

E9 231: Digital Array Signal Processing

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Eigenspace Beamformers

One way of identifying the lower dimensional subspace of \mathbf{S}_X is through the eigen decomposition.

$$\mathbf{S}_X = \sum_{i=1}^N \lambda_i \Phi_i \Phi_i^H = \mathbf{U} \Lambda \mathbf{U}^H \quad (1)$$

where \mathbf{U} is an $N \times N$ matrix of eigenvectors,

$$\mathbf{U} = [\Phi_1 \ \Phi_2 \ \cdots \ \Phi_N]$$

and Λ is the diagonal matrix of the ordered eigenvalues,

$$\Lambda = \text{diag} [\lambda_1 \ \lambda_2 \ \cdots \ \lambda_N],$$

$$\begin{aligned} \lambda_1 &\geq \lambda_2 \geq \cdots \geq \lambda_N > 0 \\ \Phi_i^H \Phi_j &= \delta_{ij} \end{aligned}$$

First r eigenvalues: Signal + Interference

Remaining eigenvalues: Noise only

$$\begin{aligned} \mathbf{U}_{S+I} &\triangleq [\Phi_1 \ \Phi_2 \ \cdots \ \Phi_r] \Rightarrow \text{eigenspace that spans signal + interference} \\ \mathbf{U}_N &\triangleq [\Phi_{r+1} \ \Phi_{r+2} \ \cdots \ \Phi_N] \Rightarrow \text{spans the noise only subspace} \end{aligned}$$

$$\mathbf{S}_X = \mathbf{U}_{S+I} \Lambda_{S+I} \mathbf{U}_{S+I}^H + \mathbf{U}_N \Lambda_N \mathbf{U}_N^H \quad (2)$$

$$\begin{aligned} \Lambda_{S+I} &\triangleq \text{diag} [\lambda_1 \ \lambda_2 \ \cdots \ \lambda_r] \\ \Lambda_N &\triangleq \text{diag} [\lambda_{r+1} \ \lambda_{r+2} \ \cdots \ \lambda_N] \end{aligned}$$

$$\begin{aligned} [\mathbf{S}_x]_T &= \mathbf{U}_{S+I} \boldsymbol{\Lambda}_{S+I} \mathbf{U}_{S+I}^H \\ [\mathbf{S}_x]_T^{-1} &= \mathbf{U}_{S+I} \boldsymbol{\Lambda}_{S+I}^{-1} \mathbf{U}_{S+I}^H \end{aligned}$$

The \mathbf{w}_{mpdr} reduces to (if we replace \mathbf{S}_x by $[\mathbf{S}_x]_T$)

$$\mathbf{w}_{mpdr,es} = \gamma_{es} [\mathbf{S}_x]_T^{-1} \mathbf{v}_m \quad (3)$$

where

$$\gamma_{es} = \frac{1}{\mathbf{v}_m^H [\mathbf{S}_x]_T^{-1} \mathbf{v}_m}$$

Rationale

Signal + D interfering plane wave model

$$\mathbf{X} = \mathbf{V}\mathbf{F} + \mathbf{N} \quad (4)$$

$$\mathbf{V} = [\mathbf{v}_s \ \mathbf{v}_{\mathbf{k}_1} \ \dots \ \mathbf{v}_{\mathbf{k}_D}]_{N \times (D+1)}$$

$$\mathbf{F} = [F_s \ F_1 \ \dots \ F_D]_{(D+1) \times 1}^T$$

$$\mathbf{S}_x = \mathbf{V}\mathbf{S}_f\mathbf{V}^H + \sigma_n^2 \mathbf{I} \quad (5)$$

where the noise is assumed to be spatially white, making \mathbf{S}_x positive definite.

Assume $(D+1) < N$.

$$\lambda_{\mathbf{S}_{x_i}} = \begin{cases} \lambda_{\mathbf{S}_{f_i}} + \sigma_n^2, & 1 \leq i \leq D+1 \\ \sigma_n^2, & D+1 < i \leq N \end{cases}$$

We can express

$$\mathbf{V}\mathbf{S}_f\mathbf{V}^H = [\mathbf{U}_S \ \mathbf{U}_S^\perp] \begin{bmatrix} \boldsymbol{\Lambda}_S & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{U}_S^H \\ \mathbf{U}_S^{\perp H} \end{bmatrix} \quad (6)$$

