Grade Nested Array with Increased Degrees of Freedom for Quasi-Stationary Signals

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Abstract—In this paper a grade nested array constituted by a uniform linear array and a grade linear array with uniformly increasing inter-element is presented. The closed-form expression of the proposed array geometries and corresponding direction-of-arrival (DOA) estimation algorithm are derived. Theory analysis certifies that the proposed grade nested array can provide higher degrees of freedom (DOF) than some existing nested arrays. Some simulations are also presented to demonstrate the improved performance of the proposed nested array for DOA estimation of quasi-stationary signals.

1. INTRODUCTION

Direction-of-arrival (DOA) estimation based on array is one of the primary contents in array signal processing, and it is extensively applied in mobile communication [1] and multiple input multiple output (MIMO) radar [2, 3]. Compared with uniform arrays, sparse arrays show distinct advantage in increasing degrees of freedom (DOF) for DOA estimation of multiple signals. Many high-performance sparse arrays including minimum-redundancy array (MRA) [4], co-prime arrays [5, 6] and two-level array [7] have been proposed. In various kinds of sparse arrays, MRA can provide the largest DOF while generate the difference co-array (DCA) with consecutive virtual sensors. However, it is difficult to give the closed-form expression of the geometries of MRA as the sensor number has been given. A co-prime array consists of two uniform linear arrays, and a pair of co-prime integers is used to set the inter-element spacing of the two sub-arrays. Although co-prime array can reduce the mutual coupling between different sensors, the DOF of co-prime array is lower than MRA and two-level array.

Two-level array also consists of two uniform linear arrays, where the inter-element spacing of the second sub-array is related to the number of the elements in the first sub-array. Because of simple structure and relatively higher DOF, two-level nested array has been regarded widely, and many modified nested arrays [8–12] have been presented one after another. In [8, 9] two improved nested array configurations have been given by changing the element spacing of the second sub-array, and modified nested arrays can offer more DOF than two-level nested array [7]. A subspace extension algorithm based L-shaped nested array is presented to estimate the azimuth and elevation simultaneously in [10]. A DOA estimation algorithm with a special two-level nested array under unknown mutual coupling is proposed in [11]. It can increase DOF and improve the accuracy of DOA estimation. In addition to this, many nested arrays based fourth-order cumulants have been presented in [13–15].

The existence of periodic stationary signals [16] such as speech and audio signals is quite extensive. In [12], a new nested array for quasi-stationary signals was proposed, and it can offer more DOF than nested arrays [7–9]. However, the redundancy of this nested array is still higher because of the partial uniform structure of the second sub-array. In addition to this, the closed-form expression of corresponding DOA estimation algorithm has not been addressed.

In this letter, a grade nested array for quasi-stationary signals is proposed. Compared with a nested array [12], it has larger virtual array apertures and can offer more DOF. Besides, the closed-form expression of corresponding DOA estimation algorithm can also be given.

Notation: $[\cdot]^T$, $[\cdot]^H$, $E[\cdot]$ and $[\cdot]$ denote transpose conjugate transpose, statistical expectation and integer part, respectively. $\odot$ and $\otimes$ stand for Khatri-Rao product and Kronecker product, respectively.

2. DATA MODEL

Consider K narrowband uncorrelated quasi-stationary signals with DOA θ_k, $k = 1, 2, 3, \ldots, K$ impinging a linear array. Denote d_l, $l = 2, 3, \ldots, L$ as the distance between the l-th sensor and the first sensor, then the observation vector $x(t) = [x_1(t), x_2(t), \ldots, x_L(t)]^T \in C^{L \times 1}$ can be presented as

$$\mathbf{x}(t) = \mathbf{A}\mathbf{s}(t) + \mathbf{n}(t) \tag{1}$$

where $\mathbf{A} = [\mathbf{a}(\theta_1), \mathbf{a}(\theta_2), \ldots, \mathbf{a}(\theta_K)] \in C^{L \times K}$ is the response matrix with $\mathbf{a}(\theta_k) = [1, e^{-i\frac{2\pi}{\lambda}d_2 \sin(\theta_k)}, \ldots, e^{-i\frac{2\pi}{\lambda}d_L \sin(\theta_k)}]^T \in C^{L \times 1}$; $\mathbf{s}(t) = [s_1(t), s_2(t), \ldots, s_K(t)]^T \in C^{K \times 1}$ is the signal vector; λ is the wavelength of incident signals; and $\mathbf{n}(t)$ represents the noise vector.

