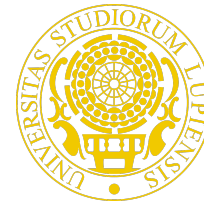


# *“Enabling Joint Localization and Synchronization in mmWave Multiple-Input Single-Output (MISO) Systems via Reconfigurable Intelligent Surfaces”*

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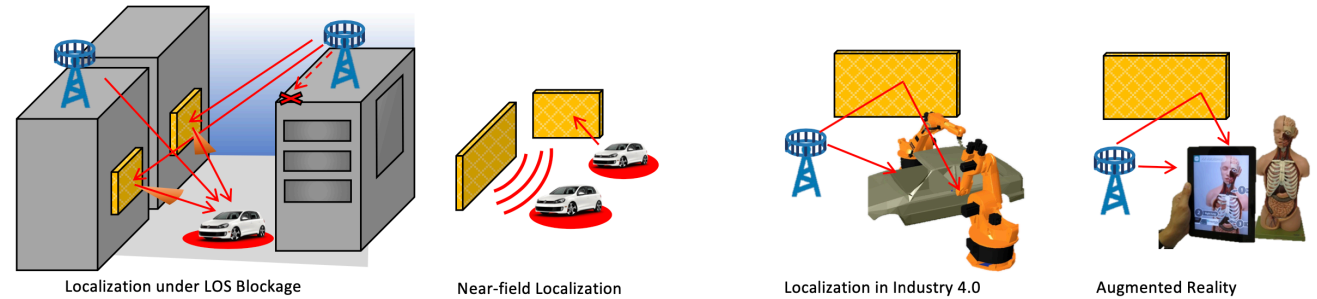
# Localization and Sensing in 5G and Beyond 5G Networks

- mmWave MIMO enable very accurate localization and more favorable multipath propagation



...*beyond 5G*

- **Reconfigurable Intelligent Surfaces:** shaping the radio environment for improved positioning



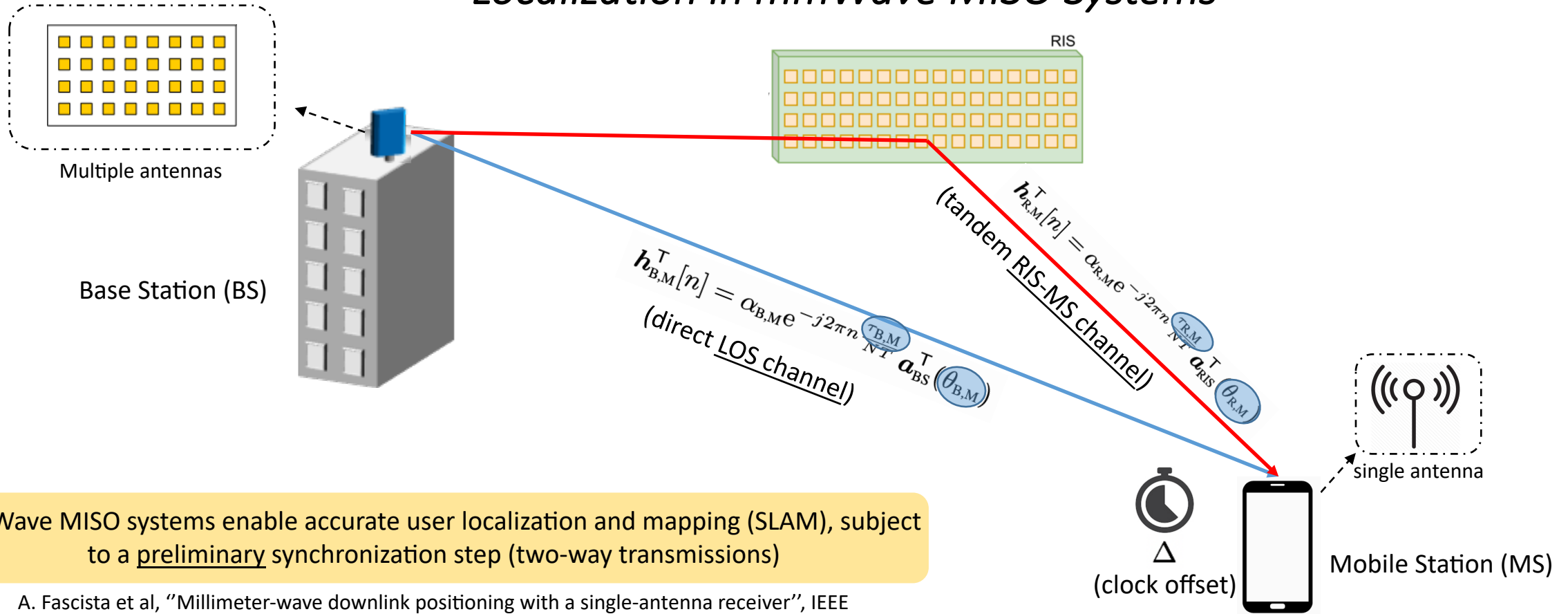
H. Wymeersch et al "Radio Localization and Mapping With Reconfigurable Intelligent Surfaces: Challenges, Opportunities, and Research Directions," IEEE Vehicular Technology Magazine, Dec. 2020

## More challenging case: single-antenna receivers (mmWave MISO)

- a) before full-fledged MIMO will be widespread
- b) limit case of MIMO, in presence of blockages/obstructions
- c) in cases of low-cost IoT single-antenna devices (e.g., WSN nodes, miniaturized sensors, ...)



