

# Multiresolution Time-of-Arrival Estimation from Multiband Radio Channel Measurements

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## Introduction

- In radio communications, the time-of-arrival (TOA) estimation starts with the estimation of the underlying non bandlimited multipath channel from bandlimited observations of communication signals.
- Modeling the channel impulse response (CIR) as a sparse sequence of Diracs pulses, the TOA estimation becomes a problem of parametric spectral inference from observed bandlimited signals.
- To increase resolution without arriving at unrealistic sampling rates, we consider a multiband sampling approach, and propose: (i) a **practical multibranch receiver for the acquisition** and (ii) an **algorithm for multiresolution TOA estimation** based on the ESPRIT algorithm.

## Multiband Sampling of the Radio Channel

- The **multipath radio channel model** assuming  $K$  propagation paths is

$$\tilde{h}(t) = \sum_{k=1}^K \tilde{\alpha}_k \delta(t - \tau_k) \xrightarrow{\text{CTFT}} \tilde{H}(\Omega) = \sum_{k=1}^K \tilde{\alpha}_k e^{-j\Omega\tau_k} \quad (1)$$

where  $\tilde{\alpha}_k \in \mathbb{R}$  and  $\tau_k \in \mathbb{R}_+$  represent the gain and time-delay of the  $k$ th resolvable path.

- **Multiband sampling** assumes probing of the channel (1) by a wide-band sensing signal  $\tilde{s}(t)$  defined by CTFT

$$\tilde{S}(\Omega) = \tilde{S}_i(\Omega), \quad \Omega \in \mathcal{W}_i, i \in [1, L]$$

- $\mathcal{W}_i = [\Omega_i - B_i/2, \Omega_i + B_i/2]$ ,
- $\Omega_i$  is the center frequency,
- $B_i$  is the bandwidth of the  $i$ th sub-band.

- The **multibranch receiver** down-converts the received signal to base-band and performs lowpass filtering and sampling of

$$X_i(\Omega) = G_i(\Omega)H_i(\Omega)S_i(\Omega) + N_i(\Omega)$$

- $G_i(\Omega)$  is the frequency response of the  $i$ th receiver chain,
- $H_i(\Omega)$  and  $S_i(\Omega)$  are the baseband equivalents of  $\tilde{H}(\Omega + \Omega_i)$  and  $\tilde{S}_i(\Omega)$ ,
- $N_i(\Omega)$  is the bandlimited white Gaussian noise.

