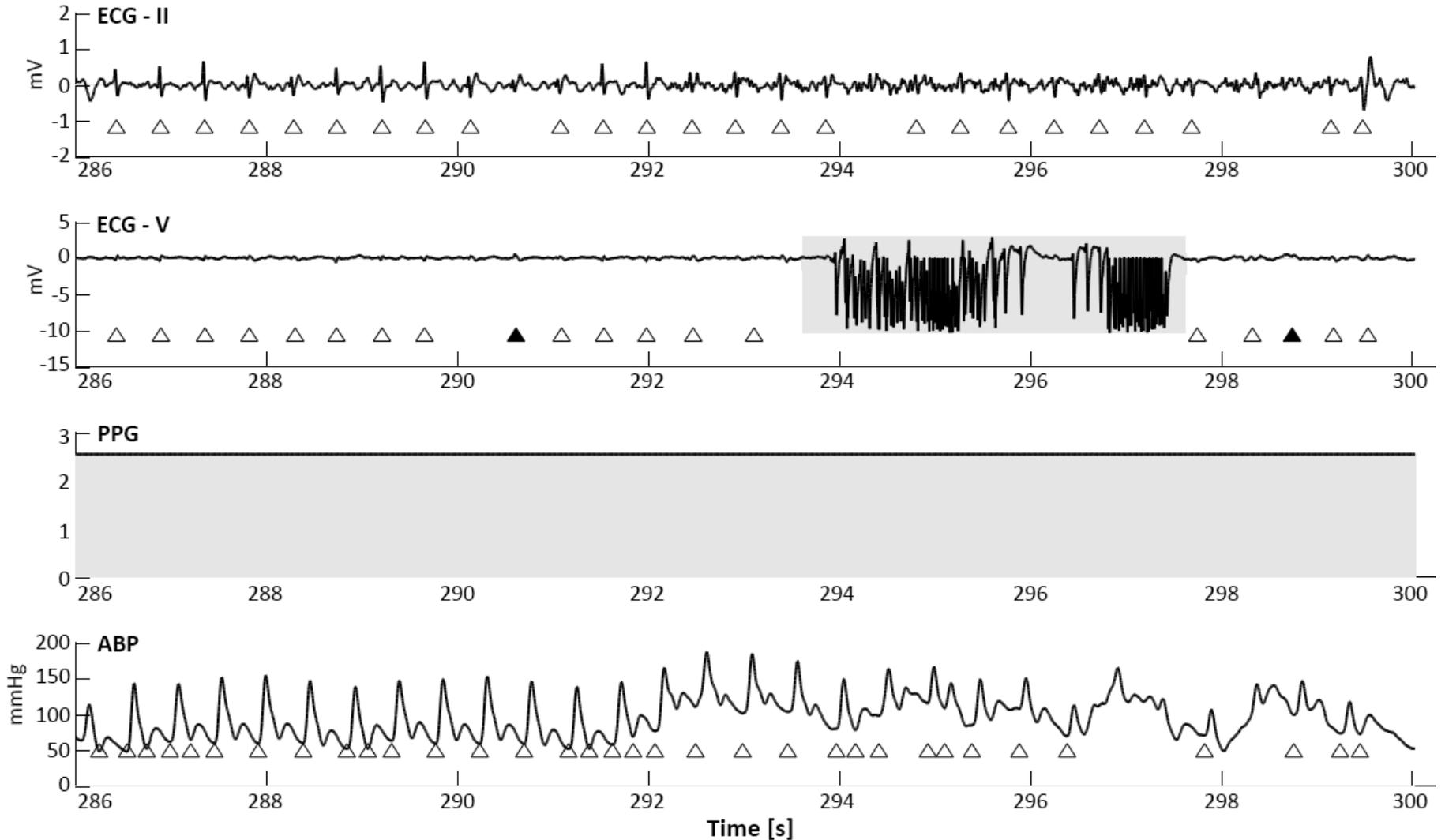




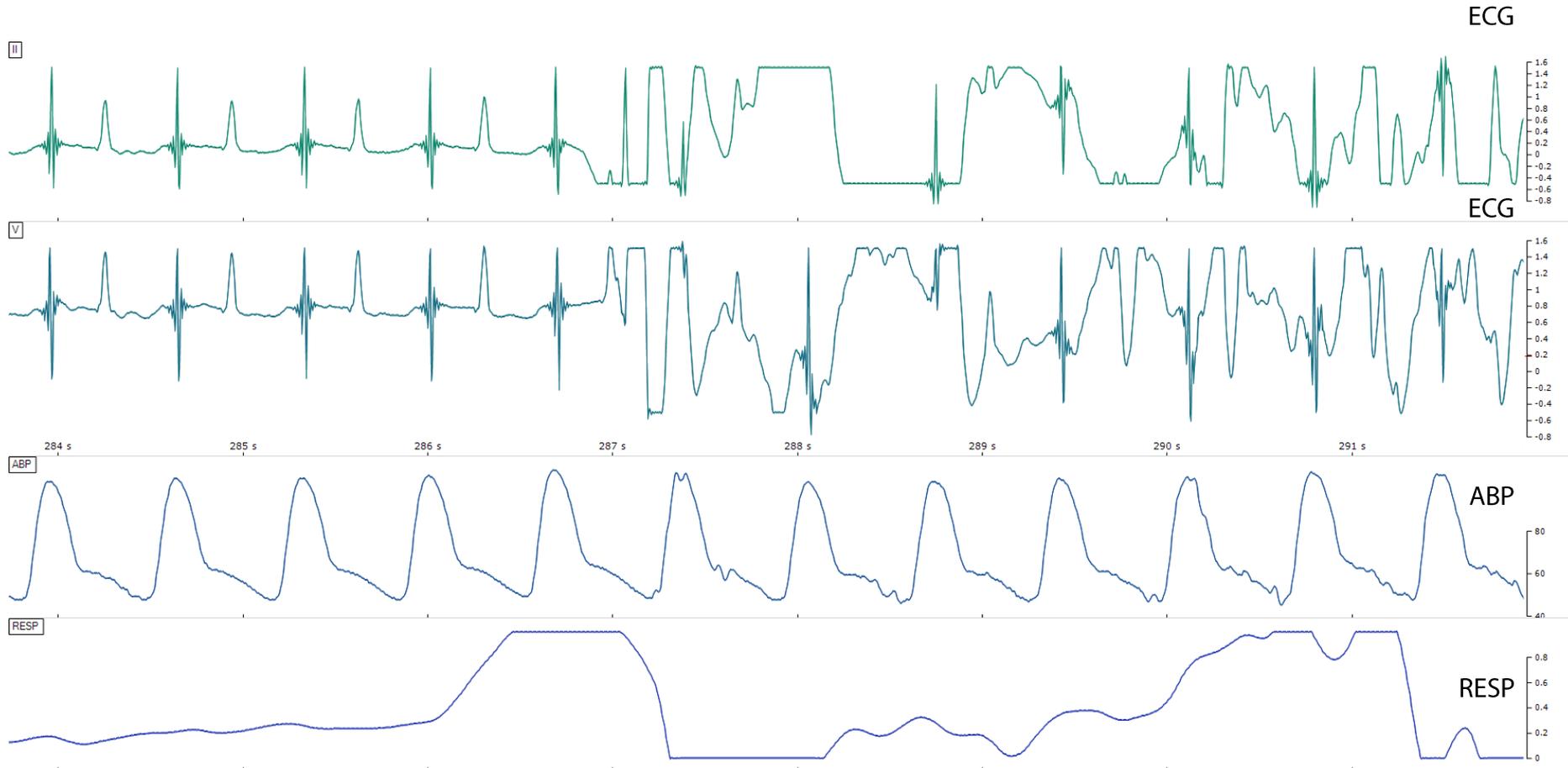
## Basic flow diagram

- A** <- Identify heavily damaged areas using FFT, descriptive statistics and histograms.
- B** <- Different beat-detection algorithms for ECG, ABP and PPG signals
- C** <- Regular heart activity test reveals some of false alarms
- D** <- Specific arrhythmia tests prove or deny alarm

