

Precedence Effect

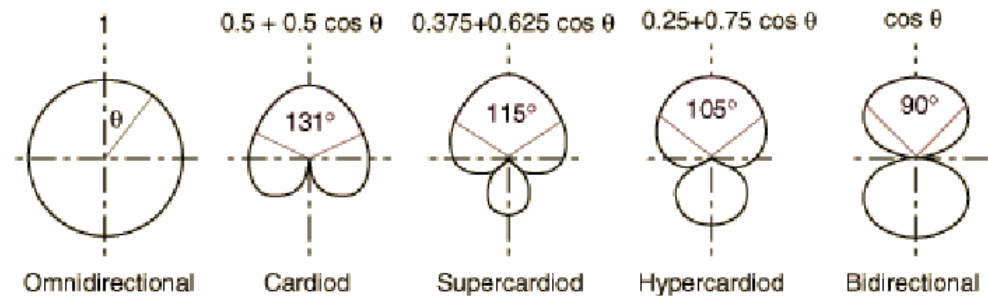
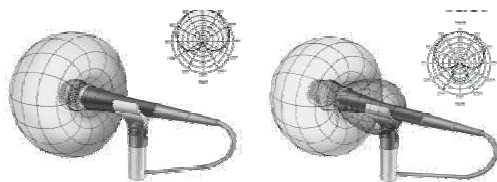
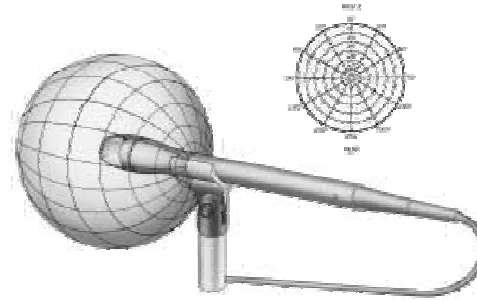
Beamforming

Demo of the Franssen effect

- Demonstrates precedence

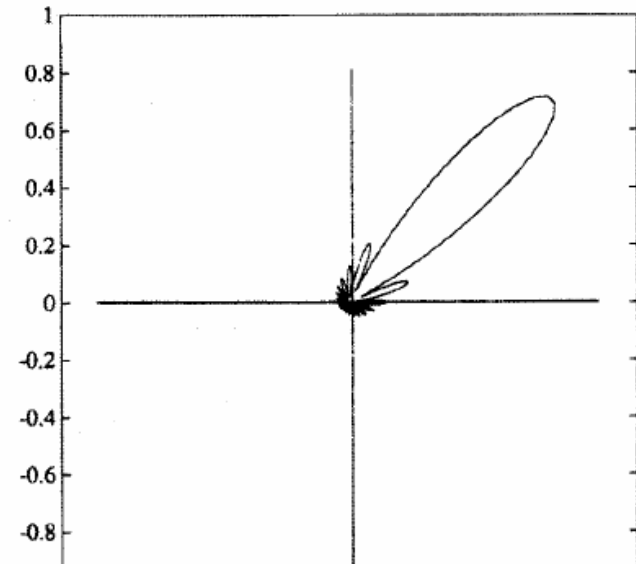
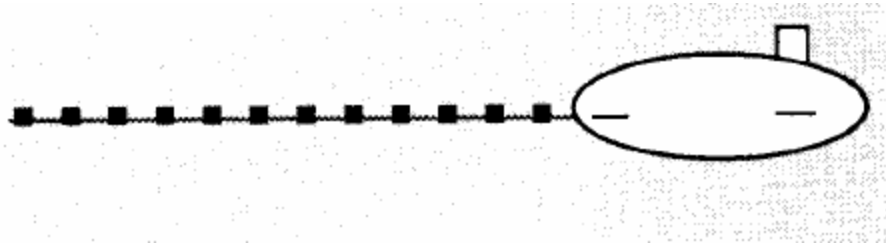
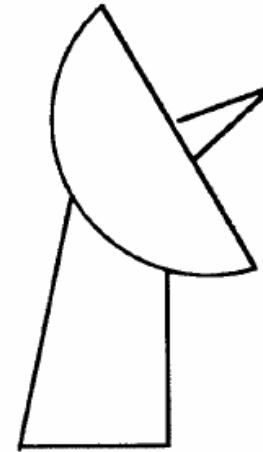
Introduction to 3D Audio (capture)

- Directivity of microphone.
 - Omni-directional
 - Advantages are that microphones capture all sound including that of interest
 - Directional
 - Capture sound from a preferred direction



Beamforming

- Given N microphones combine their signals in a way that some desired result occurs
- Word arises from the use of parabolic reflectors to form pencil “beams” for broadcast and reception
- Alternate word: “spatial filtering”
- Towed array and fixed array sonars



Delay and Sum Beamforming

- If the source location is known, delays relative to the microphone can be obtained
- Signal x at location \mathbf{s} arrives at microphone \mathbf{m}_i as

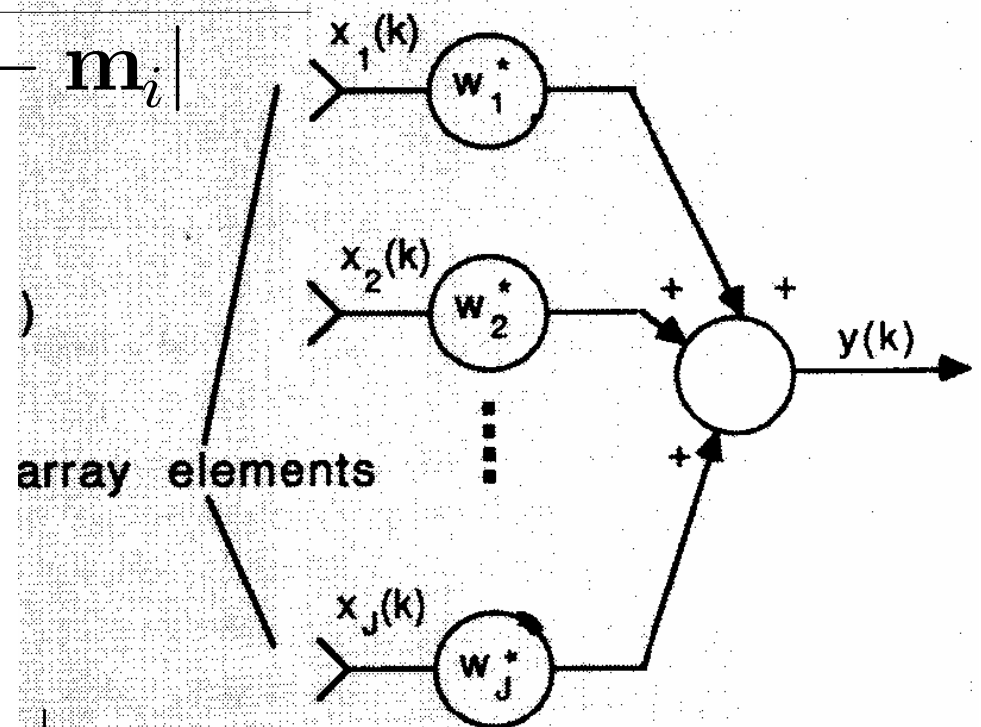
$$x \left(t - \frac{|\mathbf{s} - \mathbf{m}_i|}{c} \right)$$

- Signals at microphones can be appropriately delayed and weighted.

