DOA Estimation with Non-Uniform Linear Arrays: A Phase-Difference Projection Approach

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Abstract—Phase wrapping is a major problem in direction-of-arrival (DOA) estimation using phase-difference observations. For an antenna pair with an inter-antenna spacing greater than half of the wavelength ($\lambda/2$) of the signal, phase wrapping occurs at certain DOA angles leading to phase-difference ambiguities. Existing phase unwrapping methods exploit either frequency or spatial diversity. These techniques work by imposing restrictions on the utilized frequencies or the receiver array geometry. In addition to these restrictions, sensitivity to noise and calibration errors is another limitation of these methods. We propose a grid-less \textit{phase-difference projection (PDP) DOA algorithm to overcome these issues. The concept of \textit{wrapped phase-difference pattern (WPDP)} is introduced, which allows us to compute most of the parameters required for DOA estimation in an offline manner. This results in a superior computational speed in real-time compared to methods with similar DOA estimation performance. Simulation results demonstrate the excellent performance of the proposed algorithm, both in terms of accuracy and speed.

Index Terms—Direction of arrival, DOA, phase-difference, phase wrapping, non-uniform linear arrays, CRLB.

I. INTRODUCTION

Direction-of-arrival (DOA) estimation is an important topic for applications such as wireless sensor networks [1], indoor positioning and tracking [2], Radar [3], wireless communications [4], and so on. Many DOA estimation methods have been proposed over the years [5], [6], focusing largely on uniform linear array (ULA) configurations. On the other hand, the use of non-uniform linear arrays (NULAs) is also popular. NULA configurations are often utilized to extend the array aperture and consequently improve the array’s DOA resolution.

In this paper, we focus on single-source DOA estimation with NULAs. This scenario is motivated by a mmWave/THz multiple-input multiple-output (MIMO) communication context. In these systems, flexible arrays are adopted to alleviate the high computation and hardware costs (see switch-based MIMO [7] and array-of-subarray (AOSA) structures [8]). When an antenna/subarray selection algorithm is applied, a NULA structure will be formed. In addition to the NULA structure, the analog beamforming employed in these systems can reduce multipath and provide a dominant line-of-sight (LOS) signal [9]. Hence, DOA estimation for a single source observed at a NULA is an important problem for mmWave/THz MIMO systems.

The DOA estimation problem can be formulated as an optimization of a cost function over a feasible DOA range. Usually, the process requires evaluating the cost function for the whole DOA range, searching for that function’s optimum. Maximum likelihood estimation (MLE) [10] and MUSIC [11] are two widely used methods that exemplify this approach. A drawback of this approach is that the search process can increase the computational complexity, especially when high spatial resolution is desired.

Time-delay estimation is a fast alternative solution to DOA estimation that can produce a DOA estimate directly without applying a grid search. [12]. The linear relationship between time delay and phase-difference makes it possible to utilize phase-difference measurements for DOA estimation. Phase-difference based DOA estimation has been reported as an effective approach for multi-carrier signals [13], [14]. Nevertheless, phase-difference based DOA estimation suffers from the occurrence of \textit{phase wrapping} [15], [16].

The issue of phase wrapping can be resolved by exploiting the frequency diversity available in multi-frequency signals, or by leveraging spatial diversity in single-frequency scenarios [17]. Examples of spatial-diversity phase unwrapping methods include [15], and more recent off-grid approaches such as 2Q-order difference-set [18] and two-step offset correction [19]. A major drawback of these methods is that they require a specialized antenna setup. Besides, these methods tend to be sensitive to the phase noise effect.

This paper proposes a \textit{phase-difference projection (PDP) method for DOA estimation using non-uniform linear arrays. We capitalize on a novel concept of a \textit{wrapped phase-difference pattern (WPDP). The proposed method can be applied to an arbitrary linear array configuration of three or more sensors. Simulation results demonstrate that the proposed method offers a good trade-off between computational complexity and DOA estimation performance.}

We proceed by presenting the observation model in Section II and the proposed PDP approach for DOA estimation in Section III. Simulation results are presented in Section IV before drawing the conclusion of the paper in Section V.

