



Heart Sound Segmentation: A Stationary Wavelet Transform Based Approach

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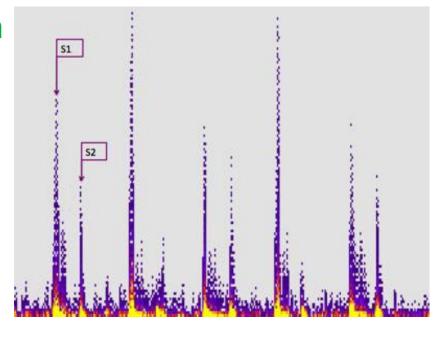
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Classifying Heart Sounds PASCAL Challenge

The challenge had 2 tasks:

Segmentation and Classification and Anomaly Detection

This work describes what i did in the first task: Segmentation and Classification

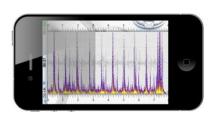


44100 Hz 20 auscultations

Datasets

4000 Hz 80 auscultations

iStethoscope



- Non-controlled environment
- No expert!
- Who was auscultated ?

Digiscope



- Controlled environement
- Done by expert!
- Auscultation were performed on infants exclusively!

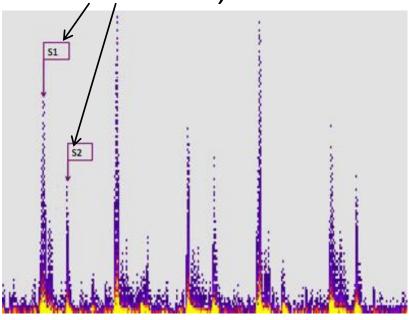
Heart Sounds

Normal Heart Sounds

Normal S1 and S2

Best with headphones.

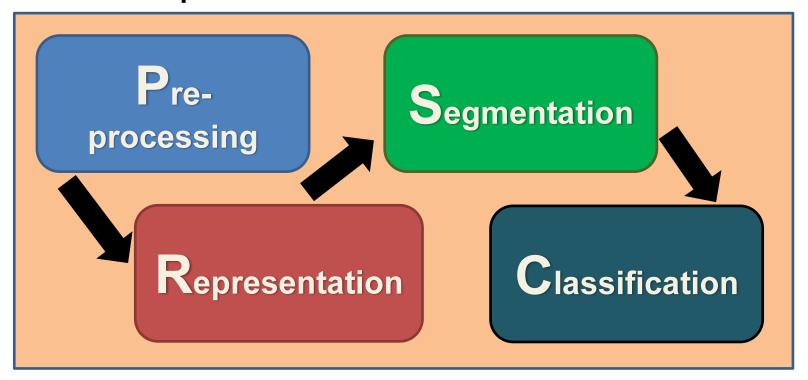
We want to detect and distinguish these two peaks! (which are the heart sounds!)



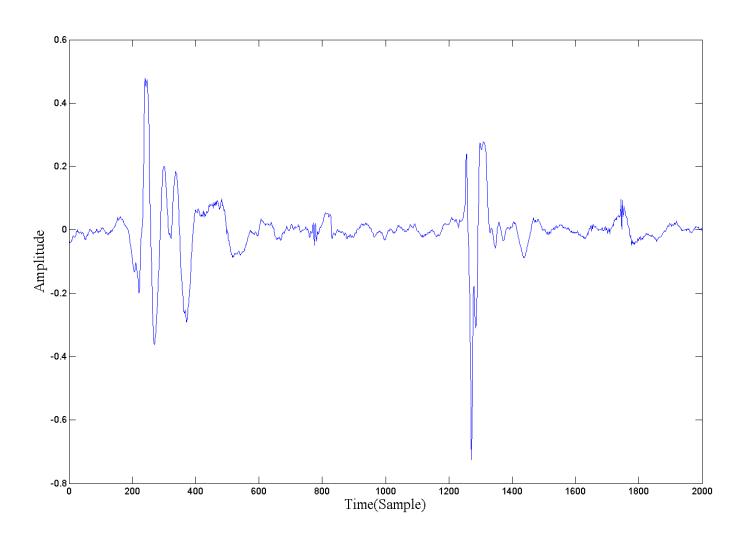
How do you detect and distinguish heart sounds?

Heart Sound Segmentation

Cardiac Segmentation algorithms can be successfully divided in 4 phases:

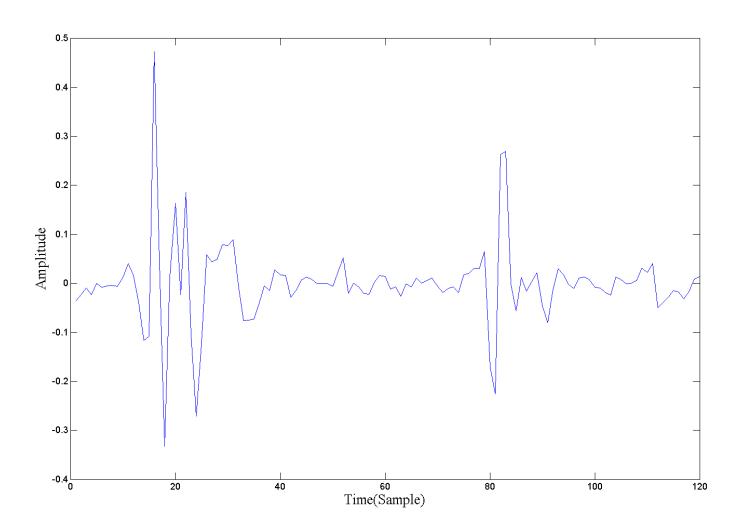


Preprocessing



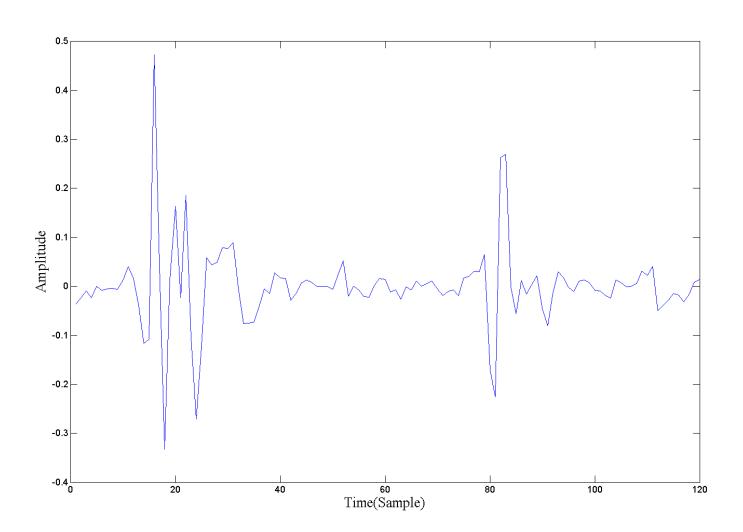


Preprocessing



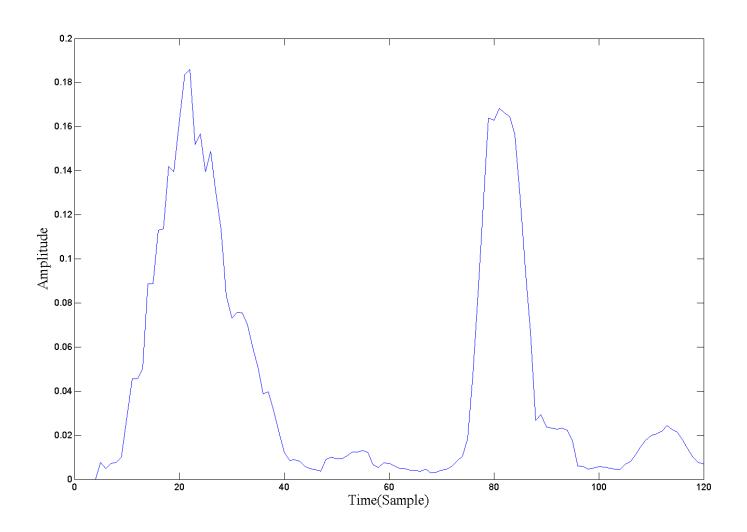


Representation

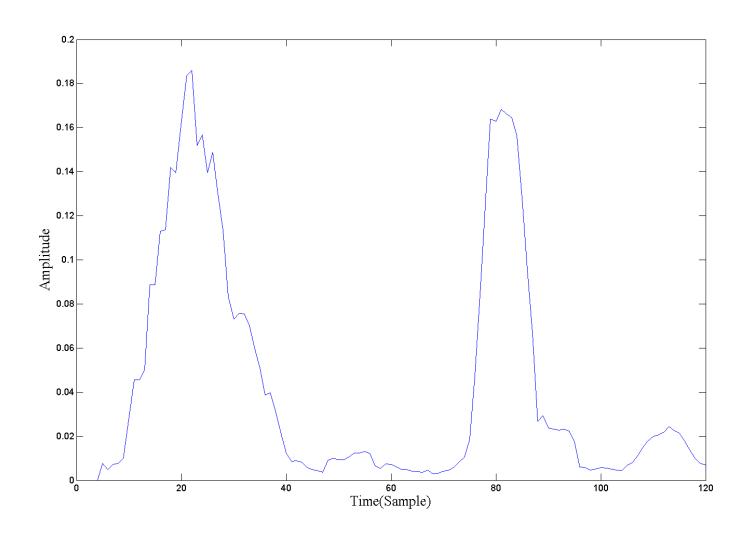




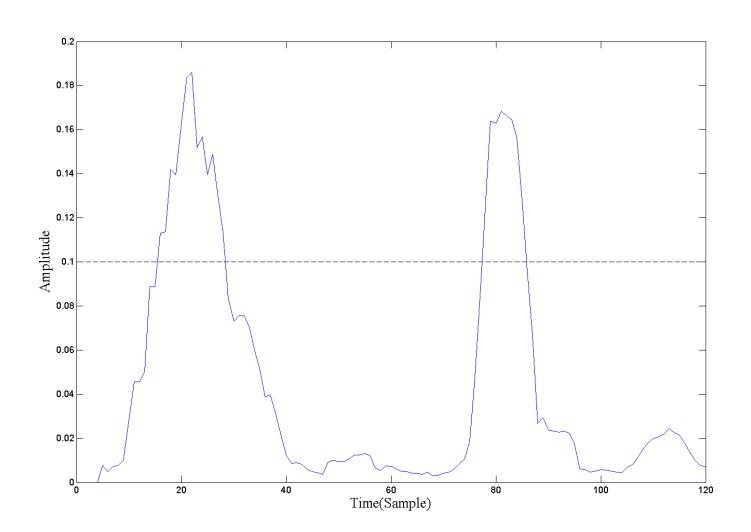
Representation



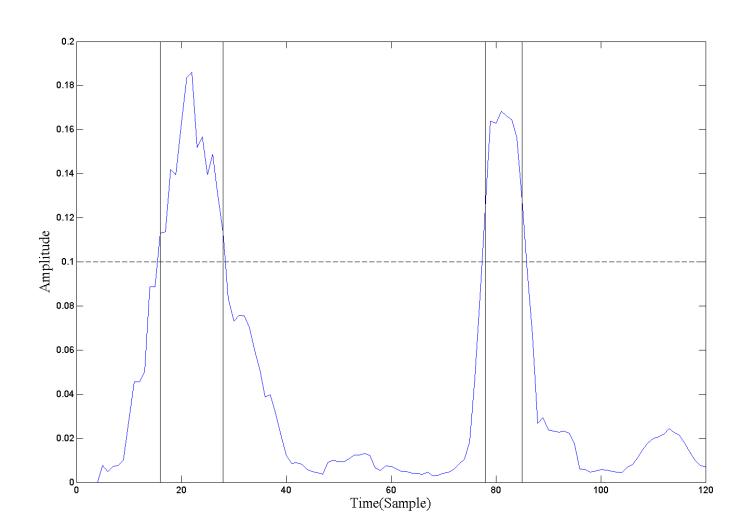




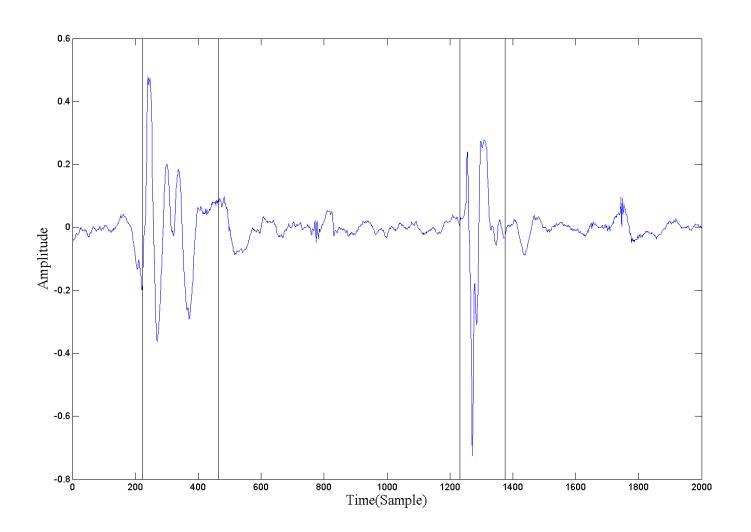






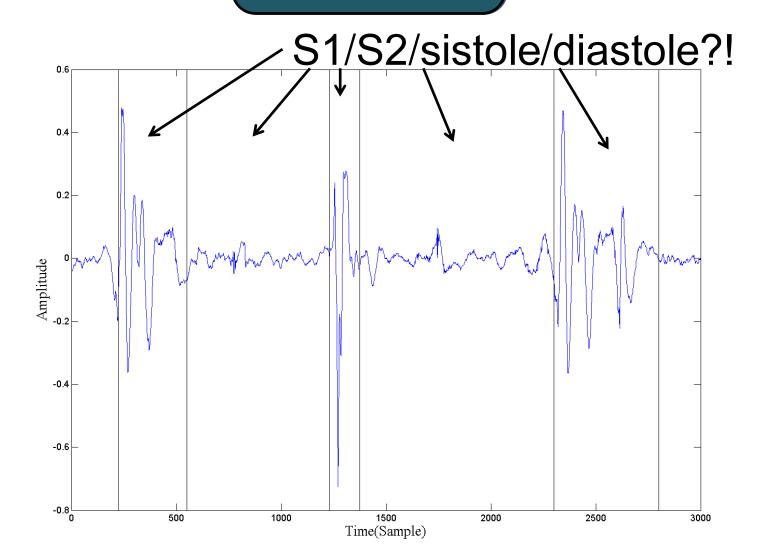