- Output signal is

$$y(k) = \frac{1}{N} \sum_{l=1}^N w_l^* x_l(k - \Delta_l)$$

$$\Delta_l = |\mathbf{s} - \mathbf{x}_l|/c \quad w_l = 1/|\mathbf{s} - \mathbf{x}_l|$$



Behavior of simple beamformer

- Usually source is assumed to be far away.
 - Weights are approximately the same in this case
- Signal from source direction adds in phase
 - So the signal is amplified N times
- Signals from other directions will add up with random phase and the power will decrease by a factor of $1/N$
- Directivity index is a measure of the gain of the array in the look direction (location of the delays) in decibels
 - For N microphones $10 \log_{10} (N)$
- Requires an ability to store the signal (at least for max $\{\Delta_1\}$)
- Jargon: “taps” number of samples in time that are stored

- Data independent beamforming:
 - Weights are fixed
- Data dependent (adaptive)
 - Weights change according to the data
- Simple example:
 - Fixed: Delay and sum looking at a particular point (direction)
 - Adaptive: Delay and sum looking at a particular moving source

More general beamforming

- Suppose we want to take advantage of the stored data
- Write the beamformer output as

$$y(k) = \sum_{l=1}^N \sum_{m=k-M}^k w_{lm}^* x_l(k-m)$$

- Can be written as $\mathbf{y} = \mathbf{w}^H \mathbf{x}$
- Take Fourier transform of the weights and the signal

Speech and Audio Processing

Microphone Array Processing

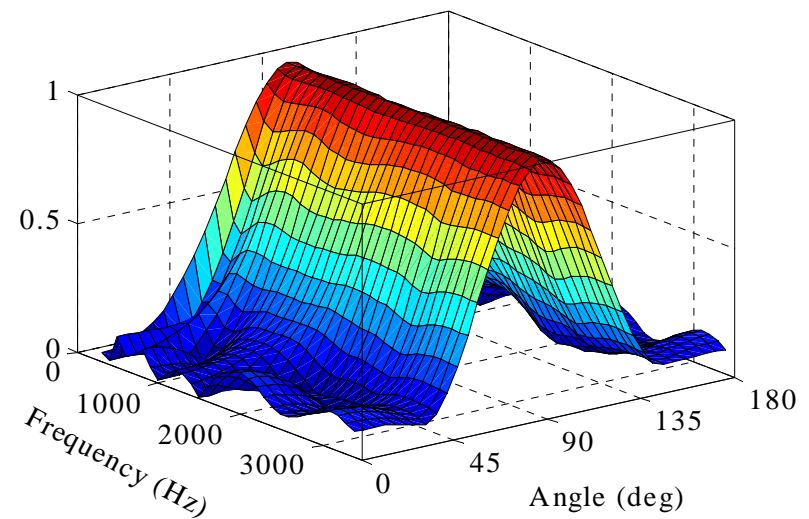
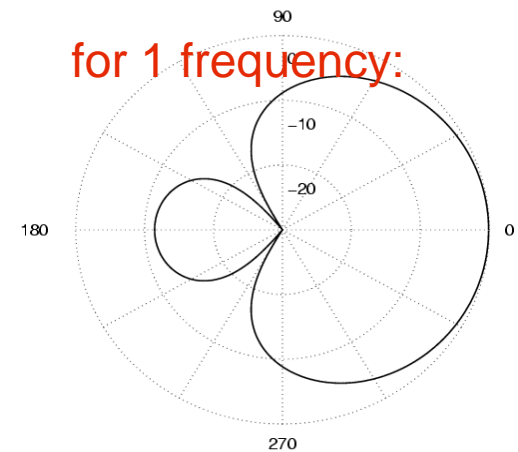
Slides adapted from those of
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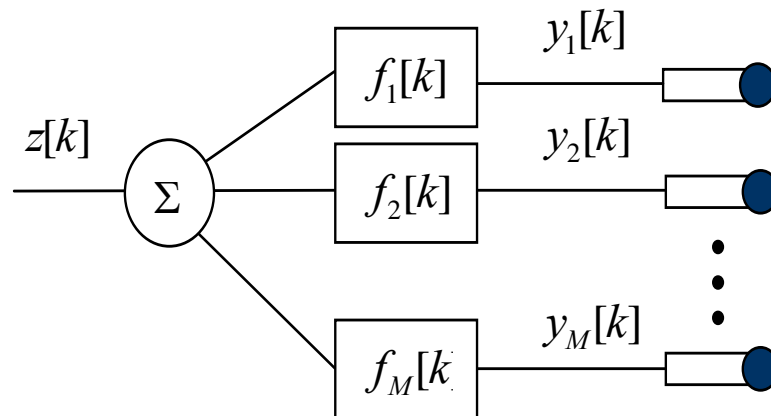
Introduction

- Each microphone is characterized by a `directivity pattern' which specifies the gain (& phase shift) that the microphone gives to a signal coming from a certain direction (`angle-of-arrival').
- Directivity pattern is a function of angle-of-arrival and frequency
- Directivity pattern is a (physical) microphone design issue.