II. OBSERVATION MODEL

We consider a complex sinusoidal source signal, with a frequency $f$ and amplitude $A$, $s(t) = Ae^{-j2\pi ft}$ in the \textit{far field} [20] of a non-uniform linear array of $N$ antennas. The source impinges on the array from a direction $\theta \in [-\pi/2, \pi/2]$ rad. Let $d_{uv}$ denote the distance between a pair of the array antennas $(u$ and $v)$ normalized by $\lambda/2$, where $\lambda$ is the signal wavelength. The received signal (vector) at time $t$ can be modeled as [21]

$$x(t) = a(\theta)s(t) + w(t),$$

(1)
where \( a(\theta) = [1, e^{-j\pi d_1 \sin(\theta)}, ..., e^{-j\pi d_N \sin(\theta)}]^T \) is the array steering vector, \((\cdot)^T\) indicates the transpose operation, and \( w(t) \) is a vector of additive white Gaussian noise (AWGN).

The wrapped phase-difference (WPD) across an antenna pair, \( u \) and \( v \), can be estimated from the \( u \)-th and \( v \)-th elements of \( x \) as

\[
\psi_{uv}(t) = \text{angle}(x_u(t) \cdot x_v^*(t)) \in [-\pi, \pi),
\]

where \((\cdot)^*\) is the complex conjugate operator. For simplicity, and without loss of generality, we will focus on single-shot scenarios. Hence, we will drop the time variable \( t \).

To develop our proposed method, we start from noise-free WPD observations, \( \psi_{uv}(\theta) \). These are related to the actual phase-difference, \( \phi_{uv}(\theta) = \pi d_{uv} \sin(\theta) \) through

\[
\psi_{uv}(\theta) = \text{mod}(\phi_{uv}(\theta) + \pi, 2\pi) - \pi = \pi d_{uv} \sin(\theta) - 2\pi q_{uv},
\]

where \( \text{mod}(\cdot, \cdot) \) is the modulus operation. The value of \( q_{uv} \) can be obtained as

\[
q_{uv} = \text{round}\left(\frac{\pi d_{uv} \sin(\theta)}{2\pi}\right),
\]

where \( \text{round}(\cdot) \) is the integer rounding operation.

Based on (3), we observe that estimating the DOA from \( \psi_{uv}(\theta) \) requires knowledge of the integer \( q_{uv} \), which is not available since we use (2) to estimate \( \psi_{uv}(\theta) \). When \( d_{uv} \leq 1 \), \( q_{uv} = 0 \) for any \( \theta \). However, for \( d_{uv} > 1 \), the latter result is not guaranteed, except for a specific range of \( \theta \) values. Since \( \theta \) is unknown, \( \psi_{uv}(\theta) \) will always be ambiguous for \( d_{uv} > 1 \), which is the case for most of the antenna pairs in a NULA.

III. THE PROPOSED PDP ALGORITHM

A. Wrapped Phase-Difference Pattern (WPD)

For an arbitrary source location \( \theta \in [-\pi/2, \pi/2] \), using (3), we can compute the WPD across receiver pairs to create a WPD vector \( \psi(\theta) = [\psi_{uv}(\theta)]^T \), \( u, v \in \{1, \ldots, N\} \), \( u < v \). Assuming that we utilize \( M \leq \binom{N}{2} \) antenna pairs, we can simplify the notations and write \( \psi(\theta) = [\psi_m(\theta)]^T \), and \( q(\theta) = [q_m(\theta)]^T \), \( m = 1, \ldots, M \). We can also arrange the inter-antenna distances that correspond to \( \psi(\theta) \) in a vector \( d = [d_m]^T \).