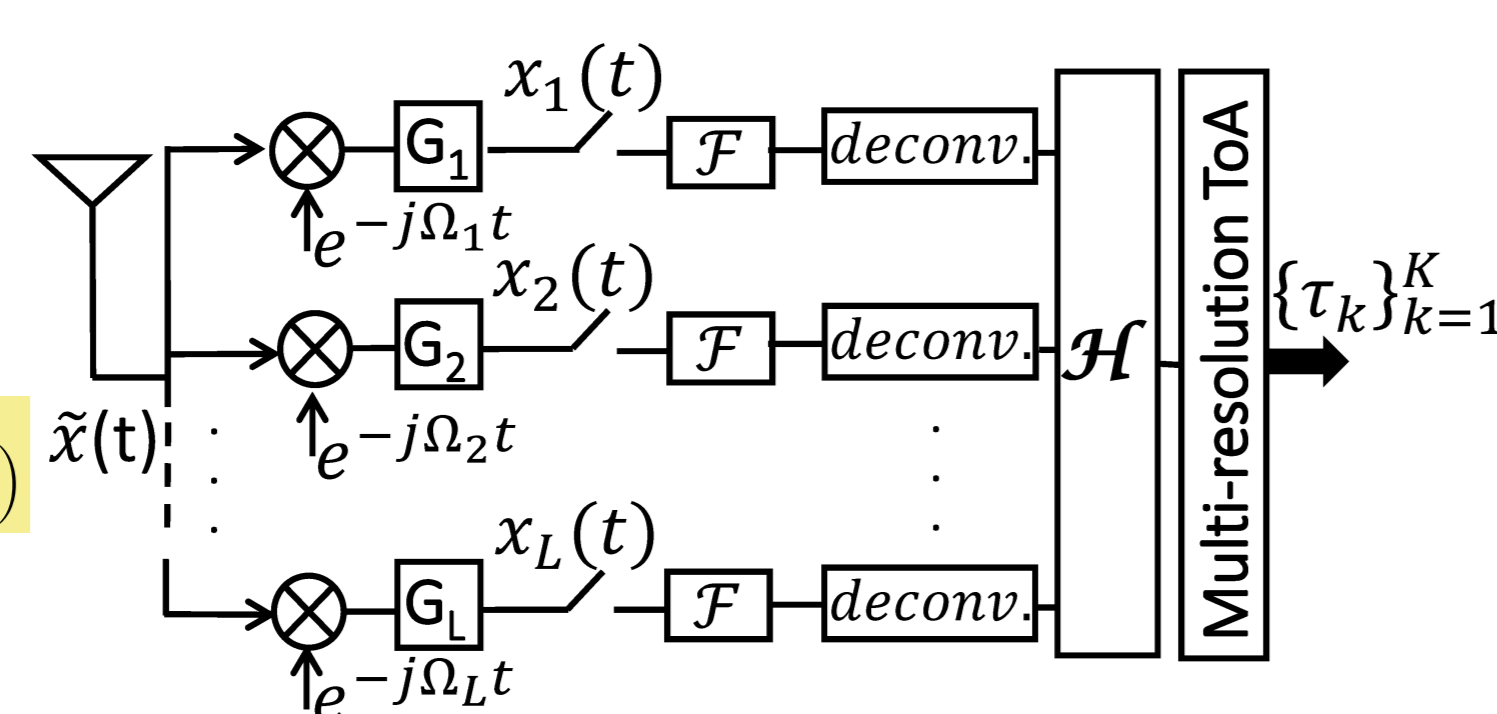
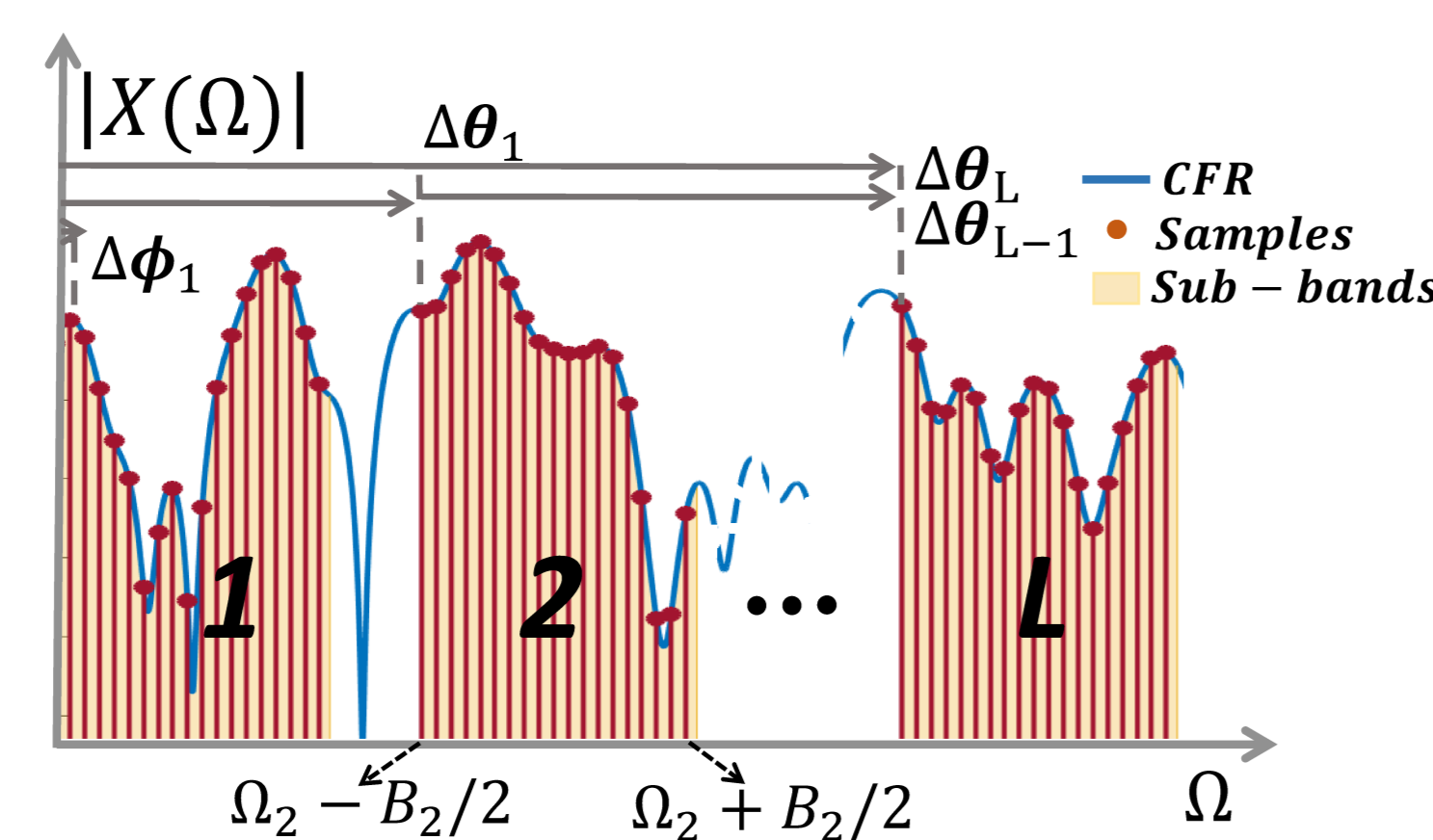
- **Discrete-time data model:** Assuming that  $x_i(t)$  has a finite duration  $T$ , and it satisfies conditions for Nyquist sampling, the data model in the frequency domain is

