Measurement and Modeling of Path Loss for GSM Signal in a Sub Urban Environment over Irregular Terrain-Review

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Abstract: This work presents measurement and modeling of path loss propagation for three GSM operators in Mubi town. We adopted COST 231-Hata model as a reference model and modified the model based on the experimental data measured in the study area. Autoregressive neural network (ARNN) predicted that the average path loss for each GSM operator is 16.96dB, 16.077dB and 24.5dB with their corresponding mean square error (MSE) as 1.76329dB, 1.2958dB and 2.37112 for AIRTEL, MTN and GLO respectively. GLO operator experienced more path loss followed by the AIRTEL andMTN is the least.

Keywords: Signal strength, Modeling, propagation, Environment, Obstacle and GSM

1. Introduction

The fundamental aim of any global system for mobile communication (GSM) service provider is to provide an outstanding service to its subscriber, and meaningful service may be achieved by having an excellent radio communication system between the base station (BS) and the mobile unit (MU). So, it becomes necessary to constantly investigate and model the path loss; as there are many factors that could hamper the effectiveness of the communication system between the BS and MU. Understanding path loss [1][13]gives us an idea on how to determine some certain parameters for an effectiveradio communication system for example, antenna; gain, height and location. GSM subscribers in Nigeria frequently complained of inter and intra connectivity, call drop, crosses talk in call conversion. This prompted us to study the path loss as one of the probable causes of these problems mentioned above.

An ideal propagation means equal propagation in all directions. Unfortunately, in real life situation, it is not feasible due to some factors between the (BS) and the (MU) that attenuates the signal [3], such factors may be responsible for reflecting, refracting, absorbing, or scattering the GSM signal before reaching the MU. Moreover, investigation of the path loss may give us almost or an accurate bill of these factors responsible for the attenuation before undertaken a design and implementation of aradio communication path.

Hata, Okuruma-Hata, Sakagami- Kuboi and Walfisch-Ikegami have developed different models using different methods [4][6][7][8][9][10], such as empirical, stochastic or deterministic method. FirstHata, COST 231-Hata and Okumura-Hata model are popularly used [5].Butyet, seems not to have met up or served efficiently with all the environments around the globe. However, these models are which used as referencemodels acceptably need improvement based on the environmental factors.In our previous work [14], we used empirical method and determined the mean square error (MSE) of the propagation, but in this type of method, the accuracy cannot be ascertained.In this work, the accuracy of the prediction is tested using the autoregressive neural network.

In this work, we propose to use the COST 231 Hata model as our reference model. Our interest is to adopt this model, modify it based on our experimental data measured in the study area. COST 231-Hata is chosen in this work because the antenna height (BS) is higher than all the roofs and the trees in the study area. We will use the autoregressive neural network to predict the losses (errors) in the attenuated signal determine its MSE in decibels and validate the prediction performance by evaluating the R-square of the regression model.We will also, use neural network clustering to determine the frequency of occurrence of the path loss measured in the study area.

1.1 Theoretical Propagation Model

Usually, propagation model starts with the free space model. Free space loss is an area whereby the GSM signal is neither reflected, diffracted nor absorbed, this may refer to as ideal propagation. But in investigating the path loss, free space loss is not adequate to quantify the amount of losses or the path loss between the BS and the MU. In order to come out with the overall path loss from source to destination the sum of the combine effect of the free space loss, reflection, diffraction and absorption signals may be needed. The power density received between the BS and the MU, may be given as

$$S = \frac{P_o}{4\pi l^2} \tag{1}$$

Where S is the power flux, P_0 is the power transmitted by the BS and the denominator of (1) is the flux area. If there are no feed line losses, the power delivered to the MU may be given by (2).

$$P_1 = SA_1 \tag{2}$$

For hypothetical isotropic antenna the MU will have an area given by

$$A_1 = \frac{\lambda^2}{4\pi} \tag{3}$$

Where λ is the wavelength of the propagation, combining (1), (2) and (3) gives us (4)

$$P_1 = P_0 \left(\frac{\lambda}{4\pi d}\right)^2 \tag{4}$$

Path loss is usually determine by the difference between the power transmitted by the BS and power received by the MU as given by (5)

$$L_0(dB) = P_0 - P_1$$
 (5)

