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#### Truncated Singular Value Decomposition Method for Mitigating Unwanted Enhancement in Active Noise Control Systems

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#### Truncated Singular Value Decomposition Method for Mitigating Unwanted Enhancement in Active Noise Control Systems

\* This presentation is a part of the INTER-NOISE 2018 Student Paper Competition.

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#### Content













# **Enhancement Problem**

- Possible Reasons:
  - System Delay
  - Feedback Path Cancellation
  - Correlated Reference Signals
  - ...
- Focus:
  - Correlated Reference Signals Problem in MIMO Feed-forward Non-adaptive Active Noise Control System



### **Experiment Setup**



#### **Experiment Setup**



#### Case 1

INTER-NOISE **2018** Impact of Noise Control Engineering



#### Case 2

INTER-NOISE **2018** Impact of Noise Control Engineering





# Feed-forward Active Noise Control System



# **Coefficients of Controller**

\*S. J. Elliott, Signal processing for active control, Elsevier, (2000).

$$J = E[\vec{e}^T(n) \cdot \vec{e}(n)] = \vec{w}^T A \vec{w} + 2 \vec{w}^T \vec{b} + c$$
$$\vec{w}_{opt} = -A^{-1} \vec{b}$$

A – Autocorrelation function of the filtered reference signals

 $\vec{b}$  – Cross-correlation function between the filtered reference signals and the disturbance signals

c – Autocorrelation function of the disturbance signals



# **Singular Value Decomposition Analysis**

SVD is used to analysis the autocorrelation function of the filtered reference signals.

$$A = U\Sigma U^{T} = \begin{bmatrix} \vec{u}_{1} \cdots \vec{u}_{N} \end{bmatrix} \begin{bmatrix} \sigma_{1} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \sigma_{N} \end{bmatrix} \begin{bmatrix} \vec{u}_{1} \\ \vdots \\ \vec{u}_{N} \end{bmatrix}$$
$$\vec{v}_{opt} = -\begin{bmatrix} \vec{u}_{1} \cdots \vec{u}_{N} \end{bmatrix} \begin{bmatrix} \sigma_{1}^{-1} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \sigma_{N}^{-1} \end{bmatrix} \begin{bmatrix} \vec{u}_{1} \\ \vdots \\ \vec{u}_{N} \end{bmatrix} \vec{b} = -\sum_{n=1}^{N} \sigma_{n}^{-1} \langle \vec{u}_{n}, \vec{b} \rangle \vec{u}_{n}$$



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# **Singular Values & Projection Product**



# **Singular Value Decomposition Analysis**

SVD is used to analysis the autocorrelation function of the filtered reference signals.

$$A = U\Sigma U^{T} = \begin{bmatrix} \vec{u}_{1} \cdots \vec{u}_{N} \end{bmatrix} \begin{bmatrix} \sigma_{1} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \sigma_{N} \end{bmatrix} \begin{bmatrix} \vec{u}_{1} \\ \vdots \\ \vec{u}_{N} \end{bmatrix}$$
$$\vec{v}_{opt} = -\begin{bmatrix} \vec{u}_{1} \cdots \vec{u}_{N} \end{bmatrix} \begin{bmatrix} \sigma_{1}^{-1} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \sigma_{N}^{-1} \end{bmatrix} \begin{bmatrix} \vec{u}_{1} \\ \vdots \\ \vec{u}_{N} \end{bmatrix} \vec{b} = -\sum_{n=1}^{N} \sigma_{n}^{-1} \langle \vec{u}_{n}, \vec{b} \rangle \vec{u}_{n}$$



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#### **Uncorrelated Noise Sources Case**



### **Correlated Noise Sources Case**



# **Comparison of Two Cases**

INTER-NOISE **2018** Impact of Noise Control Engineering



#### **Truncated Singular Value Decomposition Method**

*l* : Truncation Point

$$\vec{\boldsymbol{w}}_{opt} = -\begin{bmatrix} \vec{\boldsymbol{u}}_1 \cdots \vec{\boldsymbol{u}}_N \end{bmatrix} \begin{bmatrix} \sigma_1^{-1} & \cdots & 0\\ \vdots & \ddots & \vdots\\ 0 & \cdots & \sigma_N^{-1} \end{bmatrix} \begin{bmatrix} \vec{\boldsymbol{u}}_1\\ \vdots\\ \vec{\boldsymbol{u}}_N \end{bmatrix} \vec{\boldsymbol{b}} = -\sum_{n=1}^N \sigma_n^{-1} \langle \vec{\boldsymbol{u}}_n, \vec{\boldsymbol{b}} \rangle \vec{\boldsymbol{u}}_n$$

$$\vec{\boldsymbol{w}}^*_{opt} = -\left[\vec{\boldsymbol{u}}_1 \cdots \vec{\boldsymbol{u}}_l\right] \begin{bmatrix} \sigma_1^{-1} & \cdots & 0\\ \vdots & \ddots & \vdots\\ 0 & \cdots & \sigma_l^{-1} \end{bmatrix} \begin{bmatrix} \vec{\boldsymbol{u}}_1\\ \vdots\\ \vec{\boldsymbol{u}}_l \end{bmatrix} \vec{\boldsymbol{b}} = -\sum_{n=1}^l \sigma_n^{-1} \langle \vec{\boldsymbol{u}}_n, \vec{\boldsymbol{b}} \rangle \vec{\boldsymbol{u}}_n$$





















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# Conclusions

- The large reciprocal of singular values  $\sigma_n^{-1}$  and the projection product of  $\langle \vec{u}_n, \vec{b} \rangle$  are two causes of the noise enhancement.
- The singular vectors of some indices contribute to both noise enhancement and noise cancellation.
- Still need more tries to find the best index for truncation point to improve the performance or mitigate the unwanted enhancement.



Thank you!