

ADMM-Based Delay-Doppler Domain Channel Estimation for OTFS Systems

Can Zheng¹, Xin Wang², Chung G. Kang¹

¹Korea University, ²NUDT, China

Oct 20, 2025



Outline

1. Introduction
2. System Model
3. Proposed Method
3. Simulation Results
4. Conclusion

Introduction

- ✓ **Background**
- ✓ **Previous Works**
- ✓ **Motivations**
- ✓ **Contribution**

Background

- OFDM fails in high-Doppler scenarios (V2X, NTN) → ICI.
- OTFS operates in Delay-Doppler (DD) domain.
- Challenge: Accurate DD-domain channel estimation under noise and unknown sparsity.

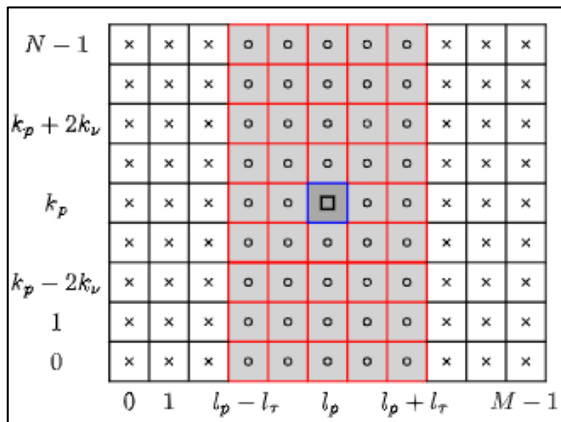
Aug, 2025: 6G meeting settles on the same old air interface.



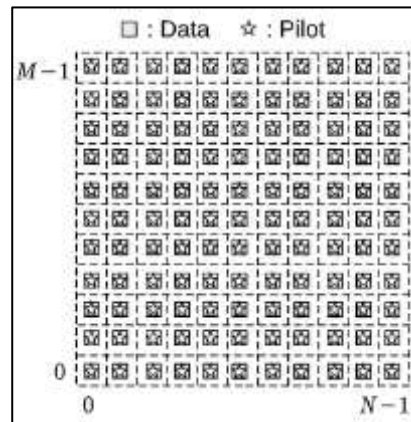
Previous Works (1/2)

- Channel Estimation

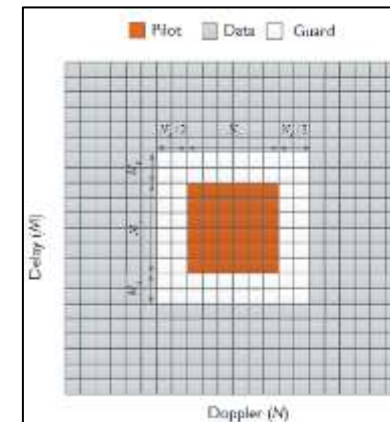
- Embedded pilot-aided: low spectral efficiency
- Superimposed pilot-aided: interference-induced
- Compressed sensing: Orthogonal matching pursuit (OMP), sparsity adaptive matching pursuit (SAMP), smoothed l_0 (SLO), ...



Embedded Pilot



Superimposed Pilot



Pseudo-random Pilot

* P. Raviteja, K. T. Phan and Y. Hong, "Embedded Pilot-Aided Channel Estimation for OTFS in Delay-Doppler Channels," in IEEE Transactions on Vehicular Technology, vol. 68, no. 5, pp. 4906-4917, May 2019.

* H. B. Mishra, P. Singh, A. K. Prasad and R. Budhiraja, "OTFS Channel Estimation and Data Detection Designs With Superimposed Pilots," in IEEE Transactions on Wireless Communications, vol. 21, no. 4, pp. 2258-2274, April 2022.

* W. Shen, L. Dai, J. An, P. Fan and R. W. Heath, "Channel Estimation for Orthogonal Time Frequency Space (OTFS) Massive MIMO," in IEEE Transactions on Signal Processing, vol. 67, no. 16, pp. 4204-4217, 15 Aug.15, 2019.

