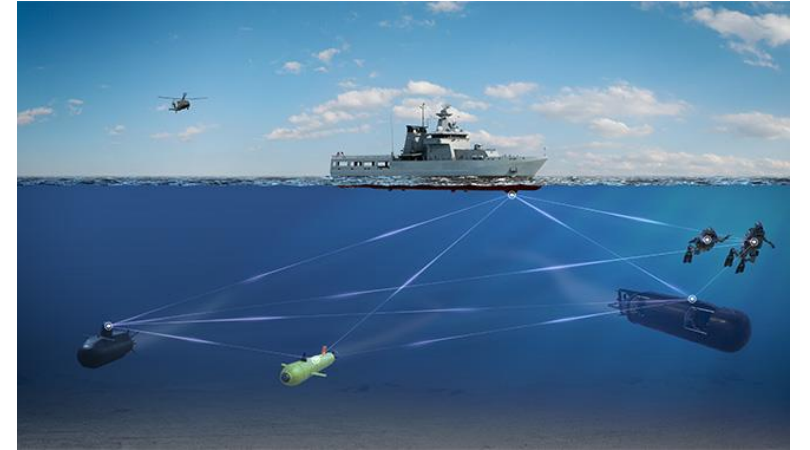
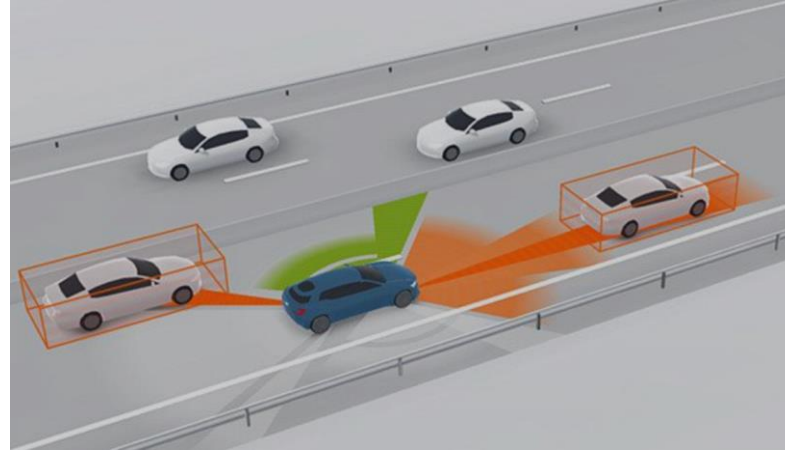


Mixed-Precision Arrangement via CRB for DOA Estimation Using SLA

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DOA Estimation

- Application
 - Radar, autonomous driving, sonar [1,2,3]
 - Array signal processing



GOAL: To determine the directions of the sources

[1]. E. Tuncer and B. Friedlander, Classical and modern direction-of-arrival estimation, Academic Press, 2009.

[2]. S. Sun, A. P. Petropulu, and H. V. Poor, "MIMO radar for advanced driver-assistance systems and autonomous driving: Advantages and challenges," IEEE Signal Process. Mag., vol. 37, no. 4, pp. 98–117, 2020.

[3]. H. Krim and M. Viberg, "Two decades of array signal processing research: The parametric approach," IEEE Signal Process. Mag., vol. 13, no. 4, pp. 67–94, 1996.

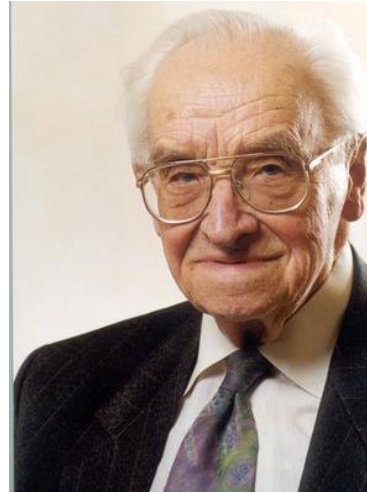
Analog-to-Digital Converters

- Sampling Theory

- Developed by Whittaker (1915), Nyquist (1928), Kotelnikov (1933), and Shannon (1949) and others



Nyquist



Kotelnikov



Shannon

- Gives mathematical proof that bandlimited signals can be perfectly reconstructed from samples taken above the Nyquist rate $f_s > 2f$

Problems

- High performance system requirements
 - High resolution
 - Wide signal bandwidth requires high speed precision ADC
 - Large number of antennas requires many high speed precision ADCs
- **Cost and power consumption** of ADCs is a challenge

$$P = f_s \cdot 10^{\frac{6b}{10}} \cdot \text{const.}$$

Diagram illustrating the power consumption equation for ADCs:

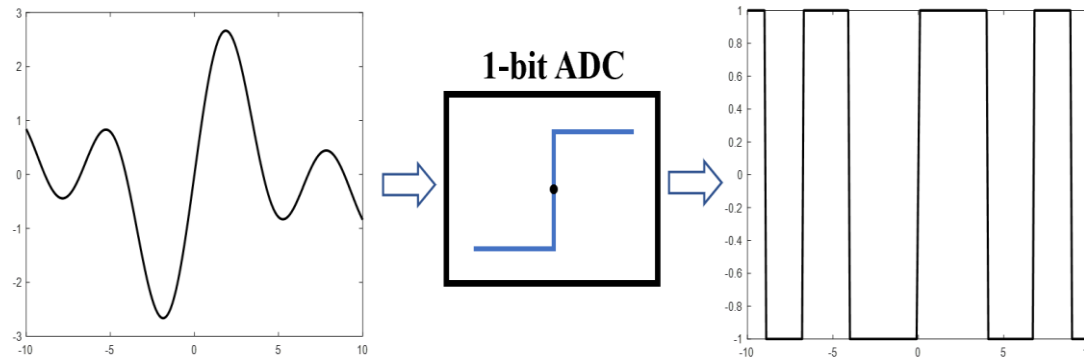
- Sampling rate** points to f_s .
- Quantization bit** points to b in the exponent.

- How to solve these issue?



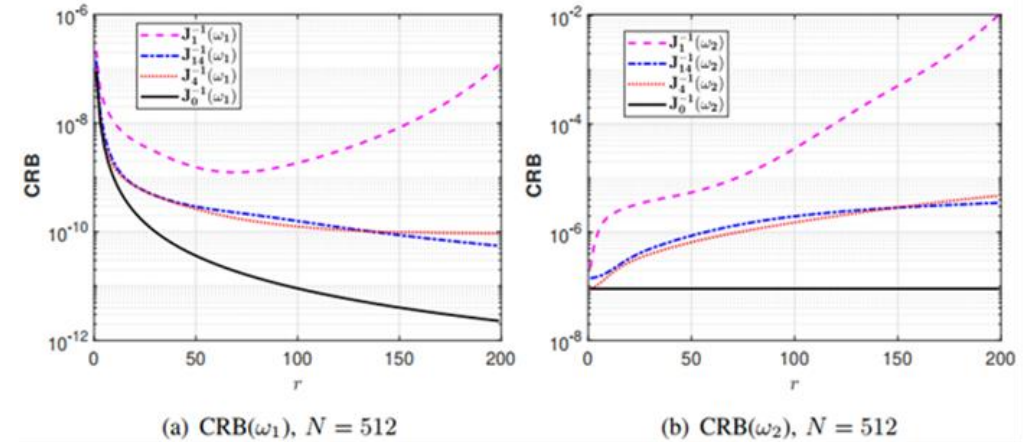
One-bit Sampling

- Quantize signal to one-bit data with known thresholds



- Low-bit quantization suffers from **dynamic range problem [4]**

Good method to save cost and energy of ADCs

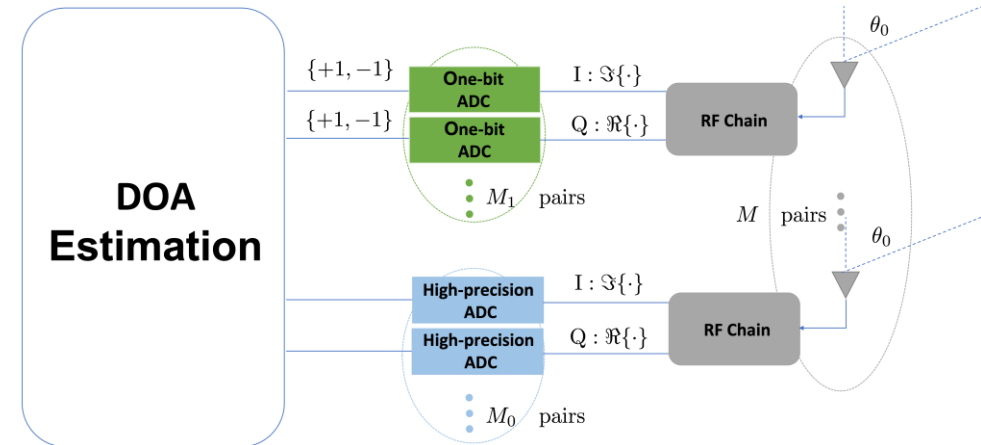
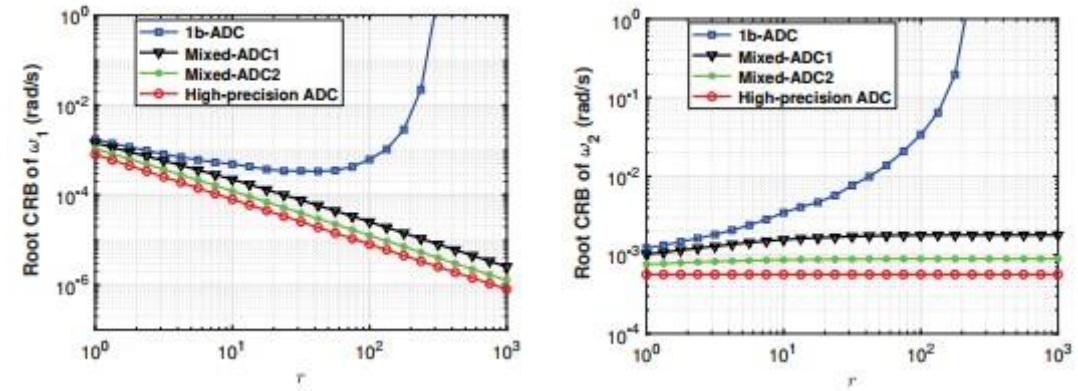


