
Channel Estimation and Sensing for OTFS Modulation using 2D MUSIC

Akshay S. Bondre and Christ D. Richmond

Center of Excellence Meeting

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Signals, Information, Inference, & Learning (SIIl) Group

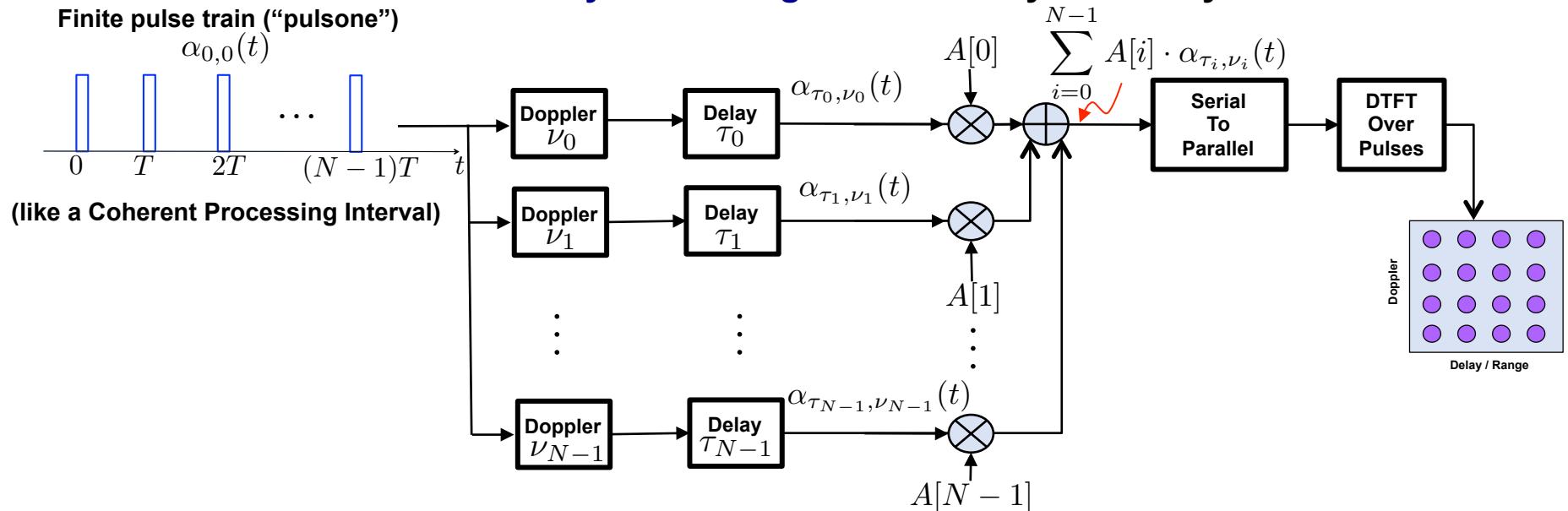




OTFS Modulation from Radar Perspective*



- OTFS modulation transmits “synthetic targets” scaled by comm. symbols



- Doppler-Delay space is populated with communication symbols
 - Similar to pulse amplitude modulation, but in Doppler-Delay space
- Implement via pre- and post-processing block overlays to existing OFDM architecture**

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Akshay S. Bondre and Christ D. Richmond*

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Joint work with Nicolò Michelusi and Ahmed Alkhateeb.



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Outline



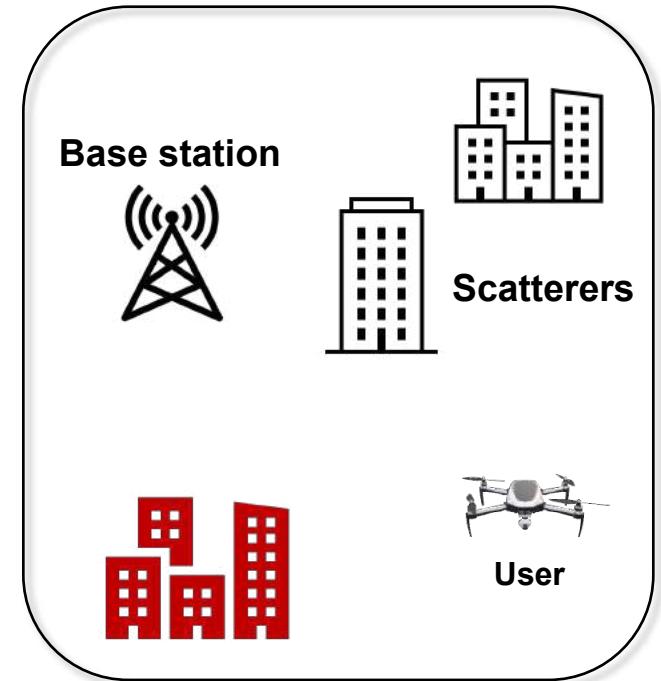
- • **Introduction**
 - MUSIC Method
 - OTFS Channel Estimation via MUSIC
 - 1D v/s 2D MUSIC
 - Numerical Results
 - Summary



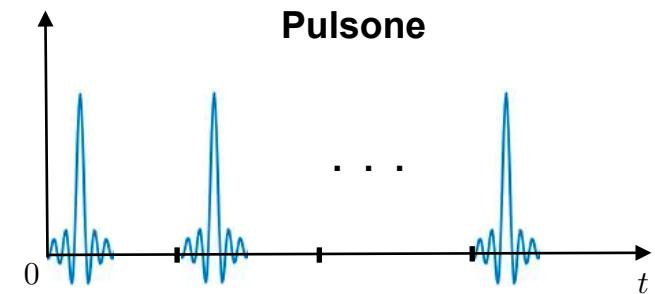
Dynamic Wireless Environments



- The wireless channel is characterized by:
 - Multipath scattering environment
 - User mobility
- The channel changes dynamically due to user mobility.
- Goal: Estimate and track the channel accurately



- The OTFS waveform provides the ability to estimate and track the channel like a radar.