Assume that noise and signals are uncorrelated and that the length of frame is T, then we can denote the covariance matrix of the q-th frame as [12, 16]

$$\begin{aligned} \mathbf{R}_q &= E\{\mathbf{x}(t)\mathbf{x}^H(t)\} \\ &= \mathbf{A}\mathbf{R}_{sq}\mathbf{A}^H + \mathbf{R}_n \in C^{L \times L}, \quad \forall t \in [(q-1)T, qT-1] \end{aligned} \tag{2}$$

for $q = 1, 2, \ldots, Q$, where Q is the number of frames.

In formula (2), $\mathbf{R}_n = E\{\mathbf{n}(t)\mathbf{n}^H(t)\}$ is the noise covariance matrix, and $\mathbf{R}_{sq} = E\{\mathbf{s}_q(t)\mathbf{s}_q^H(t)\}$ is the signal covariance matrix with the expression

$$\mathbf{R}_{sq} = \text{diag}\left\{p_{q1}^2, p_{q2}^2, \ldots, p_{qK}^2\right\} \tag{3}$$

3. KR-MUSIC ALGORITHM [16]

Denoting $\mathbf{Y} = [\mathbf{y}_1, \mathbf{y}_2, \ldots, \mathbf{y}_Q]$ and $\mathbf{P}_Q^\perp = \mathbf{I}_Q - (\mathbf{1}_Q\mathbf{1}_Q^T)/Q$, where $\mathbf{y}_q = vec(\mathbf{R}_q)$ and $\mathbf{1}_Q = [1, \ldots, 1]^T \in C^{Q \times 1}$, we can get

$$\mathbf{Y}\mathbf{P}_Q^\perp = (\mathbf{A}^* \odot \mathbf{A})(\mathbf{P}_Q^\perp \mathbf{\Psi})^T \tag{4}$$

where

$$\mathbf{\Psi} = \begin{bmatrix} p_{11}^2 & p_{12}^2 & \cdots & p_{1K}^2 \\ p_{21}^2 & p_{22}^2 & \cdots & p_{2K}^2 \\ \vdots & \vdots & \ddots & \vdots \\ p_{Q1}^2 & p_{Q2}^2 & \cdots & p_{QK}^2 \end{bmatrix} \tag{5}$$

Performing singular value decomposition (SVD) of $\mathbf{Y}\mathbf{P}_Q^\perp$, we can get the signal subspace, and MUSIC algorithm [17] can be used to estimate the DOA. More details of the KR-MUSIC and the dimension reduction KR-MUSIC algorithms can be found in [16]. To save space, the detailed procedure is not presented throughout this paper.

4. GRADE NESTED ARRAY

The proposed grade nested array consists of two linear arrays shown in Fig. 1. The first sub-array is a uniform linear array, and the inter-element spacing is d. The second sub-array is a grade array with continuously incremental inter-element spacing from Dd to $(D + M - 2)d$. The set of sensor locations of the two sub-arrays can be expressed as
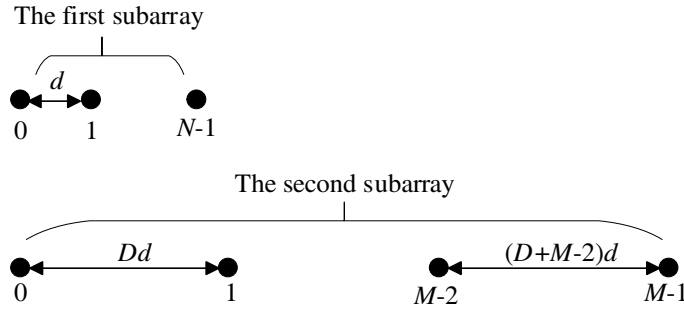
$$S_1 = \{nd \,|\, n = 0, 1, \ldots, N - 1\} \tag{6}$$