r is the amplitude ratio between two sinusoidal signals [5]

- [4]. R. H. Walden, "Analog-to-digital converter survey and analysis," IEEE J. Sel. Areas Commun., vol. 17, no. 4, pp. 539–550, 1999.
- [5]. P. Stoica, X. Shang and Y. Cheng, "The Cramér–Rao Bound for Signal Parameter Estimation From Quantized Data [Lecture Notes]," in IEEE Signal Process. Mag., vol. 39, no. 1, pp. 118-125, Jan. 2022.

Mixed-ADC Architecture [6]

- Some antenna are sampled by high-precision ADC, others are sampled by one-bit ADC
 - Save cost and energy
 - Overcome the dynamic range problem [7]
- Mixed-ADC for DOA estimation [8]
 - High-precision and one-bit ADCs are put on two hands
 - Without considering the **location arrangement**



[6]. N. Liang and W. Zhang, "Mixed-ADC massive MIMO," IEEE J. Sel. Areas Commun., vol. 34, no. 4, pp. 983997, 2016.

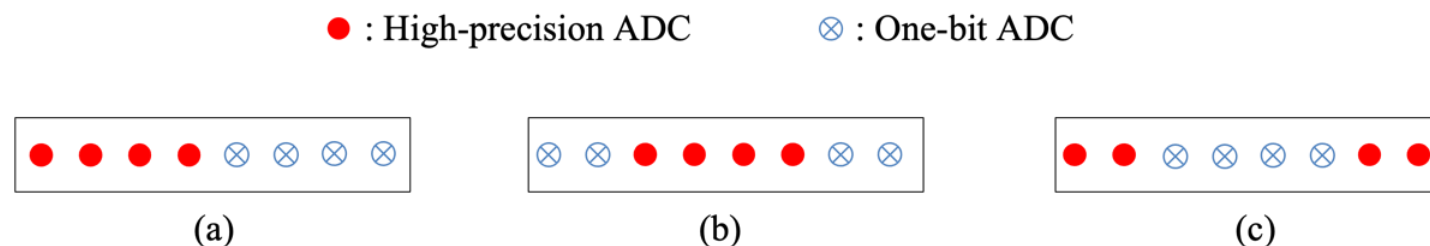
[7]. Y. Cheng, X. Shang, and F. Liu, "CRB analysis for mixed-ADC PMCW MIMO Radar," in Proc. CIE Int. Conf. Radar, Hai Kou, China, Dec. 2021, pp. 1032–1037.

[8]. B. Shi, L. Zhu, W. Cai, N. Chen, T. Shen, P. Zhu, F. Shu, and J. Wang, "On performance loss of DOA measurement using massive MIMO receiver with mixedADCs," IEEE Wireless Commun. Lett., 2022.



Location Arrangement of ADC

- Performance of different mixed-ADC arrangement is different using ULA [9]
 - The optimal mixed-precision arrangement for DOA estimation (c)



- Sparse Linear Array (SLA) with specific geometries
 - Minimum redundancy arrays [10], Nested arrays [11]
 - Higher angular resolutions and identify more sources [12]
- What is the optimal mixed-ADC arrangement using SLA?

[9]. X. Zhang, Y. Cheng, X. Shang, and J. Liu, "Optimal mixed-ADC arrangement for DOA estimation via CRB using ULA," in Proc. IEEE Int. Conf. Acoust., Speech Signal Process., Rhodes Island, Greece, Jun. 2023.