Subspace method:

$$\mathbf{S}_x = [\mathbf{U}_S \ \mathbf{U}_S^\perp] \begin{bmatrix} \boldsymbol{\Lambda}_S + \sigma_n^2 \mathbf{I}_{D+1} & \mathbf{0} \\ \mathbf{0} & \sigma_n^2 \mathbf{I}_{N-(D+1)} \end{bmatrix} \begin{bmatrix} \mathbf{U}_S^H \\ \mathbf{U}_S^{\perp H} \end{bmatrix} \quad (7)$$

- We can look at “how many equal minimum eigenvalues \mathbf{S}_x has” to find $D+1$.
 - Multiplicity of σ_n^2 is $N - D - 1$.
 - $\mathcal{R}(\mathbf{U}_S) = \mathcal{R}(\mathbf{V})$: signal + interference subspace
 - $\mathcal{R}(\mathbf{U}_S^\perp) \perp \mathcal{R}(\mathbf{V})$: noise-only subspace
- Dropping \mathbf{U}_S^\perp (or $\sigma_n^2 \mathbf{I}_{N-(D+1)}$) part is good because
 - No signal there anyway

– Inverting $[\mathbf{S}_\mathbf{X}]_T$ is easier than inverting $\mathbf{S}_\mathbf{X}$ as we dropped the smallest eigenvalues.

Note that $\mathbf{S}_\mathbf{X}^{-1}\mathbf{v}_s = [\mathbf{S}_\mathbf{X}]_T^{-1}\mathbf{v}_s$:

$$\mathbf{S}_\mathbf{X}^{-1}\mathbf{v}_s = \left[\mathbf{U}_S (\Lambda_S + \sigma_n^2 \mathbf{I})^{-1} \mathbf{U}_S^H + \sigma_n^{-2} \mathbf{U}_S^\perp (\mathbf{U}_S^\perp)^H \right] \mathbf{v}_s$$

The desired signal direction $\mathbf{v}_s \in \mathcal{R}(\mathbf{V}) = \mathcal{R}(\mathbf{U}_S)$

$$\begin{aligned} \mathbf{S}_\mathbf{X}^{-1}\mathbf{v}_s &= \mathbf{U}_S (\Lambda_S + \sigma_n^2 \mathbf{I})^{-1} \mathbf{U}_S^H \mathbf{v}_s + \mathbf{0} = [\mathbf{S}_\mathbf{X}]_T^{-1} \mathbf{v}_s \\ \mathbf{w}_{mpdr,es} &= \frac{\sum_{i=1}^r \frac{1}{\lambda_i} (\Phi_i^H \mathbf{v}_s) \Phi_i}{\sum_{i=1}^r \frac{1}{\lambda_i} |\Phi_i^H \mathbf{v}_s|^2} \end{aligned} \quad (8)$$

λ_i s are the eigenvalues of $[\mathbf{S}_\mathbf{X}]_T$.

This beamformer is called *eigenspace beamformer* or *signal + interference subspace beamformer*.

Another Interpretation:

$\mathbf{P}_{\mathbf{U}_{S+I}} \triangleq \mathbf{U}_{S+I} \mathbf{U}_{S+I}^H$ is a projection matrix as $\mathbf{U}_{S+I}^H \mathbf{U}_{S+I} = \mathbf{I}$.

$$\mathbf{v}_p \triangleq \mathbf{P}_{\mathbf{U}_{S+I}} \mathbf{v}_s = \mathbf{U}_{S+I} \mathbf{U}_{S+I}^H \mathbf{v}_s \quad (9)$$

The projection beamformer is given by

$$\begin{aligned} \mathbf{w}_p &= \gamma_p \mathbf{S}_\mathbf{X}^{-1} \mathbf{v}_p \\ &= \gamma_p \mathbf{S}_\mathbf{X}^{-1} \mathbf{U}_{S+I} \mathbf{U}_{S+I}^H \mathbf{v}_s \\ &= \gamma_p (\mathbf{U}_{S+I} \Lambda_{S+I}^{-1} \mathbf{U}_{S+I}^H + \mathbf{U}_N \Lambda_N^{-1} \mathbf{U}_N^H) \mathbf{U}_{S+I} \mathbf{U}_{S+I}^H \mathbf{v}_s \\ &= \gamma_p \mathbf{U}_{S+I} \Lambda_{S+I}^{-1} \mathbf{U}_{S+I}^H \mathbf{v}_s, \end{aligned} \quad (10)$$

which is the same as $\mathbf{w}_{mpdr,es}$ when $\gamma_p = \gamma_{es}$.

