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On the Regularization of a Low-Complexity Recursive Least-Squares Adaptive Algorithm

Cristian-Lucian Stanciu, Cristian Anghel, Camelia-Elisei Iliescu, Laura-Maria Dogariu, Ionuț-Dorinel Fîciu, and Constantin Paleologu

*Department of Telecommunications,
POLITEHNICA Bucharest, Romania*

cristian@comm.pub.ro, canghel@comm.pub.ro, camelia.elisei@romatsa.ro, ldogariu@comm.pub.ro,
ionut.ficiu22@gmail.com, pale@comm.pub.ro

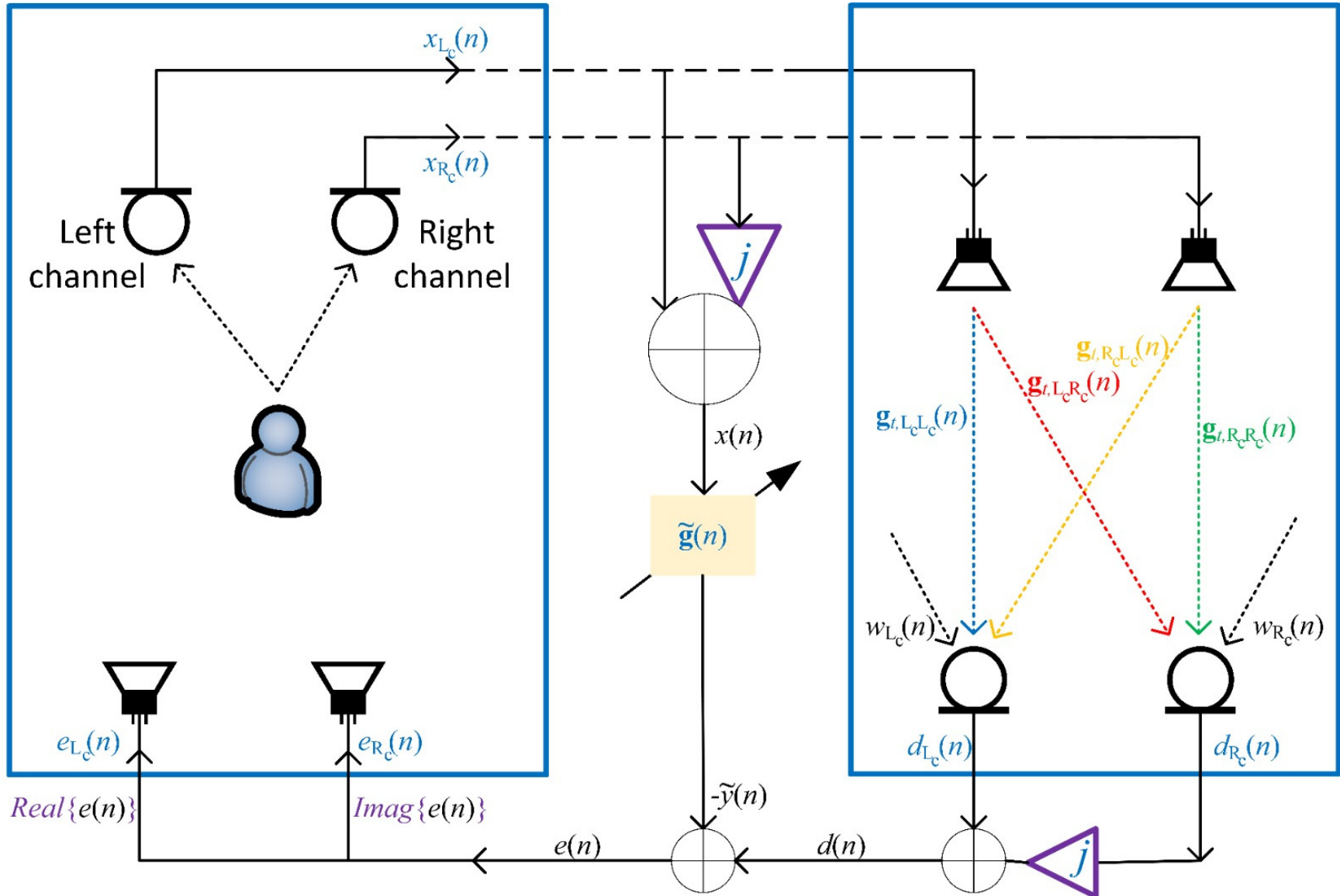
Summary

- Introduction
- System Model – WL Framework
- WL-RLS-LSM
- VR-WL-RLS-LSM
- Simulation Results
- Conclusions

Introduction

- Stereophonic Acoustic Echo Cancellation (**SAEC**):
 - Estimation of 4 echo paths => the loudspeaker-to-microphone pairs
- The **LMS** method (standard approach):
 - 4 individual systems => **low arithmetic complexity** and **limited performance for highly correlated signals due to gradient noise**
- The **WL-RLS-LSM** algorithms:
 - Complex-valued line search methods
 - Solve an auxiliary system of equations
 - Have a key parameter – performance vs complexity
- The **VR-WL-RLS-LSM** algorithms:
 - Improved performance for small echo-to-noise (ENR) values
 - Improved performance when double-talk (DT) situations occur

