

# Design and Implementation of Small Microphone Arrays

for Acoustic and Speech Signal Processing

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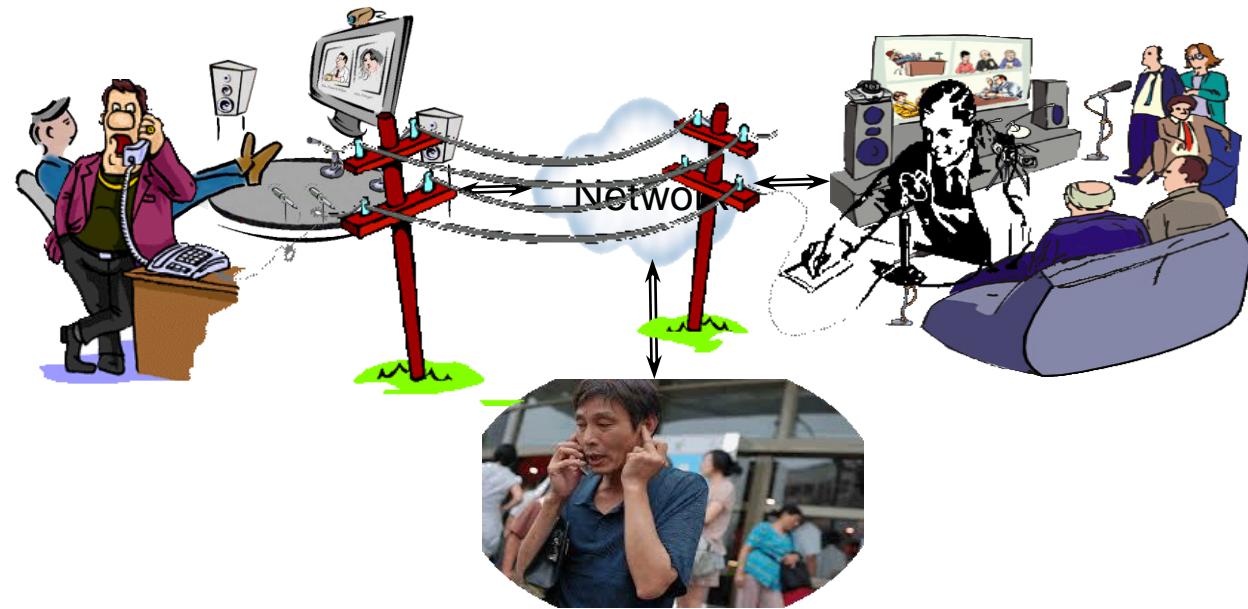


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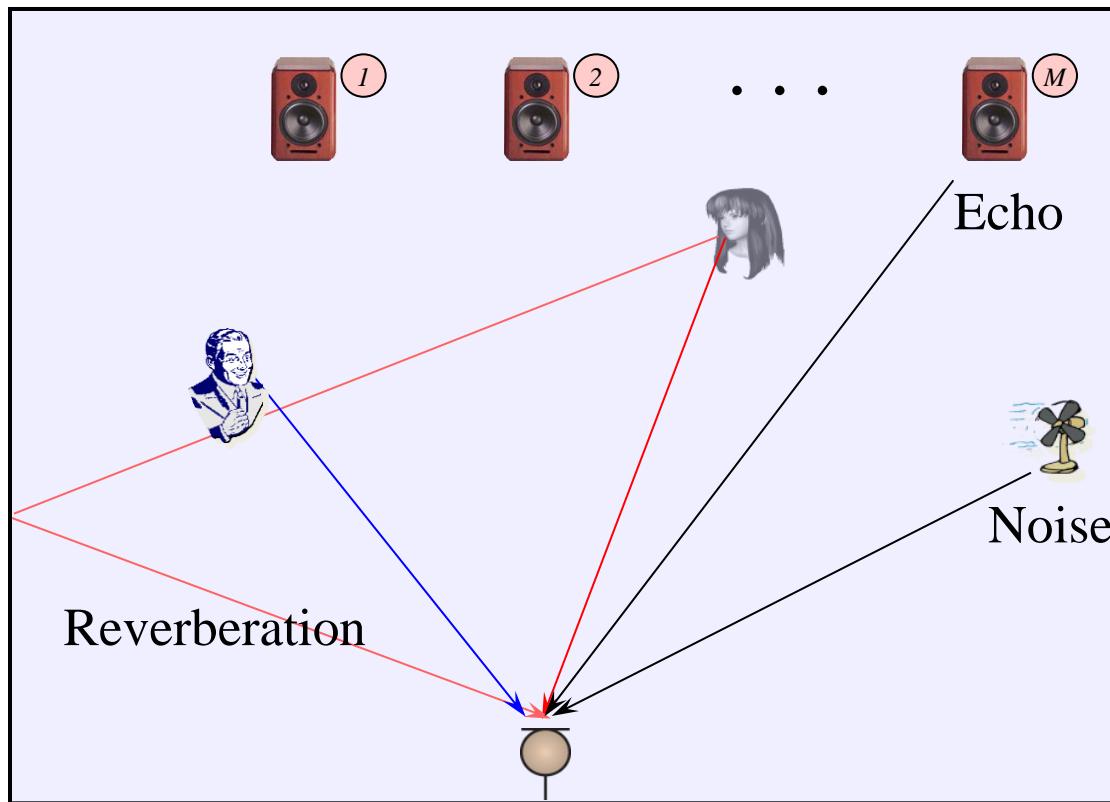
- Background

Tradition Voice Communication



# Design and Implementation of Small Microphone Arrays

## • Background



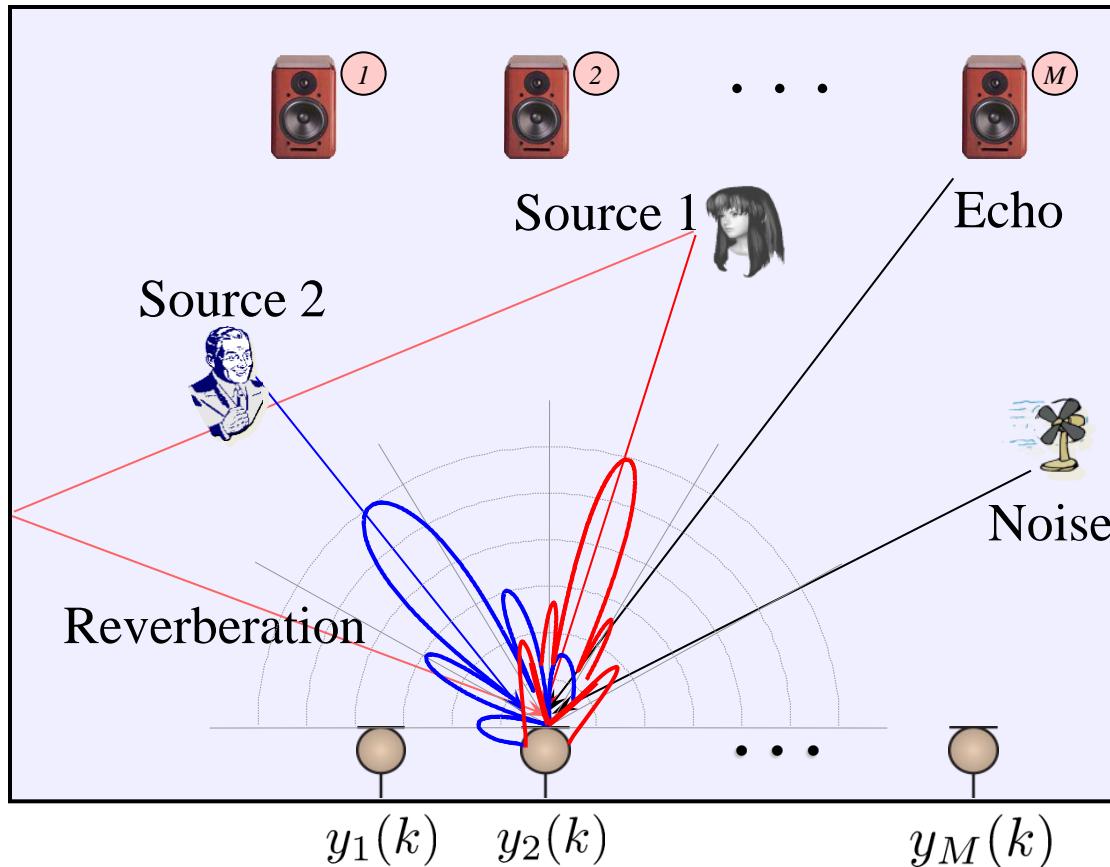
- Noise

- Reverberation

- Interference

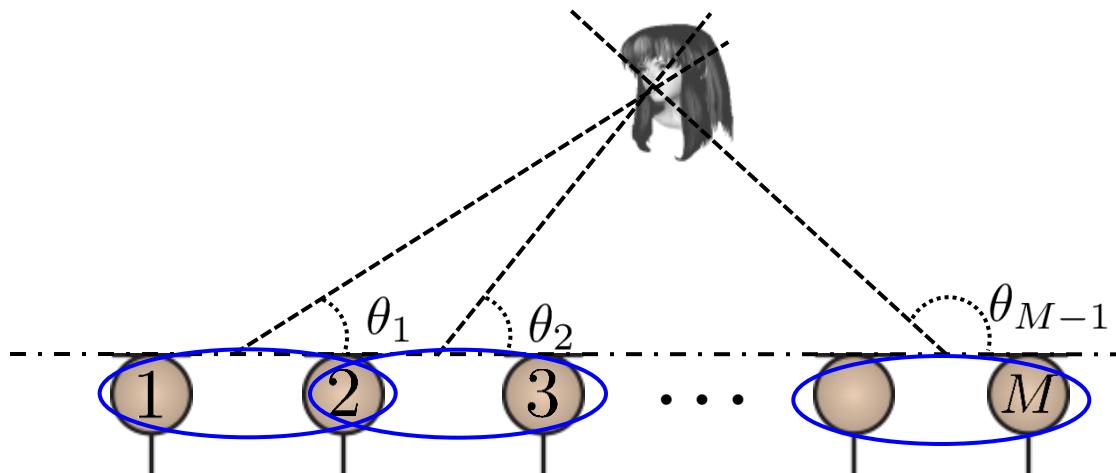
- Echo

- Why Multiple Microphones?



- Source extraction
- Noise, interference, reverberation, and echo suppression
- Source separation

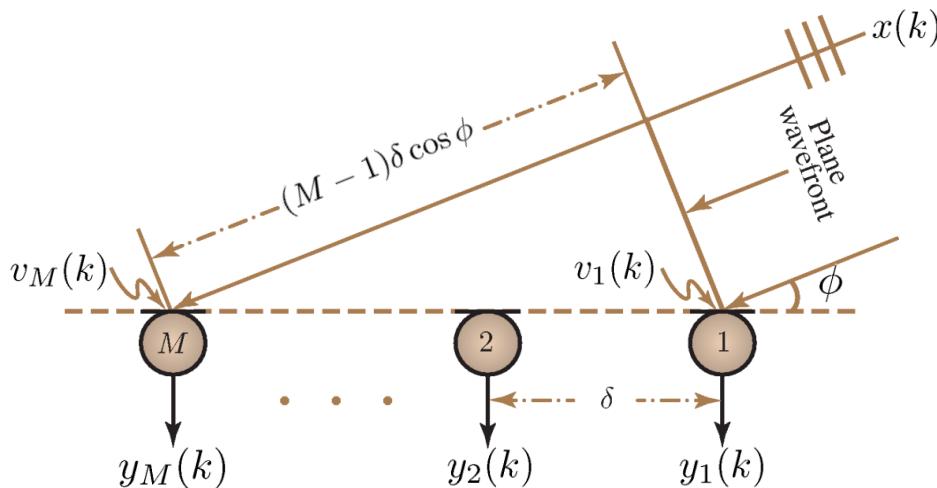
- Why Multiple Microphones?



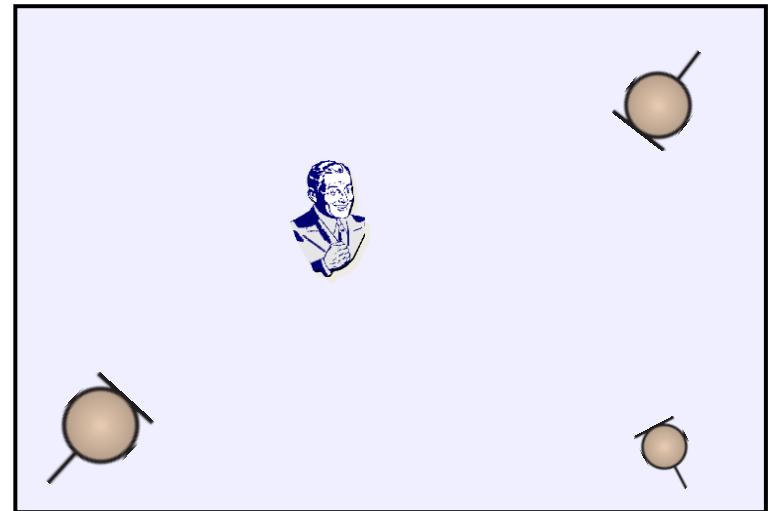
- Source extraction
- Noise, interference, reverberation, and echo suppression
- Source separation
- DOA estimation
- Source localization