[10]. A. Moffet, "Minimum-redundancy linear arrays," IEEE Trans. Antennas Propag., vol. 16, no. 2, pp. 172–175, 1968.

[11]. P. Pal and P. P. Vaidyanathan, "Nested arrays: A novel approach to array processing with enhanced degrees of freedom," IEEE Trans. Signal Process., vol. 58, no. 8, pp. 4167–4181, 2010.

[12]. P. Stoica and A. Nehorai, "Performance study of conditional and unconditional direction-of-arrival estimation," IEEE Trans. Acoust., Speech, and Signal Process., vol. 38, no. 10, pp. 1783–1795, 1990.



Signal Model

- K narrowband far-field signals from different directions (N snapshot)

$$\mathbf{X} = \mathbf{A}\mathbf{S} + \mathbf{E} \quad \mathbf{a}(\theta_k) = [e^{j\pi d_1 \sin \theta_k}, e^{j\pi d_2 \sin \theta_k}, \dots, e^{j\pi d_M \sin \theta_k}]^T$$

- One-bit sampling with known time-varying threshold

$$\mathbf{Z} = \mathcal{Q}(\mathbf{X} - \mathbf{H})$$

- Mixed-ADC output (M_1 pairs of high-precision ADCs and M_0 pairs of one-bit ADCs)

$$\mathbf{Y} = \mathbf{Z} \circ (\bar{\boldsymbol{\delta}} \otimes \mathbf{1}_N^T) + \mathbf{X} \circ (\boldsymbol{\delta} \otimes \mathbf{1}_N^T)$$

$\delta_i = 1$ High-precision ADC
 $\delta_i = 0$ One-bit ADC



Cramér-Rao Bounds of DOA

- Asymptotic (low SNR) Cramér-Rao bound of DOA [9, 13]

$$\text{CRB}(\boldsymbol{\theta}) = \sigma^2 / (2N) \Re \left\{ \left(\dot{\mathbf{A}}^H \boldsymbol{\Omega} \dot{\mathbf{A}} \right) \circ \hat{\mathbf{P}}^T \right\}^{-1}$$

$$\hat{\mathbf{P}} = \frac{1}{N} \sum_{t=1}^N \mathbf{s}(t) \mathbf{s}^H(t)$$

$$\boldsymbol{\Omega} = \boldsymbol{\Sigma}_0 - \boldsymbol{\Sigma}_0 \mathbf{A} (\mathbf{A}^H \boldsymbol{\Sigma}_0 \mathbf{A})^{-1} \mathbf{A}^H \boldsymbol{\Sigma}_0$$

$$\boldsymbol{\Sigma}_0 = \left(1 - \frac{2}{\pi}\right) \text{diag}(\boldsymbol{\delta}) + \frac{2}{\pi} \mathbf{I}_M$$

- Case of single source ($N \gg 1$)

$$\text{CRB}(\theta) = \frac{M_0 + \frac{2}{\pi} M_1}{2\pi^2 S} \frac{1}{\text{SNR} \cos^2 \theta}$$

$$g_i \in \left\{ 1, \frac{2}{\pi} \right\}$$

$$S = \sum_{i=1}^M g_i d_i^2 \sum_{i=1}^M g_i - \left(\sum_{i=1}^M g_i d_i \right)^2$$

$$\sum_{i=1}^M g_i = M_0 + \frac{2}{\pi} M_1$$

[9]. X. Zhang, Y. Cheng, X. Shang, and J. Liu, "Optimal mixed-ADC arrangement for DOA estimation via CRB using ULA," in Proc. IEEE Int. Conf. Acoust., Speech Signal Process., Rhodes Island, Greece, Jun. 2023.

[13]. P. Stoica and A. Nehorai, "MUSIC, maximum likelihood, and Cramér-Rao bound," IEEE Trans. Acoust., Speech, Signal Process., vol. 37, no. 5, pp. 720–741, 1989.



Mixed-Precision Arrangement Problem

- Mixed-precision arrangement problem using SLA

$$\begin{aligned} \max_{(g_i)_{i=1,2,\dots,M}} \quad & S = \sum_{i=1}^M \sum_{j>i}^M g_i g_j (d_j - d_i)^2 \\ \text{s.t.} \quad & g_i \in \left\{1, \frac{2}{\pi}\right\}, \quad i = 1, 2, \dots, M \\ & \sum_{i=1}^M g_i = M_0 + \frac{2}{\pi} M_1, \end{aligned}$$

This is a Combinatorial problem and is more complex than [9]

