Model on Optimizing Primary Spectrum Allocation using Cognitive Radio

By Omorogiwa, O. S. & Nwukor, F. N

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I. Introduction

The Smart Devices (SD), which are also referred to as Internet of Things (IoT) are increasing every day and are gaining focus because most organisations and individuals are using the device. Internet of Things refers to billions of physical machines around the globe that are linked to the internet, assembling and sharing data (Nidhi and Rajeev, 2019). Any physical entity can be transformed into an Internet of things machines if it can be linked to the Internet/Ethernet and controlled (Hsu et al, 2016).

As of 2016, the prediction of the Internet of things has advanced due to a convergence of numerous technologies, as well as wireless communication, real-time analytics, machine learning, product sensors, and embedded systems. The acceptance of Radio Frequency Identification (RFID) tags (low power chips that can communicate wirelessly) resolved some of this concern. The accessibility of broadband internet, cellular and wireless networking also helps in facilitating the growth of Internet of Things (Hsu et al, 2016). Internet of Things also finds application in checking electric grid, telecommunication at real time, and help to encourage healthy living by use of consumer machines such as linked scales or wearable heart check (Hus et al, 2016; Kang et al, 2017).

With the advancement in communication technology, the IoT machines have introduced a new class of low-power short-range wireless machines that uses radio spectrum for the switching of information (Asghar et al, 2015). The requirement for these machines are creating irresistible demand on the radio spectrum (secondary licensing) (Asghar et al, 2015); thereby causing shortage of frequency. Other wireless machines that use secondary spectrum have been facing interference (Otermat et al, 2015; Singh et al, 2014), since some range of spectrum are free (secondary spectrum), any user can use any spectrum that he/she assume is good for his/her machine not considering other users (Stankovic, 2014).

The number of Internet of Things machines is predicted to reach 200 billion by the year 2020 (Asghar et al, 2015; Stankovic, 2014). This rapid growth of internet of things machines are introducing high demand for the switching of information. Hence with the new discovery, it also brings the crisis; communication field frequency insufficiency which is at the present becoming extremely a main crisis as man discovers appliances every day.

To compensate this extraneous demand for radio spectrum, each application have need for spectrum to function, but with limited amount of frequency obtainable for proper throughput communication (secondary spectrum) (Asghar et al, 2015). As a result of this, Primary spectrum need to be analysed in order to locate any vacant spectrum that IoT machines need to utilise, to solve the shortage of radio frequency and throughput.

Cognitive radio (CR) device have been proven using a novel method to identify free and used radio spectrum. If IoT machines will be able to regulate their machines parameters, such as transmit (broadcast) power and frequency, in order to optimize their throughput at the same time minimizing intrusion to the primary spectrum license user, with the help of cognitive radio. We can predict the spectrum hole and use the free spectrum (Otermat et al, 2016).

This study will have a significant impact on spectrum allocation to secondary user, method that will use the underutilised spectrum in the primary spectrum through a novel use of CR.
To communicate between IoT devices over a 2.4/5GHz radio frequency it becomes problematic when other nearby devices are using that frequency (Dawid, 2017). Tan and Wang, (2010) proposed a variant of the Open Systems Interconnection (OSI) model for the IoT architecture. The first layer is split between Existed Alone Application System Layer and Edge Technology Layer/Access Layer. The second, third, fourth, and fifth layers are Backbone Network Layer, Coordination Layer, Middleware Layer, and Application Layer, respectively. Each of these different layers possesses its own enabling technology.

According to Experts, 30 billion connected devices will exist by 2020, these many devices competing for wireless spectrum will cause severe congestion (Zhang et al., 2012). Alleviating spectrum congestion is the primary reason for incorporating CR in IoT.

One of the most revolutionary applications of CR is addressing spectrum scarcity in wireless communications. The spectrum is scarce primarily because of the way it is licensed. CR provides the technical framework for spectrum sharing of the underutilised spectrum (Hassanieh et al., 2014; Haykin, 2015), the underutilised spectrum for use by CIoT devices will be key to the future success of ever increasing IoT networks.

Review of the spectrum scarcity research was conducted to prove that spectrum scarcity is indeed a pseudo situation (Omorogiuwa and omozusi, 2018). It is pseudo in the sense that the spectrum remains idle or vacant majority of the time. However, it cannot be utilised by anyone but the primary license holder.

