

# High-Resolution DOA Estimation for Contiguous Target with Large Power Difference

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Conference

## 1. Contribution

Ability to estimate Direction-of-Arrival (DoA) for contiguous targets with

- Few sensors and snapshots
- Large power difference between the targets

This is achieved by re-weighting the  $\ell_1$ -SVD with

- Capon-MUSIC, and
- Capon-MUSIC Group delay

Thus able to achieve high spatial resolution at high values of Signal-to-Interference-Noise-Ratio (SINR).

## 2. Introduction

Some common techniques for DOA estimation :

- Delay-Sum beamformer
- Capon beamformer
- MUSIC, ESPRIT and Min. Norm

The performance degrades significantly with :

- Less number of sensors and snapshots
- Large power differences between the contiguous targets.

In practical scenarios like :

- Passive SONAR, echos from the targets is several dB's lower than ships self-noise
- When one target is near and the other far away but contiguous in angular space.

## 3. Earlier Techniques

Some earlier attempts to estimate DoA's for targets with large power differences :

- Jamming method
- Constrained MUSIC and Capon-MUSIC
- Eigen-beam mCapon
- Robust sparse asymptotic minimum variance (RSAMV) algorithm
- Robust Orthogonal Projection

## 4. Signal Model

For  $J$  targets impinging upon  $N$  sensor Uniform Linear Array from  $\theta_1, \dots, \theta_J$  directions. The data received as:

$$\mathbf{y}(t) = \sum_{j=1}^J \mathbf{a}(\theta_j) s_j(t) + \mathbf{n}(t); t = 1, \dots, L \quad (1)$$

$\mathbf{a}(\theta_j) \rightarrow$  array steering vector;  $s_j(t) \rightarrow$  amplitude of  $j^{\text{th}}$  target;  $\mathbf{n}(t) \rightarrow$  Complex Gaussian noise

In matrix form Eq.1 is written as :

$$\mathbf{Y} = \mathbf{A}\mathbf{S} + \mathbf{N} \in \mathbb{C}^{N \times L}$$

Covariance matrix of  $\mathbf{y}(t)$  for  $L$  snapshots is given as:

$$\hat{\mathbf{R}} = \frac{1}{L} \sum_{t=1}^L \mathbf{y}(t)\mathbf{y}(t)^H = \frac{1}{L} \mathbf{Y}\mathbf{Y}^H$$

## 5. Proposed Method

With over-complete set of steering vectors  $\rightarrow \mathbf{A}_\Omega = \{\mathbf{a}(\theta_{\Omega,k}) | k \in [1, K]\} \in \mathbb{C}^{N \times K}$ , for  $K$  DoA bases  $\rightarrow \theta_\Omega = \{\theta_{\Omega,k} | \forall k\}$  includes the  $J$  target DoAs  $\theta$  [ $J \ll K$ ].

- Computing the SVD as  $\mathbf{Y} = \mathbf{U}\mathbf{\Lambda}\mathbf{V}$ , and Group-delay as:

$$\nabla[\angle\{\mathbf{A}_\Omega^H \mathbf{U}_N\}] = [\tau_A \ \tau_B]$$

$\mathbf{A}_\Omega = [\mathbf{A} \ \mathbf{B}]$  where  $\mathbf{A} \in \mathbb{C}^{N \times J} \rightarrow$  true DoAs and  $\mathbf{B} \in \mathbb{C}^{N \times (K-J)} \rightarrow$  other directions.

- The Group-delay weights are computed as

$$\mathbf{W}_{\text{Group-delay}}^{-1} = [[\tau_A^{(\ell_2)}]^T \ [\tau_B^{(\ell_2)}]^T]^T$$

