Fingerprint-based Positioning with Joint Antenna and Area Selection in Cell-Free Systems

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Abstract——TBD...

Index Terms—Fingerprint-based positioning, Antenna Selection, Cell-Free systems, Distributed MIMO, Adaptive Search Area, K-Means Clustering

I. INTRODUCTION

II. SYSTEM MODEL

A Cell Free System(CFS) consisting of L geographically dispersed BS/AP in the coverage area is considered for the positioning scenario. Each of the L AP is equipped with a Uniform Linear Array of N antennas. The ensemble of L APs jointly serve K single antenna UEs. It is assumed that all the L APs co-serve the K UEs by joint coherent transmission and reception. The APs are connected to edge-cloud processors called Central Processing Unit (CPU) through fronthaul links. The CPU(s) coordinate the operation of APs within the CFS. The system model is shown in Fig 1

The uplink (UL) channel from UE k to AP l, denoted by $\mathbf{h}_{kl} \in \mathbb{C}^N$ is modeled using a Coherence Block Fading model, wherein the channel coefficients are assumed remain constant for a duration corresponding to τ_c complex samples. The received combined UL signal $\mathbf{Y}_l \in \mathbb{C}^{N \times K}$ at AP l from all K UEs is given by

$$\mathbf{Y}_{l} = \sum_{k=1}^{K} \sqrt{\rho_{k}} \mathbf{h}_{kl} \boldsymbol{\phi}_{k}^{T} + \mathbf{N}_{l}$$
 (1)

where ρ_k is the transmit power of UE k, $\mathbf{N}_l \in \mathbb{C}^{N \times K}$ is the combined noise at AP l with i.i.d elements distributed as $\mathcal{N}_{\mathcal{C}}(0,\sigma_n^2)$. $\phi_k \in \mathbb{C}^{\tau_p}$, $k = \{1,2,...,K\}$ are the mutually orthogonal pilot vectors assigned to UEs when they gain access to the network i.e.

$$\phi_i^H \phi_j = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases} \tag{2}$$

It is assumed that $(K \leq \tau_p < \tau_c)$ so that there is no interference between UEs due to pilot contamination, for the

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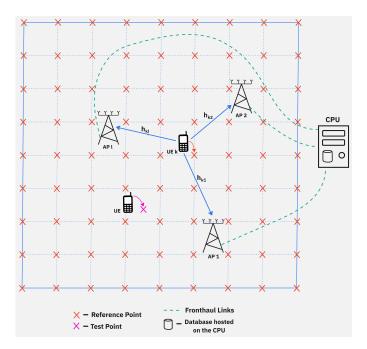


Fig. 1. System Model

purpose of positioning. The signal $\mathbf{y}_{kl} \in \mathbb{C}^N$ corresponding to UE k at AP l obtained post correlating the received signal \mathbf{Y}_l with the conjugate of the pilot ϕ_k is given by [3]

$$\mathbf{y}_{kl} = \mathbf{Y}_l \boldsymbol{\phi}_k^H = \sqrt{\rho_k} \mathbf{h}_{kl} + \mathbf{n}_{kl} \tag{3}$$

where $\mathbf{n}_{kl} \sim \mathcal{N}_{\mathcal{C}}(\mathbf{0}_N, \sigma_n^2 \mathbf{I}_N)$ and \mathbf{h}_{kl} is characterized using an uncorrelated Rayleigh fading model in each coherence block i.e.

$$\mathbf{h}_{kl} \sim \mathcal{N}_{\mathbb{C}}(\mathbf{0}_N, \mathbf{R}_{kl})$$
 (4)

Here $\mathbf{R}_{kl} = \beta_{kl} \mathbf{I}_N \in \mathbb{R}^N$ denotes the diagonal spatial correlation matrix. For β_{kl} , the large-scale fading coefficient 3GPP 38.901 Urban Micro street canyon NLOS Path Loss model [4] is adopted i.e.

$$\beta_{kl}[dB] = 22.4 + 35.3 \log_{10}(d_{kl}) + 21.3 \log_{10}(f_c) - 0.3(h_{UE} - 1.5) + \nu$$
 (5)

Where d_{kl} is the three dimensional distance between UE k and AP l, f_c is the carrier center frequency in GHz, h_{UE} is the vertical distance of the UE from the ground plane and $\nu \sim \mathcal{N}_{\mathcal{C}}(0,\sigma_{SF}^2)$ is the shadowing noise. While the shadowing noise can be mitigated through spatial averaging during the online phase, it persists during the offline phase due to the

unavailability of UE position information, rendering spatial averaging infeasible [5], [6].

To mitigate the variations in RSS due to small scale fading in the UL channel \mathbf{h}_{kl} , modeled by the complex normal distribution in 4, channel hardening is implicitly assumed in the system [1]–[3]. RSS calculation and position estimation is performed within the duration of this hardened channel by averaging the small scale variations across multiple coherence blocks.

The estimated RSS $\theta_{kl} \in \mathbb{R}$ for UE k at AP l, considering channel hardening is given by [2]

$$\theta_{kl} = \mathbb{E}\{\|\mathbf{y}_{kl}\|^2\}
= \mathbb{E}\{\|\sqrt{\rho_k}\mathbf{h}_{kl} + \mathbf{n}_{kl}\|^2\}
= \rho_k \mathbb{E}\{\|\mathbf{h}_{kl}\|^2\} + N\sigma_n^2
= \rho_k \|\beta_{kl}\mathbf{I}_N\|^2 + N\sigma_n^2
= \rho_k N\beta_{kl} + N\sigma_n^2$$
(6)

where the expectation is with respect to the channel realizations.

Equations 5 and 6 reveal that the RSS in decibels from a UE to an AP is directly proportional to the distance between the two entities. This relationship serves as a basis for estimating the unknown location of a UE during the online positioning phase.

III. FINGERPRINT-BASED POSITIONING

- 1) The Offline Phase:
- 2) The Online Phase:
- 3) Joint Antenna and Area Selection (JAAS) Algorithm:
- 4) JAAS Complexity:

IV. SIMULATION RESULTS V. CONCLUSION

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