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Published in:

2014 IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP 2014)

DOI (link to publication from Publisher): 10.1109/ICASSP.2014.6854748

Publication date: 2014

Document Version Early version, also known as pre-print

Link to publication from Aalborg University

Citation for published version (APA):

Karimian-Azari, S., Jensen, J. R., & Christensen, M. G. (2014). Fundamental Frequency and Model Order Estimation Using Spatial Filtering. In 2014 IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP 2014) (pp. 5964 - 5968). IEEE (Institute of Electrical and Electronics Engineers). https://doi.org/10.1109/ICASSP.2014.6854748

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AUDIO ANALYSIS LAB

Fundamental Frequency and Model Order Estimation Using Spatial Filtering



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Introduction

In real scenarios, a desired signal is contaminated by different levels of noise and interferers, which complicate the estimation of the signal parameters.

In most of the state-of-the-art methods for fundamental frequency and number of harmonics estimation, the desired signal

- is assumed to be degraded by additive white Gaussian noise,
- has higher power than the interferers,
- and does not have spectral overlapping with interferers.

We present an estimation procedure for harmonic-structured signals in situations with strong interference using spatial filtering featuring:

- Joint estimates of the fundamental frequency and the constrained model order
- ► A procedure to account for inharmonicity using an unconstrained model

Signal Model

$$y_{m}(t) = \sum_{n=1}^{N} \sum_{l=1}^{L_{n}} \alpha_{n,l} e^{j(l\omega_{n}t + \varphi_{n,l})} e^{-jl\omega_{n}\Delta\tau_{m,n}} + V_{m}(t)$$
 (1)

(for m = 1, ..., M microphones),

where ω_n and L_n are fundamental frequency and order of the n^{th} signal source, i.e., $x_n(t) = \sum_{l=1}^{L_n} \alpha_{n,l} e^{j(l\omega_n t + \varphi_{n,l})}$, the time difference of arrival between the m^{th} and a reference microphone is $\Delta \tau_{m,n}$, depending on direction of arrivals θ_n , and v_m is Gaussian noise.

Using a frequency-domain vector notation, the received signals are

$$\mathbf{Y}(\omega) = \sum_{n=1}^{N} \mathbf{d}(\theta_n, \omega) X_n(\omega) + \mathbf{V}(\omega), \tag{2}$$

where $\mathbf{Y}(\omega) = [Y_1(\omega) Y_2(\omega) \cdots Y_M(\omega)]^T$, and $\mathbf{d}(\theta_n, \omega) = [1 \ e^{-j\omega\Delta\tau_{2,n}} \cdots e^{-j\omega\Delta\tau_{M,n}}]^T$, for $\omega \in [0, \pi]$ and $\theta_n \in [0, \pi]$.

Spatial Filtering

A complex-valued spatial filter $\mathbf{H}(\theta,\omega)$ is applied on the microphone outputs subject to $\mathbf{H}^{H}(\theta,\omega)\mathbf{d}(\theta,\omega)=1$ like

$$Z(\theta, \omega) = \mathbf{H}^{\mathsf{H}}(\theta, \omega) \mathbf{Y}(\omega).$$
 (3)

Assuming uncorrelated signal sources and noise, the output power corresponding to the direction of the desired signal, i.e., $\theta = \theta_1$, is

$$J_Z(\theta_1,\omega) = \mathsf{E}\{Z(\theta_1,\omega)Z^\mathsf{H}(\theta_1,\omega)\} = J_{X_1}(\omega) + \Psi(\theta_1,\omega),\tag{4}$$

where $J_{X_n}(\omega) = \mathsf{E}\{|X_n(\omega)|^2\}$, and $\Psi(\theta_1, \omega) = \mathsf{H}^\mathsf{H}(\theta_1, \omega)\mathsf{R}_\mathsf{V}(\omega)\mathsf{H}(\theta_1, \omega) + \sum_{n=2}^N \mathsf{H}^\mathsf{H}(\theta_1, \omega)\mathsf{d}(\theta_n, \omega)J_{X_n}(\omega)\mathsf{d}^\mathsf{H}(\theta_n, \omega)\mathsf{H}(\theta_1, \omega).$

Proposed Method

The broadband power of the output signal and the output noise-plus-interference are, respectively,

$$J_Z(\theta_1) = \frac{1}{2\pi} \int_0^{2\pi} J_Z(\theta_1, \omega) d\omega, \qquad (5)$$

$$\Psi(\theta_1) = \frac{1}{2\pi} \int_0^{2\pi} \Psi(\theta_1, \omega) d\omega = J_Z(\theta_1) - J_{X_1}.$$
 (6)