Previous Works (2/2): SLO Channel Estimation

- **Problem:** OTFS channel estimation \rightarrow Sparse recovery: $\mathbf{y} = \Psi\mathbf{h} + \mathbf{w}$
- **SLO idea:** approximate non-convex l_0 -norm with smooth Gaussian $f_\sigma(h_i) = e^{-\frac{|h_i|^2}{2\sigma^2}}$,
minimize $F_\sigma(\mathbf{h}) = \sum_i f_\sigma(h_i)$ via gradient descent
- **Advantages**
 - No prior knowledge of sparsity
 - Fast reconstruction, low complexity
- **Limitations**
 - Non-convex, may fall into local optima, no guarantee of global convergence
 - Parameter-sensitive (σ , step size μ_0)

Motivation and Contribution

Motivations

- DD channel exhibits sparse structure → suitable for compressed sensing.
- Existing CS methods (OMP, SLO, BP): need sparsity prior, or lack stability.

Contributions

- Keep the CS framework, but solve it convexly with **Alternating Direction Method of Multipliers (ADMM)**.
 - ℓ_1 -regularized convex optimization formulation.
 - Pilot structure satisfying incoherence.
 - ADMM-based solver: no sparsity prior, stable convergence.
 - Superior NMSE, BER, runtime tradeoff.

System Model

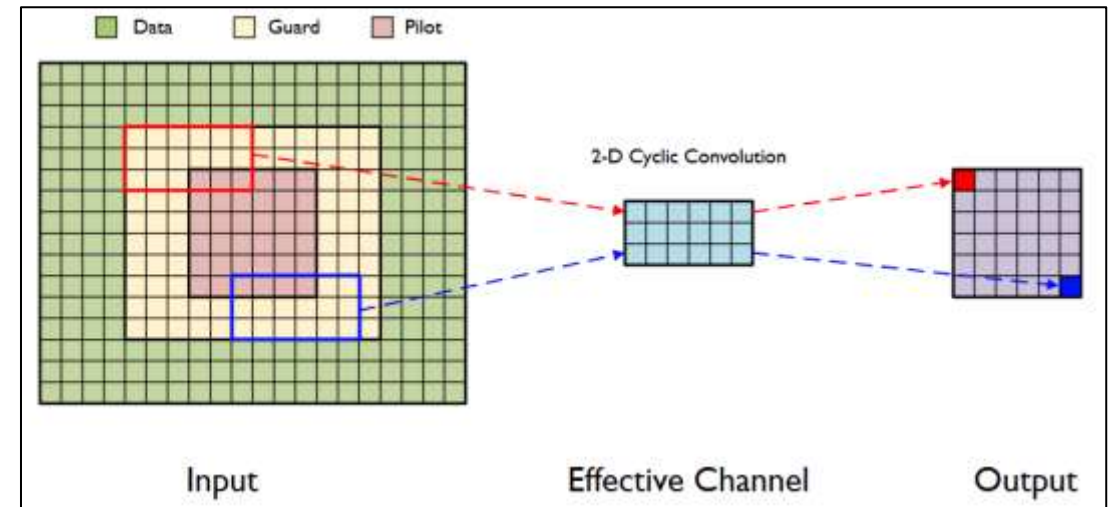
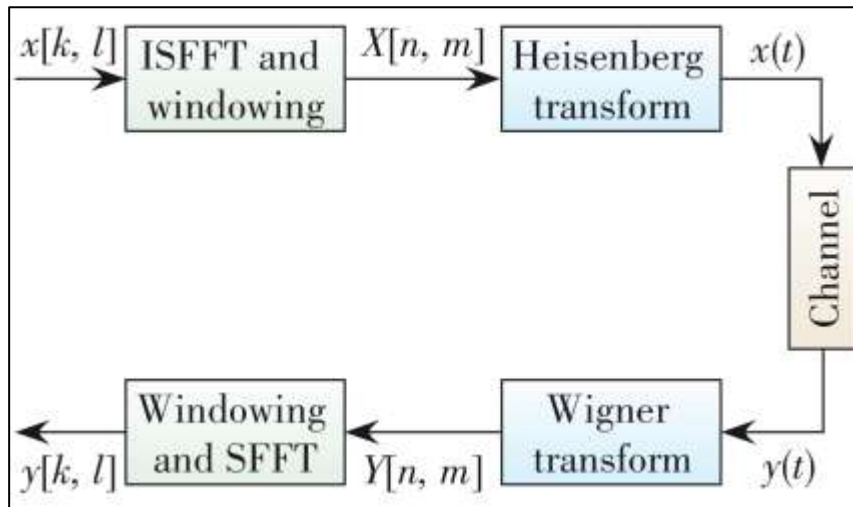
- ✓ OTFS I/O Relationship
- ✓ Pilot Design