This study focused on developing FM frequency allocation to secondary user without interference.

II. Method

a) Algorithm Model

i. System Architecture

There are two types of network architectures in the FM spectrum system: the allocated spectrum and the unallocated spectrum, as we know, the entire spectrum is made up of a number of channels. There are FM spectrum channel in total which can be represented as M and there are allocated FM which can be represented as K and there are unallocated FM spectrum which can be represented as L; with set PUs $K \in \{1, \ldots, K\}$ and SUs $L \in \{1, \ldots, L\}$. the PU operational in different channel can either be 1 or 0 which denote $PUi \ i \in K$ and the bitrates of such a channel can be represented as $Bi$ which for $K$ is not relevant in this model, Figure 1 represent M spectrum $=K + L$, which can be denoted as $M = 100$ Channel, if $K = 12$, that means L (unallocated) $= 100 – 12 = 88$ channel.

Figure 1: Denote the total number of FM stations

b) Channel Occupancy Model

A model is used to represent the status of each channel as discussed in System Architecture, in which there is intermittent channel switching between unallocated idle and unallocated busy, to avoid interference with the PUs allocated and to track the spectrum before transmission. The SU perform spectrum sensing, this is necessary because the SU are using the free spectrum of PUs and that means each SU can be represented by $SUj \ j \in L$. for a SU can access one frequency of the M channels where it is 0. If $PUi$ is 1 in the spectrum of interest $i \ i \in K$ is denote by $H1i$; if $PUi$ is absent, this denote by $H0i$, but this assumption will not be since, if the FM station allocated are made constant, that means the model will improve to $SUi$; which means when SU is busy, it will denote $H1i$ and when is idle, it will be $SU0i$. This assumption was made that every user is entitled to only one transceiver operating in half duplex; it is required of the SU to have spectrum nimbleness with embedded dynamic frequency selection. The likelihood of the $SU$ being active is connoted as $SU1i$; the likelihood of it being inactive is connoted as $SU0i$; that shows $SUH1i + SUH0i = 1$, Figure 2 shows the chain representation of the model.

Figure 2: Show the chain representation of the develop
From Figure 2 there are some probability discretions in regarding the resolution of active (busy) and inactive (idle).

First system probability discretion, that channel $i$ is idle and $PU_i$ is constant, true negative, which means the probability $P0/0i = PH0i(1 – Pf,i)$

Second system determination that the channel $SU_i$ is occupied and $PU_i$ remains constant that is a false positive occurring with a probability $P1/0i = PH1i,Pfi$.

Third system determination that channel $PU_i$ is dormant while $SU_i$ is functional, that is false negative occurring with a probability $P0/1i = PH1i(1 – Pfi)$

Fourth system determination that channel $PU_i$ is active as $SU_i$ (means all unallocated FM station) is active that means a true positive occurring with a probability $P1/1i = PH1i,Pfi$

As indicated, the algorithm presented in Channel Occupancy Model shows notable improvements in the alliance formation algorithm which forms the basis of the alliance. The model convergence time is quicker in decision making. This comes as a result of putting PU as constant, while multiple SUs are allowed to sense channel more than one time. Both the processes of channel sensing and access can be performed by Multiple SUs at a time, while the same channels in the PU (constant) spectrum are not sensed or accessed. The outcome thus is that there is no discord or interference of PU in the network.

**Code**

This code was used to display the graph using the simulation.

```plaintext
SP:
  val_F = Data
  IF Data < 100 then
    R_val = 100 – Data
  Else
    R_val = 100
  Goto sp:
  End
```

### III. Performance

**a) Simulation parameters**

The FM frequency spectrum range contains the network nodes distributed in it. The algorithm was used to compare PU and SU for access and dynamic spectrum sensing. The average utility of SUs in the spectrum used for comparison was the main matrix and this average utility of SUs depends on the constant (PU) value in each State. Simulation parameters of different state as shown in Table 2 this was use to vary and decide the resulting outcome on average spectrum SU utility. Table 1 shows the parameters of the default simulation that was used and the results is shown in Figures 3 to 9.

