A RADIO FREQUENCY IDENTIFICATION SYSTEM FOR ACCURATE INDOOR LOCALIZATION

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ABSTRACT

In this paper we present a novel Radio Frequency Identification (RFID) system for accurate indoor localization. The system is composed of a standard Ultra High Frequency (UHF), ISO-18006C compliant RFID reader, a large set of standard passive RFID tags whose locations are known, and a newly developed tag-like RFID component that is attached to the items that need to be localized. The new semi-passive component, referred to as sensatag (sense-a-tag), has a dual functionality wherein it can sense the communication between the reader and standard tags which are in its proximity, and also communicate with the reader like standard tags using backscatter modulation. Based on the information conveyed by the sensatags to the reader, localization algorithms based on binary sensor principles can be developed. We present results from real measurements that show the accuracy of the proposed system.

Index Terms— indoor localization, RFID, passive tags

1. INTRODUCTION

Radio Frequency IDentifications (RFID) is a well-known technology for real-time identification of various assets and users. One of the main goals of RFID technology is to enable ubiquitous asset visibility. Accurately determining the location of an asset is of great importance in achieving this goal. Accurate localization using RFID can enable several applications such as location of tagged items in warehouses, and location of assets and personnel in hospitals and offices [5]. State-of-the-art localization methods can be broadly classified into three categories [2]: i) distance-based methods, ii) scene-analysis, and iii) proximity-based methods. Distance-based methods rely on range measurements that can be based on Received Signal Strength (RSS), Time Of Arrival (TOA), or Time Difference of Arrival (TDOA). Using such measurements at (at least three) different reference points and upon converting them to estimated distances, one can employ simple trilateration to achieve localization. Scene-analysis methods consist of two phases. First, environmental information (fingerprints) is acquired. Then, the target location is estimated by matching the measurements with the stored fingerprints, i.e., the estimated position is the average of the k closest matches.

These methods are affected by dynamic changes in the environment. One direction of investigation for resolving this problem is to work with proximity-based methods which use binary information, i.e., information about a target being within the ranges of the reference tags or not. The location estimate is found either by associating the location of the target with that of the closest reference tag, or as the centroid obtained from the locations of all the reference tags that detected the target. For full details about RFID localization methods and quantitative comparisons, we refer the reader to [2, 5, 7].

In this paper, we introduce a novel type of semi-passive Ultra High Frequency (UHF) RFID tag that has the capability to detect and decode backscatter signals from RFID tags in its proximity and to communicate this information to a standard RFID reader [3]. We refer to this tag as sensatag (from sense-a-tag). We show that when a sensatag is attached to an object that needs to be localized and the object is in an indoor environment that is populated with passive tags with known locations, one can estimate the location of the object with high accuracy.

The paper is organized as follows. In Section 2, we introduce the sensatags and in Section 3, we present the RFID system used for localization and the processing of data collected from this system. The experimental results of our work that demonstrate the performance of the system are shown in Section 4. We conclude the paper with some final remarks in Section 5.

2. A NOVEL TYPE OF RFID COMPONENT

Passive and semi-passive UHF RFID tags do not have on-board radios. They communicate with the reader using the principle of backscatter modulation wherein, the reflection cross section (RCS) of the tag antenna is varied in accordance with the data to be conveyed to the reader [6]. This modulates the signal reflected from the tag antenna to the reader. This tag backscatter is a weak signal that is further affected by multipath reflections and other ambient interferences in cluttered indoor environments like warehouses, retail stores, libraries, and offices [1]. This results in a low signal to noise ratio (SNR) for the tag response received by the reader. Hence conventional location techniques based on the measurement of some characteristic of the tag’s response like RSS, TOA, or TDOA become highly inaccurate and unreliable for localization with passive and semi-passive RFID systems.

Our approach to localization is based on the addition of a new component to a standard reader-tag RFID system, called sensatag
This semi-passive, tag-like component has the following capabilities: i) to detect and decode backscatter signals from RFID tags in its proximity and ii) to communicate with the reader using backscatter modulation.

