

Effect of Channel Interference on Indoor Wireless Local Area Network Positioning

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Abstract—Localization systems for indoor areas have recently been suggested that make use of existing wireless local area network (WLAN) infrastructure and location fingerprinting approach. However, most existing research work ignores channel interference between wireless infrastructures and this could affect accurate and precise positioning. A better understanding of the properties of channel interference could assist in improving the positioning accuracy while saving significant amounts of resources in the location-aware infrastructure. This paper investigates to what extent the positioning accuracy is affected by channel interference between access points. Two sets of experiments compare how the positioning accuracy is affected in three different channel assignment schemes: ad-hoc, sequential, and orthogonal data is analyzed to understand what features of channel interference affect positioning accuracy. The results show that choosing an appropriate channel assignment scheme could make localization 10% more accurate and reduce the number of access points that are required by 15%. The experimental analysis also indicates that the channel interference usually obeys a right-skewed distribution and positioning accuracy is heavily dependent on channel interference between access points (APs).

I. INTRODUCTION

Wireless Local Area Networks (WLANs) are often deployed on a large-scale in a wide range of urban environments. Covering a very large urban environment requires that thousands of access points be placed and installed properly, without interference. The basic requirements of an effective WLAN are, first, adequate coverage where users wish to access location-aware (e.g., pervasive computing-enabled) applications and services. Second, the WLAN should allow accurate localization of mobile devices. The deployment of the network should reduce the interference as much as possible so as to achieve these functions in a cost-effective and resource-efficient manner.

Unfortunately, access points (APs) are usually deployed in an empirical way, manually placed and positioned on the basis of measurements of RSS (received signal strength) taken by engineers. Such an unstructured approach to WLAN infrastructure design implies poor resource utilization and strong channel interference. [1] For example, more APs may be used to improve coverage but this may still leave blind spots or places where there are too many access points packed too closely together. This can lead to signal overlap, which is wasteful and causes interference. This issue has been addressed in our previous work. [2]

Location-Fingerprinting-based approaches [1][3][4][5][6][7] locate a device by comparing its coordinates with the

received signal strengths (RSSs) and coordinates of other devices within the Wi-Fi footprint as held in an LF database. More specifically, the LF approach requires the collection of data $\{(\mathbf{Y}_n, C_i), i = 1 \dots N\}$, for N locations in an area, where C_i is the known location of the i 'th measurement and $\mathbf{Y}_i = (Y_{i1}, \dots, Y_{iN})$ is the received signal strength (RSS) vector when the transmitter is at C_i . The vector \mathbf{Y}_i is the "fingerprint" of the location C_i . When a new fingerprint \mathbf{Y} is derived from a transmitter at an unknown location A , we can locate A by searching for the fingerprint \mathbf{Y}_i that is closest to \mathbf{Y} in say d distance and estimate the location with the corresponding C_i . The drawbacks of the LF approach are (1) LF requires an initially survey with a very large training dataset and (2) LF is very sensitive to signal fluctuation due to the changes of infrastructure of buildings and channel interference among APs leading to inaccurate positioning.

The IEEE 802.11 standard establishes several requirements for radio frequency transmission, including the canalization schemes and the spectrum radiation of the signal.[8] In IEEE 802.11 b/g WLAN, there are 14 channels. In North America, the 2.4GHz frequency ISM band is divided into 11 channels. Each channel is spread over 22 MHz due to the Direct Sequence Spread Spectrum (DSSS) technique employed by IEEE 802.11b/g. These channels have only five MHz of center frequency separation. Channel interference occurs because frequency spectrum is shared with each adjacent channel. Recent research has focused on reducing channel interference by either creating a new channel assignment scheme [9][10][11] or enhancing existing MAC protocols [12][13] The goal is to improve the data transmission through wireless networks. However, none of existing work on channel interference has focused on the accuracy of the location estimation algorithm. Some researchers have even maintained that interference could increase positioning accuracy.