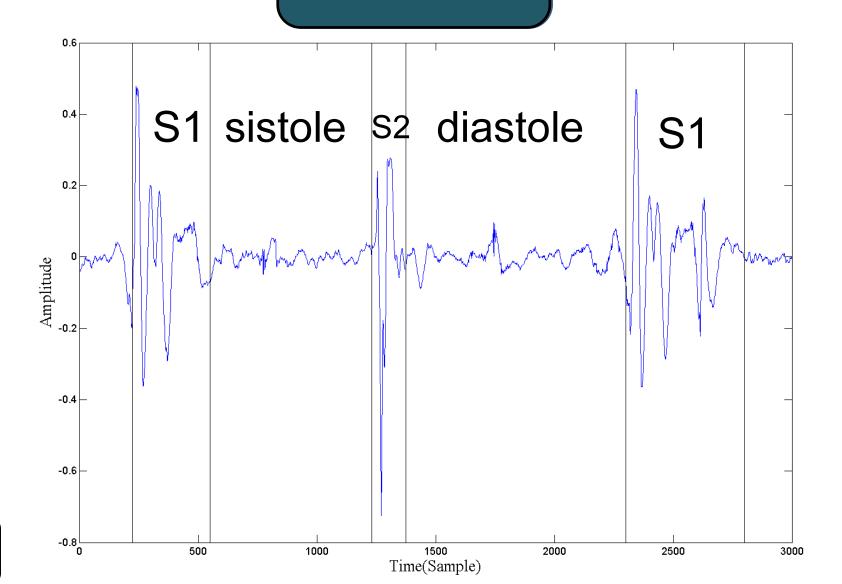


Classification



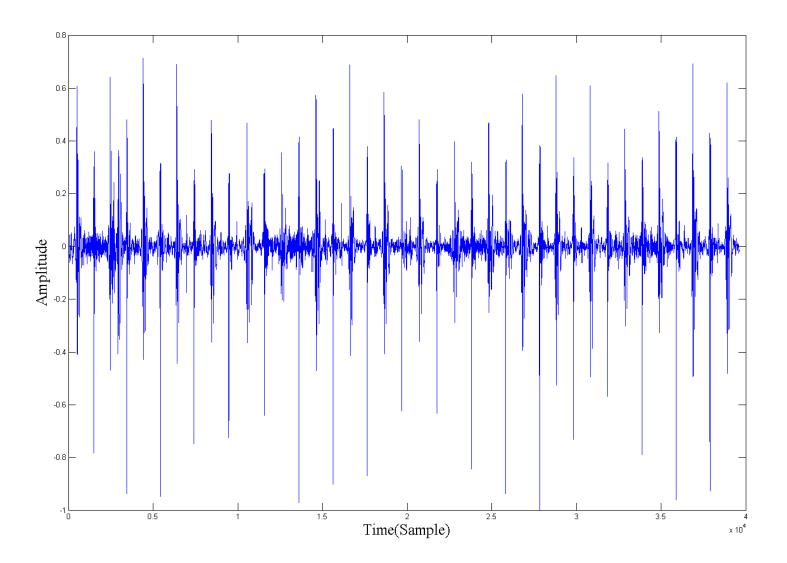


Classification



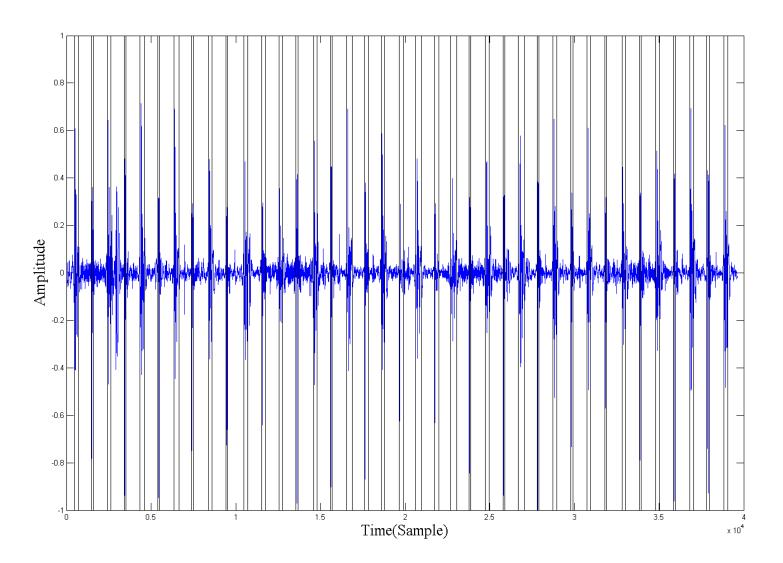


Manual Annotation



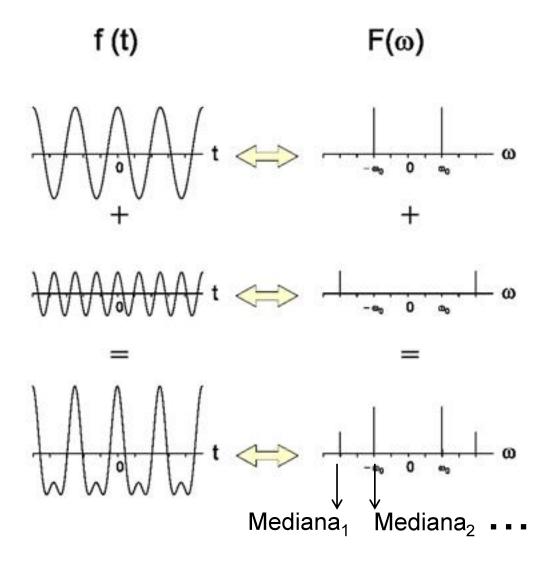


Manual Annotation



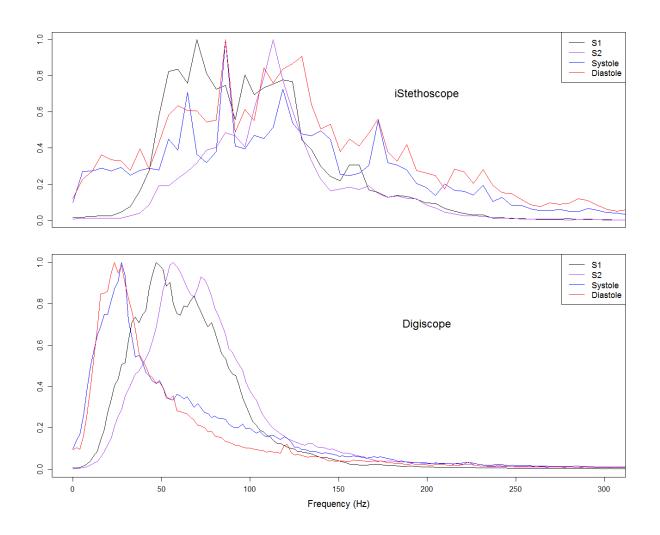


Fourier Transform



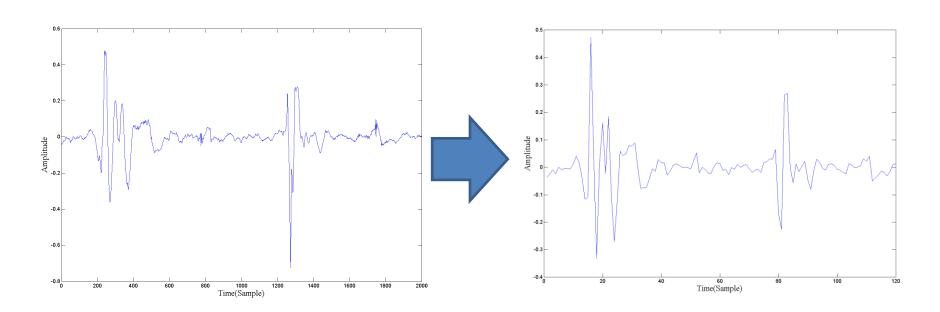


Spectral Analysis





Pre-Processing



Just downsampled iStethoscope!



