



A Data-Reuse Approach for the RLS-DCD Algorithm

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** Presenter*

Presenter's Biography

Ionut Fîciu obtained his bachelor's degree in 2018 - the Systems and Technologies in Telecommunications specialization, University Politehnica of Bucharest (UPB), Romania – valedictorian. He is a PhD. student in the Telecommunications Department at UPB. His main research topic is related to the field of adaptive systems.

He works as a member in three research grants, and during his first 18 months of doctoral studies he published more than ten papers (including in Q2/Q3 journals).

Summary

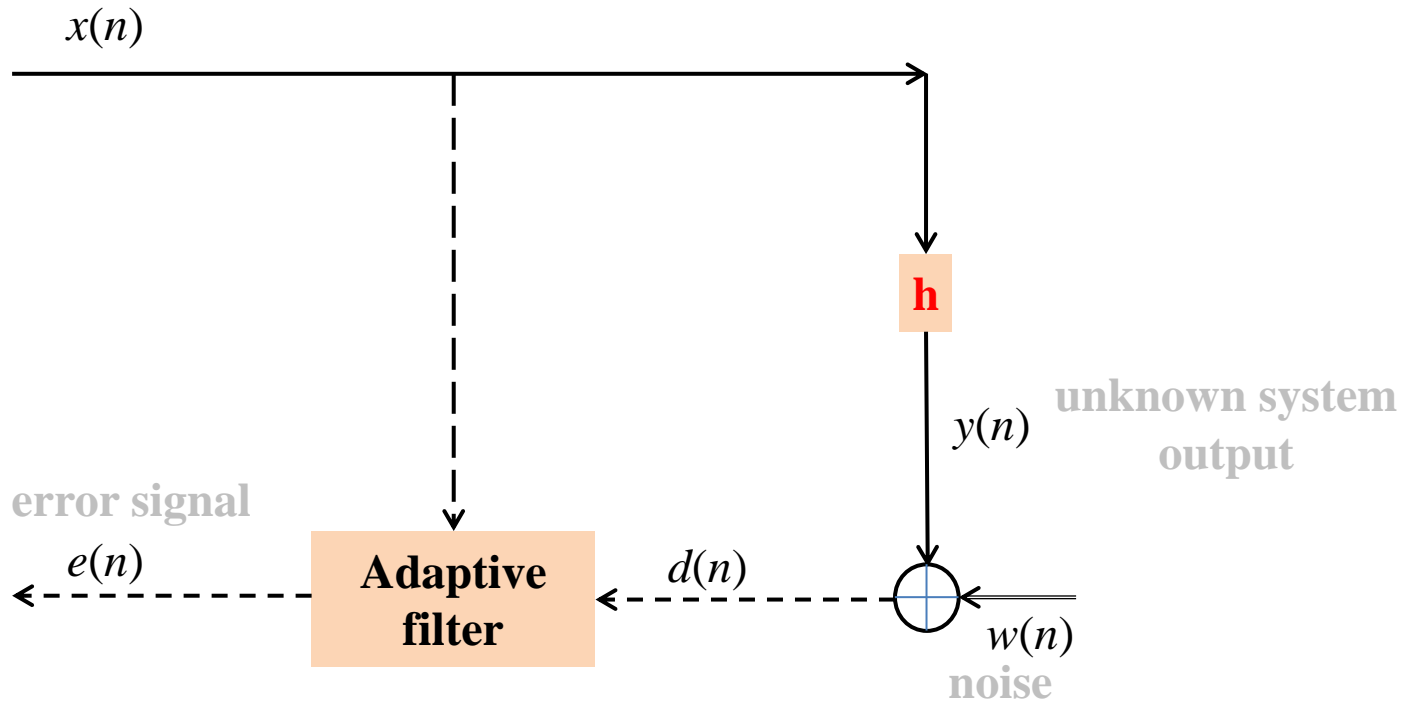
- Introduction
- System Model
- The Recursive Least Squares - Dichotomous Coordinate Descent (RLS-DCD) Algorithm
- Data-Reuse (DR) RLS-DCD Algorithm
- Simulation Results
- Conclusions