This paper investigates the influence of channel interference in a location fingerprinting approach. The study of channel interference is essential for accurate indoor positioning system. This paper also describes localization experiments and simulations on the IEEE 802.11 test-bed and investigates the channel assignment of APs, the distribution of received signal strength (RSS) values, the variation of coverage, and distances between APs. The analysis of these features provide insights into how to assign channels, how to space APs so as to reduce interference, and how many access points are

required to uniquely identify a location at a given accuracy and precision. The results would be of interest and assistance to engineers designing WLAN channel assignments specifically for positioning.

The rest of this paper is organized as follows: Section II describes the reason and algorithms of overlapping channel interference. Section III describes the positioning methodology. Section VI defines the location uncertainty. Section IV defines the characteristics of channel interference metrics. Section V presents experiment and result on channel interference metrics. Section VI presents the experiment and result on the effect of channel interference to positioning accuracy. Finally, Section VII offers our conclusion and future work.

II. OVERLAPPING CHANNEL INTERFERENCE

The bandwidth of wireless network is limited because of the property of wireless networks and stations have to share the limited bandwidth. [14] IEEE 802.11b/g has 14 overlapping frequency channels. Channel 1, 6 and 11 are non-overlapping channels.

As shown in 1, IEEE 802.11 b/g spreads through 2,401 MHz to 2,483 MHz. Each channel spreads over 22 MHz. Two adjacent channels are separated only by 5 MHz such that most of existing channels are overlapped.

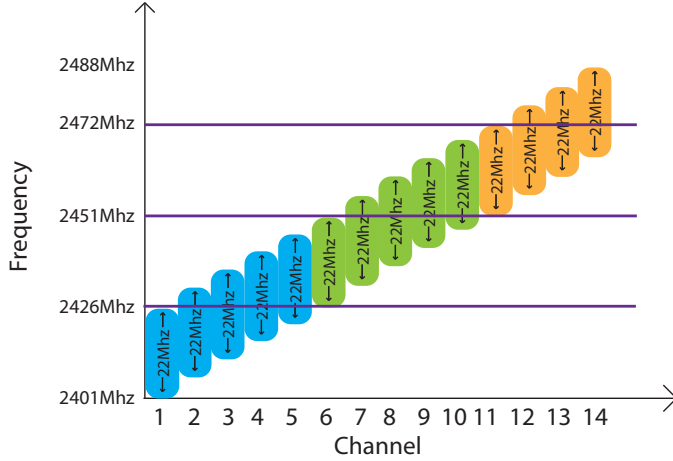


Fig. 1. IEEE 802.11b/g Frequency Spectrum to Channel Number

A. Interference-level Function

The interference-level function γ is defined as follows:

$$\gamma(\Delta c) = \max(0, 1 - k\Delta c) \quad (1)$$

where Δc is the absolute channel difference and k is the non-overlapping ratio of two channels. γ and Δc are in Db unit. When Δc increases, γ decreases. For example, if $\Delta c = 0$, then $\gamma(\Delta c) = 1$ and if $\Delta c \geq 5$, then $\gamma(\Delta c) = 0$. In other words, for channel 1 and 6, $\Delta c = 5$, $k = 0$, then $\gamma(\Delta c) = 0$, suggesting no interference. In real case, if APs are installed far enough with others, γ should be at least equal to the above threshold.

B. Signal Propagation Loss Algorithm

Signal propagation loss algorithm [6][15][16] calculates the received signal strength (RSS) with path loss as follows:

$$R = r - 10\alpha \log_{10}(d) - wallLoss \quad (2)$$

where r is initial RSS, d is a distance from APs to a location, α is the path loss exponent (clutter density factor) and $wallLoss$ is the sum of the losses introduced by each wall on the line segment drawn at Euclidean distance d .

Initially, r is the initial RSS at the reference distance of d_0 is 1 meter (this is 41.5 dBm for LOS propagation and for 37.3 dBm NLOS propagation for some report measurement). The path loss exponent α at a carrier frequency of 2.4 GHz is reported to be 2 for LOS propagation and 3.3 for NLOS propagation [8]. Under other circumstances, α can be between 1 and 6.