Substituting (4) into (5); we will yield

$$L_0 = 20\log(4\pi) + 20\log(d) - 20\log(\lambda)$$
(6)

Simplifying (6) with respect to the frequency, the generic free space path loss equation is obtained as in (7)

$$L_0 = 32.4 + 20\log(d) + 20\log(f)$$
(7)

As we know path loss is the function of distance as given in (8)

 $P_l(dB)d^{\alpha}$ (8)

Simplifying the (8) in logarithmic form, we will have (9). $P_i(dB) = L_0 + \alpha \log(d)$ (9)

Where P_i is the path loss between the BS and the MU, L_0 is the propagation constant known as free space loss and α is the propagation index, d is the distance between the BS and the MU.

1.2 COST 231- Hata Model

The COST-231-Hata model is an extension of the Hata model which is developed to address the shortcomings that Hata model is not capable of addressing [2]. This model has similar characteristics with Okumura-Hata model, but differs bythe following; frequency 1500 MHz to 2000 MHz, and the link distance, 1km to 20km [2][3]. COST 231-Hata model is given by

$$L_{i} = 46.3 + 33.9 \log f_{c} - 13.82 \log h_{B} - a(h_{m}) + \log d$$

(44.9 - 6.55 \log h_{m}) + C_H (6)

Where $a(h_m) = 3.20(\log 11.75h_m)^2 - 4.97$, for f > 400 MHz and C_H is the correction factor usually given by 3dB for urban environments.

2. Study Area and Method of Data Collection

The data is collected in a day time with MU in Mubi town, Nigeria at different locations for three GSM operators which includes AIRTEL, MTN and GLO. The city is typically a sub-urban area which consists of buildings approximately 15-25m, significant number of trees;Tamarin, Shear butter, Locust bean, Barasusaethiopus "Giginya", Neem,

Mahogany, Date palm"Dipino" cashew and Guava; their height range between 7m - 12m, river Yazaram that divides the city into two and usual human activities like vehicle movements during the time of the day. On the MU, there are 0 to 5bars signifying the signal strength received at the destination. The network bars on the MU range from 0 to 5 bars, the lowest bar is 0 and is the weakest signal while the signal strength increases as the number of the bars increases, which means that the strongest signal is 5 bars. On each location at least 100 samples of those bars were taken making a total of 2.000 observations for each GSM operator. Usually, GSM signal strength is measured in -dBm; that is, the power measured (dB) multiple by the distance between the transmitter and MU receiver. The useful range is from -50dBm to -110dBm in a frequency range of 900MHz to 1800MHz or 1900MHz depending on the environmental requirements. The smaller the number of the dB received by the MU the worse the reception or QoS. Therefore, -50dB is much better than -110dB. In this work, we assigned 0 (no bar) to -105dB, subsequently, 1bar = -95dB, 2bars = -85dB, 3bars = -75dB, 4bars = -65dB and 5bars = -55dB. The MU from transmitter is located in different positions. Starting, from; 500m to 10,000m.

3. Neural Network Prediction

The sample of the signal received is thenanalyzed using the ARNN in order to obtain a weight that could give us desired output. In the process, we trained the network several times so that the network will learn about the data and predict the error that could be responsible for the signal attenuation, as the training continues; we noticed that the successive training pairs negate the change in each stage of training to a reasonable point, that is up to the stage that the error is minimal. This now confirms the degree of the attenuation on the propagation, at that point the neural network (NN) tend to recognize and differentiate between the actual propagation energy (power) as well as the amount of the attenuation (error). Fig. 4 to 6 shows the overall outcomes of NN prediction after the training, validation and test. The accuracy of the NN prediction is validated by R-square and the MSE, each prediction is also noted.

MSE gives us the difference between values predicted by the model and the data observed which provides a good measure of accuracy. R - square sometimes refers to coefficient of determination or coefficient of multi determination for multiple regression, as mentioned earlier, this parameter shows how close the data fits the model, low value of R-square signifies poor prediction and high value give better prediction which is usually from 0 to 1, the performance of the prediction increases when R - square gets closer to 1.

