

Indoor Localization Wireless System Using RSS of IEEE 802.11b.

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Abstract—With the development of positioning in indoor wireless environments, RSS-based indoor positioning algorithm has been widely applied. In this deliverable we provide the details of building an indoor positioning system using WLAN Received Signal Strength (RSS) fingerprints. The standard deviation between measured and predicted path loss is 5.8 dB for the entire data set, and can be as small as 4 dB for specific areas within a building. Path loss contour plots for measured data are presented. In addition, contour plots for the path loss prediction error indicate that the prediction models presented in this paper are accurate to within 6 dB for a majority of locations in a building. In our case, users that carry a terminal (laptops) are able to self-locate and positioning is performed entirely on the device using the observed RSS fingerprint. Currently, one positioning method, i.e. K-Nearest Neighbours (KNN) and other one filtering method, i.e. Kalman filter, has been integrated into our positioning platform. In the future we plan to implement and integrate more positioning and filtering modules, while at the same time try to improve the performance with respect to the positioning error by tweaking and fine-tuning both modules. The rest of this report is structured as follows. The details of the four experimentation areas in the premises of the Engineering building.

Index Terms—transmission loss location; fingerprint based location;

I. INTRODUCTION

The business sector of localization based service (LBS) is extending. To locate the physical area LBS is the major premise. As GPS is the standard for outside confinement, however it doesn't function admirably in indoor environment because of the hindering of signs by dividers and roof. Remote based indoor confinement can accomplish ease and high precision in the meantime.

In this paper, an efficient examination of the remote based indoor nearby Localization is directed. Firstly, we think about the remote innovations that have been utilized for limitation as a part of late writings. The remote innovations are isolated by the separation of scope. They change in recurrence band and prominence which decide their interesting qualities when utilized for indoor restriction.

After that, we clarify the scientific systems are utilized as a part of remote based restriction. Closeness based strategy can just give inexact area in light of connection or interface data. Triangulation can be utilized to decide edge or separation in arrangement particle recover from the got signals from three or more reference point stations to acquire client location. Fingerprint expects the sign property in every point is distinctive, the area can be found by contrasting and pre-constructed radio-maps. Toward the end of the paper, we have condensed four patterns in the looks into in remote based indoor limitation. The utilization of various remote advances and participation with different advances can cover the imperfection in one innovation. Joining different numerical techniques can lessen the blunder and build the exactness. The appearance of cell telephones additionally gives a perfect gadget as client gadget for indoor restriction. To obtain high precise limitation in indoor environment, numerous systems have been produced. The vision based restriction includes camera and PC vision innovations which expand the expense. Accelerometer based confinement will gather the mistake

made by every limitation forecast. With the advancement of situating in indoor remote situations, RSS-based indoor situating calculation has been broadly connected. In this deliverable we give the subtle elements of building an indoor situating framework utilizing WLAN Received Signal Strength (RSS) fingerprints.

The standard deviation amongst measured and anticipated way misfortune is 5.8 dB for the whole information set, and can be as little as 4 dB for particular ranges inside a building. There has been number of area strategies utilizing global positioning system (GPS) or cell systems proposed in the writing. GPS is surely understood idea called triangulation system which can work precisely outside yet not appropriate for indoor because of poor radio scope. Expansion of cell phones like Laptop, personal digital assistant (PDA) has prompted quick development of wireless local area network (WLAN). WLANs are increasingly ordinarily utilized as a part of workplaces, condo, healing facilities and other indoor situations. Pervasive figuring speaks to the idea of —computing all over the place anytime and makes processing innovation points of interest straightforward to the clients. Circumstance mindfulness is alluring component of frameworks in pervasive figuring environment [1]. With the ascent of 3G innovation, indoor area gets to be prime essential, for example, area, time, commotion level, and accessible assets to perceive the present circumstance. Aside from this the framework can utilize the progressions of circumstance to adjust its conduct, for example, calculation and correspondence, without client's intercession. So area will be a standout amongst the most essential and much of the time utilized connections as a part of 3G and 4G innovation. In cutting edge innovation, indoor area will get