System Model – WL



WL-RLS-LSM

$$\mathbf{R}(n)\mathbf{g}(n) = \mathbf{p}(n)$$

- Classical Approach

RLS

$$\tilde{\mathbf{g}}(n)$$

- Direct Solution

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

- LSM Approaches

CG

$$\mathbf{r}(n), \tilde{\mathbf{g}}(n) = \tilde{\mathbf{g}}(n-1) + \Delta\tilde{\mathbf{g}}(n)$$

$$\mathbf{p}_0(n+1) = \lambda\mathbf{r}(n) + \mathbf{x}(n+1)e(n+1)$$

$$\mathbf{R}(n+1)\Delta\mathbf{g}(n+1) = \mathbf{p}_0(n+1)$$

...

$$\mathbf{r}(n+1), \tilde{\mathbf{g}}(n+1)$$

Conjugate Gradient

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

$$\Delta\tilde{\mathbf{g}} \leftarrow \mathbf{0}_{2L \times 1}$$

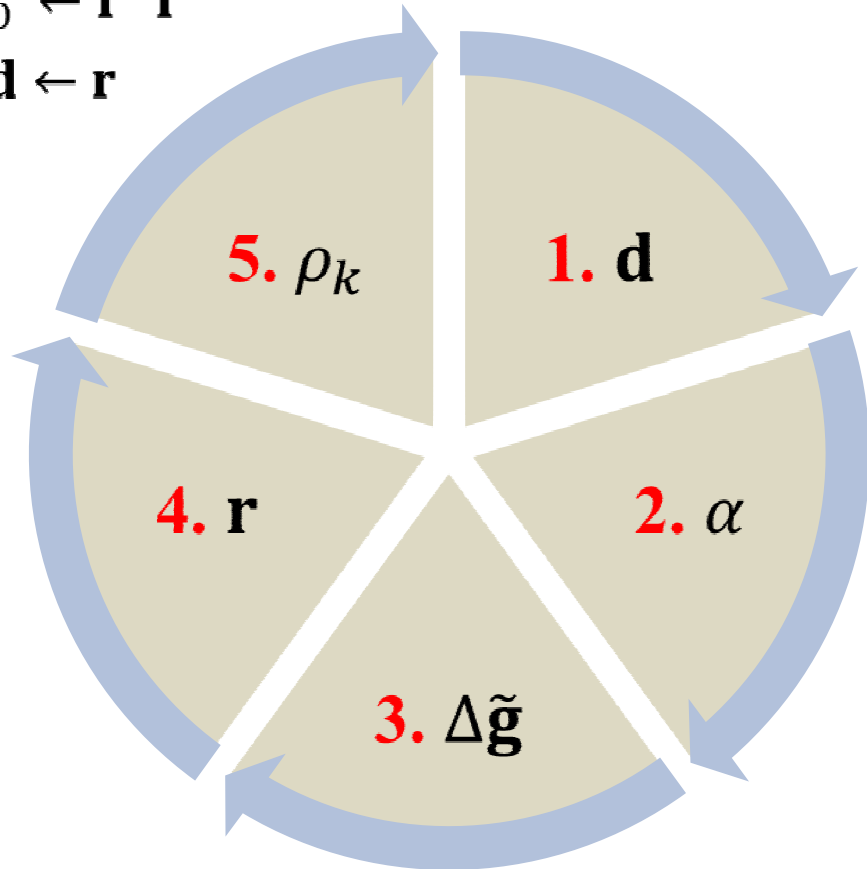
$$\mathbf{r} \leftarrow \mathbf{p}_0(n)$$

$$\rho_0 \leftarrow \mathbf{r}^H \mathbf{r}$$

$$\mathbf{d} \leftarrow \mathbf{r}$$

CG

$k \leftarrow 1 : N_u$



Conjugate Gradient

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

$$\Delta\tilde{\mathbf{g}} \leftarrow \mathbf{0}_{2L \times 1}$$

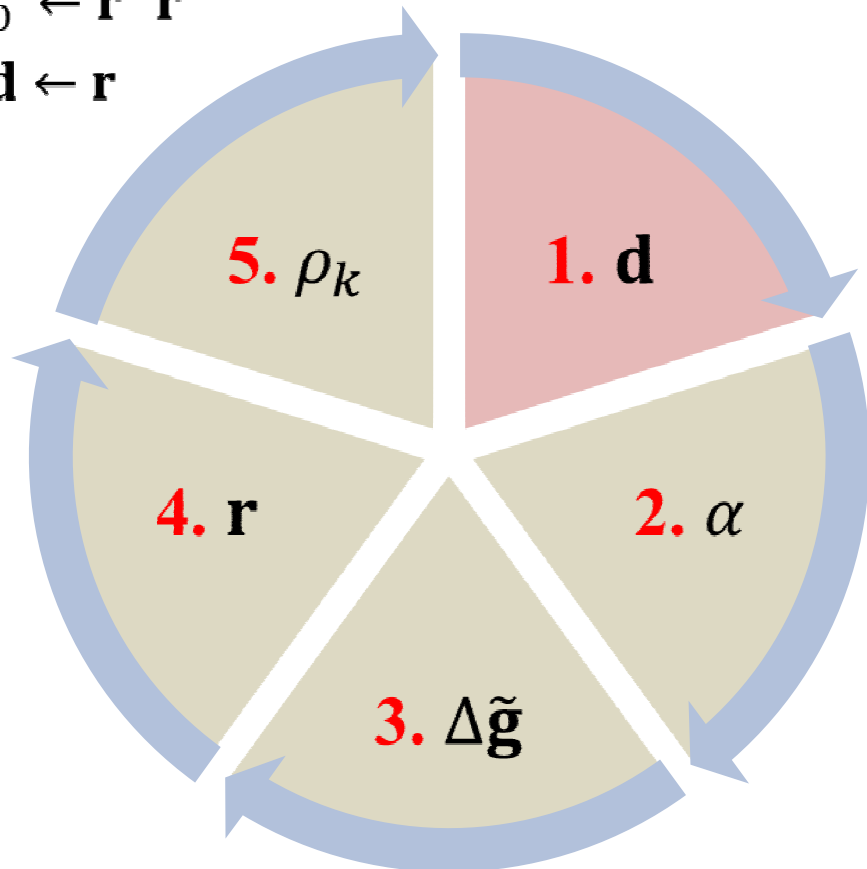
$$\mathbf{r} \leftarrow \mathbf{p}_0(n)$$