OTFS I/O and Pilot Design

- **OTFS signal chain:** DD grid \rightarrow ISFFT \rightarrow Heisenberg \rightarrow Channel \rightarrow Matched Filtering \rightarrow SFFT.
- **Vectorized model:**

$$\mathbf{y} = \Psi \mathbf{h} + \mathbf{w}.$$

- **Pilot design:** pseudo-random pilots + guard intervals ensure incoherence.



Optimization Model and ADMM

- ℓ_1 -regularized convex problem:

$$\min_{\mathbf{h}} \frac{1}{2} \|\mathbf{y} - \Psi \mathbf{h}\|_2^2 + \lambda \|\mathbf{h}\|_1.$$

- ADMM iterations:

- $\mathbf{h}^{(i+1)} := \arg \min_{\mathbf{h}} \frac{1}{2} \|\mathbf{y} - \Psi \mathbf{h}\|_2^2 + \frac{\delta}{2} \|\mathbf{h} - \mathbf{b}^{(i)} + \mathbf{v}^{(i)}\|_2^2$

- $\mathbf{b}^{(i+1)} := \arg \min_{\mathbf{b}} \frac{\lambda}{\delta} \|\mathbf{b}\|_1 + \frac{1}{2} \|\mathbf{h}^{(i+1)} - \mathbf{v}^{(i)} + \mathbf{b}\|_2^2$

- $\mathbf{v}^{(i+1)} := \mathbf{v}^{(i)} + \mathbf{h}^{(i+1)} - \mathbf{b}^{(i+1)}$

- Soft-thresholding $S_\rho(\cdot)$ enforces sparsity while retaining dominant paths.

Simulation Results

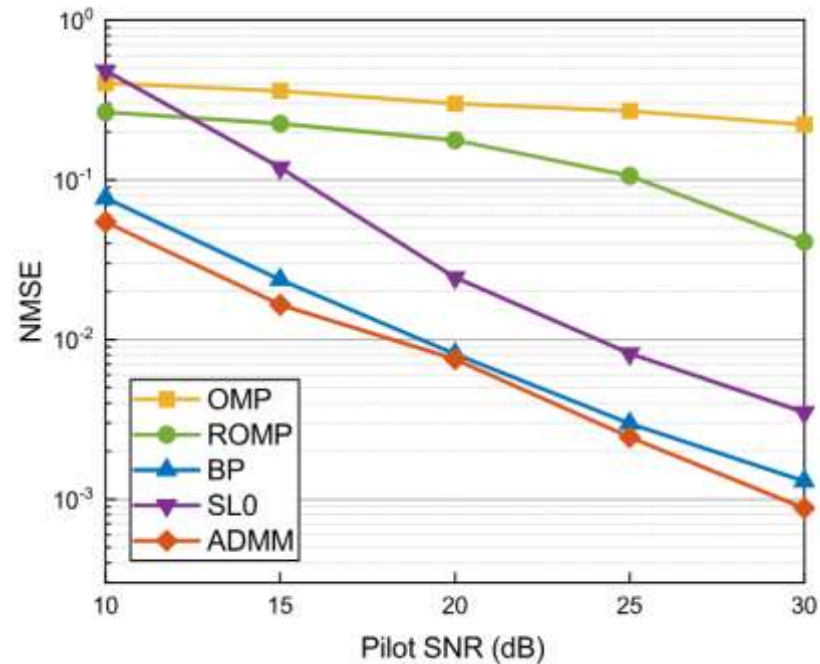
- ✓ **Simulation Results**
- ✓ **Complexity Comparison**

