

Statistical signal processing through applications

ADAPTIVE FILTERING

Problem description:

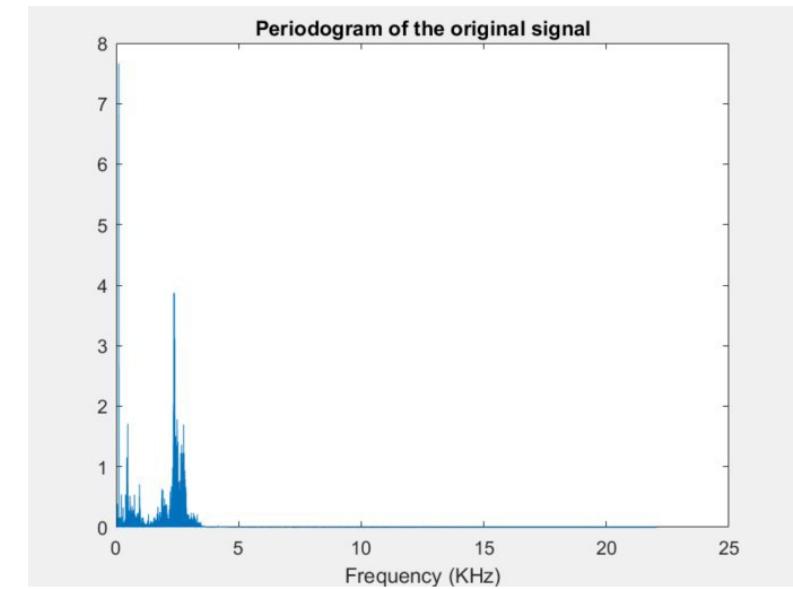
- ▶ Noisy signal (bass signal + talking noise)
 - ▶ Recorded noise (External talking noise)
 - ▶ Want to clean the signal to not hear the talking noise.
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- ▶ Difficulty: Reverb effect affects the recorded noise. Situation may not be static.

Starting point:

- ▶ Bass + talking noise:

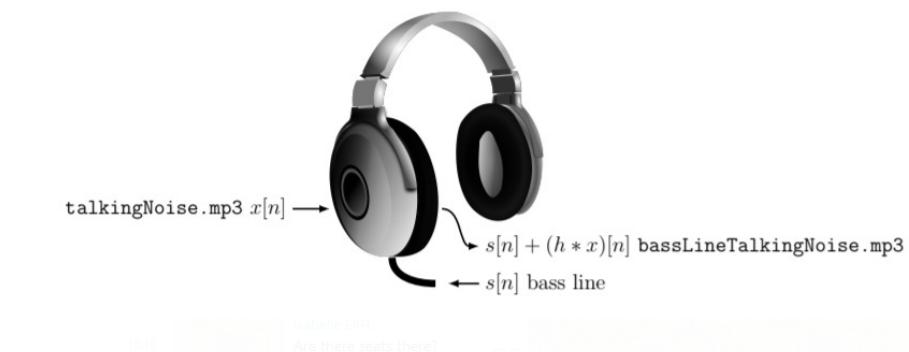
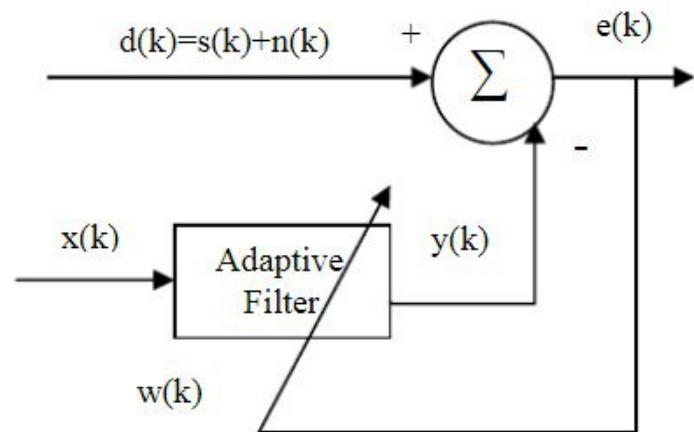


- ▶ Recorded talking noise:



Methods

- ▶ Estimate adaptive filter coefficients $w(k)$ in order to be able to clean the bass signal $e(k)$



To deal with non-static situation, adaptive solutions:

- LMS
- NLMS
- RLS

LMS algorithm summary

Parameters

- ▶ $P = \text{filter order} = 800$ (18 ms as $f_s = 44.1k$)
- ▶ $\mu = \text{step size} = 0.0003$
 - ▶ $\lambda_{\max} =$

$$0 < \mu < \frac{2}{\lambda_{\max}}$$

Initialization

- ▶ $\hat{h}(0) = 0$ or random

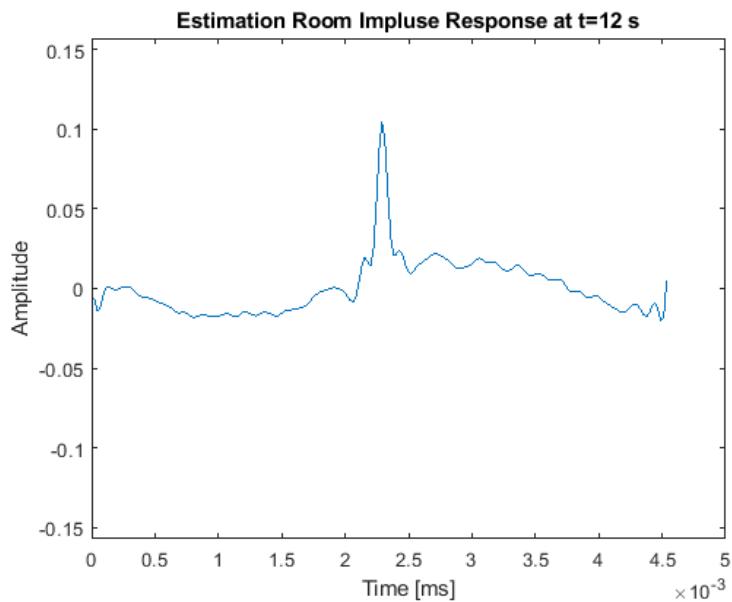
Computation

- ▶ Computation: For $n = 0, 1, 2, \dots$
- ▶ $X(n) = [x(n), x(n - 1), \dots, x(n - p + 1)]^T$
- ▶ $e(n) = d(n) - \hat{h}^H(n) X(n)$
- ▶ $\hat{h}(n+1) = \hat{h}(n) + \mu e^*(n) X(n)$