$$\mathbf{x}_i = \mathbf{h}_i \odot \mathbf{g}_i \odot \mathbf{s}_i + \mathbf{n}_i \quad (2)$$

where  $\mathbf{x}_i$ ,  $\mathbf{g}_i$ ,  $\mathbf{h}_i$ ,  $\mathbf{s}_i$ , and  $\mathbf{n}_i$  are collecting  $N$  samples of  $X_i[n]$ ,  $G_i[n]$ ,  $H_i[n]$ ,  $S_i[n]$  and  $N_i[n]$ , respectively. The elements of  $\mathbf{h}_i$  are

$$H_i[n] = \tilde{H}(n\Omega_t + \Omega_i) = \sum_{k=1}^K \tilde{\alpha}_k e^{-j\Omega_t \tau_k} e^{-jn\Omega_s \tau_k}, \quad n = 0, \dots, N-1$$

where  $\Omega_t = \frac{1}{N}\Omega_s = \frac{2\pi}{T}$ .



- After deconvolution, the channel model can be written as

$$\mathbf{h}_i = \mathbf{M}\boldsymbol{\Theta}_i\boldsymbol{\alpha} + \mathbf{n}'_i, \quad i = 1, \dots, L \quad (3)$$

where

$$\mathbf{M} = \begin{bmatrix} 1 & 1 & \dots & 1 \\ \Phi_1 & \Phi_2 & \dots & \Phi_K \\ \vdots & \vdots & \ddots & \vdots \\ \Phi_1^{N-1} & \Phi_2^{N-1} & \dots & \Phi_K^{N-1} \end{bmatrix}, \quad \boldsymbol{\Theta}_i = \begin{bmatrix} \theta_{i,1} & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & \theta_{i,K} \end{bmatrix}, \quad \boldsymbol{\alpha} = \begin{bmatrix} \tilde{\alpha}_1 \\ \vdots \\ \tilde{\alpha}_K \end{bmatrix},$$

$\Phi_k = e^{-j\phi_k}$ ,  $\phi_k = \Omega_t \tau_k$ ,  $\theta_{i,k} = e^{-j\theta_{i,k}}$ ,  $\theta_{i,k} = \Omega_i \tau_k$ , and  $\mathbf{n}'_i$  represents zero mean additive white Gaussian noise.

## Multiresolution TOA Estimation

- The Hankel matrices of size  $P \times Q$  are constructed from single  $\mathbf{h}_i$  as

$$\mathcal{H}_i = \begin{bmatrix} H_i[0] & H_i[1] & \dots & H_i[Q] \\ H_i[1] & H_i[2] & \dots & H_i[Q+1] \\ \vdots & \vdots & \ddots & \vdots \\ H_i[P-1] & H_i[P] & \dots & H_i[N-1] \end{bmatrix} \quad (4)$$

where  $P = N - Q - 1$ ,  $Q$  is a design parameter and we require  $P > K$  and  $Q \geq K$ .

- To achieve high resolution estimation of the  $\{\tau_k\}_{k=1}^K$  we form a block Hankel matrix

$$\mathcal{H} = \begin{bmatrix} \mathcal{H}_1 \\ \mathcal{H}_2 \end{bmatrix} = \begin{bmatrix} \mathbf{M}' \\ \mathbf{M}'\boldsymbol{\Theta} \end{bmatrix} \boldsymbol{\Theta}_1 \mathbf{A} + \mathbf{N} \quad (5)$$

where  $\mathbf{M}'$  is submatrix of  $\mathbf{M}$ ,  $\boldsymbol{\Theta} = \boldsymbol{\Theta}_L \boldsymbol{\Theta}_1^{-1}$ ,  $\mathbf{A} = [\boldsymbol{\alpha}, \boldsymbol{\Phi}\boldsymbol{\alpha}, \boldsymbol{\Phi}^2\boldsymbol{\alpha}, \dots, \boldsymbol{\Phi}^{Q-1}\boldsymbol{\alpha}]$  and  $\boldsymbol{\Phi} = \text{diag}(\Phi_1, \Phi_2, \dots, \Phi_K)$ .

- Exploiting the **double shift invariance structure** of  $\mathcal{H}$  similar as in the case of MI-ESPRIT: unambiguous estimates of  $\{\tau_k\}_{k=1}^K$  are obtained from  $\boldsymbol{\Phi}$ , while high resolution but ambiguous estimates are obtained from  $\boldsymbol{\Theta}$ .

