

# Prediction Characteristics of Quasi-Moment-Method Calibrated Pathloss Models

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## ABSTRACT

This paper investigates the pathloss prediction characteristics of basic models, subjected to calibration with the use of a novel technique, here referred to as the Quasi-Moment-Method, QMM. After a succinct description of the QMM calibration process, the paper presents computational results involving the calibration of three different basic models-the SUI, ECC33, and Ericsson models. The results reveal that the QMM typically reduces mean prediction (MP) and root mean square (RMS) errors by several tens of decibels. One other novelty introduced by the paper, is a comparison of contributions to total predicted pathloss, by components of the basic models, and their corresponding QMM-calibrated versions.

## General Terms

Wireless Communications, Radiowave Propagation, Empirical Modeling

## Keywords

Method of Moments, Pathloss, Least Square Solution

## 1. INTRODUCTION

In a recent publication, Zhang et. al. [1], described pathloss model calibration (or tuning) as an empirical modeling process, in which the component parameters of a basic model are systematically moderated, using field measurement data. Quite a few model calibration techniques have been reported in the open literature. Representative examples of these include the metaheuristic approach involving a swan intelligence algorithm, described by Benedicic and his associates [2], for tuning concerning an LTE network. Others are the Least Square Method utilized by Keawbunsong et. al, [3], for an urban DVB-T2 system, and the use of the ATOLL radio planning software by Popoola et. al. [4]. Keawbunsong et. al, reported RMSE values of between 6.8dB and 7.2dB, whilst Popoola et. al, informed that their approach resulted in the reduction of mean prediction error by 47.4%. Bolli and Khan [5] developed the Linear Minimum Least Square Error approach, with which pathloss in certain UHF/VHF bands was predicted within a lower RMSE error bound of 13.48dB. The quadratic regression technique adopted by Nisirat et.al [6], focused on the modification of the conventional Hata model for pathloss prediction at 900MHz. This ‘tuning’ approach replaced the sub-urban correction factor of the Hata model with a ‘terrain roughness correction factor’ to record reductions of up to 3dB in RMSE values, which nonetheless, remained greater than 7dB in the best case.

The Quasi-Moment-Method utilized in this paper for calibration with two different sets of measurement data,

recorded very impressive RMSE and MPE values. For example, in the case of the use of measurement data available from the publication by Mawjoud, [7], the best performing calibrated ECC33 models recorded RMSE values as low as 3.01dB; and when utilized with measurement data obtained by the authors for two sites in Lagos, Nigeria, RMSE values of between 0.91dB and 3.13dB were recorded for all three calibrated basic models.

## 2. FORMULATION

Let

$$P_{IB} = p_{I0} + p_{I1} + \dots + p_{In}, \quad (1)$$

represent a generic basic pathloss model to be calibrated with the use of field measurement data given as

$$P_{mea} = \left\{ p_{mea}(d_k) \right\}_{k=1}^K \quad (2)$$

The model calibration problem is then that of determining a model

$$P_{IQ} = c_0 p_{I0} + c_1 p_{I1} + \dots + c_n p_{In} = P_{mea}, \quad (3)$$

such that the weighted Euclidean semi-norm of the error function

$$\| \varepsilon \|^2 = \| P_{IB} - P_{IQ} \|^2 = \sum_{k=1}^K w_k | P_{IB} - P_{IQ} |^2, \quad (4)$$

assumes its minimum possible value. The set  $\{ c_j \}_{j=0}^n$

appearing in Eqn. (3) are unknown coefficients (here referred to as ‘model calibration coefficients’) to be determined, whilst the  $\{ w_k \}_{k=1}^K$  of Eqn. (4) are weights, set equal to 1, throughout this paper. The solution to the ‘least square approximation’ problem posed by the foregoing discussions is obtained in a manner similar to that provided by Dahlquist and Bjorck, [8], in the following manner.