► With the constrained (C) harmonic-model

$$\mathbb{X}_{n}^{\mathsf{C}}(\omega_{n}) = [X_{n}(\omega_{n}) X_{n}(2\omega_{n}) \dots X_{n}(L_{n}^{\mathsf{C}}\omega_{n})]^{\mathsf{T}}, \tag{7}$$

$$\Psi^{C}(\theta_{1}) = J_{Z}(\theta_{1}) - J_{X_{1}}^{C}(\omega_{1}) = J_{Z}(\theta_{1}) - 2 \|X_{1}^{C}(\omega_{1})\|_{2}^{2}.$$
 (8)

With the assumption of white Gaussian noise and using N_f frequency samples, we can jointly estimate the fundamental frequency and the number of harmonics using maximum a posteriori (MAP) [1] like

$$(\hat{L}_1^{\mathrm{C}}, \hat{\omega}_1^{\mathrm{C}}) \approx \arg\min_{L_1^{\mathrm{C}}, \omega_1} N_f \ln[\Psi^{\mathrm{C}}(\theta_1)] + \frac{3}{2} \ln N_f + L_1^{\mathrm{C}} \ln N_f,$$
 (9)

► With the unconstrained (UC) model

$$\mathbb{X}_{n}^{\mathsf{UC}}(\mathbf{\Omega}_{n}) = [X_{n}(\omega_{n,1}) X_{n}(\omega_{n,2}) \dots X_{n}(\omega_{n,L_{n}^{\mathsf{UC}}})]^{\mathsf{T}}, \tag{10}$$

where $\Omega_n = [\omega_{n,1} \omega_{n,2} \dots \omega_{n,L_n^{UC}}]^T$, we have

$$\Psi^{\text{UC}}(\theta_1) = J_Z(\theta_1) - J_{X_1}^{\text{UC}}(\Omega_1) = J_Z(\theta_1) - 2 \|X_1^{\text{UC}}(\Omega_1)\|_2^2.$$
 (11)

We can extend the MAP model order estimation method for estimating the number of independent sinusoids like

$$(\hat{L}_1^{\text{UC}}, \hat{\Omega}_1) \approx \arg\min_{L_1^{\text{UC}}, \Omega_1} N_f \ln[\Psi^{\text{UC}}(\theta_1)] + \frac{5}{2} L_1^{\text{UC}} \ln N_f,$$
 (12)

and apply the Markov-like weighted least-squares (WLS) method [2] to estimate the fundamental frequency.

