



# Active Control of Line Spectral Noise with Simultaneous Secondary Path Modeling Without Auxiliary Noise

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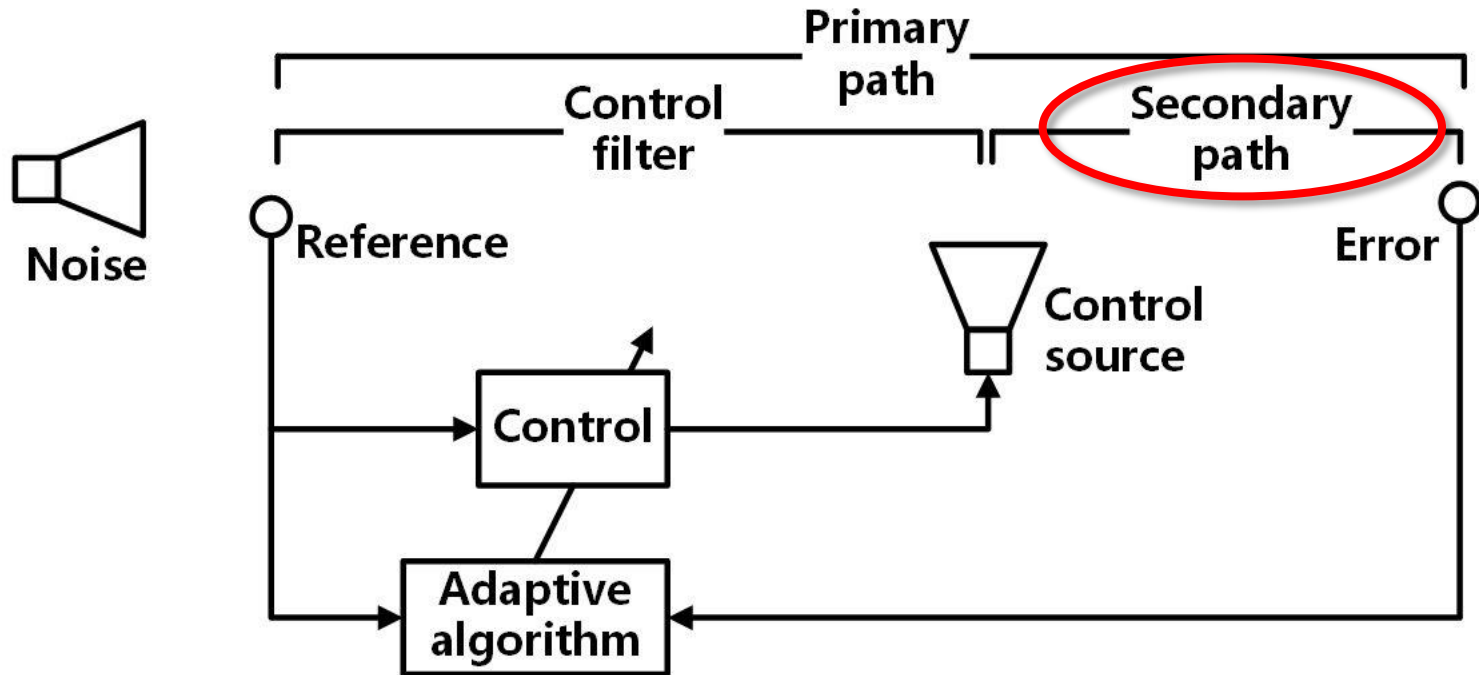
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# Outlines