First, we define the inner (or scalar) product of two real valued continuous functions  $f_1, f_2$ , as

$$\langle f_1, f_2 \rangle = \sum_{k=1}^K f_1(d_k) f_2(d_k) \quad (5)$$

Next, a set of ‘testing’ functions are prescribed as the set  $\{ p_{I0}, p_{I1}, \dots, p_{In} \}$ , which are exactly the same as the functions appearing in Eqn. (1). Then, the inner product of both sides of Eqn.(3) is taken with each ‘testing’ function, to

yield the set of equations given as

$$\begin{aligned} c_0 \langle P_{10}, P_{10} \rangle + c_1 \langle P_{10}, P_{11} \rangle + \dots + c_n \langle P_{10}, P_{1n} \rangle &= \langle P_{10}, P_{mea} \rangle \\ c_0 \langle P_{11}, P_{10} \rangle + c_1 \langle P_{11}, P_{11} \rangle + \dots + c_n \langle P_{11}, P_{1n} \rangle &= \langle P_{11}, P_{mea} \rangle \\ \dots & \dots \dots \dots \dots \dots \dots \\ c_0 \langle P_{1n}, P_{10} \rangle + c_1 \langle P_{1n}, P_{11} \rangle + \dots + c_n \langle P_{1n}, P_{1n} \rangle &= \langle P_{1n}, P_{mea} \rangle \end{aligned} \quad (6)$$

In matrix format, Eqn. (6) can clearly be rewritten as

$$[P_i](C) = (P_{mea}^*), \quad (7)$$

so that the unknown coefficients (and desired solution to the approximation problem) emerge as

$$(C) = [P_i]^{-1}(P_{mea}^*) \quad (8)$$

Because the process is similar to the method of moments originally developed by Harrington [9] for the solution of electromagnetic field problems, it is here referred to, as the ‘Quasi-Moment-Method’, QMM.

### 3. PREDICTION CHARACTERISTICS

#### 3.1 Candidate Basic Models

For the purposes of investigating the prediction characteristics of the QMM, four basic models, namely, the Ericsson, SUI, and ECC33 (medium city and large city) models, [3], [7], are selected as candidates. In the case of the ECC33 models, the outcomes of QMM calibration will be of the form

$$P_{IQECC33} = G_{fr}^* + G_{bm}^* - G_{hte}^* - G_{hre}^*, \quad (9)$$

where the free space attenuation factor,  $G_{fr}^*$  is given by

$$G_{fr}^* = c_0(92.4) + c_1(20 \log_{10} d) + c_2(20 \log_{10} f), \quad (9a)$$

the basic median pathloss factor,  $G_{bm}^*$ , by

$$G_{bm}^* = c_3(20.41) + c_4(9.83 \log_{10} d) + c_5([7.894 + 9.56 \log_{10} f] \log_{10} f) \quad (9b)$$

and  $G_{hte}^*$ , the transmitter antenna height correction factor, by

$$G_{hte}^* = c_6 \left( 13.958 \log_{10} \left( \frac{h_{re}}{200} \right) \right) + c_7 \left( 5.8 \log_{10} \left( \frac{h_{re}}{200} \right) (\log_{10} d)^2 \right) \quad (9c)$$

The receiver antenna height correction factor,  $G_{hre}^*$ , is specified differently for medium-sized cities, and large cities. It is, for medium-sized cities, given by

$$G_{hre}^* = c_8(42.57(\log_{10} h_{re} - 0.585)) + c_9(13.7 \log_{10} f (\log_{10} h_{re} - 0.585)) \quad (9d)$$

and by

$$G_{hre}^* = c_8(0.759 h_{re}) + c_9(-1.862), \quad (9e)$$

for large cities.

In Eqns. (9), frequency (f) is in GHz, and distance (d) from the transmitter, in km.

QMM-calibrated Ericsson models will be of the form

$$\begin{aligned} P_{IQ-Eric} = c_0(36.2) + c_1(30.2 \log_{10} d) + c_2(-12 \log_{10} h_{re}) + c_3(0.1 \log_{10} d \log_{10} h_{re}) \\ + c_4(-3.2 \log_{10} (11.75 h_{re})^2) + c_5(g(f)) \end{aligned} \quad (10)$$

Frequency (f) in this case, is in MHz, and distance (d) from the transmitter, in m.

