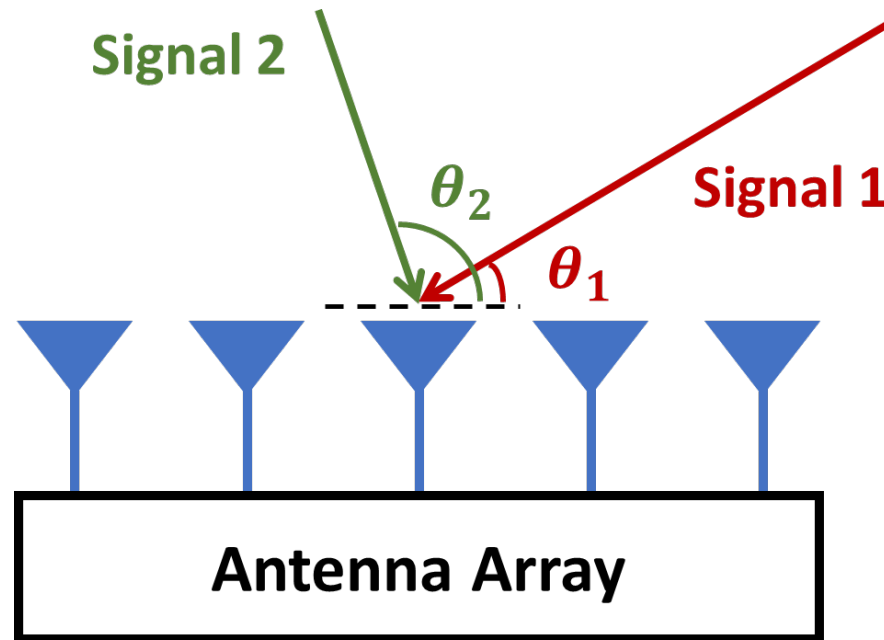


Approaches for Angle of Arrival Estimation

Wenguang Mao

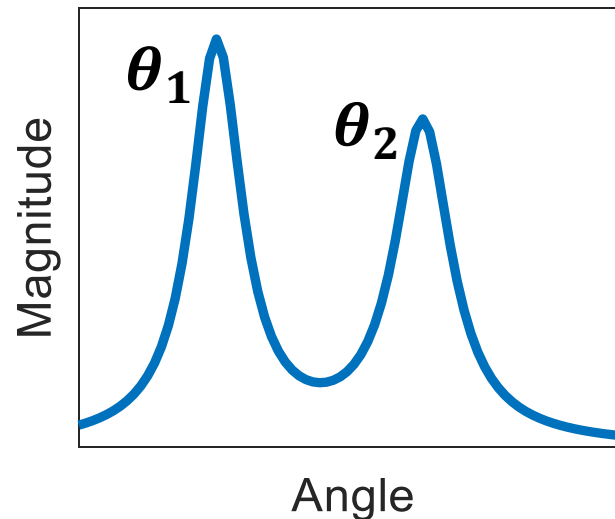
Angle of Arrival (AoA)

- Definition: the elevation and azimuth angle of incoming signals
- Also called direction of arrival (DoA)



AoA Estimation

- Applications: localization, tracking, gesture recognition,
- Requirements: antenna array
- Approaches:
 - Generate a power profile over various incoming angles
 - Determine all AoA θ_i

