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# ***Conventional Narrowband Beamforming***

**ECE 6279: Spatial Array Processing  
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Lecture 9**

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# Where We Are in J&D, Sort Of

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- **Most of the lecture material drawn from:**
  - Sec. 4.4, sort of...
  - Book covers full wideband case
  - We'll just cover the narrowband case this semester
- **Some of the lecture is based on what is usually seen in the literature**



# Inspiration

- Recall basic delay-and-sum beamformer:

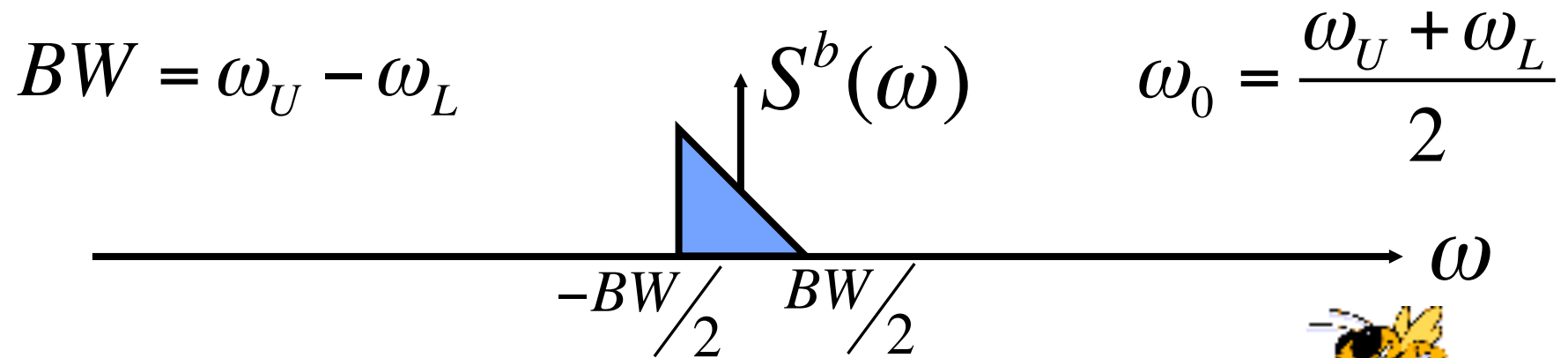
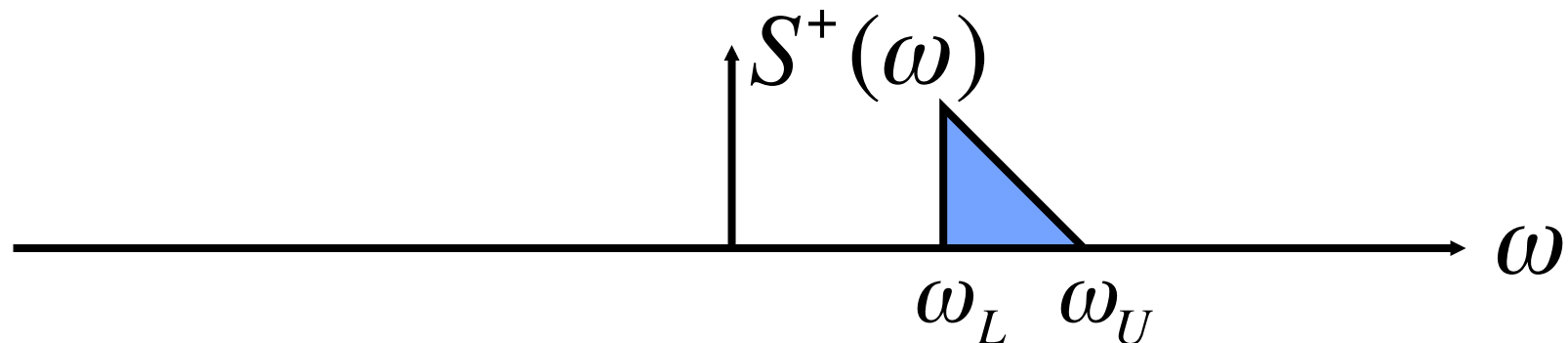
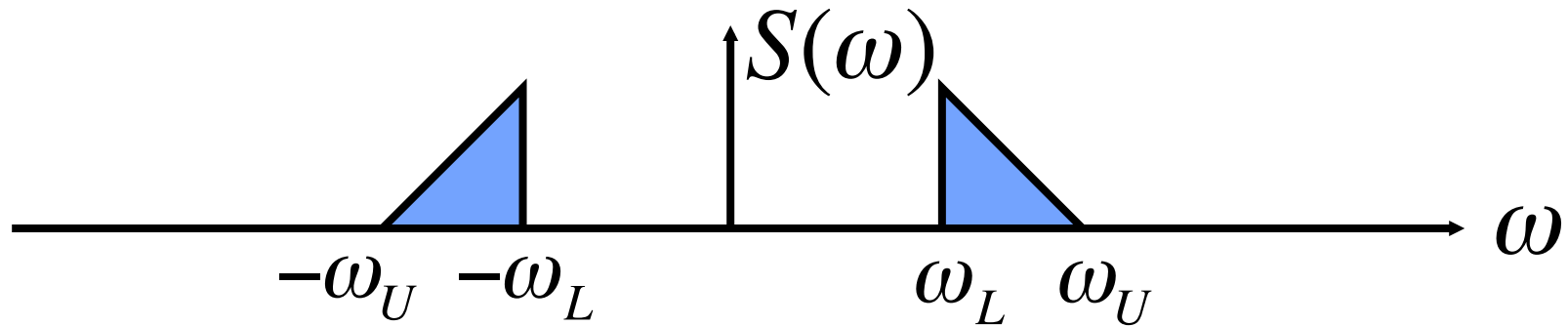
$$z(t) \equiv \sum_{m=0}^{M-1} w_m y_m(t - \Delta_m)$$

