Analysis of different digital filters for received signal strength indicator

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ABSTRACT
Due to high demand in Internet of Things applications, researchers are exploring deeper alternative methods to provide efficiency in terms of application, energy, and cost among other factors. A frequently used technique is the Received Signal Strength Indicator value for different Internet of Things applications. It is imperative to investigate the digital signal filter for the Received Signal Strength Indicator readings to interpret it into more reliable data. A contrasting analysis of three different types of digital filters is presented in this paper, namely: Simple Moving Average filter, Alpha Trimmed Mean filter and Kalman filter. There are three criteria used to observe the performance of these digital filters which are noise reduction, data proximity and delays. Based on the criteria, the choice of digital signal processing filter can be determined in accordance with its implementations in practice. For example, Alpha-Trimmed Mean filter is shown to be more efficient if used in the pre-processing of Received Signal Strength Indicator readings for physical intrusion detection due to its high data proximity. Hence, this paper illustrates the possibilities of the use of Received Signal Strength Indicator in different Internet of Things applications given a proper choice of digital signal processing filter.

Keywords:
Alpha trimmed mean
Digital filter
Energy security
Kalman filter
Received signal strength
RSSI
Simple moving average
Transceivers

1. INTRODUCTION
One of the parameters that is widely known in the area of wireless networking environment is Received Signal Strength Indicator (RSSI). In the IEEE 802.11 system, RSSI is the approximate strength of received signal in arbitrary units. The value of RSSI can be interpreted by voltage changes processing as measured by the circuit of a receiver [1]. However, IEEE 802.11 standards have not defined RSSI readings to be related with any standard physical parameter. The readings of RSSI are correlated with the distance between the transmitter and receiver. There are several factors that may affect the RSSI value perceived by the receiver. Some of the factors are reflection, diffraction, signal source and also obstruction in the path of the radio signal wave [2]. Furthermore, the human body can also cause attenuation on the wireless signal. Attenuation of the wireless signal is reflected on the reading of RSSI value received.

Numerous studies have discussed RSSI, its characteristics and implementations. One of the implementations of RSSI is in indoor tracking and positioning which are discussed in [3-5]. Furthermore, RSSI is also implemented for localization scenario and location-aware system as discussed in [6-10]. Apart from RSSI, there are several different techniques that can be used such as Angle of Arrival, Time of Arrival and Time Difference of Arrival. However, with the use of the RSSI technique, location estimated systems can be developed without the use of a specialized hardware. The implementation of RSSI for indoor
localization brings up some issues that should be overcome such as signal attenuation caused by obstructions, multipath fading effect and other noises that affect RSSI readings.

Hence, the objective of this paper is to compare and analyze a few types of digital filters that can be used for the processing of RSSI readings which can help to interpret it into more reliable data for different Internet of Things (IoT) related implementations. This paper is an extensive work as presented in [11]. This paper is also correlated to the work as presented in [12]. The organization of this paper is in the following manner. Section 1 is the Introduction that briefly explains the general background of RSSI, related researches and the aim of this paper. Section 2 explains the experimental setup and method used to analyze the different type of digital filters used on the RSSI readings. Section 3 presents the results obtained from experimentations along with its discussion. Section 4 concludes the paper. Lastly, acknowledgment, references and biographies of authors are included in the last three sections.

2. RESEARCH METHOD

The digital filters for RSSI readings are analyzed and compared by conducting experiments using ESP8266 WiFi serial transceiver module. It is a self-contained system on a chip integrated with TCP/IP protocol stack, WiFi direct Peer-to-Peer (P2P) and soft Access Point (AP) mode. It supports APSD for VoIP application and also Bluetooth co-existence interfaces. ESP8266 is also equipped with a self-calibrated RF [13]. There are various models of the ESP8266 WiFi serial transceiver module. The most basic and general version, which is ESP8266-01 is sufficient to be used for the analysis in this paper. Figure 1 shows the ESP8266-01 model with its pins distribution and size measurement. Through its general-purpose input/output (GPIO) pins, the module is able to be integrated with sensors or any other application specific devices. It has an onboard processing and storage capability enabling it to be integrated with sensors or many other devices for the Internet of Things application.

The ESP8266 WiFi serial transceiver module is programmed with the algorithm to behave as either transmitter or receiver. The transmitter ESP8266 is programmed to act as an access point that has a unique SSID and password. The receiver ESP8266 is programmed to connect to the WiFi Protected Access 2 (WPA2) network of the transmitter ESP8266 and read its RSSI value. There are three types of digital filters used in the experiment of this paper which are Simple Moving Average (SMA), Alpha Trimmed Mean (ATM) and Kalman Algorithm. The algorithm of each filter is programmed into the receiver ESP8266 to produce its filtered RSSI reading values. The values are then portrayed into graphs that are presented in Section 3.