Introduction

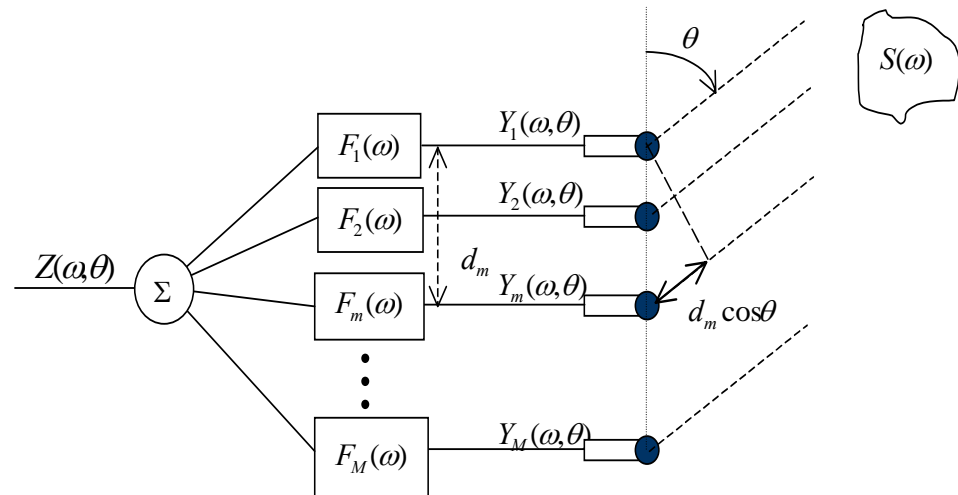
- **By weighting/filtering and summing signals from different microphones, a 'virtual' directivity pattern may be produced**



- **This is 'spatial filtering' and 'spatial filter design', based on given microphone characteristics (with correspondences to traditional (spectral) filter design)**
- **Applications: teleconferencing, hands-free telephony, hearing aids, voice-controlled systems, ...**

Introduction

- An important aspect is that different microphones in a microphone array are in different positions/locations, hence receive different signals
- Example : linear array, with uniform inter-microphone distances, under far-field (plane waveforms) conditions. Each microphone receives the same signal, but with different delays.



- Hence 'spatial filter design' based on microphone characteristics + microphone array configuration. Often simple assumptions are made, e.g. microphone gain = 1 for all frequencies and all angles.

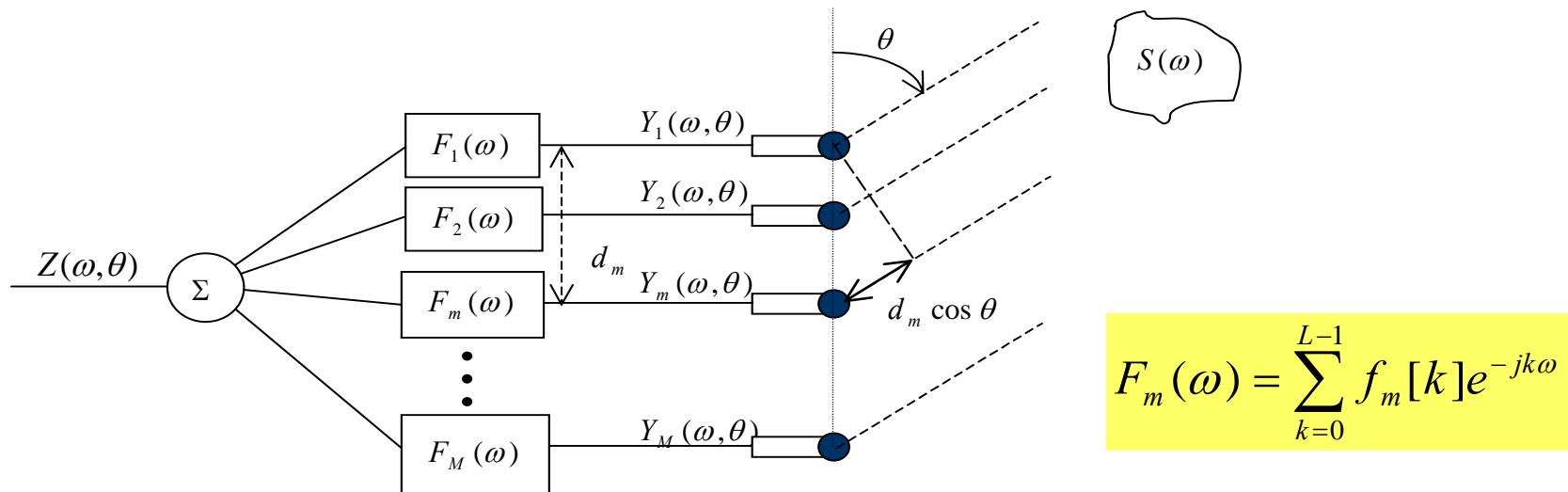
Introduction

- **Background/history: ideas borrowed from antenna array design/processing for RADAR & (later) wireless comms.**
- **Microphone array processing considerably more difficult than antenna array processing:**
 - narrowband radio signals versus broadband audio signals
 - far-field (plane wavefronts) versus near-field (spherical wavefronts)
 - pure-delay environment versus multi-path reverberant environment
- **Classification:**
 - fixed beamforming: data-independent, fixed filters $f_m[k]$
e.g. delay-and-sum, weighted-sum, filter-and-sum
 - adaptive beamforming: data-dependent, adaptive filters $f_m[k]$
e.g. LCMV-beamformer,

Beamforming basics

General form: filter-and-sum beamformer

- linear microphone array with M microphones and inter-micr. distance d_m
- Microphone gains are assumed to be equal to 1 for all freqs./angles
(otherwise, this characteristic is to be included in the steering vector, see next page)
- source $S(\omega)$ at angle θ (*far-field, no multipath*)
- filters $f_m[k]$ with filter length L



Terminology: 'Broadside' direction: $\theta = 90^\circ$ 'End-fire' direction: $\theta = 0^\circ$

Near-field beamforming

- **Far-field assumptions not valid for sources close to microphone array**
 - spherical wavefronts instead of planar waveforms
 - include attenuation of signals
 - 3 spherical coordinates θ, ϕ, r (=position \mathbf{q}) instead of 1 coordinate θ
- **Different steering vector:**

$$\mathbf{d}(\omega, \theta) \longrightarrow \mathbf{d}(\omega, \mathbf{q}) = \left[a_1 e^{-j\omega\tau_1(\mathbf{q})} \quad a_2 e^{-j\omega\tau_2(\mathbf{q})} \quad \dots \quad a_M e^{-j\omega\tau_M(\mathbf{q})} \right]^T$$

$$a_m = \frac{\|\mathbf{q} - \mathbf{p}_{ref}\|}{\|\mathbf{q} - \mathbf{p}_m\|}$$

$$\tau_m(\mathbf{q}) = \frac{\|\mathbf{q} - \mathbf{p}_{ref}\| - \|\mathbf{q} - \mathbf{p}_m\|}{c} f_s$$

with \mathbf{q} position of source

\mathbf{p}_{ref} position of reference microphone

\mathbf{p}_m position of m^{th} microphone

Beamforming basics

Data model:

- Microphone signals are delayed versions of $S(\omega)$

$$Y_m(\omega, \theta) = e^{-j\omega\tau_m(\theta)} S(\omega)$$

$$y_m[k] = s[k - \tau_m(\theta)] \quad \tau_m(\theta) = \frac{d_m \cos \theta}{c} f_s$$

- Stack all microphone signals in a vector

$$\mathbf{Y}(\omega, \theta) = \mathbf{d}(\omega, \theta) S(\omega) \quad \mathbf{d}(\omega, \theta) = \left[1 \quad e^{-j\omega\tau_2(\theta)} \quad \dots \quad e^{-j\omega\tau_M(\theta)} \right]^T$$

\mathbf{d} is 'steering vector'

- Output signal $Z(\omega, \theta)$ is

$$Z(\omega, \theta) = \sum_{m=1}^M F_m^*(\omega) Y_m(\omega, \theta) = \mathbf{F}^H(\omega) \cdot \mathbf{Y}(\omega, \theta)$$

Beamforming basics

Data model:

- Microphone signals are corrupted by additive noise

$$y_m[k] = s[k - \tau_m(\theta)] + n_m[k]$$

- Stack all noise signals in a vector

$$\mathbf{N}(\omega) = [N_1(\omega) \quad N_2(\omega) \quad \dots \quad N_M(\omega)]^T$$

- Define noise correlation matrix as

$$\Phi_{NN}(\omega) = E\{\mathbf{N}(\omega) \cdot \mathbf{N}(\omega)^H\}$$

- We assume noise field is homogeneous, i.e. diagonal elements of

$$\Phi_{NN}(\omega)$$

$$\Phi_{ii}(\omega) = \Phi_{noise}(\omega) \quad , \quad \forall i$$

- Then noise coherence matrix is

$$\Gamma_{NN}(\omega) = \frac{1}{\phi_{noise}(\omega)} \cdot \Phi_{NN}(\omega)$$

Beamforming basics

Definitions:

- **Spatial directivity pattern**: 'transfer function' for source at angle θ

$$H(\omega, \theta) = \frac{Z(\omega, \theta)}{S(\omega)} = \sum_{m=1}^M F_m^*(\omega) e^{-j\omega\tau_m(\theta)} = \mathbf{F}^H(\omega) \cdot \mathbf{d}(\omega, \theta)$$

- **Steering direction** θ_{max} = angle θ with maximum amplification (for 1 freq.)
- **Beamwidth** = region around θ_{max} with amplification $> -3\text{dB}$ (for 1 freq.)
- **Array Gain** = improvement in SNR

$$G(\omega, \theta) = \frac{SNR_{Output}}{SNR_{Input}} = \frac{|\mathbf{F}^H(\omega) \cdot \mathbf{d}(\omega, \theta)|^2}{\mathbf{F}^H(\omega) \cdot \mathbf{\Gamma}_{NN}(\omega) \cdot \mathbf{F}(\omega)}$$

Beamforming basics

Definitions:

- **Array Gain** = improvement in SNR

$$G(\omega, \theta) = \frac{SNR_{Output}}{SNR_{Input}} = \frac{|\mathbf{F}^H(\omega) \cdot \mathbf{d}(\omega, \theta)|^2}{\mathbf{F}^H(\omega) \cdot \mathbf{\Gamma}_{NN}(\omega) \cdot \mathbf{F}(\omega)}$$

- **Directivity** = array gain for θ_{max} and **diffuse** noise (=coming from all directions)

$$DI(\omega) = \frac{|\mathbf{F}^H(\omega) \cdot \mathbf{d}(\omega, \theta_{max})|^2}{\mathbf{F}^H(\omega) \cdot \mathbf{\Gamma}_{NN}^{diffuse}(\omega) \cdot \mathbf{F}(\omega)}$$

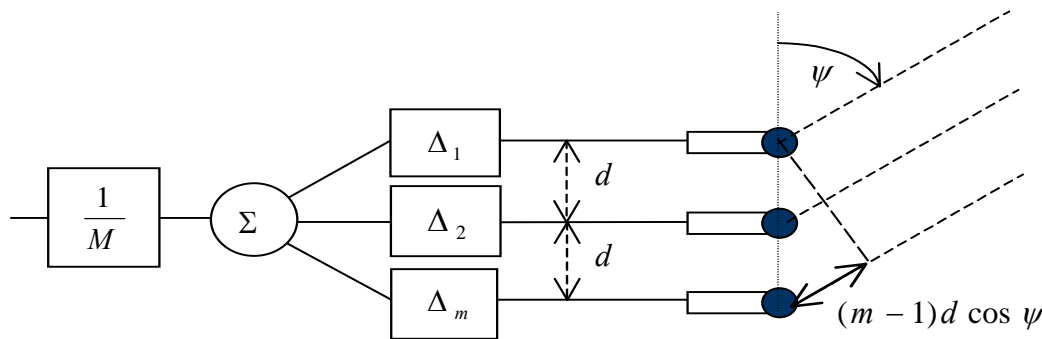
- **White Noise Gain** = array gain for θ_{max} and **spatially uncorrelated** noise ($\mathbf{\Gamma}_{NN} = \mathbf{I}$)
(e.g. sensor noise)

$$WNG(\omega) = \frac{|\mathbf{F}^H(\omega) \cdot \mathbf{d}(\omega, \theta_{max})|^2}{\mathbf{F}^H(\omega) \cdot \mathbf{F}(\omega)}$$

ps: often used as a measure for robustness

Delay-and-sum beamforming

- **Microphone signals are delayed and summed together**
Array can be virtually steered to angle ψ



$$z[k] = \frac{1}{M} \cdot \sum_{m=1}^M y_m[k + \Delta_m]$$

$$F_m(\omega) = \frac{e^{-j\omega\Delta_m}}{M}$$

$$\Delta_m = \frac{d_m \cos \psi}{c} f_s$$

- **Angular selectivity is obtained, based on constructive (for $\theta = \psi$) and destructive (for $\theta \neq \psi$) interference**