A third interpretation

$$\begin{aligned} \mathbf{X}_{es} &\triangleq \mathbf{U}_{S+I}^H \mathbf{X} \\ \mathbf{v}_{es} &\triangleq \mathbf{U}_{S+I}^H \mathbf{v}_s \end{aligned}$$

If we build an MPDR beamformer with \mathbf{X}_{es} replacing \mathbf{X} and \mathbf{v}_{es} replacing \mathbf{v}_s ,

$$\mathbf{w}_{mpdr} = \frac{\Lambda_{S+I}^{-1} \mathbf{v}_{es}}{\mathbf{v}_{es}^H \Lambda_{S+I}^{-1} \mathbf{v}_{es}} \quad (11)$$

With diagonal loading:

$$\mathbf{w}_{es,mpdr,dL}^H = \frac{\mathbf{v}_{es}^H (\Lambda_{S+I} + \sigma_L^2 \mathbf{I})^{-1}}{\mathbf{v}_{es}^H (\Lambda_{S+I} + \sigma_L^2 \mathbf{I})^{-1} \mathbf{v}_{es}} \quad (12)$$

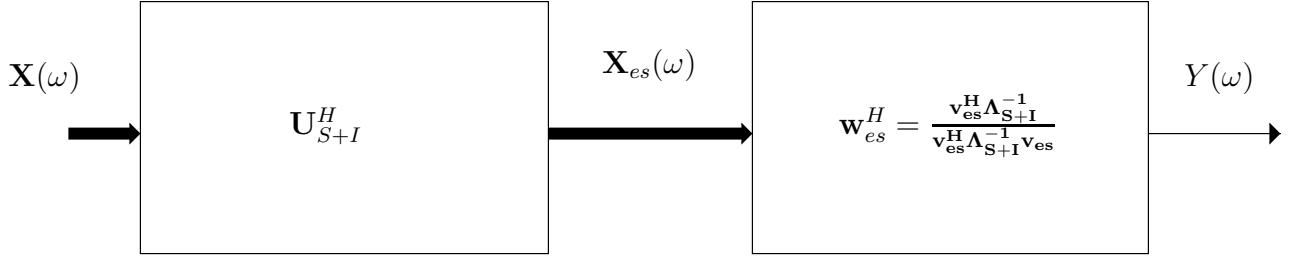


Figure 1: Eigenspace beamformer

LCMP Eigenspace Beamforming

$$\begin{aligned}\mathbf{C}^H \mathbf{w} &= \mathbf{g}^H \\ \mathbf{w} &= \mathbf{U}_{S+I} \mathbf{w}_{es}\end{aligned}$$

Constraint becomes

$$\begin{aligned}\mathbf{C}^H \mathbf{U}_{S+I} \mathbf{w}_{es} &= \mathbf{g}^H \\ \Rightarrow \mathbf{C}_{es}^H \mathbf{w}_{es} &= \mathbf{g}^H, \text{ where } \mathbf{C}_{es} = \mathbf{U}_{S+I}^H \mathbf{C} \\ \mathbf{w}_{LCMP,es}^H &= \mathbf{g}^H (\mathbf{C}_{es}^H \Lambda_{S+I}^{-1} \mathbf{C}_{es})^{-1} \mathbf{C}_{es}^H \Lambda_{S+I}^{-1}\end{aligned}\tag{13}$$

Performance of Eigenspace Beamformer

1. Critically dependent on knowing $\dim(S+I)$.
2. At high SNR ($>$ INR), underestimating $\dim(S+I)$ is OK.
3. At low SNR, underestimating $\dim(S+I)$ is bad as pushing the signal into the noise-only subspace will result in a drastic reduction of performance.

Dominant Mode Rejection (DMR) Beamformers

Eigen decomposition of \mathbf{S}_X gives

$$\mathbf{S}_X = \sum_{i=1}^{D_m} \lambda_i \Phi_i \Phi_i^H + \sum_{i=D_m+1}^N \lambda_i \Phi_i \Phi_i^H\tag{14}$$

Replace the eigenvalues $\lambda_{D_m+1}, \dots, \lambda_N$ by $\alpha \triangleq \frac{1}{N-D_m} \sum_{i=D_m+1}^N \lambda_i$.

Note 1: $\alpha = \frac{1}{N-D_m} \left(\text{tr}(\mathbf{S}_X) - \sum_{i=1}^{D_m} \lambda_i \right)$.
So it is enough to find $\lambda_1, \dots, \lambda_{D_m}$.

Note 2: If the noise is spatially white and there are $< D_m + 1$ desired signal + interferers, $\alpha = \sigma_n^2 \Rightarrow$ same as eigenspace decomposition.