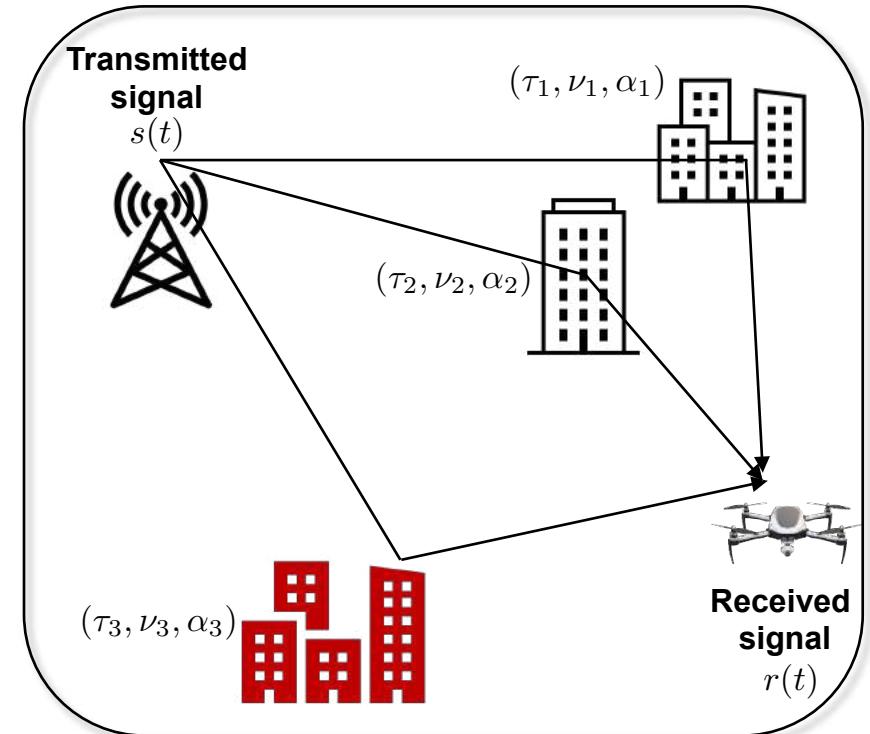


Estimating Sparse Linear Time-Varying Channels



- D multipath components parameterized by $\{(\tau_i, \nu_i, \alpha_i)\}_{i=1}^D$
 - Delays τ_i , Doppler shifts ν_i
 - Path gains α_i
- $r(t) = \sum_{i=1}^D \alpha_i s(t - \tau_i) e^{j2\pi\nu_i t}$
- Goal: Estimate $\{(\tau_i, \nu_i, \alpha_i)\}_{i=1}^D$
- Doppler Rayleigh resolution limit:
$$\Delta\nu = \frac{1}{\text{Signal duration}}$$

Ex.: $\Delta\nu = 1/(1 \text{ ms}) = 1 \text{ kHz}$



- Fractional delays and Dopplers → Need for parametric super-resolution methods



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Line Spectral Estimation



- **Received Signal:** $y[n] = \sum_{i=1}^D \alpha_i e^{j\omega_i n} + w[n], \quad n = 0, \dots, N-1$
 $w[n] \sim \mathcal{CN}(0, \sigma^2)$
- **Goal: Estimate frequencies** $\{\omega_i\}_{i=1}^D$ **and complex gains** $\{\alpha_i\}_{i=1}^D$.
- **Vector form:**
$$\mathbf{y} = \sum_{i=1}^D \alpha_i \mathbf{a}(\omega_i) + \mathbf{w} = \mathbf{A}\boldsymbol{\alpha} + \mathbf{w}$$
 - $\mathbf{y} = [y[0], \dots, y[N-1]]^T, \mathbf{w} = [w[0], \dots, w[N-1]]^T \sim \mathcal{CN}(\mathbf{0}, \sigma^2 \mathbf{I})$
 - $\mathbf{a}(\omega_i) = [e^{j\omega_i(0)}, e^{j\omega_i(1)}, \dots, e^{j\omega_i(N-1)}]^T, \mathbf{A} = [\mathbf{a}(\omega_1), \dots, \mathbf{a}(\omega_D)]$
and $\boldsymbol{\alpha} = [\alpha_1, \alpha_2, \dots, \alpha_D]^T$.
- **Max. likelihood (ML) solution:** D - dimensional search
 - **Pros:** Achieves Cramér-Rao bound asymptotically
 - **Cons:** Computationally expensive



The MUSIC (Multiple Signal Classification) Method



- MUSIC only requires a single 1D search.
- $\mathbf{y} = \mathbf{A}\boldsymbol{\alpha} + \mathbf{w} \rightarrow$ Covariance matrix
$$\mathbf{K} = E [\mathbf{y}\mathbf{y}^H]$$
- $\{\lambda_i, \mathbf{v}_i\}_{i=1}^N$: Eigenvalues and eigenvectors of \mathbf{K} , $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_N$
 - Eigenvector matrix: $\mathbf{V} = [\underbrace{\mathbf{v}_1, \dots, \mathbf{v}_D}_{\text{Spans the column space of } \mathbf{A}}, \underbrace{\mathbf{v}_{D+1}, \dots, \mathbf{v}_N}_{\text{G Noise subspace matrix}}]$
- The true frequency values $\omega_i, i = 1, \dots, D$ are the only solutions of

$$\mathbf{a}^H(\omega_i) \mathbf{G} \mathbf{G}^H \mathbf{a}(\omega_i) = 0$$

Knowledge of number of scatterers D required



The MUSIC Method

- Practically, the covariance matrix must be estimated from data
 - Single snapshot case: Consider subsequences of \mathbf{y} as snapshots