Finally, the QMM-calibrated SUI models will admit representation according to

$$\begin{aligned} P_{IQ-SUI} = c_0 \left( 20 \log_{10} \left( \frac{400\pi}{\lambda} \right) \right) + c_1 10\gamma \log_{10} \left( \frac{d}{100} \right) + c_2 \left( 6 \log_{10} \left( \frac{f}{2000} \right) \right) \\ + c_3 \left( -10.8 \log_{10} \left( \frac{h_{re}}{2000} \right) \right) + c_4(S) \end{aligned} \quad (11)$$

All the parameters (including pathloss exponent ( $\gamma$ ) in Eqn. (11) are as defined in [7], for ‘terrain type-B’, and for all examples considered in this paper, correction factor for shadowing (denoted by ‘S’) is taken as 8.5. Distance (d) from transmitting antenna is in meters, and operating frequency (f) is in MHz

Using measurement data available (through ‘GETDATA’ <https://getdata-graph-digitizr.com>), a commercial graph digitizer software) from Figs (1) and (3) of [7], as well as data from field measurements by the authors in Lagos Island, Nigeria, the four candidate basic models were calibrated to yield the results described in the ensuing discussions.

#### 3.2 Faysala and Industrial Zone [7]

Outcomes of the QMM-calibration of the four candidate models, using field measurement data from Figure (1) of [7] are described by the following solutions to Eqn. (8), for the model calibration coefficients for the four models.

$$\begin{aligned} (c_{m SUI})_{m=0}^4 &= [0.4952, 0.4870, 1.2953, 0.4770, 5.1634] \\ (c_{m Eric})_{m=0}^5 &= [0.2954, 0.7763, -1.4289, -6.4964, 7.7587, 0.5817] \\ (c_{m ECC-L})_{m=0}^9 &= \begin{bmatrix} 0.9007, -0.6581, -2.7957, -0.1217, 6.4635, \\ 7.8773, 2.7915, 3.1839, -10.8165, -5.1786 \end{bmatrix} \\ (c_{m ECC-M})_{m=0}^9 &= \begin{bmatrix} 0.5619, -0.5027, -5.4348, -0.2971, 6.1436 \\ -3.8484, 2.4002, 3.1838, 2.0197, -4.8028 \end{bmatrix} \end{aligned} \quad (12)$$

The corresponding pathloss prediction profiles are displayed in Fig. (1) below. And it is apparent from the profiles that all four calibrated models very closely match measurement data, and perform considerably much better than the basic models from which they derive. It is a matter of interest to mention here in that connection that the profile for the basic Ericsson model displayed in Mawjoud’s [7], Figure 1 (and indeed, in all other Figures in the publication) is incorrect. It has been verified by the authors that the development owes to use of km (rather than m) for the computations that informed the profile.