C. Signal-to-Interference-plus-Noise-Ratio

Signal-to-Interference-plus-Noise-Ratio (SINR) is a very common indicator to measure interference. SINR is defined as follows:

$$SINR = \frac{R_b}{\gamma(\Delta c) \sum R + n} \quad (3)$$

where R_b is the highest RSS after path loss calculation. R is the remaining set of RSS after path loss calculation. n is the noise signal strength. R_b , R , n are in dBm unit. Usually, n should have the value of -100dbm. Again SINR should be at least equal to above calculated threshold which depends on the distance among APs, the transmission rate, the modulation scheme and the required bit-error rate.

III. POSITIONING METHODOLOGIES

Two positioning methodologies are typically applied in WLANs, propagation based approaches and location fingerprinting (LF). Our previous works [17][18] make use of LF to track a WLAN-enabled device. In our later simulations we will use only LF approach. For the completeness in the following we briefly describe both.

A. Propagation-based Approach

Propagation-based approaches measure the received signal strength (RSS), angle of arrival (AOA), or time difference of arrival (TDOA) of received signals and apply mathematical models a set of triangulation algorithms to determine the location of the device.

Using the propagation-based theorem in (2), we can denote the distance between the access points and object location as d_1 , d_2 , and d_3 , where d_0 is the initial RSS at the reference distance. To estimate the location of the object, we use the tri-lateration with angle of arrival (AOA) method. We can calculate the distance and find the angle θ_1 , θ_2 and θ_3 between the object location and APs. Then we are able to calculate the possible location matrix of the object as follows:

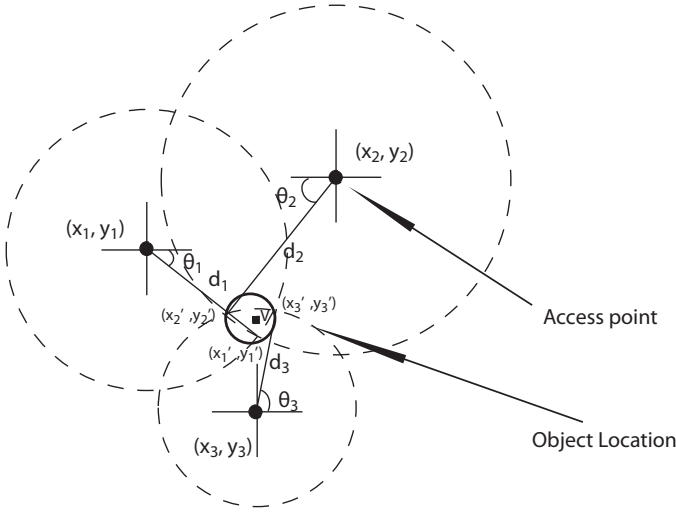


Fig. 2. Triangular Algorithm with Angle of Arrival

$$\begin{pmatrix} x'_1 & y'_1 \\ x'_2 & y'_2 \\ x'_3 & y'_3 \end{pmatrix} = \begin{pmatrix} x_1 + d_1 \cos \theta_1 & y_1 + d_1 \sin \theta_1 \\ x_2 + d_2 \cos \theta_2 & y_2 + d_2 \sin \theta_2 \\ x_3 + d_3 \cos \theta_3 & y_3 + d_3 \sin \theta_3 \end{pmatrix} \quad (4)$$

B. Location Fingerprinting Approach

There are few types of location Fingerprinting approaches, the K-Nearest Neighbor (K-NN), the probabilistic, support vector model (SVM) and neural networks (NN). Among these methods, we choose the probabilistic approach.

Probabilistic LF applies Baye's theorem to calculate the most probable location out of the pre-recorded LF database, $F = [f_1, f_2, f_3 \dots f_N]$. We can estimate d by

$$\arg \max_d [P(d/F)] = \arg \max_d \left[\frac{P(F/d) P(d)}{P(F)} \right] \quad (5)$$

Since $P(F)$ is constant for all d , the algorithm can be rewritten as:

$$\arg \max_d [P(d/F)] = \arg \max_d [P(F/d) P(d)] \quad (6)$$

As $P(d)$ can be factored out from the maximization process, the probabilistic positioning algorithm is as

$$P(F/d) = \prod_{i=1}^N P(f_i/d) \quad (7)$$

We make use of probabilistic location fingerprinting to estimate the position in our later part of experiment and simulation.