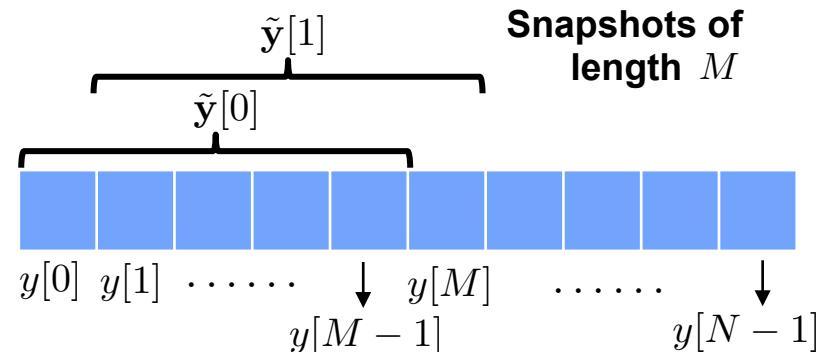
- Let $\tilde{\mathbf{y}}(l), l = 0, \dots, N_{\text{snap}} - 1$
 - N_{snap} snapshots of length M

- Sample covariance matrix

$$\hat{\mathbf{K}} = \frac{1}{N_{\text{snap}}} \sum_{l=0}^{N_{\text{snap}}-1} \tilde{\mathbf{y}}(l) \tilde{\mathbf{y}}(l)^H, \text{ noise subspace matrix } \hat{\mathbf{G}}$$

- Spectral MUSIC: $\{\hat{\omega}_i\}_{i=1}^D$ are the D peaks of the pseudospectrum

$$P_{\text{MUSIC}}(\omega) = \frac{1}{\mathbf{a}^H(\omega) \hat{\mathbf{G}} \hat{\mathbf{G}}^H \mathbf{a}(\omega)} \rightarrow 1\text{D search}$$





Outline

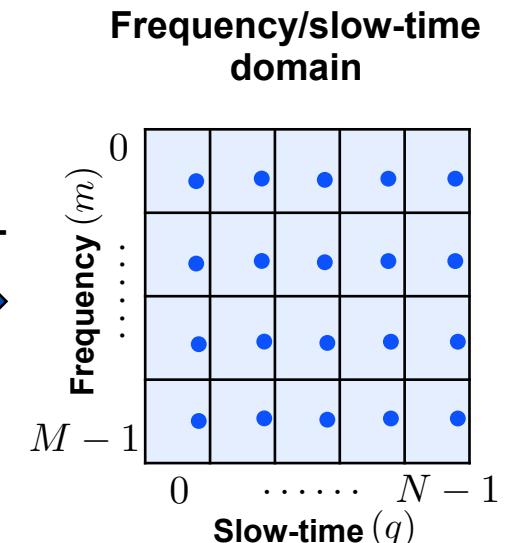
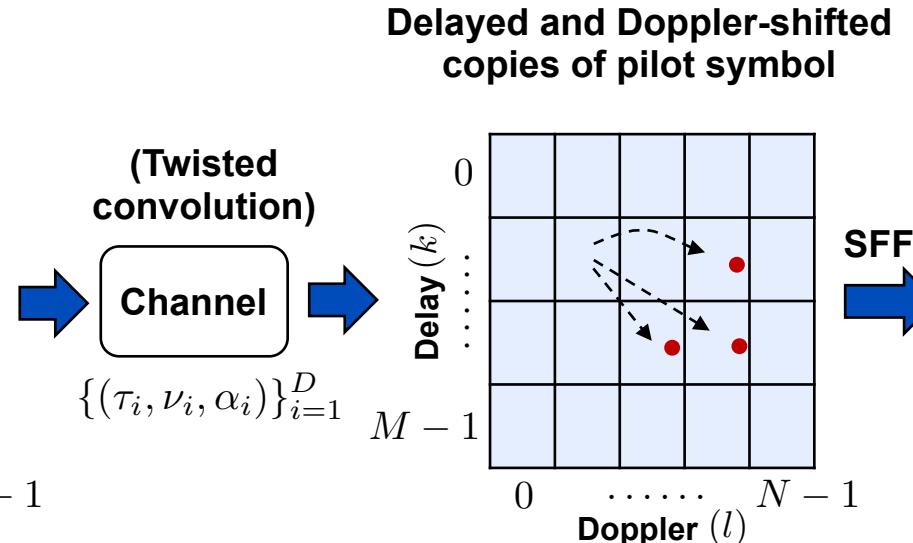
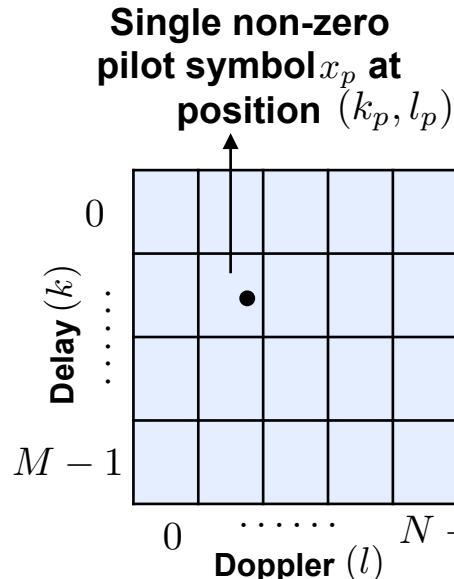


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OTFS Signal Model

- $M \times N$ OTFS delay-Doppler grid:



Transmitter

Receiver

Sum of 2D complex exponentials

x_p, k_p, l_p
are known pilot parameters

$$R[m, q] = \frac{x_p}{\sqrt{MN}} \sum_{i=1}^D \alpha_i e^{-j2\pi\left(\frac{k_p}{M} + \frac{\tau_i}{T}\right)m} e^{j2\pi\left(\frac{l_p}{N} + \nu_i T\right)q}$$

$$\begin{aligned} m &= 0, \dots, M-1 \\ q &= 0, \dots, N-1 \end{aligned}$$



2D MUSIC in the Frequency/Slow-time Domain



- Frequency/slow-time domain signal:

$$R[m, q] = \underbrace{\frac{x_p}{\sqrt{MN}} \sum_{i=1}^D \alpha_i e^{-j2\pi\left(\frac{k_p}{M} + \frac{\tau_i}{T}\right)m} e^{j2\pi\left(\frac{l_p}{N} + \nu_i T\right)q}}_{[\mathbf{R}]_{m,q}} + \underbrace{W[m, q]}_{[\mathbf{W}]_{m,q}}$$

Noise term

$$W[m, q] \sim \mathcal{CN}(0, \sigma^2)$$

$$m = 0, \dots, M - 1$$

$$q = 0, \dots, N - 1$$

- Matrix form: $\mathbf{R} = \frac{x_p}{\sqrt{MN}} \sum_{i=1}^D \alpha_i \mathbf{B}(\tau_i, \nu_i) + \mathbf{W}$
- Vectorizing,

$$\mathbf{y} = \text{vec}(\mathbf{R}) = \frac{x_p}{\sqrt{MN}} \sum_{i=1}^D \alpha_i \mathbf{b}(\tau_i, \nu_i) + \mathbf{w}$$

$$\mathbf{b}(\tau, \nu) = \mathbf{b}_1(\nu) \otimes \mathbf{b}_2(\tau) = \begin{bmatrix} 1 \\ e^{j2\pi\left(\frac{l_p}{N} + \nu T\right)} \\ \vdots \\ e^{j2\pi\left(\frac{l_p}{N} + \nu T\right)(N-1)} \end{bmatrix} \otimes \begin{bmatrix} 1 \\ e^{-j2\pi\left(\frac{k_p}{M} + \frac{\tau}{T}\right)} \\ \vdots \\ e^{-j2\pi\left(\frac{k_p}{M} + \frac{\tau}{T}\right)(M-1)} \end{bmatrix}$$

- Thus, we can apply the MUSIC method for estimating $\{\tau_i, \nu_i\}_{i=1}^D$:

$$\mathbf{b}^H(\tau, \nu) \mathbf{G} \mathbf{G}^H \mathbf{b}(\tau, \nu) = 0$$



2D Root-MUSIC



- **2D spectral-MUSIC:**

- 2D search over (τ, ν) grid for finding D peaks of:

$$P_{\text{MUSIC}}(\tau, \nu) = \frac{1}{\mathbf{b}^H(\tau, \nu)\mathbf{E}_n\mathbf{b}(\tau, \nu)}$$

$$\mathbf{b}^H(\tau, \nu)\mathbf{E}_n\mathbf{b}(\tau, \nu) = 0$$

$$\mathbf{G}\mathbf{G}^H = \mathbf{E}_n$$

- **2D root-MUSIC: Let** $z_1 = e^{j2\pi\left(\frac{l_p}{N} + \nu T\right)}$ **and** $z_2 = e^{-j2\pi\left(\frac{k_p}{M} + \frac{\tau}{T}\right)}$

$$\mathbf{b}(\tau, \nu) = \tilde{\mathbf{b}}(z_2, z_1) = \tilde{\mathbf{b}}_1(z_1) \otimes \tilde{\mathbf{b}}_2(z_2), \text{ where } \tilde{\mathbf{b}}_1(z_1) = \begin{bmatrix} 1 \\ z_1 \\ \vdots \\ z_1^{N-1} \end{bmatrix} \quad \tilde{\mathbf{b}}_2(z_2) = \begin{bmatrix} 1 \\ z_2 \\ \vdots \\ z_2^{M-1} \end{bmatrix}$$

$$\mathbf{b}^H(\tau, \nu)\mathbf{E}_n\mathbf{b}(\tau, \nu) = \tilde{\mathbf{b}}^T(z_2^{-1}, z_1^{-1})\mathbf{E}_n\tilde{\mathbf{b}}(z_2, z_1) = 0 \rightarrow \text{Polynomial in two variables}(z_1, z_2)$$

- **Solutions for (z_1, z_2) can be obtained by solving two separate polynomial rooting problems*.**