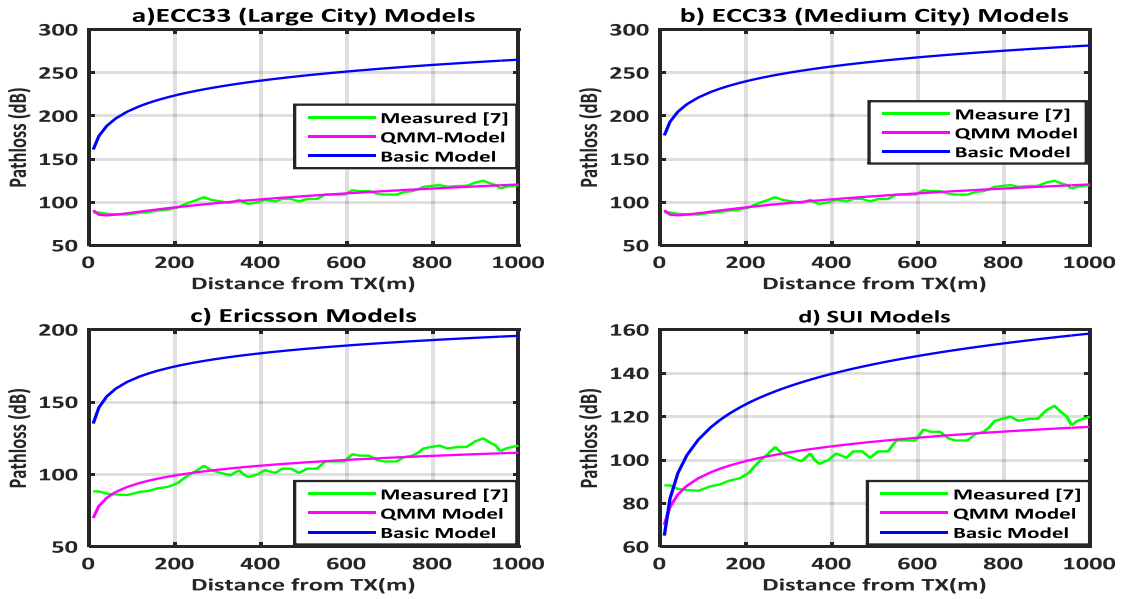


Figure 1. Pathloss profiles for Calibrated and corresponding basic models-Faysala (Fig. (1) of [7])

In the case of the ‘Industrial Zone’ (Fig. (3) of [7]), the model calibration coefficients were obtained as

$$\begin{aligned}
 (c_{m\ SUI})_{m=0}^4 &= [0.2065, 0.5540, 1.0380, 1.0316, 5.1870] \\
 (c_{m\ Eric})_{m=0}^5 &= [-0.4109, 0.8795, -0.2293, 2.6857, -4.7948, 0.2929] \\
 (c_{m\ ECC-L})_{m=0}^9 &= [-0.5473, -3.8383, 18.4073, 3.5545, 13.7333, 16.2363, \\
 &\quad -3.5935, 4.8484, -3.7268, 1.2714] \\
 (c_{m\ ECC-M})_{m=0}^9 &= [0.5443, -5.0415, 0.5966, 1.1186, 16.1532, 20.5622, \\
 &\quad -0.4552, 4.8482, -0.9866, 13.3668]
 \end{aligned}
 \tag{13}$$

Profiles of the pathloss predicted by these models as well as those of the corresponding basic models are displayed in Fig. (2).

Comments earlier made for Fig. (1) also generally apply in the case of Fig. (2) and these comments are supported by the statistical performance metrics of Table (1).

It is readily observed from the table that the best performing calibrated model is the ECC (large city) model, which reduced MPE from -134.77dB to -0.008dB, and RMSE from 135.53dB to 3.016dB, in the case of Faysala. These metrics are closely