Figure 1. The configuration of grade nested array.

and

$$S_2 = \left\{ \frac{(m+1)m}{2} Dd \,|\, m = 0, 1, \ldots, M-1 \right\} \tag{7}$$

where $d \leq \lambda/2$ and D is a positive integer.

The difference co-array [12] set can be expressed as

$$D_c = D_{c11} \cup D_{c12} \cup D_{c22} \tag{8}$$

where D_{c11} and D_{c22} are two self-difference sets, and D_{c12} is the cross-difference set.

As [12], we denote the number of unduplicated elements in D_c as DOF of the array. According to Eqs. (6) and (7), D_{c11}, D_{c22} and D_{c12} can be written as

$$\begin{cases} D_{c11} = \{\pm(n_1 d - n_2 d)\} \\ D_{c12} = \left\{ \pm(nd - \dfrac{(m+1)m}{2} Dd) \right\} \\ D_{c22} = \left\{ \pm \left(\dfrac{(m_1+1)m_1}{2} Dd - \dfrac{(m_2+1)m_2}{2} Dd \right) \right\} \end{cases} \tag{9}$$

where n_1, n_2, $n = 0, 1, \ldots, N-1$ and m_1, m_2, $m = 0, 1, \ldots, M-1$.

In order to reduce the number of repetitive elements between D_{c12} and D_{c22}, we let

$$\begin{cases} D \geq N + M^2/4 - 3M/2 + 1, & M \text{ is even} \\ D \geq N + M^2/4 - M + 3/4, & M \text{ is odd} \end{cases} \tag{10}$$

Assume that the number of array elements is $L(L \geq 5)$. Because the first sensor of the first sub-array has the same location as the first sensor of the second sub-array, we let $N = [L/2]$ and $M = L - N + 1$. When L is even, we have $N = L/2$ and $M = L/2 + 1$. The number of nonnegative elements in D_{c11} is $L/2$. Since $0 \in D_{c11}$, we only need to know the positive elements of D_{c12} and D_{c22}. It is easy to know that the number of positive elements being different from D_{c11} in D_{c12} is $L^2/4$. According to Eq. (10), removing the same elements with D_{c12}, the number of positive elements in D_{c22} is $(L^2-2L)/8$. When L is odd, we have $N = (L-1)/2$ and $M = (L+1)/2 + 1$. The number of nonnegative elements in D_{c11} is $(L-1)/2$, and the positive elements being different from D_{c11} in D_{c12} are $(L^2-1)/4$. According to Eq. (10), except the same elements with D_{c12}, the number of positive elements in D_{c22} is $(L^2-1)/8$.

From the analysis above, combining the symmetry of D_{c11}, D_{c22}, D_{c12}, we can get the DOF of proposed array as

$$\text{DOF} = \begin{cases} \dfrac{3L^2 + 2L}{4} - 1, & L \text{ is even} \\ \dfrac{3L^2 + 4L - 7}{4} - 1, & L \text{ is odd} \end{cases} \tag{11}$$

For the two-level nested arrays [7], the second sub-array is also a uniform array, which limits the number of elements in self-difference set of the second sub-array. For the improved nested arrays [8, 12],

the second sub-array is replaced by a uniform linear array and an isolated sensor. In [9], the second sub-array is replaced by two different uniform linear arrays. However, the DOF of the nested arrays [8, 9, 12] is still affected by the partial uniform structure of the second sub-array.