IV. CHARACTERIZATION OF CHANNEL INTERFERENCE METRIC

This section investigates whether interference is normally or log normally distributed. The signal difference between the preset (maximum) RSS value of AP and sample RSS of an AP

could be seen as interference strength value (in dB). Two APs are placed and interfere with each other. We assume that there is a zero (or very short) distance between receiver and APs. In other words, the signal should not be reduced by propagation-loss. The preset RSS value is denoted by ρ and the interference strength value by $F = \{f_1, f_2, \dots, f_n\}$. The interference could then be by:

$$Y = \sqrt{\sum_{i=1}^n (\rho - r_i)^2} = \sqrt{\sum_{i=1}^n f_i^2} \quad (8)$$

where n represents the number of collected sample. In (8), the difference between a preset RSS value and measured RSS, r_i considered to be a signal interference strength. In fact, the random variable of interference f_i should have a zero mean if and only if the wireless RSS obeys a normal distribution. In other words, the random variable of interference f_i has a non-zero mean when the wireless RSS does not obey a normal distribution.

It is assumed that the random variable $X = Y^2$ where the random variable X is the square of the difference between the sample RSS and the preset RSS. Assuming that the RSS is normally distributed, the random variable $X = Y^2$ has a central chi-squared distribution with n degrees of freedom, i.e. $E\{r_i\} = \rho$ or the mean of sample RSS is equal to a preset RSS value. Thus, the difference-squared f_i obeys a zero mean Gaussian distribution. A probability density function (PDF) of X will be the chi-square distribution:

$$P_{X_n^2}(x) = \frac{1}{2^{\frac{n}{2}} \sigma^n \Gamma(\frac{n}{2})} x^{\frac{n}{2}-1} e^{-\frac{x}{2\sigma^2}} \quad (9)$$

where the variance of each Gaussian component in X is σ^2 , and Γ denotes Gamma function which has closed-form values at the half-integers.

However, in some real-world scenarios, the distribution of the RSS is not usually Gaussian and it is often left-skewed. The standard deviation for this real distribution varies according to the signal level. This has been verified in [19]. Actually, the distribution of the RSS is a non-central chi-squared distribution. In this case, the random variable of interference f_i will have a non-zero mean equal to $\mu = \rho - E\{r_i\}$.

Here, λ is defined as a non-centrality parameter of the non-central chi-squared distribution. Parameter λ could be defined as $\lambda = \sum_{i=1}^n \mu_i^2$. A larger value of λ indicates that some regions are experiencing higher signal interference. The PDF of non-central chi-square distribution is seen to be a Poisson-weight mixture of central chi-squared distribution. It could be defined by:

$$P_X(x; n, \lambda) = \sum_{i=0}^{\infty} \frac{e^{-\lambda/2\sigma^2} (\lambda/2\sigma^2)^i}{i!} P_{P_{X_n^2}(n+2i)}(x) \quad (10)$$

Alternatively, the PDF can be written as

TABLE I
EXPERIMENT SETTINGS (SET A)

Item	Description
Number of APs	2 APs
Sampling time	2 hours
Samples of Signal Strength	1,000
Wi-Fi coverage from each APs	80 meters
Range of signal strength	-70dBm to -30dBm

$$P_X(x; n, \lambda) = \frac{1}{2\sigma^2} e^{-\frac{x+\lambda}{2\sigma^2}} \left(\frac{x}{\lambda}\right)^{\frac{n-2}{4}} I_{\frac{n-2}{2}}\left(\frac{\sqrt{\lambda x}}{\sigma^2}\right) \quad (11)$$

where $I_k(x)$ is the k th-order modified Bessel function of the first kind given by

$$I_k(x) = \left(\frac{x}{2}\right)^k \sum_{i=0}^{\infty} \frac{x^{2i}}{4^i i! \Gamma(k+i+1)} \quad (12)$$