## Signal validity (example 1 – statistical properties of 0.8 s window)

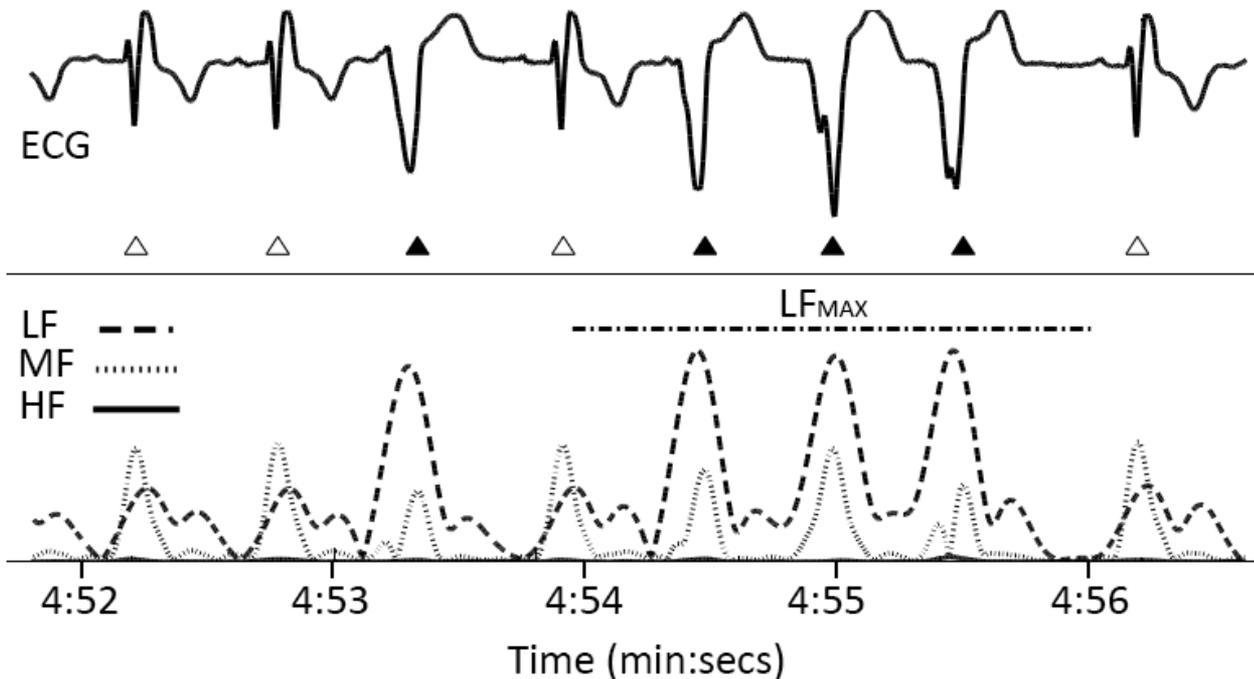


## Signal validity (example 2 – saturation, PCM)



## QRS detection from ECG

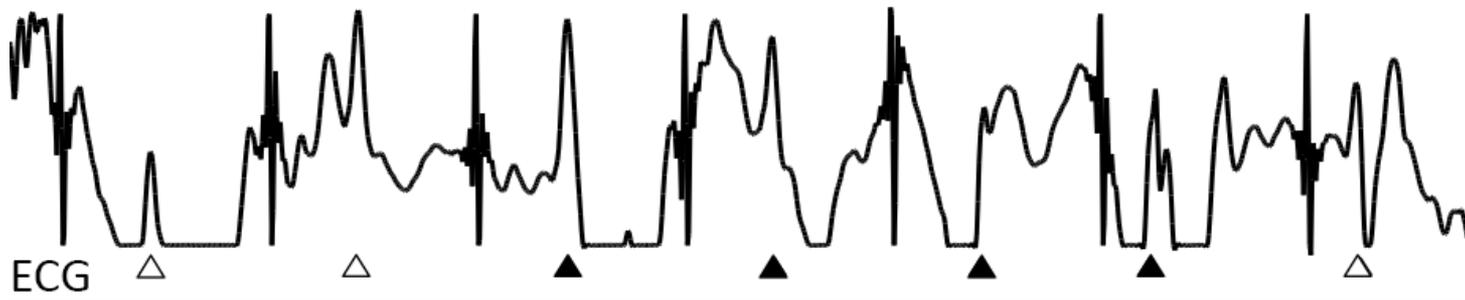
- Based on amplitude envelopes (frequency ranges: 1-8Hz, 5-25Hz, 50-70Hz)
- Automatically suppresses noise and pacemaker stimuli ( $R = MF - HF \times k$ ).
- Distinguish QRS type by LF/MF ratio (white triangles vs. black triangles)
- Output: QRS list with „**Group**“ and **max. low-frequency dominant time** (LF max)



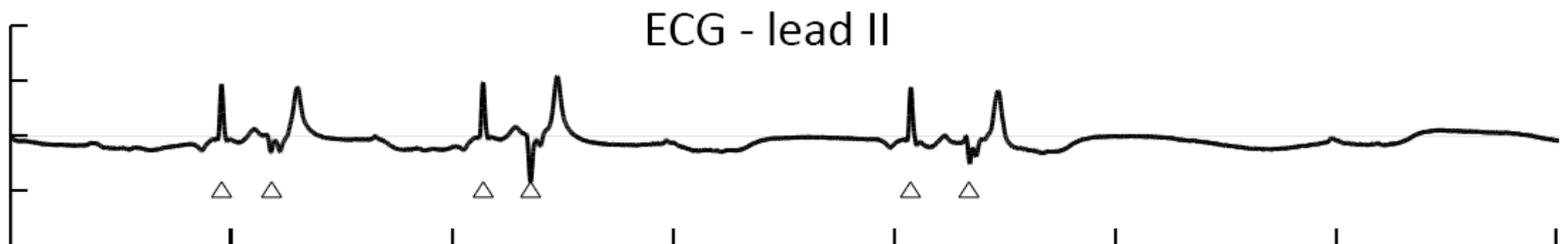
## QRS detection from ECG ... may be very challenging task

- Based on amplitude envelopes (frequency ranges: 1-8Hz, 5-25Hz, 50-70Hz)
- Automatically suppresses noise and pacemaker stimuli ( $R=MF-HF \times k$ ).
- Distinguish QRS type by LF/MF ratio (white triangles vs. black triangles)
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Example A