- Taking FT on both sides, we see the delay corresponds to a phase shift in Fourier space:

$$Z(\omega) = \sum_{m=0}^{M-1} w_m Y_m(\omega) \exp(-j\omega\Delta_m)$$



# Analytic Representations (Fourier)



# Analytic Representations (Time)

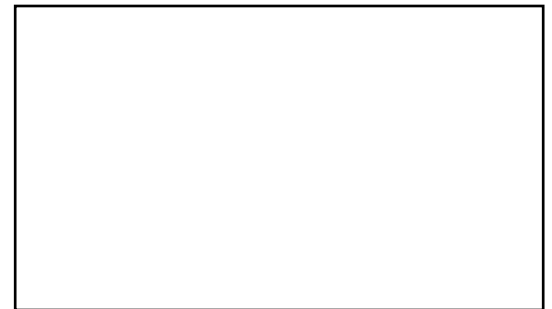
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- **Easiest to represent a real narrowband signal in terms of the analytic signal:**

$$s(t) = \text{Re}\{s^+(t)\}$$

- **The analytic signal is:**

$$s^+(t) = s_b(t)\exp(j\omega^0 t)$$



# The Narrowband Approximation

$$s^+(t) = s_b(t) \exp(j\omega^0 t)$$

- **Narrowband assumption** means that:

$$s_b(t - \tau) \approx s_b(t) \text{ for largest possible } \tau$$

- **Consider plane waves:**

$$\begin{aligned} y_m^+(t) &\equiv f^+(\vec{x}_m, t) = s^+(t - \vec{\alpha}^0 \cdot \vec{x}_m) \\ &= s_b(t - \vec{\alpha}^0 \cdot \vec{x}_m) \exp(-j\omega^0 \vec{\alpha}^0 \cdot \vec{x}_m) \exp(j\omega^0 t) \\ &\approx \exp(-j\vec{k}^0 \cdot \vec{x}_m) \underbrace{s_b(t) \exp(j\omega^0 t)}_{s^+(t)} \end{aligned}$$