$$\rho_0 \leftarrow \mathbf{r}^H \mathbf{r}$$

$$\mathbf{d} \leftarrow \mathbf{r}$$

CG

$k \leftarrow 1 : N_u$



$$k > 1 ? \mathbf{d} \leftarrow \mathbf{r} + \frac{\rho_{k-1}}{\rho_{k-2}} \mathbf{d}$$

Conjugate Gradient

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

$$\Delta\tilde{\mathbf{g}} \leftarrow \mathbf{0}_{2L \times 1}$$

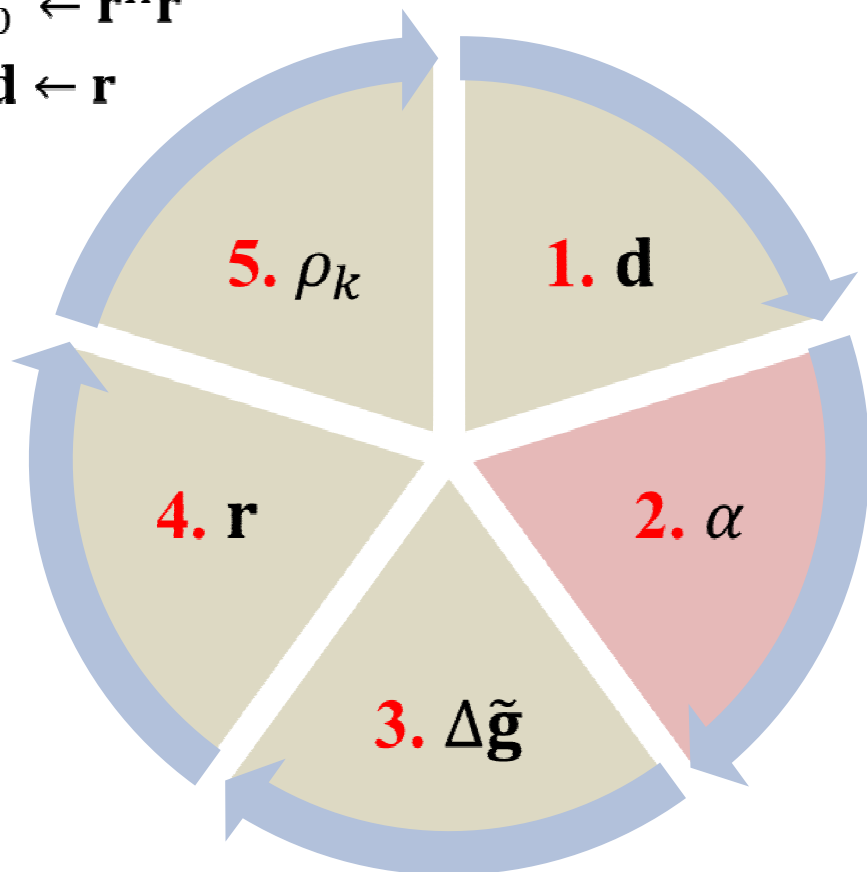
$$\mathbf{r} \leftarrow \mathbf{p}_0(n)$$

$$\rho_0 \leftarrow \mathbf{r}^H \mathbf{r}$$

$$\mathbf{d} \leftarrow \mathbf{r}$$

CG

$k \leftarrow 1 : N_u$



$$\alpha \leftarrow \frac{\rho_{k-1}}{\mathbf{d}^H \mathbf{R}_{\tilde{\mathbf{x}}}(n) \mathbf{d}}$$