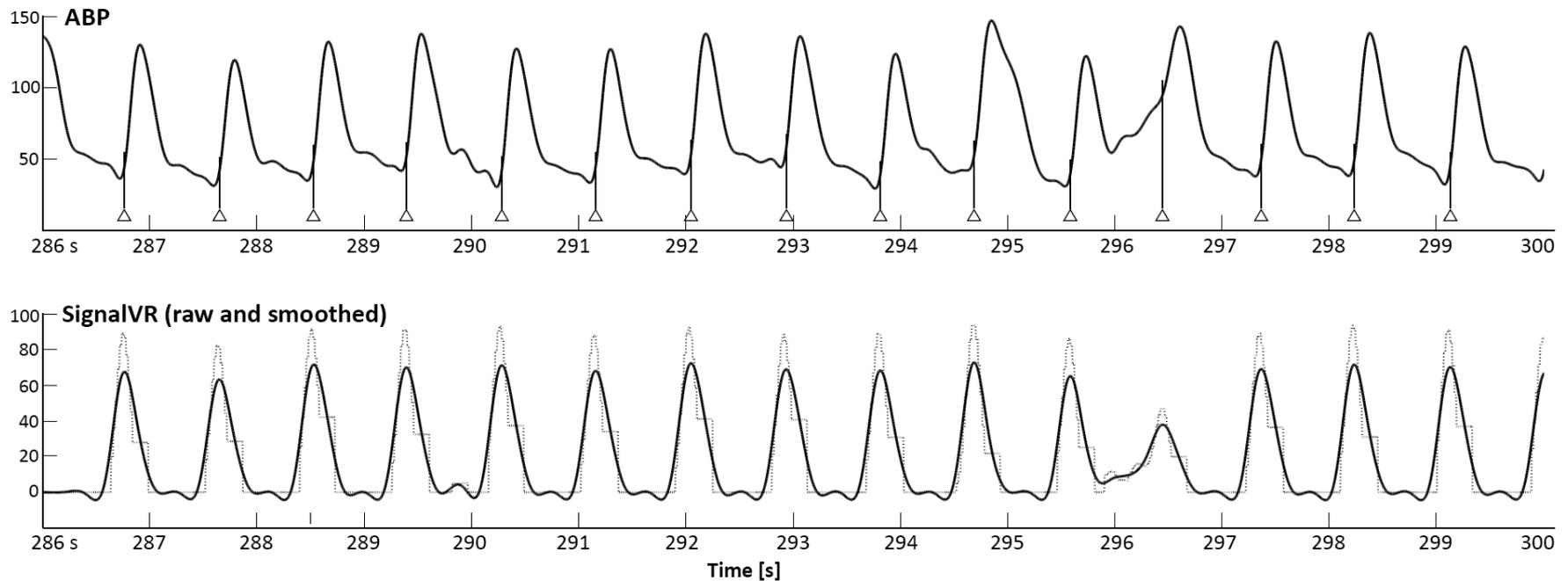


Example B



## Beat detection from ABP signal

Based on signal variation range in a 150 ms window and comparison of the left and right half of the window.



## Beat detection from ABP signal

Based on signal variation range in a 150 ms window and comparison of the left and right half of the window.

## Beat detection from PPG signal

Based on slope recognition on the left and right of local minima.

## Regular activity test

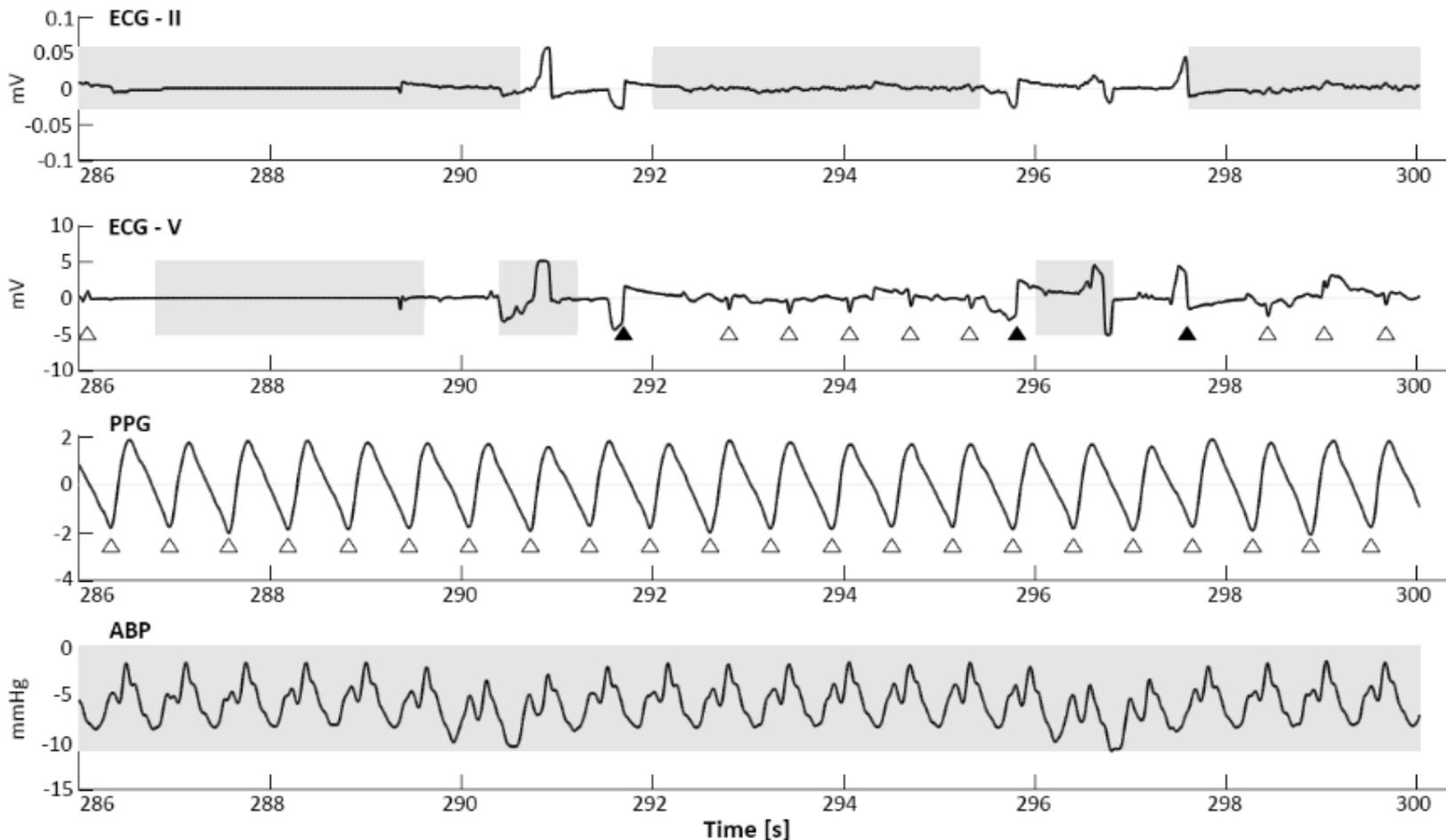
Independently for each channel:

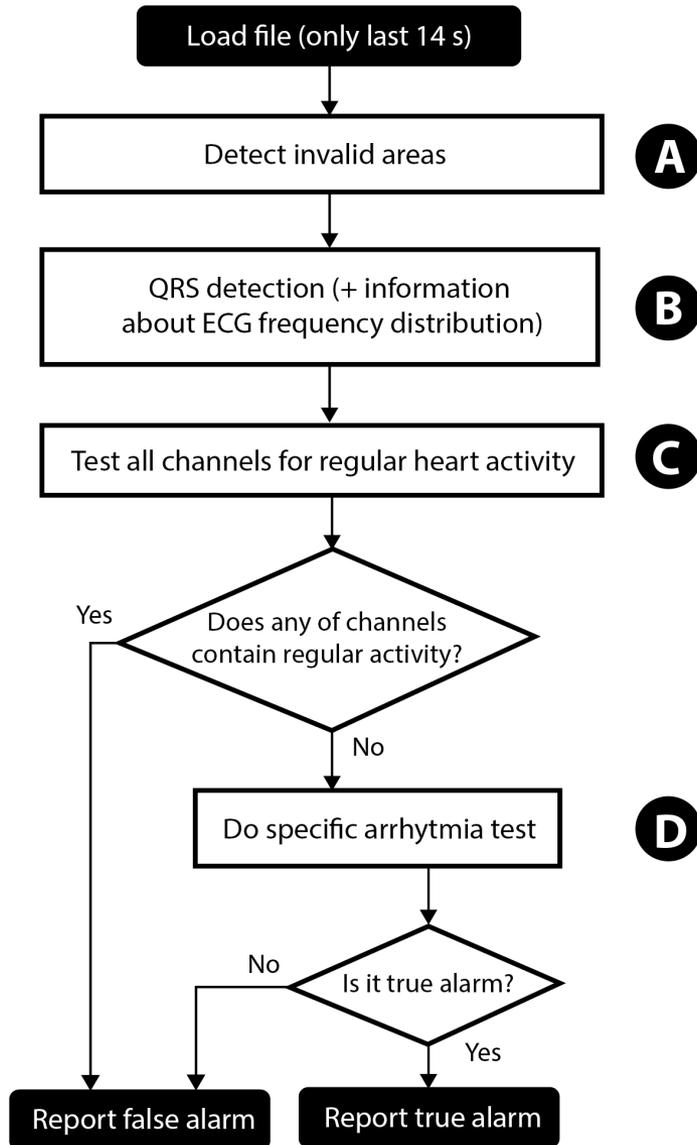
- Beat temporal distribution (count, standard deviation and time coverage)
- Mean heart rate
- Presence of low-frequency dominance
- Presence of invalid data

**Regular activity test filters out 35% of false alarms (training set)**

## Regular activity test (Independent across channels)

- QRS temporal distribution (count, standard deviation and time coverage)
- Mean heart rate; Presence of low-frequency dominance; only valid data



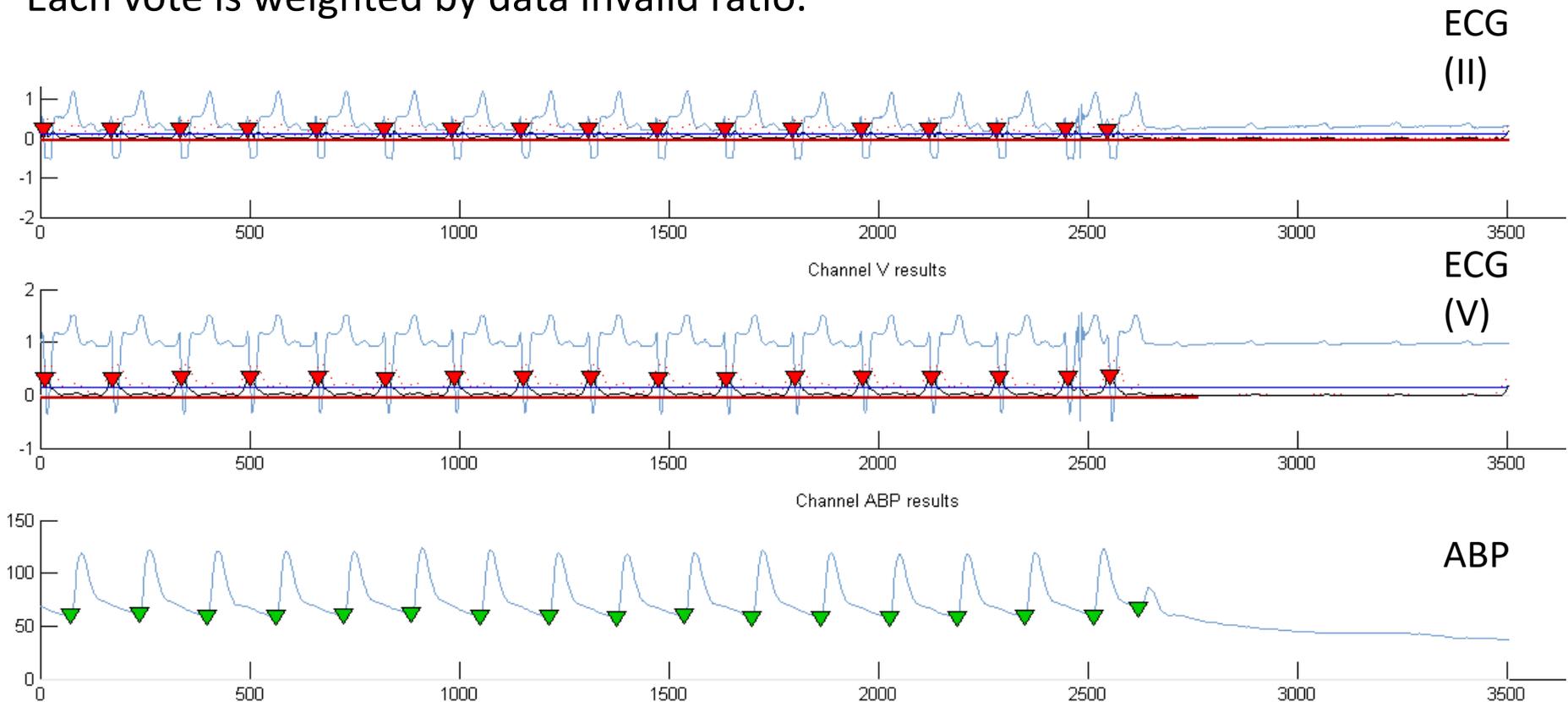


## Basic flow diagram

- <- Identify heavily damaged areas using FFT, descriptive statistics and histograms.
- <- Different beat-detection algorithms for ECG, ABP and PPG signals
- <- Regular heart activity test reveals some of false alarms
- <- Specific arrhythmia tests prove or deny alarm