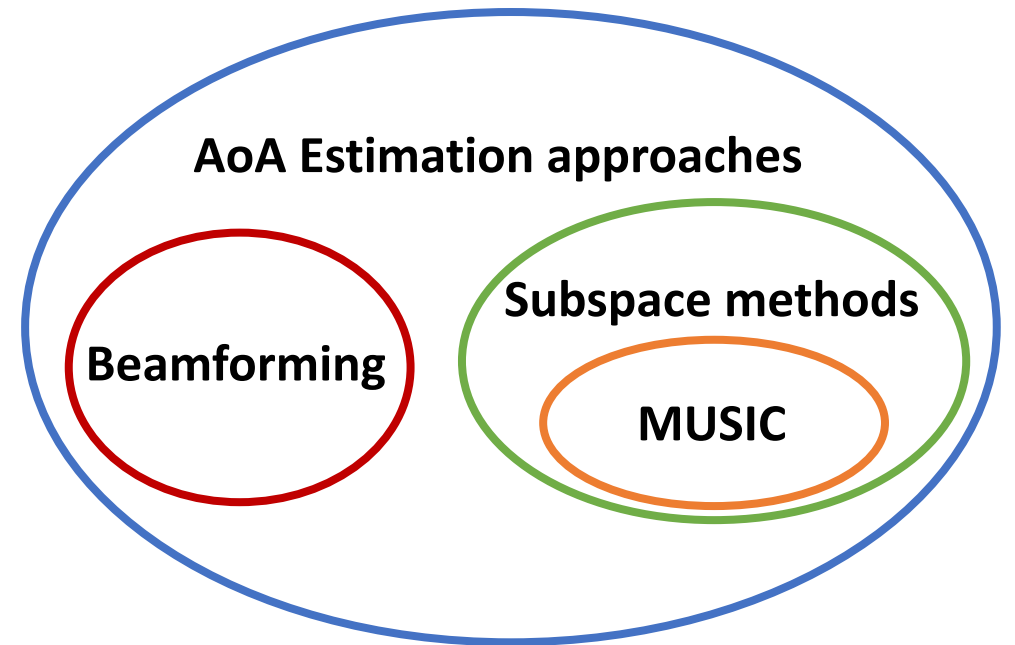


Related Concepts

- Synthetic aperture radar (SAR)
 - Using a moving antenna to emulate an array
 - Alternative way of using physical antenna array
 - **NOT** an estimation approach in the context of AoA
 - Most AoA estimation methods can be applied to both physical antenna array and SAR
- In this presentation, we only focus on antenna array
 - May require some modification when applied to SAR

Related Concepts

- **Beamforming**
 - A class of AoA estimation approaches
- **MUSIC**
 - A specific algorithm in subspace-based approaches



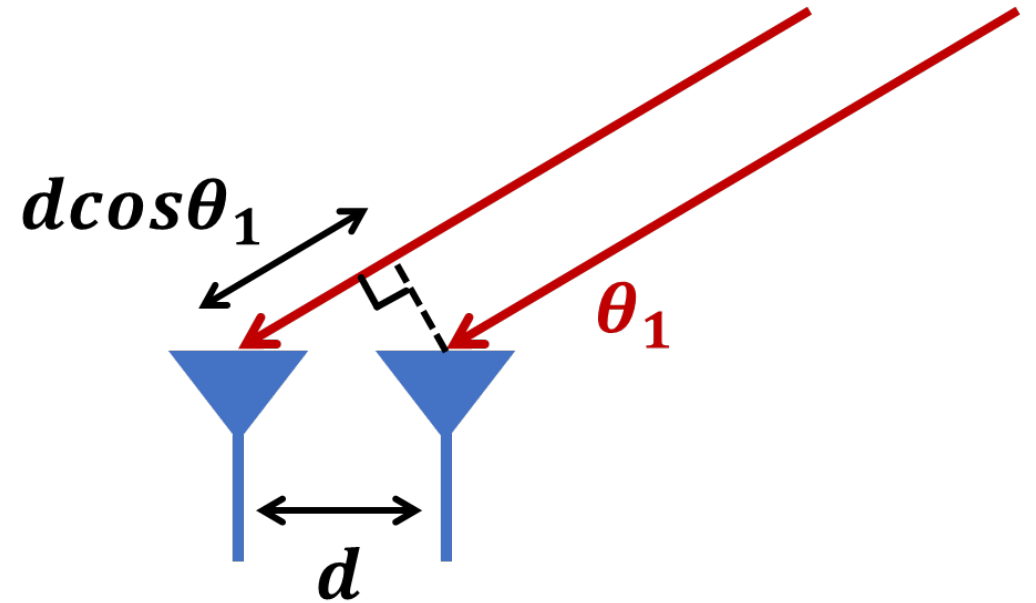
Approaches for AoA Estimation

- **Naïve approach**
- Beamforming approaches
 - **Bartlett method**
 - **MVDR**
 - Linear prediction
- Subspace based approaches
 - **MUSIC** and its variants
 - ESPRIT
- Maximum likelihood estimator
-