- $\boldsymbol{\Phi}$  and  $\boldsymbol{\Theta}$  are estimated by finding Least Squares (LS) approximate solutions to

$$\begin{aligned} \mathbf{U}_{\Phi_1} &= \mathbf{J}_{\Phi_1}^{(1)} \mathbf{U} = \begin{bmatrix} \mathbf{M}'' \\ \mathbf{M}''\boldsymbol{\Theta} \end{bmatrix} \boldsymbol{\Theta}_1 \mathbf{T}^{-1}, & \mathbf{U}_{\Theta_1} &= \mathbf{J}_{\Theta_1} \mathbf{U} = \mathbf{M}' \mathbf{T}^{-1}, \\ \mathbf{U}_{\Phi_2} &= \mathbf{J}_{\Phi_2}^{(1)} \mathbf{U} = \begin{bmatrix} \mathbf{M}'' \\ \mathbf{M}''\boldsymbol{\Theta} \end{bmatrix} \boldsymbol{\Phi} \boldsymbol{\Theta}_1 \mathbf{T}^{-1}, & \mathbf{U}_{\Theta_2} &= \mathbf{J}_{\Theta_2} \mathbf{U} = \mathbf{M}' \boldsymbol{\Theta} \mathbf{T}^{-1}, \end{aligned} \quad (6)$$

where  $\mathbf{U}$  is the  $K$  dimensional orthonormal basis of the column span of  $\mathcal{H}$ ,  $\mathbf{M}''$  is a submatrix of  $\mathbf{M}'$ , and the selection matrices are

$$\begin{aligned} \mathbf{J}_{\Phi_1}^{(r)} &= \mathbf{I}_2 \otimes [\mathbf{I}_{P-r} \quad \mathbf{0}_{P-r,r}], & \mathbf{J}_{\Theta_1} &= [\mathbf{1} \quad \mathbf{0}] \otimes \mathbf{I}_P, \\ \mathbf{J}_{\Phi_2}^{(r)} &= \mathbf{I}_2 \otimes [\mathbf{0}_{P-r,r} \quad \mathbf{I}_{P-r}], & \mathbf{J}_{\Theta_2} &= [\mathbf{0} \quad \mathbf{1}] \otimes \mathbf{I}_P. \end{aligned}$$

- The LS solutions of (6) satisfy

$$\begin{aligned} \boldsymbol{\Psi} &:= \mathbf{U}_{\Phi_1}^\dagger \mathbf{U}_{\Phi_2} = \mathbf{T} \boldsymbol{\Phi} \mathbf{T}^{-1}, \\ \boldsymbol{\Upsilon} &:= \mathbf{U}_{\Theta_1}^\dagger \mathbf{U}_{\Theta_2} = \mathbf{T} \boldsymbol{\Theta} \mathbf{T}^{-1}. \end{aligned}$$

- To pair estimates of  $\{\tau_k\}_{k=1}^K$  from  $\boldsymbol{\Phi}$  and  $\boldsymbol{\Theta}$ , we find  $\mathbf{T}$  that is jointly diagonalizing  $\boldsymbol{\Psi}$  and  $\boldsymbol{\Upsilon}$  and estimate

$$\tau_k = \Omega_t^{-1} \phi_k = (\Omega_2 - \Omega_1)^{-1} (\theta_k + 2\pi n_k). \quad (7)$$

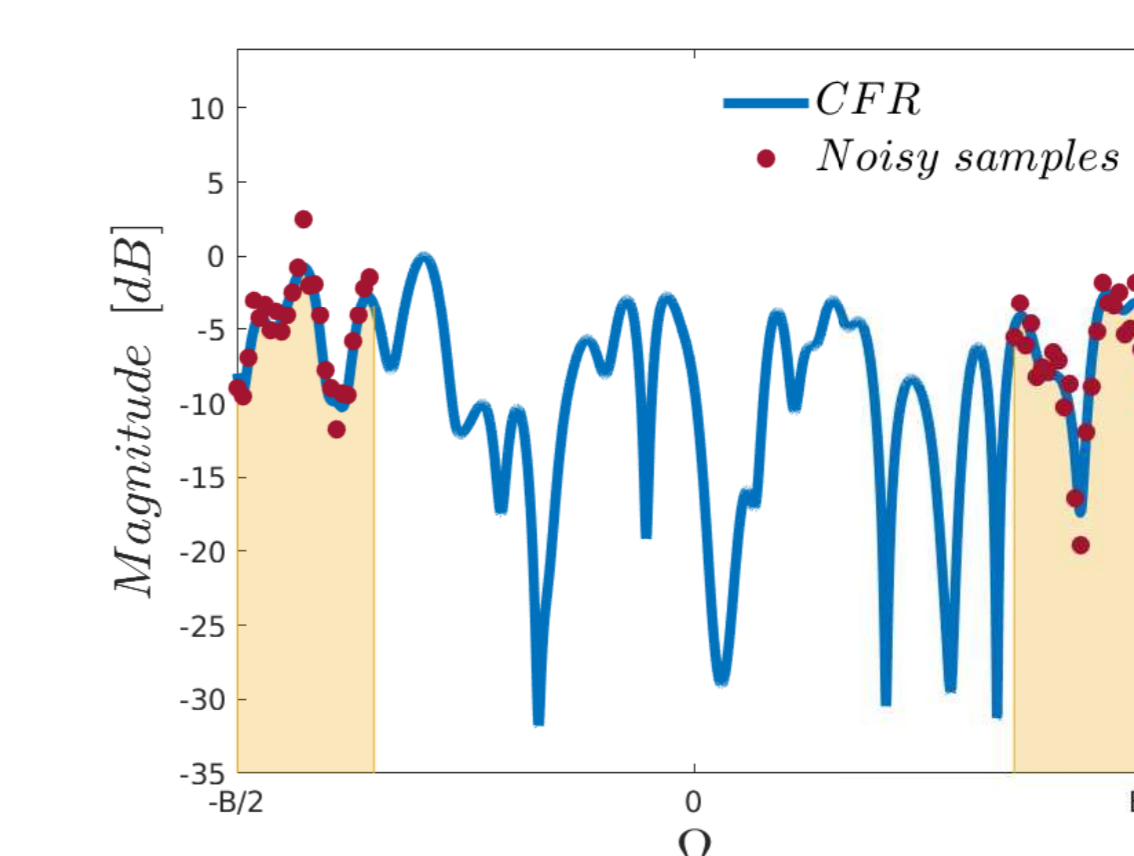
The final estimate of  $\tau_k$  is obtained based on  $\theta_k$ , by finding the best fitting integer satisfying (7), which is  $n_k = \text{round}\{1/2\pi (\Omega_t^{-1}(\Omega_2 - \Omega_1)\phi_k - \theta_k)\}$ .