# Localization in mmWave MISO Systems



mmWave MISO systems enable accurate user localization and mapping (SLAM), subject to a preliminary synchronization step (two-way transmissions)

- A. Fascista et al, "Millimeter-wave downlink positioning with a single-antenna receiver", IEEE Transactions on Wireless Communications, 2019
- A. Fascista et al, 'Downlink Single-Snapshot Localization and Mapping with a Single-Antenna Receiver', IEEE Transactions on Wireless Communications, 2021

With the aid of a RIS, joint localization and synchronization becomes possible using one-way transmission!

Jointly estimate user position and clock offset from signals received from a single BS and a single RIS in unknown MISO environment



# Maximum Likelihood Joint Localization and Synchronization

- Desired parameters:  $\Theta = [p_x \ p_y \ \Delta]^T$  (user position and clock offset)
- Stack all the N samples collected for each transmission g

$$\mathbf{y}^g = \sqrt{P} \mathbf{B}^g(\Theta) \alpha + \nu^g$$

- Joint ML Estimation Problem

$$\hat{\Theta}^{\text{ML}} = \arg \min_{\Theta} \left[ \min_{\alpha} \sum_{g=1}^G \|\mathbf{y}^g - \sqrt{P} \mathbf{B}^g \alpha\|^2 \right]$$

Inner minimization possible in closed-form

$$\hat{\alpha}^{\text{ML}} = \frac{1}{\sqrt{P}} \mathbf{B}^{-1} \sum_{g=1}^G (\mathbf{B}^g)^H \mathbf{y}^g$$



$$\hat{\Theta}^{\text{ML}} = \arg \min_{\Theta} \sum_{g=1}^G \|\mathbf{y}^g - \sqrt{P} \mathbf{B}^g(\Theta) \hat{\alpha}(\Theta)\|^2$$

- **Problem:** highly-non linear and requires exhaustive 3D grid search
- ✓ **Solution:** find a good initial estimate of  $\theta$  and solve joint ML via iterative optimization (e.g., Nelder-Mead)



# Good Initialization via Relaxed ML Estimation

- More convenient rewriting to **decouple** dependencies on AODs and delays

$$\underbrace{\begin{bmatrix} \mathbf{y}^1 \\ \vdots \\ \mathbf{y}^G \end{bmatrix}}_{\mathbf{y} \in \mathbb{C}^{GN \times 1}} = \underbrace{\begin{bmatrix} \Phi_{B,M}^1(\theta_{B,M}(\mathbf{p})) & \Phi_{R,M}^1(\theta_{R,M}(\mathbf{p})) \\ \vdots & \vdots \\ \Phi_{B,M}^G(\theta_{B,M}(\mathbf{p})) & \Phi_{R,M}^G(\theta_{R,M}(\mathbf{p})) \end{bmatrix}}_{\Phi(\theta_{B,M}(\mathbf{p}), \theta_{R,M}(\mathbf{p})) \stackrel{\text{def}}{=} \Phi(\mathbf{p}) \in \mathbb{C}^{GN \times 2N}} \underbrace{\begin{bmatrix} \mathbf{e}_{B,M} \\ \mathbf{e}_R \end{bmatrix}}_{\mathbf{e} \in \mathbb{C}^{2N \times 1}} + \begin{bmatrix} \boldsymbol{\nu}^1 \\ \vdots \\ \boldsymbol{\nu}^G \end{bmatrix}$$

depends on  $\mathbf{p}$  via AODs

with  $\mathbf{e}_{B,M} = \sqrt{P}\alpha_{B,M}$ ,  $\mathbf{e}_R = \sqrt{P}\alpha_R$

$$\begin{bmatrix} 1 \\ e^{-j\kappa_1 \tau_{B,M}} \\ \vdots \\ e^{-j\kappa_{N-1} \tau_{B,M}} \end{bmatrix}$$

$$\begin{bmatrix} 1 \\ e^{-j\kappa_1 \tau_R} \\ \vdots \\ e^{-j\kappa_{N-1} \tau_R} \end{bmatrix}$$

depends on  $\mathbf{p}$  and  $\Delta$  via delays

- Relax dependency of  $\mathbf{e}$  on delays  $\rightarrow$  relaxed ML estimator of position-only

$$\hat{\mathbf{p}}^{\text{RML}} = \arg \min_{\mathbf{p}} \left[ \min_{\mathbf{e}} \|\mathbf{y} - \Phi \mathbf{e}\|^2 \right]$$

$$\hat{\mathbf{e}}^{\text{RML}} = (\Phi^H(\mathbf{p})\Phi(\mathbf{p}))^{-1} \Phi^H(\mathbf{p})\mathbf{y}$$