- ◆ Background
- ◆ Theory
- ◆ Simulations
- ◆ Conclusions

# Background

- ◆ Scheme of ANC



# Background

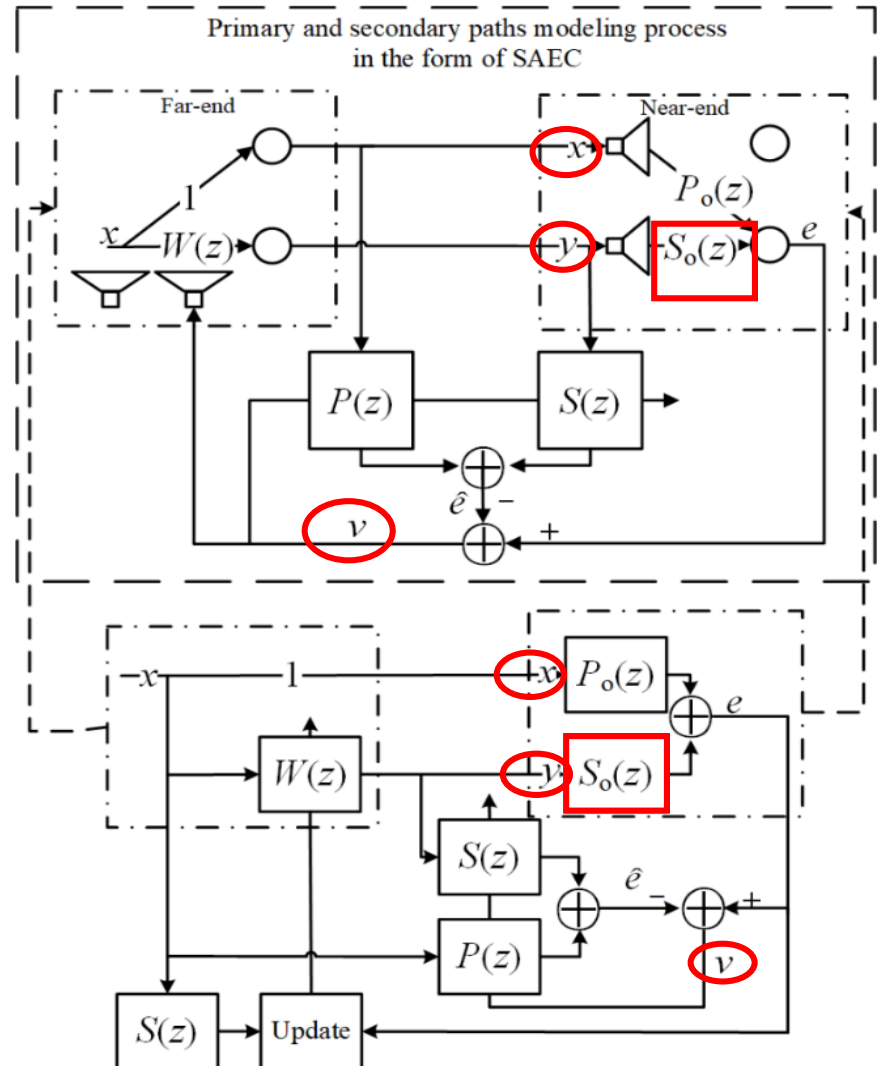
- ◆ Off-line modeling
  - **pros** Easy implementation
  - **cons** Incapable of tracking varying secondary path
  - **cons** Have to be executed before ANC
- ◆ On-line modeling
  - **pros** Theoretically good tracking ability
  - **cons** Require additive noise
  - **cons** Some methods suspend the control process
- ◆ Proposed method
  - **pros** Simple structure
  - **pros** No additive noise

# Background

- ◆ Previous work
  - **Proof** of the effectiveness of the proposed simple structure to **model** and **control simultaneously** under the basic **assumption** that the primary noise is **not line spectral process**
- ◆ This work
  - **Proof** of the effectivity of the proposed simple structure to **model** and **control simultaneously** under the basic assumption that the primary noise is **line spectral process**