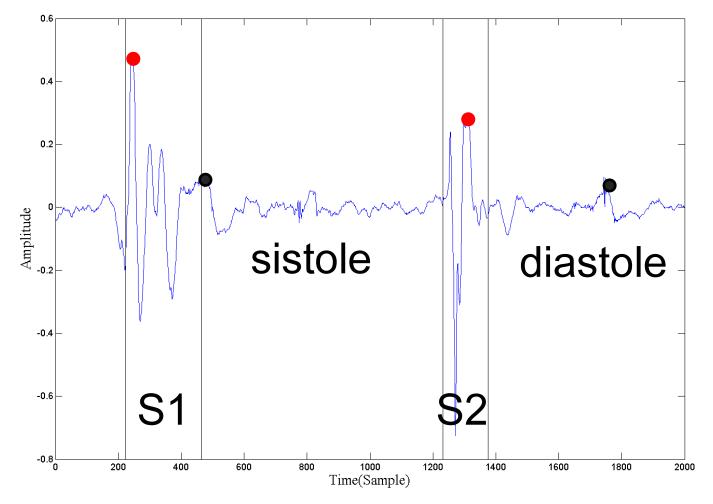
Representation

A good cardiac signal representation should have 2 characteristics g₁ e g₂



g₁. Accentuate the difference between S1/S2 and sistole/diastole

 $g_1 = |median(max(w_{S1,S2}(t))) - median(max(w_{Systole,Diastole}(t)))|$





Accentuate the difference between S1 and S2

 $g_2 = |median(max(w_{S1}(t))) - median(max(w_{S2}(t)))|$ 0.4 0.2 Amplitude -0.4 -0.6 -0.8 0

1000

Time(Sample)

1200

1400

1600

1800

2000



200

400

600

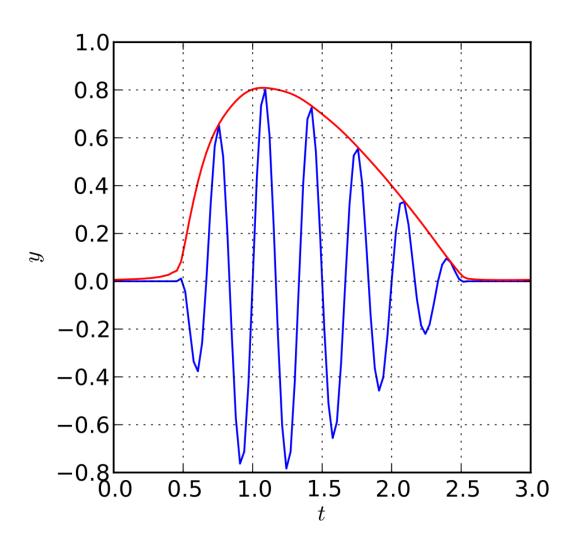
Representation

- Shannon Energy Envelope
- Shannon Entropy Envelope

Domínio do tempo

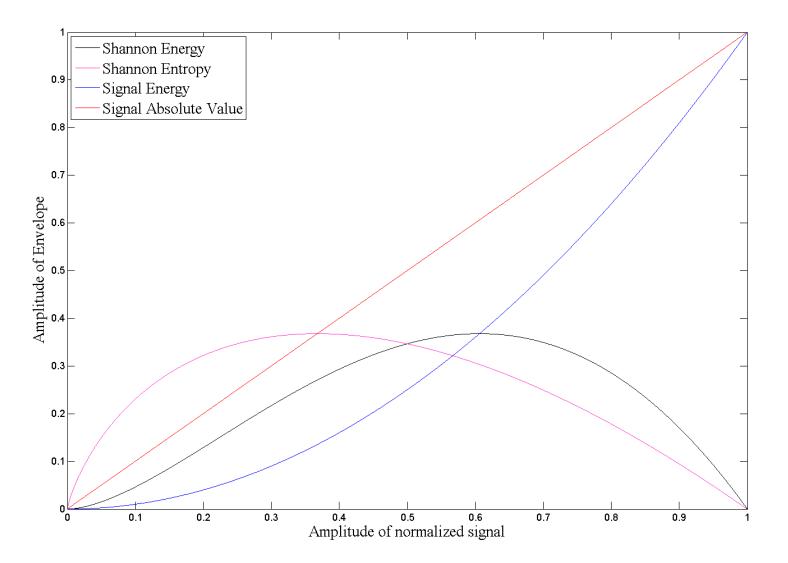


Shannon Energy Envelope





Shannon Energy/Entropy





Representations

- Continuous Wavelet Transform
- Discrete Wavelet Transform
- Stationary Wavelet Transform
- S-Transform
- Empirical Mode Decomposition
- Hilbert-Huang Transform

Time-Frequency Domain



Digiscope Results

Representation	Order	Scale	Coef	g_1	g_2
DWT	38	3	c_a	0,63	0,014
SWT	1	3	c_a	0,59	0,26
CWT	2	60(*)		0,57	0,22
S-T		380(*)		0,42	0,25
Original Signal				0,57	0,2875
Shannon Energy				0,70	0,18
Shannon Entropy				0,35	0,03
HHT				0,28	0,17
EMD				0,31	0,17
DWT	5	3	c_d	0,32	0,42
SWT	15	3	c_d	0,36	0,48
CWT	13	240(*)		0,40	0,50
S-T		500(*)		0,42	0,33



iStethoscope Results

Representation	Order	Scale	Coef	g_1	g_2
DWT	23	3	c_a	0,49	0,02
SWT	2	5	c_d	0,48	0,25
CWT	4	60(*)		0,49	0,29
S-T		500(*)		0,40	0,27
Original Signal				0,40	0,34
Shannon Energy				0,61	0,31
Shannon Entropy				$0,\!45$	0,09
ННТ				0,12	0,13
EMD				0,12	0,15
DWT	23	4	c_a	0,11	0,41
SWT	2	5	c_a	0,41	0,39
CWT	4	20(*)		0,31	0,41
S-T		380(*)		0,37	0,38

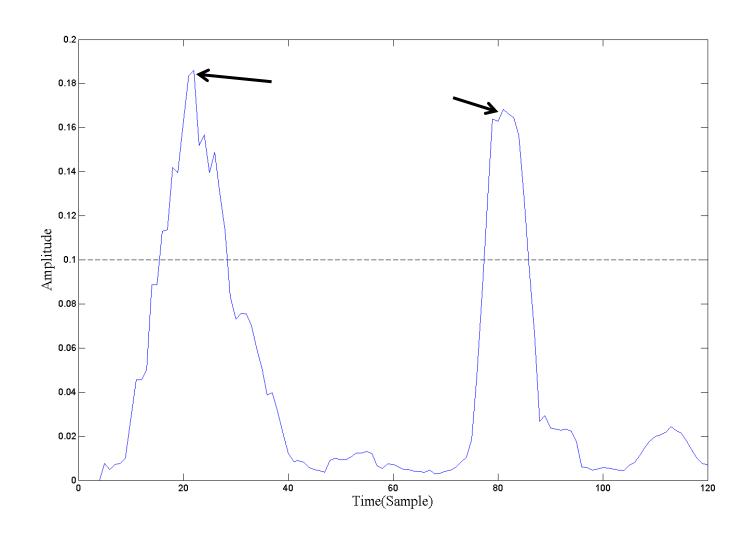


We can divide the Segmentation phase into 2 sub-phases:

- Peak Detection
- Boundary Detection

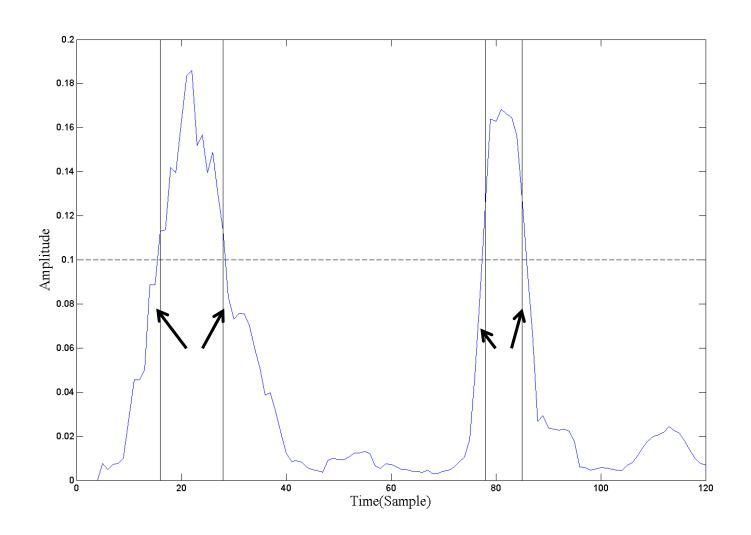


Peak Detection



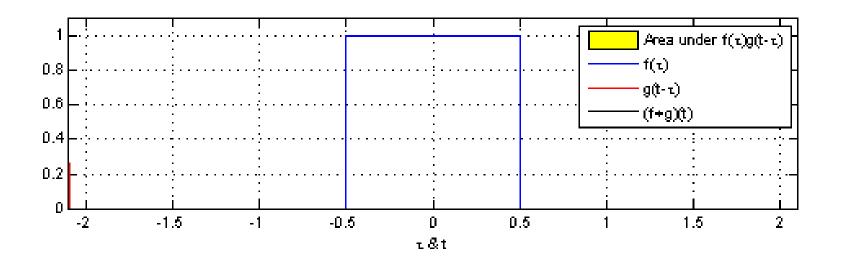


Boundary Detection



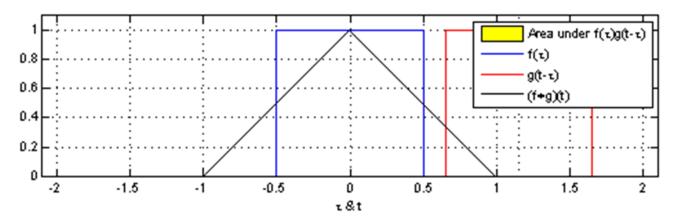


Convolution





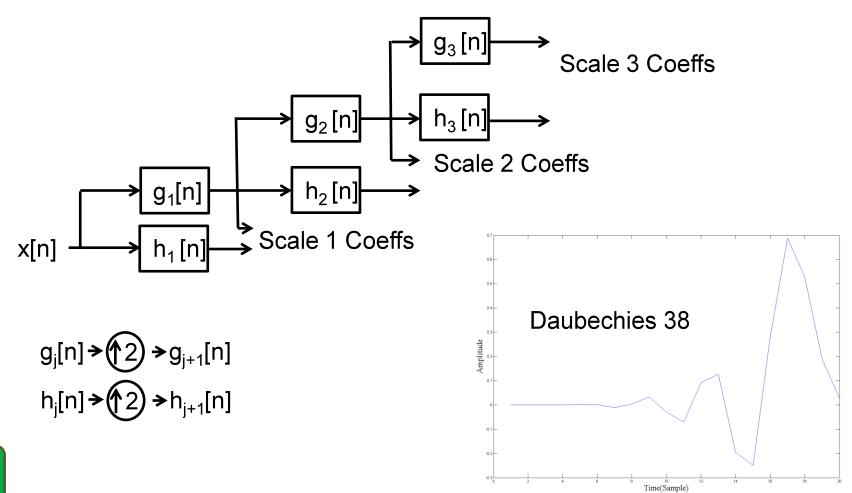
Idea!



Use a filter in the SWT that looks like the S1/S2 in order to determine their boundaries!

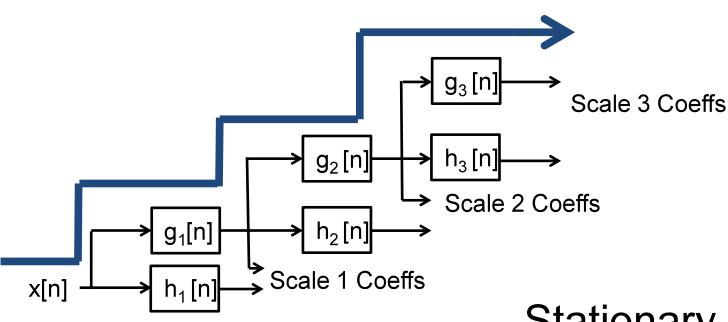


Stationary Wavelet Transform





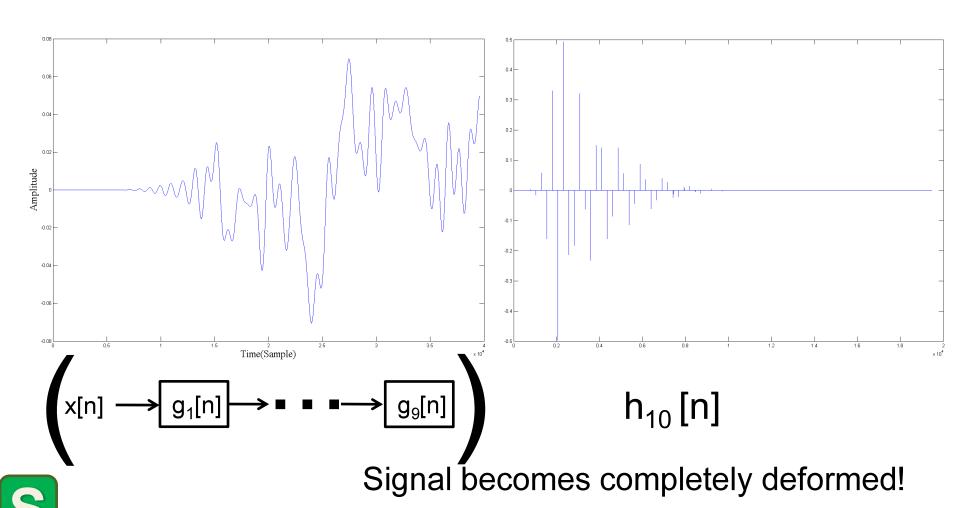
Problem



Stationary Wavelet Transform



Problem



Solution

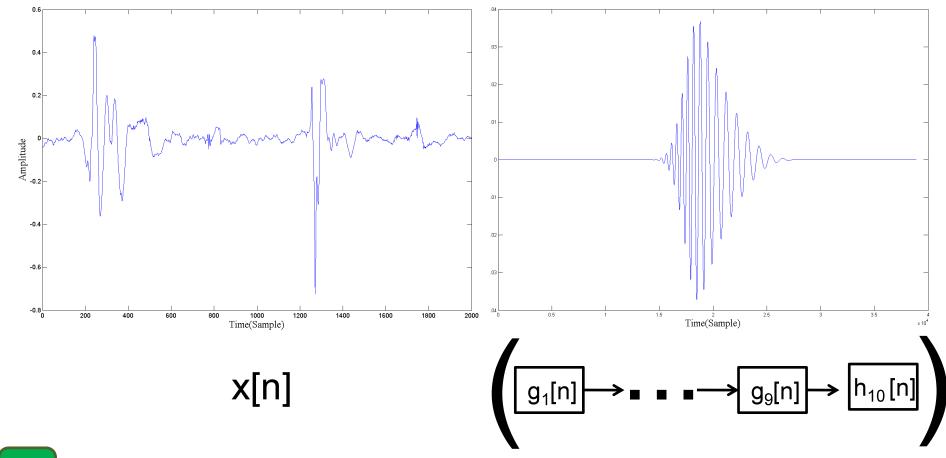
Lets use the Convolution's Associative Property!

$$x[n] \longrightarrow g_1[n] \longrightarrow \bullet \bullet \bullet \bullet g_9[n] \longrightarrow h_{10}[n]$$

$$x[n] \longrightarrow g_1[n] \longrightarrow \bullet \bullet \bullet \bullet g_9[n] \longrightarrow h_{10}[n]$$