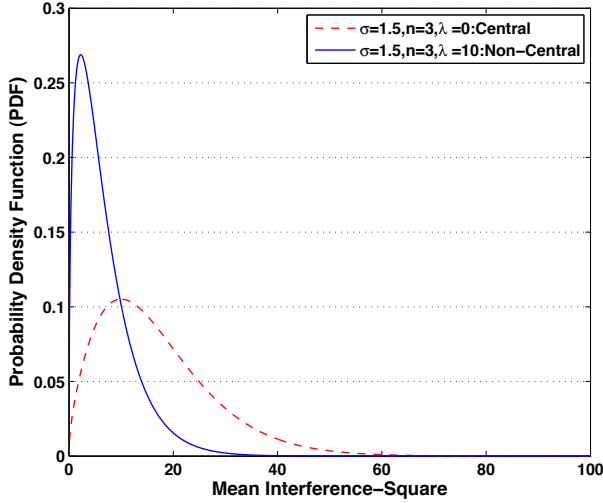


Fig. 3. Theoretical comparison of the PDFs of central and non-central chi-squared distribution of interference under $\sigma = 1.5$, $n = 3$

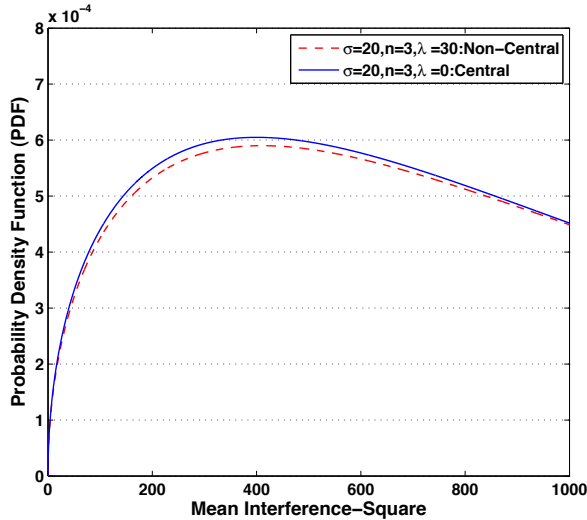


Fig. 4. Theoretical comparison of the PDFs of central and non-central chi-squared distribution of interference under $\sigma = 20$, $n = 3$

In Figure 3 we have the theoretical distribution formed by (9) and (11). The dotted purple line represents the central chi-squared distribution used by (9). The blue line represents a

non-central chi-squared distribution used by (11). As can be seen, the use of a central chi-square distribution produces a more evenly distributed interference-square. In other words, the strength of the interference fluctuates widely. When $\sigma = 1.5$, $n = 3$, the interference-square of the non-central distribution would be mostly around 10. When a non-central distribution is used, the distribution is skewed more to the right and interference strength is more likely to be a smaller value. When $\sigma = 1.5$, $n = 3$ and $\lambda = 10$, the interference-square of non-central distribution would be mostly at 4.

Figure 4 shows the PDF versus mean interference-square under $\sigma = 20$, $n = 3$. The non-central and central chi-square becomes very similar when the standard deviation of the received signal strength has a larger value. This indicates that the larger standard deviation of received signal strength causes the mean interference-square of central and non-central chi-distribution to move closer together.

In conclusion, depending on whether RSS is a central or non-central normal distribution, the interference distribution could be defined by either (9) or (11). Having said this, it should also be noted that experiments with RSS distributions could vary, with some experiment results showing that RSS obeys a normal distribution and some showing otherwise. To our knowledge, however, the distribution of interference has not yet been studied. The later experimental section will show that interference usually follows the distribution described in (11).

V. EXPERIMENT & RESULT ON CHANNEL INTERFERENCE METRIC

The following section describes an experiment on channel interference metrics and discusses the experiment results. The purpose of the experiment is to determine whether interference distribution obeys a normal, mean chi-square, or non-mean chi-square distribution.

The experiment places two APs within a short distance of an RSS receiver. The assumption is that all signal fluctuations are caused by interference between two APs. Two APs were set to the same channel and emitted a WLAN signal at the maximum strength -30dBm. A receiver recorded 1,000 samples of signal strength from two APs over two hours. The sample result was used to form a distribution and the theoretical distribution was compared using (9) and (11). Table I summarizes the experimental settings of Set A. Two parameters, σ and λ are input to adjust the shape of the distribution curve. σ is

the standard distribution of interference. λ is a non-centrality parameter.

