Robust Beamforming Algorithm Based on Linear Constrained Minimum Variance and Diagonal Loading

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Abstract. In order to improve the robustness of microphone array beamformer in existing mismatch errors, robust beamforming algorithm based on linear constrained minimum variance and diagonal loading method is proposed. The proposed algorithm incorporates the spatial response variation function into the linear constrained minimum variance criterion to design the frequency invariant beamformer and improve the robustness against microphone array mismatch errors in the actual environment by diagonal loading method. Simulation results and analyses illustrate that the array response within passband is robust, the stopband amplitude is controlled in the appropriate range, and the proposed algorithm based closed solution is better than the proposed algorithm based on convex optimization algorithm.

Introduction

As one of the key technologies for microphone arrays, broadband beamforming has been used in a wide range of audio and speech processing applications[1]. In many popular methods, the traditional linear constrained minimum variance(LCMV) beamforming method is mainly used to design narrow-band antenna beamformer[2]. If it is used to design a broadband beamformer of microphone array, it isn’t fit. It is necessary for microphone array to design broadband beamformer without specific arrays[3][4]. In order to achieve broadband frequency invariant beamformer[3], it is necessary to consider frequency invariant beamformer with mismatch errors caused by the gain, phase and position of the microphone[5][6]. For this purpose, in this paper, frequency invariant beamforming algorithm based on the linear constrainted minimum variance (LCMV), the spatial response variation function(SRV) of the microphone array, and diagonal loading method are used to design frequency invariant beamformer and improve the robustness on the given frequency range and the region.

Array Model

Consider a $M$-element linear microphone array in the far-field (not general, this method can be applied to arbitrary array structure), FIR subfilter with tap length $L$ is added to microphone. The array output signal can be expressed as $y(k) = w^H x(k)$, where $H$ denotes conjugate transpose. The weight vector of microphone array response is written as $w(k) = [w_{11}(k),L,w_{m1}(k),L,w_{12}(k),L,w_{m2}(k)]^T$, $T$ denotes transpose. $x(k) = [x_{11}(k),L,x_{m1}(k),L,x_{12}(k),L,x_{m2}(k)]^T$ and it denotes the received signal of microphone array, the definition of the angle between the direction of the sound source and the line array is $\theta$, the beam-pattern at frequency $f$ and angle $\theta$ of arrival can be expressed as $H(f,\theta) = w^T d(f,\theta)$, where frequency $f$ is belongs to $[f_l,f_u]$, the angle $\theta$ of the received signal is in$[\theta_l,\theta_u]$, and $d(f,\theta)$ denotes the array response vector.
Frequency Invariant Beamforming Algorithm and Diagonal Loading Method

The spatial response variation function (SRV) is given by

\[ \text{SRV} = \sum_{n=0}^{N-1} \sum_{q=0}^{Q-1} |w^H d(f_n, \theta_q) - w^H d(f_r, \theta_q)|^2 = w^H R_{ps} w = w^H [(1-\beta)R_p + \beta R_s]w \]  

where \( f_r \) is reference frequency, \( R_{ps} \) is the equilibrium matrix of array spatial response variation function. \( 0 < \beta < 1 \) is an equilibrium parameter between frequency invariance and stopband attenuation. \( R_p \) is the matrix of the reference array response variation vector and \( R_s \) is reference array response matrix. The frequency invariant beamforming problem based on LCMV criterion and SRV can be expressed as

\[ \min_w w^H (R_{sx} + \alpha R_{ps})w \quad \text{s.t.} \quad C^H w = F \]  

where \( R_{sx} = \mathbb{E}[x(k)x^T(k)] \), \( C = [d(f_0, \theta), d(f_1, \theta), \ldots, d(f_{N-1}, \theta)] \), \( F = [e^{-j\pi f_k (t-1)f_s}, \ldots, e^{-j\pi f_k (t-1)f_s}, \ldots, e^{-j\pi f_k (t-1)f_s}]^T \), \( \alpha \) is the matrix weighting factor and a positive number. When some uncertainties in microphone gain, phase, and position are considered, the microphone characteristics of the \( m \)-th microphone the array response of the beamforming with the mismatch errors can be expressed as

