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Hot-Spot Wireless Access Exploiting Shadowing Diversity of Distributed Antennas

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Abstract—In this paper, we investigate throughput improvement of hot-spot wireless access with distributed antennas. To exploit shadowing diversity of multiple distributed antennas, we propose an antenna selection approach, which takes both desired signal strength and interference strength into consideration. We also develop a power allocation approach to further enhance the throughput of wireless access. We then use wireless stadium as an example to demonstrate the effectiveness of the proposed approaches.

Key Words: Coverage, Signal Processing for Transmission, Modulation & Multiplexing.

I. INTRODUCTION

Hot-spot wireless access is required by more and more applications. One of the critical characteristics of hot-spot wireless access is that there are a lot of mobile users crowded in a small area while the available wireless spectrum is limited.

To improve the throughput of indoor wireless access, distributed antenna techniques have been proposed in [1] - [6] over ten years ago. Recently, it has taken on renewed interesting. In this paper, we will develop an optimal antenna selection approach, which takes both desired signal strength and interference strength into consideration when selecting antennas. We will also propose a pseudo-capacity based power allocation approach.

The rest of the paper is organized as follows. In Section II, we will first introduce wireless propagation model used in our study and then describe wireless access for a stadium. Then in Sections III and IV, we develop an antenna selection approach and a power allocation approach, respectively. Finally, we demonstrate performance improvement through computer simulation.

II. SYSTEM DESCRIPTION

In this section, we will first introduce wireless propagation model, which will be used in our study and then we describe wireless stadium as an example of hot-spot wireless access.

A. Wireless Propagation Model

Radio propagation is often very complicated and different in different environments. In general, it can be characterized by three phenomena: path loss, shadowing, and multipath fading. The fading characteristics of distributed antennas have been investigated in [4]. Here we only briefly describe a simplified model that will be used in our study.

Path loss depends on the distance between the transmit and receive antennas. Denote P_t and \bar{P}_r as the transmitted signal power and the average received signal power, respectively. Pass loss is defined as

$$L = \frac{P_r}{P_t}.$$

It can be modeled as [7]

$$L = K \frac{1}{d^{\alpha}},$$

where K is a constant determined by antenna configuration, d is the distance between the transmit antenna and the receive antenna, and α is called path loss exponent. The pass loss exponent, α , is 2 in free space, is usually between 2 and 4, and may be as large as 6 sometimes.

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Due to path loss and shadowing, the actual received signal power can be expressed as

$$P_r$$
 (dB) = P_r (dB) + ϵ (dB),

where ϵ represents the effect of shadowing and is a random variable. The shadowing, ϵ , in dB is often modeled as normal distribution with variance σ_s^2 . σ_s is usually between 4 dB and 10 dB and its typical value is 8 dB.

Besides path loss and shadowing, radio signal also experiences multipath fading, which causes received signal level changes quickly with time or position of mobile. Consequently, the instantaneous power of the received signal can be expressed as

$$P = P_r |h|^2 = \beta |h|^2 P_t$$

where h is a base-band complex channel gain and $\beta = L\epsilon$. For simplicity, the complex channel gain is frequently modeled as complex Gaussian with zero mean and unit variance. As a result, $|h|^2$ is with exponential distribution. Furthermore, we assume that $|h|^2$ is with unit average.

It should be indicated that both shadowing and multipath fading cause signal power to change with time. However, time-varying due to shadowing is much slower than due to multipath. Therefore, in our study, we assume that shadowing parameter ϵ can be fed back to the transmitter for performance optimization while instantaneous complex channel gain is not available at the transmitter.

B. Wireless Stadium

We use wireless stadium as an example in this paper even through the approaches developed here can be directly used to all indoor environments.

As in Figure 1, the stadium is modeled as a circle. The sports are played in the inner circle with a radius of R = 150 meters. The audience are between the outer and the inner circles and need wireless access so that they can track their favorite players clearly. The radius difference between the two circles is D = 30 meters.