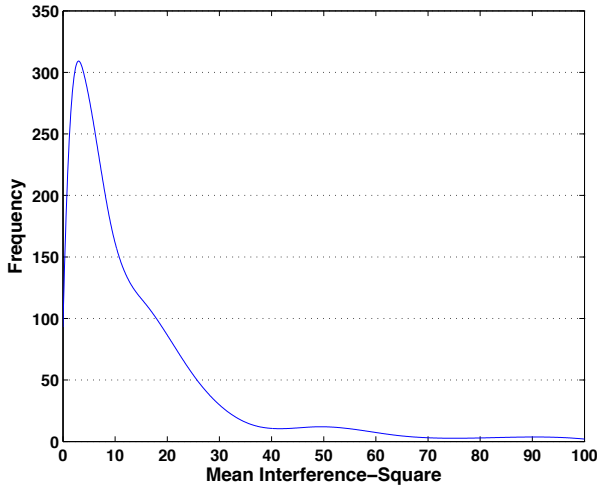


Fig. 5. Result of Frequency to Mean Interference-Square experiment

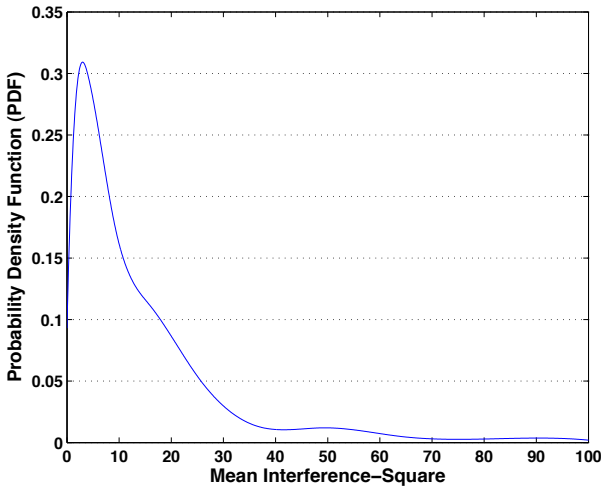


Fig. 6. Result of PDF to Mean Interference-Square experiment

Section VI discussed the theoretical distribution of interference. Here, it is compared with the actual experimental result. Figure 5 shows the relationship of the mean interference-square to frequency. Along the Y-axis, frequency represents the number of occurrences of a particular value of the interference-square. The interference-square is mostly at 7 and is a right-skewed distribution.

Figure 6 shows the relationship of the mean interference-square to the probability density function (pdf). As can be seen in Figure 6, it is again a right-skewed distribution. A smaller value for the mean interference-square means a larger value of the pdf. When the mean interference-square is 4, the pdf has the largest value of 0.31.

In following paragraphs, we could now further discuss on the impact of parameters of the interference based on the visual results presented so far. The visual results so far suggest that interference is a random variable if the received signal strengths from APs are also all random variables. Intuitively this makes sense. The results in these experiments indicate that the interference mostly occurs in a right-skewed distribution, which implies a smaller value for the mean interference-square. It also shows that a non-central chi-distribution (11) could represent the distribution of interference. It would be ideal to have less interference and a smaller fluctuation. The above work helps to understand the features of interference and it could be used to decide a better positioning system.

The following sections will continue to look at some factors that affect interference: channel assignment, number of APs, the SINR value, and distribution of location uncertainty.

VI. THE EFFECT OF CHANNEL INTERFERENCE ON POSITIONING ACCURACY: EXPERIMENT AND RESULTS

This section describes experiments on the effect of channel interference on positioning accuracy. The experiments were conducted in a 1000 m^2 testing area. Accuracy was measured using only the probabilistic LF approach, which is defined by (7). The radius of coverage of each AP is 80m. The signal strength ranges between -85dBm and -30dBm. The positioning resolution is set to three meters. Table II summarizes the experiment settings of Set B.

TABLE II
EXPERIMENT SETTINGS (SET B)

Item	Description
Total testing area	1000 square meter
RSS variation	Between -90 dBm and -35 dBm
Number of sample points	4880 sample points
Wi-Fi channel	1, 6, and 11
Facing direction of each sampling	North, South, East, and West

The following discusses the results and compares the positioning accuracy under three typical channel allocation schemes. Subsection A looks at the impact of channel interference on positioning accuracy. Subsection B investigates the positioning accuracy by varying the number of access points under different channel allocations. Subsections C and D look at how channel interference is affected by varying the number of APs and SINR values.