For $\theta = \psi$, this is referred to as a 'matched filter' :

$$F(\omega) = \frac{\mathbf{d}(\omega, \psi)}{M}$$

- For **uniform** linear array :

$$d_m = (m-1)d \quad \Delta_m = (m-1)\Delta$$

$$H(\omega, \theta = \psi) = 1$$

- **PS:** $H(\omega, \theta) = H(\omega, -\theta)$ plain!) (if microphone characteristics are ignored)

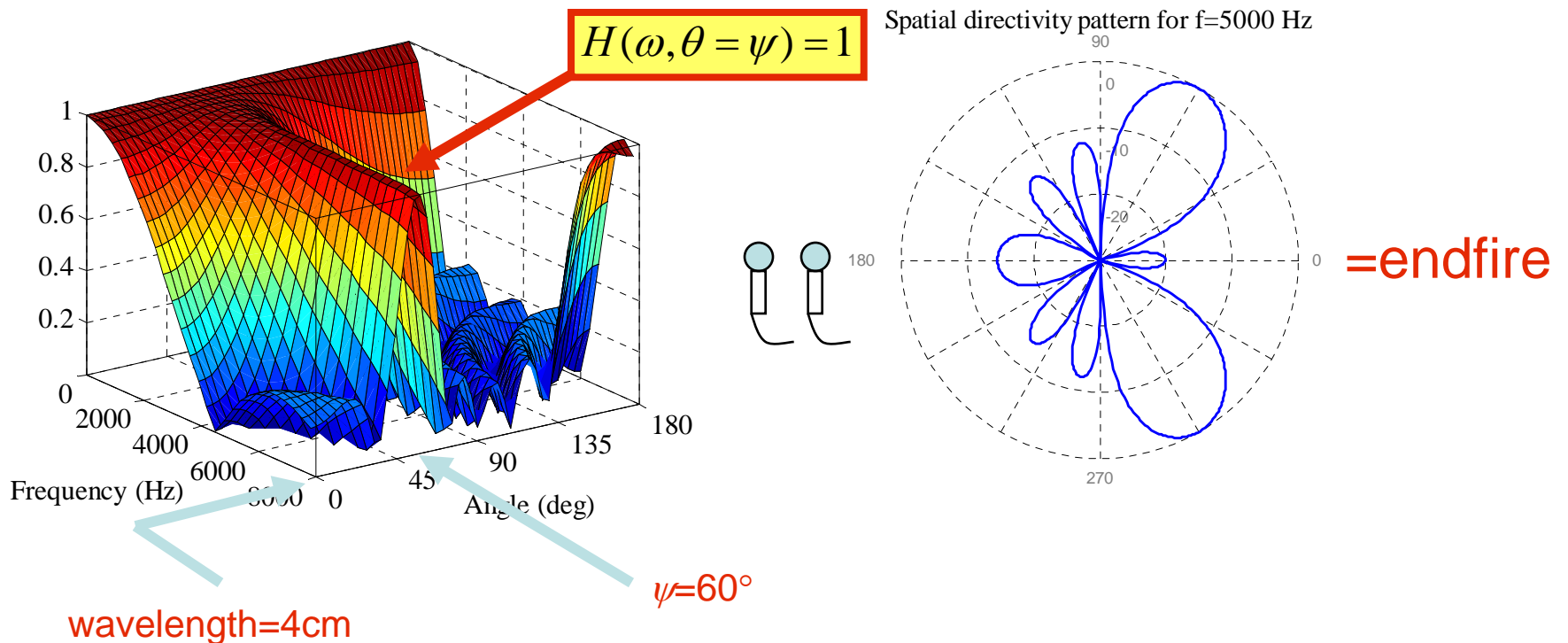
Delay-and-sum beamforming

- Spatial directivity pattern $H(\omega, \theta)$ for uniform DS-beamformer

$$\begin{aligned}
 H(\omega, \theta) &= \sum_{m=1}^M e^{-j(m-1)\omega \frac{d(\cos\theta - \cos\psi)}{c} f_s} \\
 &= \frac{e^{-jM\gamma/2} \sin(M\gamma/2)}{e^{-j\gamma/2} \sin(\gamma/2)}
 \end{aligned}$$

$M=5$ microphones
 $d=3$ cm inter-microphone distance
 $\psi=60^\circ$ steering angle
 $f_s=16$ kHz sampling frequency