**Table 1: Simulation Parameters**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>$M$</td>
<td>Number of channels</td>
<td>100</td>
</tr>
<tr>
<td>$K$</td>
<td>Number of primary users</td>
<td>Depends on the state</td>
</tr>
<tr>
<td>$L$</td>
<td>Number of secondary users</td>
<td>Depends on the state</td>
</tr>
<tr>
<td>$PH1,i$</td>
<td>Probability of PU active</td>
<td>Constant</td>
</tr>
<tr>
<td>$PH1,i$</td>
<td>Probability of SU active</td>
<td>Varies</td>
</tr>
<tr>
<td>$PH0,i$</td>
<td>Probability of SU inactive</td>
<td>Varies</td>
</tr>
</tbody>
</table>

**Table 2: FM radio stations and their percentage utilisation**

<table>
<thead>
<tr>
<th>S/N</th>
<th>State</th>
<th>FM Stations</th>
<th>$S_{unallocated}$ MHz</th>
<th>Percentage Underutilisation</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Abia</td>
<td>10</td>
<td>18</td>
<td>90</td>
</tr>
<tr>
<td>2</td>
<td>Adamawa</td>
<td>6</td>
<td>18.8</td>
<td>94</td>
</tr>
<tr>
<td>3</td>
<td>Akwa Ibom</td>
<td>8</td>
<td>18.4</td>
<td>92</td>
</tr>
<tr>
<td>4</td>
<td>Anambra</td>
<td>19</td>
<td>16.2</td>
<td>81</td>
</tr>
<tr>
<td>5</td>
<td>Bauchi</td>
<td>4</td>
<td>19.2</td>
<td>96</td>
</tr>
<tr>
<td>6</td>
<td>Bayelsa</td>
<td>6</td>
<td>18.8</td>
<td>94</td>
</tr>
</tbody>
</table>
Each state in the Country has an average of 5-12 FM radio stations. Since there are 100 possible radio channels that could be occupied at any given location (latitude and longitude coordinate), that show the FM radio channels in Nigeria is 89% underutilised.

b) Alliance formation

The number of channels available determines the number of alliance available to join each channel. To aid visualization of the alliances, charts were drawn to show the total FM spectrum. SUs and PUs in the network were generated with the assumption as discussed in Algorithm Model and Channel Occupancy Model and the colour of the node as shown in Figures 3 to 9.

Source: Omorogiuwa and Nwukor
Figure 3: Network representation of Abia State total FM spectrum, PU and SU representation

Figure 4: Network representation of Adamawa State total FM spectrum, PU and SU representation

Figure 5: Network representation of Akwa Ibom State total FM spectrum, PU and SU representation

Figure 6: Network representation of Anambra State total FM spectrum, PU and SU representation

Figure 6: Network representation of Bauchi State total FM spectrum, PU and SU representation

Figure 8: Network representation of Bayelsa State total FM spectrum, PU and SU representation
From the chart that was shown in Figure 3-9, First system decision that channel i is idle and PU is constant, true negative, which means the probability \( P_0/0i = PH_0i(1 - Pf,i) \) as described in Channel Occupancy Model was used to formulate the results in Figures 3 – 9. It was an indicator that there is free spectrum in the FM station. This indicates how many SUs and PUs are in each alliance. The organisations of PUs and SUs across the alliances are what the spectrum sensing and access algorithm seeks to optimize, in reflecting the average utility of the spectrum.

Second system determination that the channel SU is occupied and PU remains constant that is a false positive occurring with a probability \( P_1/0i = PH_1i,Pfi \), using Abia, Adamawa, Delta State. The assumptions are represented in Figures 10 to 12.

From the chart in Figures 10 – 12, it was shown that if any spectrums from the secondary user are allocated it becomes constant that means the spectrum cannot be reallocated to any user.

IV. Conclusion

In this paper, better approaches of spectrum sharing were obtained and analysis was done. Algorithm model with channel occupancy model for multi-channel dynamic access was presented. The details of a collaborative spectrum sharing technique were presented from an analytical perspective and an adapted algorithm for fast convergence in a hedonic coalition technique was soon. From the technique shown, if compared with other model, that put the primary license holder as parity 1 and secondary user as parity 2, the system will use time to scan if parity 1 is available or when using the spectrum, when parity 1 is
active (primary user), the secondary user will disconnect, which will lead to loss of packet. But the performance of the model can operate in the primary user spectrum and secondary user spectrum without causing interference to the primary spectrum. It will be better if NCC approve this model because if implemented, it will protect the primary license holder and will also protect the secondary user.

**References Références Referencias**