(Pseudoinverse only requires  $G \geq 2$ )



$$\hat{\mathbf{p}}^{\text{RML}} = \arg \min_{\mathbf{p}} \|\mathbf{P}_{\Phi(\mathbf{p})}^\perp \mathbf{y}\|^2$$

Use  $\hat{\mathbf{p}}^{\text{RML}}$  in  $\hat{\mathbf{e}}^{\text{RML}}(\hat{\mathbf{p}}^{\text{RML}})$  and estimate delays via FFT

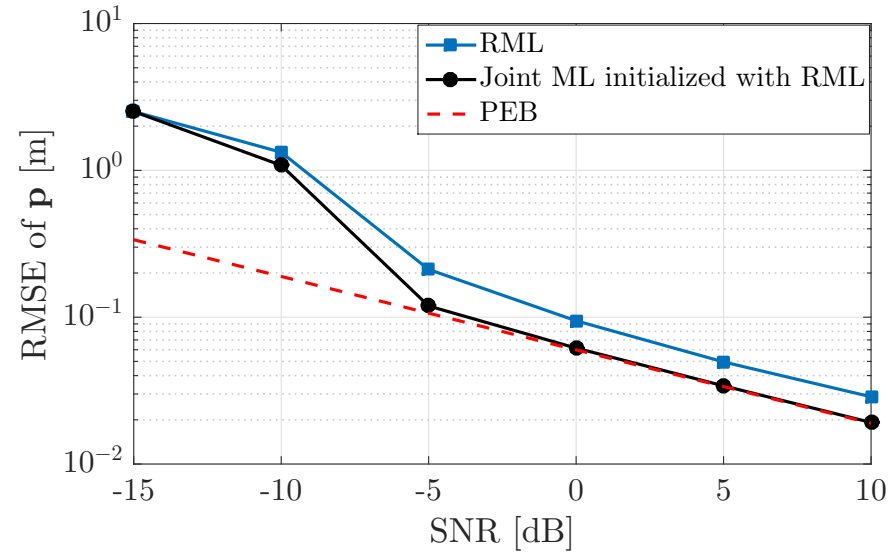
$$\hat{\Delta}^{\text{RML}} = \frac{1}{2} \left[ \hat{\tau}_{B,M}^{\text{RML}} - \|\hat{\mathbf{p}}^{\text{RML}}\|/c + \hat{\tau}_R^{\text{RML}} - (\|\mathbf{r}\| + \|\mathbf{r} - \hat{\mathbf{p}}^{\text{RML}}\|)/c \right]$$



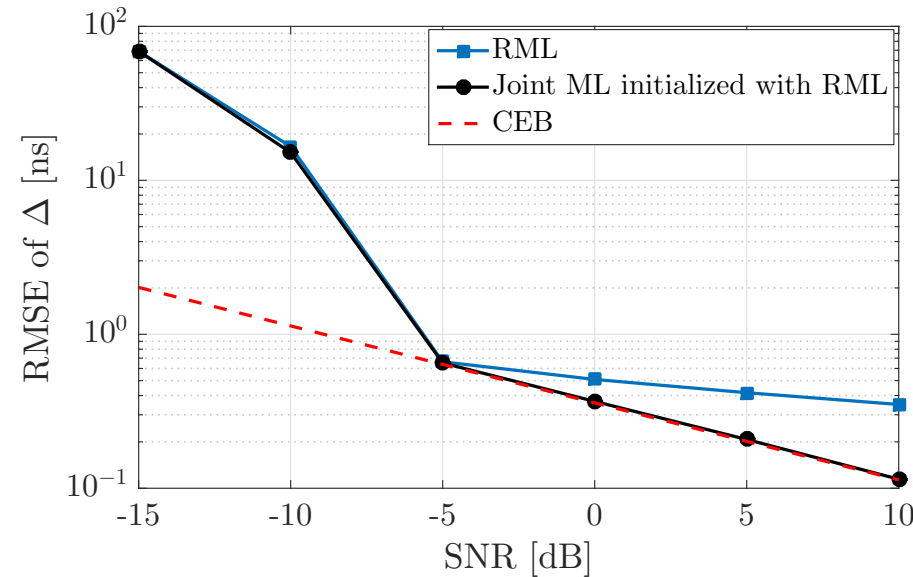
# Results

## Simulation setup:

$f_c$	60 GHz
$B$	40 MHz
$G$	5
$N$	30
$N_{BS}$	20
$N_R$	20
$M$	$N_{BS}/2$
$\mathbf{F}$	1 beam towards $\theta_{B,R}$ and $M - 1$ uniform coverage
RIS phase shifts	0 or $\pi$



- RML achieves good performance already at -5 dB SNR
- ..but does not strictly attain the PEB
- JML exhibits excellent performance



- almost same trend as for  $\mathbf{p}$
- slightly wider gap of RML with CEB due to suboptimality of FFT

...in case of random RIS weights, smaller  $N_R$  can be compensated by increasing  $G$



