# Downlink Channel Covariance Estimation in Realistic FDD Massive MIMO Systems





# **Spatial Covariance Matrix**

- Knowledge of  $\mathbf{R}^d = \mathbb{E}[\mathbf{h}^d(\mathbf{h}^d)^H]$  at the base station:
- Crucial for many DL CSI acquisition and beamforming algorithms in FDD Massive MIMO systems.
- However, difficult to obtain in practical FDD Massive MIMO systems.

# **DL** Covariance Estimation

- Conventional techniques based on DL training and feedback unfeasible in Massive MIMO.
- At the BS,  $\mathbf{R}^u$  is generally easier to estimate.
- In FDD systems,  $\mathbf{R}^d \neq \mathbf{R}^u$ .
- However, a weaker form of channel reciprocity in the angular domain can be assumed.



# FDD UL/DL Covariance **Conversion Problem**

 $\blacktriangleright$  Estimation of  $\mathbf{R}^d$  from  $\mathbf{R}^u$ .

# **Existing Approaches: Issues**

 $oldsymbol{R} = \mathbb{E}[oldsymbol{h}oldsymbol{h}^{I}]$ 

- coordinate system.
- $\boldsymbol{a}_V, \boldsymbol{a}_H : \Omega \to \mathbb{C}^{N \times 1}$  are the BS antenna array responses for the vertical and for the horizontal polarizations.
- $\rho_V, \rho_H : \Omega \to \mathbb{R}^+$  are the frequency invariant angular power spectra for the vertical and for the horizontal polarizations (V-APS, H-APS).
- $\blacksquare$  It can be derived from 3GPP-3D-like channel models, both for narrow-band and wide-band OFDM systems.
- |1| L. |2| L.

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• Most of the available solutions rely on simple channel models and/or require specific array geometries (e.g. ULA). They often fail to address important effects such as propagation in 3Denvironments, polarization, and non-ideal array geometries.

#### A Novel Covariance Model for Dual-polarized Arrays

$${}^{H}] = \int_{\Omega} \rho_{V}(\boldsymbol{\theta}) \boldsymbol{a}_{V}(\boldsymbol{\theta}) \boldsymbol{a}_{V}(\boldsymbol{\theta})^{H} d\boldsymbol{\theta} + \int_{\Omega} \rho_{H}(\boldsymbol{\theta}) \boldsymbol{a}_{H}(\boldsymbol{\theta}) \boldsymbol{a}_{H}(\boldsymbol{\theta})^{H} d\boldsymbol{\theta}$$

•  $\Omega = [-\pi, \pi] \times [0, \pi]$  is a spherical

Core idea: Joint V-APS and H-APS estimation formalized as a convex feasibility problem: very effective solutions based on projection methods on an infinite-dimensional Hilbert space.

- covariance model.
- equipped with the inner product
- We estimate  $(\rho_V, \rho_H) \in \mathcal{H}$  by solving

where

$$V_m := \{ \langle (\rho_V$$

- $r_m^u$  is the *m*th element of vec  $(\left[\Re\{\mathbf{R}^u\} \Im\{\mathbf{R}^u\}\right])$ .
- variety  $P_V(0)$  (Algorithm 1).
- type  $\langle (g_{V,m}^{u}, g_{H,m}^{u}), (g_{V,l}^{u}, g_{H,l}^{u}) \rangle$ .

Miretti, Renato L.G. Cavalcante and Slawomir Stanczak, "FDD massive MIMO channel spatial covariance conversion using projection methods", IEEE ICASSP, 2018 . Miretti, Renato L.G. Cavalcante and Slawomir Stanczak, "Downlink channel spatial covariance estimation in realistic FDD massive MIMO systems", IEEE GlobalSIP, 2018

### Covariance Conversion using Projection Methods

$$\begin{aligned} \boldsymbol{R}^{u} &= \int_{\Omega} \rho_{V}(\boldsymbol{\theta}) \boldsymbol{a}_{V}^{u}(\boldsymbol{\theta}) \boldsymbol{a}_{V}^{u}(\boldsymbol{\theta})^{H} d\boldsymbol{\theta} + \int_{\Omega} \rho_{H}(\boldsymbol{\theta}) \boldsymbol{a}_{H}^{u}(\boldsymbol{\theta}) \boldsymbol{a}_{H}^{u} \\ \boldsymbol{R}^{d} &= \int_{\Omega} \rho_{V}(\boldsymbol{\theta}) \boldsymbol{a}_{V}^{d}(\boldsymbol{\theta}) \boldsymbol{a}_{V}^{d}(\boldsymbol{\theta})^{H} d\boldsymbol{\theta} + \int_{\Omega} \rho_{H}(\boldsymbol{\theta}) \boldsymbol{a}_{H}^{d}(\boldsymbol{\theta}) \boldsymbol{a}_{H}^{d} \end{aligned}$$

• We obtain an estimate  $(\hat{\rho}_V, \hat{\rho}_H)$  of  $(\rho_V, \rho_H)$  based on the knowledge of  $\mathbf{R}^u$ , expression (1), and known properties of  $(\rho_V, \rho_H)$ . **2** We compute an estimate of  $\mathbf{R}^d$  from (2), and by substituting  $(\rho_V, \rho_H)$  with its estimate  $(\hat{\rho}_V, \hat{\rho}_H)$ .

#### **Proposed Covariance Conversion Scheme**

Extension of the ideas in [1] to the considered realistic

• Derived by focusing on the Hilbert space  $\mathcal{H} = L^2[\Omega] \times L^2[\Omega]$ 

 $\langle (f_V, f_H), (g_V, g_H) \rangle := \int_{\Omega} f_V(\boldsymbol{\theta}) g_V(\boldsymbol{\theta}) d^2 \boldsymbol{\theta} + \int_{\Omega} f_H(\boldsymbol{\theta}) g_H(\boldsymbol{\theta}) d^2 \boldsymbol{\theta}.$ find  $(\rho_V, \rho_H)^* \in V := \bigcap_{m=1}^M V_m$ ,

 $\langle r, \rho_H \rangle, (g^u_{V,m}, g^u_{H,m}) \rangle = r^u_m \}.$ 

•  $g^u_{(\cdot),m}$  is the *m*th element of vec  $\left( \left[ \Re\{\boldsymbol{a}^u_{(\cdot)}(\theta)\boldsymbol{a}^u_{(\cdot)}(\theta)^H\} \Im\{\boldsymbol{a}^u_{(\cdot)}(\theta)\boldsymbol{a}^u_{(\cdot)}(\theta)^H\} \right] \right)$ . Among the solutions, we choose the projection onto the linear

• Closed-form solution available in terms of inner products of the

• A more accurate but more complex variant (Algorithm 2) is available, which takes into account the positivity of the APS.

- No specific array geometry is required.



estimation error.



$(\boldsymbol{\theta})^H d\boldsymbol{\theta}$	(1)
$(\boldsymbol{ heta})^H d\boldsymbol{ heta}$	(2)

#### Main Advantages

• Algorithm 1 is a simple matrix/vector multiplication  $\hat{\boldsymbol{r}}^d = \boldsymbol{F} \boldsymbol{r}^u$  over vectorized covariances.

• **F** depends only on the array geometry and it is

computed once for the entire system lifetime.

• It takes into account polarization and 3D propagation.