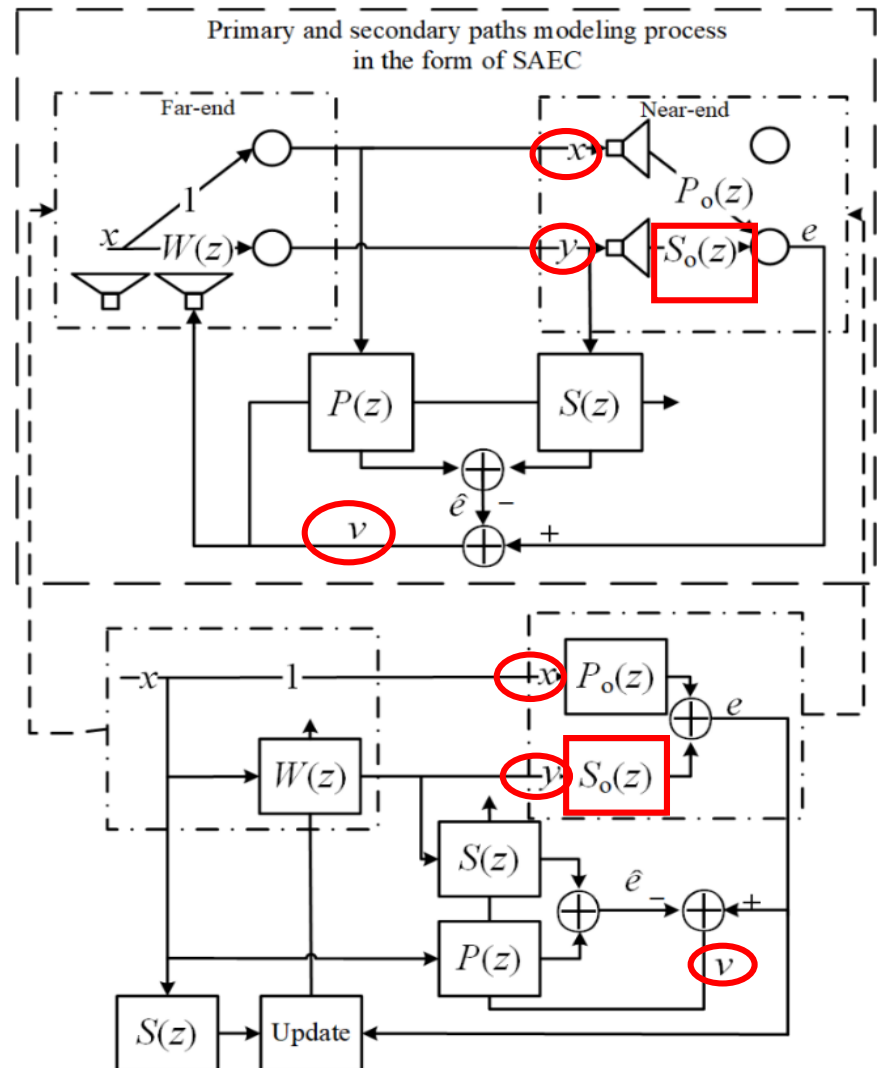
# Theory

- ◆ Note the connection between SAEC and ANC
- ◆ Analysis of non-uniqueness problem of SAEC is helpful in ANC
  - Control filter acts as a time-varying transfer function



# Theory

- ◆ The goal is to prove that the joint auto-correlation matrix of  $x$  and  $y$  is full-rank
  - $x$  is the reference signal
  - $y$  is the output of the control source
  - $x$  and  $y$  are the modeling inputs of the primary and secondary path
  - Full rank joint auto-correlation matrix of  $x$  and  $y$  leads to a wiener function of unique solution while minimizing the cost  $v$ ,  $E[v^2]$