INTRODUCTION

- RLS adaptive algorithm:
 - Mitigation of highly correlated input signals
 - High arithmetic costs
- The RLS-DCD method:
 - The RLS algorithm => the **matrix inversion** problem
 - » The Dichotomous Coordinate Descent (DCD) iterations
- The Data-Reuse (DR) RLS-DCD:
 - Increased tracking/convergence speed
 - » The use of the same input information at time index n
 - » Repeat the update process (*reuse*) the **same data set** N_{it} times

System Model

input signal



$$\mathbf{x}(n) = [x(n) \quad x(n-1) \quad \dots \quad x(n-L+1)]^T$$

$$\mathbf{h} = [h_0 \quad h_1 \quad \dots \quad h_{L-1}]^T$$

$$d(n) = y(n) + w(n) = \mathbf{h}^T \mathbf{x}(n) + w(n)$$



$$e(n) = d(n) - \hat{y}(n) = d(n) - \hat{\mathbf{h}}(n-1) \mathbf{x}(n)$$

RLS-DCD Algorithm

- The least-squares (LS) error criterion:

$$J[\hat{\mathbf{h}}(n)] = \sum_{i=1}^n \lambda^{n-i} |d(i) - \hat{\mathbf{h}}^T(n)\mathbf{x}(i)|^2 \quad \dashrightarrow \quad 0 \leq \lambda < 1$$

$$\mathbf{R}(n)\hat{\mathbf{h}}(n) = \mathbf{p}(n)$$

where

$$\begin{aligned} \mathbf{R}(n) &= \sum_{i=1}^n \lambda^{n-i} \mathbf{x}(i)\mathbf{x}^T(i) && \dashrightarrow L \times L \\ \mathbf{p}(n) &= \sum_{i=1}^n \lambda^{n-i} \mathbf{x}(i)d(i) && \dashrightarrow L \times 1 \end{aligned}$$

- Matrix inversion:

{	RLS	\dashrightarrow	$\sim L^2$	\dashrightarrow RLS-DCD
	FRLS	\dashrightarrow	$\sim L$	

RLS-DCD Algorithm

$$0: \hat{\mathbf{h}}(0) = \mathbf{0}, \mathbf{r}(0) = \mathbf{0}, \mathbf{R}(0) = \delta \mathbf{I}_L$$

For $n = 1, 2 \dots$

$$1: \mathbf{R}(n) = \lambda \mathbf{R}(n-1) + \mathbf{x}(n)\mathbf{x}^T(n)$$

$$2: e(n) = d(n) - \hat{\mathbf{h}}^T(n-1)\mathbf{x}(n)$$

$$3: \mathbf{p}_0(n) = \lambda \mathbf{r}(n-1) + \mathbf{x}(n)e(n) \\ = \lambda \mathbf{r}(n-1) + \mathbf{r}_{e,x}(n)$$

$$4: \mathbf{R}(n)\Delta\hat{\mathbf{h}}(n) = \mathbf{p}_0(n) \Rightarrow \Delta\hat{\mathbf{h}}(n), \mathbf{r}(n)$$

$$5: \hat{\mathbf{h}}(n) = \hat{\mathbf{h}}(n-1) + \Delta\hat{\mathbf{h}}(n)$$

Only the first columns has to be computed

$$\mathbf{R}^{(1)}(n) = \lambda \mathbf{R}^{(1)}(n-1) + x(n)\mathbf{x}(n)$$

- Solved with Dichotomous Coordinate Descent (DCD):
 - Uses only **additions**
 - Important parameters:
 - » N_u (maximum number of allowed updates/ “successful iterations”)
 - » M_b (number of bits used for the representation of the solution vector)
 - Arithmetic complexity $\sim LN_u$, $N_u \ll L$

DR-RLS-DCD Algorithm

- At each time index “ n ”, N_{it} :

$$e(n) = \begin{cases} d(n) - \hat{\mathbf{h}}^T(n-1)\mathbf{x}(n) = e_0(n), & k = 0 \\ e_{k-1}(n) - \Delta \hat{\mathbf{h}}_{k-1}^T(n)\mathbf{x}(n), & k = 1 \dots N_{it} - 1 \end{cases}$$