Conjugate Gradient

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

$$\Delta\tilde{\mathbf{g}} \leftarrow \mathbf{0}_{2L \times 1}$$

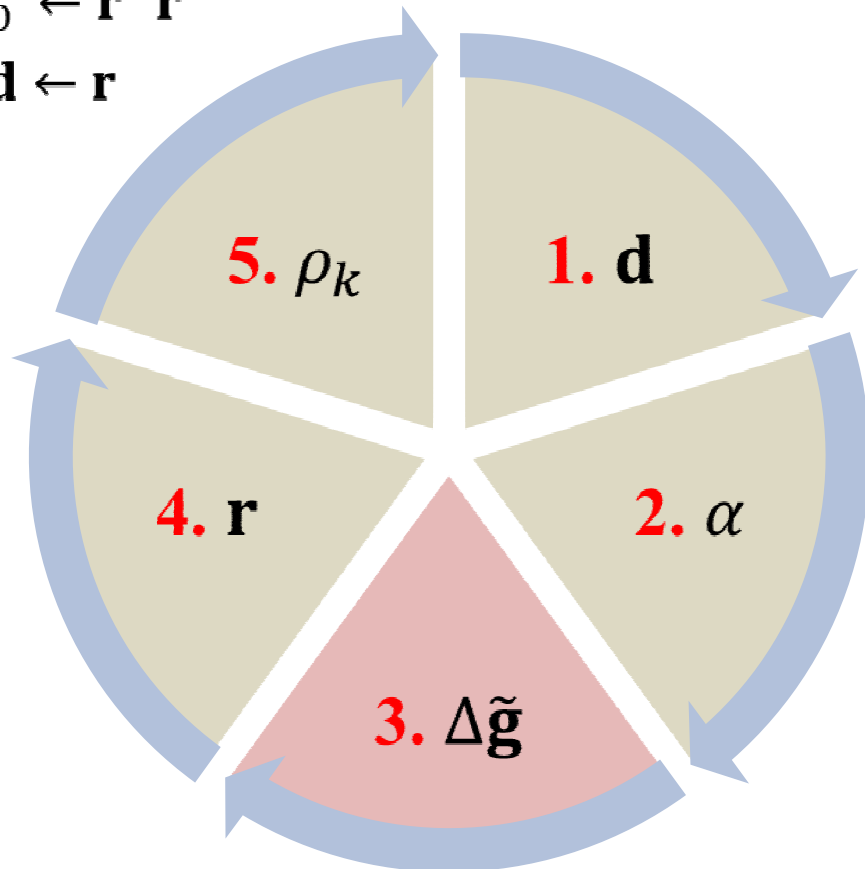
$$\mathbf{r} \leftarrow \mathbf{p}_0(n)$$

$$\rho_0 \leftarrow \mathbf{r}^H \mathbf{r}$$

$$\mathbf{d} \leftarrow \mathbf{r}$$

CG

$k \leftarrow 1 : N_u$



$$\Delta\tilde{\mathbf{g}} \leftarrow \Delta\tilde{\mathbf{g}} + \alpha\mathbf{d}$$

Conjugate Gradient

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

$$\Delta\tilde{\mathbf{g}} \leftarrow \mathbf{0}_{2L \times 1}$$

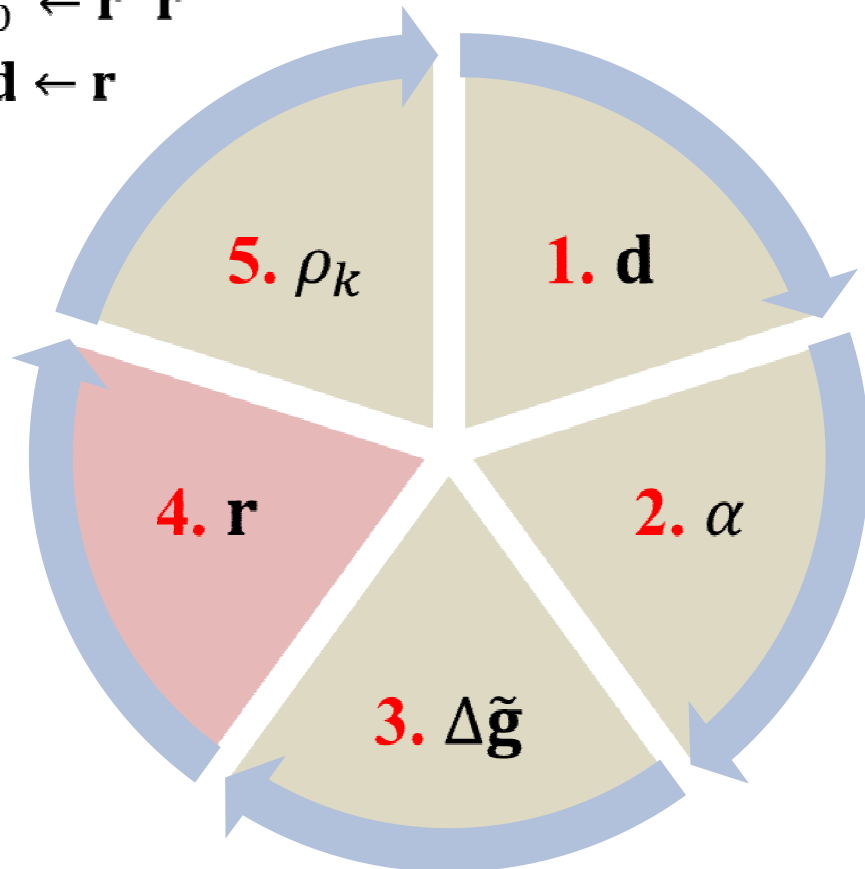
$$\mathbf{r} \leftarrow \mathbf{p}_0(n)$$

$$\rho_0 \leftarrow \mathbf{r}^H \mathbf{r}$$

$$\mathbf{d} \leftarrow \mathbf{r}$$

CG

$k \leftarrow 1 : N_u$



$$\mathbf{r} \leftarrow \mathbf{r} - \alpha \mathbf{R}_{\tilde{\mathbf{x}}}(n) \mathbf{d}$$