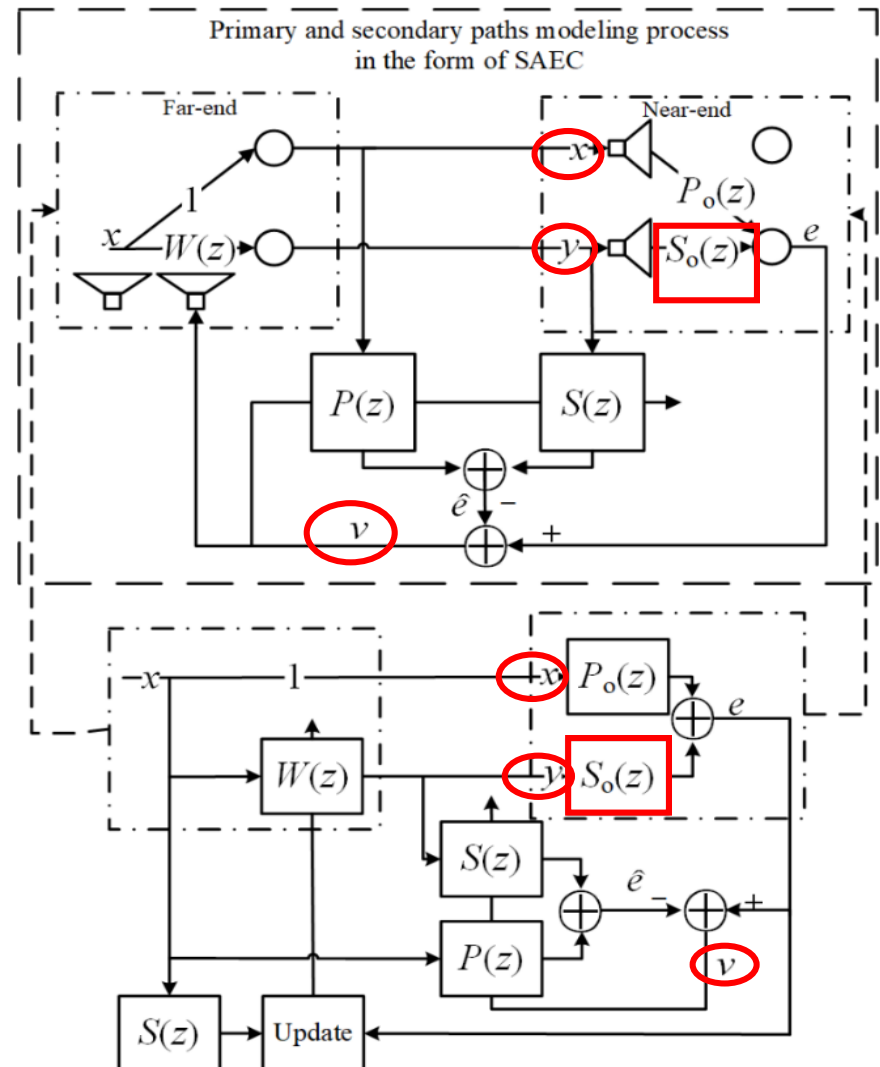
# Theory

## ◆ Assumption

- The reference noise is of line spectral property and composed of  $N/2$  frequencies
- The impulse responses of  $P$ ,  $S$ ,  $W$  are all set as  $N$

## ◆ Used facts

- 1. Time-varying control filter  $W(z)$
- 2. Signal with non-zero power spectrum at  $N/2$  frequencies is full-rank for a correlation matrix of dimension  $N \times N$
- 3. The eigenvalues of the circulant matrix are determined by the Fourier transform of its first row

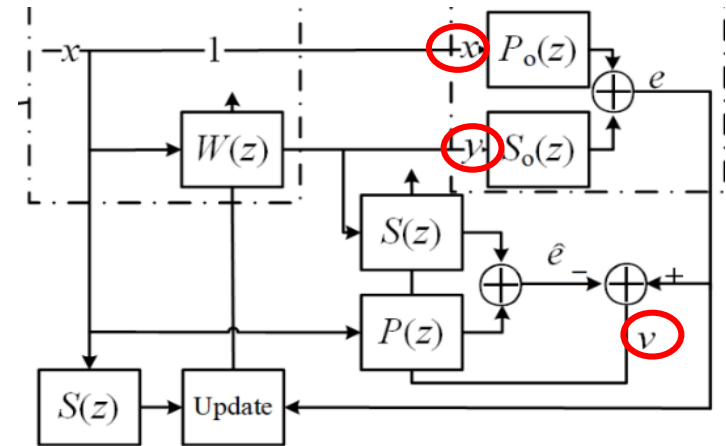




# Theory

## ◆ Cost function

$$E[v^2] = \mathbf{c}^T E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T] \mathbf{c}$$