## • Multiple-Microphone Systems

### Organized arrays



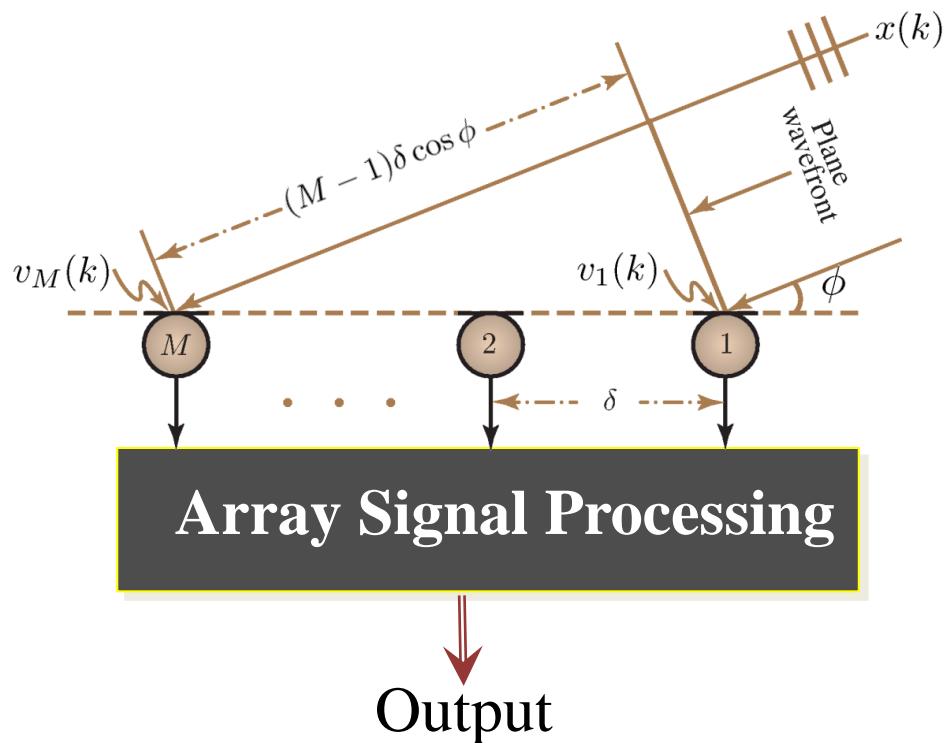
### Disorganized systems



- Geometry is known
- Sensors are uniform in responses
- Sampled with the same clock

- Geometry is unknown or time varying
- Sensors may be different in responses
- Clock skew

## • Microphone Array and Beamforming

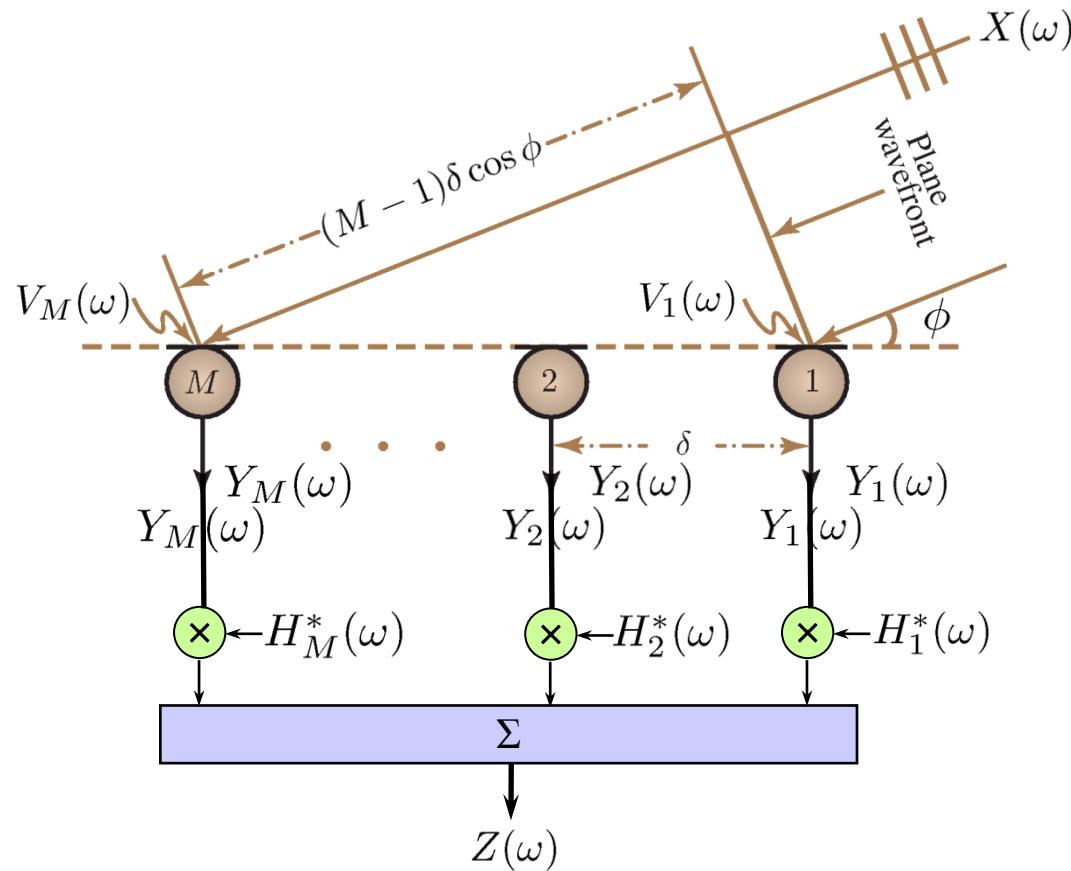


- Geometry
- Selection of sensors
- Calibration
- A/D

- Beamforming
- Multichannel NR
- Source Separation
- Source Localization

# Design and Implementation of Small Microphone Arrays

- Beamforming



- Beamforming

$$Z(\omega) = \sum_{m=1}^M H_m^*(\omega) Y_m(\omega) = \mathbf{h}^H(\omega) \mathbf{y}(\omega)$$

$$= \mathbf{h}^H(\omega) \mathbf{d}(\omega, \cos \phi) X(\omega) + \mathbf{h}^H(\omega) \mathbf{v}(\omega)$$

$$\mathbf{h}(\omega) = [ H_1(\omega) \quad H_2(\omega) \quad \cdots \quad H_M(\omega) ]^T$$

$$\mathbf{d}(\omega, \cos \theta) = [ 1 \quad e^{-j\omega\delta \cos \phi/c} \quad \dots \quad e^{-j(M-1)\omega\delta \cos \phi/c} ]^T$$

The core problem of beamforming design  
is to the optimal beamforming filter  $\mathbf{h}(\omega)$



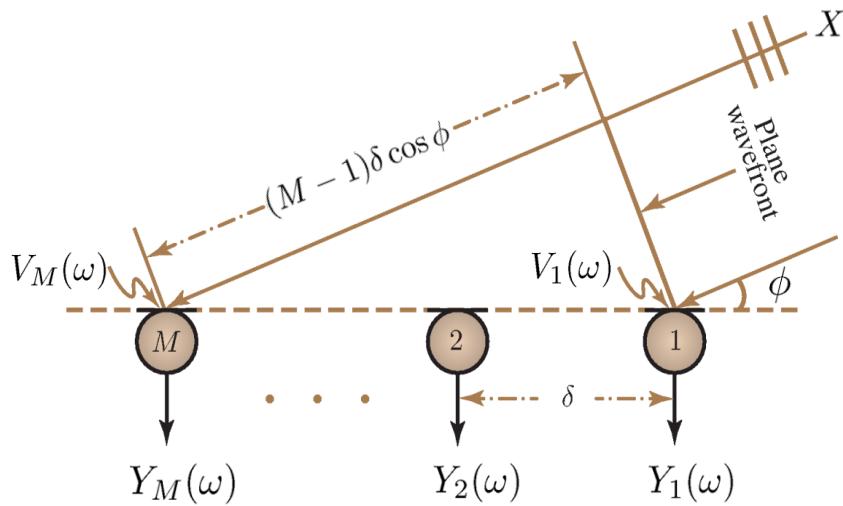
- Performance Criterion

Signal model:

$$Y_m(\omega) = e^{-j(m-1)\omega\tau_0 \cos \phi} X(\omega) + V_m(\omega)$$

Beampattern:

Describes the sensitivity of a beamformer to a plane wave impinging on the array from the direction  $\theta$ )



$$Z(\omega) = \mathbf{h}^H(\omega) \mathbf{d}(\omega, \cos \phi) X(\omega)$$

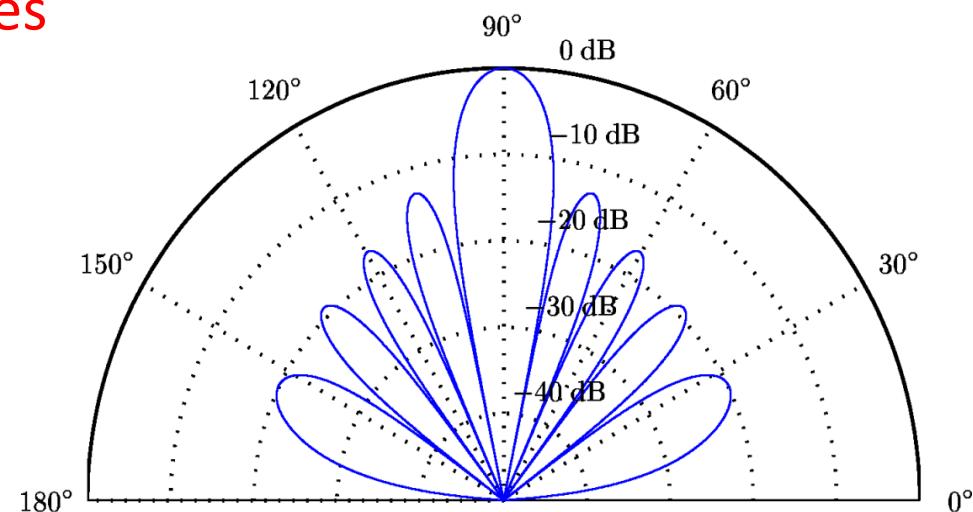
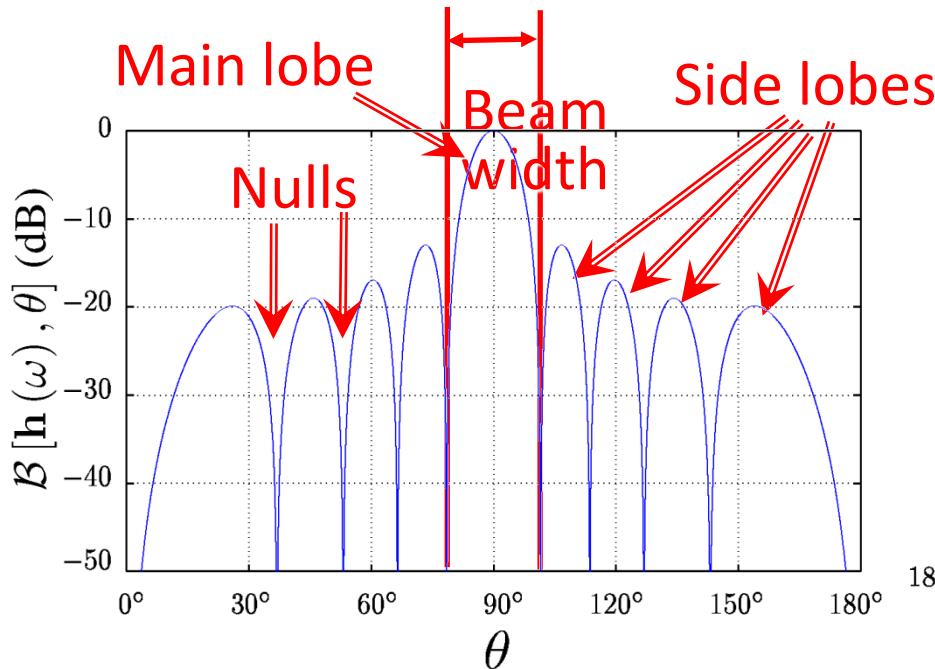
$$\mathcal{B}[\mathbf{h}(\omega), \theta] = \mathbf{d}^H(\omega, \cos \phi) \mathbf{h}(\omega)$$

$$= \sum_{m=1}^M H_m(\omega) e^{j(m-1)\omega\tau_0 \cos \phi}$$

# Design and Implementation of Small Microphone Arrays

- Beamforming

DS Beampattern:  $\mathcal{B}[\mathbf{h}(\omega), \theta] = \left| \frac{m \sin[M\delta(\cos\phi - \cos\theta)/(2c)]}{M \sin[\omega\delta(\cos\phi - \cos\theta)/(2c)]} \right|$

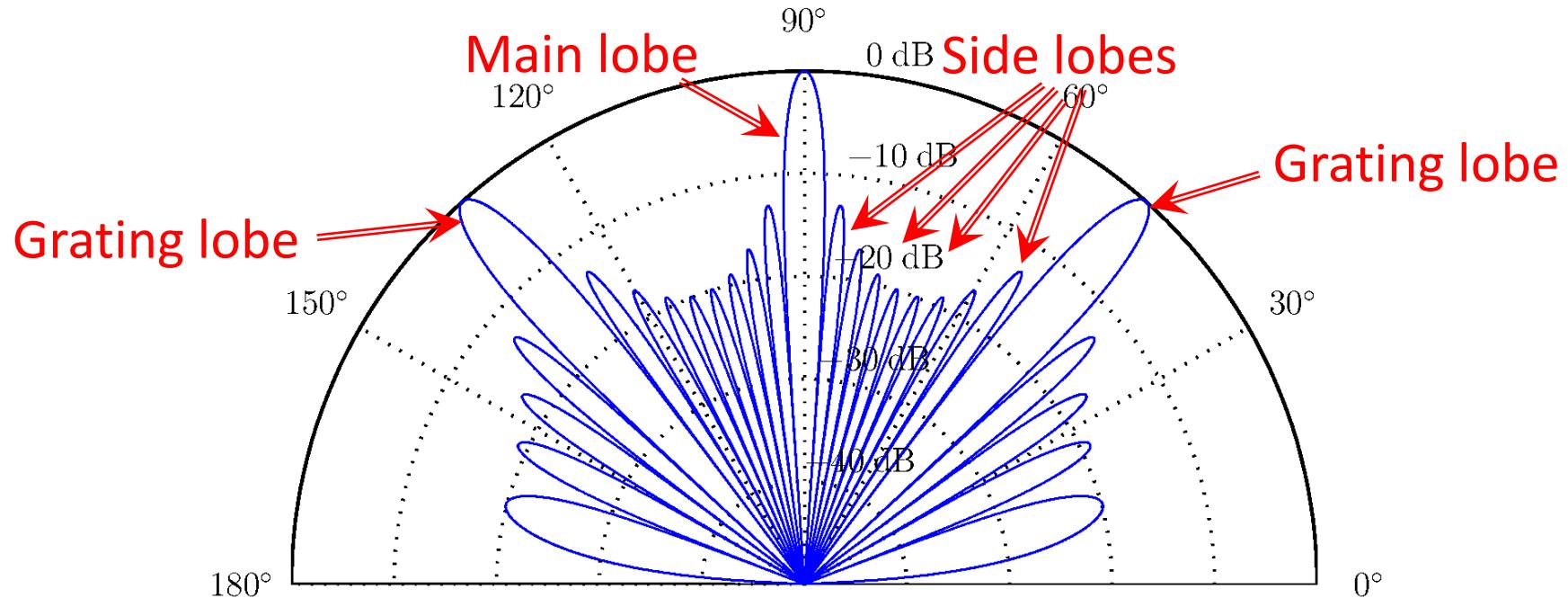


Beampattern of a DS beamformer with a ten-sensor array when  $\phi = 90^\circ$ ,  $\delta = 8$  cm, and  $f = 2$  kHz

