# High Accuracy Acoustic Estimation of Multiple Targets



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## Introduction

- Precise estimation of multiple targets is required in many of nowadays technologies.
- Many types of signals have been used for multiple targets estimation:



- The update of the parameter vector is:  $\theta^{(k+1)} = \theta^{(k)} + p_k$ .
- For multiple targets, the Expectation-Maximization (EM) algorithm is used [1] to solve the following optimization problem:

$$\min_{\theta_m} \|q - \sum_{m=1}^{M} s(\theta_m)\|^2 \tag{4}$$

where q is the observed data vector which represents the cross-correlation.

 The EM algorithm defines *M* unobserved data vectors as *h<sub>m</sub>* = *s*(θ<sub>m</sub>) + *n<sub>m</sub>*, where *n<sub>m</sub>* is an AWGN sequence. The relation between the observed and the unobserved data is: *q* = ∑<sup>M</sup><sub>m=1</sub> *h<sub>m</sub>*.
 The expectation of the unobserved data *h<sub>m</sub>* can be computed as [1] ∩(k) = 1 <sup>M</sup><sub>m</sub> (k)

(a) Infrared (b) Radar TOA (c) Ultrasound

Time of arrival (TOA) estimation using cross-correlation is simple but its resolution is dictated by the available bandwidth:

$$R = \frac{2v_s}{BW} \tag{1}$$

where  $v_s$  is the wave propagation speed in the medium and BW is the bandwidth of the system.



- The challenge is to resolve the TOAs of the reflected signals from the cross-correlation.
- The deconvolution method provides high resolution estimation TOAs of the reflected signals at high SNR, but the accuracy significantly degrades under low SNR.

- $\hat{h}_{m}^{(k)} = s(\theta_{m}^{(k)}) + \frac{1}{M}(q \sum_{l=1}^{M} s(\theta_{l}^{(k)}))$ (5)
- The M-step iterates the parameter vector  $\theta_m^{(k)}$  by minimizing:  $\theta_m^{(k+1)} = \arg_{\theta_m} \min \|\hat{h}_m^{(k)} - s(\theta_m)\|^2$
- The initial guess of  $\beta_m^{(0)}$ , and  $f_{cm}^{(0)}$  depends on the available hardware bandwidth and center frequency and the initial phase  $\phi_m^{(0)}$  is set to zero.



(6)

- Determine if there are overlapping peaks based on the width of each peak and the inverse-square law (amplitude  $\propto \frac{1}{\text{distance}^2}$ ).
- The initial guess for the overlapping targets is the same as the initial guess of the m<sup>th</sup> target except that the TOAs are perturbed.
- The deconvolution method provides high resolution:

$$Z(f) = \frac{Q(f)}{S(f)} \implies z(t) = \mathsf{IFFT}(Z(f)) = \sum_{k=1}^{M} c_k \delta(t - \tau_k)$$
(7)

where Q(f) is the FFT of the estimated GEM of the cross-correlation, z(t) is the high resolution output of the deconvolution method,  $c_k$  is the amplitude of the  $k^{\text{th}}$  target signal, and  $\delta(t)$  is the Dirac delta function.



Denoising the cross-correlation vector is required before applying the deconvolution method.

#### **Multiple TOAs Estimation**

- Transmitting an m-sequence and cross-correlating Rx with Tx gives the system impulse response (SIR): r<sub>yx</sub>(I) = h(I) \* r<sub>xx</sub>(I)
  The Gaussian echo model (GEM) is a good model for SIR [1]: s(θ; t(nT<sub>s</sub>)) = αe<sup>-β(t(nT<sub>s</sub>)-τ)<sup>2</sup></sup> cos(2πf<sub>c</sub>(t(nT<sub>s</sub>) τ) + φ), where θ = [β, τ, f<sub>c</sub>, φ, α], β is a BW factor, τ is the TOA, f<sub>c</sub> is the center
- frequency,  $\phi$  is the phase and  $\alpha$  is the amplitude of the target signal.
- The SIR in the presence of multiple targets can be modeled as:
  - $M \qquad M \qquad M \qquad M$

### **Experimental Setup**



(a) Static Ranging



(b) Movement Tracking

### **Results**

									T						
									Proposed Algorithm			GEM estimation			
	Proposed Algorithm,			GEM estimation			CC	Rx-Tx	RMSE	σ	% error <	RMSE	$\sigma$	% error<	
Exps	RMSE	$\sigma$	Μ	RMSE	$\sigma$	Μ	M		(mm)	(mm)	5 mm	(mm)	(mm)	5 mm	
	(mm)	(mm)		(mm)	(mm)			1-1	2.30	1.48	99.40	1.07	0.97	100	
Exp 1	2.11	2.10	3	2.86	2.26	3	3	2-1	3.11	3.10	88.80	3.60	3.01	84.43	
Exp 2	5.89	5.81	3	10.78	5.91	3	2	3-1	2.86	2.38	96.80	2.26	2.20	95.41	
Exp 3	3.52	3.48	3	7.32	4.54	3	2	4-1	2.35	2.35	97.40	3.28	2.97	90.82	
Exp 4	5.11	4.13	3	8.67	3.15	3	1	1-2	3.98	2.53	76.65	3.02	2.51	93.41	
Exp 5	3.39	2.43	4	3.46	3.34	4	2	2-2	8.30	6.35	27.54	9.03	6.15	27.74	
Exp 6	3.99	3.08	3	8.52	5.61	3	2	3-2	7.21	5.70	50.9	6.21	5.75	42.32	
Table: Estimated multiple								4-2	6.02	4.56	66.67	11.13	10.78	57.68	
taraa	te RN		$\sigma$ and					Ta	Table: Tx1 and Tx2 tracking results						

 $q(t(nT_s)) = \sum_{m=1}^{\infty} h(t(nT_s))_m = \sum_{m=1}^{\infty} s(\theta_m; t(nT_s)) + n(t(nT_s)), \quad (2)$ where *M* is the number of targets,  $h(t(nT_s))$  is the impulse response for a single target, and  $n(t(nT_s))$  is the noise coming from the discrepancy between the system impulse response and the GEM.

For a single target, we need to solve:

$$\begin{split} \min_{\theta} \|h(t(nT_s)) - s(\theta, t(nT_s))\|^2 \quad (3) \\ &\blacktriangleright \text{ Using Levenberg-Marquardt (LM) algorithm [2], the step vector:} \\ &p_k = \begin{cases} -(H_k^T H_k)^{-1} H_k^T r_k, & \text{if } \|p_k^{GN}\| \leq \Delta_k, \\ -(H_k^T H_k + \lambda \text{diag}(H_k^T H_k))^{-1} H_k^T r_k, & \text{if } \|p_k^{GN}\| > \Delta_k. \end{cases} \end{split}$$

where  $H_k$  is the gradient vector of the model with respect to the parameter vector  $\theta$ ,  $r_k$  is the residual function,  $\Delta_k$  is the trust region radius,  $\lambda$  is the damping factor, and  $p_k^{GN} = -(H_k^T H_k)^{-1} H_k^T r_k$ .



#### References

- 1 Ramazan Demirli and Jafar Saniie, "Model-based estimation of ultrasonic echoes. part i: Analysis and algorithms," IEEE transactions on ultrasonics, ferroelectrics, and frequency control, vol. 48, no. 3, pp. 787-802, 2001.
- 2 Stephen J Wright and Jorge Nocedal, "Numerical optimization," Springer Science, vol. 35, no. 67-68, pp. 7, 1999.