## ◆ where (assume noise with $N/2$ frequencies)

$$\mathbf{c} = \mathbf{G}^T (\mathbf{p}_o - \mathbf{p}) + \mathbf{W}^T (\mathbf{s}_o - \mathbf{s})$$

$$\mathbf{G} = [\mathbf{I} \quad \mathbf{0}]_{N \times 2N}$$

$$\tilde{\mathbf{x}}(n) = \begin{bmatrix} \mathbf{x}^T(n) & \mathbf{x}^T(n) \end{bmatrix}^T \quad \mathbf{x}(n) = \begin{bmatrix} x(n) & \cdots & x(n-N+1) \end{bmatrix}^T$$

$$\mathbf{W} = \begin{bmatrix} w(1) & w(2) & \cdots & w(N) & 0 & \cdots & 0 & 0 \\ 0 & w(1) & w(2) & \cdots & w(N) & \ddots & \vdots & 0 \\ \vdots & 0 & \ddots & \cdots & \cdots & \ddots & 0 & \vdots \\ 0 & \cdots & 0 & w(1) & w(2) & \cdots & w(N) & 0 \end{bmatrix}$$

# Theory

- ◆ Take the derivative of the cost function with respect to  $\mathbf{s}$

$$\frac{\partial E[v^2]}{\partial \mathbf{s}} = -2\mathbf{W}E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T]\mathbf{c}$$

- ◆ Initialize  $\mathbf{w}$  as a non-zero vector, then  $\mathbf{W}$  has full row rank

$$\mathbf{W}E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T]\mathbf{c} = 0 \Leftrightarrow E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T]\mathbf{c} = 0$$

- ◆ Utilize time-varying property of  $\mathbf{w}$

$$\mathbf{c} = \mathbf{G}^T(\mathbf{p}_o - \mathbf{p}) + \mathbf{W}^T(\mathbf{s}_o - \mathbf{s})$$

$$\mathbf{c}_* = \mathbf{G}^T(\mathbf{p}_o - \mathbf{p}) + \mathbf{W}_*^T(\mathbf{s}_o - \mathbf{s})$$

$$E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T]\mathbf{c} = 0, E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T]\mathbf{c}_* = 0$$

**Fact 1**

Time-varying control filter  $W(z)$

# Theory

- ◆ Subtraction of  $E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T]\mathbf{c} = 0$  and  $E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T]\mathbf{c}_* = 0$  leads to

$$E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T](\mathbf{c} - \mathbf{c}_*) = E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T]\tilde{\mathbf{W}}^T(\mathbf{s}_o - \mathbf{s}) = 0$$

$$\tilde{\mathbf{W}} = (\mathbf{W} - \mathbf{W}_*)$$

- ◆ Divide  $\tilde{\mathbf{W}}$  into left and right part  $\tilde{\mathbf{W}}_{(1)}$  and  $\tilde{\mathbf{W}}_{(2)}$
- ◆  $E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T]$  can be denoted as

$$E[\tilde{\mathbf{x}}\tilde{\mathbf{x}}^T] = \begin{bmatrix} \mathbf{R} & \mathbf{R} \\ \mathbf{R} & \mathbf{R} \end{bmatrix} \quad \mathbf{W} = \begin{bmatrix} w(1) & w(2) & \cdots & w(N) & | & 0 & \cdots & 0 & 0 \\ 0 & w(1) & w(2) & \cdots & | & w(N) & \ddots & \vdots & 0 \\ \vdots & 0 & \ddots & \tilde{\mathbf{W}}_{(1)} & \cdots & \cdots & \tilde{\mathbf{W}}_{(2)} & 0 & \vdots \\ 0 & \cdots & 0 & w(1) & | & w(2) & \cdots & w(N) & 0 \end{bmatrix}$$

# Theory

- ◆  $\mathbf{R} \left( \tilde{\mathbf{W}}_{(1)} + \tilde{\mathbf{W}}_{(2)} \right)^T (\mathbf{s}_o - \mathbf{s}) = 0$