## Numerical evaluation

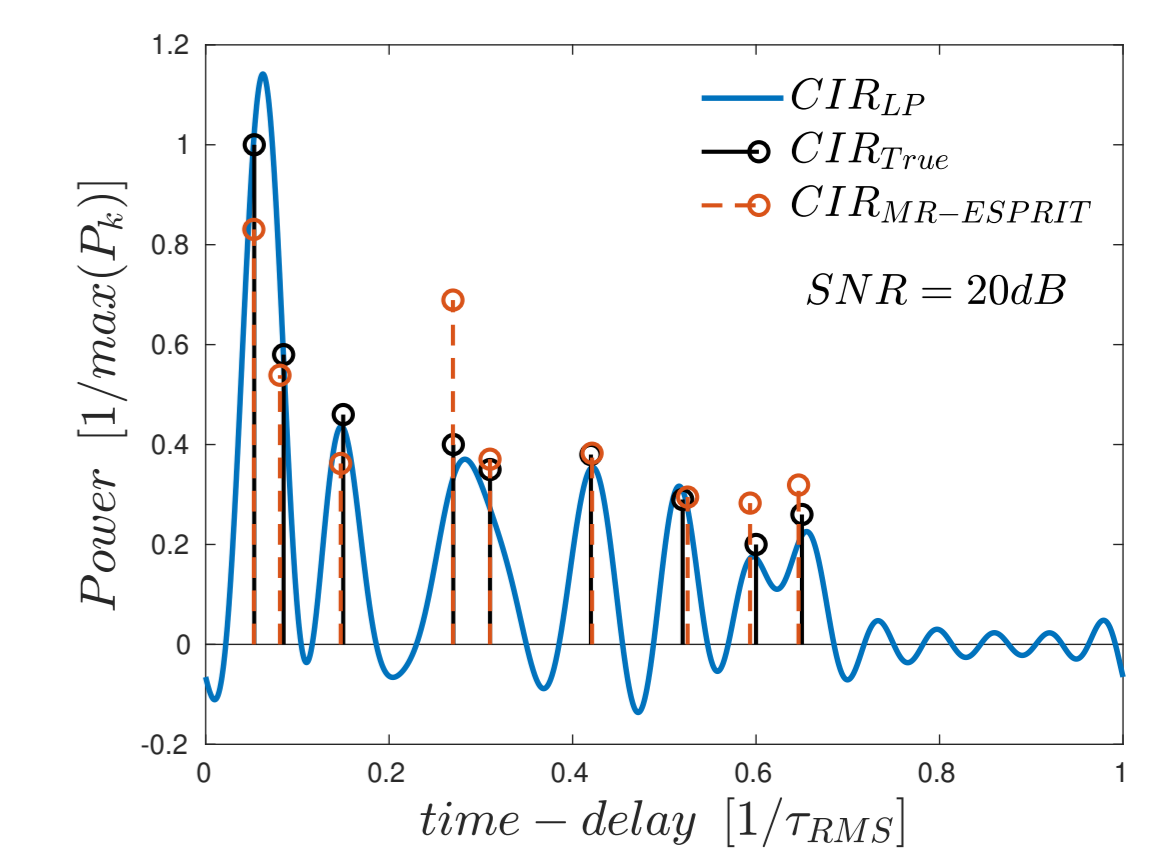
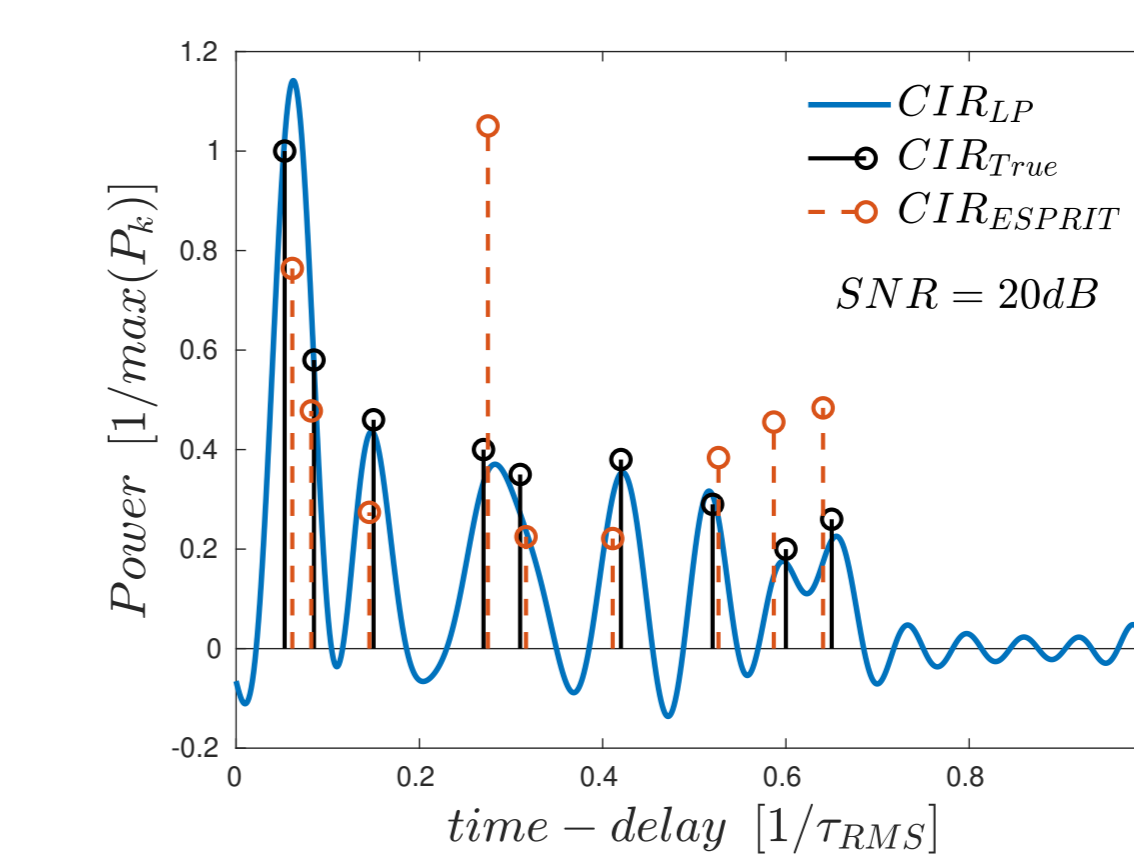
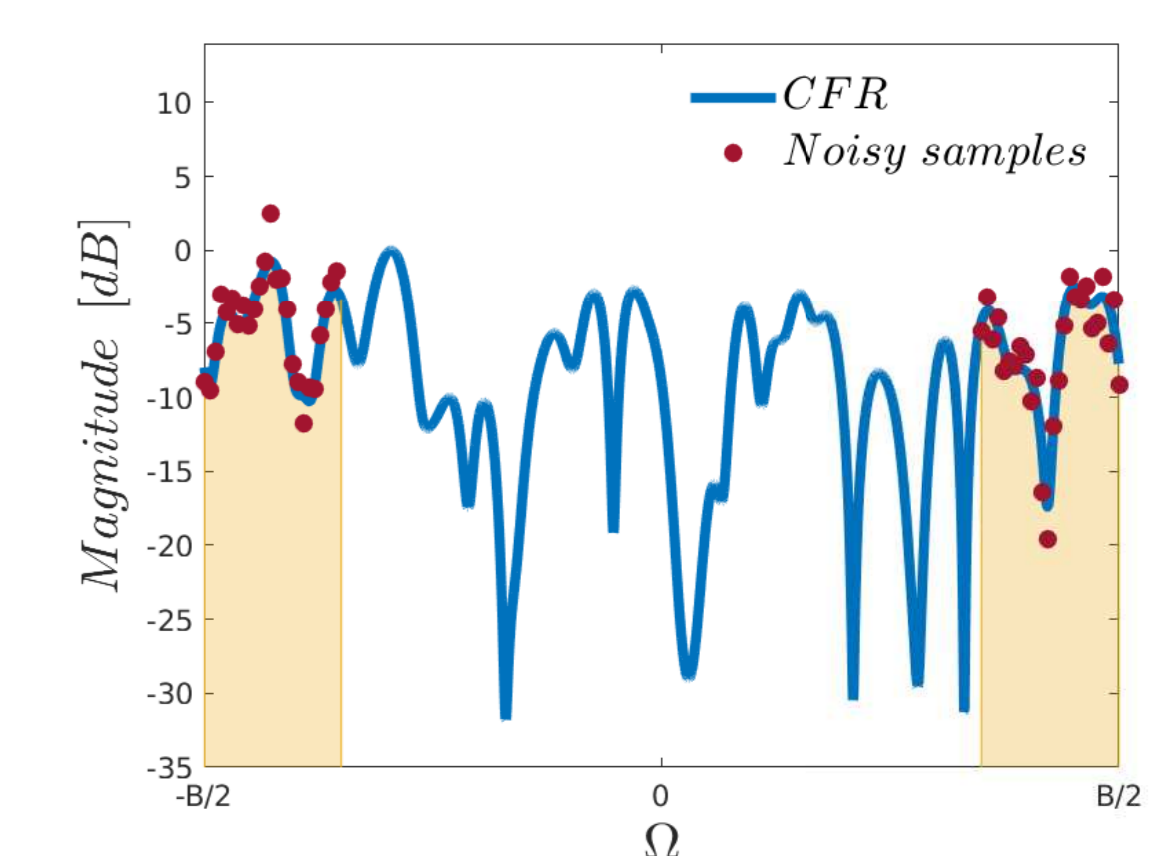
- A standard outdoor UWB channel model with eight dominant multipath components (MPCs) is considered with  $\tau_{RMS} = 40ns$ .
- The continuous time is modeled using a 3 GHz grid, where the channel tap delays are spaced at 333.33 ps.
- The Root Mean Square Error (RMSE) is used as a metric for evaluation, which is obtained over  $10^4$  independent Monte Carlo runs.