The regular error expression; use results from time index $n-1$

Reuse of the same input information (for $N_{it}-1$ times)
 $\sim N_{it}N_u$

$$\mathbf{r}_k(n) = \begin{cases} \lambda \mathbf{r}_{N_{it}-1}(n-1) + e_0(n)\mathbf{x}(n), & k = 0 \\ \mathbf{p}_k(n) = \mathbf{r}_{k-1}(n) + e_k(n)\mathbf{x}(n), & k = 1 \dots N_{it} - 1 \end{cases}$$

Update the **residual component**: use results from time index $n-1$

Update the **residual component** for the **reuse** process.
 $\sim L$

DR-RLS-DCD Algorithm

$$0: \hat{\mathbf{h}}(0) = \mathbf{0}, \mathbf{r}(0) = \mathbf{0}, \mathbf{R}(0) = \delta \mathbf{I}_L$$

For $n = 1, 2 \dots$

$$1: \mathbf{R}(n) = \lambda \mathbf{R}(n-1) + \mathbf{x}(n) \mathbf{x}^T(n)$$

$$2: e(n) = d(n) - \hat{\mathbf{h}}^T(n-1) \mathbf{x}(n)$$

$$3: \mathbf{p}_0(n) = \lambda \mathbf{r}(n-1) + \mathbf{x}(n) e(n) \\ = \lambda \mathbf{r}(n-1) + \mathbf{r}_{e,x}(n)$$

$$4: \mathbf{R}(n) \Delta \hat{\mathbf{h}}(n) = \mathbf{p}(n) \Rightarrow \Delta \hat{\mathbf{h}}(n), \mathbf{r}(n)$$

$$5: \hat{\mathbf{h}}(n) = \hat{\mathbf{h}}(n-1) + \Delta \hat{\mathbf{h}}(n)$$

$$0: \hat{\mathbf{h}}(0) = \mathbf{0}, \mathbf{r}(0) = \mathbf{0}, \mathbf{R}(0) = \delta \mathbf{I}_L$$

For $n = 1, 2 \dots$

$$1: \mathbf{R}(n) = \lambda \mathbf{R}(n-1) + \mathbf{x}(n) \mathbf{x}^T(n)$$

$$2: e(n) = d(n) - \hat{\mathbf{h}}^T(n-1) \mathbf{x}(n)$$

3: For $k = 1, 2 \dots N_{it}$

$$a) \text{ Compute } e_k(n)$$

$$b) \text{ Compute } \mathbf{p}_k(n)$$

$$c) \mathbf{R}(n) \Delta \hat{\mathbf{h}}_k(n) = \mathbf{p}_k(n) \Rightarrow \Delta \hat{\mathbf{h}}_k(n), \mathbf{r}_k(n)$$

$$d) \hat{\mathbf{h}}_k(n) = \hat{\mathbf{h}}_k(n-1) + \Delta \hat{\mathbf{h}}_k(n)$$