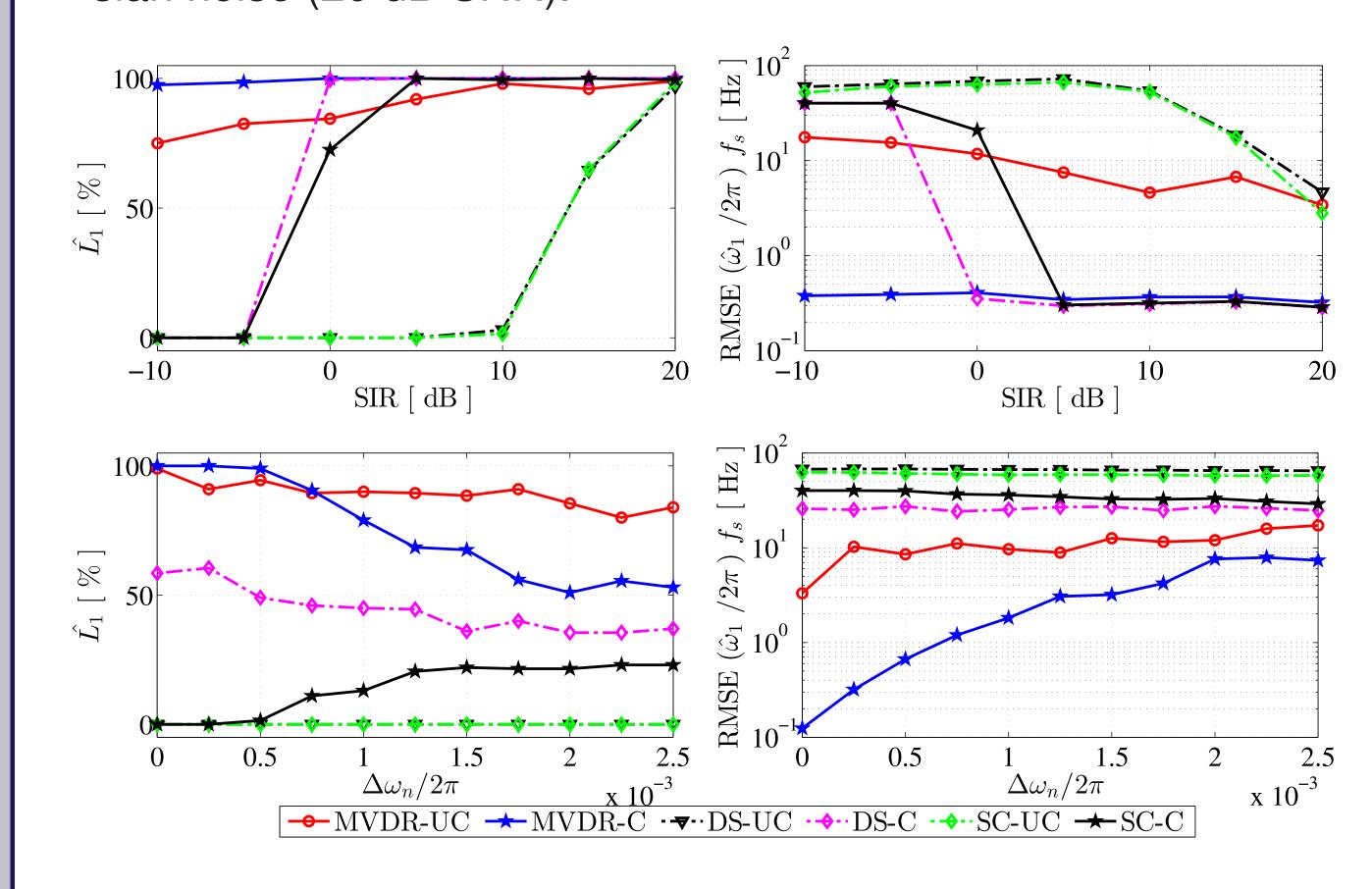
Conclusion

- ► In situations with spatially separated interference signals with low SIRs, the joint fundamental frequency and model order estimation can be facilitated using spatial filters.
- ➤ Simulations indicate that the UC model order estimates are more accurate than the C model. However, the fundamental frequency estimates via the C model are more accurate than the UC based estimator.

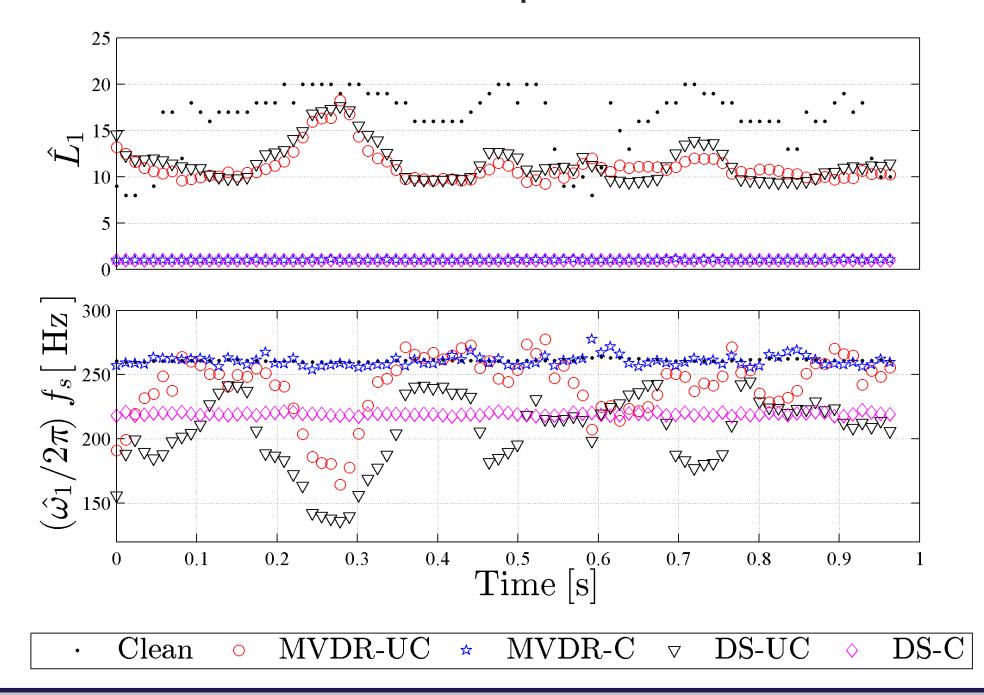
Experimental Results

We compared the results of single-channel (SC) parameter estimators with the proposed method, using the delay-and-sum (DS) and the minimum variance distortionless response (MVDR) beamformers.

► Two synthetic signals: $\theta_1 = 60^\circ$, $\omega_1 = 0.0450\pi$, $L_1 = 5$, and $\theta_2 = 40^\circ$, $\omega_2 = 0.0550\pi$, $L_2 = 7$, with unit amplitudes. Harmonic frequencies were perturbed by a normal distribution ($\Delta \omega_{n,l} = 0.0005\pi$), and the received signals were distorted by white Gaussian noise (20 dB SNR).



A real trumpet signal with vibrato (SIR = -1.5 dB and SNR = 10 dB), and estimates of order and pitch.



References

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