Take 7-element array as an example to compare the DOF of different nested arrays. Since the DOF of array [9] is lower than the nested array [8], and the configurations of two nested arrays are similar, we only compare the proposed array with a two-level nested array [7], Yang's array [8] and Huang's array [12]. For the proposed array, we let $D = 5$, $N = 3$ and $M = 5$. Fig. 2 shows the configurations of four nested arrays.
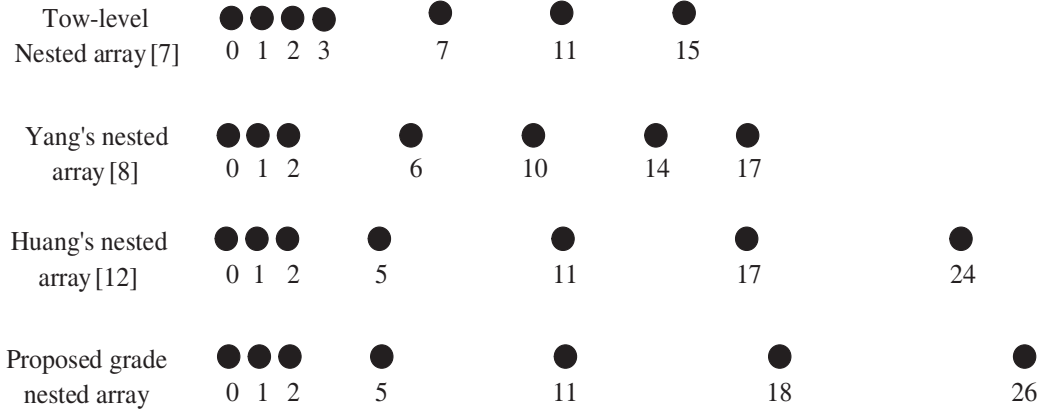

Figure 2. The configurations of four 7-element nested arrays.

The difference co-array set of sensor locations for two-level nested array [7], Yang's array [8], Huang's array [12] and proposed grade nested array are $D_{1c} = \{-15, -14, -13, -12, -11, -10, -9, -8, -7, -6, -5, -4, -3, -2, -1, 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15\}d$, $D_{2c} = \{-17, -16, -15, -14, -13, -12, -11, -10, -9, -8, -7, -6, -5, -4, -3, -2, -1, 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17\}d$, $D_{3c} = \{-24, -23, -22, -19, -17, -16, -15, -13, -12, -11, -10, -9, -7, -6, -5, -4, -3, -2, -1, 0, 1, 2, 3, 4, 5, 6, 7, 9, 10, 11, 12, 13, 15, 16, 17, 19, 22, 23, 24\}d$, and $D_{4c} = \{-26, -25, -24, -21, -18, -17, -16, -15, -13, -11, -10, -9, -8, -7, -6, -5, -4, -3, -2, -1, 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 13, 15, 16, 17, 18, 21, 24, 25, 26\}d$, respectively. The DOF of the four arrays are 31, 35, 39, 41, respectively.

We must acknowledge that there are "holes" in the virtual array for the proposed nested array and Huang's nested array [12]. We know that "holes" may bring about angle ambiguity. However, when the number of consecutive virtual sensors is large enough, the "holes" will not affect the performance of DOA estimation. More relevant details are described in [12]. In addition, from the above example, we can also find that the number of consecutive virtual sensors of the proposed array is larger than the nested array [12].

5. DIMENSION REDUCTION KR-MUSIC FOR PROPOSED ARRAY

In this section, we derive the closed-form expression of DOA estimation algorithm based on the characteristic of proposed grade nested array. We construct a selection matrix $G \in C^{F \times L^2}$ with the expression

$$\mathbf{G} = \left[\begin{array}{cccccccc} \mathbf{G}_{M-1}^T & \mathbf{G}_{M-2}^T & \cdots & \mathbf{G}_0^T & \bar{\mathbf{G}}_0^T & \cdots & \bar{\mathbf{G}}_{M-2}^T & \bar{\mathbf{G}}_{M-1}^T \end{array}\right]^T \tag{12}$$

where F is the DOF of the proposed array.