- Beamforming

DS Beampattern:

$$\mathcal{B} [\mathbf{h} (\omega), \theta] = \left| \frac{\sin [M\omega d(\cos \phi - \cos \theta)/(2c)]}{M \sin [\omega \delta(\cos \phi - \cos \theta)/(2c)]} \right|$$



Beampattern of a DS beamformer with a ten-sensor array when  
 $\phi = 90^\circ$ ,  $\delta = 24$  cm, and  $f = 2$  kHz

- Performance Measures (Cont'd)
  - SNR Gain

Signal model:  $Y_m(\omega) = e^{-j(m-1)\omega\tau_0 \cos \phi} X(\omega) + V_m(\omega)$

Input SNR:  $iSNR(\omega) = \frac{\phi_X(\omega)}{\phi_{V_1}(\omega)}$

BF output:  $Z(\omega) = \mathbf{h}^H(\omega) \mathbf{y}(\omega) = \mathbf{h}^H(\omega) \mathbf{x}(\omega) + \mathbf{h}^H(\omega) \mathbf{v}(\omega)$

Output:  $\Phi_v(\omega) = E[\mathbf{v}(\omega) \mathbf{v}^H(\omega)]$

$$\omega) \frac{|\mathbf{h}^H(\omega) \mathbf{d}(\omega, \cos 0^\circ)|^2}{\mathbf{h}^H(\omega) \Phi_v(\omega) \mathbf{h}(\omega)}$$

$$\Gamma_v(\omega) = \frac{\Phi_v(\omega)}{\phi_{V_1}(\omega)}$$

$$\omega) \frac{|\mathbf{h}^H(\omega) \mathbf{d}(\omega, \cos 0^\circ)|^2}{\mathbf{h}^H(\omega) \Gamma_v(\omega) \mathbf{h}(\omega)}$$



- Performance Measures (Cont'd)

- SNR Gain

$$\text{SNR Gain: } \mathcal{G} [\mathbf{h} (\omega)] = \frac{\text{oSNR} [\mathbf{h} (\omega)]}{\text{iSNR} (\omega)} = \frac{\left| \mathbf{h}^H (\omega) \mathbf{d} (\omega, \cos 0^\circ) \right|^2}{\mathbf{h}^H (\omega) \boldsymbol{\Gamma}_v (\omega) \mathbf{h} (\omega)}$$

**SNR gain depends on the noise  
pseudo-coherence matrix**

- White Noise Gain (SNR gain in white noise)

$$\text{White Noise: } \boldsymbol{\Gamma}_v (\omega) = \mathbf{I}_M$$

WNG:

$$\mathcal{G}_{\text{wn}} [\mathbf{h} (\omega)] = \frac{\left| \mathbf{h}^H (\omega) \mathbf{d} (\omega, \cos 0^\circ) \right|^2}{\mathbf{h}^H (\omega) \mathbf{h} (\omega)}$$



- Performance Measures (Cont'd)

- SNR Gain

$$\text{SNR Gain: } \mathcal{G} [\mathbf{h} (\omega)] = \frac{\text{oSNR} [\mathbf{h} (\omega)]}{\text{iSNR} (\omega)} = \frac{\left| \mathbf{h}^H (\omega) \mathbf{d} (\omega, \cos 0^\circ) \right|^2}{\mathbf{h}^H (\omega) \boldsymbol{\Gamma}_v (\omega) \mathbf{h} (\omega)}$$

- Directivity Index (SNR gain in diffuse noise)

$$\text{Diffuse Noise: } [\boldsymbol{\Gamma}_v (\omega)]_{ij} = \frac{\sin [\omega(j-i)\tau_0]}{\omega(j-i)\tau_0} = \text{sinc} [\omega(j-i)\tau_0]$$

Directivity Index:  $\mathcal{G} [\mathbf{h} (\omega)] \rightarrow \mathcal{G}_{dn} [\mathbf{h} (\omega)]$

$$\mathcal{D} [\mathbf{h} (\omega)] = 10 \log_{10} \mathcal{G}_{dn} [\mathbf{h} (\omega)]$$



- Performance Measures (Cont'd)

- SNR Gain

SNR Gain:  $\mathcal{G} [\mathbf{h} (\omega)] = \frac{\text{oSNR} [\mathbf{h} (\omega)]}{\text{iSNR} (\omega)} = \frac{\left| \mathbf{h}^H (\omega) \mathbf{d} (\omega, \cos 0^\circ) \right|^2}{\mathbf{h}^H (\omega) \boldsymbol{\Gamma}_{\mathbf{v}} (\omega) \mathbf{h} (\omega)}$

- SNR gain in point-source noise

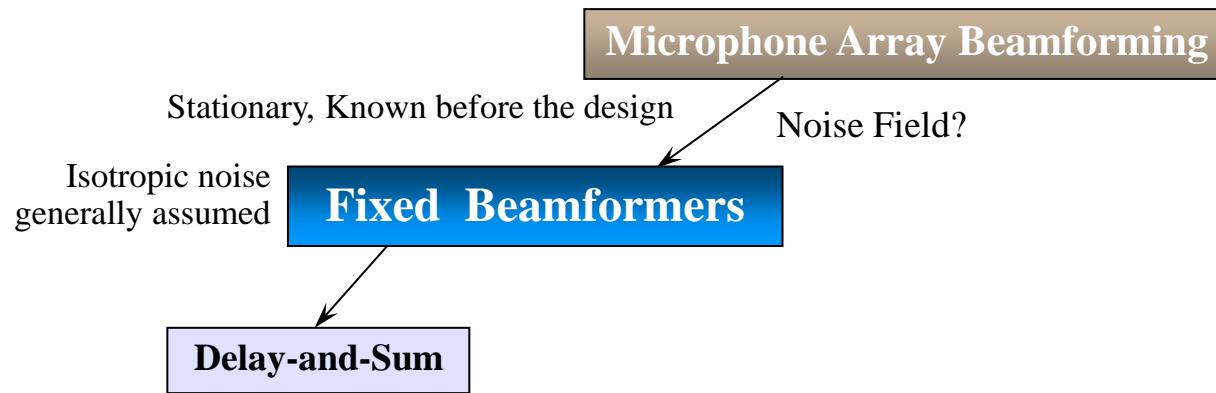
Point Noise:  $\boldsymbol{\Gamma}_{\mathbf{v}} (\omega) = \mathbf{d} (\omega, \cos \theta_n) \mathbf{d}^H (\omega, \cos \theta_n)$

SNR Gain:  $\mathcal{G} [\mathbf{h} (\omega)] \rightarrow \mathcal{G}_{pn} [\mathbf{h} (\omega), \theta_n]$

$$\mathcal{G}_{pn} [\mathbf{h} (\omega), \theta_n] = \frac{1}{|\mathcal{B} (\mathbf{h} (\omega), \theta_n)|^2}$$