There are 30 access antennas with 15 uniformly distributed in the outer and the other 15 in the inner circles. The audience area is equally divided into 30 cells, which are marked $C_{n,k}$ for n = 1, 2, 3 and $k = 1, \dots, 10$, as in the figure. The available frequency set is equally divided into three different subsets and the 10 cells with the same first index, n, use the same frequency subset. Consequently, the same frequency may be used by as many as 10 cells simultaneously. It will cause *multi-user*



Fig. 1. Antenna distribution for a stadium.

interference (MUI). In the next section, we will develop approaches to reduce MUI.

Without loss of generality, we only consider 10 cells using the same set of frequency, $C_{1,k}$'s for $k = 1, \dots 10$, and those users in the edge of these cells, M_k 's for $k = 1, \dots 10$, as in Figure 1. Denote β_{mk} the power gain that only takes path loss and shadowing into account and corresponds to the *m*-th transmit antenna, F_m , and the mobile user, M_k . Then the received signal at M_k can be expressed as

$$y_k = \sum_{m \in \mathcal{M}} \sqrt{\beta_{mk}} h_{mk} s_{i_m} + n_k, \qquad (1)$$

where \mathcal{M} is the index set of access antennas used to transmit at the same frequency simultaneously, h_{mk} is the complex channel gain due to multipath fading, s_{i_m} is the transmitted signal at access antenna F_m for user i_m , and n_k is additive white Gaussian noise (AWGN).

III. ANTENNA SELECTION

After introducing a fixed antennas selection approach that considers only pass loss, we develop an optimal antenna selection approach that takes pass loss, shadowing, and interference into consideration.

A. Fixed Antenna Selection

We can select the access point/anatnna that is the closest to each mobile user, which we call *fixed*

antenna selection (FAS). It is obvious that FAS is optimal if there is no shadowing. Consequently, mobile M_k will use access point F_{m_k} , where

$$m_k = \begin{cases} 3(k-1) + 2, \text{if } k \text{ is odd,} \\ 3(k-1) + 1, \text{if } k \text{ is even} \end{cases}$$

Therefore, the receive signal at mobile M_k can be expressed as

$$y_k = \sqrt{\beta_{m_k k}} h_{m_k k} s_k + \sum_{i \neq k} \sqrt{\beta_{m_i k}} h_{m_i k} s_i + n_k.$$
(2)

Consequently, the average signal power and interference-plus-noise power over multipath fading, h_{mk} , are

$$P_S = \beta_{m_k k} p_k,$$

and

$$P_I = \sum_{i \neq k} \beta_{m_i k} p_i + \sigma_N^2,$$

respectively, where

$$p_k = E|s_k|^2$$
, and $\sigma_N^2 = E|n_k|^2$.

If all access points transmit signals with the same power, *i.e.*, $p_i = P_t$ for all *i*'s, then *signal-to-interference-plus-noise ratio* will be

$$\operatorname{SINR}_{k} = \frac{\beta_{m_{k}k}P_{t}}{\sum_{i \neq k} \beta_{m_{i}k}P_{t} + \sigma_{N}^{2}},$$
(3)

the average capacity for FAS will be¹

$$C_{\text{FSE}} = E \left\{ \log_2 \left(1 + \frac{\beta_{m_k k} |h_{m_k k}|^2 P_t}{\sum_{i \neq k} \beta_{m_i k} |h_{m_i k}|^2 P_t + \sigma_N^2} \right) \right\}$$
(4)

Note that in the above expression, the average is over all random variables, β_{m_ik} 's, consisting of constant pass loss and random shadowing fading, h_{m_ik} 's, complex Gaussian random variables with zero mean and unit variance.