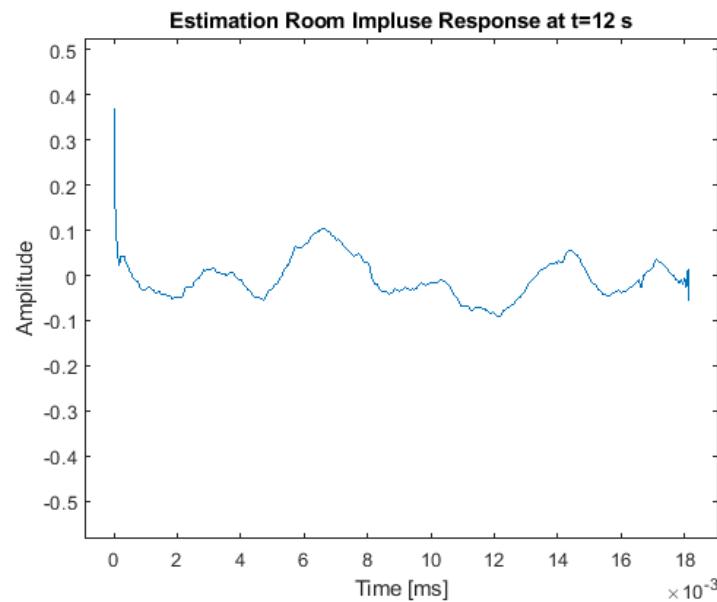
LMS results

Estimated Impulse Response

Offline

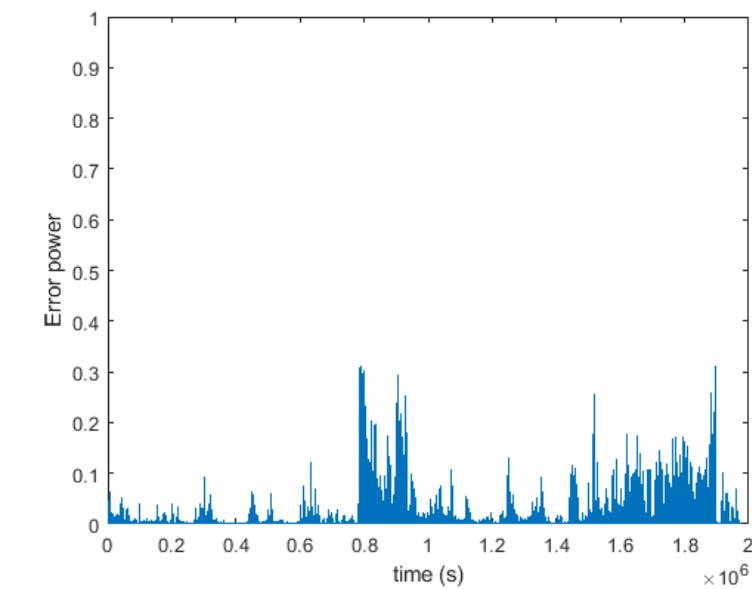


On the fly



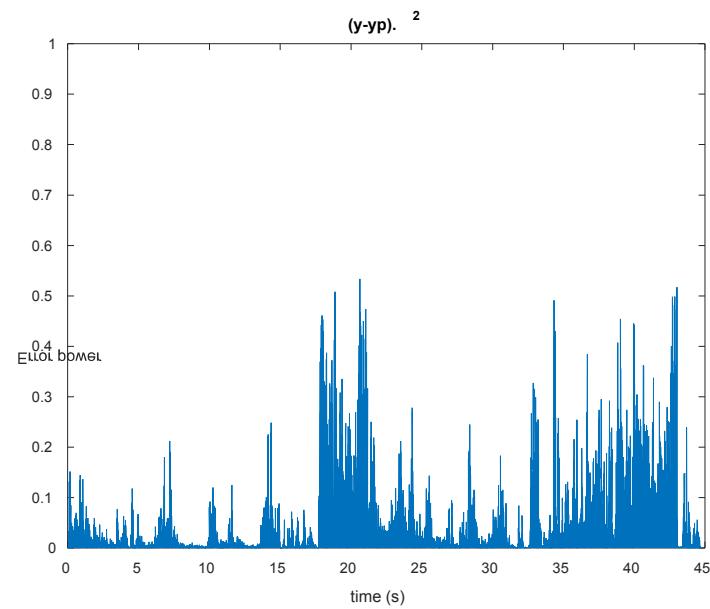
LMS results

Mean square error



Offline

Mean square error



On the fly

NLMS algorithm summary

Parameters

- ▶ $P = \text{filter order} = 800$ (18 ms as $fs = 44.1k$)
- ▶ $\mu = \text{step size} = 0.003 * Nf * \text{mean}(x.^2)$