**Fact 2**



Signal with non-zero power spectrum at  $N/2$  frequencies is full-rank for a correlation matrix of dimension  $N \times N$

- ◆  $\left( \tilde{\mathbf{W}}_{(1)} + \tilde{\mathbf{W}}_{(2)} \right)^T (\mathbf{s}_o - \mathbf{s}) = 0$

- ◆ Where

$$\tilde{\mathbf{W}}_{(1)} + \tilde{\mathbf{W}}_{(2)} = \begin{bmatrix} \tilde{w}(1) & \tilde{w}(2) & \cdots & \tilde{w}(N) \\ \tilde{w}(N) & \tilde{w}(1) & \cdots & \tilde{w}(N-1) \\ \vdots & \ddots & \ddots & \vdots \\ \tilde{w}(2) & \tilde{w}(3) & \cdots & \tilde{w}(1) \end{bmatrix}$$

- ◆ is full-rank

**Fact 3**

The eigenvalues of the circulant matrix are determined by the Fourier transform of its first row

# Simulations

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## Initializing $\mathbf{x}$ , $\mathbf{y}$ , $\mathbf{u}$ , $\mathbf{h}$ , $\mathbf{f}$ , $\mathbf{Q}$ , $\lambda$ , $\mu$ , and $\mathbf{w}$

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for  $n = 0, 1, 2, \dots$  do

(a) for  $i = 0$  to  $N - 1$ ,  $y(n - i) = \mathbf{w}^T \mathbf{x}(n - i)$ ; stack  $\mathbf{x}(n)$  and  $\mathbf{y}(n)$  into  $\mathbf{u}(n)$ ;

(b) **modeling process** using the RLS algorithm

$$\mathbf{k} = \mathbf{Q}\mathbf{u}(n) / (\lambda + \mathbf{u}^T(n)\mathbf{Q}\mathbf{u}(n))$$

$$\mathbf{h} = \mathbf{h} + \mathbf{k}(e(n) - \mathbf{h}^T \mathbf{u}(n))$$

$$\mathbf{Q} = (\mathbf{Q} - \mathbf{k}\mathbf{u}^T(n)\mathbf{Q}) / \lambda$$

assign the last  $N$  taps of  $\mathbf{h}$  to  $\mathbf{s}$ ;

(c) for  $i = 0$  to  $N - 1$ ,  $f(n - i) = \mathbf{s}^T \mathbf{x}(n - i)$ ;

(d) **control process** using the LMS algorithm

$$\mathbf{w} = \mathbf{w} - 2\mu\mathbf{f}(n)\mathbf{e}(n).$$

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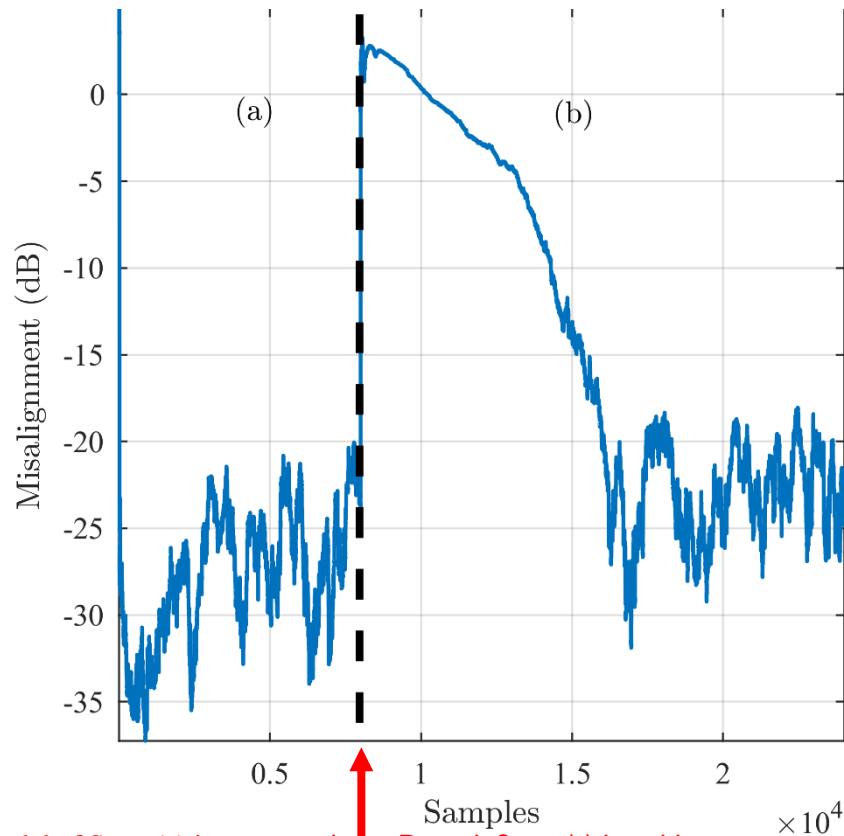
Excited frequency  
number is 3 so  $N = 6$