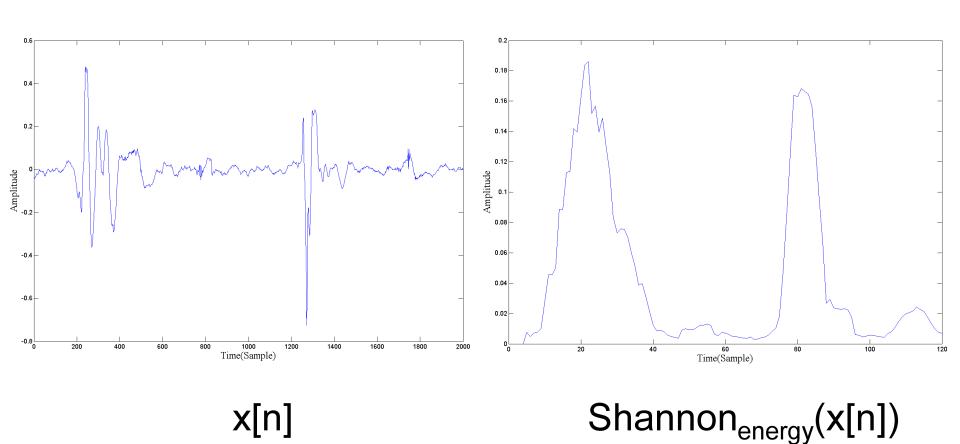


Solution



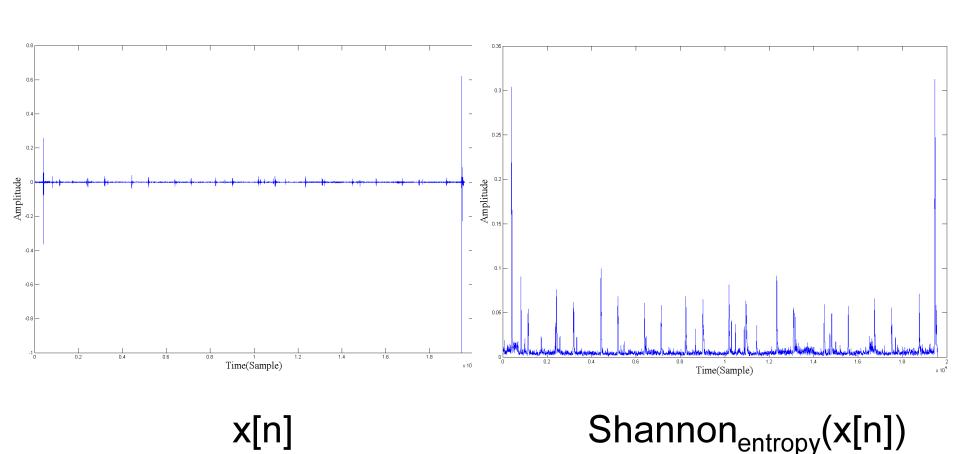


Signal Transformation: Digiscope



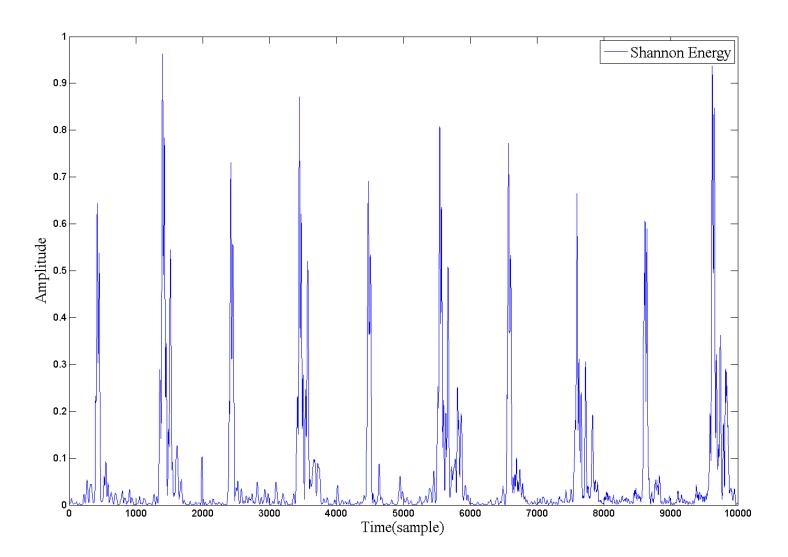


Signal Transformation: iStethoscope



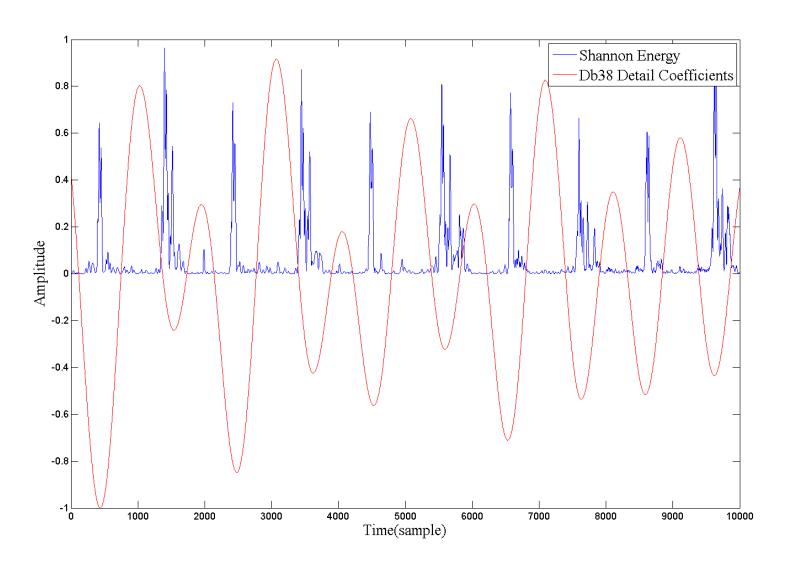


Shannon Energy



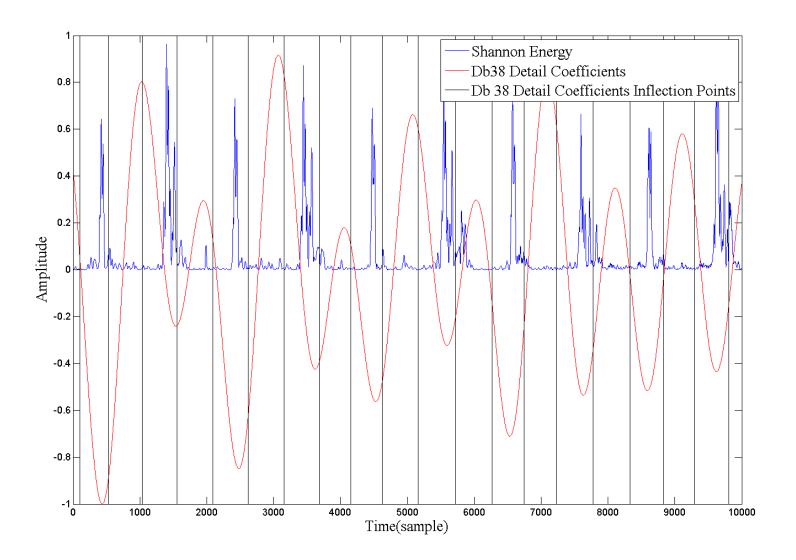


Wavelet Coefficients



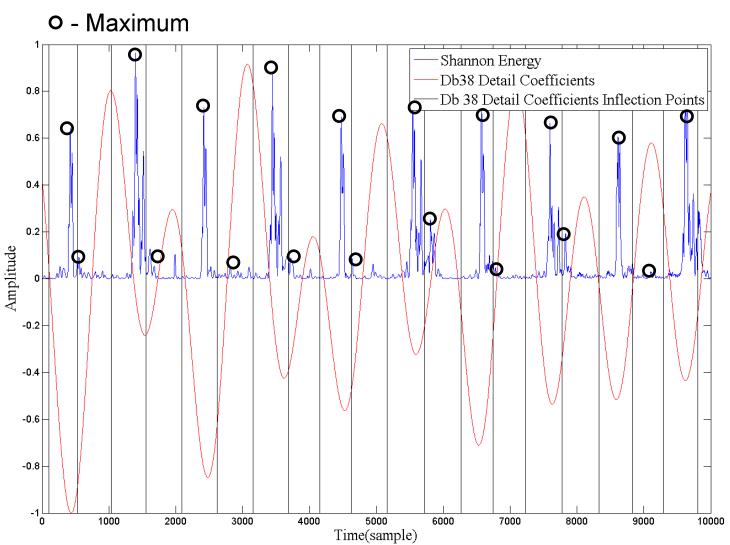


Inflection Points



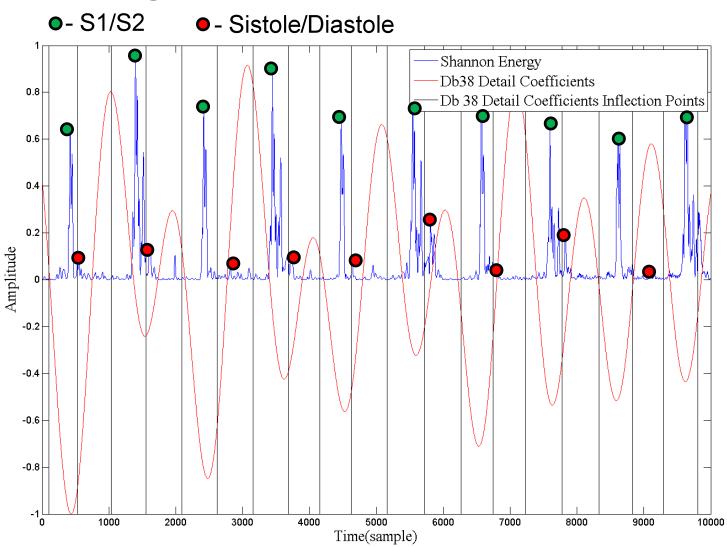


Segment Descriptors





Segment Descriptors





Dendrogram Second S1/S2 0.5 First S1/S2 Candidates 0.4 Candidates 0.3 0.2 0.1 First S1/S2 Candidates 0.5 -0.5 -1₀ 0.5 2.5 3.5 x 10⁴ First and Second S1/S2 Candidates 0.5 -0.5 -1₀ 0.5 1.5 3.5 2.5 x 10⁴