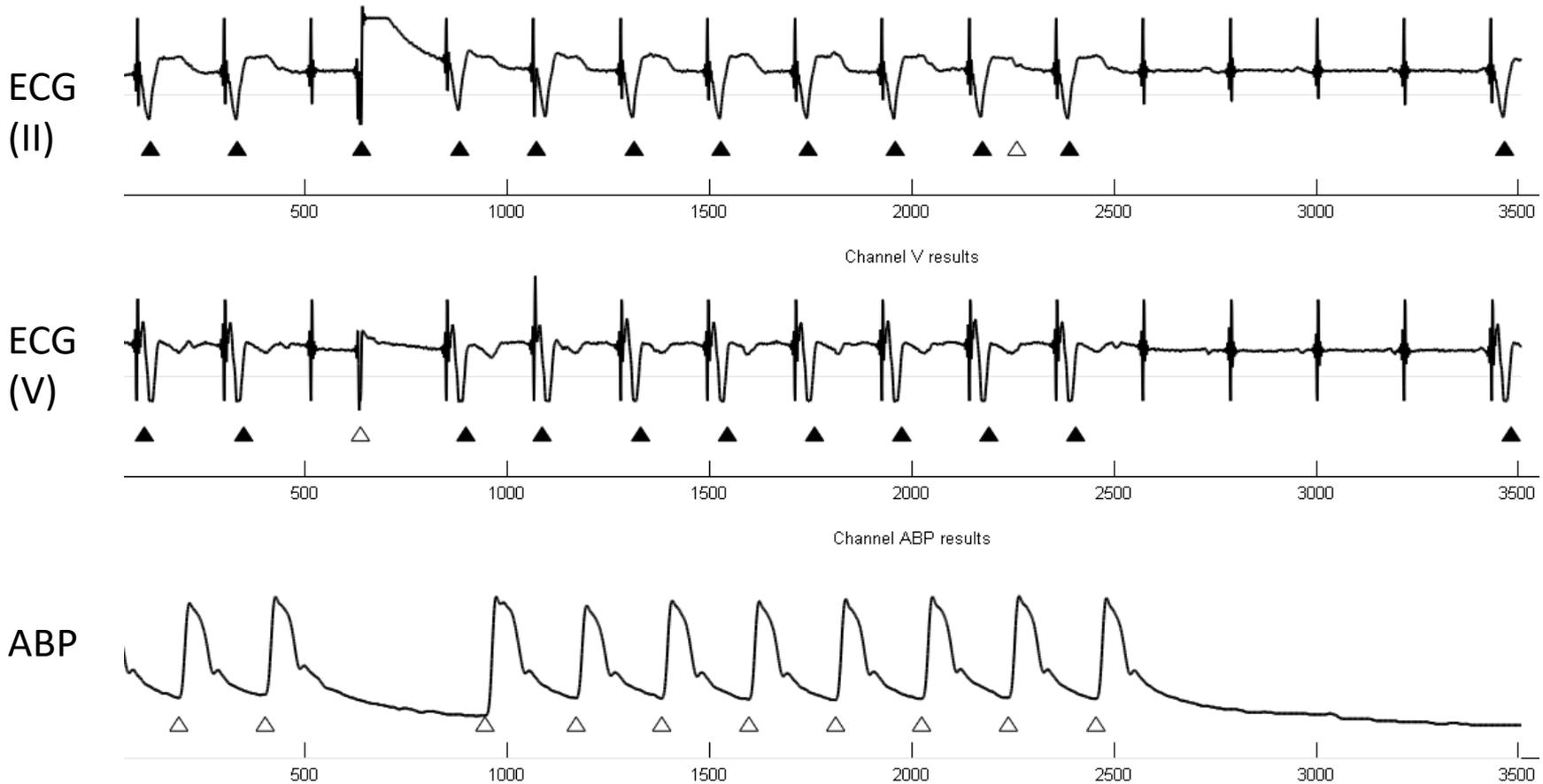
## Asystole test (semi-democratic elections)

Count QRSs in moving 3 secs window, then all channels vote for alarm presence. Each vote is weighted by data invalid ratio.



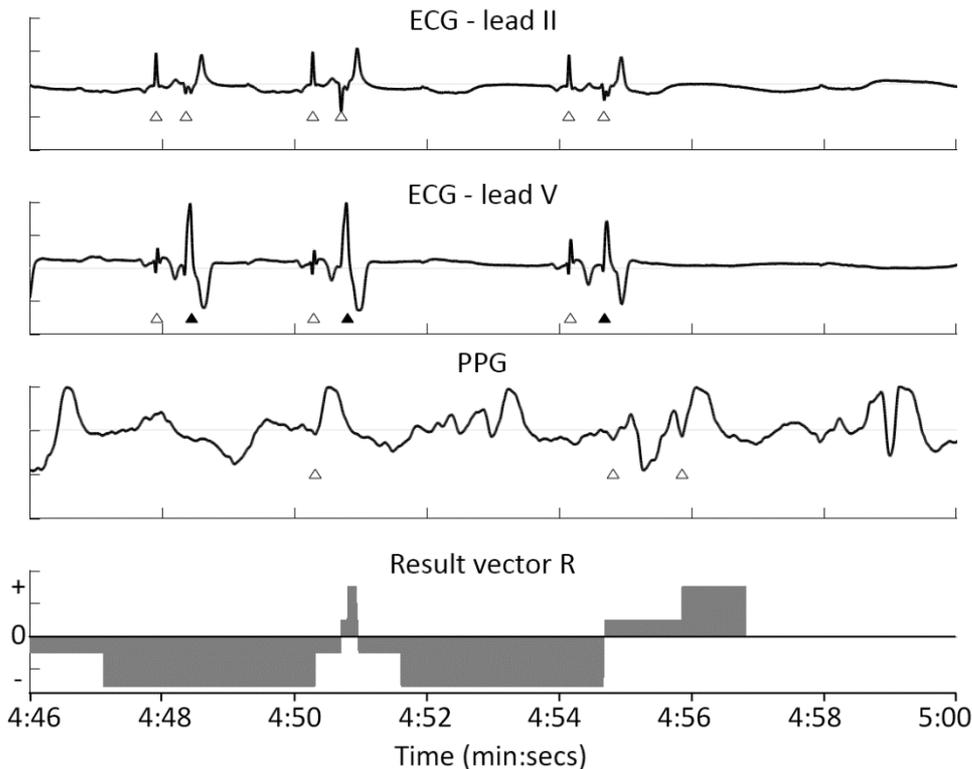
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Count QRSs in moving 3 secs window, then all channels vote for alarm presence. Each vote is weighted by data invalid ratio.



## Asystole results

### Training set

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Sensitivity  
**96%**

Specificity  
**92%**

### Test set

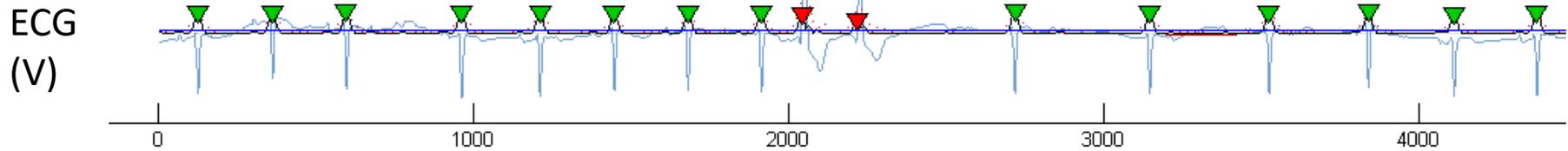
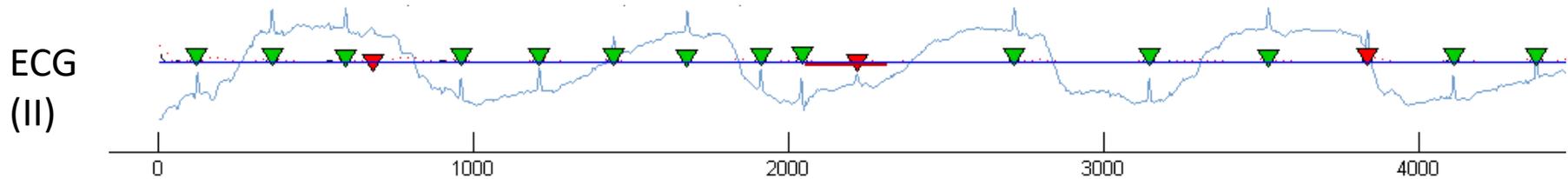
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Sensitivity  
**100%**

