

# Review of filtering methods for artifacts removal in ECG signals

Ivan Markovsky

School of Electronics and Computer Science  
University of Southampton

In collaboration with

S. Van Huffel (K.U.Leuven) and A. Amann (Innsbruck Medical Univ.)

## Notation

$$y = v + \alpha c \quad (*)$$

$y$  — measured signal       $v$  — desired signal

$c$  — “noise” signal,  $\|c\| = 1$        $\alpha$  — noise level

$$\text{SNR}(y, v) := 20 \log_{10} \left( \frac{\|v\|}{\|v - y\|} \right) \quad \text{— data SNR in db}$$

Two players:

- **Nature**: chooses  $c$ ,  $v$ ,  $\alpha$ , and constructs  $y$  according to  $(*)$
- **Estimator (we)**: given  $y$  and some partial knowledge of how nature chooses  $c$ ,  $v$ ,  $\alpha$ , find a good estimate of  $v$

## Outline

Introduction

Filtering methods

Evaluation of the methods

Reproducible research

## The filtering problem

Given  $y$ , find in real-time an approximation  $\hat{v}$  of  $v$ .



Choose the filter, so that the approx. error  $\|v - \hat{v}\|$  is “small”.

$$\text{SNR}(\hat{v}, v) = 20 \log_{10} \left( \frac{\|v\|}{\|v - \hat{v}\|} \right) \quad \text{— restored SNR in db}$$

## Filtering methods

Filtering is based on prior knowledge about  $v$ ,  $c$ ,  $\alpha$ .

- **Bandpass filtering**  $v$  and  $c$  are separated in frequency and the cutoff frequency  $f_0$  is known
- **Kalman filtering**  $\text{col}(v, c)$  is modeled as an ARMA(X) process  $M$  and that model is known
- **Adaptive filtering** uses an additional signal  $u$  that is well correlated with  $c$  but uncorr. with  $v$

## Biomedical context (cont.)

We study the first question empirically.

A database of separately recorded  $v$ ,  $c$ ,  $u$  signals allows us to

1. **verify what properties  $v$ ,  $c$ ,  $u$  have**
2. **test the methods** (by constructing  $y$  and checking the SNR improvement  $\text{SNR}(\hat{v})/\text{SNR}(y)$ )

The second question is currently unexplored.

We use the default **2-norm**, however, can choose weights that make the result medically relevant.

Which aspects of the signal are important for the doctor?

## Biomedical context

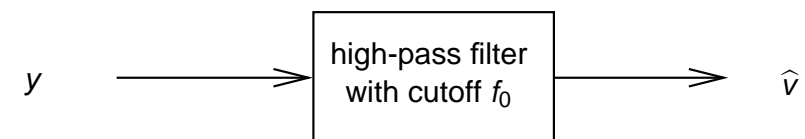
Meaning of the signals:

- $v$  — **pure ECG** signal
- $c$  — **artifact** on the ECG caused by resuscitation
- $u$  — **arterial blood pressure** signal  
 $u$  is “well” correlated with  $c$  and is measured

Main questions from the application point of view:

1. **What is the most relevant prior knowledge about  $v$ ,  $c$ ,  $\alpha$ ?**  
(How nature “chooses” these signals?)
2. **In what norm  $\|\cdot\|$  should the est. error  $v - \hat{v}$  be small?**  
(In what sense we want the “best” approximation of  $v$ ?)

## Band-pass filtering

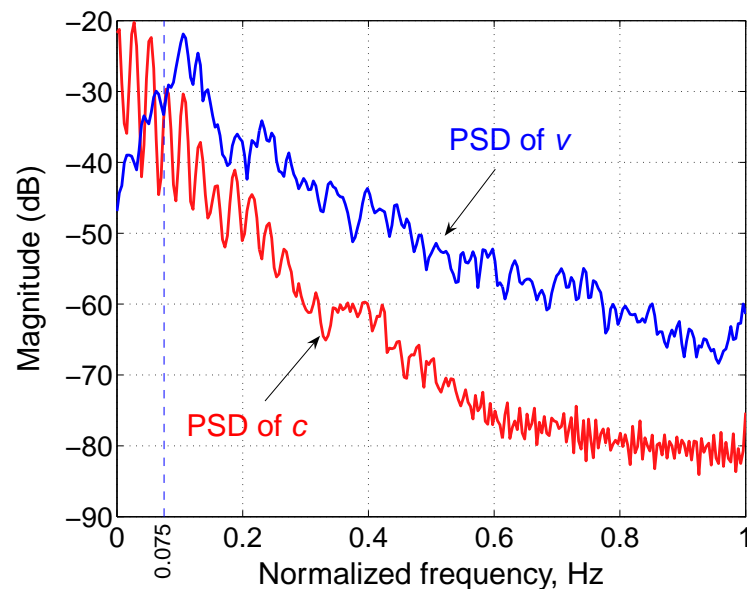


Notes:

- the arterial blood pressure signal  $u$  is not used
- the cutoff frequency  $f_0$  is the only parameter

We choose  $f_0$  empirically from the identification data. It turns out that  $f_0$  doesn't vary much from one pair  $(v, c)$  to another.

## Frequency separation of $v$ and $c$ ( $f_0 = 0.075\text{Hz}$ )



## Filtering based on a model $u \mapsto c$

$M$  is a **prior knowledge**. However, it is not given in practice!