Key Insights

- Phase changes over antennas are determined by the incoming angle
- Far-field assumption
- Phase of the antenna 1: ϕ_1
- Phase of the antenna 2: ϕ_2
- Then the difference is given by

$$\phi_2 - \phi_1 = 2\pi \frac{d \cos \theta_1}{\lambda} + 2k\pi$$



Naïve approach

- Determine AoA based on the phase difference of two antenna

$$\cos\theta_1 = \left(\frac{\Delta\phi}{2\pi} - k\right) \frac{\lambda}{d}$$

- Problems:
 - Works for only one incoming signals
 - Phase measurement could be noisy
 - Ambiguity
- Adopted and improved by **RF-IDraw**

Using Antenna Array

- Received signals at m -th antenna:

$$\mathbf{x}_m(t) = \sum_{n=1}^N \mathbf{s}_n(t) e^{j \cdot 2\pi \cdot \tau_n \cdot (m-1)} + \mathbf{n}_m(t)$$

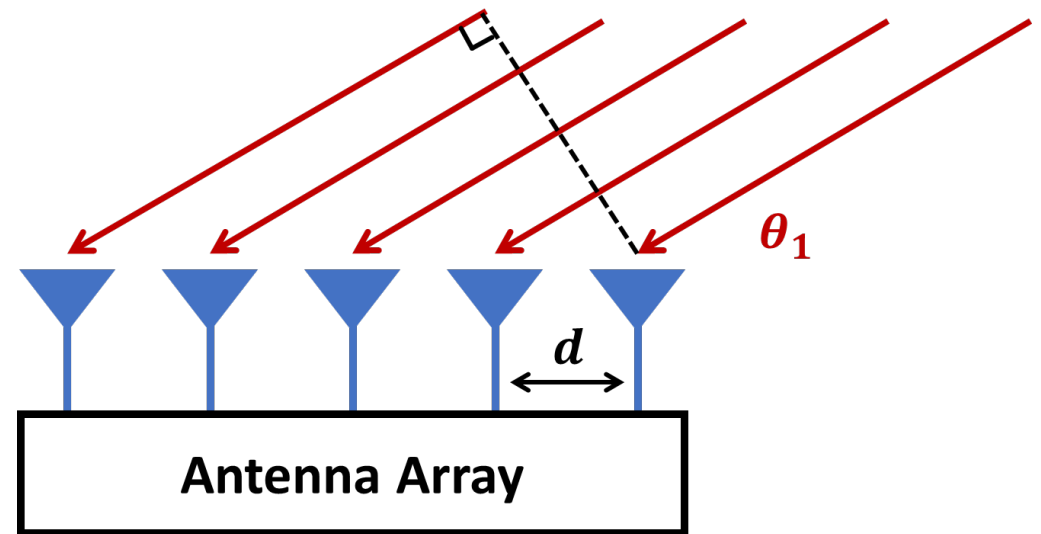
$\mathbf{s}_n(t)$: n -th source signals

$\tau_n = \frac{d \cos \theta_n}{\lambda}$: phase shift per antenna

N : the number of sources

M : the number of antennas

$\mathbf{n}_m(t)$: noise terms



Using Antenna Array

- Matrix form:

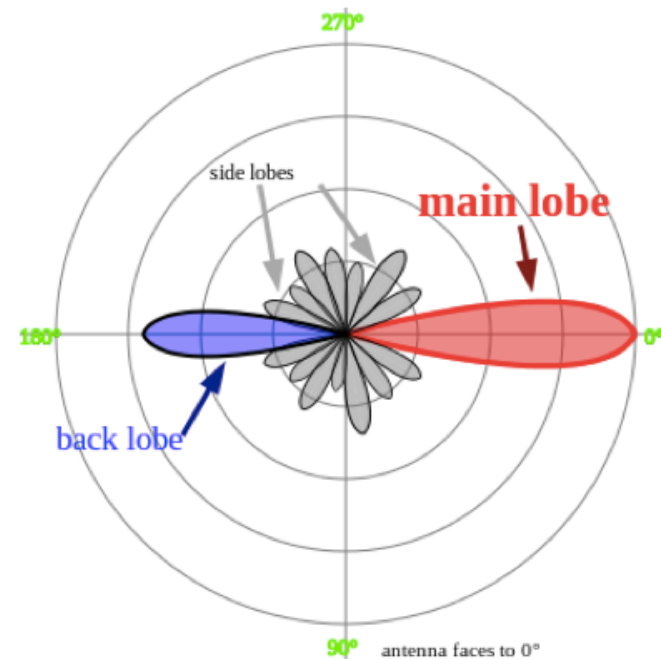
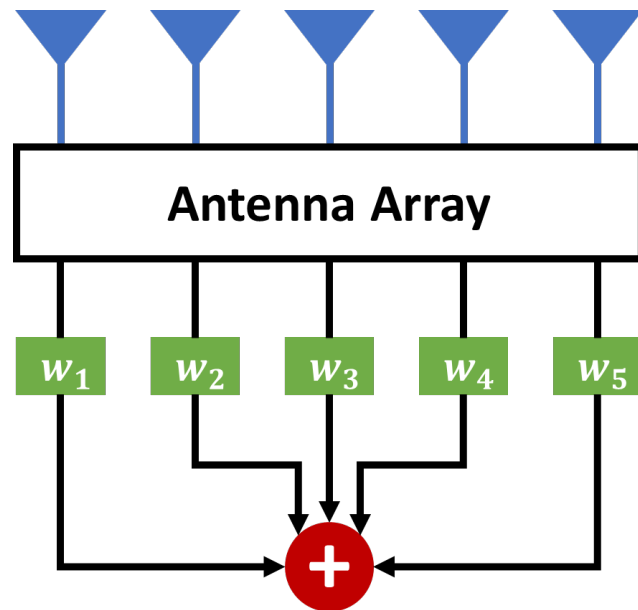
$$\begin{bmatrix} x_1(t)^T \\ x_2(t)^T \\ \vdots \\ x_M(t)^T \end{bmatrix} = [a(\theta_1) \quad a(\theta_2) \quad \dots \quad a(\theta_N)] \begin{bmatrix} s_1(t)^T \\ s_2(t)^T \\ \vdots \\ s_N(t)^T \end{bmatrix} + \begin{bmatrix} n_1(t)^T \\ n_2(t)^T \\ \vdots \\ n_M(t)^T \end{bmatrix}$$

$$X = AS + N$$

$$\text{Steering vector: } a(\theta) = [1 \quad e^{j2\pi\tau(\theta)} \quad e^{j2\pi\tau(\theta)\cdot 2} \quad \dots \quad e^{j2\pi\tau(\theta)(M-1)}]^T$$

Beamforming at the Receiver

- Definition: a method to create certain **radiation pattern** by combining signals from different antennas with different **weights**.
- Will magnify the signals from certain direction while suppressing those from other directions



Beamforming at the Receiver

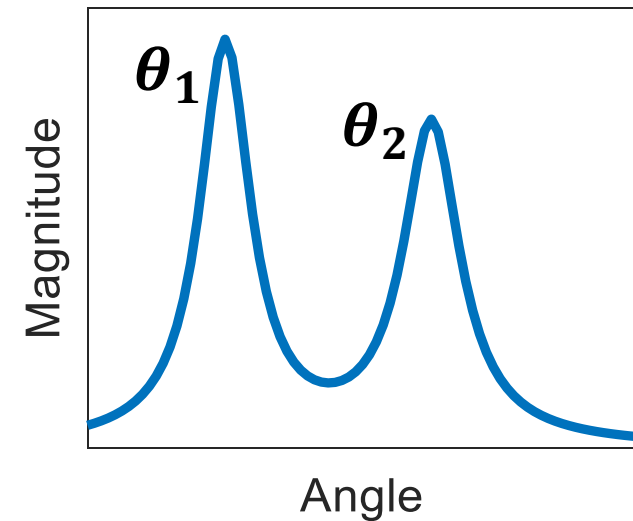
- Signals after beamforming using a **weight vector w**

$$Y = w^H X$$

- By selecting different w , the received signal Y will contain the signal sources arrived from different direction.
- Beamforming techniques are widely used in wireless communications