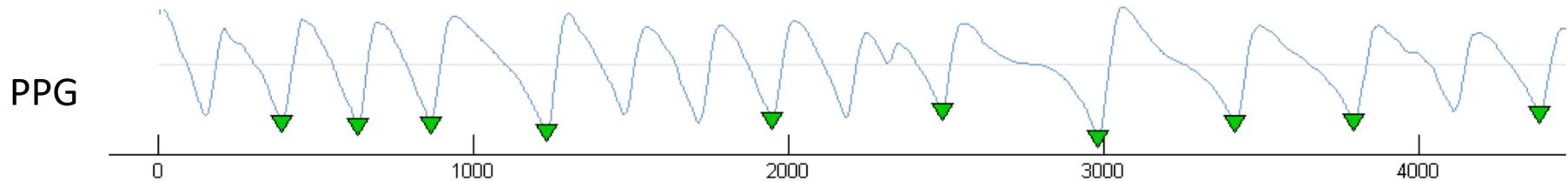
Specificity  
**97%**

## Bradycardia – heart is working too slowly

Channel II results



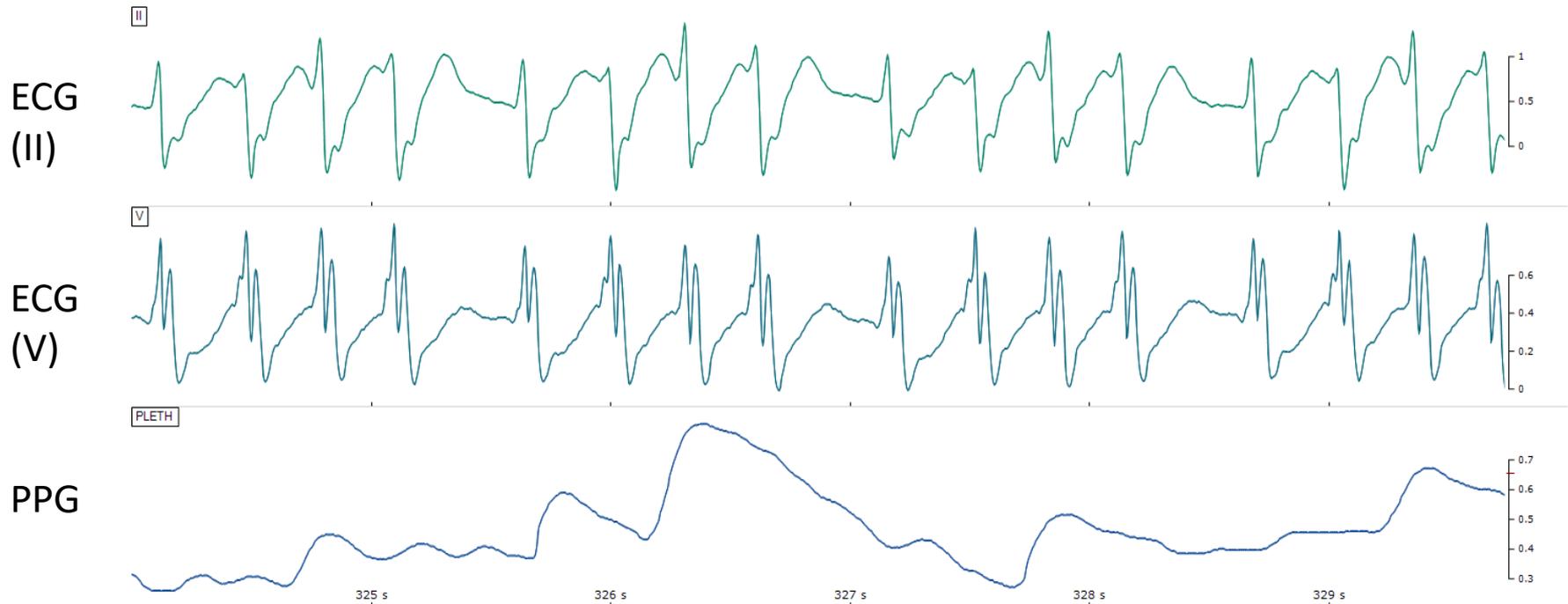
Channel PLETH results



Channel III results



## Tachycardia – heart is working too quickly



Compare ECG above with the „nice“ ECG:



## Bradycardia and Tachycardia tests

Choose the most reliable channel (using descriptive statistics) and compare heart rate to limits (**HR < 46 BPM** for bradycardia, **HR > 135** for tachycardia).