- Convert to a 0-1 integer programming problem

$$\begin{aligned} \min_{\mathbf{z}} \quad & -\mathbf{z}^T \mathbf{D} \mathbf{z} - \mathbf{b}^T \mathbf{z} \\ \text{s.t.} \quad & \mathbf{z} \in \{0, 1\}^M, \\ & \mathbf{1}_M^T \mathbf{z} = M_0. \end{aligned}$$
$$\begin{aligned} \mathbf{b} &= \frac{4}{\pi - 2} \mathbf{D} \mathbf{1}_M \\ \mathbf{D}_{ij} &= (d_i - d_j)^2 \end{aligned}$$

[9]. X. Zhang, Y. Cheng, X. Shang, and J. Liu, "Optimal mixed-ADC arrangement for DOA estimation via CRB using ULA," in Proc. IEEE Int. Conf. Acoust., Speech Signal Process., Rhodes Island, Greece, Jun. 2023.



Optimal Solution

- Convert binary constraint to continuous constraint [14]

$$\mathbf{z} \in \{0, 1\}^M \Leftrightarrow \mathbf{z} \in S_b \cap \mathbf{z} \in S_p$$

Box constraint: $S_b = [0, 1]^M$ ℓ_2 -sphere constraint: $S_p = \left\{ \mathbf{z} : \left\| \mathbf{z} - \frac{1}{2} \mathbf{1}_M \right\|_2^2 = \frac{M}{4} \right\}$

- Introduce two additional variables to decompose the constraints on \mathbf{z}

$$\begin{aligned} \min_{\mathbf{z}} \quad & -\mathbf{z}^T \mathbf{D} \mathbf{z} - \mathbf{b}^T \mathbf{z} \\ \text{s.t.} \quad & \mathbf{1}_M^T \mathbf{z} = M_0, \quad \mathbf{z} = \mathbf{z}_1, \quad \mathbf{z} = \mathbf{z}_2, \\ & \mathbf{z}_1 \in S_b, \quad \mathbf{z}_2 \in S_p, \end{aligned}$$

[14]. B. Wu and B. Ghanem, "p-box ADMM: A versatile framework for integer programming," IEEE Trans. Pattern Anal. Mach. Intell., vol. 41, no. 7, pp. 1695–1708, 2018.



Optimization Solution

- **Augmented Lagrangian function**

$$\begin{aligned}\mathcal{L}(\mathbf{z}, \mathbf{z}_1, \mathbf{z}_2, \boldsymbol{\eta}_1, \boldsymbol{\eta}_2, \eta_3) \\&= -\mathbf{z}^T \mathbf{D} \mathbf{z} - \mathbf{b}^T \mathbf{z} + h_1(\mathbf{z}_1) + h_2(\mathbf{z}_2) + \boldsymbol{\eta}_1^T (\mathbf{z} - \mathbf{z}_1) + \boldsymbol{\eta}_2^T (\mathbf{z} - \mathbf{z}_2) \\&+ \eta_3 (\mathbf{1}_M^T \mathbf{z} - M_0) + \frac{\rho_1}{2} \|\mathbf{z} - \mathbf{z}_1\|_2^2 + \frac{\rho_2}{2} \|\mathbf{z} - \mathbf{z}_2\|_2^2 + \frac{\rho_3}{2} \|\mathbf{1}_M^T \mathbf{z} - M_0\|_2^2\end{aligned}$$

Dual variables: $\boldsymbol{\eta}_1 \in \mathbb{R}^M, \boldsymbol{\eta}_2 \in \mathbb{R}^M, \eta_3 \in \mathbb{R}$ **Positive penalty parameters:** (ρ_1, ρ_2, ρ_3)

Indicator functions: $h_1(\mathbf{z}_1) = \mathbb{I}_{\{\mathbf{z}_1 \in S_b\}}$ $h_2(\mathbf{z}_2) = \mathbb{I}_{\{\mathbf{z}_2 \in S_p\}}$ $\mathbb{I}_a = \begin{cases} 0 & \text{when } a \text{ is true} \\ +\infty & \text{when } a \text{ is false} \end{cases}$

- **Update the variables iteratively like conventional ADMM**



Optimization Solution

- Update \mathbf{z}^{t+1}

$$(-2\mathbf{D} + (\rho_1 + \rho_2)\mathbf{I} + \rho_3\mathbf{1}_M\mathbf{1}_M^T)\mathbf{z}^{t+1} = \rho_1\mathbf{z}_1^t + \rho_2\mathbf{z}_2^t + M_0\rho_3\mathbf{1}_M + \mathbf{b} - \boldsymbol{\eta}_1^t - \boldsymbol{\eta}_2^t - \boldsymbol{\eta}_3^t\mathbf{1}_M$$