Beamforming at the Receiver

- Adjust the weight vector to rotate the radiation pattern to angle θ
- Measure the received signal strength $P(\theta)$
- Repeat this process for any θ in $[0, \pi]$
- Plot $(\theta, P(\theta))$
- Peaks in the plot indicates the angle of arrival



Bartlett Beamforming

- Also called: correlation beamforming, conventional beamforming, delay-and-sum beamforming, or Fourier beamforming
- **Key idea**: magnify the signals from certain direction by compensating the phase shift
- Consider one source signal $s(t)$ arrived at angle θ_0
- Signal at m -th antenna: $x_m(t) = s(t) \cdot e^{j \cdot 2\pi \cdot \tau(\theta_0)(m-1)}$
- Weight at m -th antenna: $w_m = e^{j \cdot 2\pi \cdot \tau(\theta)(m-1)}$
- Only when $\theta = \theta_0$, the received signal $Y = w^H X = \sum w_m^* x_m(t)^T$ is maximized

Phase shift



Bartlett Beamforming

- Weight vector for beamforming angle θ :

$$\mathbf{w} = \mathbf{a}(\theta)$$

This is why it is called
steering vector

- Signal power at angle θ :

$$P(\theta) = \mathbf{Y}\mathbf{Y}^H = (\mathbf{w}^H \mathbf{X})(\mathbf{w}^H \mathbf{X})^H = \mathbf{w}^H \mathbf{X}\mathbf{X}^H \mathbf{w} = \mathbf{w}^H \mathbf{R}_{XX} \mathbf{w} = \mathbf{a}^H(\theta) \mathbf{R}_{XX} \mathbf{a}(\theta)$$

Covariance matrix

- Used by **Ubicarse** with SAR

Bartlett Beamforming

- Works well when there is only one source signal
- Suffers when there are multiple sources: **very low resolution**

Minimum Variance Distortionless Response (MVDR)

- Also called Capon's beamforming
- **Key idea**: maintain the signal from the desired direction while minimizing the signals from other direction
- Mathematically, we want to find such weight vector \mathbf{w} for the beaming angle θ

$$\min\{YY^H\} = \min\{\mathbf{w}^H \mathbf{R}_{XX} \mathbf{w}\}$$

$$\text{s.t. } \mathbf{w}^H (\mathbf{a}(\theta) s(t)^T) = s(t)^T \longrightarrow \mathbf{w}^H \mathbf{a}(\theta) = 1$$



Maintain the signals from angle θ

MVDR

- Weight vector for beamforming angle θ :

$$\mathbf{w} = \frac{\mathbf{R}_{XX}^{-1} \mathbf{a}^H(\theta)}{\mathbf{a}(\theta) \mathbf{R}_{XX}^{-1} \mathbf{a}^H(\theta)}$$

- Signal power at angle θ :

$$P(\theta) = \mathbf{Y} \mathbf{Y}^H = \mathbf{w}^H \mathbf{R}_{XX} \mathbf{w} = \frac{1}{\mathbf{a}(\theta) \mathbf{R}_{XX}^{-1} \mathbf{a}^H(\theta)}$$

MVDR

- Resolution is significantly enhanced compared to Bartlett method
- But still not good enough
- Better beamforming approaches are developed, e.g., Linear Prediction
- Or resort to **subspace based approaches**

Subspace Based Approaches

- Beamforming is a way of shaping received signals
 - Can be used for estimating AoA
 - Can also be used for directional communications
- Subspace based approaches are specially designed for parameter (i.e., AoA) estimation using received signals
 - Cannot be used for extracting signals arrived from certain direction
- Subspace based approaches **decompose** the received signals into “signal subspace” and “noise subspace”
 - Leverage special properties of these subspaces for estimating AoA