to be prime critical for crisis salvage, visit guides, in doctor's facilities, open spots, commercial ventures, shopping centers, for seeking and following the articles and work force. Being an area mindful, a framework can simply do right thing at correct time. According to the circumstance, the framework ought to do or act as needs be. A PDA rings or vibrate contingent on circumstance whether it is on table or in pocket. The area of versatile client can be evaluated utilizing radio sign transmitted or got by the terminal [2,3,4]. Among numerous indoor area techniques, WLAN based indoor area innovation has turned into a decent research and creating course for its wide scope on the planet and without extra offices.

Past work can be sorted into two wide ways: 1) deterministic, 2) probabilistic. In deterministic e. g. RADAR utilizes closest neighbor and triangulation method and best flag match is found and arrived at the midpoint of. This has precision of around 3 meter with 50 % likelihood. In probabilistic system [10][11]. Here signal quality guide is built over various areas and utilizations probabilistic methodology. Another related work is LANDMARC taking into account RFID innovation. LANDMARC utilizes idea of reference labels to keep away from impact of vacillations of R. F. signal quality. This strategy first registers separation between sign quality got vector from following labels and those from reference labels separately. It then uses k-closest reference labels directions to figure the surmised direction of following labels. Here precision of LANDMARC can be ensured just when reference labels are thickly conveyed. Creator reported that one reference tag is required 1 meter square, which is not doable in numerous applications. Additionally FRID's are costly. In this paper we display our way to deal with indoor area framework utilizing existing IEEE802.11 radio recurrence remote system, which requires least extra equipment. Existing RSS based indoor area calculations chiefly exploit RSSI recorded by clients to accomplish particular area. This methodology does not require any synchronization amongst transmitter and beneficiary.

Two types of techniques are used: one is based on mathematical modeling of wireless channel and second is fingerprinting technique. In mathematical modeling, received signal strength (RSS), angle of arrival (AOA), time of arrival (TOA), time difference of arrival (TDOA) parameters are measured. Then for particular environment, mathematical model is prepared to determine user's location [7-8]. In fingerprinting technique, it is fact that [2][6] characteristics of propagation signal are different at each location of interest. Thus each location has unique fingerprint. From measurement data base of fingerprint (RSS) of different location is built. This observed

fingerprint is compared with on line data using KNN algorithm. This technique gives better location accuracy than mathematical modeling multipath that is difficult to model [2]. Most of the WLAN positioning systems use RSS as input to estimate location because it is most relevant data and easy to obtain from WLAN infrastructure. There are two ways to obtain RSS data. 1) Collect RSS at access point i. e. signal transmitted from mobile host and collected at access point. 2) Collect RSS at mobile host i. e. access point will transmit and mobile host will receive. We employed second method to collect RSS data as a input to location estimator. One drawback of WLAN based location which uses RSS based technique is the extensive calibration phase to build signal fingerprint. Recalibration is also needed if there is major change in propagation environment.

Overview of Wireless Based Indoor Localization

The architecture of wireless based indoor localization system usually requires two parts, the beacon stations that emit the wireless signal and the user devices that receives the signal or versa vice. The computation of localization can be resident in either parts. GPS can also be included as a wireless based localization technology. It uses wireless signals to communicate between satellites and the GPS devices. The 24 satellites are 24 beacon stations. The GPS devices calculate their locations based on received signals from satellites.

However, indoor environment is quite different from outdoor environment. The propagation of wireless wave can be influenced by reflection, scattering, and diffraction. The signal strength can be affected by multi path fading or shadow fading. In indoor environment, the walls, furnitures or walking people will change the propagation of the wireless wave and introduce variance to the wireless signal received by the user. For indoor localization, there are several criteria for evaluating a localization system.