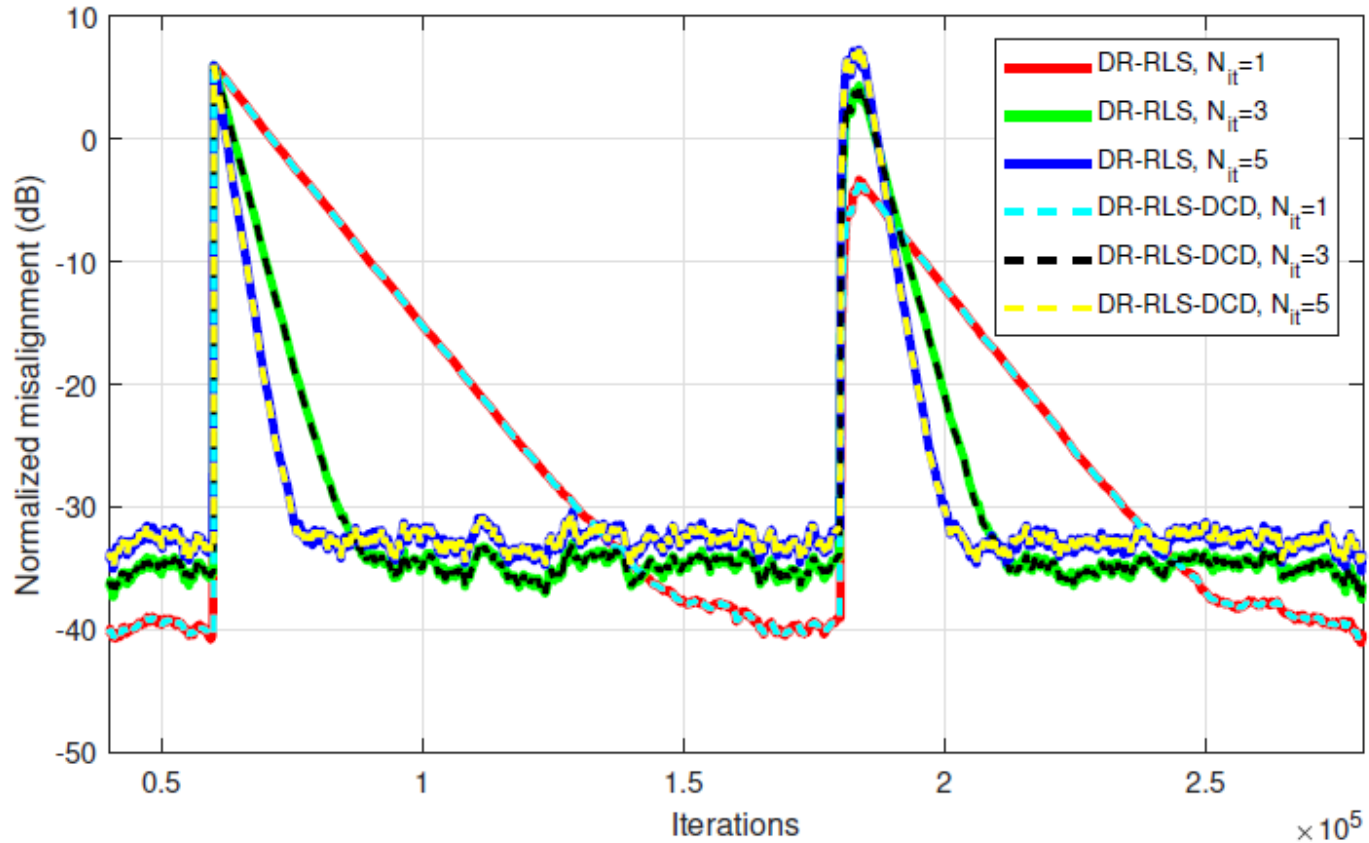
- Run the DCD for N_{it} times
- Low complexity => acceptable cost
- Overall complexity $\sim (N_{it}-1)N_u L$



Simulation Results

- Configuration: system identification (tracking) + low SNR conditions
- DR-RLS vs DR-RLS-DCD
- Unknown system:
 - Fourth impulse response from the G.168 ITU-T Recommendation ($L=128$)
- Input signal:
 - AR(1) sequence with pole 0.9
- Forgetting factor: $\lambda = 1 - 1/(KL)$; $K = 128$
- DCD: $N_u = 4$, $M_b = 16$, $H = 1$
- SNR = 20 dB
- Normalized misalignment: $20 \log_{10} \left[\frac{\|\mathbf{h} - \hat{\mathbf{h}}(n)\|_2}{\|\mathbf{h}\|_2} \right]$ (dB)

Simulation Results



Performance of the DR-RLS and DR-RLS-DCD algorithms for different values of N_{it} . The unknown system changes at time index 60001, and the SNR is decreased for 5000 iterations, starting with index 180001.

Conclusions

- DR-RLS-DCD: **compromise** between **tracking speed** and **accuracy** at steady-state.
- **DR-RLS-DCD matches** the performance of **DR-RLS**.
- **Arithmetic workload** ~ proportional to a small multiple of $L \rightarrow (N_{it}-1)N_uL$
 - **Attractive for hardware implementations**
- Future research:
 - **DR-RLS-DCD with variable N_{it}**
 - ***Regularized/Variable-Regularized* DR-RLS-DCD versions**

Thank you !

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