### Bradycardia results

#### Training set

---

Sensitivity	Specificity
<b>98%</b>	<b>74% (-&gt; 89%)</b>

#### Test set

---

Sensitivity	Specificity
<b>100% (-&gt;95%)</b>	<b>72% (-&gt;76%)</b>

### Tachycardia results

#### Training set

---

Sensitivity	Specificity
<b>99%</b>	<b>89%</b>

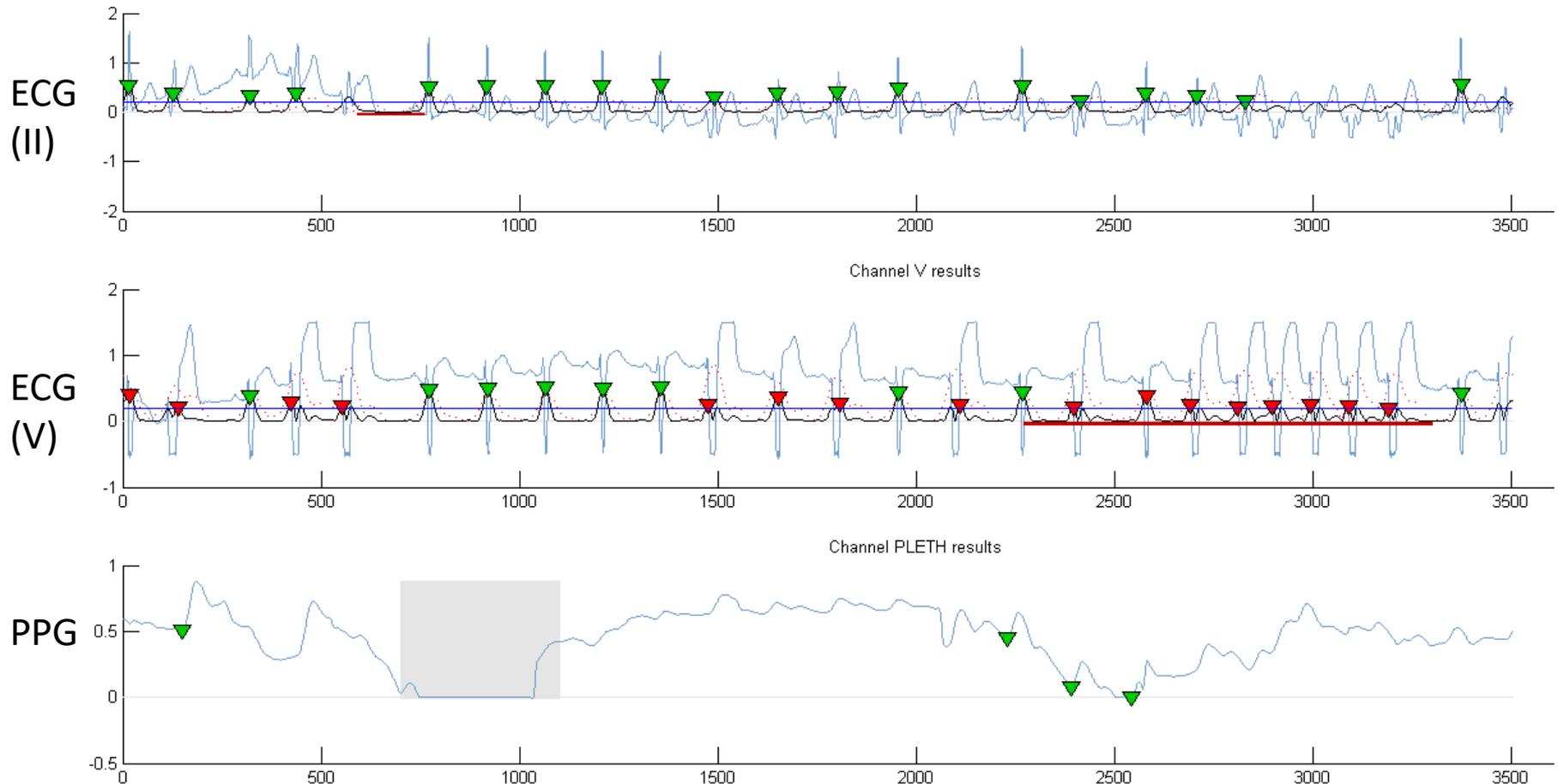
#### Test set

---

Sensitivity	Specificity
<b>97%</b>	<b>100%</b>

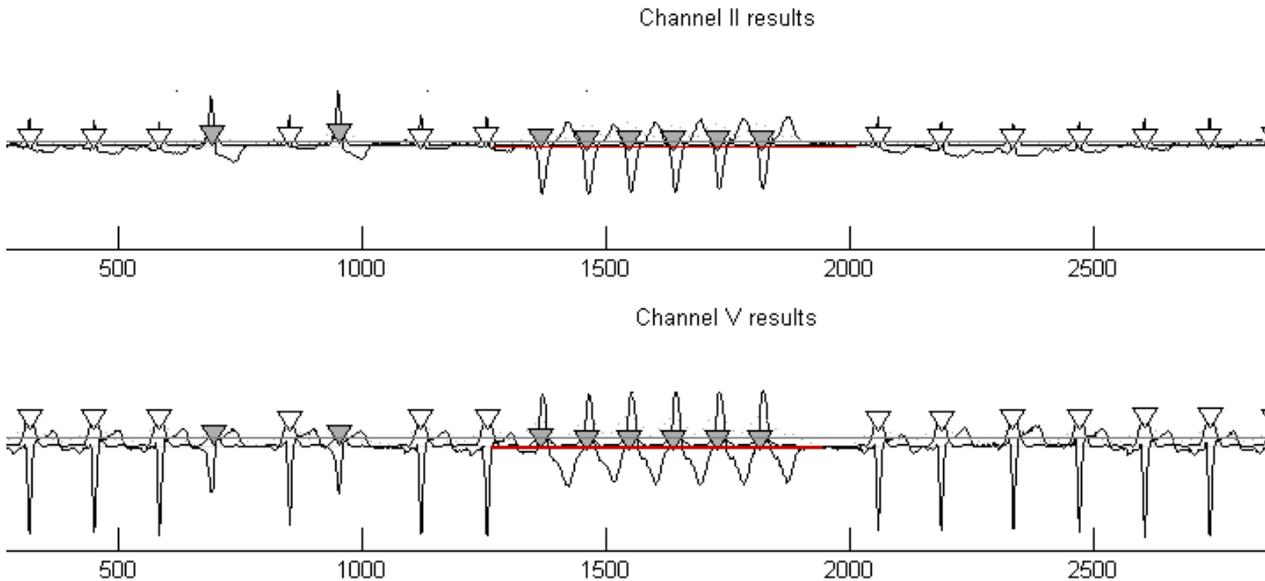
## Ventricle Tachycardia

Searchs sequence of 3 consecutive LF beats at heart rate  $> 95$  BPM using „Group“ information from ECG detection. Uses variation range of ABP signal.



## Ventricle Tachycardia test

Searchs sequence of 3 consecutive LF beats at heart rate > 95 BPM using „Group“ information from ECG detection. Uses variation range of ABP signal.