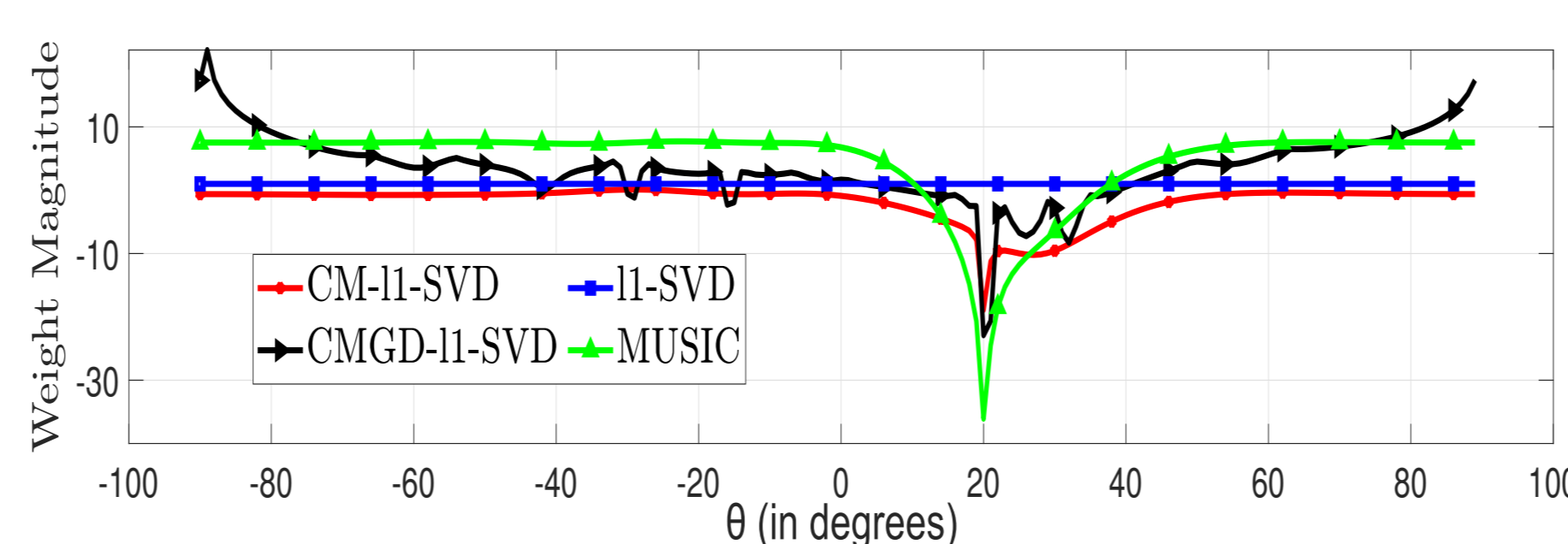


Figure 1: Angular spectrum for different weights at  $N=6$ ,  $\text{SINR}=20$  dB and targets at  $20^\circ$  &  $26^\circ$

- Instability due to Group-delay spectrum (CMGD- $\ell_1$ -SVD), seen in Fig.1.
- Compensated by Hadamard product of the Capon-MUSIC [2] spectrum and GD weights as:

$$\mathbf{W}_{\text{CMGD}} = \frac{1}{\mathbf{W}_{\text{Group-delay}} \odot \mathbf{P}_{\text{C-music}}}$$

where

$$\mathbf{P}_{\text{C-music}}(\theta) = \frac{\mathbf{a}_\theta^H \hat{\mathbf{R}}^{-1} \mathbf{a}_\theta}{\mathbf{a}_\theta^H \mathbf{U}_n \mathbf{U}_n^H \mathbf{a}_\theta}$$

- Diagonalizing weights  $\rightarrow \mathbf{W} = \text{diag}(\mathbf{W}_{\text{CMGD}})$  the minimization equation is solved as

$$\min \|\mathbf{W}\mathbf{s}^{(\ell_2)}\|_1 \text{ subject to } \|\mathbf{Y}\mathbf{V}\mathbf{E} - \mathbf{A}_\Omega \hat{\mathbf{S}}_\Omega \mathbf{V}\mathbf{E}\|_F^2 \leq \eta^2$$

$\mathbf{W}\mathbf{s}^{(\ell_2)} = \hat{\mathbf{s}}_\Omega \in \mathbb{R}^K \rightarrow$  row-wise  $\ell_2$ -norm of  $\hat{\mathbf{S}}_\Omega \in \mathbb{C}^{K \times L}$ , indicating DoAs,  $\mathbf{E} = [\mathbf{I}^J \ \mathbf{0}]^L \in \mathbb{C}^{L \times J}$  with  $\mathbf{I}^J \rightarrow$  identity matrix and  $\mathbf{0} \rightarrow (J \times (L - J))$  zero matrix, and  $\eta \rightarrow$  regularization coefficient [1].