- $H(\omega, \theta)$ has sinc-like shape and is frequency-dependent



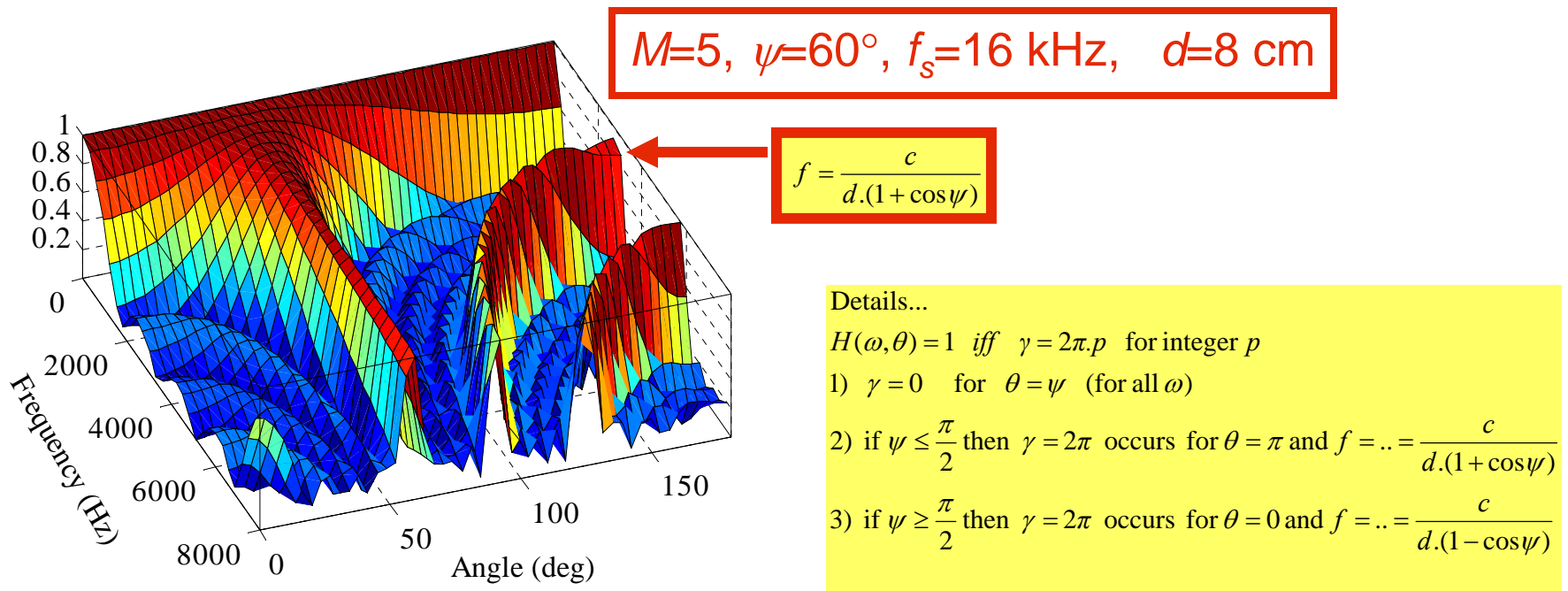
Delay-and-sum beamforming

- For $f \geq \frac{c}{d(1+|\cos\psi|)}$ an ambiguity, called *spatial aliasing*, occurs.

This is analogous to time-domain aliasing where now the spatial sampling ($=d$) is too large.

Aliasing does not occur (for any ψ) if

$$d \leq \frac{c}{f_s} = \frac{c}{2 \cdot f_{\max}} = \frac{\lambda_{\min}}{2}$$



Delay-and-sum beamforming

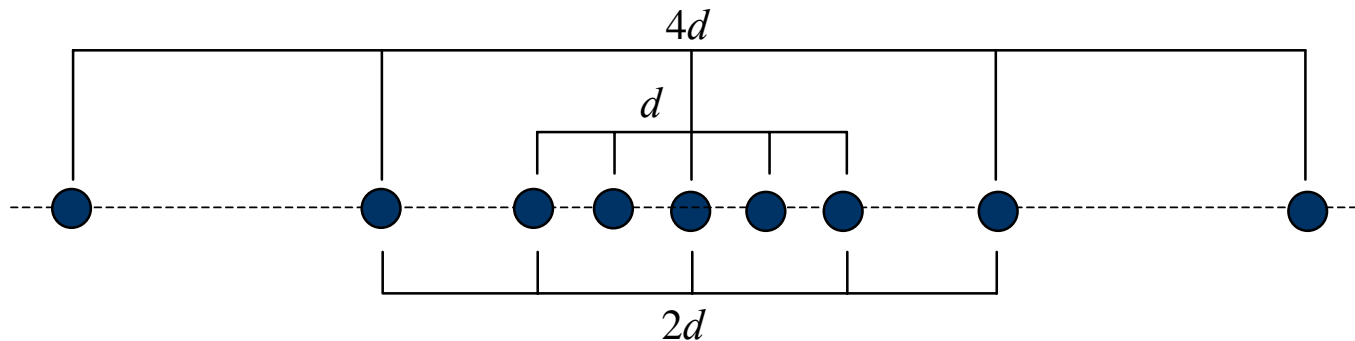
- **Beamwidth:** for a uniform delay-and-sum beamformer

$$BW \approx c \frac{\sqrt{96(1-\nu)}}{\omega d M} \sec \psi \quad \text{with e.g. } \nu = \sqrt{\frac{1}{2}} \quad (-3 \text{ dB})$$

hence large dependence on # microphones, distance (compare p14 & 15) and frequency (e.g. BW infinitely large at DC)

- **Array topologies:**

- Uniformly spaced arrays
- Nested (logarithmic) arrays (small d for high ω , large d for small ω)
- Planar / 3D-arrays

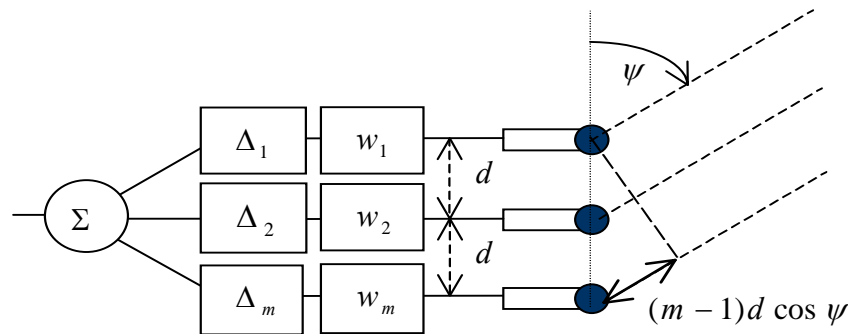


Weighted-sum beamforming

'delay-

and-weight/sum'

- **Sensor-dependent complex weight + delay (compare to p. 13)**

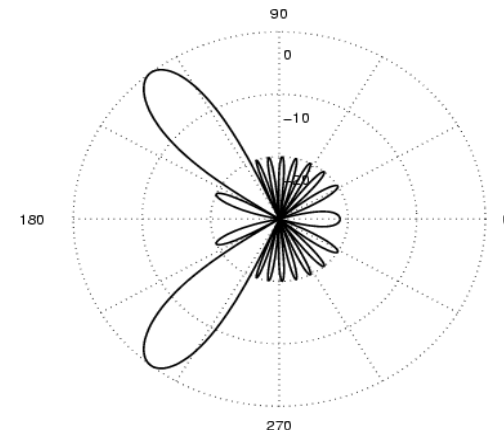


$$z[k] = \sum_{m=1}^M w_m \cdot y_m[k + \Delta_m]$$

$$H(\omega, \theta) = \sum_{m=1}^M w_m \cdot e^{-j(m-1)\omega \frac{d(\cos\theta - \cos\psi)}{c} f_s}$$

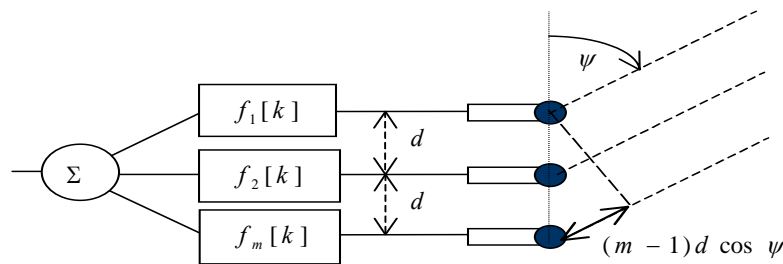
- **Weights added to allow for better beam shaping**
- **Design similar to traditional (spectral) filter design**