\[ g_m^2(f, \theta) = [1 + g_m^e]e^{-j\pi f_k d_m \cos \theta / \gamma}, \quad \mathcal{K}(f, \theta) = \mathcal{K}_0(f, \theta)H(f, \theta) \]  

where \( g_m^e \) denotes gain error, \( \delta_m \) denotes position error, and \( \phi_m \) denotes phase error. The diagonal loading LCMV frequency invariant beamforming problem is expressed as

\[ \min_w w^H (R_{sx} + \alpha R_{ps})w \quad \text{s.t.} \quad C^H w = F, \quad w^H w \leq \zeta \]  

where \( \zeta \) is a constraint value and must meet \( \zeta \geq 1/M \). Using Lagrange method, the weight vector of the beamformer is given by

\[ w_{opt} = (R_{sx} + \alpha R_{ps} + \lambda I)^{-1} C (C^H (R_{sx} + \alpha R_{ps} + \lambda I)^{-1} C)^{-1} F \]  

In order to overcome the defect of the diagonal loading method, according to the convex optimization toolbox, the problem in Eq.4 can be expressed as

\[ \min_w ||Lw||^2 \quad \text{s.t.} \quad \|C^H w - F\|^2 \leq \zeta_0, \quad \|w\|^2 \leq \zeta \]  

where \( R_{sx} + \alpha R_{ps} = L^H L, w^H (R_{sx} + \alpha R_{ps})w = ||Lw||^2 \), \( \zeta_0 \) is a desired signal constraint value. The diagonal loading can be obtained by

\[ w^H w = (CF)^H (R_{sx} + \alpha R_{ps} + \lambda I)^{-2} CF [C^H (R_{sx} + \alpha R_{ps} + \lambda I)^{-1} C]^{-2} = \zeta \]  

where

\[ \zeta \leq \frac{(\gamma_1 + \lambda)^2}{ML(\gamma_{ML} + \lambda)^2} \]  

\[ \lambda \leq \frac{\gamma_1 - (ML\zeta)^{1/2} \gamma_{ML}}{(ML\zeta)^{1/2} - 1} \]  

The diagonal loading values in Eq.9 may be obtained by the Newton iteration method. \( \gamma_i \) denotes the \( i \)-th eigenvalue of \( R \), \( 1 \leq i \leq ML \).
Simulation Tests

The effectiveness of the proposed algorithm was verified in existing microphone mismatch errors. An uniform linear array with 15 identical omnidirectional microphones was used. The order of FIR filter was 31, the array element spacing was the half wavelength of the highest frequency of 5cm, the sampling frequency was 8000Hz, the sound velocity was 340m/s, the noise variance $\sigma^2_n = 0.01$, the interference variance $\sigma^2_i = 0.01$. The direction of the expected signal was 90° and the limit of the gain error of the microphone was $\beta_m = 0.2$, the phase error bound was $\phi_m = \pi/18$, the error bound for the X axis component of the position vector of the microphone was $\delta_m = 0.02$. Simulation results was shown in Fig.1, (a) 3D array response of closed solution, (b) side view of array response of the closed solution, (c) 3D array response of the convex optimization solution, (d) side view of array response of the convex optimization.

![Figure 1](image)

From Fig.1, it can be seen that the array response within passband is robust, the stopband amplitude is controlled in the appropriate range (about -20dB or -17dB), and the robust convex optimization algorithm has poor performance in the expected direction comparison with closed solution.

Conclusion

The robust linear constrained minimum variance beamforming algorithm against microphone mismatches has been proposed, which takes into consideration of diagonal loading method that imposes norm constraint to weight vector. In this proposed algorithm, the linear constrained minimum variance beamforming algorithm is combined with the SRV to obtain better frequency invariant property, the optimal weight vector is obtained by using Lagrange and convex optimization toolbox, respectively. Simulation results show that the proposed algorithm based on closed solution is better than the proposed algorithm based on convex optimization algorithm.

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