Conjugate Gradient

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

$$\Delta\tilde{\mathbf{g}} \leftarrow \mathbf{0}_{2L \times 1}$$

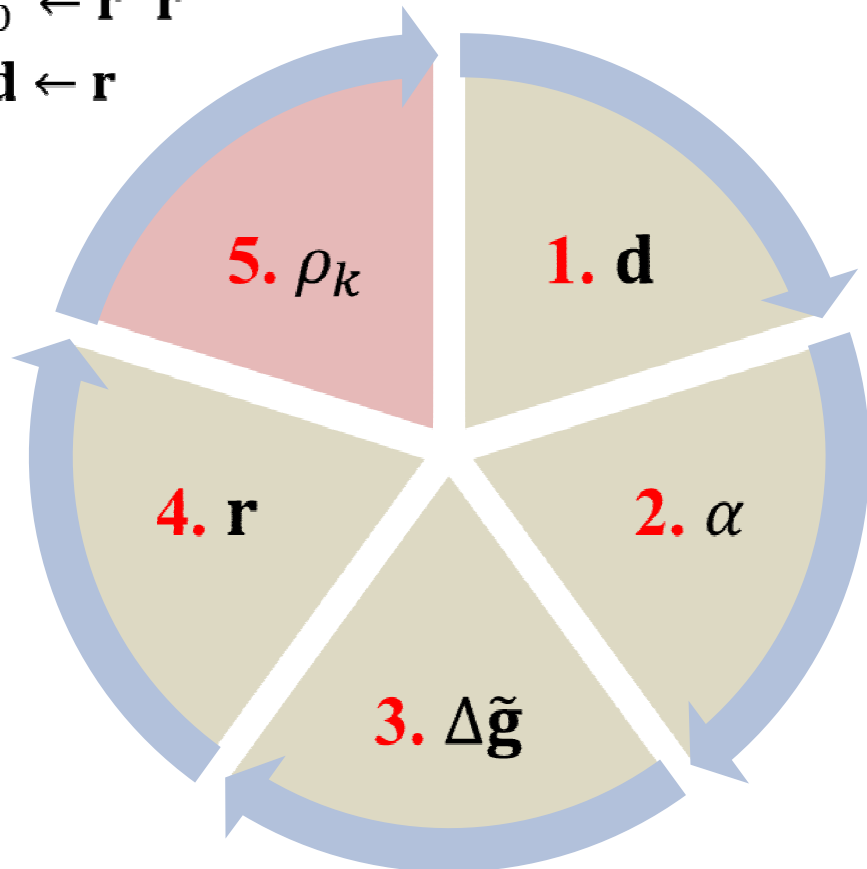
$$\mathbf{r} \leftarrow \mathbf{p}_0(n)$$