- Beamforming



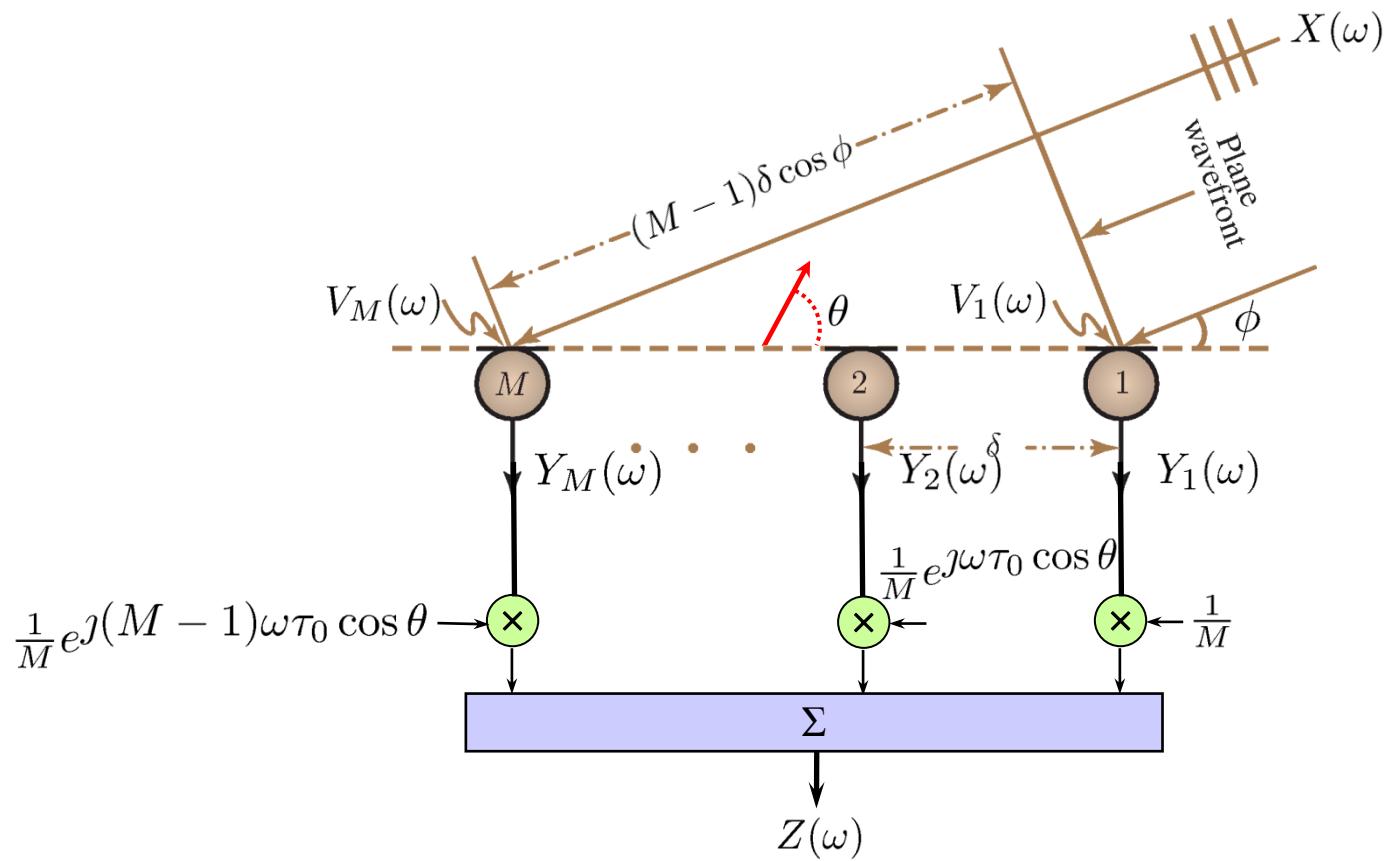
## Delay-and-Sum

- Simple
- Non-uniform directional responses over a wide spectrum of frequencies

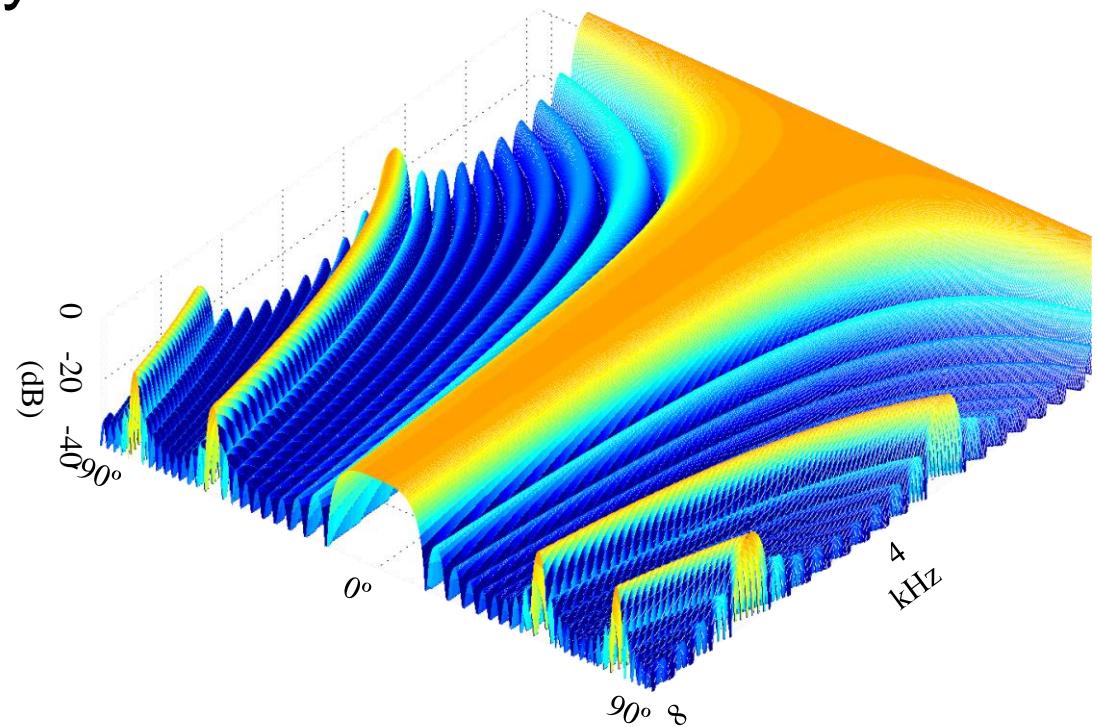


# Design and Implementation of Small Microphone Arrays

- Delay and Sum



- Delay and Sum

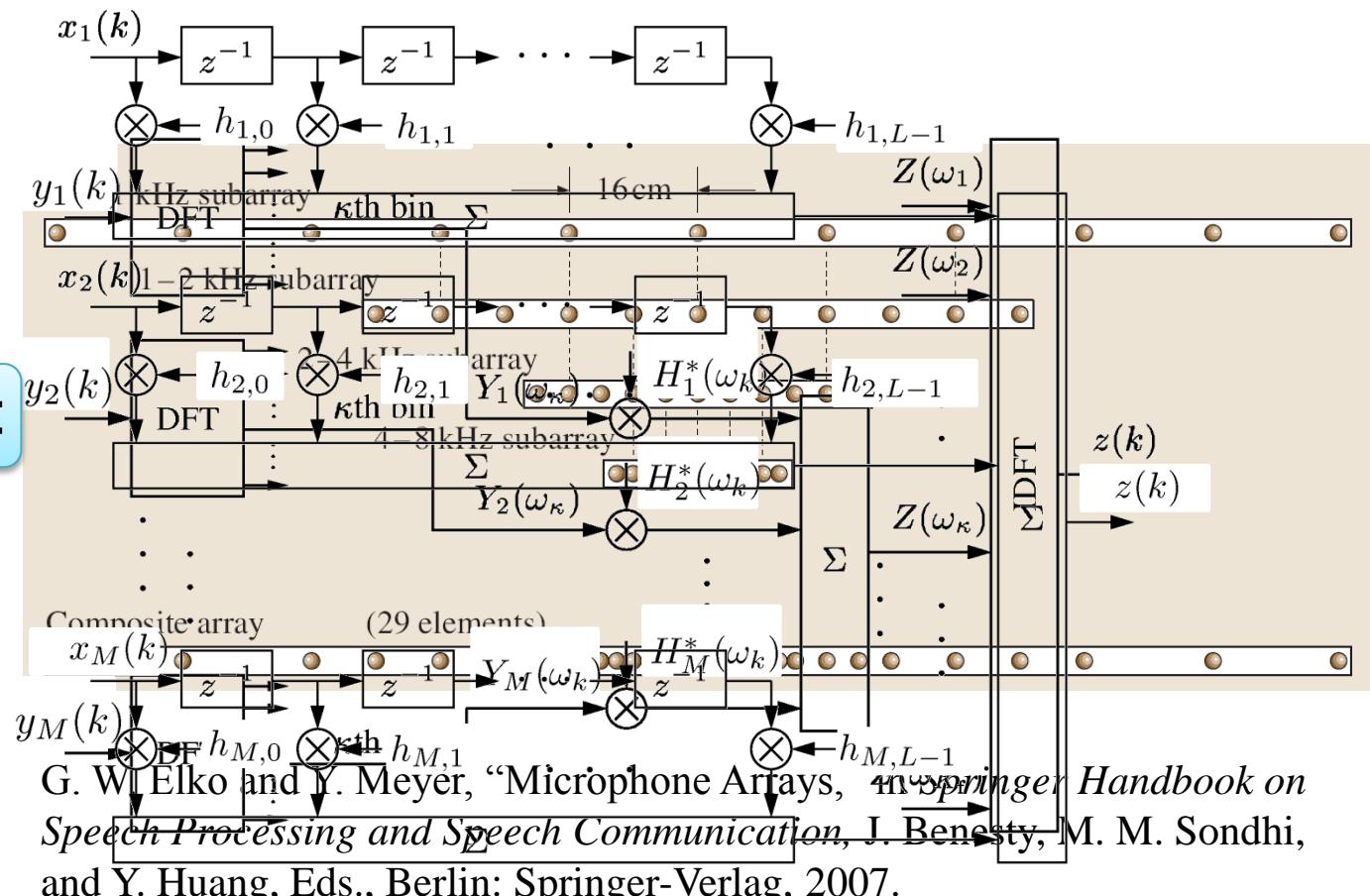


- Nonuniform beam width (spectral tilt, lowpass filtering the desired speech signal)
- Not very effective in reducing the reverberation effect.

# Design and Implementation of Small Microphone Arrays

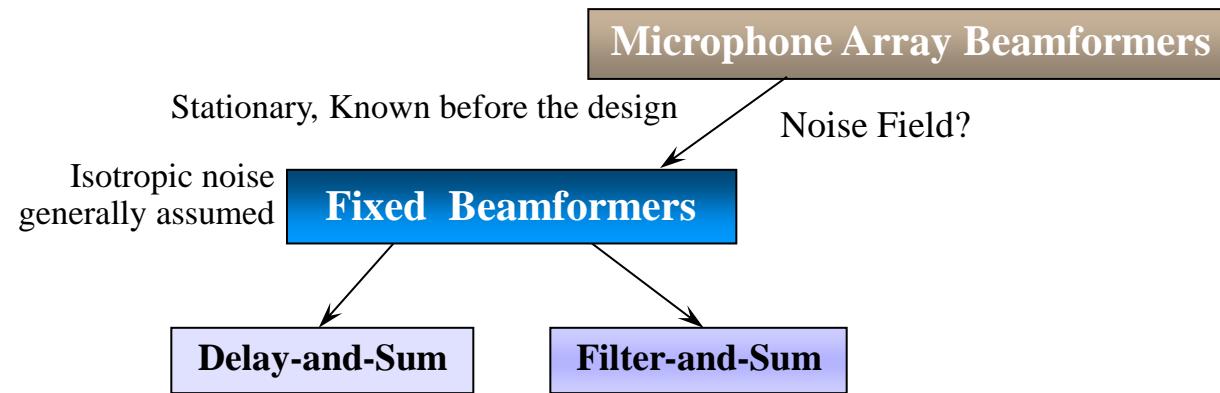
- Constant Beamwidth Beamforming

Filter-and-Sum:



# Design and Implementation of Small Microphone Arrays

## • Beamforming



### Delay-and-Sum

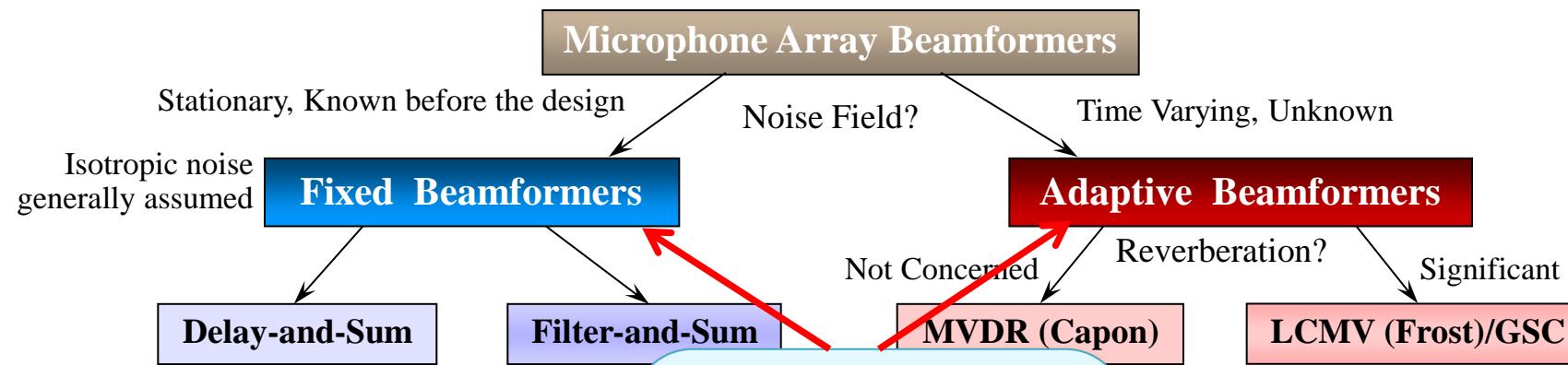
- Simple
- Non-uniform directional responses over a wide spectrum of frequencies

### Filter-and-Sum

- Uniform directional responses over a wide spectrum of frequencies: good for wideband signals, like speech



## • Beamforming



### Delay-and-Sum

- Simple
- Non-uniform directional responses over a wide spectrum of frequencies

### Filter-and-Sum

- Complicated
- Uniform directional responses over a narrow spectrum of frequencies
- For wideband signals, like speech

## Super Gain

(assuming noise

Correlation matrix is known)

- Reverberation causes the signal cancellation problem.
- Time-domain or frequency-domain

### LCMV (Frost)/GSC

- The impulse responses (IRs) from the source to the microphones have to be known or estimated.
- Errors in the IRs lead to the signal cancellation problem.



- Performance consistency over frequencies
- Performance consistency in different environments
- Working with other Processors
- Working with modern devices



- Small Microphone Arrays



Differential Microphone Arrays

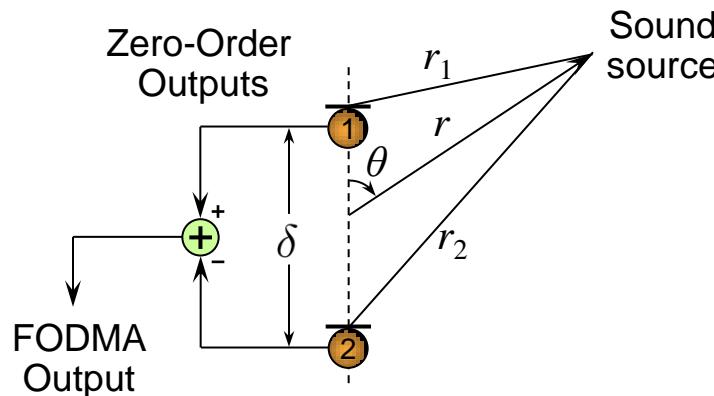
- Additive Array vs Differential Array
  - Both sensors and the array are responsive to the pressure field
  - Size is large (spacing from a couple of centimeters to a few decimeters)
  - Optimal gain is on the broadside
  - Signal extraction is achieved by steering the main lobe to the signal direction



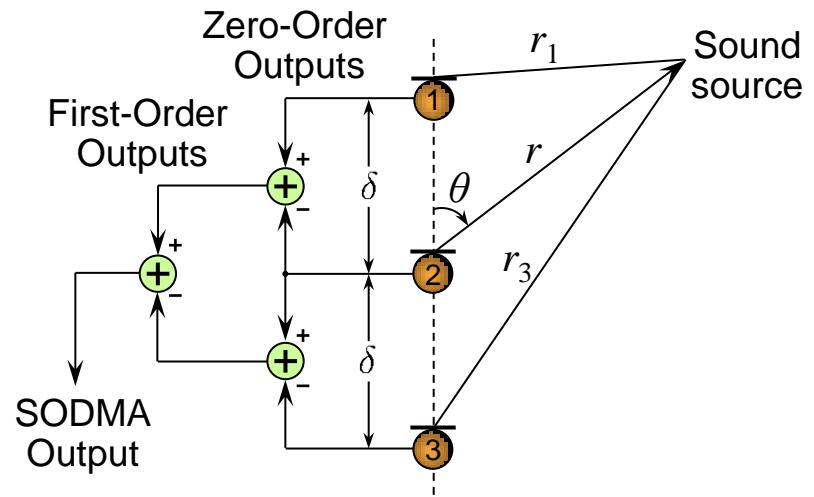
- Additive Array vs Differential Array
  - Sensors measure the pressure field; while the array is responsive to the spatial derivatives of the acoustic pressure field
  - Size is small: the sensor spacing,  $\delta$ , is much smaller than the acoustic wavelength, so that the true acoustic pressure differentials can be approximated by finite differences of the microphones' outputs.
  - Optimal gain is on the endfire direction



- Traditional Design of DMA



1<sup>st</sup>-Order DMA

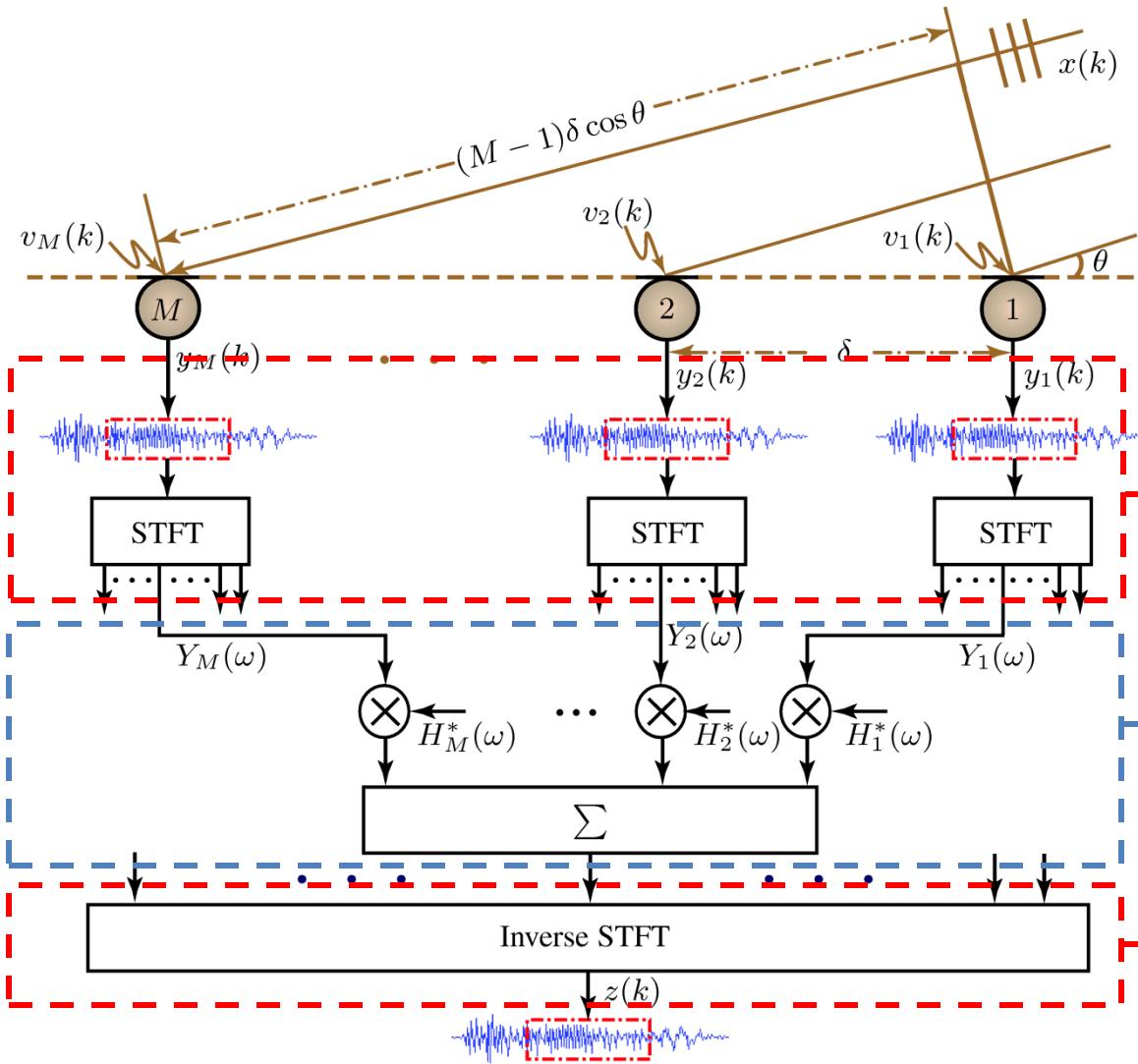


2<sup>nd</sup>-Order DMA

- An  $N$ th order DMA is formed by subtractively combining the outputs of two DMAs of order  $N-1$
- Not flexible to design different beampatterns
- Not flexible in dealing with white noise amplification