1. Accuracy: The accuracy is indicating the reliability of the system. How likely will it give a right position?
2. Precision: Precision is similar to accuracy which indicates the correctness of the localization. However accuracy means the mean of distance error, whereas precision can be seemed as the derivation of the distance error. The precision of different systems or methods can varies from 5 m to 20 cm.
3. Coverage: Coverage means the area covered for accurate localization. Since different wireless technology has different distance, hence short range wireless technology might need more devices to cover the same area.
4. Update interval: Update interval indicates how frequently the location information updates and reflect the possible power consumption. A simple solution for reducing the power consumption of the localization system is to increase the update interval.
5. Computational cost: The computational cost is an important criterion for evaluation a localization method. More computational cost means more power consumption and more cost on user device.

6. Infrastructure: Whether it needs to build extra infrastructure for the deployment of the system or not is import to the budget cost. The reuse and maintenance of the infrastructure is also significant for the deployment. The infrastructure cost also includes the user device cost.

7. Offline Computing: Some of the indoor localization methods requires of fine computing or site survey which need labor-intensive work and more deployment time. Such requirement increases the cost for deployment as well as maintenance cost. The offline computing needs to recalibrate every other interval to keep the accuracy. Localization time: The time needed for localization for wireless based indoor localization varies for different methods. For methods that support localization for immobile object, the localization time can be very fast.

I. MODELS/ALGORITHMS A. PATH LOSS MODEL

Due to the blocking objects in the environment, received signal strength vibrates up and down. Through multiple measurements in a certain environment, the path loss model is concluded, and the relationship between spread of distance and path loss can be shown in equation (1)

$$PL(dB) = PL(d_0) + 10n \log \left(\frac{d}{d_0} \right) + X_\sigma$$

Where n is the path loss parameter, which denotes pass loss rate as increasing distance which depends on the types of surrounding environment is the shelter factor, a random variable with standard deviation X_σ . d_0 Denotes the distance between reference points, and denotes path loss for 1st meter d .

B. FINGERPRINT LOCATION ALGORITHMS

Location fingerprinting based positioning systems usually work in two phases (see Fig. 1): calibration phase (also called offline phase) and working phase (also called online phase or run-time phase). In the calibration phase, a mobile device is used to measure RSS values (in dBm) from several APs at the chosen calibration points in the area of interest. Each of the n measurements becomes a part of the radio map and is a tuple $(q_i, r_i) i = 1, 2, n$

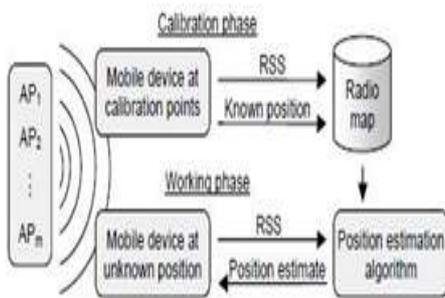


Figure 1. The phases of location fingerprinting

where $q_i = x_i, y_i$ are the geographical coordinates of the i th location and $(r_i) i = r_{i1}, r_{i2}, \dots, r_{im}$ $r =$ are the m RSS values from m APs at that location. Usually, an average of several samples recorded per location is stored. In the working

phase, a mobile device measures the RSS values in an unknown location and applies a location estimation algorithm to estimate its current location using the previously created radio map. As indoor environments have unique signal propagation characteristics, it can be assumed that each location can be associated with a unique combination of RSS values.

C. K NEAREST NEIGHBORHOOD (KNN)

$\dots, m,$

m_n

n, m

$$n \quad | \quad | \quad \} \quad (2)$$

Where, s denotes measured RSS values \dots denote vector in database. \dots is called Manhattan distance if \dots and Euclidean distance if \dots the accuracy does not necessarily higher as q increases. \dots is j th sample value in i th base station, j is a measured value in i th base station,

\dots , m is number of base stations, n is number sample data. The distance between \dots and \dots is defined as

$$\sqrt{\sum (\dots)}, \quad (3)$$

Electing K samples since the smallest value and calculate average coordinates as outputs in equation (4):

$$(\dots) = \frac{1}{K} \sum (\dots) \quad (4)$$

Where \dots is coordinated corresponding to i th sample. Another optimized algorithm based on KNN is weighted KNN algorithm. After selecting K vectors \dots , it calculates output coordinates by adding a weighted coefficient for each vector in database.