$$\rho_0 \leftarrow \mathbf{r}^H \mathbf{r}$$

$$\mathbf{d} \leftarrow \mathbf{r}$$

CG

$k \leftarrow 1 : N_u$

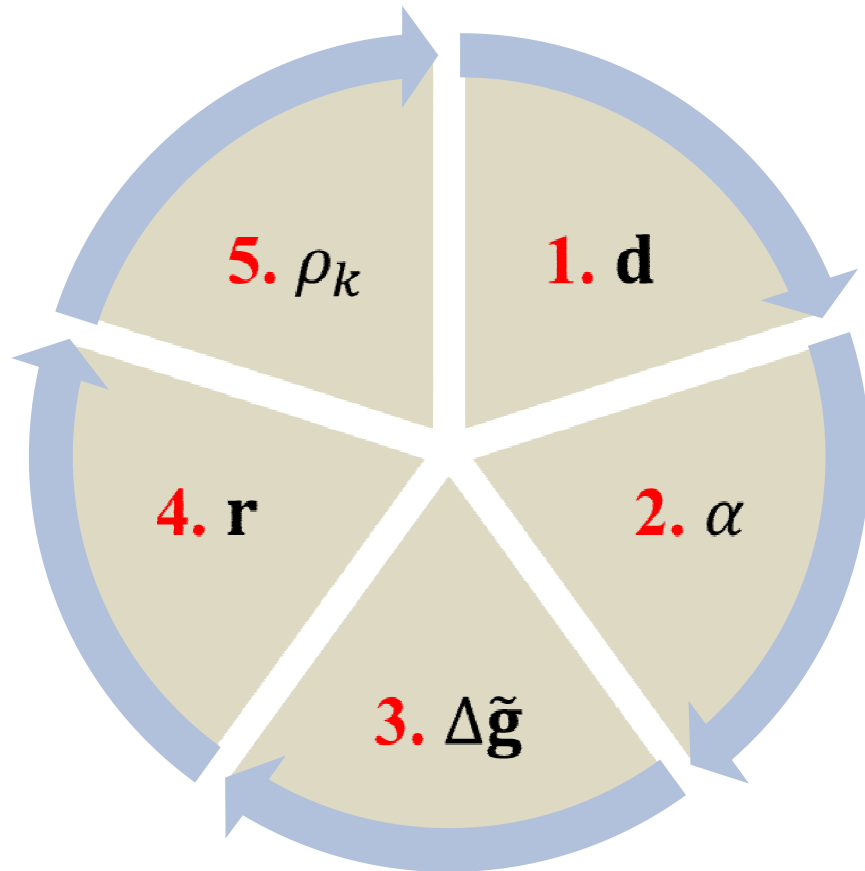


$$\rho_k \leftarrow \mathbf{r}^H \mathbf{r}$$

Conjugate Gradient

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

CG $k \leftarrow 1 : N_u$



Complexity: $O(8L^2)$ multiplications/it.

$$\tilde{\mathbf{g}}(n+1) = \tilde{\mathbf{g}}(n) + \Delta\tilde{\mathbf{g}}$$

Dichotomous Coordinate Descent

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

Complexity: no multiplications/it.

$$\Delta\tilde{\mathbf{g}} \leftarrow \mathbf{0}_{2L \times 1}$$

$$\mathbf{r} \leftarrow \mathbf{p}_0(n)$$

$$\alpha \leftarrow H$$

$$m \leftarrow 0$$

DCD

$$k \leftarrow 1 : N_u$$

Step	Action
1	$\{val_1; pos_1\} \leftarrow \max_{i=1 \dots 2L} \{ Re\{r_i\} \}$ $\{val_2; pos_2\} \leftarrow \max_{i=1 \dots 2L} \{ Im\{r_i\} \}$ $(p; \eta) \leftarrow (val_1 > val_2) ? (pos_1; 1) : (pos_2; j)$ Jump to step 4
2	$\alpha \leftarrow \alpha/2; \quad m \leftarrow m + 1$
3	If $(m > M_b)$ RETURN
4	$r_{temp} \leftarrow (\eta == 1) ? Re\{r_p\} : Im\{r_p\}$
5	If $(r_{temp} \leq \frac{\alpha}{2} R_{p,p})$ Jump to step 2
6	$\Delta\tilde{g}_p \leftarrow \Delta\tilde{g}_p + \text{sign}\{r_{temp}\}\eta\alpha$
7	$\mathbf{r} \leftarrow \mathbf{r} - \text{sign}\{r_{temp}\}\eta\alpha\mathbf{R}^{(p)}(n)$

$$\tilde{\mathbf{g}}(n+1) = \tilde{\mathbf{g}}(n) + \Delta\tilde{\mathbf{g}}$$

Variable Regularization

$$\mathbf{R}(n)\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

VR



$$[\mathbf{R}(n) + \phi(n)\mathbf{I}_{2L}]\Delta\mathbf{g}(n) = \mathbf{p}_0(n)$$

$$\Phi(n) = 2L\tilde{\sigma}_x^2(n)(1 + \sqrt{1 + \widetilde{\text{ENR}}})/(\widetilde{\text{ENR}})$$