# Design and Implementation of Small Microphone Arrays

## • New Method of DMA Design



The new paradigm is for processing nonstationary broadband signals like speech

Analysis (STFT)

DMA Processing

Synthesis (ISTFT)



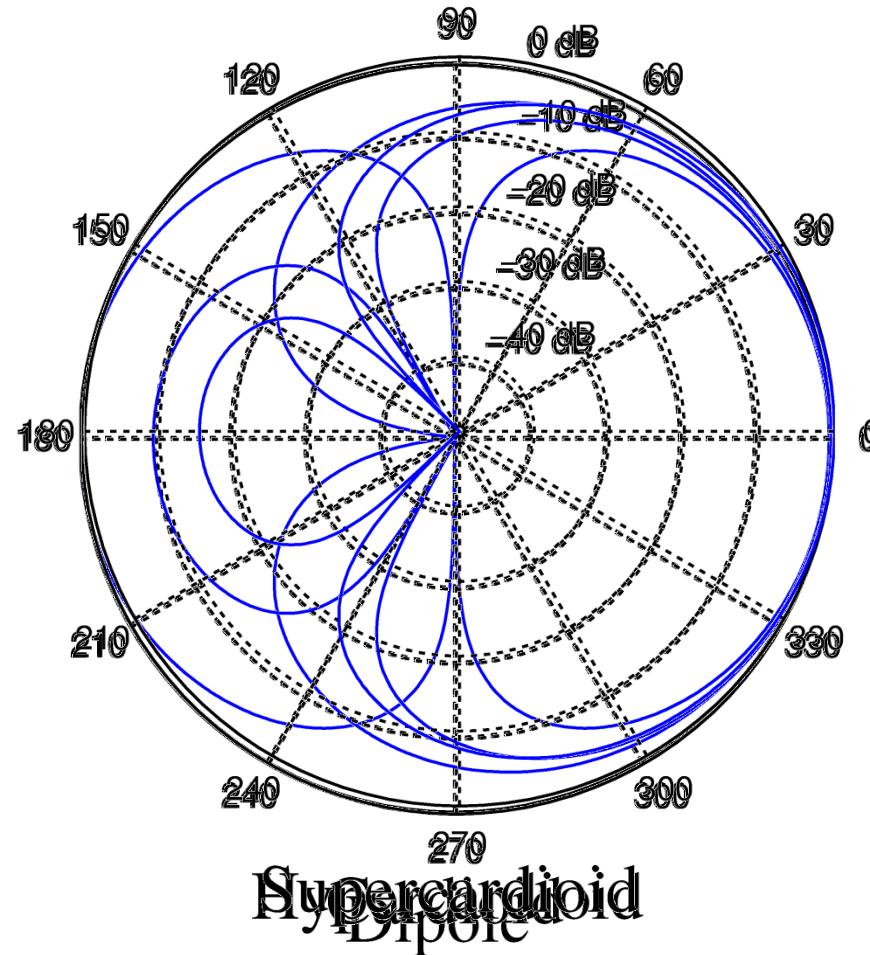
- What is a DMA
  - DMA Beampatterns:  $\mathcal{B}_N(\theta) = \sum_{n=0}^N a_{N,n} \cos^n(\theta)$



- What is a DMA

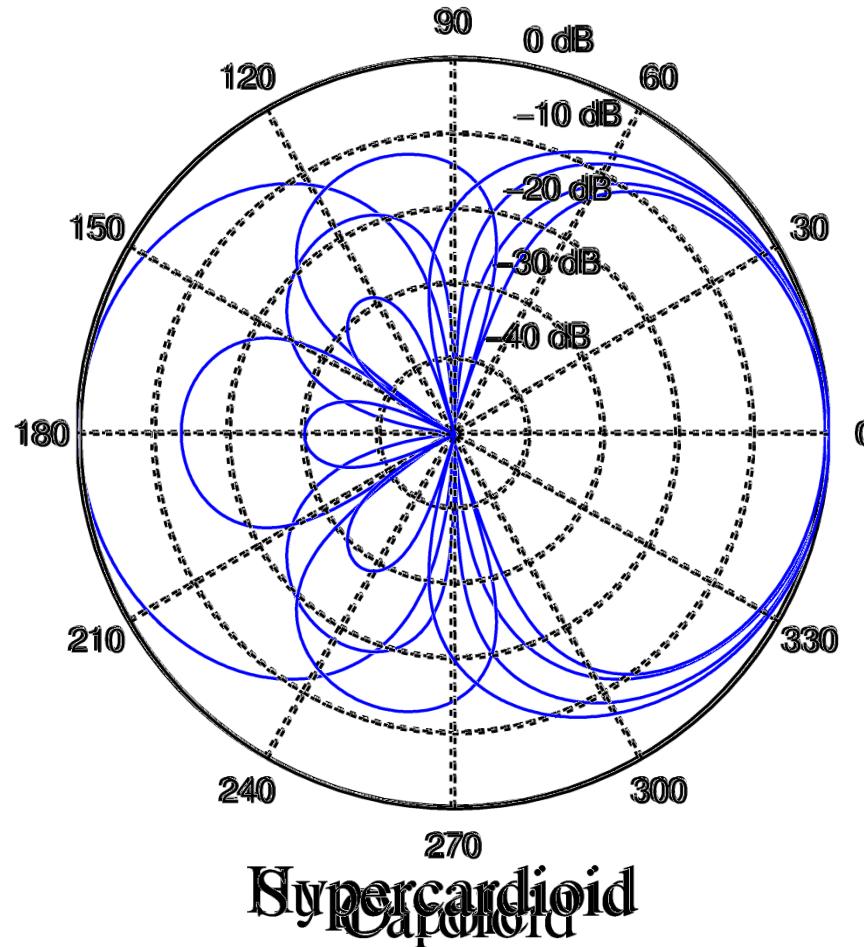
- 1<sup>st</sup>-order DMA:  $\mathcal{B}_1(\theta) = (1 - a_{1,1}) + a_{1,1} \cos \theta$

$$a_{1,1} = 2 - \sqrt{2}$$



- What is a DMA

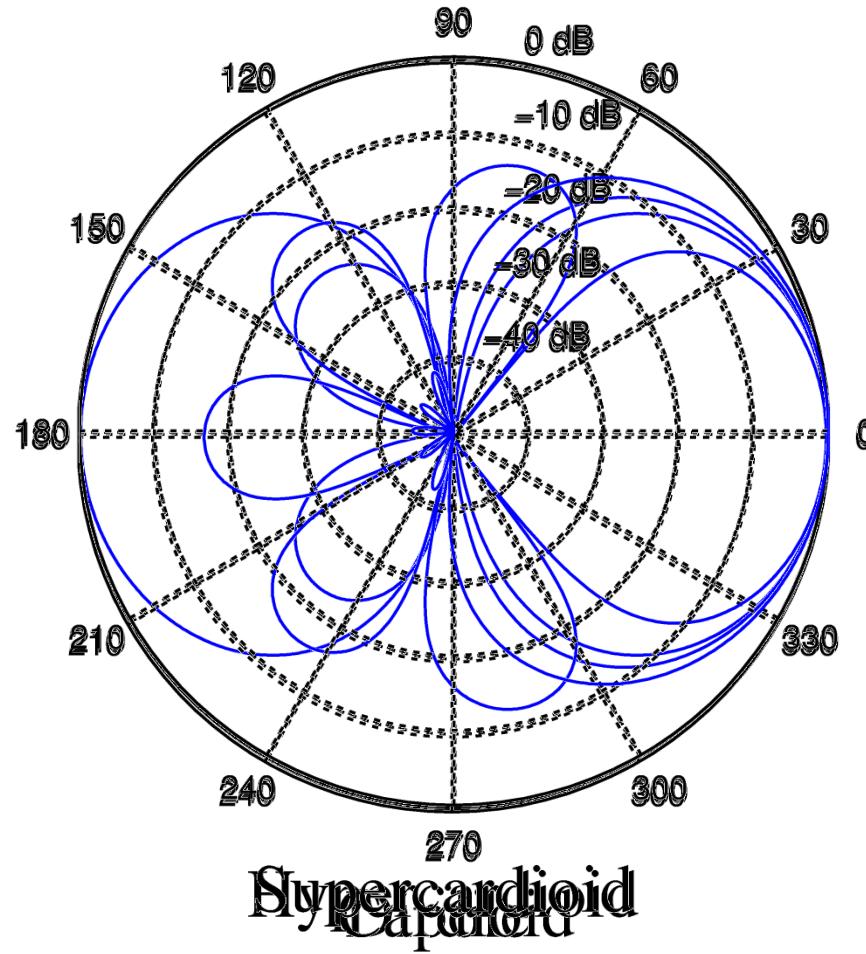
- 2<sup>nd</sup>-order DMA:  $\mathcal{B}_2(\theta) = (1 - a_{2,1} - a_{2,2}) + a_{2,1} \cos \theta + a_{2,2} \cos^2 \theta$



- What is a DMA

- 3<sup>rd</sup>-order DMA:  $\mathcal{B}_3(\theta) = (1 - a_{3,1} - a_{3,2} - a_{3,3}) + a_{3,1} \cos \theta + a_{3,2} \cos^2 \theta + a_{3,3} \cos^3 \theta$

$a_{3,1} = 0.217$   
 $a_{3,2} = 0.475$   
 $a_{3,3} = 0.286$



- DMA Beamforming
  - Ideal DMA Beampatterns:

$$\mathcal{B}_N(\theta) = \sum_{n=0}^N a_{N,n} \cos^n(\theta)$$

- Beampattern with  $\mathbf{h}(\omega)$

$$\mathcal{B}[\mathbf{h}(\omega), \theta] = \mathbf{d}^H(\omega, \cos \theta) \mathbf{h}(\omega)$$

Given  $M$ ,  $N$ , and the array geometry, finding coefficients in  $\mathbf{h}(\omega)$  so that

$$\mathcal{B}[\mathbf{h}(\omega), \theta] \rightarrow \mathcal{B}(\theta)$$



- DMA Beamforming for a 2<sup>nd</sup>-Order Cardioid

Beampattern:  $\mathcal{B} [\mathbf{h} (\omega), \theta] = \mathbf{d}^H (\omega, \cos \theta) \mathbf{h} (\omega)$

$$\mathbf{h} (\omega) = [ H_1 (\omega) \quad H_2 (\omega) \quad H_3 (\omega) ]^T$$

$$\mathbf{d} (\omega, \cos \theta) = [ 1 \quad e^{-j\omega\delta \cos \theta / c} \quad e^{-j2\omega\delta \cos \theta / c} ]^T$$



- DMA Beamforming for a 2<sup>nd</sup>-Order Cardioid

$$\mathbf{d}^H(\omega, \cos 0^\circ) \mathbf{h}(\omega) = 1$$

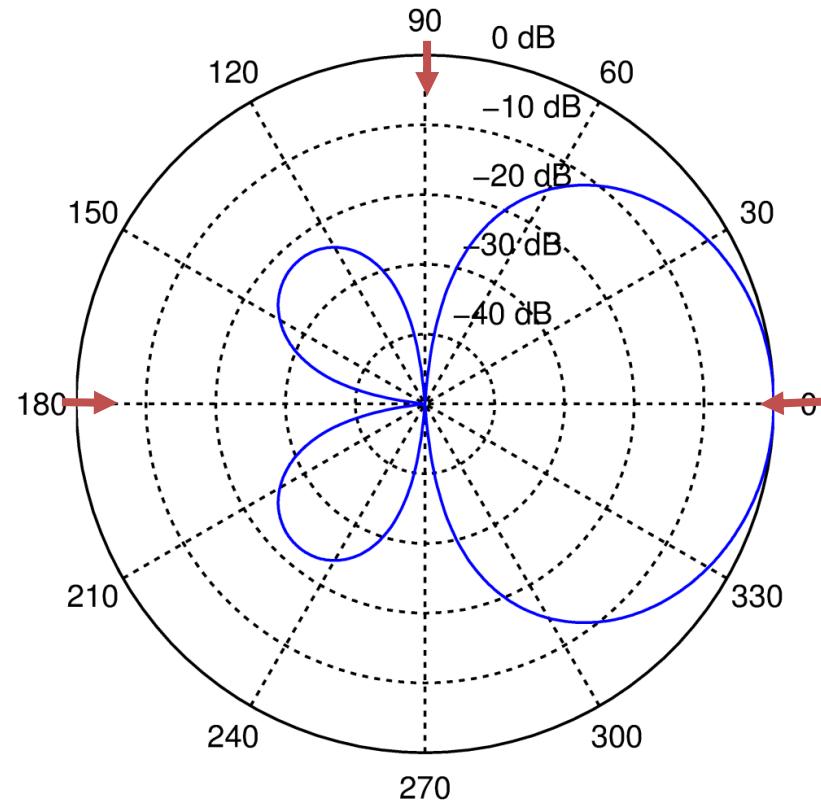
$$\mathbf{d}^H(\omega, \cos 90^\circ) \mathbf{h}(\omega) = 0$$

$$\mathbf{d}^H(\omega, \cos 180^\circ) \mathbf{h}(\omega) = 0$$

$\mathbf{D}(\omega, \boldsymbol{\alpha}) \mathbf{h}(\omega) = \boldsymbol{\beta}$

$$\mathbf{D}(\omega, \boldsymbol{\alpha}) = \begin{bmatrix} \mathbf{d}^H(\omega, \cos 0^\circ) \\ \mathbf{d}^H(\omega, \cos 90^\circ) \\ \mathbf{d}^H(\omega, \cos 180^\circ) \end{bmatrix}$$

$$\boldsymbol{\alpha} = \begin{bmatrix} \cos 0^\circ \\ \cos 90^\circ \\ \cos 180^\circ \end{bmatrix} \quad \boldsymbol{\beta} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$



2<sup>nd</sup>-order Cardioid

- DMA Beamforming for a 2<sup>nd</sup>-Order Cardioid
  - Is the design beampattern the same as the ideal beampattern?
  - Is the method generalizable?