Modified \mathbf{S}_X :

$$\tilde{\mathbf{S}}_X = \sum_{i=1}^{D_m} \lambda_i \Phi_i \Phi_i^H + \alpha \sum_{i=D_m+1}^N \Phi_i \Phi_i^H \quad (15)$$

Dominant-mode subspace,

$$\mathbf{U}_{dm} \triangleq [\Phi_1 \ \cdots \ \Phi_{D_m}]$$

Noise subspace,

$$\begin{aligned} \mathbf{U}_{dm}^\perp &\triangleq [\Phi_{D_m+1} \ \cdots \ \Phi_N] \\ \tilde{\mathbf{S}}_X &= \mathbf{U}_{dm} \Lambda_{dm} \mathbf{U}_{dm}^H + \alpha \mathbf{U}_{dm}^\perp (\mathbf{U}_{dm}^\perp)^H \end{aligned}$$

where $\Lambda_{dm} = \text{diag}[\lambda_1, \dots, \lambda_{D_m}]$.

$$\begin{aligned} \tilde{\mathbf{S}}_X^{-1} &= \mathbf{U}_{dm} \Lambda_{dm}^{-1} \mathbf{U}_{dm}^H + \frac{1}{\alpha} \mathbf{U}_{dm}^\perp (\mathbf{U}_{dm}^\perp)^H \\ &= \frac{1}{\alpha} (\mathbf{U}_{dm} \alpha \Lambda_{dm}^{-1} \mathbf{U}_{dm}^H + \mathbf{P}_{dm}^\perp) \end{aligned} \quad (16)$$

$\mathbf{P}_{dm}^\perp \triangleq \mathbf{U}_{dm}^\perp (\mathbf{U}_{dm}^\perp)^H$ is a projection matrix.

The weight vector becomes

$$\begin{aligned} \mathbf{w}_{dm} &= \frac{\tilde{\mathbf{S}}_X^{-1} \mathbf{v}_m}{\mathbf{v}_m^H \tilde{\mathbf{S}}_X^{-1} \mathbf{v}_m} \\ &= \frac{\alpha \sum_{i=1}^{D_m} \frac{1}{\lambda_i} (\Phi_i^H \mathbf{v}_m) \Phi_i + \mathbf{P}_{dm}^\perp \mathbf{v}_m}{\alpha \sum_{i=1}^{D_m} \frac{1}{\lambda_i} |\Phi_i^H \mathbf{v}_m|^2 + \mathbf{v}_m^H \mathbf{P}_{dm}^\perp \mathbf{v}_m} \end{aligned} \quad (17)$$

When $\mathbf{P}_{dm}^\perp \mathbf{v}_m = \mathbf{0}$, i.e., $\mathbf{v}_m \in (S + I)$ subspace,

$$\mathbf{w}_{dm} = \frac{\sum_{i=1}^{D_m} \frac{1}{\lambda_i} (\Phi_i^H \mathbf{v}_m) \Phi_i}{\sum_{i=1}^{D_m} \frac{1}{\lambda_i} |\Phi_i^H \mathbf{v}_m|^2} \quad (18)$$

i.e., the same as the eigenspace beamformer.

If the input spectrum matrix was $\lambda_k \Phi_k \Phi_k^H$,

$$\begin{aligned} \mathbf{P}_{\Phi_k} &= \text{output power due to the } k\text{th eigenmode} \\ &= \frac{\frac{1}{\lambda_k} |\mathbf{v}_m^H \Phi_k|^2}{\sum_{i=1}^{D_m} \frac{1}{\lambda_i} |\Phi_i^H \mathbf{v}_m|^2} \\ \lambda_k &= \lambda_k^s + \alpha, \end{aligned} \quad (19)$$

where λ_k^s is the signal component and α is the noise component.

- If λ_k is large, \mathbf{P}_{Φ_k} will be small. Hence the name “dominant mode rejection” beamformer.
- As the projection of \mathbf{v}_m on Φ_k increases, the depth of the null on the k th eigenmode decreases.
- As \mathbf{v}_m changes (from \mathbf{v}_a), the projection of \mathbf{v}_m in the noise-only subspace increases.
- As the eigenspace beamformer totally discarded the noise-only subspace, there would be no hope of recovering the signal if we underestimated D_m and the signal happened to be in the noise-only subspace. The DMR beamformer works even when D_m is underestimated and the signal lies fully in the noise-only subspace.
- For high SNR ($>\text{INR}$), use eigenspace beamformer. For low SNR, use the DMR beamformer.
- DMR fails when SNR is high and there is a significant mismatch between \mathbf{v}_m and \mathbf{v}_a .