

# Revised System Model: Massive MIMO-OFDM with Rayleigh Fading

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April 1, 2025

## 1 System Model

We consider a single-cell massive MIMO-OFDM downlink system where the base station (BS) employs  $N_t \gg 1$  transmit antennas, and the user equipment (UE) has a single receive antenna ( $N_r = 1$ ). The system utilizes  $N_c$  orthogonal subcarriers with frequency-division duplexing (FDD).

### 1.1 Channel Model

The time-domain channel impulse response between the  $i$ -th transmit antenna and the UE follows a Rayleigh fading model with  $L$  multipath components:

$$h_i(\tau) = \sum_{l=1}^L \alpha_{i,l} \delta(\tau - \tau_l), \quad \alpha_{i,l} \sim \mathcal{CN}(0, \sigma_l^2), \quad (1)$$

where  $\alpha_{i,l}$  is the complex path gain and  $\tau_l$  is the delay of the  $l$ -th path.

The frequency-domain channel matrix for the  $n$ -th subcarrier is:

$$\mathbf{H}_n = [H_{1,n}, H_{2,n}, \dots, H_{N_t,n}] \in \mathbb{C}^{1 \times N_t}, \quad (2)$$

where each element is derived from the DFT of the CIR:

$$H_{i,n} = \sum_{l=1}^L \alpha_{i,l} e^{-j2\pi n \tau_l / N_c}. \quad (3)$$

### 1.2 Pilot Transmission and LS Estimation

The BS transmits orthogonal pilot sequences  $\mathbf{X}_{p,n} \in \mathbb{C}^{N_t \times T_p}$  ( $T_p \geq N_t$ ) for each subcarrier. The received signal at the UE is:

$$\mathbf{Y}_{p,n} = \mathbf{H}_n \mathbf{X}_{p,n} + \mathbf{Z}_n, \quad \mathbf{Z}_n \sim \mathcal{CN}(0, \sigma^2 \mathbf{I}). \quad (4)$$

The least-squares (LS) channel estimate is:

$$\hat{\mathbf{H}}_n = \mathbf{Y}_{p,n} \mathbf{X}_{p,n}^\dagger, \quad \text{where } \mathbf{X}_{p,n}^\dagger = \mathbf{X}_{p,n}^H (\mathbf{X}_{p,n} \mathbf{X}_{p,n}^H)^{-1}. \quad (5)$$

### 1.3 CSI Feedback Compression

The full CSI matrix  $\mathbf{H}_{\text{freq}} \in \mathbb{C}^{N_c \times N_t}$  is constructed as:

$$\mathbf{H}_{\text{freq}} = \left[ \hat{\mathbf{H}}_1^T, \hat{\mathbf{H}}_2^T, \dots, \hat{\mathbf{H}}_{N_c}^T \right]^T. \quad (6)$$

To exploit sparsity, we transform  $\mathbf{H}_{\text{freq}}$  to the angle-delay domain via 2D-DFT:

$$\tilde{\mathbf{H}} = \mathbf{F}_d \mathbf{H}_{\text{freq}} \mathbf{F}_a^H, \quad (7)$$

where  $\mathbf{F}_d \in \mathbb{C}^{N_c \times N_c}$  (delay DFT) and  $\mathbf{F}_a \in \mathbb{C}^{N_t \times N_t}$  (angular DFT) are unitary matrices. Only the first  $N_i$  delay bins are retained:

$$\tilde{\mathbf{H}}_f = \left[ \tilde{\mathbf{H}} \right]_{1:N_i,:}. \quad (8)$$

### 1.4 Deep Learning-Based Compression

The UE compresses  $\tilde{\mathbf{H}}_f$  into a codeword  $\mathbf{s}$  using an encoder network  $f_{\text{enc}}$ :

$$\mathbf{s} = f_{\text{enc}}(\tilde{\mathbf{H}}_f), \quad \mathbf{s} \in \mathbb{R}^{N_s}. \quad (9)$$

The BS reconstructs the CSI using a decoder  $f_{\text{dec}}$ :

$$\hat{\tilde{\mathbf{H}}}_f = f_{\text{dec}}(\mathbf{s}). \quad (10)$$

The compression ratio is:

$$\gamma = \frac{N_s}{2N_i N_t}. \quad (11)$$

## 2 References

1. D. Tse and P. Viswanath, *Fundamentals of Wireless Communication*. Cambridge University Press, 2005.
2. R. W. Heath Jr. and A. Lozano, *Foundations of MIMO Communication*. Cambridge University Press, 2018.