Ex: Dolph-Chebyshev design:
beampattern with uniform sidelobe level ('equiripple')



Filter-and-sum beamforming

- **Sensor-dependent filters implement frequency-dependent complex weights to obtain a desired response over the whole frequency/angle range of interest**



$$z[k] = \sum_{m=1}^M f_m[-k] \otimes y_m[k]$$

$$H(\omega, \theta) = \sum_{m=1}^M F_m^*(\omega) \cdot e^{-j(m-1)\omega \frac{d \cos \theta}{c} f_s}$$

- **Design strategies : desired beampattern is $P(\omega, \theta)$**

– Non-linear:

$$\min_{f_m[k], m=1 \dots M} \int_{\theta_1}^{\theta_2} \int_{\omega_1}^{\omega_2} (|H(\omega, \theta)| - |P(\omega, \theta)|)^2 d\omega d\theta$$

– Quadratic:

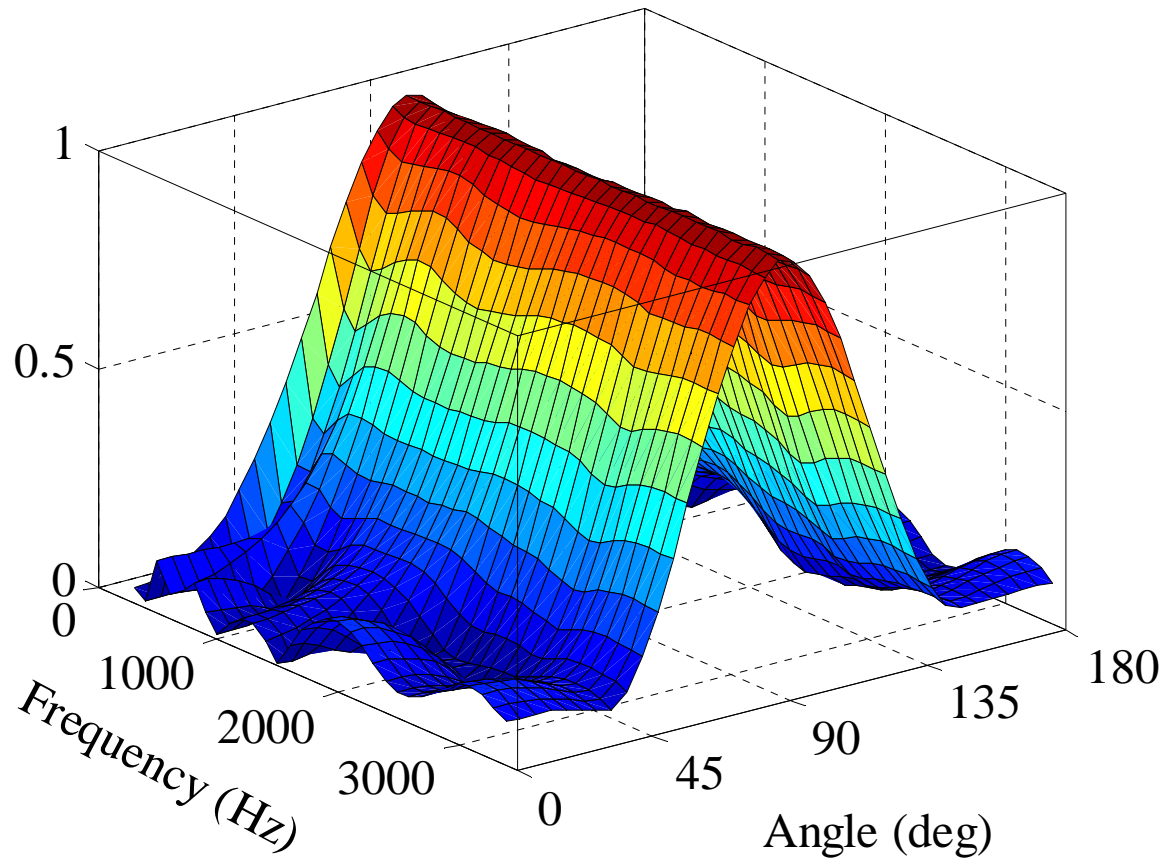
$$\min_{f_m[k], m=1 \dots M} \int_{\theta_1}^{\theta_2} \int_{\omega_1}^{\omega_2} |H(\omega, \theta) - P(\omega, \theta)|^2 d\omega d\theta$$

– Frequency sampling, i.e. design weights for sampling frequencies ω_i and then interpolate :

$$\min_{F_m(\omega_i), m=1 \dots M} \int_{\theta_1}^{\theta_2} |H(\omega_i, \theta) - P(\omega_i, \theta)|^2 d\theta$$

Filter-and-sum beamforming

- **Example-1: frequency-independent beamforming** (continued)



$M=8$
Logarithmic array
 $L=50$
 $\psi=90^\circ$
 $f_s=8$ kHz

Filter-and-sum beamforming

- **Example-2: 'superdirective' beamforming**

- Maximize directivity for known (diffuse) noise fields
- Maximum directivity $=M^2$ obtained for diffuse noise & endfire steering ($\theta=0^\circ$)

Design: find $\mathbf{F}(\omega)$ that maximizes

for given steering angle θ_{\max}

- Optimal solution is

$$DI(\omega) = \frac{|\mathbf{F}^H(\omega) \cdot \mathbf{d}(\omega, \theta_{\max})|^2}{\mathbf{F}^H(\omega) \cdot \Gamma_{NN}^{diffuse}(\omega) \cdot \mathbf{F}(\omega)}$$

$$\mathbf{F}(\omega) = \alpha \cdot \Gamma_{NN}^{-1}(\omega) \cdot \mathbf{d}(\omega, \theta_{\max})$$

- This is equivalent to minimization of noise output power, subject to unit response for steering angle (**)

$$\min_{\mathbf{F}(\omega)} \mathbf{F}^H(\omega) \cdot \Gamma_{NN}(\omega) \cdot \mathbf{F}(\omega), \text{ s.t. } \mathbf{F}^H(\omega) \cdot \mathbf{d}(\omega, \theta_{\max}) = 1$$