On top of these basic capabilities, we have incorporated into the sensatag, a novel locator protocol, which is fully compatible with the EPC Global Class 1 Gen 2 standard (ISO-18006C). This protocol enables the sensatag to communicate with a standard reader and convey binary information about the presence or absence of a responding tag in its proximity.

A block diagram of the sensatag hardware is shown in Figure 1. The sensatag communicates passively without an on board radio. An on board battery is used for powering up the sensatag circuitry. Thus, in its current form, the sensatag is a semi-passive device. We will now briefly describe the various functional blocks that make up the sensatag.

2.1. RF Front End

The RF front end of the device consists of a passive envelope detector that is built using a Schottky Diode with corresponding matching circuit. When a passive RFID tag in the vicinity of the sensatag backscatters, the sensatag receives a signal that is a superposition of the tag backscatter and the continuous wave (CW) signal that the reader is transmitting during this time. The sensitivity of the sensatag to tags in its vicinity depends upon its ability to detect small changes in resultant power in this superimposed signal. This corresponds to the $\Delta$ RCS of the tag, i.e. the difference in tag antenna RCS when the tag backscatters a 1 vs when it backscatters a 0. This means that the detector circuit needs to be optimized not for the maximum value of output voltage for a stated input power, but for maximum slope of the input power ($P_{in}$) vs output voltage ($V_{out}$) characteristic around the typical power levels of operation. This optimization was done by appropriately tuning the matching circuit and the time constant of load on the baseband side of the diode detector circuit. The signal at the output of the diode detector is shown in Figure 2.

2.2. Analog Section

The sensatag analog section has the ability to process both the reader signal as well as the tag backscatter in order to produce a digital signal that can be processed by the digital section.

The analog processing of the reader signal is exactly the same as in a standard passive tag. It consists of a buffer followed by a hysteresis comparator that generates the digital output. The processing of the tag backscatter is a bit more complex since the backscatter is a weak signal that has a significant DC offset due to the presence of the CW signal from the reader. The circuit consists of a band-pass filter (or a high-pass filter) for removing the DC offset, followed by a comparator that is configured as a data slicer. The filter parameters and the threshold generation circuit for the comparator are adaptive.

2.3. Digital Section Implementation:

The digital section runs the sensatag protocol and as such is the brain of the device. In the current version, the digital section is implemented on an FPGA platform. This platform is chosen to allow for rapid prototyping and verification of the digital section, particularly keeping in mind that ultimately, the sensatag will be implemented as an ASIC. The present embodiment uses a Xilinx Spartan 3AN FPGA. This device has an internal configuration memory which results in significant space saving on the digital section of the board. The current embodiment of the sensatag, used in the system described herein, is shown in Figure 3.

2.4. Locator Protocol

As mentioned earlier, the sensatag implements a novel locator protocol which enables it to convey binary association information about tags in its vicinity to a standard reader. In order to implement this functionality, the locator protocol specifies two states of operation for the sensatag. In the first state or the listen state, the sensatag listens for backscattering tags in its vicinity. In the second state or the
respond state, the sensatag itself functions as an RFID tag and conveys the information of the tags detected when it was in the listen state as part of its EPC ID payload. The transition between the two states is done based on different types of queries \((Q_1 \cdot \text{tag query and } Q_s \cdot \text{sensatag query})\) received from the reader using the Select functionality provided by the Gen 2 standard. In the query round \(Q_s\), the sensatag acts as a sensor detecting, decoding and storing information about the responding tags within its vicinity. In the subsequent \(Q_s\) query round, the sensatag conveys the binary tag association, along with its own unique identifier information to the reader using backscatter modulation. The localization algorithm running on the reader side aggregates the binary association information from successive query rounds and determines the location of the sensatag with respect to the pre-deployed tags in the environment. The localization algorithms are described in detail in the next section.

3. SYSTEM DESCRIPTION AND DATA PROCESSING

In the system described in this paper, passive RFID tags are deployed at pre-defined locations within the environment where localization is to be performed. A sensatag is attached to the target of interest. The reader is programmed to send out alternating queries for the tags and sensatags using the Select functionality. The sensatag attached to the target operates using the locator protocol described above and conveys binary information about presence or absence of responding tags to the reader.