# Data in Baseband Representation

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$$y_m^+(t) \approx \exp(-j\vec{k}^0 \cdot \vec{x}_m) s_b(t) \exp(j\omega^0 t)$$

- **We'll work with the baseband signal at the sensors**

$$y_m^b(t) = \exp(-j\vec{k}^0 \cdot \vec{x}_m) s_b(t)$$

**remember this is really an approximation, but people usually write an equals sign**



# Narrowband Beamforming

$$y_m^b(t) = \exp(-j\vec{k}^0 \cdot \vec{x}_m) s_b(t)$$

$$z(t) = \sum_{m=0}^{M-1} w_m \exp(j\vec{k} \cdot \vec{x}_m) \overset{\swarrow}{y_m^b(t)}$$

$$= \sum_{m=0}^{M-1} w_m \exp(j\vec{k} \cdot \vec{x}_m) \exp(-j\vec{k}^0 \cdot \vec{x}_m) s_b(t)$$

$$= \sum_{m=0}^{M-1} w_m \exp\{j(\vec{k} - \vec{k}^0) \cdot \vec{x}_m\} s_b(t)$$





# Vector Notation for the Data

$$y_m^b(t) = \exp(-j\vec{k}^0 \cdot \vec{x}_m) s_b(t)$$

$$\begin{bmatrix} y_0^b(t) \\ \vdots \\ y_{M-1}^b(t) \end{bmatrix} = \begin{bmatrix} \exp(-j\vec{k}^0 \cdot \vec{x}_0) \\ \vdots \\ \exp(-j\vec{k}^0 \cdot \vec{x}_{M-1}) \end{bmatrix} s_b(t)$$

$$\mathbf{y}(t) = \mathbf{e}(\vec{k}^0) s_b(t)$$

**steering vector**



# Vector Notation for Beamforming (1)

$$z(t) = \sum_{m=0}^{M-1} w_m \exp(j\vec{k} \cdot \vec{x}_m) y_m^b(t)$$
$$= \underbrace{\begin{bmatrix} \exp(j\vec{k} \cdot \vec{x}_0) & \cdots & \exp(j\vec{k} \cdot \vec{x}_{M-1}) \end{bmatrix}}_{\mathbf{e}^H(\vec{k})} \begin{bmatrix} w_0 y_0(t) \\ \vdots \\ w_{M-1} y_{M-1}(t) \end{bmatrix}$$
$$\mathbf{e}(\vec{k}) = \begin{bmatrix} \exp(-j\vec{k} \cdot \vec{x}_0) \\ \vdots \\ \exp(-j\vec{k} \cdot \vec{x}_{M-1}) \end{bmatrix}$$



# Vector Notation for Beamforming (2)

$$\begin{bmatrix} w_0 y_0^b(t) \\ \vdots \\ w_{M-1} y_{M-1}^b(t) \end{bmatrix} = \underbrace{\begin{bmatrix} w_0 & & 0 \\ & \ddots & \\ 0 & & w_{M-1} \end{bmatrix}}_{\mathbf{W} = \text{diag}(w_0, \dots, w_{M-1})} \begin{bmatrix} y_0^b(t) \\ \vdots \\ y_{M-1}^b(t) \end{bmatrix}$$
$$= \mathbf{W} \mathbf{y}(t)$$



# Vector Notation for Beamformer Output

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$$\mathbf{y}(t) = \mathbf{e}(\vec{k}^0) s_b(t)$$

$$\begin{aligned} z(t) &= \mathbf{e}^H(\vec{k}) \mathbf{W} \mathbf{y}(t) \\ &= \mathbf{e}^H(\vec{k}) \mathbf{W} \mathbf{e}(\vec{k}^0) s_b(t) \end{aligned}$$