B. Optimal Antenna Selection

As indicated in Section II.A, radio signal will experience three almost independent phenomena: pass loss, shadowing, and multipath fading. The pass loss is fixed and is determined by the distance between the transmit and the receive antennas. The shadowing is random and independent for different points/antennas. Since shadowing changes with time very slowly, shadowing parameters can be fed back to the transmitters to exploit shadowing diversity of distributed access antennas.

Due to shadowing, FAS described above is not necessarily optimal. With shadowing parameters, we can select access antenna for each mobile to maximize the SINR. From the above discussion, we should find a mapping from k to m_k such that SINR_k in (3) is maximized for all k, which turns out to be a complicated non-linear programming problem.

To reduce the complexity of the above problem, we select access antenna for each mobile to maximize *signal-to-leakage radio* (SLR),

$$SLR_k = \frac{\beta_{m_k k} P_t}{\sum_{i \neq k} \beta_{m_k i} P_t + \sigma_N^2}.$$
 (5)

It should be indicated that optimizing SLR has been used [12], [13] to find optimal steering vector for down-link beamforming of spatial multiplexing systems.

As in Figure 1, the fixed access point, F_3 , is close to mobiles, M_1 and M_2 , and it is sometimes possible, even through seldom happens, that F_3 is selected to transmit (different) signals to the two users at the same frequency simultaneously, that is, $m_1 = m_2 = 3$, which will cause a problem. First, it implies that the fixed access point send different signals to two different mobile users using the same frequency at the same time, which is impossible. Secondly, at least of one of $SINR_1$ and $SINR_2$ is less that 1, which implies that access point, F_3 , is not good for either of M_1 and M_2 and we should exclude it when search for optimal access points for M_1 and M_2 . To reduced the search complexity, for M_1 , we only compare SLR for two fixed access points, F_1 and F_2 , that is, $m_1 = 1$ or 2, depending on which gives a larger SLR. Similarly, we search F_4 and F_5 to find an optimal access point for M_2 , and so on.

Once an optimal access point for each mobile, m_k , is determined, we can find the average capacity using Equation (4).

IV. POWER ALLOCATION

The performance of the system can be further improved by optimally allocating transmission power for each mobile and taking both signal and interference strengthes into consideration.

In this section, we assume the average transmission power for each mobile, p_k , is P_t , that is,

$$\frac{1}{10}\sum_{i=1}^{10} p_k = P_t \text{ with } p_k \ge 0, \ k = 1, \cdots, 10.$$

¹We approximate interference as Gaussian to obtain the capacity formula.

Ideally, we should allocate power to maximize the average capacity of all mobiles, that is,

$$C = \frac{1}{10} \sum_{k=1}^{10} E \left\{ \log_2 \left(1 + \frac{\beta_{m_k k} |h_{m_k k}|^2 p_k}{\sum_{i \neq k} \beta_{m_i k} |h_{m_i k}|^2 p_i + \sigma_N^2} \right) \right\}$$
(6)

However, the implementation of the criterion is very complicated since the expectation should be taken over $h_{m_i,k}$'s. Therefore, we come up with a of suboptimal criterion.

Instead of using the average capacity, we use what we call *pseudo-capacity* (PC) criterion to allocate transmission power, that is, allocate the power to maximize

$$C_{pc} = \frac{1}{10} \sum_{k=1}^{10} \log_2 \left(1 + \frac{\beta_{m_k k} p_k}{\sum_{i \neq k} \beta_{m_i k} p_i + \sigma_N^2} \right).$$
(7)

The difference between (6) and (7) is that the expectation is removed in (7) and $|h_{m_ik}|^2$ is substituted by its average.