Contrary to  $f_0$ ,  $M$  differs a lot from one pair  $(u, c)$  to another

$\Rightarrow$  computing  $M$  from one part of the data and using it on another part is not an option.

**Possible solution:** use adaptive filter; it identifies  $M$  in real-time from the given data  $(u, y)$ .

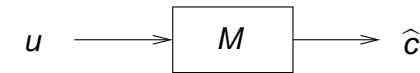
In the simulations, we identify  $M$  from the testing data (unrealistic)

Reason: this serves as a **reference method**, i.e.,

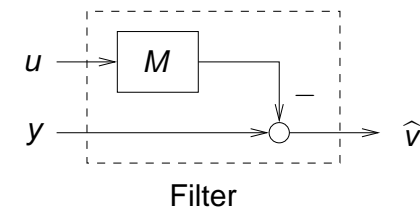
gives an upper bound for the achievable performance.

## Filtering based on a model $u \mapsto c$

$u$  is well correlated with  $c$  and is measured  $\Rightarrow \exists$  model  $M$



such that  $\hat{c}$  is a “good” approximation of  $c$ . Then  $\hat{v}$  can be constructed as follows:



Actually,  $\hat{c}$  is computed by a Kalman filter and not by simulation.

## Adaptive filters proposed in the literature

**Main idea:** simultaneously identify  $M$  from past data  $(u, y)$  and filter  $(u, y)$ , using  $M$ .

Various implementations. Often  $M$  is **assumed FIR** (simplification).

**Main publications:**

- Husøy *et al.* (Stavanger, Norway):  
matching pursuit FIR adaptive filtering
- Amann *et al.* (Innsbruck, Austria):  
FIR adaptive filter implemented by a Kalman filter

This work does not address the questions:

- 1) relevant prior knowledge
- 2) relevant estimation criterion

Moreover, adaptive filters also depend on prior knowledge (smoothness of parameter variation) and its choice is heuristic.

## Outline

Introduction

Filtering methods

Evaluation of the methods

Reproducible research

## Simulation setup

This give us 49  $\text{SNR}(\hat{v}_{i,j})$  for each method and for each  $\alpha$ .

As a final result, we show

a plot of the average  $\text{SNR}(\hat{v})$  as a function of  $\text{SNR}(y)$ .

A method being above the  $45^\circ$  line passing through (0,0) means that this method improves the given signal  $y$ .

## Simulation setup

We use a database of triplets  $(u_i, c_i, v_i)$ ,  $i = 1, \dots, 7$  to construct

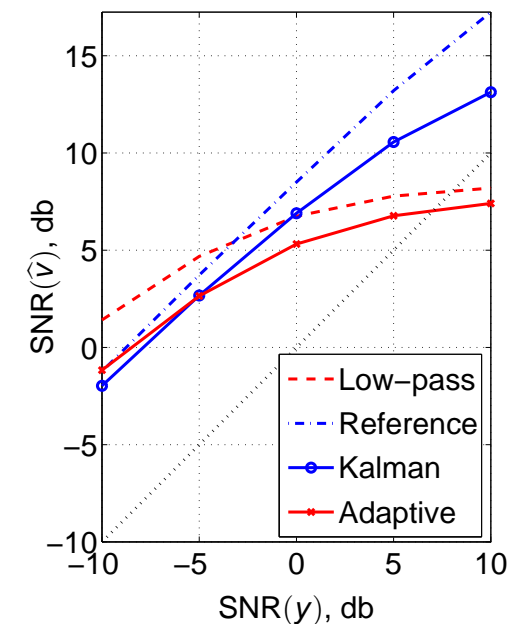
$$y_{\alpha,i,j} = v_i + \alpha c_i, \quad i, j = 1, \dots, 7 \text{ and } \alpha = -10, -5, 0, 5, 10 \quad (*)$$

and apply the considered methods on the data  $(u_i, y_{\alpha,i,j})$ .

Notes:

- The database is collected at the Innsbruck Medical Univ.
- These are “real” measurement  $(u, v)$  from human,  $c$  from pigs
- $y$  is a construction, but the “real” artifacts corrupted human ECG signal is likely to be an additive mixture (\*)

## Comparison of methods



## Observations

1. For low SNR ( $< -6\text{db}$ ), the low-pass filter gives the best performance.
2. The low-pass filter uniformly outperforms the adaptive filter.
3. The reference method has a steady increase of performance as the SNR increases.

The good performance of the low-pass filter may be due to its superior robustness.

## Reproducible research

For a field to qualify as a science, it is important published work be reproducible by others.

J. Buckheit and D. Donoho (Stanford Univ.) advocate:

“When we publish articles containing figures which were generated by computer, we also publish the complete software environment which generates the figures.”

## Conclusions

- **Reference filter:** computing a “good” model for  $u \mapsto c$  is challenging even when using the unknown  $c$  signal.  
The optimization methods have **convergence problems**. Typically, the best model is **unstable**.
- **Adaptive filtering:** computing a good model from  $(u, y)$  online is of course harder than offline.
- The simulation results suggest that the simplest method—low-pass filtering—is overall the best one.
- The filter robustness is important for ECG artifact removal.

## Is research in the biomedical field reproducible?

Usefulness of methods is normally and acceptably “proved” by experiments on “real” data.

However, the methods and/or data being used are rarely given.

The field will benefit from having publicly available benchmark data sets against which the methods can be tested.

Currently this is not the case for ECG artifacts removal.