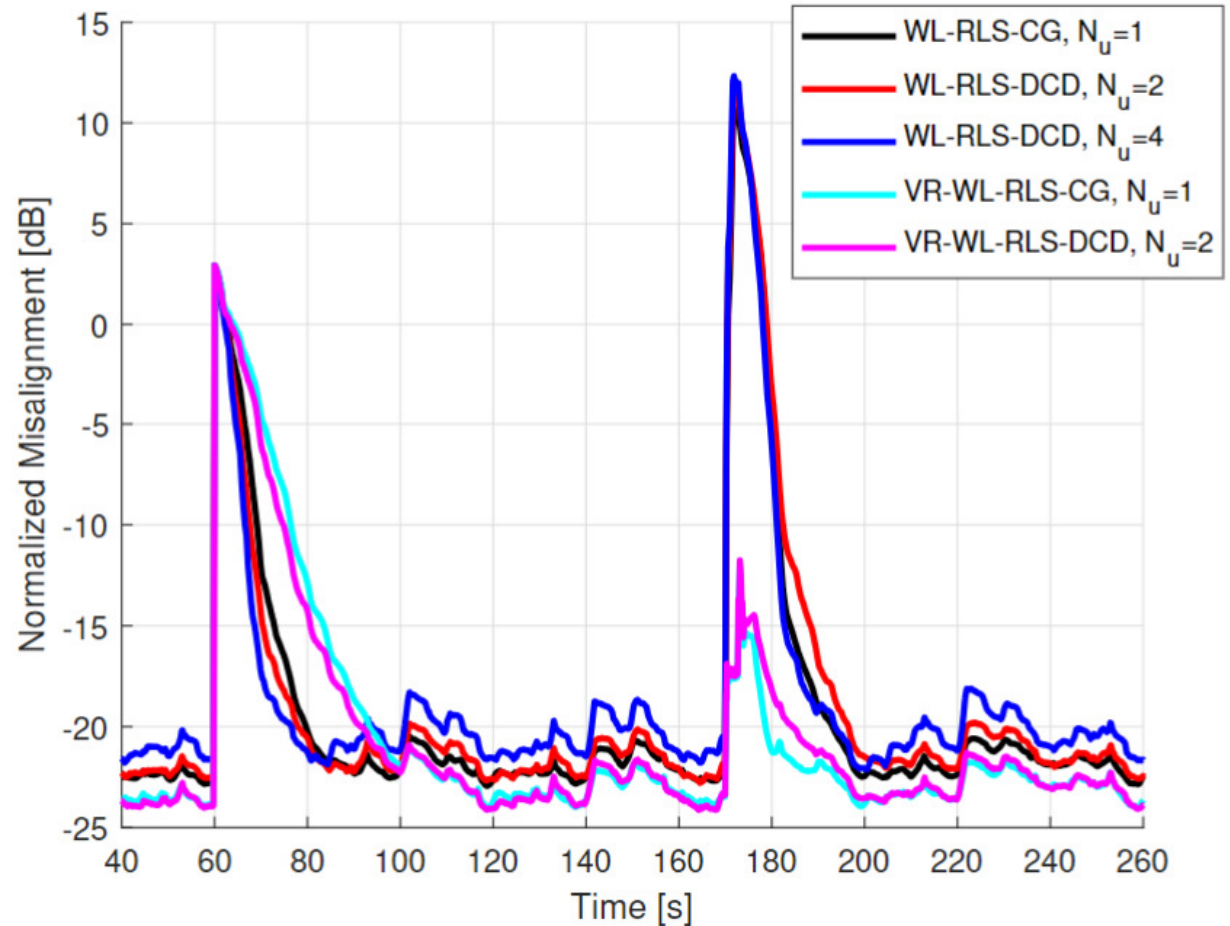
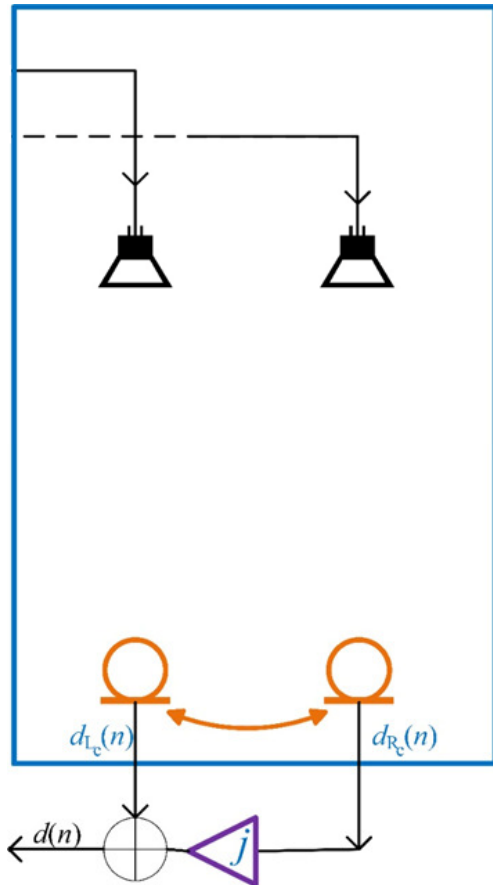
$$\tilde{\mathbf{g}}(n + 1) = \tilde{\mathbf{g}}(n) + \Delta\tilde{\mathbf{g}}$$

Simulation Results

- **Tracking & Double Talk:**
 - WL-RLS-CG vs WL-RLS-DCD vs VR-WL-RLS-CG vs VR-WL-RLS-DCD
- **Input signal:**
 - High quality speech sequence
- **Unknown systems (*left* and *right*):**
 - Real measured impulse responses (decimated)
 - $L = 256$
 - ENR = 25 dB & 10 dB
- **Performance measure:**