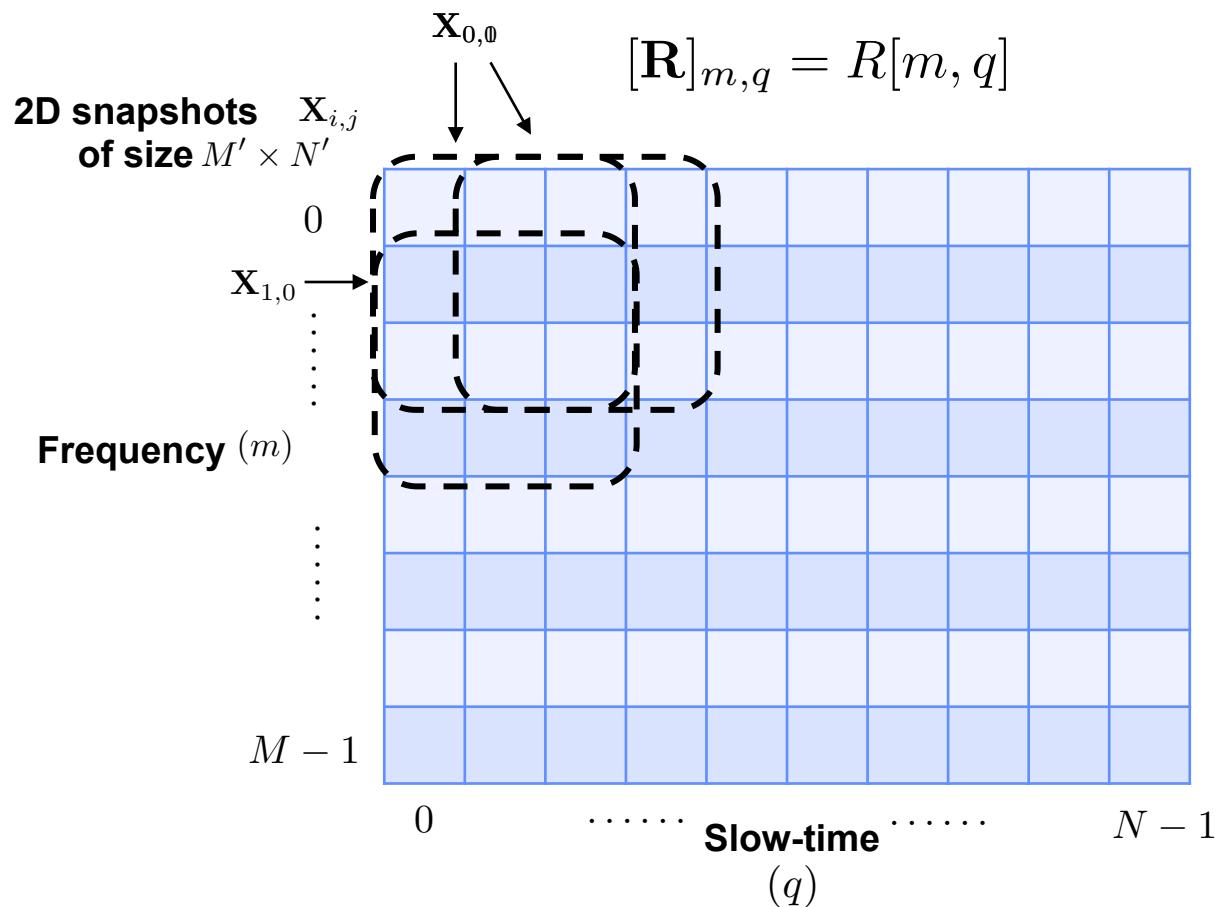
- 2D search not required \rightarrow Computationally feasible
(Consequence of the complex exponential steering vector structure)



Snapshots for Covariance Estimation



- 2D subsequences are considered as snapshots



- Consecutive snapshots shifted by one sample

Vectorized version

$$\bullet \mathbf{y} = \text{vec}(\mathbf{R}) = \frac{x_p}{\sqrt{MN}} \sum_{i=1}^D \alpha_i \mathbf{b}(\tau_i, \nu_i) + \mathbf{w}$$

- Snapshots: $\mathbf{x}_{i,j} = \text{vec}(\mathbf{X}_{i,j})$
 $i = 0, \dots, M - M', j = 0, \dots, N - N'$
 $N_{\text{snap}} = (M - M' + 1)(N - N' + 1)$

Sample covariance matrix

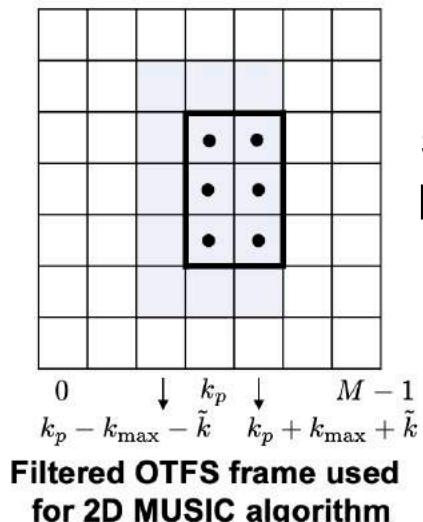
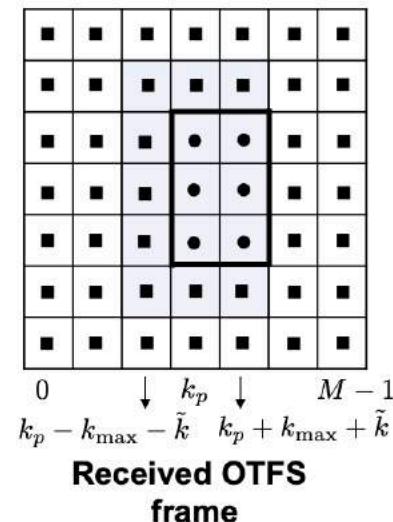
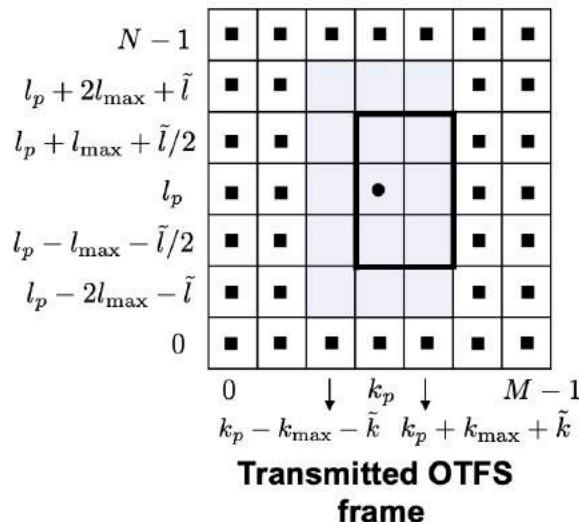
$$\hat{\mathbf{K}} = \frac{1}{N_{\text{snap}}} \sum_{i=0}^{M-M'} \sum_{j=1}^{N-N'} \mathbf{x}_{i,j} \mathbf{x}_{i,j}^H$$