Figure 2 shows the experimental setup and layout to investigate the effect of different digital filters to RSSI readings. The transmitter and receiver are positioned 1m above the ground and 1m apart from each other. They are located in between a pathway to let a human subject cross the LoS between the transmitter and receiver. The same human subject is used throughout the experiment. The crossing of a human subject is included in the experiment condition in order to observe the noises produced.
2.1. Simple Moving Average

Simple Moving Average (SMA) is a simple low pass Finite Impulse Response (FIR). It is suitable to be used on RSSI readings due to its simplicity and efficiency. SMA comes in handy to filter unwanted noise from the raw RSSI data. SMA works by taking n readings of RSSI and calculating its average to produce a single averaged reading of RSSI. In this paper, the parameter is set at n = 5. The SMA algorithm can be expressed in the equation 1 as follows:

$$Y[i] = \frac{1}{n} \sum_{j=0}^{n-1} X[i + j]$$  \hspace{1cm} (1)

where; $Y[\ ]$=Filtered RSSI value
$X[\ ]$=Raw RSSI values
$n=$Number of readings

2.2. Alpha trimmed mean

Alpha Trimmed Mean (ATM) is one of the suitable digital filters to be used in the pre-processing of RSSI readings. ATM filters unwanted noises especially random spikes. The algorithm of ATM is portrayed in the flowchart of Figure 3. The algorithm starts by initializing the window size, n and alpha, $\alpha$. Suitable parameter can be determined through offline testing phases according to the implementation of RSSI readings. In this paper, the parameters are set at n = 5 and $\alpha = 2$. The array of size n is filled with raw RSSI readings. The raw RSSI readings are then sorted from smallest to largest or vice versa. The algorithm proceeds with discarding $\alpha/2$ elements from each end of the array. The value of alpha should be as in equation 2. The filtered RSSI value is obtained by averaging the remaining elements in the array by $n-\alpha$.

$$\text{Alpha, } \alpha < \frac{1}{2} \text{ (Window Size, } n)$$  \hspace{1cm} (2)

![Algorithm of Alpha Trimmed Mean digital filter](image)

2.3. Kalman

Kalman filter is an exceptional pre-processing tool that significantly removes noises from raw RSSI readings. It is a recursive algorithm that takes the previous RSSI readings into account. A regular Kalman filter that assumes linear models is sufficient as RSSI reading is a static system. The equations for computing Kalman filter is defined by (3) to (7) as follows [14, 15]:

Prediction step:

$$\hat{x} = x$$  \hspace{1cm} (3)

$$p = p + Q$$  \hspace{1cm} (4)
Measurement update:

\[ k = \frac{p}{p+R} \]  

\[ x = \bar{x} + k(RSSI - \bar{x}) \]

\[ p = p(1 - k) \]

where; \(x\) = Filtered RSSI value
\(p\) = Estimated error
\(Q\) = Process noise
\(R\) = Measurement noise
\(k\) = Kalman gain

The value of \(R\) in this paper is determined by the standard deviation of a sample of RSSI values and \(Q\) is half of the value of \(R\) to reduce time lag. The value of \(Q\) and \(R\) are tweaked during the offline testing phase.

3. RESULTS AND ANALYSIS

The experiment as explained in Section 2 is conducted. The resulting graphs and data are presented in this section. To observe the effect of the digital filters, four samples of 500 RSSI readings are obtained with the sampling rate of 10Hz from receiver ESP8266. The result of the experiment is as shown in the graph of Figure 4.

![RSSI READINGS FROM DIGITAL FILTERS](image)

Figure 4. Received Signal Strength Indicator readings for no filter, SMA filter, ATM filter and Kalman filter

The RSSI readings obtained from the experiment are categorized under four conditions which are settling period, before crossing of a human subject, during the crossing of a human subject and after the crossing of a human subject. The period of each condition is as in Table 1.

| Category of Received Signal Strength Indicator Conditions | Start | Setting Period | Before Crossing | During Crossing | After Crossing |
|--------------------------------------------------------|------|----------------|----------------|----------------|--------------|
| Reading Start                                        | 1    | 5              | 156            | 286            | 500          |
| Reading End                                          | 4    | 155            | 285            | 500            |              |

3.1. Noise reduction

The noise reduction is observed by computing the standard deviation of the RSSI readings samples. Standard deviation is a measurement that quantifies the dispersion of data values. In this paper, the standard deviation value can indicate how close the RSSI readings are to its mean. Hence, the observation of the standard deviation value of the sample in this experiment is the best indicator of noises. It is safe to assume that the higher the value of the sample standard deviation, the higher the noises. The standard deviation values of the data samples are tabulated in Table 2.
From the values observed in Table 2, Kalman filter is observed to have the least values of standard deviation throughout the three conditions which are 0 before crossing, 0.49 during crossing and 0.00 after crossing. On the other hand, ATM filter is observed to have the highest values of standard deviation throughout the three conditions which are 0 before crossing, 2.37 during crossing and 0.51 after crossing.