- DMA Beamforming with Distinct Nulls

- $N+1$  Constraints:

$$\alpha_{N,n} = \cos \theta_{N,n}$$

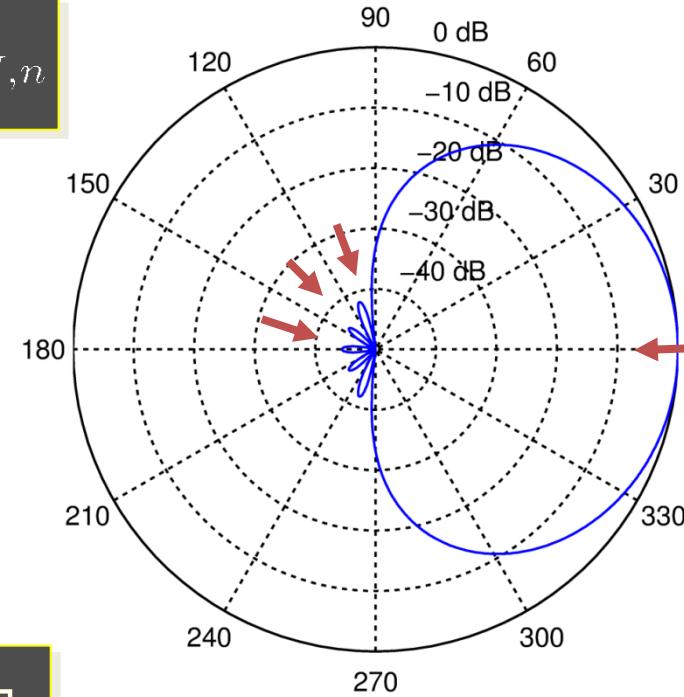
$$\boldsymbol{\alpha} = [1 \quad \alpha_{N,1} \quad \dots \quad \alpha_{N,N}]^T$$

$$\boldsymbol{\beta} = [1 \quad 0 \quad \dots \quad 0]^T$$

- Linear system with  $N+1$  equations

$$\mathbf{D}(\omega, \boldsymbol{\alpha}) \mathbf{h}(\omega) = \boldsymbol{\beta}$$

$$\mathbf{D}(\omega, \boldsymbol{\alpha}) = \begin{bmatrix} \mathbf{d}^H(\omega, 1) \\ \mathbf{d}^H(\omega, \alpha_{N,1}) \\ \vdots \\ \mathbf{d}^H(\omega, \alpha_{N,N}) \end{bmatrix}$$



- DMA Beamforming with Distinct Nulls: Examples

- DMA filter:  $\mathbf{h}(\omega) = \mathbf{D}^{-1}(\omega, \boldsymbol{\alpha}) \boldsymbol{\beta}$

- 1<sup>st</sup>-order DMA filter

$$\begin{bmatrix} 1 & e^{j\omega\tau_0} \\ 1 & e^{j\omega\tau_0}\alpha_{1,1} \end{bmatrix} \mathbf{h}(\omega) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\mathbf{h}(\omega) = \frac{1}{1 - e^{j\omega\tau_0}(1 - \alpha_{1,1})} \begin{bmatrix} 1 \\ -e^{-j\omega\tau_0}\alpha_{1,1} \end{bmatrix}$$


$$\tau_0 = \delta/c$$

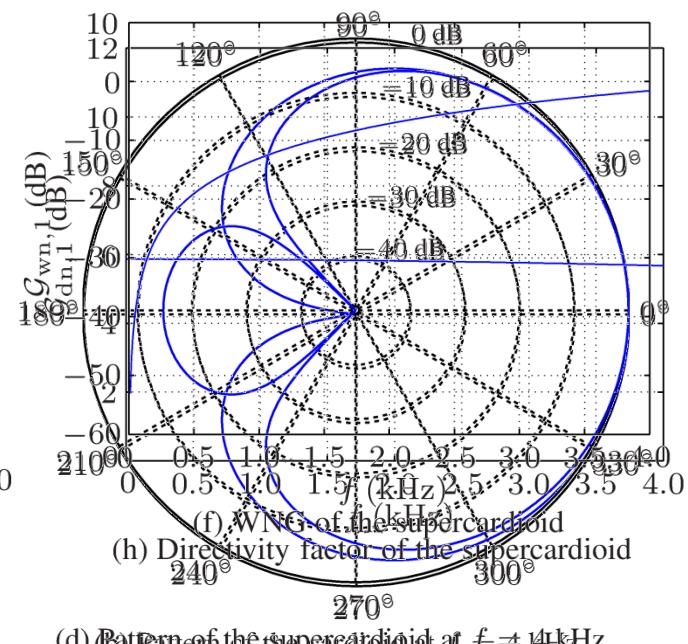
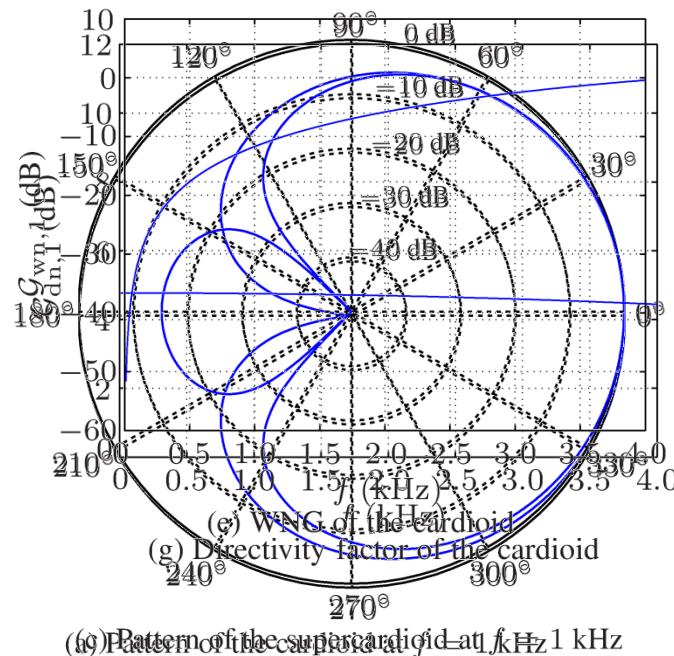


# Design and Implementation of Small Microphone Arrays

-1<sup>st</sup>-order DMA filter:

$$\mathbf{h}(\omega) = \frac{1}{1 - e^{j\omega\tau_0}(1 - \alpha_{1,1})} \begin{bmatrix} 1 \\ -e^{-j\omega\tau_0}\alpha_{1,1} \end{bmatrix}$$

WNG



# Design and Implementation of Small Microphone Arrays

## -2<sup>nd</sup>-order DMA filter

$$\begin{bmatrix} 1 & e^{j\omega\tau_0} & e^{j2\omega\tau_0} \\ 1 & e^{j\omega\tau_0\alpha_{2,1}} & e^{j2\omega\tau_0\alpha_{2,1}} \\ 1 & e^{j\omega\tau_0\alpha_{2,2}} & e^{j2\omega\tau_0\alpha_{2,2}} \end{bmatrix} \mathbf{h}(\omega) = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$



$$\mathbf{h}(\omega) = \frac{1}{\left[1 - e^{j\omega\tau_0}(1 - \alpha_{2,1})\right] \left[1 - e^{j\omega\tau_0}(1 - \alpha_{2,2})\right]} \begin{bmatrix} 1 \\ -e^{-j\omega\tau_0\alpha_{2,1}} - e^{-j\omega\tau_0\alpha_{2,2}} \\ e^{-j\omega\tau_0(\alpha_{2,1} + \alpha_{2,2})} \end{bmatrix}$$

Cardioid:  $\alpha_{2,1} = 0, \alpha_{2,2} = -1$  (i.e.,  $\theta_{2,1} = 90^\circ, \theta_{2,2} = 180^\circ$ )

Hypercardioid:  $\alpha_{2,1} = 0.31, \alpha_{2,2} = -0.81$  (i.e.,  $\theta_{2,1} \approx 72^\circ, \theta_{2,2} \approx 144^\circ$ )

Supercardioid:  $\alpha_{2,1} = -0.28, \alpha_{2,2} = -0.89$  (i.e.,  $\theta_{2,1} \approx 106^\circ, \theta_{2,2} \approx 153^\circ$ )

Quadrupole:  $\alpha_{2,1} = \sqrt{2}/2, \alpha_{2,2} = -\sqrt{2}/2$  (i.e.,  $\theta_{2,1} = 45^\circ, \theta_{2,2} = 135^\circ$ )

# Design and Implementation of Small Microphone Arrays

## -3<sup>rd</sup>-order DMA filter

$$\begin{bmatrix} \mathbf{d}^H(\omega, 1) \\ \mathbf{d}^H(\omega, \alpha_{3,1}) \\ \mathbf{d}^H(\omega, \alpha_{3,2}) \\ \mathbf{d}^H(\omega, \alpha_{3,3}) \end{bmatrix} \mathbf{h}(\omega) = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$



$$\mathbf{h}(\omega) = \frac{1}{\left[1 - e^{j\omega\tau_0}(1 - \alpha_{3,1})\right] \left[1 - e^{j\omega\tau_0}(1 - \alpha_{3,2})\right] \left[1 - e^{j\omega\tau_0}(1 - \alpha_{3,3})\right]} \begin{bmatrix} 1 \\ \gamma_1 \\ \gamma_2 \\ \gamma_3 \end{bmatrix}$$

$$\gamma_1 = -e^{-j\omega\tau_0}\alpha_{3,1} - e^{-j\omega\tau_0}\alpha_{3,2} - e^{-j\omega\tau_0}\alpha_{3,3}$$

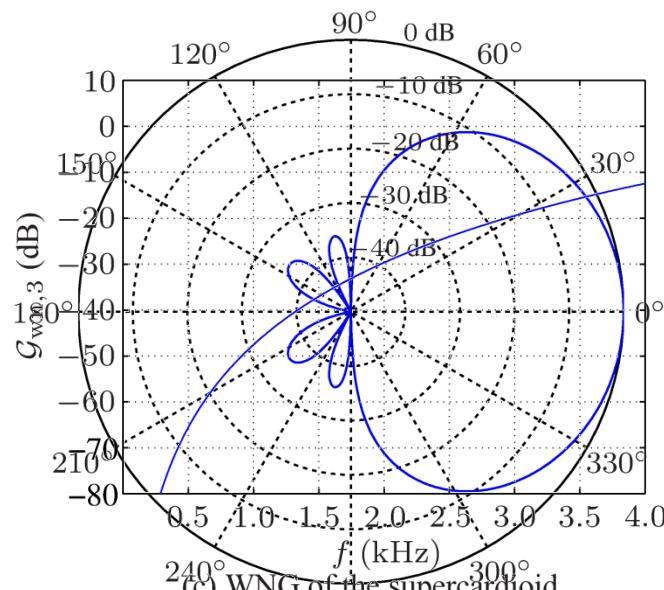
$$\gamma_2 = e^{-j\omega\tau_0}(\alpha_{3,1} + \alpha_{3,2}) + e^{-j\omega\tau_0}(\alpha_{3,2} + \alpha_{3,3}) + e^{-j\omega\tau_0}(\alpha_{3,1} + \alpha_{3,3})$$

$$\gamma_3 = -e^{-j\omega\tau_0}(\alpha_{3,1} + \alpha_{3,2} + \alpha_{3,3})$$

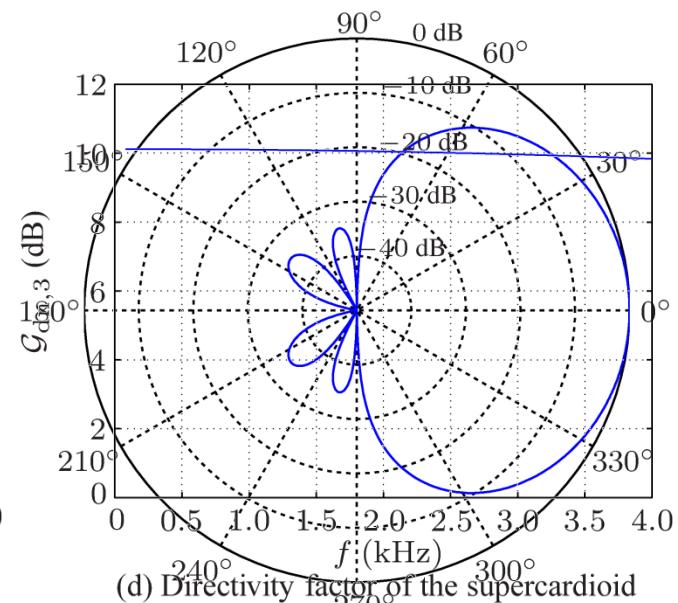


- DMA Beamforming with Distinct Nulls: Examples
  - 3<sup>rd</sup>-order DMA filter

$$\begin{aligned}\alpha_{3,1} &= 0 \\ \alpha_{3,2} &= -1/2 \\ \alpha_{3,3} &= -1 \\ \delta &= 1 \text{ cm}\end{aligned}$$