From traditional gradient algorithm, we have the following algorithm to find the optimal power allocation to maximize C_{pc} :

Step (i): Initialization:

$$\mathbf{p}^{(0)} = \mathbf{1} P_t.$$

Step (ii): *Iteration*: for $i = 1, 2, \dots$,

$$C_{pc}^{(i-1)} = \frac{1}{10} \sum_{k=1}^{10} \log_2(1 + \frac{\beta_{m_k k} p_k^{(i-1)}}{\sum_{j \neq k} \beta_{m_j k} p_j^{(i-1)} + \sigma_N^2})$$

$$oldsymbol{\Delta} ilde{\mathbf{p}}^{(i)} \;=\; \left(egin{array}{c} rac{\partial C_{pc}^{(i-1)}}{\partial p_1^{(i-1)}} \ dots \ rac{\partial C_{pc}^{(i-1)}}{\partial p_{10}^{(i-1)}} \end{array}
ight)$$

$$\mathbf{\Delta p}^{(i)} = \begin{cases} \mathbf{\Delta \tilde{p}}^{(i)}, & \text{if } \|\mathbf{\Delta \tilde{p}}^{(i)}\| \leq \delta_1, \\ \frac{\mathbf{\Delta \tilde{p}}^{(i)}}{\|\mathbf{\Delta \tilde{p}}^{(i)}\|}, & \text{if } \|\mathbf{\Delta \tilde{p}}^{(i)}\| > \delta_1, \end{cases}$$

and

$$\tilde{\mathbf{p}}^{(i)} = \tilde{\mathbf{p}}^{(i-1)} + \mu \Delta \mathbf{p}^{(i)}.$$

Step (iii): Checking Constraint:

$$\hat{p}_{k}^{(i)} = \begin{cases} \tilde{p}_{k}^{(i)}, & \text{if } \tilde{p}_{k}^{(i)} > \delta_{2} , \\ \delta_{2}, & \text{if } \tilde{p}_{k}^{(i)} \le \delta_{2} , \end{cases}$$



Fig. 2. Performance versus the normalized transmission power when $\alpha = 3.5$ and $\sigma_s = 8$ dB.

for $k = 1, \dots 10$, and

$$\mathbf{p}^{(i)} = \frac{P_t}{\operatorname{mean}(\hat{\mathbf{p}}^{(i)})} \hat{\mathbf{p}}^{(i)}.$$

Step (iv): Repeating Steps (ii) and (iii) if $\parallel \mathbf{p}^{(i)} - \mathbf{p}^{(i-1)} \parallel \leq \delta_3$.

In the above δ_1 , δ_2 , and δ_3 are small positive numbers and μ is called step size..

V. SIMULATION RESULTS

To demonstrate the performance of the proposed approaches, computer simulation is performed. In all of our simulation, the transmission power per Hz is normalized by the power spectral density of noise. When calculating the average capacity, the fast fading due to multipath also takes into consideration.

Figure 2 shows the average capacity versus transmission power for different algorithms when the path loss exponent of channel is $\alpha = 3.5$ and the standard deviation of shadowing is $\sigma_s = 8$ dB. From the figure, optimal antenna selection can significantly improve the system performance compared with fixed antenna selection. Without power allocation, optimal antenna selection can increase the throughput by about 30% when the normalized transmission power, $P_t = 120$ dB. The pseudo-capacity based power allocation can further improve the network throughput, especially when there is no antenna selection.

Figures 3 (a) and (b) show the average capacity versus the path loss exponent and versus shadowing, respectively. From Figure 3 (a), the network



Fig. 3. Performance versus (a) the path loss exponent when $P_t = 120$ dB and $\sigma_s = 8$ dB and (b) the standard deviation of shadowing when $P_t = 120$ dB and $\alpha = 3.5$.

reaches the best performance for the environment with a pass loss exponent of around 4.0. From Figure 3 (b), the optimal antenna selection can improve the network performance and the performance improvement increases with the standard deviation of shadowing.

VI. CONCLUSIONS

In this paper, we have described our research results in antenna selection and power allocation for hot-spot communications. We have developed an optimal antenna selection approach, which takes both desired signal strength and interference strength into consideration when selecting antennas. We have also proposed a power allocation approach that can significantly improve the average capacity of the network. Even through we only use a wireless stadium as an example in the paper. The developed techniques can be directly used to other indoor environments.

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