In formula (12), $\mathbf{G}_m$ and $\bar{\mathbf{G}}_m$ can be written as

$$\mathbf{G}_m = \begin{cases} [\mathbf{0}_{(N+m-1) \times (N+m-1)L}, \mathbf{I}_{N+m-1}, \mathbf{0}_{(N+m-1) \times (L^2-(N+m-1)(L+1))}], & m = 1, \ldots, M-1 \\ [\mathbf{0}_{N \times (N-1)L}, \mathbf{I}_N, \mathbf{0}_{N \times (L^2-(N-1)L-N)}], & m = 0 \end{cases} \tag{13}$$

and

$$\bar{\mathbf{G}}_m = \left[\begin{array}{cc} \mathbf{J}_{N+m-1} \otimes \mathbf{e}_{N+m} & \mathbf{0}_{(N+m-1) \times [L^2-(N+m-1)L]} \end{array}\right], m = 0, \ldots, M-1 \tag{14}$$

where $e_{N+m} = [\ 0 \quad \ldots \quad 1 \quad \ldots \quad 0\ ] \in C^{1 \times L}$ is a vector with one on the $(N+m)$th component and zero for other components; $\mathbf{I}_{N+m-1}$ indicates an $N + m - 1$ order identity matrix; and $\mathbf{J}_{N+m-1}$ denotes a matrix with one on back-diagonal and zero for other elements.

According to Eq. (4), we denote $\bar{\mathbf{Y}} \in^{F \times Q}$ as

$$\bar{\mathbf{Y}} = \mathbf{G}\mathbf{Y}\mathbf{P}_Q^{\perp} = \mathbf{G}(\mathbf{A}^* \odot \mathbf{A})(\mathbf{P}_Q^{\perp}\mathbf{\Psi})^T \tag{15}$$

As [16], SVD of $\bar{\mathbf{Y}}$ can get the noise subspace $\mathbf{U}_n$, where $\mathbf{U}_n$ is a matrix consisting of the left singular vectors of the smallest $F - k$ singular values. Minimizing cost function

$$f(\theta) = \frac{1}{(\mathbf{G}(\mathbf{a}^*(\theta) \otimes \mathbf{a}(\theta)))^H \mathbf{U}_n \mathbf{U}_n^H \mathbf{G}(\mathbf{a}^*(\theta) \otimes \mathbf{a}(\theta))} \tag{16}$$

by multiple signal classification (MUSIC) algorithm [17], we can get the DOA estimation of all signals.

In fact, the configuration of proposed array has something in common with the other four nested arrays [7–9, 12]. All the five arrays consist of a uniform array with small inter-element spacing and a sparse array with large inter-element spacing. Meanwhile, all the difference co-array sets of five arrays consist of two self-difference sets and a cross-difference set. So, a similar method can also be given for the other four arrays.

6. SIMULATION

In order to prove the improved performance of proposed grade nested array, we give three sets of simulation results. The nested array [8] has clear advantage over two-level nested array [7] in DOF and the performance of DOA estimation, which has been proved in [8] and [12]. Hence, we only compare the nested array [8] and nested array [12] with the proposed grade nested array.