Forgetting factor of  
RLS is 0.999

Learning rate  $\mu = 0.01$

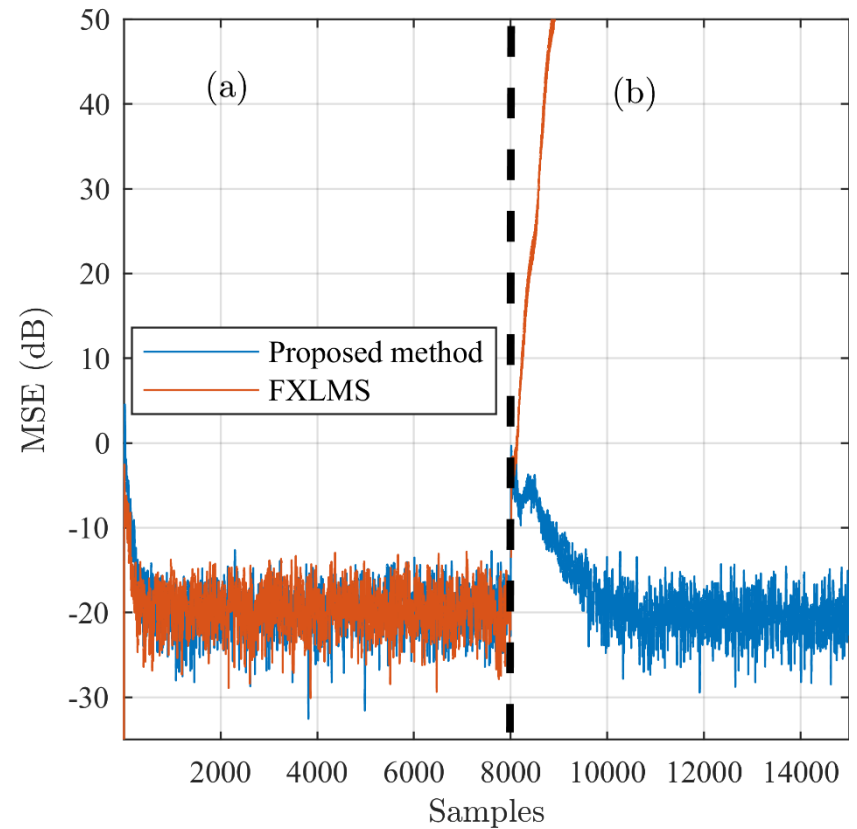
$\mathbf{w}$  is initialized as  
zero except for  
0.001 at the first tap

# Simulations



Model of State (a) is measured in the listening room of the Institute of Acoustics, Nanjing University

Disturb State (a) by adding random noise with SNR = 20 dB to give a different set of transfer functions, denoted by State (b)



# Conclusions

- ◆ For the noise of line spectral process, the modeling process of the secondary path that uses only the output of the control filter is possible.
- ◆ The secondary path is guaranteed to converge to the optimum as long as the number of non-zero spectrum frequencies is known and the control filter is set as twice this number.



Thanks for your attention!