Multiple Signal Classification (MUSIC)

- Key ideas: we want to find a vector q and a vector function $f(\theta)$
- Such that $q^H f(\theta) = 0$ if and only if $\theta = \theta_i$ (i.e., one of AoA)
- Then we can plot $p(\theta) = \frac{1}{\|q^H f(\theta)\|^2} = \frac{1}{f^H(\theta) q q^H f(\theta)}$
- The peaks in the plot indicates AoA
- We can expect **very sharp peak** since $q^H f(\theta) = 0$, so the inverse of its magnitude is infinity

How to find q and $f(\theta)$

Multiple Signal Classification (MUSIC)

- MUSIC gives a way to find a pair of q and $f(\theta)$
- The signals from antenna array

$$X = AS + N$$

- Covariance matrix of the signals

$$R_{XX} = E[XX^H] = E[ASS^H A^H] + E[NN^H]$$

$$R_{XX} = AE[SS^H]A^H + \sigma^2 I$$

$$R_{XX} = \underbrace{AR_{SS}A^H}_{\text{Signal terms}} + \underbrace{\sigma^2 I}_{\text{Noise terms}}$$

MUSIC

- Consider the signal term
- R_{SS} is $N \times N$ matrix, where N is the number of source signals
 - R_{SS} has the rank equal to N if source signals are independent
- A is $M \times N$ matrix, where M is the number of antenna
 - A has full column rank
 - The signal term is $M \times M$ matrix, and its rank is N
 - The signal term has N positive eigenvalues and $M - N$ zero eigenvalues, if $M > N$
 - There are $M - N$ eigenvectors q_i such that $AR_{SS}A^H q_i = 0$
 - Then $A^H q_i = 0$, where $A = [a(\theta_1) \quad a(\theta_2) \quad \cdots \quad a(\theta_N)]$
 - Then $q_i^H a(\theta) = 0$ if $\theta = \theta_i$ ← **What we want !!!**

