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# **Fundamental Frequency and Model Order Estimation** Using Spatial Filtering



# 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

$$y_m(t) = \sum_{n=1}^N \sum_{l=1}^{L_n} \alpha_{n,l} e^{j(l\omega_n t + \varphi_{n,l})} e^{-j l\omega_n \Delta \tau_{m,n}} + V_m$$

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

where  $\omega_n$  and  $L_n$  are fundamental frequency and order of the n<sup>th</sup> 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  $m^{\text{th}}$  and a reference microphone is  $\Delta \tau_{m,n}$ , depending on direction of arrivals  $\theta_n$ , and  $v_m$  is Gaussian noise.

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

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

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

# **Spatial Filtering**

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

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

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

$$J_{Z}(\theta_{1},\omega) = \mathsf{E}\{Z(\theta_{1},\omega)Z^{\mathsf{H}}(\theta_{1},\omega)\} = J_{X_{1}}(\omega) + \Psi(\theta_{1},\omega)\}$$

where  $J_{X_n}(\omega) = \mathbb{E}\{|X_n(\omega)|^2\}$ , and  $\Psi(\theta_1, \omega) = \mathbb{H}^{H}(\theta_1, \omega)\mathbb{R}_{V}(\omega)\mathbb{H}(\theta_1, \omega) + \mathbb{E}\{|X_n(\omega)|^2\}$  $\sum_{n=2}^{N} \mathbf{H}^{\mathsf{H}}(\theta_{1}, \omega) \mathbf{d}(\theta_{n}, \omega) J_{X_{n}}(\omega) \mathbf{d}^{\mathsf{H}}(\theta_{n}, \omega) \mathbf{H}(\theta_{1}, \omega).$ 

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(1) n(t)

(2)

(3)

(4)  $\nu_1,\omega),$ 

### **Proposed Method**

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

$$J_{Z}(\theta_{1}) = \frac{1}{2\pi} \int_{0}^{2\pi} J_{Z}(\theta_{1},\omega) d\omega, \qquad (5)$$
$$\Psi(\theta_{1}) = \frac{1}{2\pi} \int_{0}^{2\pi} \Psi(\theta_{1},\omega) d\omega = J_{Z}(\theta_{1}) - J_{X_{1}}. \qquad (6)$$

With the constrained (C) harmonic-model  $\mathbb{X}_{n}^{\mathsf{C}}(\omega_{n}) = [X_{n}(\omega_{n}) X_{n}(2\omega_{n}) \dots X_{n}(2\omega_{n})]$ 

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

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

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

$$\Psi^{UC}(\theta_1) = J_Z(\theta_1) - J_{X_1}^{UC}(\mathbf{\Omega}_1) = J_Z(\theta$$

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

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

$$[L_n^{\mathsf{C}}\omega_n)]^{\mathsf{T}},\tag{7}$$

) ... 
$$X_n(\omega_{n,L_n^{\text{UC}}})]^{\mathsf{T}}$$
, (10)

 $(\theta_1) - 2 \| \mathbb{X}_1^{UC}(\Omega_1) \|_2^2.$ (11)

### **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.

sian noise (20 dB SNR).



dB), and estimates of order and pitch.



### References

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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 Gaus-

► A real trumpet signal with vibrato (SIR = -1.5 dB and SNR = 10

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