First, we compare the DOF of the three nested arrays. DOF of three nested arrays versus the number of sensors is provided in Fig. 3. Obviously, the proposed grade nested array can offer more DOF than other two nested arrays.
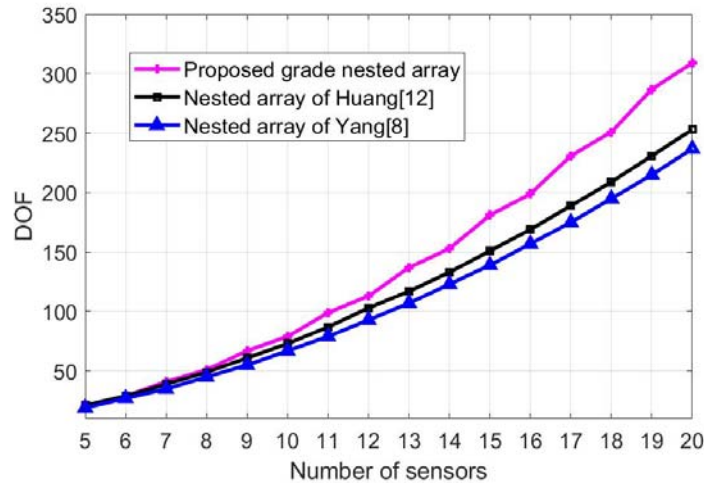

Figure 3. DOF against the number of sensors.

Second, we compare the MUSIC spectra of dimension reduction KR-MUSIC for three nested arrays. 11 narrowband uncorrelated quasi-stationary signals come from the directions $[-50°, -40°, -30°, -20°, -10°, 0°, 10°, 20°, 30°, 40°, 50°]$. Fix the signal to noise ratio (SNR) at $-5\,\text{dB}$, and Fig. 4 shows the MUSIC spectra of three arrays with $T = 500$, $Q = 40$ and $L = 7$. Because of larger virtual aperture, the proposed nested array can provide sharper spectral peak. From Fig. 4, we can find that the proposed nested array can provide higher resolution than other two nested arrays.
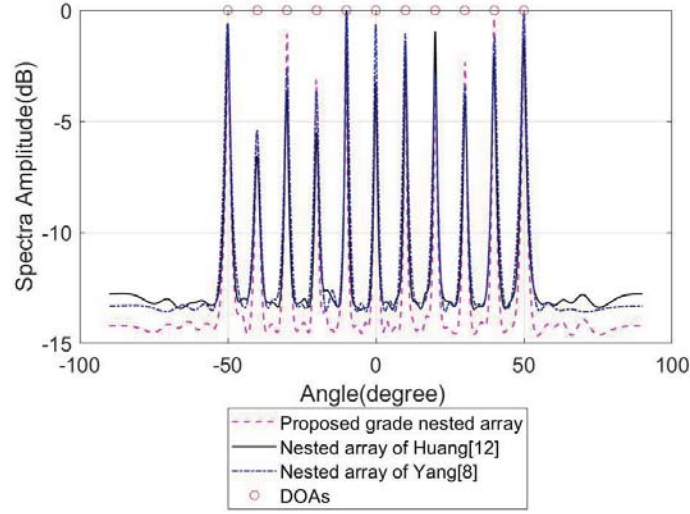
Figure 4. MUSIC spectra of three nested arrays.

At last, we compare the RMSE of dimension reduction KR-MUSIC for three nested arrays. The root mean square error (RMSE) is expressed as

$$\text{RMSE} = \sqrt{\frac{1}{KJ}\sum_{j=1}^{J}\sum_{k=1}^{K}(\hat{\theta}_{kj} - \theta_k)^2} \tag{17}$$

where $\hat{\theta}_{kj}$ is the estimation in the jth experiment for the kth signal, and $J = 200$ is the number of experiments.