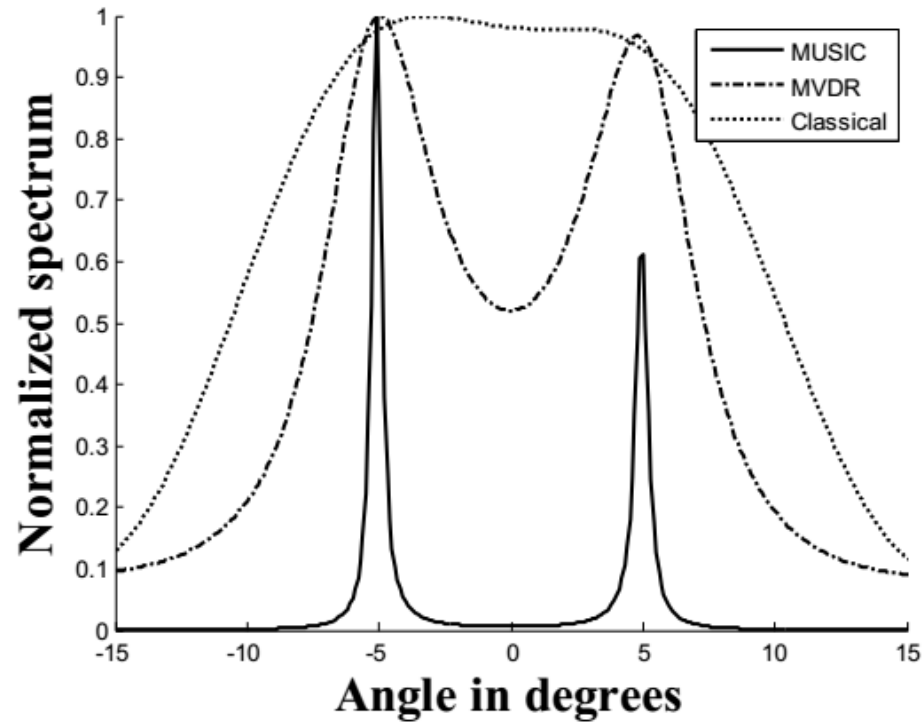
MUSIC

- $a(\theta)$ is the steering function, so it is known
- Needs to determine q_i , which needs the eigenvalue decomposition of the signal term.
- We don't know the signal term; we only know the sum of the signal term and the noise term, i.e., R_{XX}
- All of eigenvectors of the signal term are also ones for R_{XX} , and corresponding eigenvalues are added by σ^2
- Only need to find the eigenvectors of R_{XX} with eigenvalues equal to σ^2

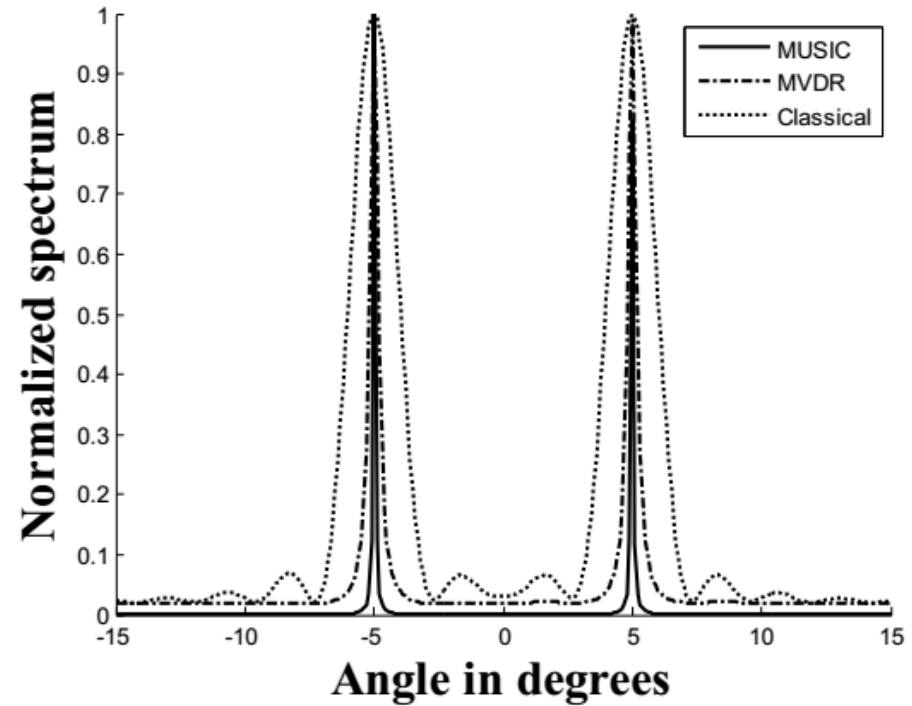
MUSIC

- Derive R_{XX}
- Perform eigenvalue decomposition on R_{XX}
- Sort eigenvectors according to their eigenvalues in descent order
- Select last $M - N$ eigenvectors q_i
- **Noise space matrix** $Q_N = [q_{M+1} \ q_{M+2} \ \dots \ q_N]$
- $Q_N^H a(\theta) = 0$ for any AoA θ_i
- Plot $p(\theta) = \frac{1}{a^H(\theta)Q_N Q_N^H a(\theta)}$ and find the peaks