## Vent. tachycardia results

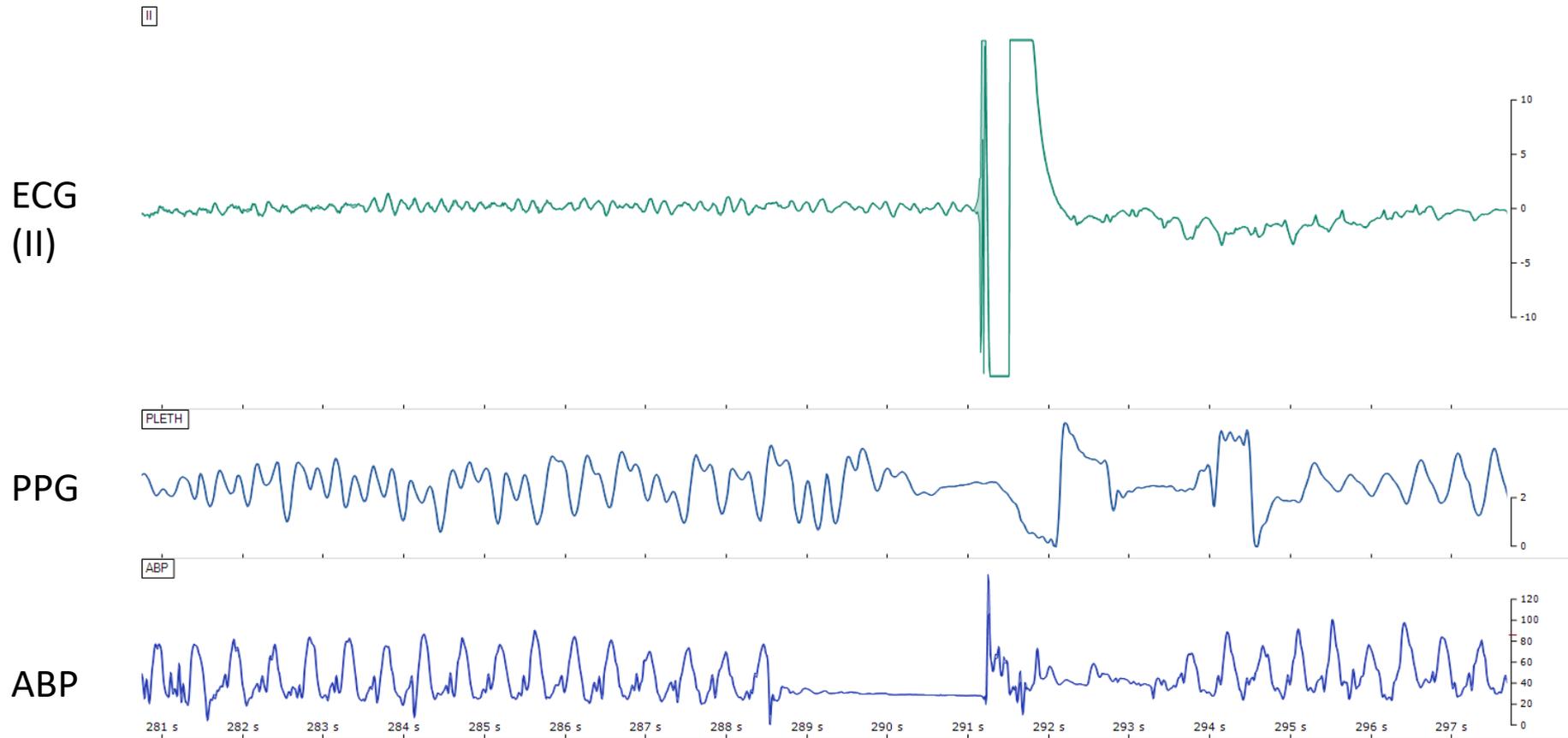
### Training set

Sensitivity	Specificity
<b>83%</b>	<b>82%</b>

### Test set

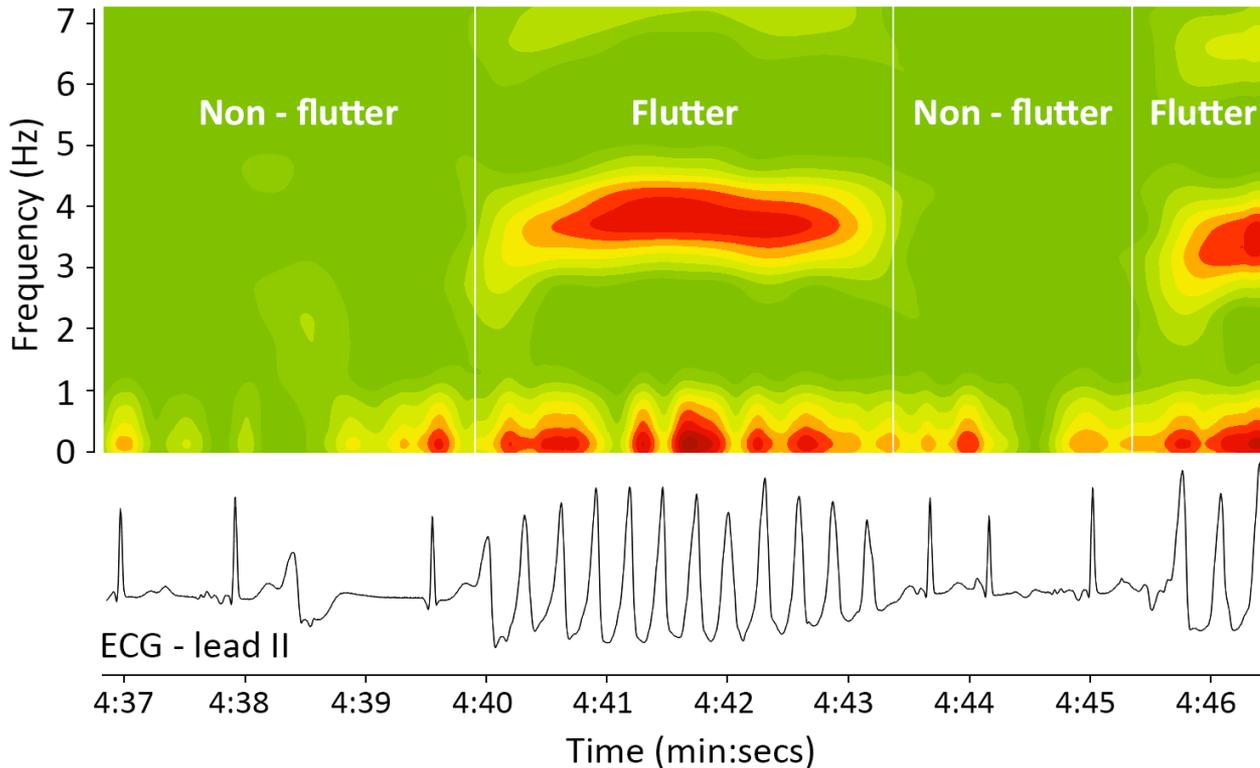
Sensitivity	Specificity
<b>85%</b>	<b>84%</b>

## Ventricle Fibrillation/ Flutter test



## Ventricle Fibrillation/ Flutter test

Finds out maximal frequencies in ECG spectra. ABP channel can raise an alarm.



### Flutter/ Fib. results

#### Training set

Sensitivity	Specificity
83% (->100%)	100%

#### Test set

Sensitivity	Specificity
67%	100% (->98%)

**Overall results – score 81.39 (real-time) and 84.96 (retrospective)**

## Training set results (750 records)

	Sens.	Spec.
Asystole	96%	92%
Bradycardia	98%	74%
Tachycardia	99%	89%
V. Tachycardia	83%	82%
Flutter/ Fib.	83%	100%
Average	96%	89%
Gross	93%	86%

## Test set results (500 records)

	Sens.	Spec.
Asystole	100%	97%
Bradycardia	100%	72%
Tachycardia	97%	100%
V. Tachycardia	85%	84%
Flutter/ Fib.	67%	100%
Real-time	92%	88%
Retrospective	95%	88%

## Physionet/CinC Challenge results - comparison with other competitors

Entrant	Event 1 (Real-time)			Event 2 (Retrospective)		
	TPR (%)	TNR (%)	Score	TPR (%)	TNR (%)	Score
Plešinger <i>et al</i> (2015)	92	88	<b>81.39</b>	95	88	84.96
Kalidas and Tamil (2015)	<b>94</b>	82	79.44	94	86	81.85
Krasteva <i>et al</i> (2015b) <sup>a</sup>	93	83	79.41 <sup>a</sup>	93	84	79.56 <sup>a</sup>
Couto <i>et al</i> (2015)	89	<b>91</b>	79.02	88	<b>92</b>	78.28
Fallet <i>et al</i> (2015)	<b>94</b>	77	76.11	<b>99</b>	80	<b>85.04</b>
Hoog Antink and Leonhardt (2015)	93	77	75.55	90	82	75.18
Eerikäinen <i>et al</i> (2015)	90	82	75.54	89	85	75.52
Ansari <i>et al</i> (2015)	89	84	74.48	89	87	76.57
Liu <i>et al</i> (2015)	89	79	71.68	93	78	75.91

+ 34 others

## **Thank you for your attention**

**[fplesinger@isibrno.cz](mailto:fplesinger@isibrno.cz)**

### **Acknowledgements**

This was supported by project no. P102/12/2034 from the Grant Agency of the Czech Republic and by MEYS CR (LO1212), its infrastructure by MEYS CR and EC (CZ.1.05/2.1.00/01.0017) and by ASCR (RVO:68081731).

## More information regarding this work:

### Easy, popular article (MathWorks website article):

Developing Detection Algorithms to Reduce False Alarms in Intensive Care Units

<https://www.mathworks.com/company/newsletters/articles/developing-detection-algorithms-to-reduce-false-alarms-in-intensive-care-units.html>

### Brief scientific paper (CinC conference paper):

False alarms in intensive care unit monitors: detection of life-threatening arrhythmias using elementary algebra, descriptive statistics and fuzzy logic

<http://www.cinc.org/archives/2015/pdf/0281.pdf>

### Detailed scientific paper (Full paper in the Physiological Measurement journal):

Taming of the monitors: reducing false alarms in intensive care units

<http://iopscience.iop.org/article/10.1088/0967-3334/37/8/E5/pdf>