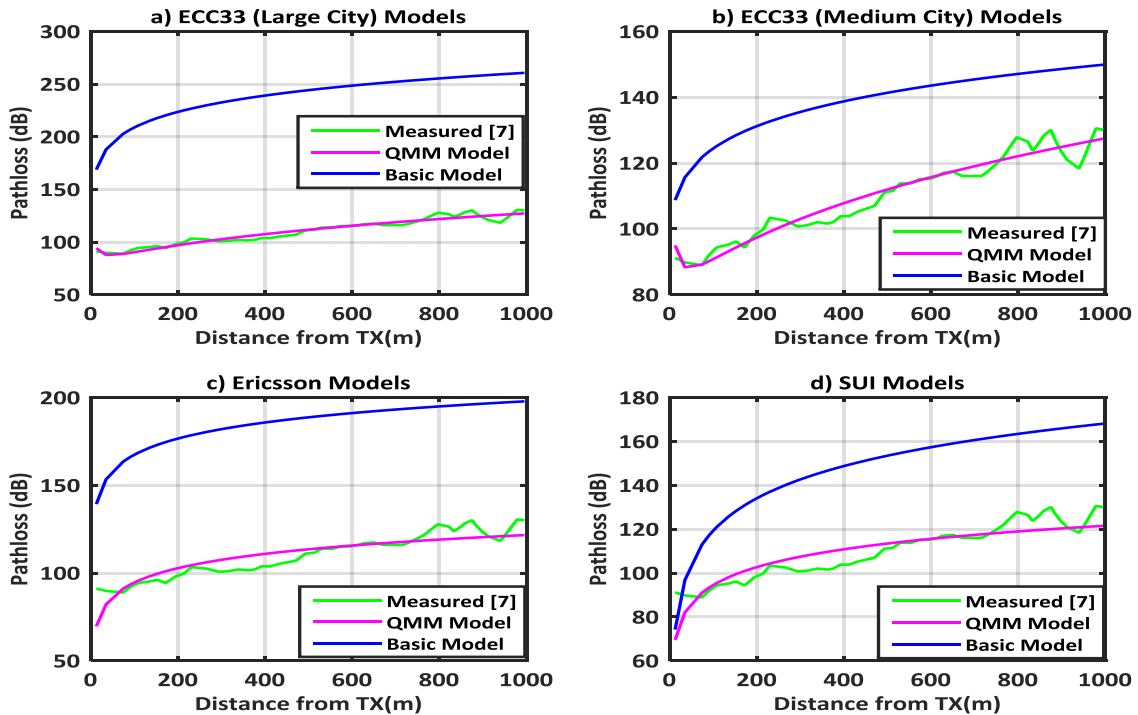


Figure 2. Pathloss profiles for Calibrated and corresponding basic models-Industrial Zone (Fig. (3) of [7])

Table (1). MPE and RMSE Metrics for the pathloss profiles of Figs. (1) and (2)

Site	Faysala- Figure 1 [7]							
Model	Ericsson		SUI		ECC33-L		ECC3-M	
	Basic	QMM	Basic	QMM	Basic	QMM	Basic	QMM
MPE	-77.265	0.242	-32.830	0.022	-134.77	-0.008	-151.20	-0.027
RMSE	77.532	5.530	34.832	5.523	135.53	3.015	151.87	3.015

Site	Industrial Zone- Figure 3 [7]							
Model	Ericsson		SUI		ECC33-L		ECC3-M	
	Basic	QMM	Basic	QMM	Basic	QMM	Basic	QMM
MPE	-75.008	-0.102	-38.10	0.000 2	129.17	-0.066	-28.598	-0.343
RMSE	75.252	5.908	39.48	5.981	129.58	3.026	29.071	3.046

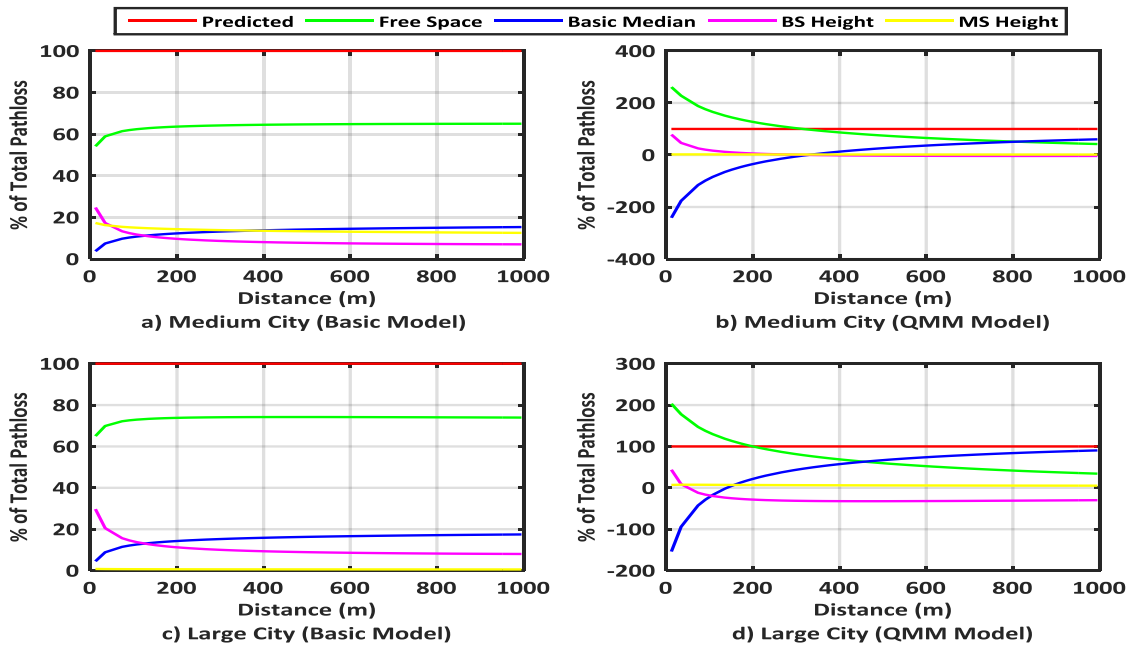


Figure 3. Contributions to predicted pathloss of Fig. (2) by component parameters of the ECC-33 models