### Channel sampling and reconstruction:

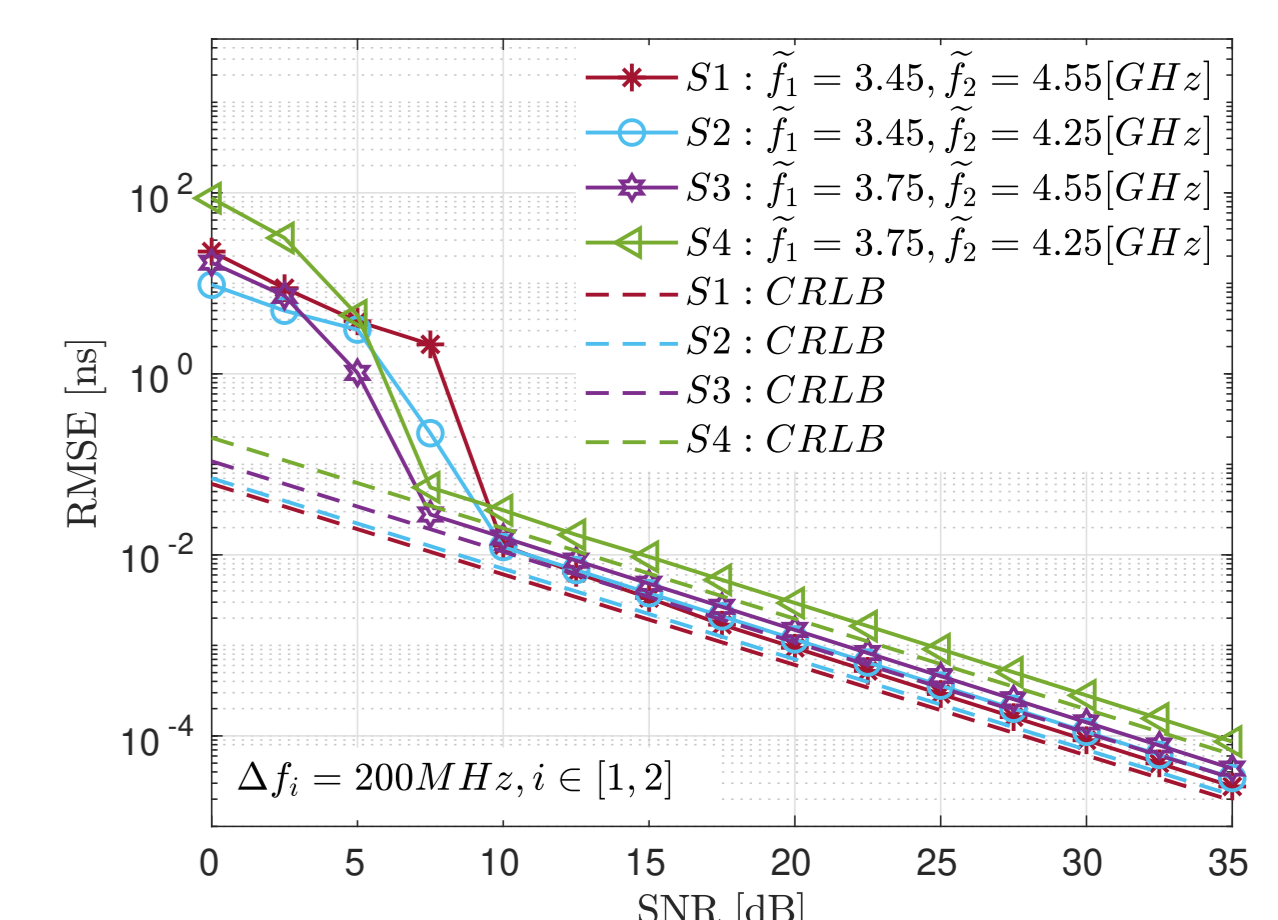
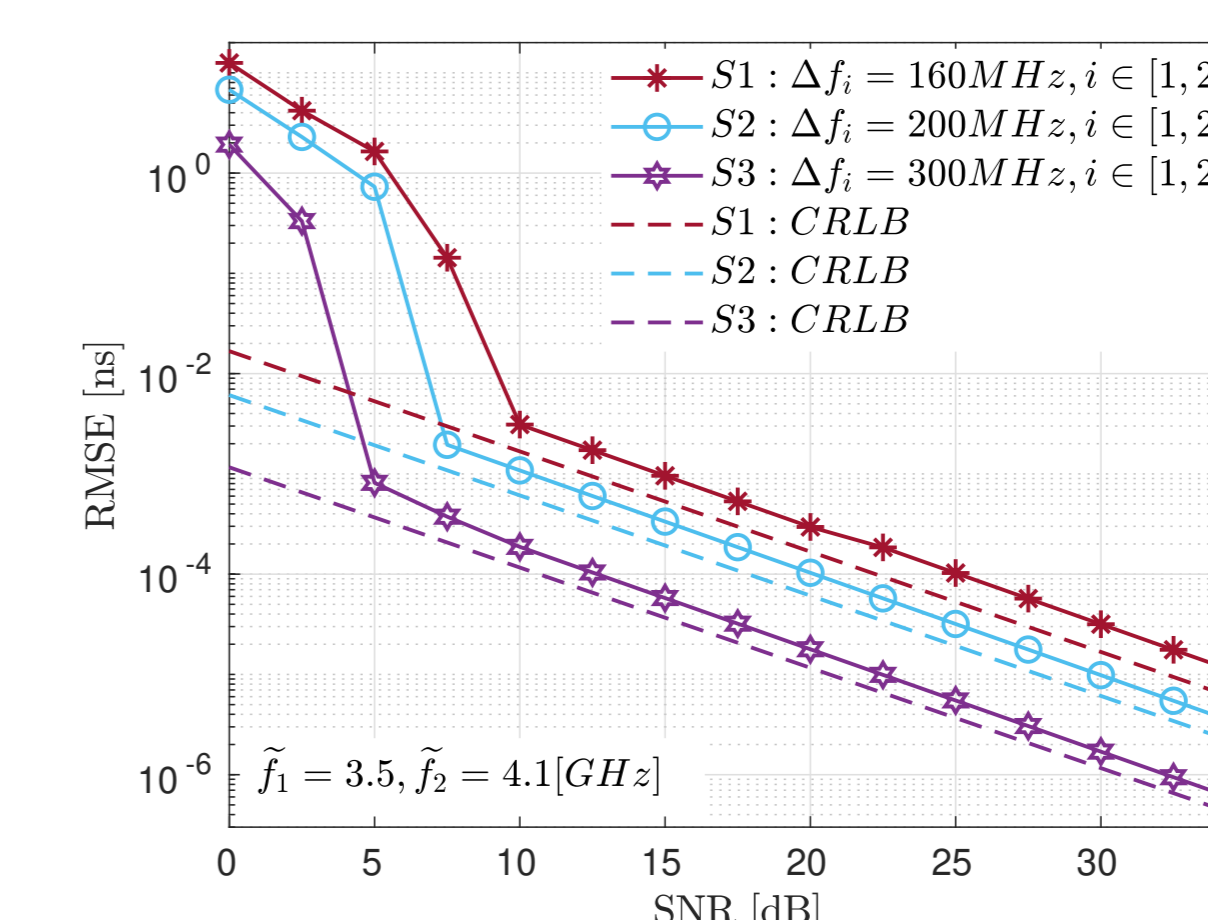
(a): Single band sampling - ESPRIT



(b): Multiband sampling - MR-ESPRIT



### Performance of the algorithm:



## Conclusions

- The block Hankel matrix formed from multiband radio channel samples has a multiple shift invariance structure.
- The invariance structure of a single sub-band provides coarse parameter estimates, while the invariance structure of the lowest against the highest frequency sub-band provides high-resolution, but phase wrapped estimates.
- The multiresolution TOA estimation from multiband channel samples increases the resolution of TOA estimates while it reduces the spectral occupancy and sampling costs.