A. Effect of Interference on the positioning accuracy

Figure 7 shows the relationship of channel interference to positioning accuracy. In order to see how channel interference affects positioning accuracy specifically, the number of APs is set to 13. The channel interference is varied from 0 to 25 dBm and the accuracy is in a scale from 0 to 1 (1 represents 100% accuracy). The three different channel allocations do not have major difference of positioning accuracy when the interference value is still small. When the interference strength increases above 15dBm, positioning accuracy deteriorates seriously. It

is thus clear that when channel interference increases, the positioning accuracy decreases. However, this interference value is difficult to control because it depends on the environment. One way of improving this is to take more iteration. As can be seen in Figure 7, the positioning performance of orthogonal channel allocation is the most accurate. This result indicates that orthogonal channel allocation is 10% more accurate when the system is burdened with high channel interference.

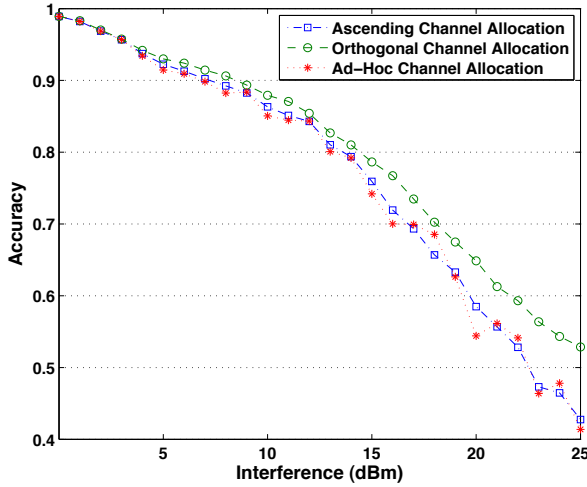


Fig. 7. Relationship of interference to accuracy under ascending, orthogonal and ad-hoc channel allocation

B. Effect of number of APs on the positioning accuracy

This section considers the impact of the number of APs. Figure 8 shows the relationship of the number of access points to accuracy using each of the three allocation schemes. The resolution is 2m. A higher number of APs improves the precision dramatically up to the point that nine APs are used. If more than nine APs are used, the accuracy does not increase significantly due to the interference between them.

The channel interference between APs increases when the number of APs increases. Figure 8 shows that orthogonal channel allocation with only nine APs achieve 90% accuracy. Perhaps the most important point to note is that orthogonal channel allocation could require 15% fewer APs than either ascending channel allocation or ad-hoc channel allocation. Again, the smaller the interference is, the more accurate the positioning is. This is because orthogonal channel allocation provides less interference than the other two channel allocations. The next subsection further investigates this issue.

C. Effect of number of APs on the interference

Figure 9 shows the relationship of the number of APs to interference. The result suggests that more APs cause in more interference. Orthogonal channel allocation is associated with less interference in any case, an average of 10.9 dBm, whereas ascending channel allocation and ad-hoc channel allocation respectively average of 13.1 dBm and 14.24 dBm of interference.

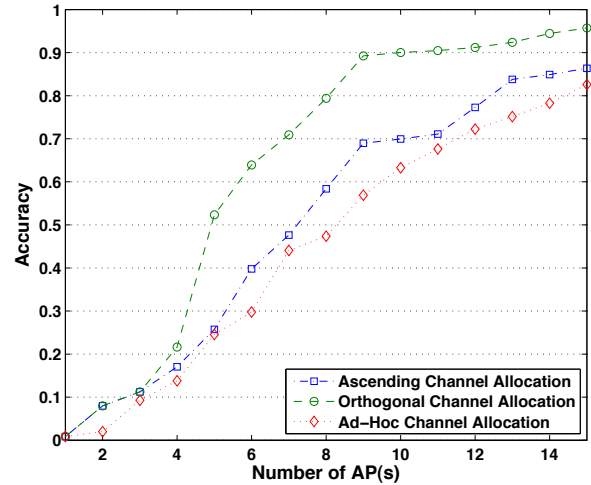


Fig. 8. Relationship of number of access points to accuracy under ascending, orthogonal and ad-hoc channel allocation