Now, let us think of \( \psi(\theta) \) as a point in an \( M \)-dimensional space. From (3), and for \( \theta = 0 \), we can see that \( \psi_m(\theta) = 0, q_m(\theta) = 0 \), \( m = 1, \ldots, M \). By gradually increasing \( \theta \) starting from \( \theta = 0 \), we can see that all \( \psi_m(\theta) \) increase linearly with \( \sin(\theta) \). The entries of the vector \( q(\theta) \) remain constant (all zeros) up to a certain \( \theta \) value at which the entry corresponding to the largest inter-antenna spacing will have an increment of +1. Then, again, \( q(\theta) \) will remain constant until another entry changes its value. The elements of \( q(\theta) \) will successively change their value until we reach \( \theta = \pi/2 \). A similar phenomenon is observed when \( \theta \) is varied in the negative direction starting from zero—the entries of \( q(\theta) \) successively be incremented by -1. This process creates different intervals of \( \theta \), each interval with a distinct vector \( q \) that remains unchanged throughout that interval. Let us denote these intervals as \( \Theta_k, k \in \{1, \ldots, K\} \). For any \( \theta_a, \theta_b \in \Theta_k \),

\[
q(\theta_a) = q(\theta_b) = q_k.
\]

Based on (3) and (5), we can write

\[
\delta_{ab} = \psi(\theta_a) - \psi(\theta_b) = \pi d [\sin(\theta_a) - \sin(\theta_b)],
\]

which indicates that each continuum given by \( \psi(\Theta_k), k \in \{1, \ldots, K\} \), is a straight line; and that all the \( K \) straight lines point in the same direction \( d/||d||_2 \), where \( || \cdot ||_2 \) is the Euclidean norm. That is, we have \( K \) parallel lines in \( M \)-dimensional space, with \( K \) given by

\[
K = 2^M \sum_{i=1}^{M} \text{ceil}\left(\frac{\psi_i(\pi/2) - \pi}{2\pi}\right) + 1,
\]

where \( \text{ceil}(\cdot) \) returns the nearest integer greater than or equal to the argument. These \( K \) lines result from abrupt changes in the linear relationship between the entries of the vector \( \psi(\theta) \) that occur when an entry of \( q(\theta) \) changes its value. We refer a plot of \( \psi(\theta) \) as a wrapped phase-difference pattern (WPD).

An illustration of a WPD for an array of 3 elements is depicted in Fig. 1. We use \( M = 2 \) and \( \psi(\theta) = [\psi_{12}(\theta), \psi_{23}(\theta)]^T \). The inter-antenna spacing vector is \( r = [0, 2.3, 5.18] \) (relative to antenna-1). We can see \( K = 5 \) WPD lines displayed together with the corresponding projection points, \( p_k, k = 1, \ldots, 5 \) (will be discussed shortly). These five lines represent the relationship between the entries of \( \psi(\theta) \) as \( \theta \) changes (see (3)). Sample \( \theta \) values (in degrees) are indicated.

![Fig. 1. An example of a WPD for \( r = [0, 2.3, 5.18] \), \( M = 2 \).](image-url)

Using simple geometry, we can see that all the WPD lines \( \psi(\Theta_k) \) (dotted lines) are perpendicular to a hyperplane (solid green line in this 2-dimensional WPD) that contains the origin. This hyperplane is formed by the points \( \psi \) that satisfy

\[
d^T \psi = d_1 \psi_1 + d_2 \psi_2 + \cdots + d_M \psi_M = 0.
\]

Each line \( \psi(\Theta_k) \) has a projection point, \( p_k \), at which the line intersects with the hyperplane. In the following discussion, we show how to compute these projection points.

Given a noise-free WPD vector \( \psi(\theta) \), the distance between this vector, treated as a point in \( M \)-dimensional space, and the hyperplane (8), is given by

\[
\text{dist} (\psi(\theta), d^T \psi = 0) = \frac{d^T \psi(\theta)}{||d||_2}.
\]

The projection point \( p(\theta) \) of \( \psi(\theta) \) on the hyperplane along the direction \( d/||d||_2 \) can be obtained as

\[
p(\theta) = \text{proj}(\psi(\theta)) = \psi(\theta) - \frac{d^T \psi(\theta)}{||d||_2} \frac{d}{||d||_2}.
\]
It is obvious that all points on the same WPD line are projected on the same point in the projection hyperplane. However, for an observed noisy wrapped phase-difference vector $\hat{\psi}$, (10) returns a perturbed projection point, possibly $\hat{p} \notin \{p_k\}$. In this case, we pick the nearest projection point $p_z$, where

$$z = \arg\min_k ||p_k - \hat{p}||_2.$$  \hspace{1cm} (11)