Let us assume that we have \(M\) reference (passive) tags with known two-dimensional positions, \(x_i\) \((i = 1, 2, \cdots, M)\) and one sensatag with unknown position \(l\). A reference tag can be detected by a sensatag with probability \(p_i\). This probability depends on various factors, but primarily on the distance between the reference tag and the sensatag, orientation, and the power of the reader. This probability is easily estimated by counting the number of detections of a tag by a sensatag in a fixed number of reader queries.

Our main goal is to develop an algorithm that can perform well in environments with dynamical changes, and therefore we decided to use three simple localization methods that should work well in such circumstances. They are based on i) association, ii) centroids, and iii) weighted centroids.

With association we simply associate the sensatag with the nearest passive tag. The proximity is measured by comparing the \(p_i\)'s of each reference tag. The main drawback of association is when more passive tags are detected by the sensatag, the \(p_i\)s may not correctly reflect the distance from the sensatag, which will imply that the sensatag will be associated with a wrong passive tag. As a result, the position error will be larger.

One simple way of building a more robust method is to implement averaging of the positions of all the passive tags that have been detected by the sensatag. In that case, the position of the sensatag is computed by: 
\[
\hat{l} = \frac{1}{N} \sum x_i
\]
where the summation is over the locations of the tags that have been detected and \(N\) is the total number of detected tags by the sensatags. Therefore, the estimated position will be the centroid of the positions of the detected passive tags. This approach does not take into account the number of detections.

A natural extension of the centroid method, is the weighted centroid, where the estimated position is the weighted average of the positions of the detected tags. Since it is expected that the closer tags will be detected more times than more distant ones, the weights are proportional to probabilities of detection of the tags. So, the estimated position is found as 
\[
\hat{l} = \sum p_i x_i
\]
A similar idea is already used for target tracking in binary sensor networks [4].

4. EXPERIMENTAL RESULTS

Here we provide details of the experiments for studying our system and the performance of the methods for localization. We deployed 12 passive tags in 6 reference points, where at each point we deployed two passive tags.\(^1\) The overall area was 1.6m x 1.3m. The difference between the reader antenna and the center of the plane in which the sensatag is placed is 1.8m. The sensatag was placed somewhere inside the area of interest. The objective was to estimate its position in the area.

In the first set of experiments, we studied the accuracy of the estimate as a function of the reader power. We carried out localization of the sensatag at 10 different positions and computed the average error (defined as the Euclidean distance between the true and the estimated position) as a function of power. The results are shown in Figure 4, where we see that the association method has the worst performance but is almost constant in the studied range of reader powers. The best performance of all the methods was by the weighted centroid with a reader power of 28 dBm (the accuracy was about 14cm).

\[\text{Fig. 4: The effect of reader power on the average position error of the three methods.}\]

\[\text{Fig. 5: Estimated probability of detection and the corresponding four-degree polynomial fitting.}\]

\(^1\)The reason for having two passive tags at the same location was to prevent missing a tag by the sensatag because of destructive superposition of the signals from the reader and the tag.
grid points. The results are shown in Figure 5. We can see that the probability of detection can vary considerably even for the same distance. We, however, expect this variability; it is due to the different multipath components and other factors that play role in formation of the signal received by the sensatag. We fitted the data with a fourth-degree polynomial function, which is also shown in the figure. The curve shows how the probability of detection decreases monotonically with distance.

The decrease of the probability of detection with distance is the main motivation for using the weighted centroid method. Clearly, with weighting the locations of the detected passive tags, we give higher emphasis to the detected tags that are closer than the ones that are further away from the sensatag.

For comparison of the performance of the methods, we also used the empirical cumulative distribution function (CDF) of the location error of the three methods. The results are shown in Figure 6. The CDFs of the errors confirm that the weighted centroid performs significantly better than the other two methods. For example, the probability of the error being less than 40cm is about 0.95 for the weighted centroid, 0.82 for the centroid, and 0.4 for the association-based method.

Finally, we also show the average error and the standard deviation for all the methods as functions of space (Figures 7-8). The weighted centroid-based method shows the smallest variability of the three methods across space. This is a feature that gives it an additional advantage over the other two methods.

6. REFERENCES