The expression is in equation (5):

$$\sum \{ \dots \} \quad (5)$$

Where, i is the distance between measured RSS values and i th vector in database. And \dots is a minimal positive constant which prevents division of zero.

D. AUGMENTING THE RADIO MAP

Collecting large numbers of fingerprints in the calibration phase is labour-intensive, which makes a large-scale deployment of accurate indoor positioning non-trivial.



Figure 2. Client based positioning and tracking. Therefore, a variety of techniques have been proposed in order to generate synthetic calibration points with predicted RSS values for adding to the radio map, allowing to collect only a limited number of field measurements. Many proposed techniques (Hossain et al., 2007; Pechac and Klepal, 2001; Widyawan et al., 2007) predict the RSS values using a radio signal propagation model requiring knowing exact locations of all used APs or even complete plans of the whole deployment area with precise locations of all walls. And even with this information available, the derived propagation models may be inadequate for the environment and therefore may not bring the desired positioning accuracy. A number of other proposed techniques do not rely on any propagation model, instead the RSS values are predicted via local interpolation of the original calibration points, approximating the behavior of the radio signal (Ferris et al., 2006; Li et al., 2006). This study adopts the latter approach and proposes generating synthetic calibration points using LWR with locality of interpolation optimized individually for each AP. The technique does not require knowing locations of neither APs nor walls. The augmentation of the radio map is done using only the original calibration points. This allows performing augmentation also when the available information on the environment is incomplete, e.g when locations of some APs are unknown, and without additional software for environment layout analysis. LWR (Cleveland and Devlin, 1988; Atkeson et al., 1997) is designed to address situations in which the models of global behavior do not perform well or cannot be effectively applied without undue effort. The LWR interpolation is carried out by point-wise fitting of low-degree polynomials to localized subsets of the data. The advantage of this method is that it is not required to specify a global function of the data. The assumption of the LWR is that near the query point (i.e. the to-be-added synthetic calibration point) the approximated value changes smoothly and can be approximated using a low-degree polynomial. The coefficients of the polynomial are calculated using the

weighted least-squares method giving the largest weights to the nearest (according to the Euclidean distance) calibration points and the smallest weights to the farthest calibration points.

II. FINGERPRINT LOCATION ALGORITHM

Multi-path propagation of signal is special and depends on environment. For each position, multi-path structure of the channel is unique. Radio waves sent by terminals generate multi-path signals with specific modes relevant to surroundings

through reflection and refraction, which are called fingerprints. There are two phases for the implementation of fingerprint location: training / off-line phase and location / on-line phase. In off-line phase, record signal characteristic parameters for reference points to establish a location fingerprint radio map. In on-line phase, take advantage of parameters measured by received signals to match the data in the database and output the actual coordinates of users by matching algorithms.

III. EXPERIMENTAL TEST BED

The measurement campaign necessary to build the fingerprints' database has been carried out on second floor in Electronics Telecommunication Engineering Department; of Siddhant College of Engineering is five storey building. Figure 1 shows the layout of the department. It has a width of about 19.5 m and a length of 34 m. Four access points 802.11b (D-Link wireless router) have been placed at various locations as shown in figure. The used mobile host consisted of a Pentium-based laptop computer running Microsoft Windows 7 installed within SSIDer software. This software monitors signal of access points in the vicinity of mobile host.