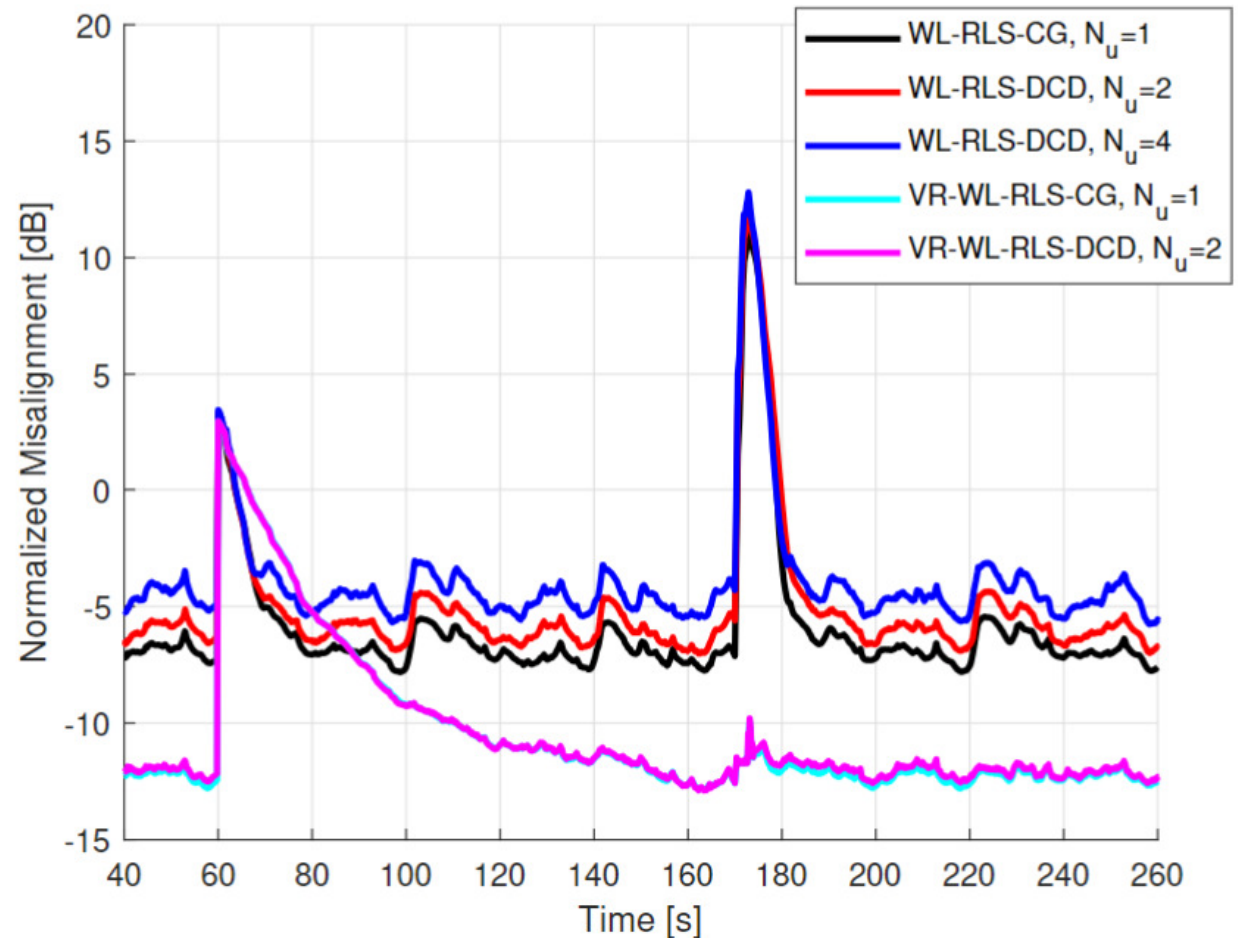
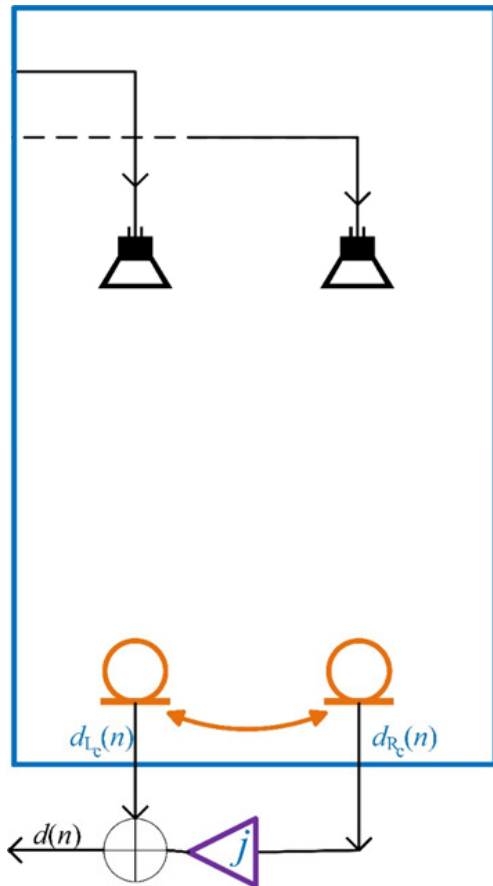
$$\text{Mis}(n) = 20 \log_{10} \left[\frac{\|\tilde{\mathbf{g}}_t - \tilde{\mathbf{g}}(n)\|_2}{\|\tilde{\mathbf{g}}_t\|_2} \right] (\text{dB})$$

Simulation Results



Performance of the LSM algorithms for $L=256$ and $\text{ENR} = 25\text{dB}$. The unknown system changes at $t_0 = 60\text{s}$ due to the interchanging of the microphone positions. **Double-talk** situation in the **time interval [170; 174]s**.

Simulation Results

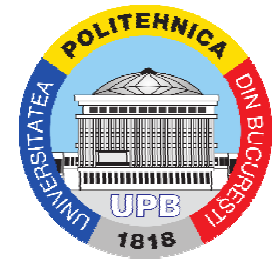


Performance of the LSM algorithms for $L=256$ and $\text{ENR} = 10\text{dB}$. The unknown system changes at $t_0 = 60\text{s}$ due to the interchanging of the microphone positions. **Double-talk** situation in the **time interval [170; 174]s**.

Conclusions

○ VR-WL-RLS-LSM:

- Improves the performance of WL-RLS-LSM working in the SAEC setup with small ENR values and/or when double-talk situations occur
- Extra arithmetical effort
- Tracking speed loses



Thank you!

Please refer any questions to:

cristian@comm.pub.ro,
canghel@comm.pub.ro,
camelia.elisei@romatsa.ro,
ldogariu@comm.pub.ro,
ionut.ficiu22@gmail.com,
pale@comm.pub.ro

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