Mandatory 2 IN5450

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Power estimation from classical and adaptive beamforming

- R
- Classical / DAS
- Capon's / MV
- Eigenvalues / vectors (signal+noise / noise spaces)
- Eigenvalue
- MUSIC

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The problem

- 10 element ULA array, with $\lambda/2$ spacing.
- Two incoherent signals at -10 and 0 degrees, with added spatially white noise.
- SNR is 0 for both sources.
- 100 time samples per element.
- Data generated from (fixed) 'generate_data.m'

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Spatial correlation matrix R

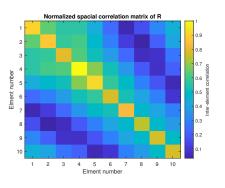
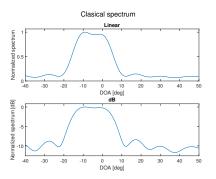


Figure: Caption

 $R = x \cdot x^H$ Correlation is a statistical relationship between two variables.

Classical / DAS beamformer

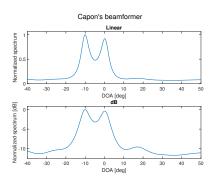


$$P(w) = w^{H}Rw$$

 $max_{w} E(w^{H}Rw) = max_{w} w^{H}(E(R))w \rightarrow w_{bf} = \frac{a(\theta)}{\sqrt{a(\theta)^{H}a(\theta)}}; \quad |w| = 1$
 $P_{DAS}(\theta) = \frac{a^{H}Ra}{a^{H}a}$
 $a_{j} = exp(j*-kd*sin(\theta*pi/180)*(j-1)); \quad j = 1,...,10$

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Capon's / Minimum variance beamformer



$$P(w) = w^H R w$$

 $min_w E(w^H R w); \quad w^H a(\theta) = 1$
 $w_{CAP} = \frac{R^{-1}a}{a^H R^{-1}a}$

$$P_{MV}(\theta) = \frac{1}{a^H R^{-1} a}$$

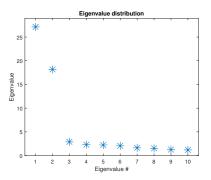
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Eigenvalues of R

$$R = U\Lambda U^H$$

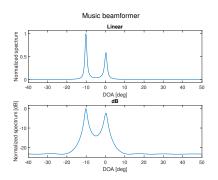


Based on the signal and noise model we know that the eigenvectors corresponding to the largest eigenvalues span the signal+noise space. While the smallest span the noise space.

Largest != signal. But signal vectors are linear combinations.

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MUSIC beamformer with 2 sources



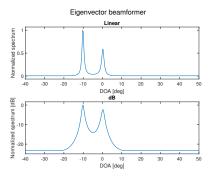
$$R = U_s \Lambda_s U_s^H + U_n \Lambda_n U_n^H$$

 $U_n^H a(\theta) = 0, \quad \theta \in (1, ..., M), \quad \text{(But we have estimates)}$
2 Sources \rightarrow We use 8 of 10 eigenvectors.
 $\Pi = U_n U_n^H \quad \text{Orthogonal projector onto noise subspace}$
 $P_{MUSIC} = \frac{a^H a}{a^H \Pi a}$

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Eigenvector beamformer with 2 sources

Again we use 8 of 10 eigenvectors.



Similar to MUSIC and uses noise space eigenvalue matrix as well

$$P_{EV} = \frac{a^H a}{a^H U_n \Lambda_n^{-1} U_n^H a}$$

Diagonal elements of Λ_n^{-1}

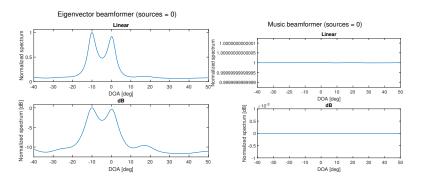
0.3469, 0.4354, 0.4481, 0.4883, 0.6131, 0.6589, 0.7917, 0.8437

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Assuming 0 sources

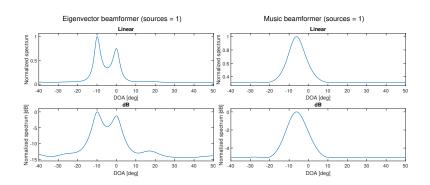
0 Sources \rightarrow We use 10 of 10 eigenvectors.



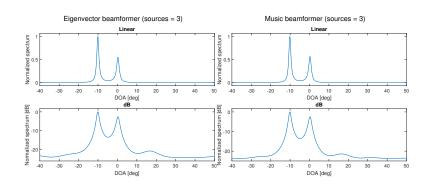
For 0 sources, $U_n\Lambda_n^{-1}U_n^H=R^{-1}$. So it is like the MV beamformer. MUSIC whitens noise, 0 sources corresponds to only noise.

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Assuming 1 source



Assuming 3 sources



References

H. Krim, M. Viberg, Two decades of array signal processing research
 The parametric approach, IEEE Signal Processing Magazine,
 pp. 67–94, July 1996, http://dx.doi.org/10.1109/79.526899

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