↑
Positive semi-definite

$$-\mathbf{z}^T\mathbf{D}\mathbf{z} = \mathbf{z}^T(\alpha\mathbf{I} - \mathbf{D})\mathbf{z} - \alpha\mathbf{1}_M^T\mathbf{z}$$

- Update $(\mathbf{z}_1^{t+1}, \mathbf{z}_2^{t+1})$

$$\mathbf{z}_1 = \arg \min_{\mathbf{z}_1} h_1(\mathbf{z}_1) + \frac{\rho_1}{2} \|\mathbf{z} - \mathbf{z}_1\|_2^2 + \boldsymbol{\eta}_1^T(\mathbf{z} - \mathbf{z}_1),$$

$$\mathbf{z}_2 = \arg \min_{\mathbf{z}_2} h_2(\mathbf{z}_2) + \frac{\rho_2}{2} \|\mathbf{z} - \mathbf{z}_2\|_2^2 + \boldsymbol{\eta}_2^T(\mathbf{z} - \mathbf{z}_2).$$



$$\mathbf{z}_1^{t+1} = \mathbf{P}_{S_b} \left(\mathbf{z}^{t+1} + \frac{1}{\rho_1} \boldsymbol{\eta}_1^t \right),$$

$$\mathbf{z}_2^{t+1} = \mathbf{P}_{S_p} \left(\mathbf{z}^{t+1} + \frac{1}{\rho_2} \boldsymbol{\eta}_2^t \right),$$

$$\mathbf{P}_{S_b}(\mathbf{x}) = \min(\mathbf{1}_M, \max(\mathbf{0}_M, \mathbf{x}))$$

$$\mathbf{P}_{S_p}(\mathbf{x}) = \frac{\sqrt{M}}{2} \frac{\mathbf{x} - \frac{1}{2}\mathbf{1}_M}{\|\mathbf{x} - \frac{1}{2}\mathbf{1}_M\|} + \frac{1}{2}\mathbf{1}_M$$



Optimization Solution

- **Update** $(\eta_1^{t+1}, \eta_2^{t+1}, \eta_3^{t+1})$ **by conventional gradient ascent method**

$$\eta_1^{t+1} = \eta_1^t + \rho_1(\mathbf{z}^{t+1} - \mathbf{z}_1^{t+1}),$$

$$\eta_2^{t+1} = \eta_2^t + \rho_1(\mathbf{z}^{t+1} - \mathbf{z}_2^{t+1}),$$

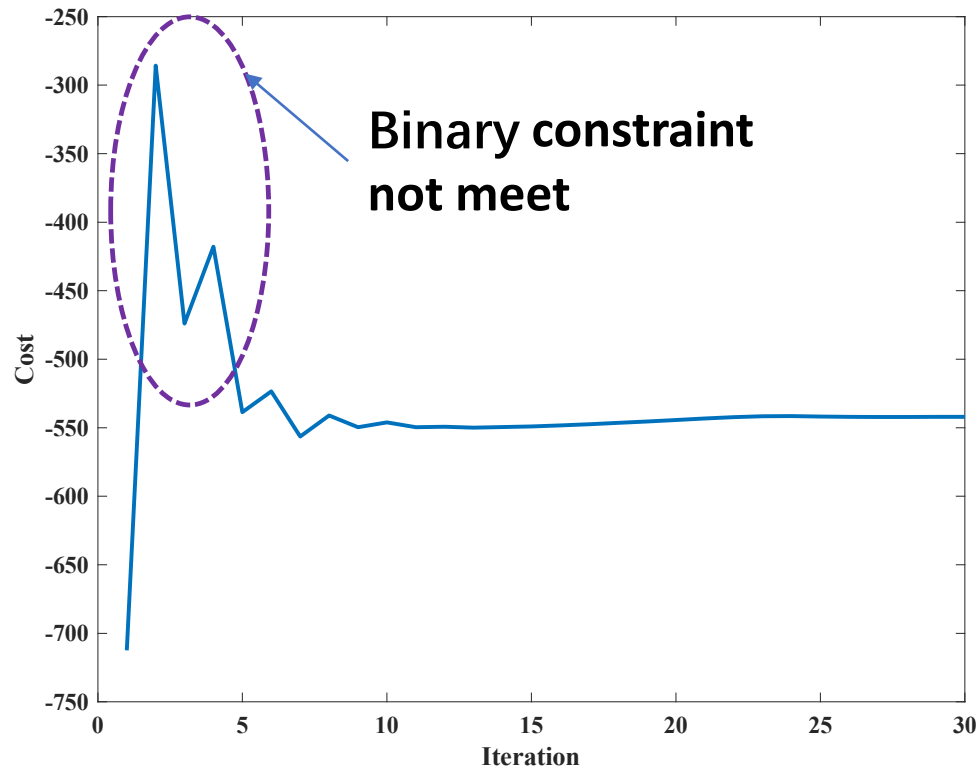
$$\eta_3^{t+1} = \eta_3^t + \rho_3(\mathbf{1}_M^T \mathbf{z} - M_0).$$