Performance Comparison

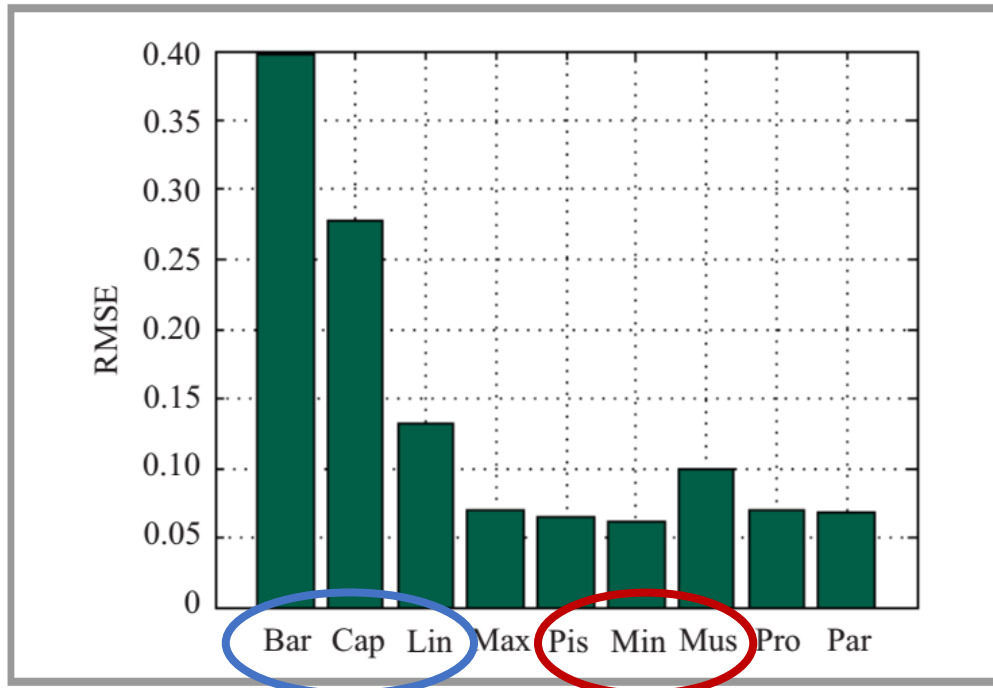


(a) 10 antennas

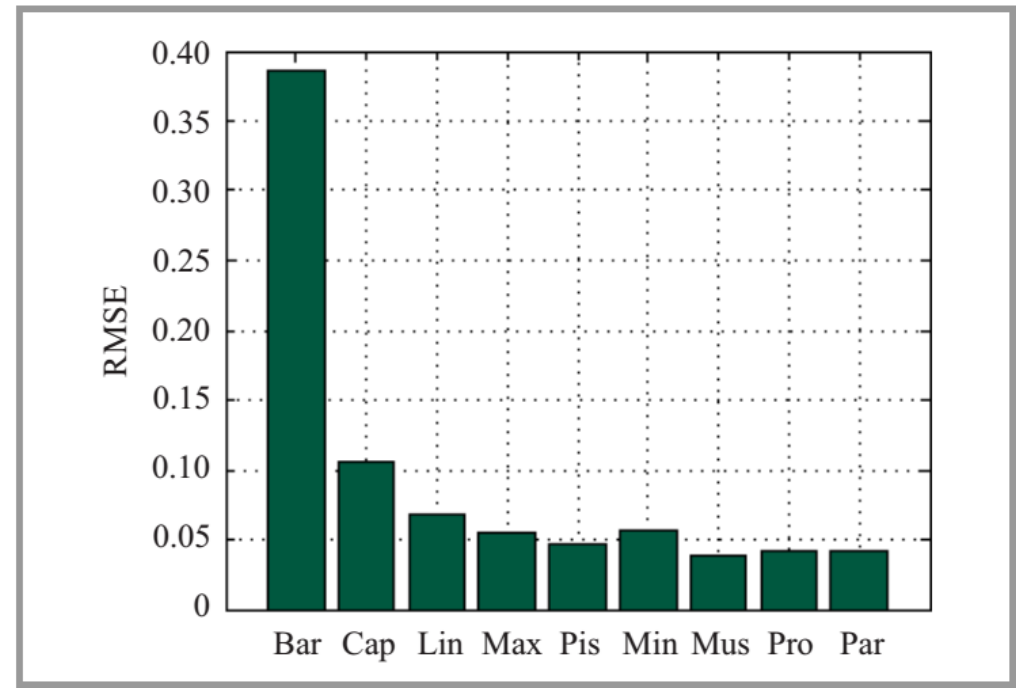


(a) 50 antennas

Performance Comparison



(a) SNR 1dB



(b) SNR 20dB

Beamforming approaches

Music variants