Simulation Results

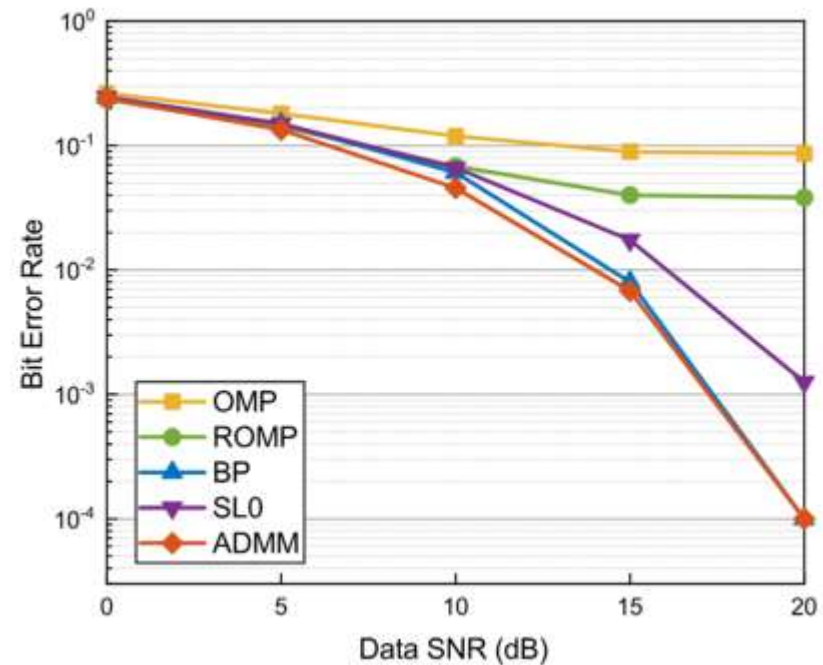
- Setup: 4GHz, 15kHz spacing, 400 km/h (3GPP ETU), 9 paths, QPSK.

- Results:

➤ NMSE vs. SNR



➤ BER vs. SNR



Complexity Comparison

- Dominant cost: matrix inversion $\rightarrow \mathcal{O}((k_\nu l_\tau)^3)$,
feasible as $k_\nu \ll N, l_\tau \ll M$.

- Comparison:

- OMP – Fast, inaccurate
- SL0 – Non-convex, unstable
- BP – Accurate, slow
- ADMM – Convex, efficient, robust

Algorithm	Average Time (s)
OMP	0.332×10^{-2}
ROMP	4.630×10^{-2}
BP	382.2×10^{-2}
SL0	0.349×10^{-2}
ADMM	5.258×10^{-2}

(Monte Carlo experiments on device of Intel (R)
Core (TM) i5-12400F CPU and 32GB of RAM)

Conclusions

- ✓ **Conclusions**
- ✓ **Limitations**

Conclusions

- **Conclusions:**

- ADMM-based OTFS channel estimation → accurate, robust.
- Better NMSE and BER performance without sparsity prior than other CS methods.

- **Limitations:**

- Relatively high complexity → 2D structure
- Fractional Doppler, Practical waveforms not considered.

- **Acknowledgment:** National Research Foundation of Korea (NRF), Grant RS-2025-00517140 (MSIT).

Thanks!