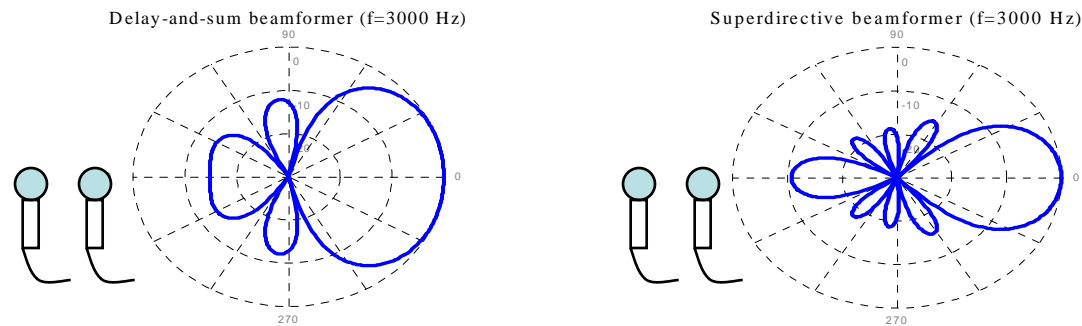
PS: Delay-and-sum beamformer similarly maximizes $W_{\text{WNG}} = 1$

$$\mathbf{F}(\omega) = \alpha \cdot \mathbf{d}(\omega, \theta_{\max})$$

Filter-and-sum beamforming

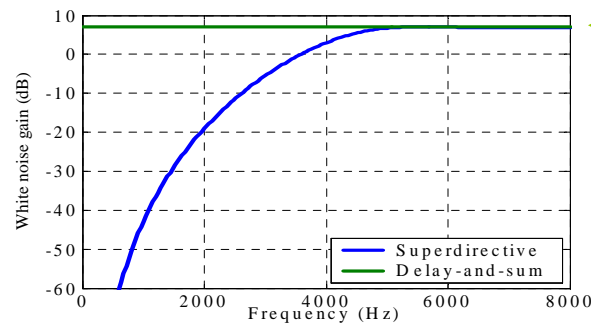
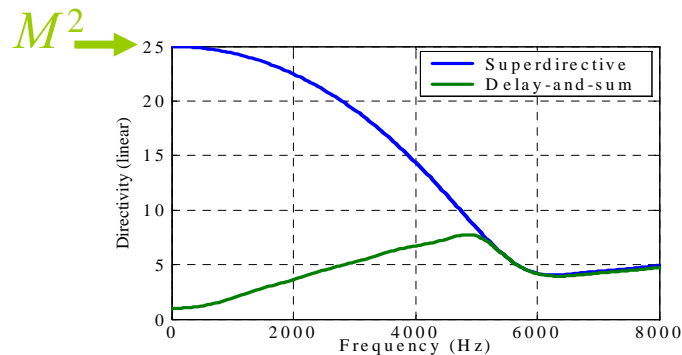
- **Example-2: 'superdirective' beamforming (continued)**

Directivity patterns for endfire steering:



$M=5$
 $d=3$ cm
 $\theta_{max}=0^\circ$
 $f_s=16$ kHz

Superdirective beamformer has highest DI, but very poor WNG hence problems with *robustness* (e.g. sensor noise) !



$\leftarrow 6.99=10.\text{Log}(5)$

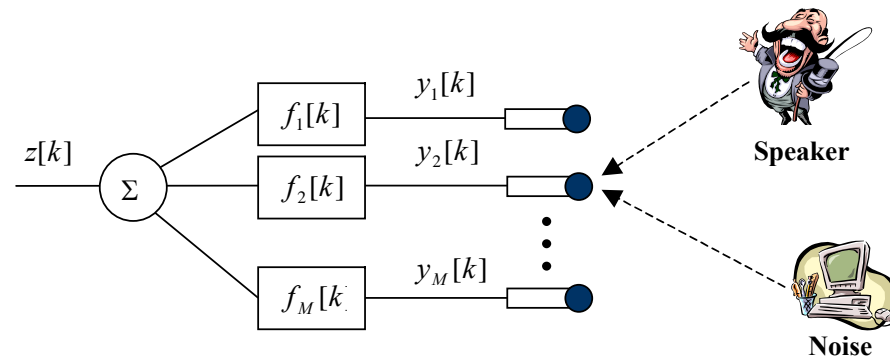
PS:
diffuse noise =
white noise for
high frequencies

LCMV-beamforming

- **Adaptive filter-and-sum structure:**

- Aim is to minimize noise output power, while maintaining a chosen frequency response in a given look direction (and/or other linear constraints, see below)
- This corresponds to operation of a superdirective array (see (**) p25), but now noise field is unknown
- Implemented as adaptive filter (e.g. constrained LMS algorithm)
- Notation:

$$z[k] = \mathbf{f}^T \mathbf{y}[k] = \sum_{m=1}^M \mathbf{f}_m^T \mathbf{y}_m[k]$$



$$\mathbf{y}[k] = [y_1^T[k] \quad y_2^T[k] \quad \dots \quad y_M^T[k]]^T$$

$$\mathbf{f} = [\mathbf{f}_1^T \quad \mathbf{f}_2^T \quad \dots \quad \mathbf{f}_M^T]^T$$

$$\mathbf{y}_m[k] = [y_m[k] \quad y_m[k-1] \quad \dots \quad y_m[k-L+1]]^T$$

$$\mathbf{f}_m = [f_m[0] \quad f_m[1] \quad \dots \quad f_m[L-1]]^T$$

LCMV-beamforming

LCMV = Linearly Constrained Minimum Variance

- \mathbf{f} designed to minimize variance of output $z[k]$:

$$\min_{\mathbf{f}} E\{z^2[k]\} = \min_{\mathbf{f}} \mathbf{f}^T \cdot \mathbf{R}_{yy}[k] \cdot \mathbf{f}$$

- to avoid desired signal distortion/cancellation, add linear constraints:

$$\mathbf{C}^T \cdot \mathbf{f} = \mathbf{b}, \text{ with } \mathbf{C} \in \mathbb{R}^{ML \times J}, \mathbf{b} \in \mathbb{R}^J$$

- if noise and speech are uncorrelated, constrained output power minimization corresponds to constrained noise power minimization

- Type of constraints:

- Frequency response in look-direction.
- Point, line and derivative constraints

Ex: $\sum_{m=1}^M F_m(z) = 1$ (for broadside)
(=L constraints)

- Solution is (obtained using Lagrange-multipliers, etc.):

$$\mathbf{f}_{opt} = \mathbf{R}_{yy}^{-1}[k] \cdot \mathbf{C} \cdot (\mathbf{C}^T \cdot \mathbf{R}_{yy}^{-1}[k] \cdot \mathbf{C})^{-1} \mathbf{b}$$