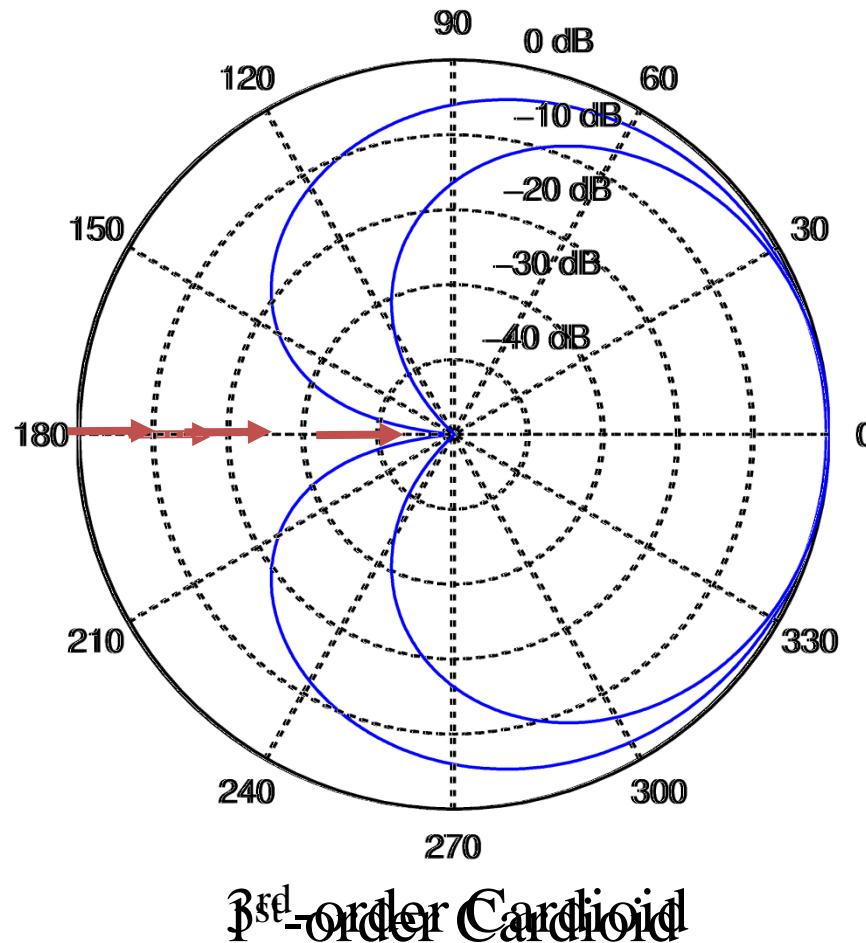


(a) Pattern of the supercardioid at  $f = 1 \text{ kHz}$



(b) Pattern of the supercardioid at  $f = 4 \text{ kHz}$

- DMA Beamforming with Nulls of Multiplicity More Than One



- DMA Beamforming with Nulls of Multiplicity More Than One

An  $N$ th-order DMA pattern that has  $N$  nulls but one of multiplicity  $P$  ( $1 \leq P \leq N$ ) at  $\alpha_{N,n} = \cos \theta_{N,n}$

$$\mathcal{B}_N(\theta) = \mathcal{B}_{N-P}(\theta) \times (\cos \theta - \alpha_{N,n})^P$$



$$\left. \frac{\partial^p \mathcal{B}_N(\theta)}{\partial \alpha^p} \right|_{\alpha=\alpha_{N,n}} = 0, \quad p = 1, 2, \dots, P-1$$



- DMA Beamforming with Nulls of Multiplicity More Than One

– Beampattern with  $\mathbf{h}(\omega)$

$$\begin{aligned}\frac{\partial^p \mathcal{B} [\mathbf{h}(\omega), \theta]}{\partial \alpha^p} &= \frac{\partial^p [\mathbf{d}^H(\omega, \alpha) \mathbf{h}(\omega)]}{\partial \alpha^p} \\ &= (\jmath \omega \tau_0)^p [\Sigma^p \mathbf{d}(\omega, \alpha)]^H \mathbf{h}(\omega)\end{aligned}$$



$$\Sigma = \text{diag}(0, 1, \dots, M - 1)$$

- DMA Beamforming with Nulls of Multiplicity More Than One

$$\mathbf{D}'(\omega, \boldsymbol{\alpha}) \mathbf{h}(\omega) = \boldsymbol{\beta}$$

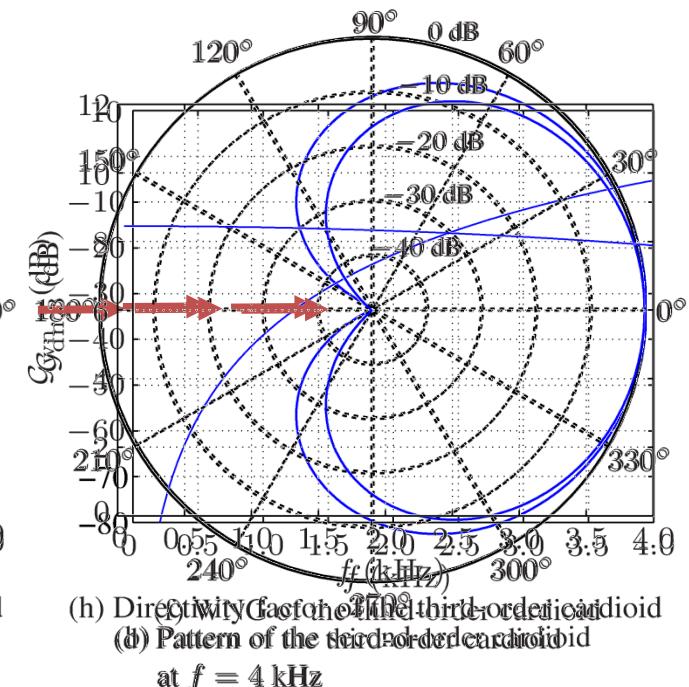
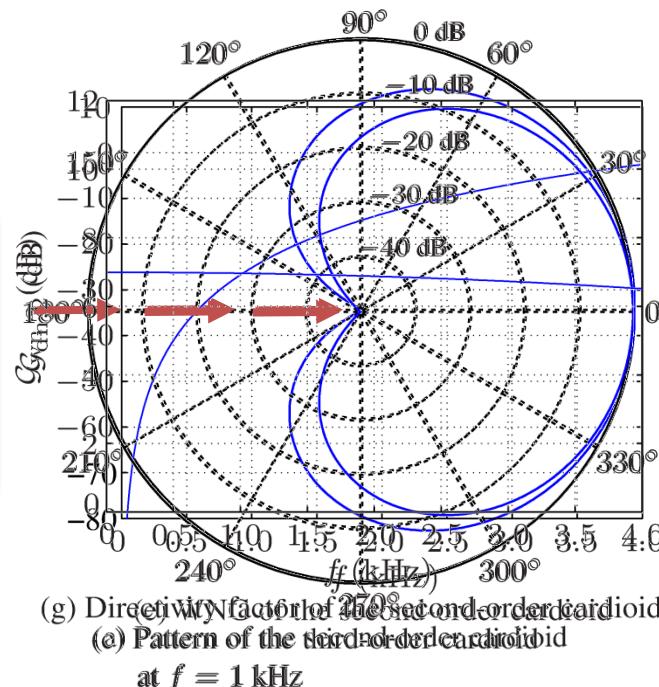
$$\mathbf{D}'(\omega, \boldsymbol{\alpha}) = \begin{bmatrix} \mathbf{d}^H(\omega, 1) & | & 1 \\ \mathbf{d}^H(\omega, \alpha_{N,1}) & | & 0 \\ \vdots & | & \vdots \\ \mathbf{d}^H(\omega, \alpha_{N,n}) & | & 0 \\ \mathbf{d}^H(\omega, \alpha_{N,n}) \Sigma & | & 0 \\ \vdots & | & \vdots \\ \mathbf{d}^H(\omega, \alpha_{N,n}) \Sigma^{P-1} & | & 0 \\ \hline \mathbf{d}^H(\omega, \alpha_{N,n+P}) & | & 0 \\ \vdots & | & \vdots \\ \mathbf{d}^H(\omega, \alpha_{N,N}) & | & 0 \end{bmatrix} \quad \boldsymbol{\beta} = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \\ 0 \\ \vdots \\ 0 \\ \hline 0 \\ \vdots \\ 0 \end{bmatrix}$$



# Design and Implementation of Small Microphone Arrays

- DMA Beamforming with Nulls of Multiplicity More Than One (Examples)

Directivity Factor



- DMA Beamforming with Ideal Pattern Information

–  $N+1$  Constraints:

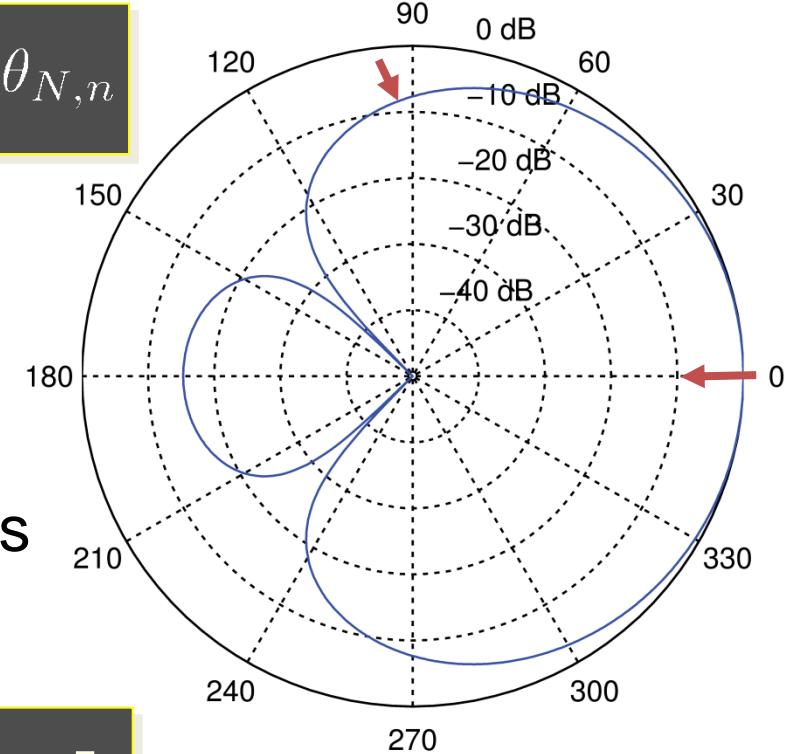
$$\boldsymbol{\alpha} = [1 \ \alpha_{N,1} \ \dots \ \alpha_{N,N}]^T$$

$$\boldsymbol{\beta} = [1 \ \beta_{N,1} \ \dots \ \beta_{N,N}]^T$$

– Linear system with  $N+1$  equations

$$\mathbf{D}(\omega, \boldsymbol{\alpha}) \mathbf{h}(\omega) = \boldsymbol{\beta}$$

$$\mathbf{D}(\omega, \boldsymbol{\alpha}) = \begin{bmatrix} \mathbf{d}^H(\omega, 1) \\ \mathbf{d}^H(\omega, \alpha_{N,1}) \\ \vdots \\ \mathbf{d}^H(\omega, \alpha_{N,N}) \end{bmatrix}$$



- DMA Beamforming with Ideal Pattern Information
  - Linear system with  $N+1$  equations
$$\mathbf{D}(\omega, \boldsymbol{\alpha}) \mathbf{h}(\omega) = \boldsymbol{\beta}$$
  - Fundamental constraints

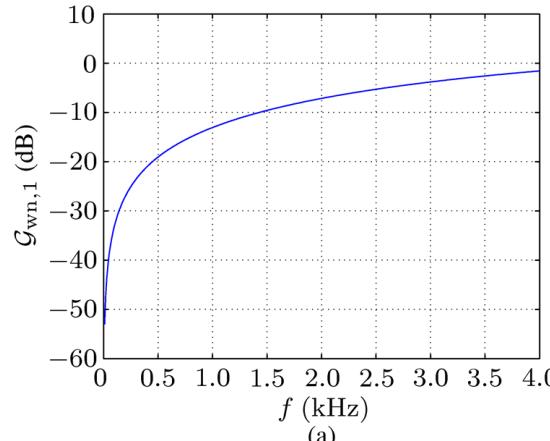
The  $N$  coefficients  $\alpha_{N,n}$  should be chosen in such a way that the  $\mathbf{D}(\omega, \boldsymbol{\alpha})$  matrix is well conditioned so that its inverse can be computed without any numerical problem.

The  $N$  pairs of coefficients  $(\alpha_{N,n}, \beta_{N,n})$  should take values from a desired “ideal” DMA beampattern. In general, They should correspond to the nulls of the desired “ideal” DMA beampattern; but they can take other values as well. For example, if the ”ideal” beampatten has multiple nulls at the same angle, we should choose a different angle. Then  $\beta_{N,n}$  will no longer be 0.