PASCAL Challenge Results

Approach	Total Error		
	Digiscope	iStethoscope	
Our Proposed Method	56732	706535	
Stanford	76444	1243640	
UCL	75569	3394378	
ISEP	72242	3905581	



Determining Boundaries

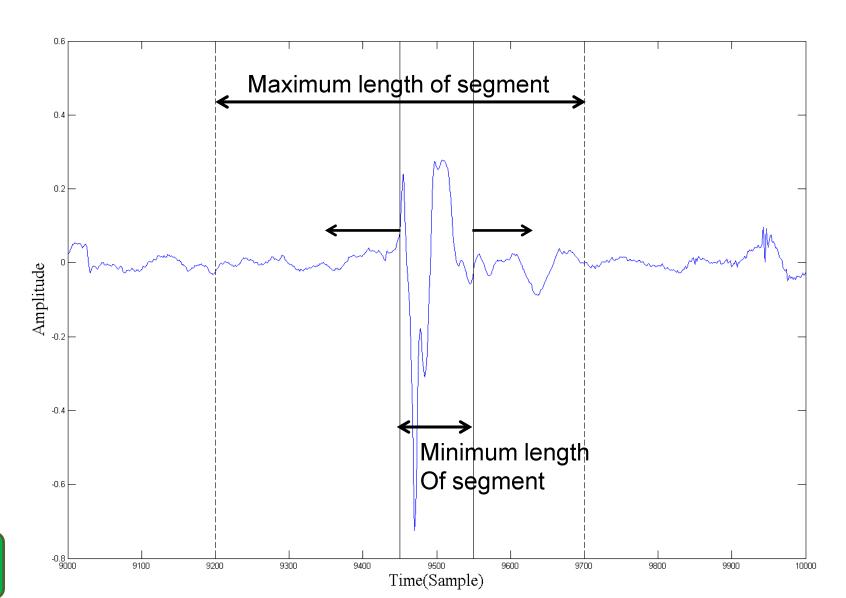


Determining Boundaries

- Variation between Segments
- Longest Increasing/Decreasing
 Sub-sequence

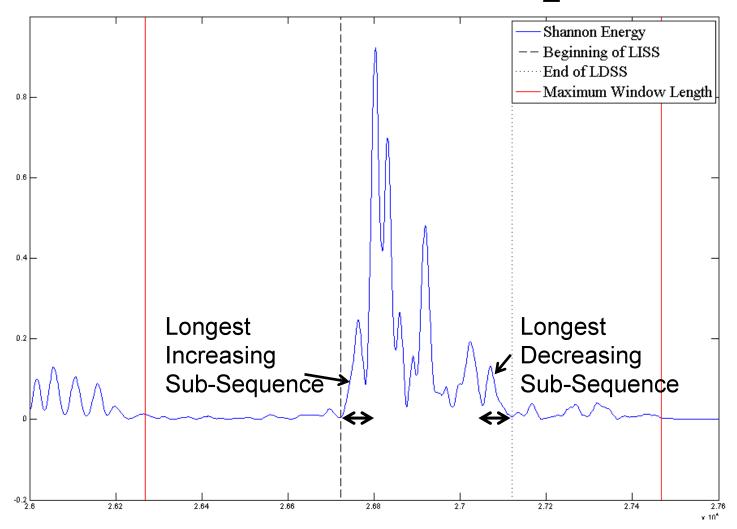


Variation Between Segments(a₁)



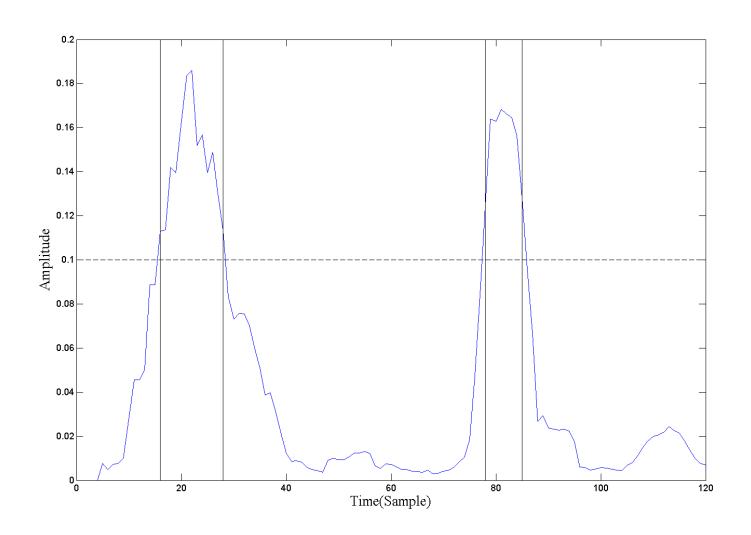


Longest Increasing/Decreasing Sub-sequence(a₂)





Baseline Method(a₃)