A. Neural Network Clustering

Clustering simply means grouping elements together based on their properties, appearance, contexts and other features. We trained the experimental data collected using neural network clustering, the network learned about the data and produced clusters of the power measured. Figure 1, is the clusters of the path loss measured for the AIRTEL GSM operator. International Journal of Science and Research (IJSR) ISSN (Online): 2319-7064 Impact Factor (2012): 3.358



Figure1: SOM of weight positions path loss for the AIRTEL

As we can see, the path loss measured is mostly $135 \le P_i \le 140$. Figure 2, shows the clusters of the path loss measured for the MTN GSM operator.



Figure 2: SOM of weight positions path loss for the MTN

Here also the path loss measured is almost similar to that of the AIRTEL mostly from $135 \le P_i \le 140$. Figure 3, gives the path loss measured for the GLO GSM operator.



Figure 3: SOM of weight positions path loss for the GLO

In this case, it seems GLO operators experienced more path loss than the two other GSM operators; the path loss here is mostly $136 \le P_i \le 145$.

4. Results and Discussion

The NN path loss predictive models for each of the GSM operators are given in (10) to (12).

$$P_{airtel} = 1.04 * t \arg et + 0.32 \tag{10}$$

Where α corresponds to 1.04in (10), this seems to have a good propagation index. In literature the best propagation index is between 0 to 1, [11] target corresponds to the log (d), 0.32 corresponds to the L₀ and P_{airtel}is the path loss experienced by the AIRTEL GSM operator. Figure 4, presents the NNautoregressive model.



Figure 4: NN autoregressive prediction for AIRTEL operator

NN predictive model for the MTN GSM network operator is given in (11).



Figure 5:.NN autoregressive model for MTN operator NN path loss predictive model for the GLO network operator is given in (11).

$$P_{olo} = 1.1 * t \arg et + 6.9 \tag{12}$$

Parameter α =1.1, L₀=6.9



Figure 6: NN autoregressive model for GLO operator

Table 1, shows the R-square, MSE values predicted using the models (10) to (12). This gives the average path loss for the entire GSM operator.

Table 1: R-square, MSE and P₁ values

Tuble 1. R square, most and r values			
Operators	R-square	MSE (dB)	$P_l(dB)$
AIRTEL	0.99628	1.76329	16.9600
MTN	0.99945	1.29580	16.0770
GLO	0.99454	2.37112	24.5000

The COST 231 Hata model is then modified using the values of MSE in Table 1 for each GSM operator as given in (13), (14) and (15) for AIRTEL, MTN and GLO respectively.

$$L_{airtel} = 44.54 + 33.9 \log f_c - 13.82 \log h_B - a(h_m) + \log d (44.9 - 6.55 \log h_m) + C_H$$
(13)
$$L_{mtn} = 45.01 + 33.9 \log f_c - 13.82 \log h_B - a(h_m) + \log d (44.9 - 6.55 \log h_m) + C_H$$
(14)

$$L_{glo} = 43.93 + 33.9 \log f_c - 13.82 \log h_B - a(h_m) + \log d (44.9 - 6.55 \log h_m) + C_H$$
(15)

The modified models and COST 231 are shown in Figure 7, the blue color graph represents the COST 231-Hata model, red color represents the MTN modified model, green represents the AIRTEL modified model and the brown color represents the GLO modified model, as we can see the GLO operator experienced more losses followed by AIRTEL operator then MTN is the least. The reason may be features of the environment studied is different from where COST 231-Hata is modeled and the variation of the losses among the three GSM operators in the same city may be due to the fact that the BSs are not located in the same area.



Figure 7: The COST 231 and the modified models

5. Conclusion

Path loss is an important parameter that one needs to know before undertaking the design or improving the existing radio frequency communication path. In this work, we used the autoregressive neural network to the predict path loss for three GSM operators in Mubi town. The neural network predicted that the average GSM signal strength is attenuated at 16.96dB, 16.077dB and 24.5dB with the corresponding MSE of 1.76329dB, 1.2958dB and 2.37112dB for AIRTEL, MTN and GLO respectively. Then each MSE value is used to modify the reference model adopted and neural network clustering determined the weight position of the power received. This work demonstrated that autoregressive neural network is a good tool for path loss prediction with precision.

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