The transmitted power from the access point is 100 mW at 2.4 GHz. The wireless device in the laptop record signal strength at that location. The range of access point is 100 meter in free space. The measurements are taken at different points in laboratories, class rooms, staff rooms, HOD room, and corridors of the test bed. The set of RSSs have been collected at 80 different mobile user's locations. All measurement locations are shown on the test bed. Each location coordinates are measured with respect to bottom left corner. The grid spacing between two mobile user locations was set to 2.5 meter to 2.9 meter. At each location RSS

CONCLUSION

In RSS based indoor location system, the accuracy is hard to improve further because it is difficult to establish suitable propagation models with strong expansion corresponding to complex environment. However, RSS based fingerprint location technology is adaptable to dynamic indoor environment. The sample values are shown in table 2. Values of current applied fingerprint indoor

location algorithms are proposed in the paper, and there are also many problems to be deal with. As two important branches of machine learning algorithm in indoor location system, ANN and SVM are prospective in resolving complex problems for indoor location. ANN has made great progress while SVM initially shows its potency in indoor location.

was recorded for 30 seconds. The record speed was one record per second. Then from four access points (A, B, C, D) received signal has been recorded. This forms training data which has been stored on test bed.

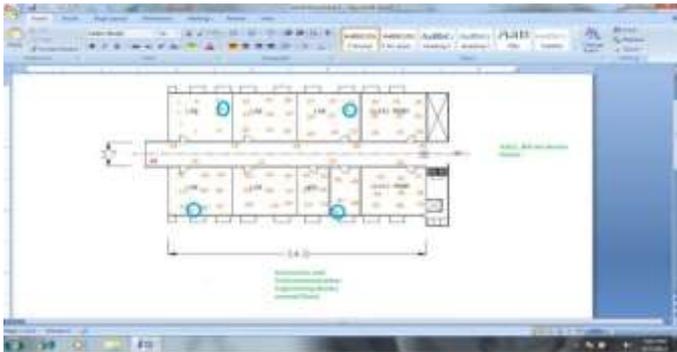


figure 3 The test bed



figure 4 inSSIDer software

	Location in		A - dBm	B - dBm	C - dBm	D - dBm
	X	Y				
1	1.36	16.3	76	77	60	42
2	1.36	14	75	89	53	49

3	1.36	12	72	80	59	52
4	4.76	12	74	83	71	48
5	4.76	14	75	86	77	35

Query node	No.of neighbour	Position of query node	Averages of k coordinates			Error in estimation in mtrs
			x	y		
1	3,4, 9	1.36	16	3.6	13	3.9
2	1,3, 4	1.36	14	2.2	13	1.789
3	1,2, 4	1.36	12	2.2	13	1.789
4	13,4, ,11	4.76	12	5.2	13	1.655
5	6,8, 7	4.76	14	6	15	1.716

Therefore it will have far-reaching significance if indoor accuracy can be improved further by taking advantage of severe theory and non-linear optimized samples of errors are shown in table 3.

V. ADVANTAGES

- Doesn't suffer from multi-path interference.
- High data carrying capacity.
- It need only low power.
- Low energy density.
- Minimum complexity.
- Low cost.

VI. ULTRA-WIDEBAND

- Allow police to detect the movements of a hostage taker through a wall
- Help cars avoid collisions by sensing the location and speed of oncoming vehicles. This can greatly enhance accident avoidance.
- Spawn wireless home networks, linking cable set-top boxes or computers. UWB goes a step beyond Bluetooth and other current home wireless systems by transmitting video and other high-bandwidth

content.

- It also can be used to wirelessly download video from a camcorder to a TV.
- Track the precise location of retail products in stores or keep track of military equipment.
- Provide low-cost security systems that could distinguish between, say a pet and an intruder.
- The table shows the results of the examinations of Cassoli.
- However, users of Global Positioning Systems (GPS) say that by traversing many frequencies,
- UWB might interfere with GPS systems, such as those used by airplanes to navigate over oceans.
- Satellite-based GPS signals are very sensitive. But FCC officials say UWB emits about as much energy as a laptop PC, and interference is unlikely.

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