Now, the unbiased WPD $\tilde{\psi}$, which is the nearest point on the line with the projection point $p_k$ can be estimated as

$$\tilde{\psi} = p_z + \frac{d^T \hat{\psi}}{||d||_2} \cdot \frac{d}{||d||_2}. \hspace{1cm} (12)$$

Based on (5), all points on the $k$-th wrapped phase-difference (WPD) line can be compensated/unwrapped with the same unwrapping vector $h_k = 2\pi q_k$, i.e., $\phi(\theta) = \psi(\theta) + h_k$, $\theta \in \Theta_k$. Hence, the estimated unwrapped phase difference can be obtained as

$$\hat{\phi} = \tilde{\psi} + h_z. \hspace{1cm} (13)$$

A procedure to compute the unwrapping vector $h_k$ and the projection points $p_k$ will be detailed in the next subsection.

**B. Computing the Unwrapping Vectors and Projection Points**

The unwrapping vector $h_k$ can be obtained by tracing the WPD lines. Together with their projection points, these lines are easily identified by their (known) direction unit vectors and starting points. We can start from the point $\psi(-\pi/2)$, which, let us say, falls on the first line. The point where this line intersects with the M-cube whose boundaries are $-\pi$ and $\pi$ can easily be calculated. The intersection point determines the next line’s starting point, which is obtained by wrapping the coordinate of $\psi$ that crosses the cube’s surface. A pseudocode for calculating $p_k$ and $h_k$ is listed in Algorithm 1, where $\hat{\phi}_0$, $\psi_0$, and $h_0$ are intermediate variables. The procedure in Algorithm 1 is performed completely offline, which reduces the online complexity of the proposed algorithm.

**Algorithm 1 – Calculate $h_k$, $p_k$**

1. $\hat{\phi}_0 \leftarrow \pi d \sin(-\pi/2)$, $\phi_{max} \leftarrow \pi d \sin(\pi/2)$
2. $j \leftarrow 1$, $h_0 \leftarrow \phi_0 - \text{wrap}(\phi_0)$
3. while $\phi_0(m) < \phi_{max}(m)$ for all $m = [1, 2, ..., M]$ do
4. $h_j \leftarrow h_0$
5. $p_j \leftarrow \text{project}(\psi_0)$
6. $j \leftarrow j + 1$
7. $i \leftarrow \arg\min_k [||\pi - \psi_c(i)||/d_i]$
8. $\psi_0 \leftarrow d(\pi - \psi_0(i))/(d_i + \psi_0)$
9. $\psi_0(i) \leftarrow \psi_0(i) - 2\pi$
10. $h_0(i) \leftarrow h_0(i) + 2\pi$
11. $\phi_0 = \psi_0 + h_0$
12. return $h_k, p_k(k \in \{1, 2, ..., K = j - 1\})$

**C. PDP DOA Estimation Algorithm**

Given a noisy WPD $\hat{\psi}$, the estimated projection point $\hat{p}$ can be computed using (10) and (11). Then, the unbiased WPD $\tilde{\psi}$ can be obtained using (12). Next, the estimated unwrapped phase-difference vector $\hat{\phi}$ can be obtained using (13). Finally, the DOA of the source can be calculated using (3). A pseudocode of the proposed algorithm is listed in Algorithm 2 (Matlab codes available in Github).

**Remark 1:** In Algorithm 2, we note that the bulk of the computational complexity lies in Step-1. This step needs to be performed only once at the initial setup (offline). The rest of the algorithm’s (online) steps involve simple computations. This, along with its grid-less nature, greatly enhances the online computational complexity of the proposed algorithm.