Results

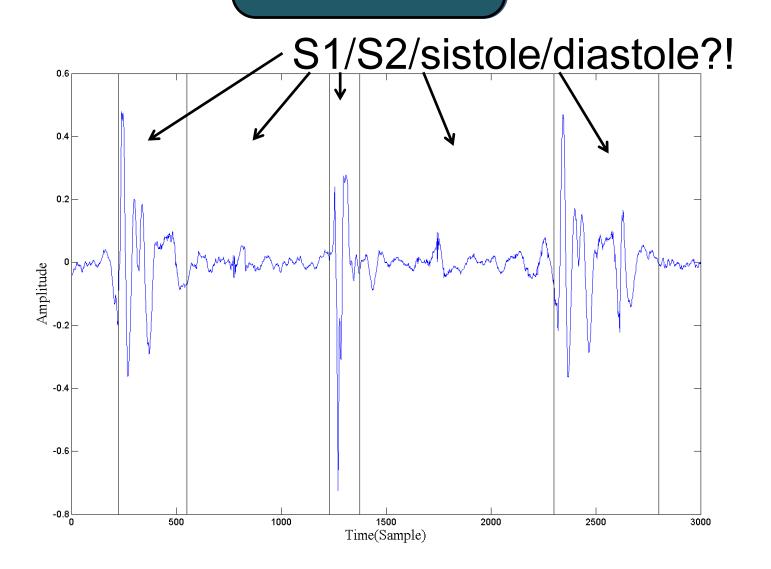
Approach	Annotation Error			
	Digiscope	iStethoscope		
a_1	$29,1\pm14,3$	$37,1 \pm 13,4$		
a_2	$41,4\pm\ 10,8$	$67,1 \pm 15,2$		
a_3	$46,8 \pm 15,2$	$83,2 \pm 20,4$		

Média +- desvio padrão (ms)

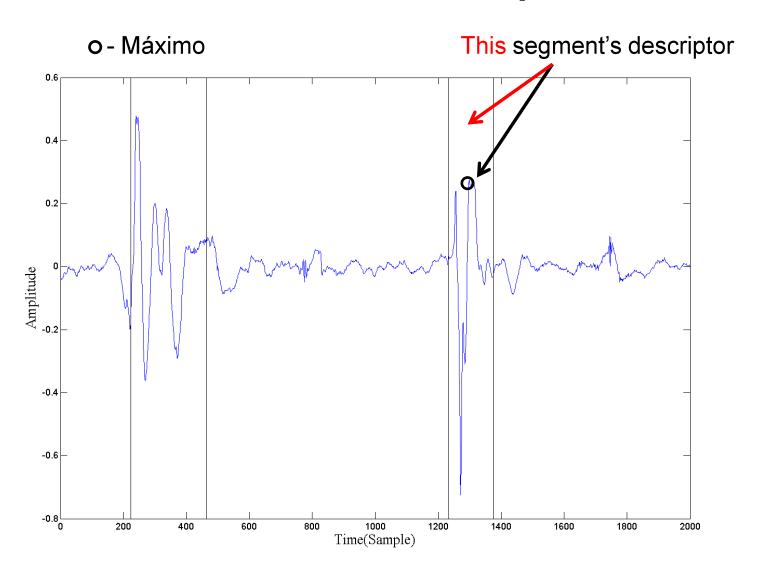


Classification

Classification

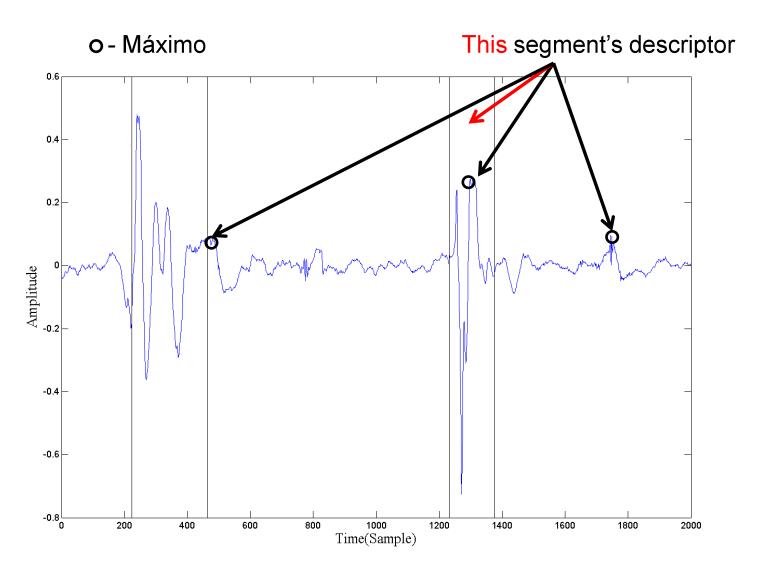


Individual descriptor



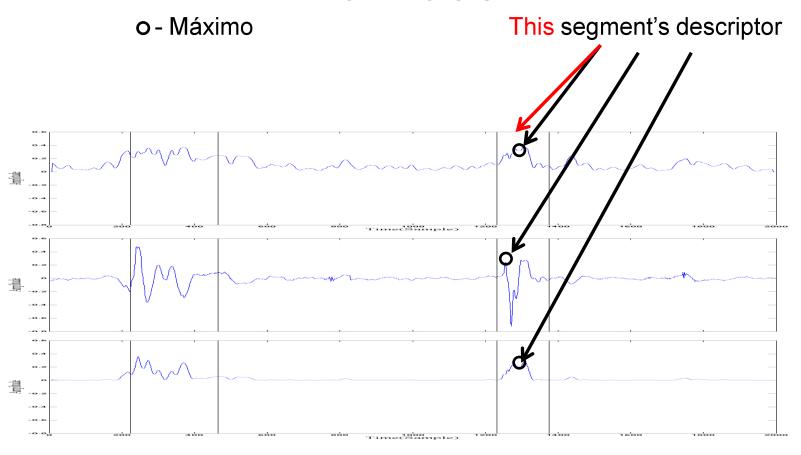


Expanded Descriptor



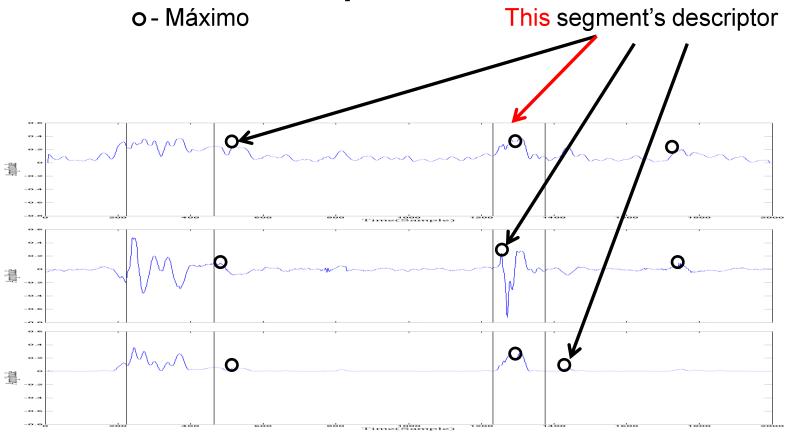


Combination of descriptors: Individual





Combination of descriptors: Expanded





Results: Combination of Descriptors

Type of Feature	Approach	Accuracy	Sensitivity	Specificity	
Individual	CWT+ST	0.86	0.88	0.84	
Neighbourhood	SWT+DWT+ST	0.83	0.86	0.80	
	Digiscope				
Individual	CWT+DWT+HHT+ST+EMD	0.90	0.91	0.89	
Neighbourhood	CWT+DWT+ST	0.92	0.90	0.94	
Istethoscope					



Conclusion

Conclusion

- Spectral Analysis
- Evaluation of different types of Representations
- New peak detection algorithm
- 2 new boundary detection algorithms
- Article publication in Computing in Cardiology 2013

Thank you!