The DOAs of six narrowband uncorrelated quasi-stationary signals are $[-50°, -40°, -30°, -20°, 30°, 40°]$. Let $T = 500$, $Q = 40$ and $L = 7$, and the RMSEs with respect to the SNR are described jointly in Fig. 5. Fig. 6 describes the RMSEs versus the number of snapshots with SNR $= 10$ dB, $Q = 40$ and $L = 7$. Fig. 7 shows the RMSEs versus the number of frames with SNR $= 10$ dB, $T = 500$ and $L = 7$. It can be indicated from the three figures that the estimation precision of proposed nested array is higher than other two nested arrays in different situations. But we must point out that the performance of three arrays degrade sharply when the number of frames is less than 25, which can be seen in Fig. 7.
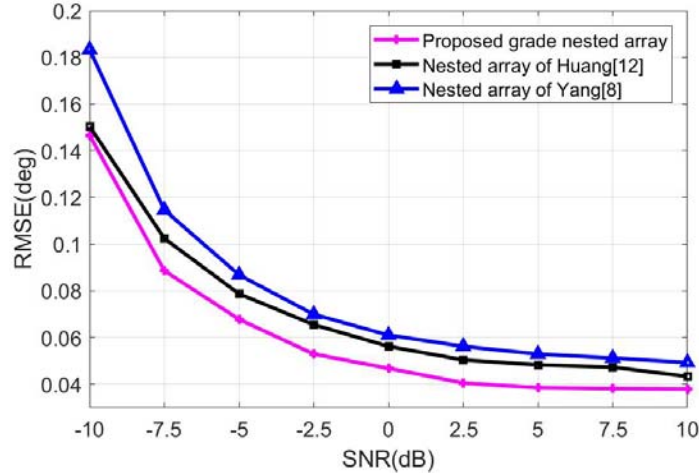

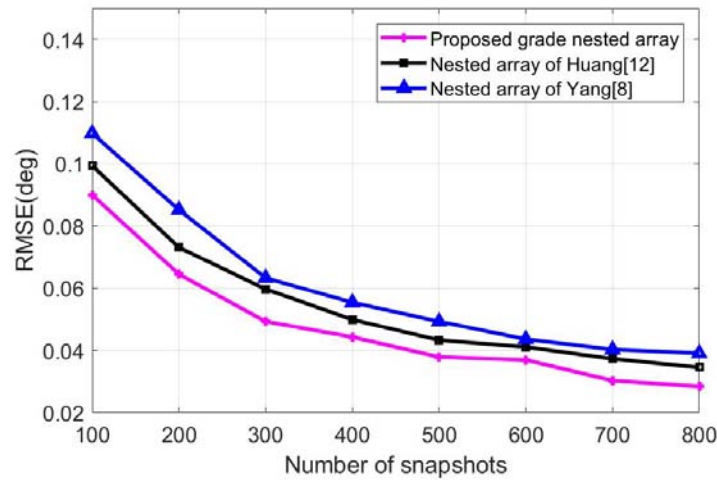**Figure 5.** RMSE against SNR for three nested arrays.

Figure 6. RMSE against the number of snapshots for three nested arrays.
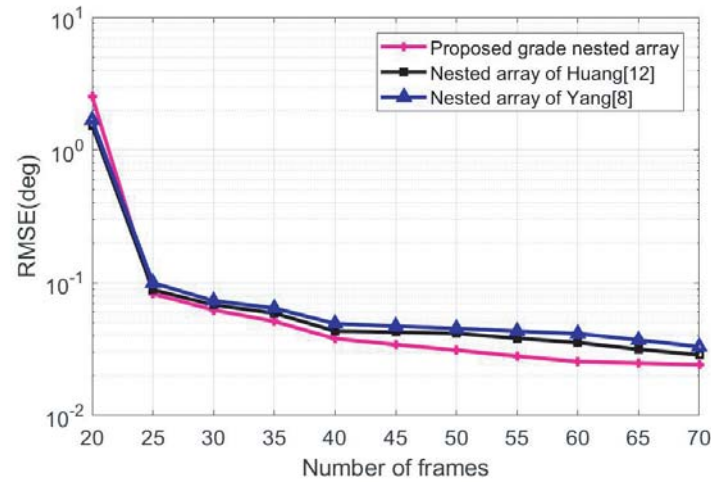

Figure 7. RMSE against the number of frames for three nested arrays.

7. CONCLUSION

We present a grade nested array for quasi-stationary signals. The closed-form expression of array geometries and corresponding DOA estimation algorithm are obtained. Because of the reduction of redundancy, it can offer more DOF than many pre-existing nested arrays. Simulation results testify the improved performance of proposed nested array in DOA estimation for quasi-stationary signals.

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