**Algorithm 2 – PDP DOA Estimation Algorithm**

1. Initialize $h_1, h_K, p_1 - p_K$ using Algorithm 1
2. $\psi \leftarrow \text{angle}(x_u(t_0) \cdot x_u^*(t_0))$ for selected pairs (2)
3. $\tilde{p} \leftarrow \psi - d^T \hat{\psi} \frac{d}{||d||}$ (10)
4. $z \leftarrow \arg\min_k ||p_k - \tilde{p}||$ (11)
5. $\hat{\psi} \leftarrow p_z + \frac{d^T \hat{\psi}}{||d||} \cdot \frac{d}{||d||}$ (12)
6. $\hat{\phi} \leftarrow \tilde{\psi} + h_z$ (13)
7. $\hat{\theta} \leftarrow \sin^{-1}\left(\frac{\hat{\phi}}{\pi d}\right)$ (3)
8. return $\hat{\theta}$

Fig. 2. Visualization of two different WPDPs.

**Remark 2:** Noise perturbation in WPD tends to drift a noisy WPD point away from its original (noise-free) WPD line. The proposed PDP algorithm associates a noisy point with the closest projection point. Depending on the noise level, and the distance between the projection points, this might lead to a wrong (hard) decision and erroneous DOA estimation. For a specific signal frequency, the distance between the projection points is solely determined by the array layout, which can easily be deduced from (10). We illustrate the impact of array configuration in Fig. 2, which shows WPDP examples for two different array configurations with the same number of elements. From the figure, we can see a significant difference in WPDP structures for the two arrays, as reflected in the number of projection points and the inter-point distances. As an example, it is expected that recovering a projection point correctly from noisy observations to be easier for $p_4$ in Fig. 2 (b) compared to the rest of the projection points in both Fig. 2 (a) and (b). On the other hand, in situations where two WPD lines overlap, the proposed algorithm, or any other DOA...
estimation algorithm, will fail to identify the source location. This happens when the array configuration is ambiguous [22].

IV. PERFORMANCE EVALUATION

In this section, we evaluate the performance of the proposed PDP algorithm given in Algorithm 2 along with several benchmark methods. The benchmark methods are the 2Q-order algorithm [18], the two-step offset correction method [19], the expectation maximization ESPRIT (EM-ESPRIT) algorithm for NULAs [23], the MUSIC algorithm [11], and the MLE [10]. We also compare with Cramér-Rao lower bound (CRLB) given by [10]

\[
\text{CRLB} = \frac{1}{2\pi^2 NSU \sin(\theta)}, \quad U = \frac{1}{N} \sum_{n=1}^{N} (r_n - \bar{r})^2, \quad (14)
\]

where \( \theta \) is the source location, \( S \) is the linear signal-to-noise ratio (SNR), and \( r \) is an \( N \times 1 \) antenna spacing vector, \( r_n \) and \( \bar{r} \) are the \( n \)-th element and the average value of \( r \), respectively.

| Algorithms | Off-Grid | Number of Multiplications |
|------------|----------|--------------------------|
| PDP        | Yes      | \( (K+6)N^2/(N+1) \)     |
| Two-step   | Yes      | \( (K_c+1)N \)           |
| 2Q-order   | Yes      | \( N \)                  |
| EM-ESPRIT  | Yes      | \( K_c(2N^2 + \frac{3}{2}N_f^2) \) |
| MUSIC      | No       | \( \frac{2}{3}N^3 + (K_c+K_f+\frac{5}{2})N(N+1) \) |
| MLE        | No       | \( (K_c+K_f+\frac{5}{2})N(N+1) \) |

| Algorithms | \( r_{1,3} \) | \( r_{1,5} \) | \( r_{1,8} \) | \( r_{2,3} \) | \( r_{2,5} \) | \( r_{2,8} \) |
|------------|--------------|--------------|--------------|--------------|--------------|--------------|
| PDP        | 81           | 1150         | 14588        | 33           | 470          | 5796         |
| MLE        | 8670         | 21675        | 52020        | 8670         | 21675        | 52020        |

Table II

**TABLE II**

**ONLINE MULTIPLEMENTS REQUIRED WITH DIFFERENT CONFIGURATIONS**

**A. Computational Complexity Analysis**

We use the number of multiplication operations to characterize each algorithm’s computational complexity, as summarized in Table I. The symbol \( K \) denotes the number of projection points of the PDP algorithm, which is array-layout dependent, as given in (7). The symbols \( N_e \) and \( K_i \) are the virtual array size and the number of iterations for EM-ESPRIT. For grid-search-based methods, the parameters \( K_c \) and \( K_f \) are the numbers of grid points used in the coarse search and fine search, respectively. From the table, we observe that 2Q-order has the lowest computational cost of all algorithms, which depends only on \( N \). However, this algorithm requires specific array configurations. The two-step method has the second-lowest complexity, which is \( K_cN \). For PDP, EM-ESPRIT, MUSIC, and MLE, the complexity comparison depends on the values of \( K, K_i, K_c, \) and \( K_f \). PDP can outperform the former two algorithms in computational complexity when \( K \) is small compared to \( K_i \) and \( K_c+K_f \).