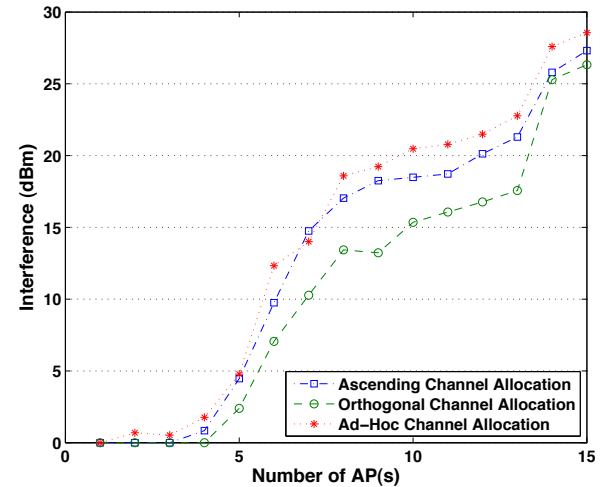


Fig. 9. Relationship of number of APs to interference under ascending, orthogonal and ad-hoc channel allocation

D. Effect of Signal-to-Interference-plus-Noise-Ratio on the positioning accuracy

In order to see how SINR affects the positioning accuracy, the number of APs is set to 13. The SINR is varied from 0 to 1 and the accuracy was in a scale from 0 to 1 (1 represents 100% accuracy). Figure 10 shows the relationship of SINR to positioning accuracy. The result indicates that the higher the value of SINR, the more accurate the positioning. Orthogonal channel allocation has the best positioning performance and was 8% more accurate than the other two channel allocations.

VII. CONCLUSION

We present a comprehensive analysis of the channel interference to positioning accuracy based on location fingerprinting. The result indicates that channel interference would worsen

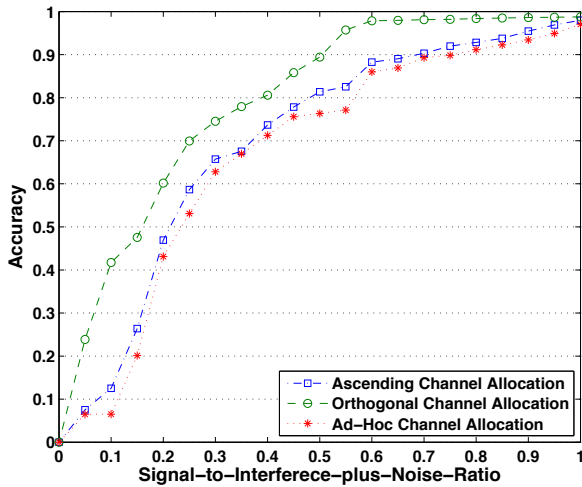


Fig. 10. Relationship of Signal-to-Interference-plus-Noise-Ratio to accuracy under ascending, orthogonal and ad-hoc channel allocation

the positioning accuracy. Higher channel interference implies less accurate positioning. We clear from falsehood that interference could help to achieve more accurate positioning. From our experiment, channel interference among APs using same frequency channel have a significant (worsening) impact on the positioning system. Our experimental analysis also verifies that the channel interference usually obeys a right-skewed distribution. This verification helps to model the environment of channel interference virtually.

The effect of channel interference to positioning accuracy depends on number of APs, SINR and channel assignment scheme. A higher number of APs gives more accurate positioning. Moreover, as the number of APs increases, the channel interferences among APs increase and decrease the rate of increasing positioning accuracy. We also point out that a higher value of SINR is, more accurate positioning could be achieved. In this paper, we emphasize that channel assignment is critical issue to positioning system. Choosing orthogonal channel allocation could save 15% of APs and meanwhile achieve 10% more accurate positioning in average. The results in this paper provide more insight on the stable and robust indoor positioning system based on LF. Our studies address that the positioning accuracy is seriously dependent on channel interference. Thus, reducing channel interference is essential to improve the positioning accuracy and meanwhile could save significant amount of resources of localization infrastructure.

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