Initialization

- ▶ $\hat{h}(0) = 0$ or random

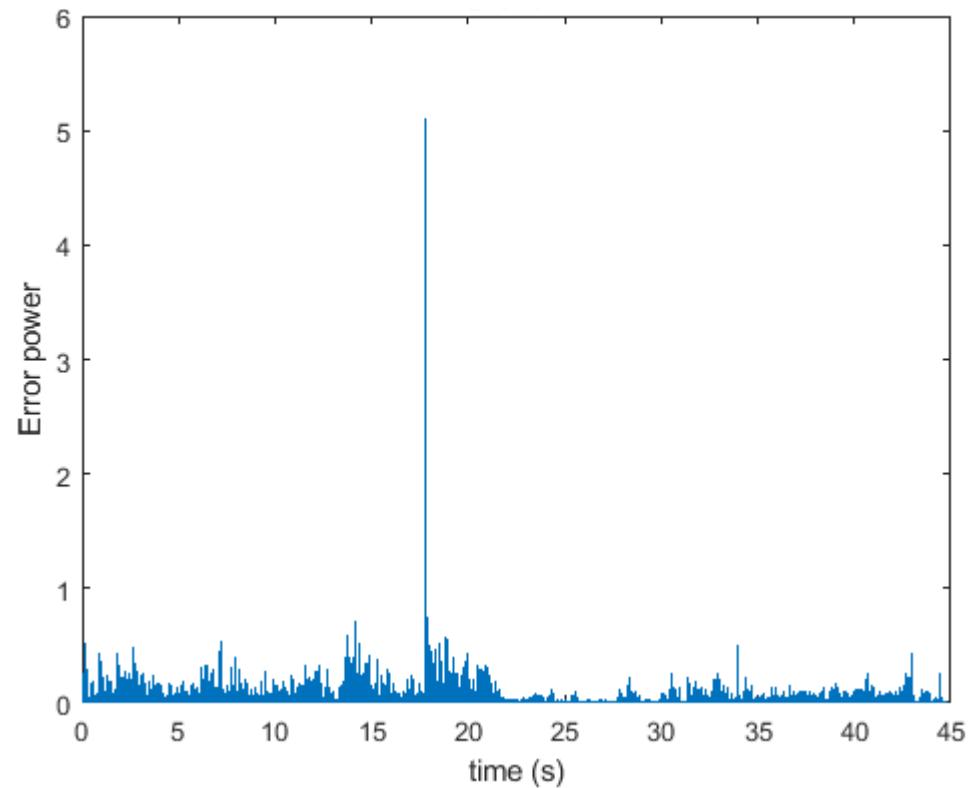
Computation

- ▶ Computation: For $n = 0, 1, 2, \dots$
- ▶ $X(n) = [x(n), x(n - 1), \dots, x(n - p + 1)]^T$
- ▶ $e(n) = d(n) - \hat{h}^H(n) X(n)$
- ▶ $\hat{h}(n+1) = \hat{h}(n) + \frac{\mu e(n) X(n)}{X^H(n) X(n)}$

NLMS results

NLMS

- ▶ Play sound



RLS Algorithm Summary

Parameters

- ▶ $P = 300$ samples,
- ▶ Signal downsampled to 11kHz
- ▶ $\lambda = \text{forgetting factor} = 0.999$

Initialization

- ▶ $\hat{h}(0) = \text{random (cannot be zero)}$

Computation

- ▶ P is correlation matrix
- ▶ Batch is the u vector

$$\begin{aligned}k(n) &= \frac{\lambda^{-1}P(n-1)\underline{u}(n)}{1 + \lambda^{-1}\underline{u}^T(n)P(n-1)\underline{u}(n)} \\ \alpha(n) &= d(n) - \underline{u}(n)^T \underline{w}(n-1) \\ \underline{w}(n) &= \underline{w}(n-1) + k(n)\alpha(n) \\ P(n) &= \lambda^{-1}P(n-1) - \lambda^{-1}k(n)\underline{u}^T(n)P(n-1)\end{aligned}$$

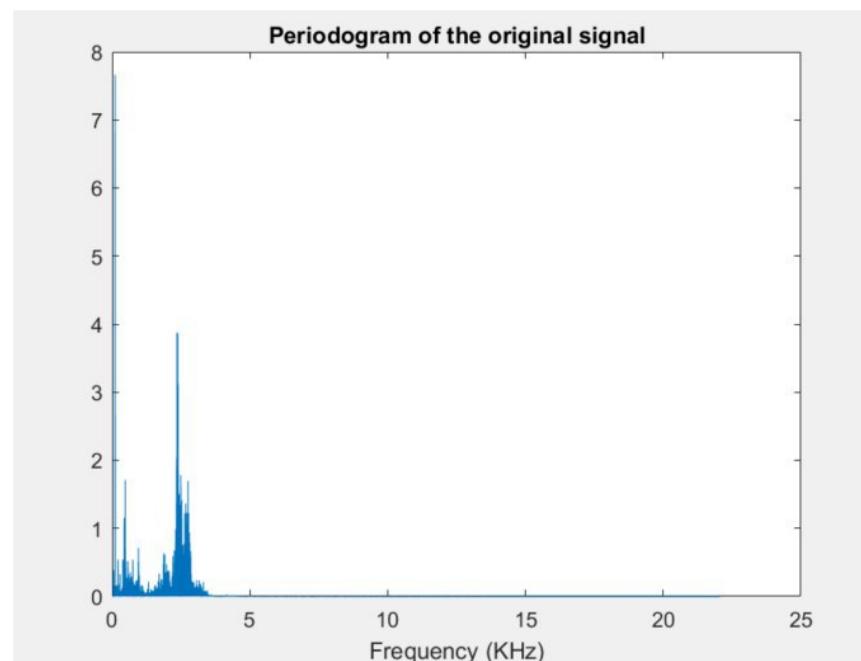
RLS results

RLS

- ▶ Play sound

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graphs



Comparison

	Filter order	Mse using only past samples	Mse using past&future samples
Noisy signal		0,0568	
LMS	800 (offline 200)	0,0338	0.004
NLMS	800	0,0435	
RLS	300	0.0022	

Thank you, any questions?