**B. Simulation Results**

In our simulations, we utilize two array configurations, namely, \( r_1 = [0, 5, 10, 15, 16.5, 23, 30, 37.5, 45.5] \) (setup from [18]) and \( r_2 = [0, 0.4, 2.4, 4.9, 2, 10.4, 13.6, 16.4] \) (a non-redundant array from [24]). From these arrays, we create six scenarios using three subarrays of each of \( r_1 \) and \( r_2 \) with different number of antennas \( N = 3, 5, 8 \) (e.g., \( r_{1,3} = [0, 5, 10.5] \) is a subarray of the first three elements of \( r_1 \)). For each scenario, we plot the root mean squared error (RMSE) versus SNR calculated from 1000 simulation trials at each SNR value.
In each trial, the source location is generated randomly from a uniform distribution between $39.5^\circ$ to $40.5^\circ$. For the proposed method, $M = \binom{N}{2}$ phase-difference estimates are computed using (2). For the MUSIC and MLE algorithms, the search is implemented in two stages [10]--a coarse search in the interval $[-70^\circ, 70^\circ]$ using a $0.2^\circ$ step followed by a fine search using a $0.01^\circ$ step. An initial estimation is needed for the EM-ESPRIT [23], which is chosen randomly from $[\theta - 2^\circ, \theta + 2^\circ]$. Also, we set $K_j = 20$. In all simulation trials, and for all methods, a single snapshot is used to estimate the source location. The RMSE performance for all scenarios is presented in Fig. 3. The 2Q-order algorithm only works for $\nu_1$ due to a specific geometry requirement, and hence is not shown in the results based on $\nu_2$ (Fig. 3 (d)-(f)).

For the scenarios of Fig. 3, the proposed PDP algorithm mostly matches the RMSE of the MLE (fine search resolution of $0.01^\circ$), with some deviations that occur at low SNRs, especially for relatively large arrays. Both PDP and MLE ($0.01^\circ$ resolution) achieve the CRLB at high SNRs in all the tested scenarios. The other benchmark methods tend to lack consistency as they deviate from the CRLB in the high SNR regime (exceptions are the 2Q-order and EM-ESPRIT algorithms in Fig. 3 (a)). The EM-ESPRIT algorithm tends to outperform the rest of the methods at low SNRs. However, this can be attributed to the extra information available to this algorithm in the form of a good initial point. The performance of the on-grid methods (MUSIC and MLE) highly depends on the search step. As an example, the MLE performance using only a (coarse) search step of $0.5^\circ$ deviates from the CRLB in scenarios (e) and (f).

As for computational complexity, the two-step and 2Q-order methods require fewer computations than all the other methods. Based on the above discussion, these methods, together with MUSIC and EM-ESPRIT, offer inferior performance on average (compared to PDP and MLE). Therefore, in Table II, we show the actual number of multiplications for only PDP and MLE (with $0.01^\circ$ search step) in all six scenarios. From the table, we observe a substantial advantage for the proposed PDP algorithm. This advantage is attributed mainly to the avoidance of grid search and that the proposed algorithm performs a good amount of its computations offline.

V. CONCLUSION

A phase-difference projection (PDP) direction of arrival (DOA) algorithm is proposed. The proposed algorithm projects the phase-difference observations measured across antenna pairs on a predefined hyperplane determined by the array geometry. Based on this projection, DOA estimation can be achieved in a simple and computationally efficient manner. Simulation results show that the proposed algorithm can match maximum likelihood estimation while maintaining a significant computational-complexity advantage.

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