2D MUSIC for Embedded Pilot Structure



- Pilot symbol separated from data symbols with a guard interval
 - Max. Doppler tap $l_{\max} = \nu_{\max} T_w / 2$, Max. delay tap $k_{\max} = \tau_{\max} B$
 - \tilde{k} and \tilde{l} chosen to control interference between data and pilot symbols
 - : Pilot symbols
 - : Data symbols
 - : Guard symbols





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Cramér-Rao Bound (CRB) for Delay-Doppler Estimation



- Unknown parameter vector: $\theta = [\theta_1, \dots, \theta_P]^T$
- Received signal having the form: $\mathbf{y} = \mu(\theta) + \mathbf{w}$ $\mathbf{w} \sim \mathcal{CN}(\mathbf{0}, \sigma^2 \mathbf{I})$

- Fisher information matrix (FIM) $\mathbf{J}(\theta)$:

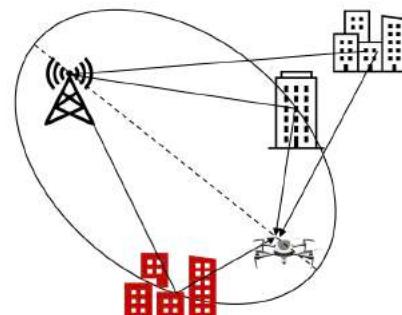
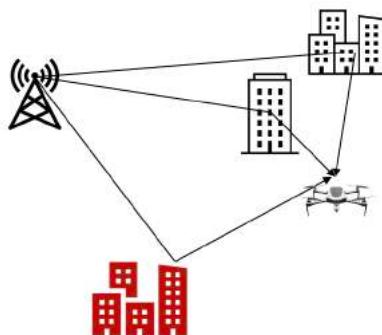
$$[\mathbf{J}(\theta)]_{i,k} = \frac{2}{\sigma^2} \cdot \text{Re} \left\{ \left(\frac{\partial \mu}{\partial \theta_i} \right)^H \left(\frac{\partial \mu}{\partial \theta_k} \right) \right\} \quad \mathbf{y} = \frac{x_p}{\sqrt{MN}} \sum_{i=1}^D \alpha_i \mathbf{b}(\tau_i, \nu_i) + \mathbf{w}$$

- CRB matrix $= \mathbf{J}^{-1}(\theta)$
 - Diagonal elements of CRB provide a lower bound on variance of any unbiased estimator $\rightarrow \mathbb{E}[(\hat{\theta}_i - \theta_i)^2] \geq [\mathbf{J}^{-1}(\theta)]_{i,i}$
- Non-diagonal elements \rightarrow Possible coupling between unknown parameters (Motivation for joint estimation)



1D v/s 2D Estimation

- Turns out the coupling between delays and Dopplers is negligible*
- Delays and Dopplers can be estimated individually, but pairing procedures are necessary to find the correct delay-Doppler pair
 - 2D MUSIC pairs delays and Dopplers automatically
- Estimating Dopplers for a specific delay requires the knowledge of the number of Dopplers with that particular delay



Two scatterers with same delay and different Doppler shift

Frequent estimation of model order required as the channel changes due to user mobility