- Another attempt by re-weighting  $\ell_1$ -SVD using Capon-MUSIC weights as:

$$\mathbf{W}_{\text{C-music}} = \frac{1}{\mathbf{P}_{\text{C-music}}}$$

## 6. Performance Evaluation

Performance evaluations settings

- Multi-snapshot processing with  $L = 200$
- Targets at  $20^\circ$  and  $26^\circ$  with step-size  $\Delta=1^\circ$
- The number of sensors  $N = 6$
- Evaluated over  $S=10^3$  Monte-Carlo simulations

### 6.1 Average Root Mean Square Error

- Capon-MUSIC Group-delay- $\ell_1$ -SVD (CMGD- $\ell_1$ -SVD) and Capon-MUSIC  $\ell_1$ -SVD (CM- $\ell_1$ -SVD) achieve low ARMSE score than state-of-the-art methods with less sensors.
- Achieves lowest ARMSE score for less snapshots for contiguous targets ( $20^\circ$  and  $26^\circ$ ) with a power difference of 20 dB,
- In terms of SINR (strong target  $\rightarrow 20$  dB, weak target  $\rightarrow [20$  to  $-5]$  dB), proposed methods achieve minimum ARMSE score.
- Target  $T_1 \rightarrow 20^\circ$  and  $T_2 \rightarrow [40^\circ - 22^\circ]$ , proposed methods achieve lowest ARMSE score.

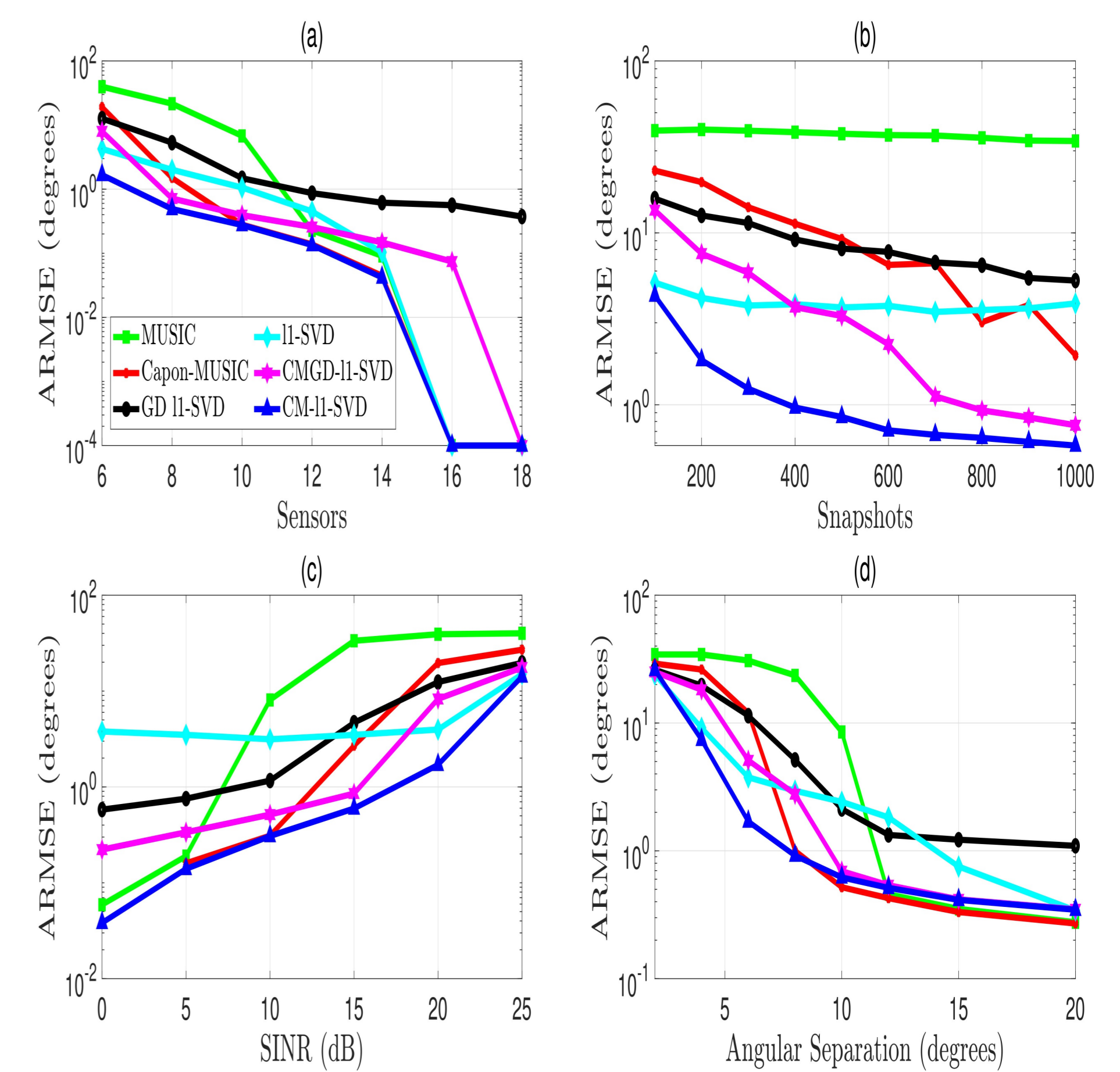


Figure 2: Comparison of proposed methods with state-of-the-art methods using (a) ARMSE Vs. No. of Sensors, (b) ARMSE Vs. Snapshots, (c) ARMSE Vs SINR and, (d) ARMSE Vs Angular Separation

### 6.2 Resolution Probability

- CM- $\ell_1$ -SVD outperforms all other algorithms in comparison.
- For angular separation  $\rightarrow 2^\circ$ , resolution probability  $\rightarrow 0.2$  achieved.
- Observed both CM- $\ell_1$ -SVD and CMGD- $\ell_1$ -SVD achieve good resolution probability in SINR ranges.