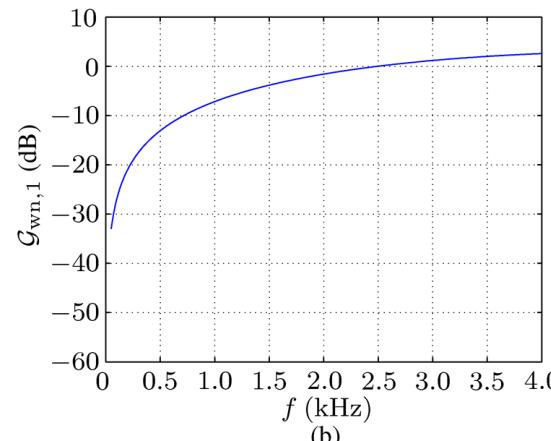


# Design and Implementation of Small Microphone Arrays

- Problem of White Noise Amplification

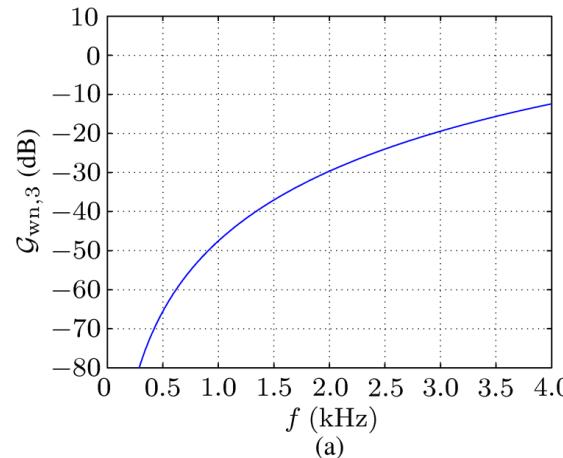


(a)

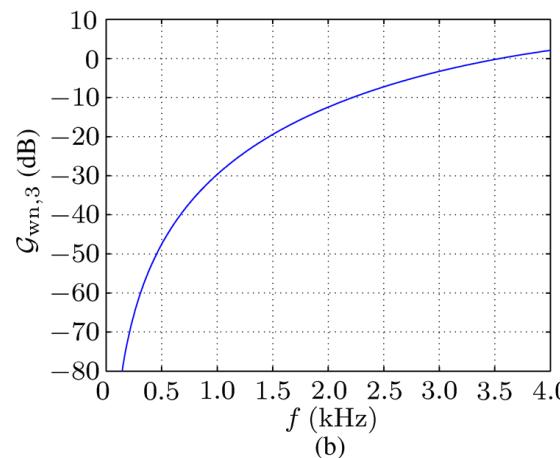


(b)

WNG of a 1st-order supercardioid at: (a)  $\delta = 1$  cm and (b)  $\delta = 2$  cm



(a)



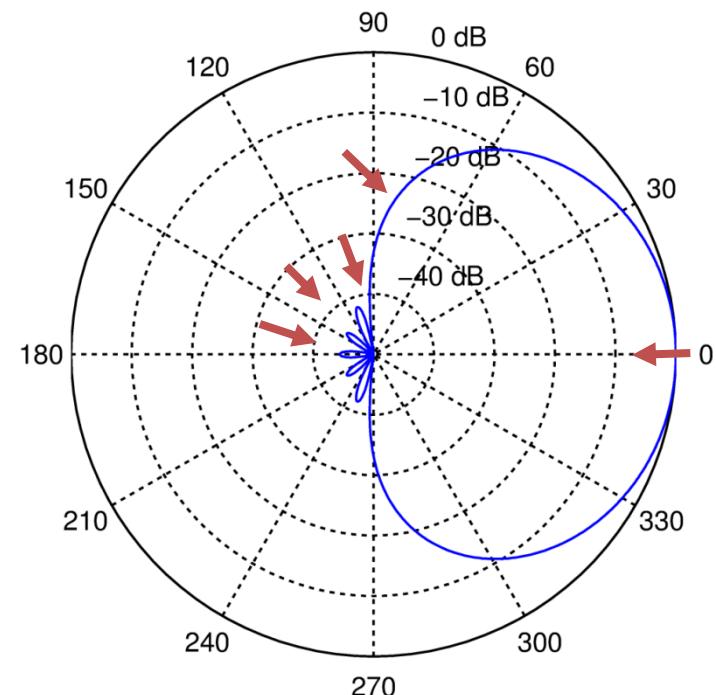
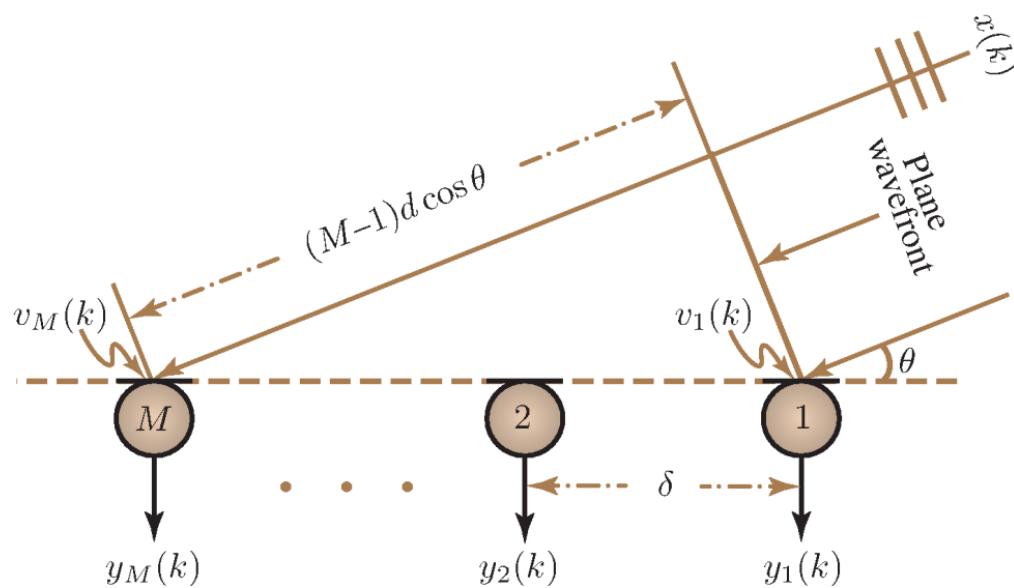
(b)

WNG of a 3rd-order supercardioid at: (a)  $\delta = 1$  cm and (b)  $\delta = 2$  cm



- WNG Improvement

Use more than  $N+1$  microphones to design a DMA with order of less than  $N$  (i.e.,  $M > N+1$ ).



- WNG Improvement
  - Linear system with  $M (\geq N+1)$  equations

$$\mathbf{D}(\omega, \alpha) \mathbf{h}(\omega) = \beta$$

$$\mathbf{D}(\omega, \alpha) = \begin{bmatrix} \mathbf{d}^H(\omega, 1) \\ \mathbf{d}^H(\omega, \alpha_{N,1}) \\ \vdots \\ \mathbf{d}^H(\omega, \alpha_{N,N}) \end{bmatrix}_{(N+1) \times M}$$

$$\mathbf{d}(\omega, \alpha_{N,n}) = \left[ 1 \quad e^{-j\omega\tau_0\alpha_{N,n}} \quad \dots \quad e^{-j(M-1)\omega\tau_0\alpha_{N,n}} \right]^T$$



- WNG Improvement
  - Maximization of the WNG

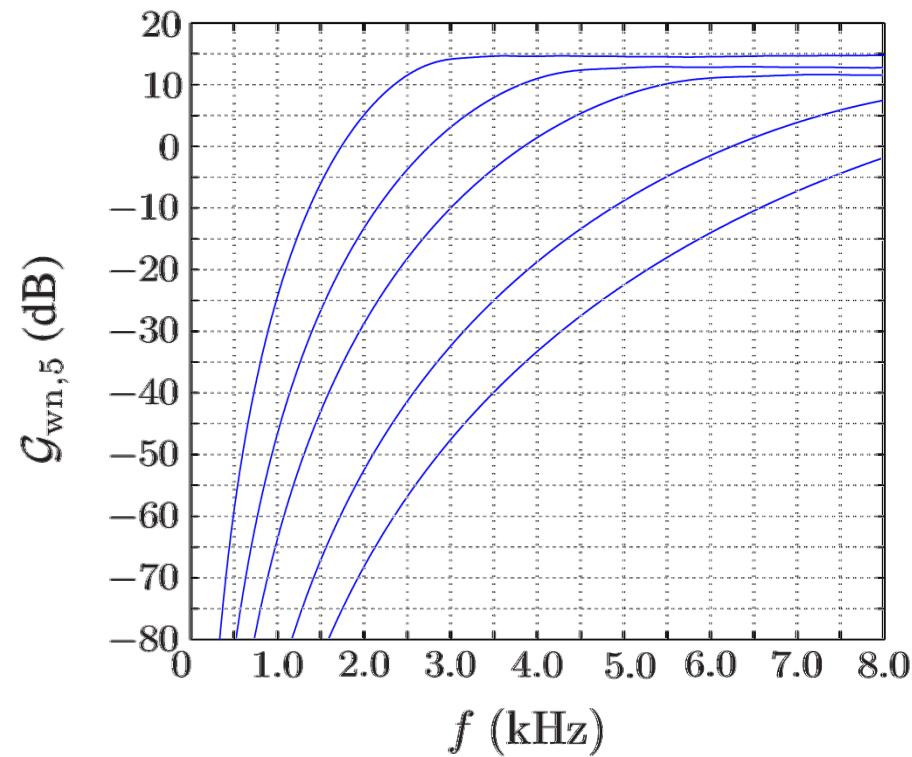
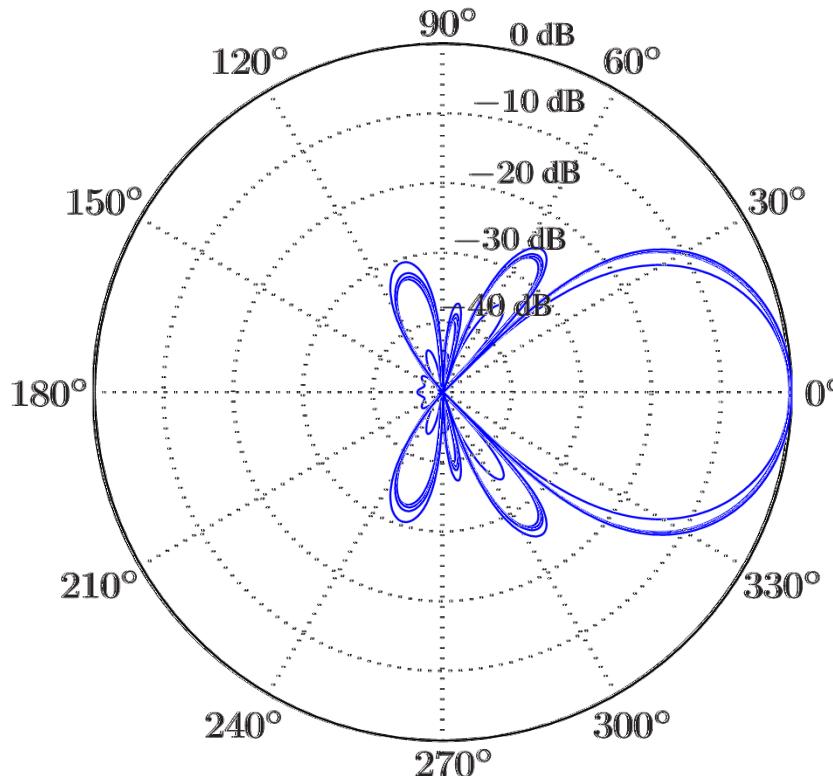
$$\max_{\mathbf{h}(\omega)} \frac{1}{\mathbf{h}^H(\omega) \mathbf{h}(\omega)} \quad \text{subject to} \quad \mathbf{D}(\omega, \boldsymbol{\alpha}) \mathbf{h}(\omega) = \boldsymbol{\beta}$$



$$\mathbf{h}_{\text{MaxWNG}}(\omega) = \mathbf{D}^H(\omega, \boldsymbol{\alpha}) [\mathbf{D}(\omega, \boldsymbol{\alpha}) \mathbf{D}^H(\omega, \boldsymbol{\alpha})]^{-1} \boldsymbol{\beta}$$

# Design and Implementation of Small Microphone Arrays

- WNG Improvement (Examples: a 5<sup>th</sup>-order DMA,  $\delta = 1$  cm)

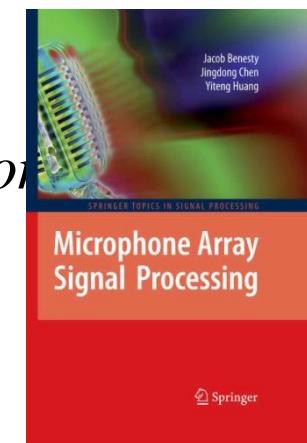
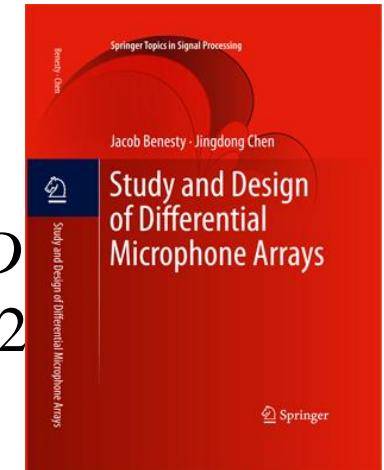


30 microphones

# Design and Implementation of Small Microphone Arrays

## • References

- J. Benesty and J. Chen, *Study and Design of Differential Microphone Arrays*. Berlin: Springer-Verlag, 2013
- J. Chen and J. Benesty, “A general approach to the design and implementation of linear differential microphone arrays,” in *Proc. APSIPA Annual Summit and Conference*, Oct. 2013
- J. Benesty, J. Chen, and Y. Huang, *Microphone Array Processing*. Berlin: Springer-Verlag, 2008



- Microphone arrays, particularly the ones with small aperture are more and more popularly used.
- Consistency in performance and flexibility in working with other processors are very important.
- A general approach to DMA Implementation is discussed, which converts the DMA design into a linear system solving problem. This method is very flexible and can design any desired beampattern.
- A robust method is discussed, that can use more microphones to design a given order DMA with less white noise amplification.