Outline

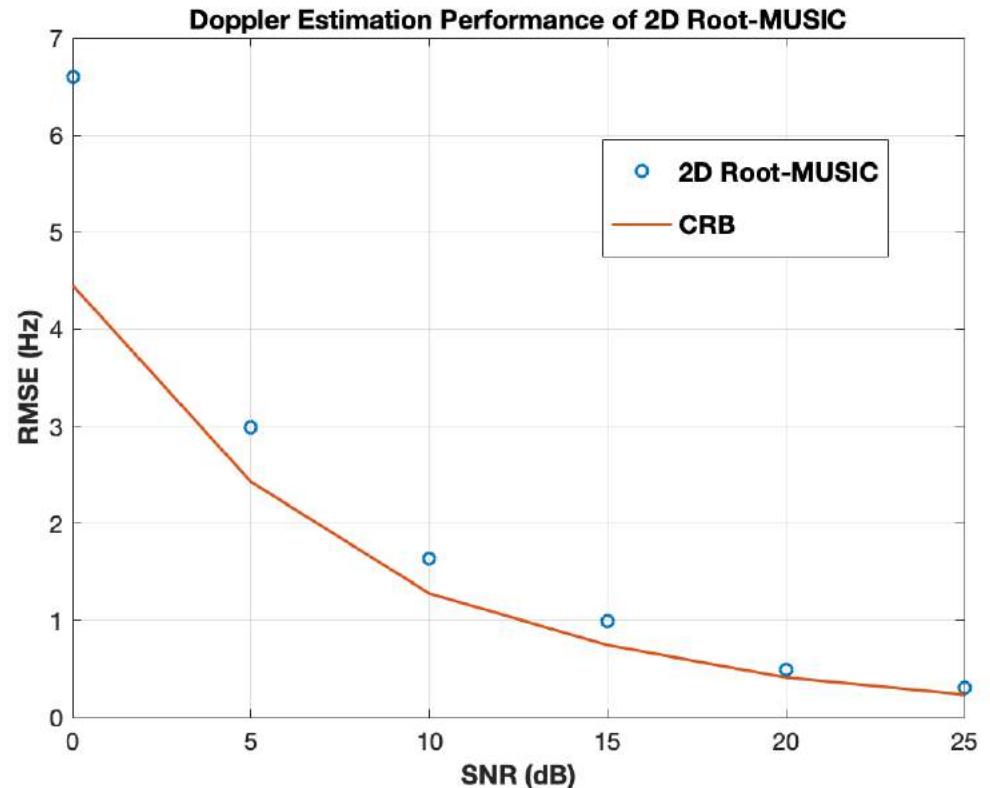


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Numerical Results

- **Simulation parameters:**
 - $M = N = 64$
 - $B = 2 \text{ MHz}$
 - $D = 4 \text{ paths}$
 - $\tau_{\max} = 10T_s$
 - $\nu_{\max} = 3000 \text{ Hz}$
 - $\tau_i \sim \mathcal{U}(0, \tau_{\max})$
 - $\nu_i \sim \mathcal{U}(-\nu_{\max}, \nu_{\max})$
 - $h_i \sim \mathcal{CN}(0, 1)$
- **2D Root MUSIC method*, no 2D search involved.**





Outline



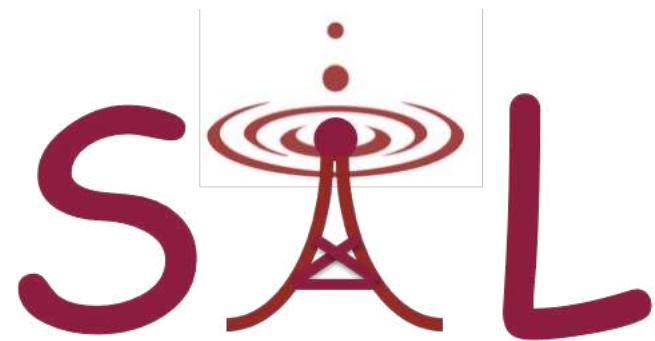
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Summary



- Estimating delays and Doppler shifts reduces to a 2D sinusoidal frequency estimation problem in the frequency/slow-time domain.
- Computationally feasible 2D root-MUSIC can be applied.
- 2D MUSIC can be applied to an embedded pilot OTFS frame by filtering out the data symbols.
- 2D MUSIC automatically gives paired delay-Doppler estimates and requires less frequent model order estimation.
- Future work:
 - Exploring more general OTFS pilot sequences and the application of 2D root-MUSIC-type methods to such pilot sequences



Thank You



Backup Slides

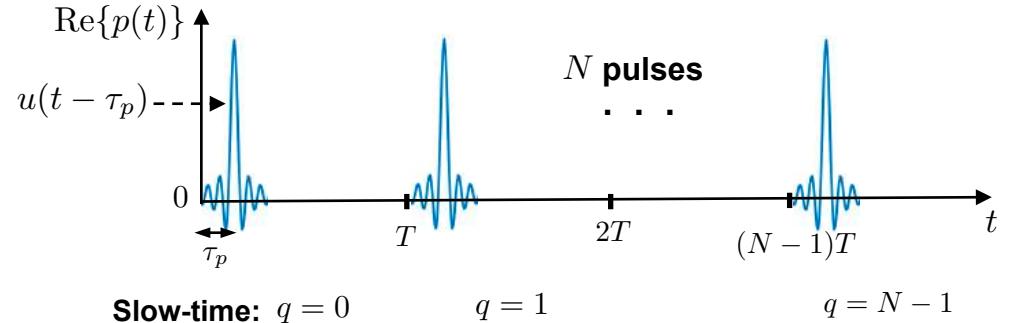


OTFS Received Signal Model



- Transmitted pilot signal:

$$p(t) = x_p \sum_{q=0}^{N-1} u(t - \tau_p - qT) e^{j2\pi\nu_p qT}$$



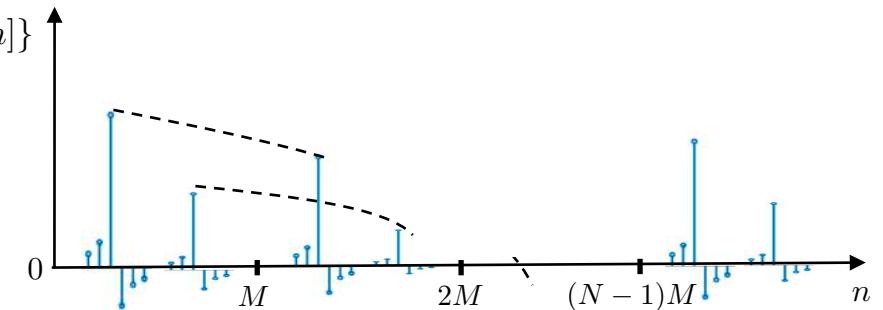
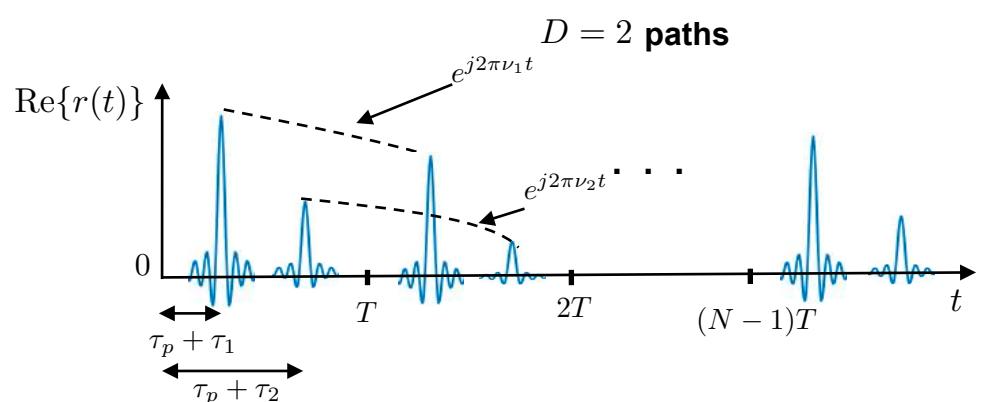
- Received signal:

$$r(t) = \sum_{i=1}^D \alpha_i p(t - \tau_i) e^{j2\pi\nu_i t}$$

- Sampling at $T_s = 1/B$,

$$r[n] = r(nT_s) \approx \sum_{i=1}^D \alpha_i p(nT_s - \tau_i) e^{j2\pi\nu_i \lfloor \frac{n}{M} \rfloor M T_s}$$

$$n = 0, \dots, MN - 1$$





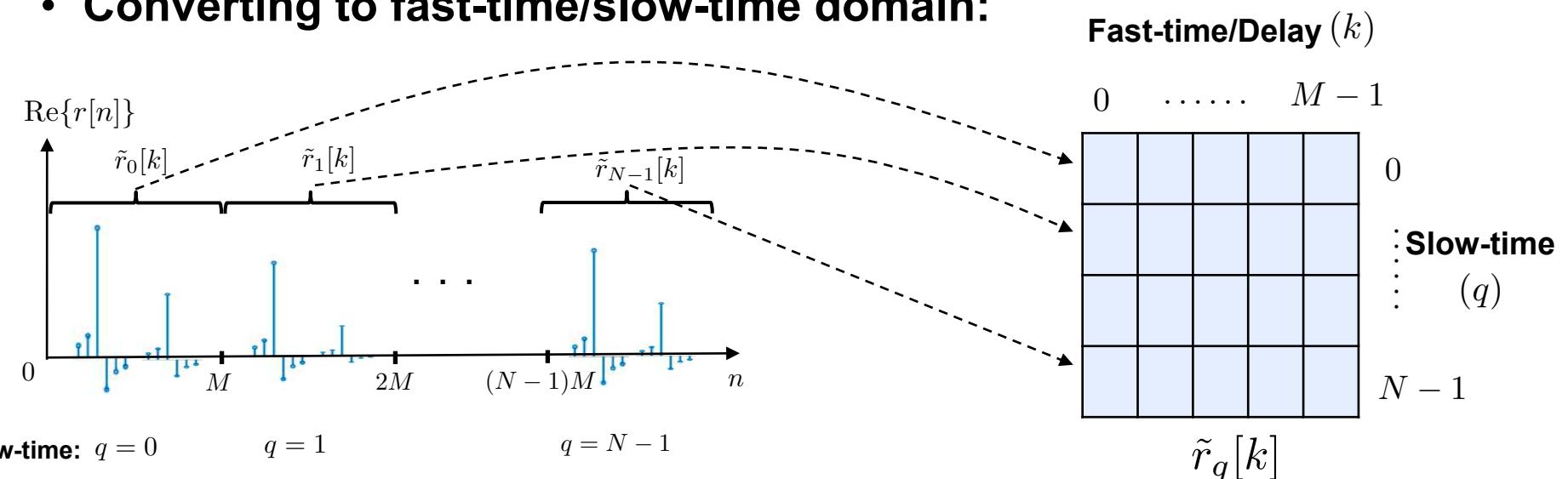
Fast-time/Slow-time Domain



- Received signal:

$$r[n] = r(nT_s) \approx \sum_{i=1}^D \alpha_i p(nT_s - \tau_i) e^{j2\pi\nu_i \lfloor \frac{n}{M} \rfloor MT_s} \quad n = 0, \dots, MN - 1$$

- Converting to fast-time/slow-time domain:



$$\tilde{r}_q[k] \approx x_p \sum_{i=1}^D \alpha_i u(kT_s - \tau_p - \tau_i) e^{j2\pi(\nu_p + \nu_i)qT}$$

$$\begin{aligned} k &= 0, \dots, M - 1 \\ q &= 0, \dots, N - 1 \end{aligned}$$



Frequency/Slow-time Domain



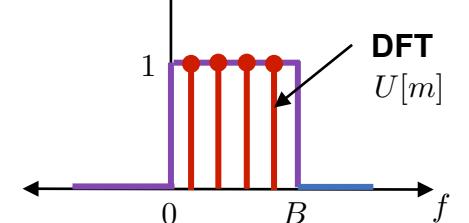
$$\tilde{r}_q[k] = x_p \sum_{i=1}^D \alpha_i u(kT_s - \tau_p - \tau_i) e^{j2\pi\nu_p qT} e^{j2\pi\nu_i qT} \quad k = 0, \dots, M-1 \\ q = 0, \dots, N-1$$

- DFT of fast-time/slow-time matrix along fast-time (delay):

$$\tilde{r}_q[k] \xrightarrow[k]{\text{DFT}} R[m, q] = \sum_{k=0}^{M-1} \tilde{r}_q[k] e^{-j2\pi km/M}$$

$$u(t) \longleftrightarrow U(f) = \begin{cases} 1 & f \in [0, B] \\ 0 & \text{otherwise.} \end{cases}$$

- Frequency variable $m = 0, \dots, M-1$
- $R[m, q]$: Frequency/slow-time domain



- Using the time-shift property of the DFT,

$$u[k] = u(kT_s) \longleftrightarrow U[m] = 1 \quad m = 0, \dots, M-1$$

$$R[m, q] = x_p \sum_{i=1}^D \alpha_i e^{-j2\pi \frac{(\tau_p + \tau_i)m}{T}} e^{j2\pi(\nu_p + \nu_i)qT} \quad m = 0, \dots, M-1 \\ q = 0, \dots, N-1$$