**ADMM can provide a guarantee of convergence to a KKT point
(local optimal point)**

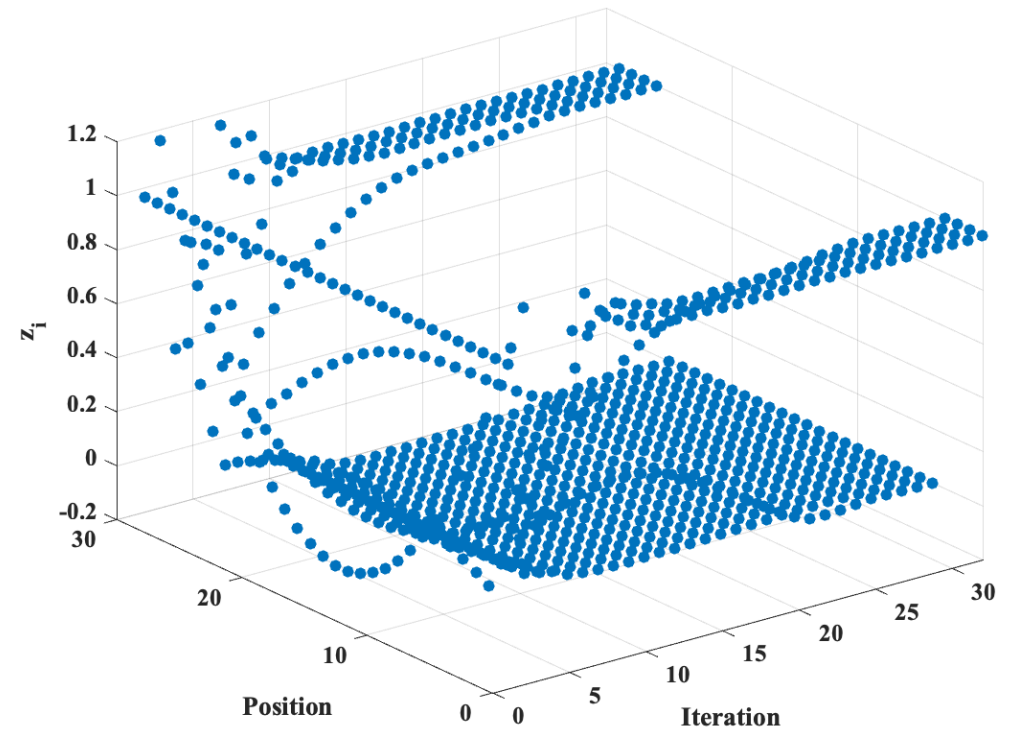


Example. 1

- The optimization procedure of ADMM using ULA $M = 30, M_0 = 8$



Objective value versus iteration

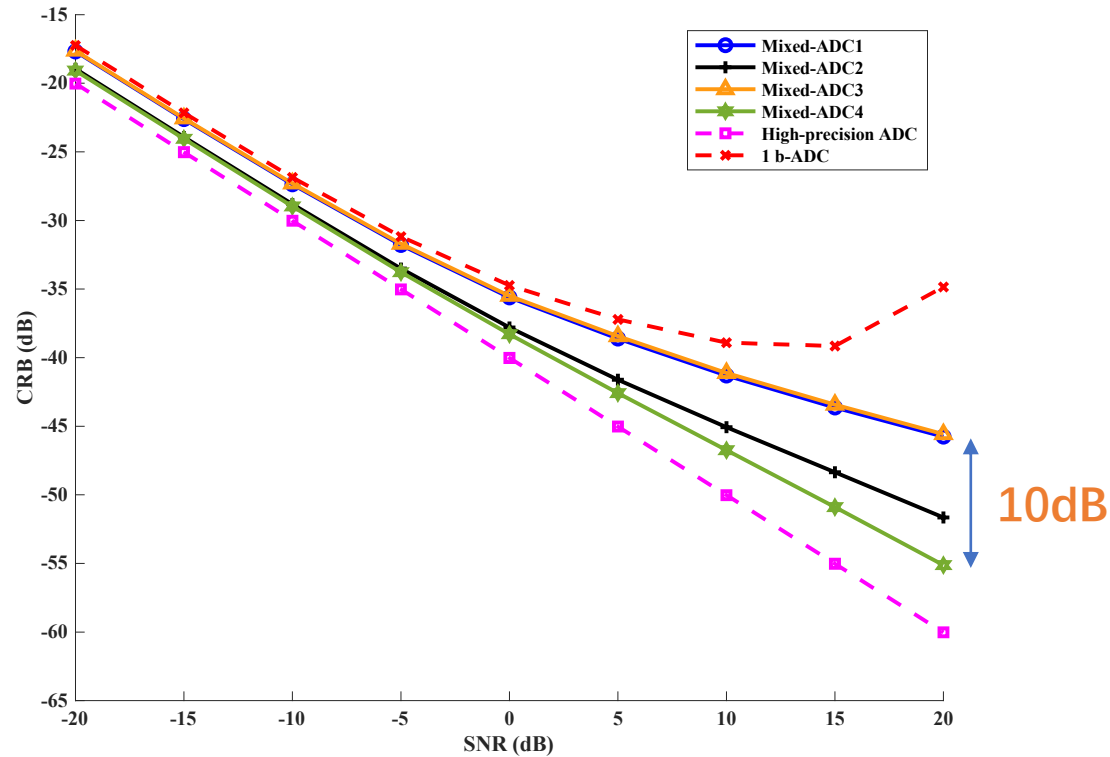


Solution changes versus iteration

Result is consistent with our previous work [9]

Example. 2

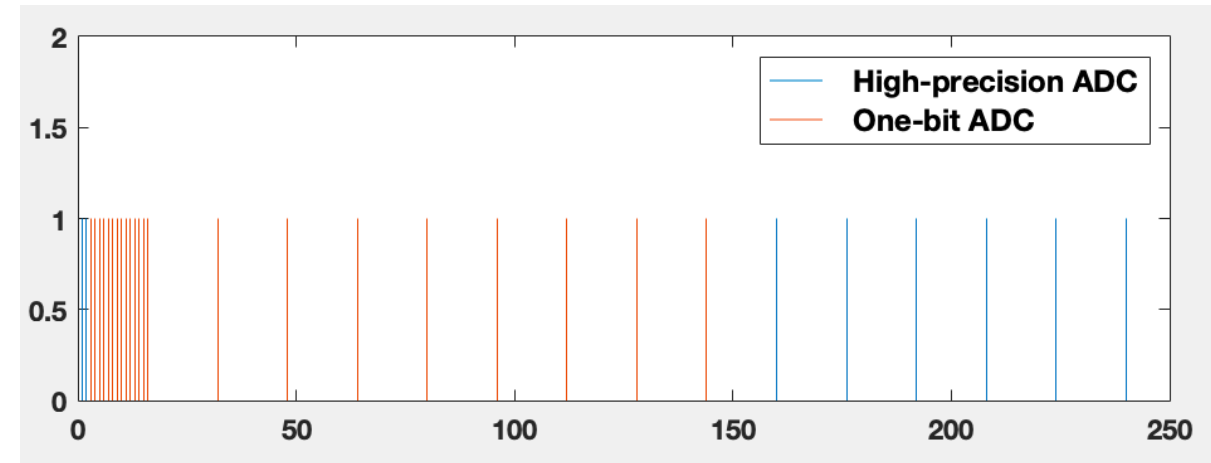
- CRB comparisons on Nested array $M = 30, M_0 = 8$



CRB versus SNR on nested array

$$N_1 = N_2 = 15$$

- 1) $\{\delta_i = 1\}_{i=1}^8$ and $\{\delta_i = 0\}_{i=9}^{30}$;
- 2) $\{\delta_i = 0\}_{i=1}^{22}$ and $\{\delta_i = 1\}_{i=23}^{30}$;
- 3) $\{\delta_i = 0\}_{i=1}^{10}$, $\{\delta_i = 1\}_{i=11}^{18}$ and $\{\delta_i = 0\}_{i=19}^{30}$;
- 4) The optimal mixed-precision arrangement computed by the ADMM.



Optimal mixed-precision arrangement

$$\{\delta_i = 1\}_{i=1}^2, \{\delta_i = 0\}_{i=3}^{24} \quad \text{and} \quad \{\delta_i = 1\}_{i=25}^{30}$$

Findings

- **Although the optimization goal is asymptotic CRB, the true CRB performs better, especially in large SNR.**
- **Through experiments, the optimal mixed-precision arrangement still works when extending to multiple sources case.**

Summary

- **We explore the arrangement of one-bit and high-precision ADCs in SLAs with arbitrary structure under a mixed-ADC based architecture.**
- **We simplified the problem into a 0-1 integer quadratic programming and solve it efficiently using the ADMM algorithm to obtain the optimal mixed-precision arrangement.**



Thanks everyone !