Table 2. Standard deviation of data samples

| Digital Filter | Sample Standard Deviation, S |
|----------------|-------------------------------|
|                | Before Crossing | During Crossing | After Crossing |
| No Filter      | 0              | 2.49            | 0.50           |
| SMA            | 0              | 2.33            | 0.51           |
| ATM            | 0              | 2.37            | 0.51           |
| Kalman         | 0              | 0.49            | 0.00           |

3.2. Data proximity

The data proximity is a criterion to observe how close the mean values of the filtered RSSI readings to the mean value of unfiltered RSSI readings. This is to assume how accurate the filtered readings are to its actual value. The higher the difference between the mean values of filtered RSSI readings and unfiltered RSSI readings, the further it is to the actual value. The mean values of the data samples are tabulated in Table 3. From the values observed in Table 3, Kalman filter is observed to have the highest difference in mean value throughout the three conditions which are 1 before crossing, 0.99 during crossing and 0.56 after crossing. On the other hand, ATM filter is observed to have the least difference in mean value throughout the three conditions which are 0 before crossing, 0.13 during crossing and 0.03 after crossing.

Table 3. Mean of data samples

| Digital Filter | Mean (dBm)       |
|----------------|------------------|
|                | Before Crossing | During Crossing | After Crossing |
| No Filter      | -59             | -59.57          | -59.56        |
| SMA            | -59             | -59.26          | -59.50        |
| ATM            | -59             | -59.44          | -59.53        |
| Kalman         | -58             | -58.58          | -59.00        |

3.3. Delays

The delays observed in this paper are Settling Delay and Response Delay. Settling delay is the time taken for a filter to collect sufficient RSSI readings to produce an acceptable data. Response delay is the time taken for filtered readings to produce changes as affected by unfiltered input readings. The time delay can be calculated using equation 8. The sampling rate used in the experiment of this paper is 10Hz.

\[
\text{Time Delay} = \frac{1}{\text{Sampling Rate}} \times \text{Number of Reading} \tag{8}
\]

Figure 5 shows the graph of 10 RSSI readings as sampled from the result in Figure 4. It can be observed that the SMA filter has the highest settling delay by 0.5s and Kalman filter has the least settling delay by 0s. To analyze the response delay, RSSI readings between the condition of before crossing and during the crossing of a human subject as sampled from the result in Figure 4 are observed. Figure 6 shows the graph of RSSI readings with SMA filter. Figure 7 shows the graph of RSSI readings with ATM filter. Figure 8 shows the graph of RSSI readings with Kalman filter.

It can be observed that SMA filter has the highest response delay by 0.6s and Kalman filter has the least response delay by 0s. To envision the purpose of this analysis, an example of RSSI readings that is implemented in physical intrusion detection is discussed. In intrusion detection, the data collected during human crossing is considered to be very sensitive. From the graph of Kalman filtered RSSI readings in Figure 9, it can be observed that the noises during the crossing of a human subject (reading 156 to 285) are significantly smoothed where an intrusion is unlikely to be detected. Furthermore, there are no significant negative peak value that can be extracted during the period which make the data useless for a radio frequency signature identification. On the other hand, SMA filter may seem to be equally favourable to ATM filter. SMA filter has a slightly higher performance in noise reduction compared to ATM filter by 1.72%. However, referring back to the criteria requirement of the implementation example, noise reduction is weighted lesser compared to delays. ATM filter has a higher performance in delays where the data is more responsive to
changes as shown in graph of Figure 10. Higher data responsiveness will result in faster physical intrusion detection. Thence, ATM Filter is more efficient to be used in the pre-processing of RSSI readings for physical intrusion detection.

Figure 5. Graph of Received Signal Strength Indicator readings from reading 1 to reading 10 as sampled from readings in graph of Figure 4

Figure 6. Graph of Received Signal Strength Indicator readings with SMA filter

Figure 7. Graph of Received Signal Strength Indicator readings with ATM filter

Figure 8. Graph of Received Signal Strength Indicator readings with Kalman filter

Figure 9. Graph of Received Signal Strength Indicator readings with Alpha-Trimmed Mean filter

Figure 10. Graph of Received Signal Strength Indicator readings with Kalman filter

4. CONCLUSION
The comparison analysis of three different digital filters for RSSI readings has been carried out and presented in this paper. There are many options for digital signal processing filters that can be used. However, the choice of digital signal processing filters has to be aligned with the implementations and

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applications. In this paper, there are three performance evaluation criteria were observed during the experiment. They are noise reduction, data proximity and delays. The results have shown that the filter with the highest performance in noise reduction is Kalman filter followed by SMA filter and lastly ATM filter. Secondly, the filter with the highest performance in data proximity is ATM filter followed by SMA filter and lastly Kalman filter. Finally, the filter with the highest performance in delays is Kalman filter followed by ATM filter and lastly SMA filter. All the filters are highly beneficial subjected to its implementation. For example, ATM Filter is more efficient to be used in the pre-processing of RSSI readings for physical intrusion detection due to its high data proximity and responsiveness. Hence, the paper recommends that depending on the RSSI application, a specific digital filter may be used individually or in combination with different other filters.

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