# Another Frequent Notation

Look  
direction  
vector

$$\mathbf{e}(\vec{k}) = \begin{bmatrix} \exp(-j\vec{k} \cdot \vec{x}_0) \\ \vdots \\ \exp(-j\vec{k} \cdot \vec{x}_{M-1}) \end{bmatrix}$$

$$\vec{l}(\phi, \theta) = (\sin \phi \cos \theta, \sin \phi \sin \theta, \cos \phi)$$

$$\mathbf{e}(\phi, \theta, \lambda) = \begin{bmatrix} \exp\{jk\vec{l}(\theta, \phi) \cdot \vec{x}_0\} \\ \vdots \\ \exp\{jk\vec{l}(\theta, \phi) \cdot \vec{x}_{M-1}\} \end{bmatrix}$$

$\nwarrow$   
 $2\pi / \lambda$



# Linear Array (Phi Version)

$$\begin{aligned} \mathbf{e}(\phi, 0, \lambda) &= \begin{bmatrix} \exp\left\{-jk \sin(\phi) \frac{M-1}{2} d\right\} \\ \vdots \\ \exp\left\{jk \sin(\phi) \frac{M-1}{2} d\right\} \end{bmatrix} \\ &= \exp\left\{-jk \sin(\phi) \frac{M-1}{2} d\right\} \begin{bmatrix} 1 \\ \exp\{jkd \sin(\phi)\} \\ \vdots \\ \exp\{jkd \sin(\phi)(M-1)\} \end{bmatrix} \end{aligned}$$



# Vandermonde Structure (Phi)

$$\exp\left\{-jk \sin(\phi) \frac{M-1}{2} d\right\} \begin{bmatrix} 1 \\ \exp\{jkd \sin(\phi)\} \\ \vdots \\ \underbrace{\exp\{jkd \sin(\phi)(M-1)\}}_{\gamma} \end{bmatrix}$$

$$= \text{const} \begin{bmatrix} (e^{j\gamma})^0 \\ (e^{j\gamma})^1 \\ \vdots \\ (e^{j\gamma})^{M-1} \end{bmatrix} = \text{const} \begin{bmatrix} z^0 \\ z^1 \\ \vdots \\ z^{M-1} \end{bmatrix}$$

**“electrical angle”** ←  $\gamma$

**Vandermonde structure** ←



# Linear Array (Theta Version)

- That  $\phi$  version is popular with textbook authors, and what we used in previous lectures
  - Elevation, sort of
- More common in practice to analyze linear arrays in terms of  $\theta$ : Let  $\phi = \pi/2$

– Azimuth

$$\mathbf{e}(\pi / 2, \theta, \lambda) = \begin{bmatrix} \exp \left\{ -jk \cos(\theta) \frac{M-1}{2} d \right\} \\ \vdots \\ \exp \left\{ jk \cos(\theta) \frac{M-1}{2} d \right\} \end{bmatrix}$$





# Vandermonde Structure (Theta)

$$\exp\left\{-jk \cos(\theta) \frac{M-1}{2} d\right\} \begin{bmatrix} 1 \\ \exp\{jkd \cos(\theta)\} \\ \vdots \\ \exp\{jkd \cos(\theta)(M-1)\} \end{bmatrix}$$

$\underbrace{\hspace{10em}}_{\gamma}$

$\swarrow$  **“electrical angle”**

$\longleftarrow$  **Vandermonde structure**

$$= \text{const} \begin{bmatrix} (e^{j\gamma})^0 \\ (e^{j\gamma})^1 \\ \vdots \\ (e^{j\gamma})^{M-1} \end{bmatrix} = \text{const} \begin{bmatrix} z^0 \\ z^1 \\ \vdots \\ z^{M-1} \end{bmatrix}$$