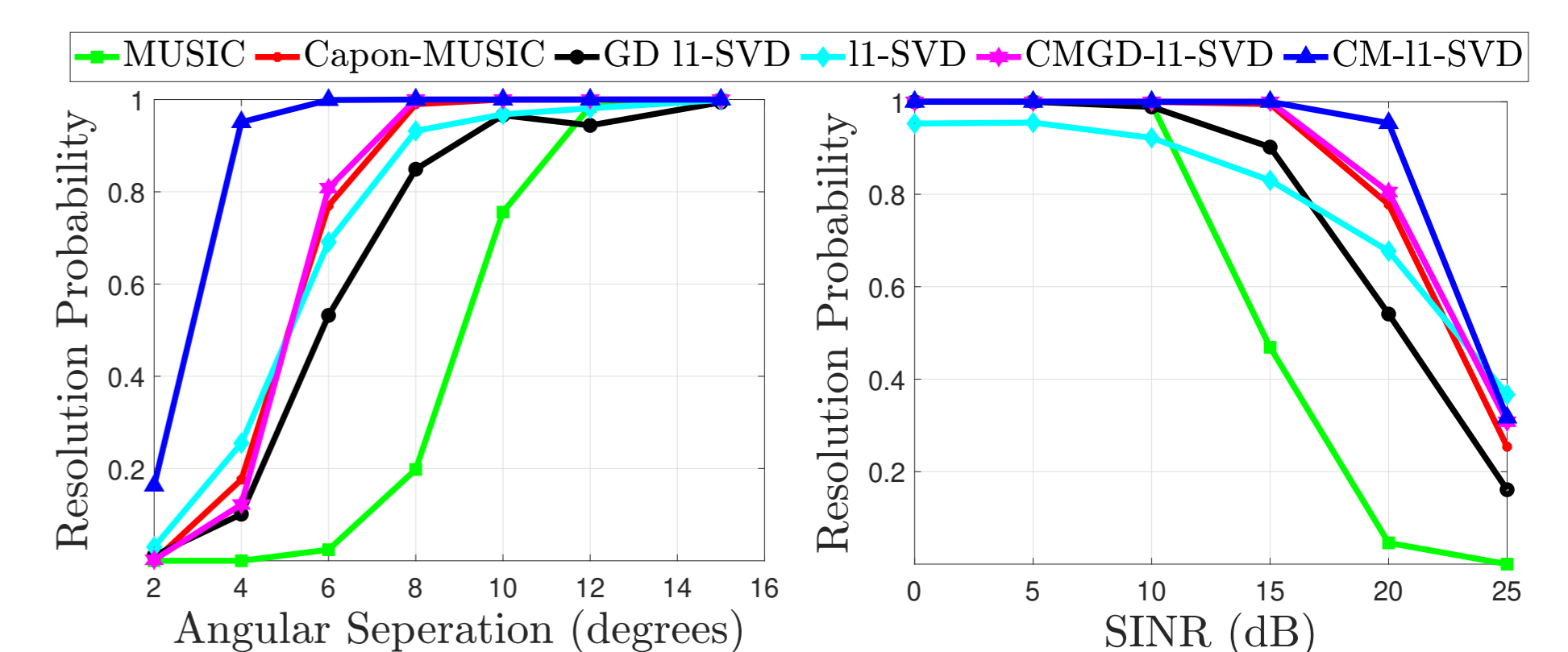


Figure 3: Plots for Resolution Probability evaluated at  $N = 6$ ,  $L = 200$ .

## References

- [1] M. Ali, A. Koul, and K. Nathwani. Significance of group delay spectrum in re-weighted sparse recovery algorithms for doa estimation. *Digital Signal Processing*, 122:103388, 2022.
- [2] Y. Gao, X. Jia, J. Xu, T. Long, and X. Xia. A novel doa estimation method for closely spaced multiple sources with large power differences. In *2015 IEEE Radar Conference (RadarCon)*, pages 1276–1279, 2015.