# Power in the Beamformed Signal

$$\begin{aligned} |z(t)|^2 &= \mathbf{e}^H(\vec{k}) \mathbf{W} \mathbf{y}(t) [\mathbf{e}^H(\vec{k}) \mathbf{W} \mathbf{y}(t)]^H \\ &= \mathbf{e}^H(\vec{k}) \mathbf{W} \underbrace{\mathbf{y}(t) \mathbf{y}^H(t)} \mathbf{W}^H \mathbf{e}(\vec{k}) \end{aligned}$$

**spatial correlation matrix  $\rightarrow \mathbf{R}(t)$**

$$= \mathbf{e}^H(\vec{k}) \mathbf{W} \mathbf{R}(t) \mathbf{W}^H \mathbf{e}(\vec{k})$$



# Reviewing the Strategy

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- To find the direction of arrival of a single source at time  $t$ , find the parameters that maximize the power in the beamformed signal:

$$|z(t)|^2 = \left| \mathbf{e}^H(\vec{k}) \mathbf{W} \mathbf{y}(t) \right|^2 = \left| \mathbf{e}^H(\phi, \theta, \lambda) \mathbf{W} \mathbf{y}(t) \right|^2$$

- If the source isn't moving, we can average over time
  - We'll revisit this issue in detail in a later lecture
- If multiple targets, we can find multiple peaks...
  - ...but this isn't really “optimal” anymore
  - We'll revisit this too



# SCM for a Plane Wave

- **Spatial correlation matrix for a single plane wave:**

$$\begin{aligned}\mathbf{R}(t) &= \mathbf{y}(t)\mathbf{y}^H(t) \approx \mathbf{e}(\vec{k}^0)s_b(t)s_b^H(t)\mathbf{e}^H(\vec{k}^0) \\ &= \underbrace{\mathbf{e}(\vec{k}^0)\mathbf{e}^H(\vec{k}^0)}_{\text{Spatial Correlation Matrix}} |s_b(t)|^2\end{aligned}$$

$$\begin{bmatrix} \exp(j\vec{k}^0 \cdot \vec{x}_0) \\ \vdots \\ \exp(j\vec{k}^0 \cdot \vec{x}_{M-1}) \end{bmatrix} \begin{bmatrix} \exp(-j\vec{k}^0 \cdot \vec{x}_0) & \cdots & \exp(-j\vec{k}^0 \cdot \vec{x}_{M-1}) \end{bmatrix}$$

$$[\mathbf{R}(t)]_{m_1, m_2} = |s_b(t)|^2 \exp\{j\vec{k}^0 \cdot (\vec{x}_{m_1} - \vec{x}_{m_2})\}$$



# SCM for a Plane Wave/Linear Array

$$[\mathbf{R}(t)]_{m_1, m_2} = |s_b(t)|^2 \cdot \exp\{j\vec{k}^0(\vec{x}_{m_1} - \vec{x}_{m_2})\}$$

- For a linear array with spacing  $d$ :

$$[\mathbf{R}(t)]_{m_1, m_2} = |s_b(t)|^2 \cdot \exp\{j\vec{k}^0(m_1 - m_2)d\}$$

$$\mathbf{R}(t) = \begin{bmatrix} R_0 & R_1 & R_2 & \cdots & R_{M-1} \\ R_1^* & R_0 & R_1 & \cdots & R_{M-2} \\ R_2^* & R_1^* & R_0 & \cdots & R_{M-3} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ R_{M-1}^* & R_{M-2}^* & R_{M-3}^* & \cdots & R_0 \end{bmatrix}$$

← Toeplitz structure



# Unweighted Case

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- **Suppose  $\mathbf{W}=\mathbf{I}$ :**

$$\begin{aligned} |z(t)|^2 &= \mathbf{e}^H(\vec{k}) \mathbf{W} \mathbf{R}(t) \mathbf{W}^H \mathbf{e}(\vec{k}) \\ &= \mathbf{e}^H(\vec{k}) \mathbf{R}(t) \mathbf{e}(\vec{k}) \end{aligned}$$