followed by corresponding metrics for the ECC33 (medium-sized city). The calibrated SUI model recorded the best MPE of 0.0002dB in this case, but its QMM- RMSE performance of 5.523dB is some 2dB poorer than those of the ECC-33 models. Similar comments apply for the profiles of the 'Industrial Zone', though in this case, the QMM-MPE performance of SUI model was slightly lower than the corresponding ECC-33 models.

The illustrations of Figs. (3) and (4) describe contributions (as

percentages of the net) to net predicted pathloss of Fig. (2), by component parameters of the candidate basic models and their QMM-calibrated versions.

According to the computational results utilized for the curves of Fig. (3), contributions to the net predicted pathloss by calibrated ECC-33 large city model's free-space attenuation factor varied from 203% of the net, close to the transmitter, and decreased to 34.24%, 1km away from the transmitter. These

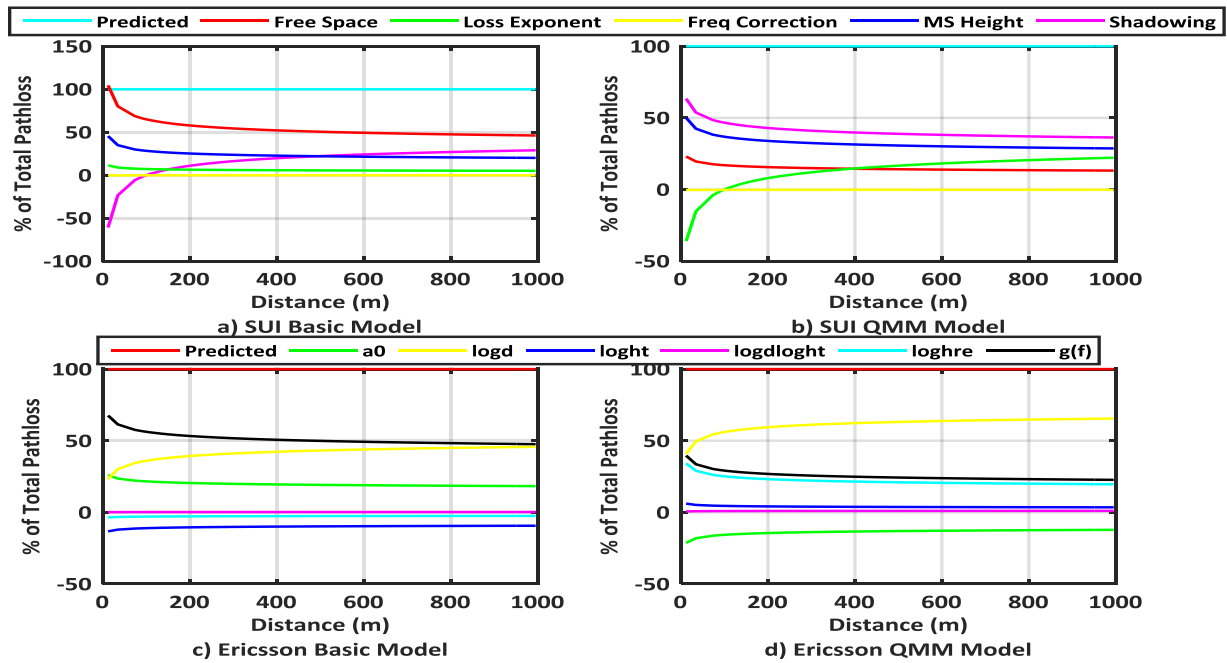


Figure 4. Contributions to predicted pathloss of Fig. (2) by component parameters of the Ericsson and SUI models

Table 2: Identity and location of candidate Lagos BTS'

Site ID	City/District	Capacity Dimension Area	Longitude	Latitude	BSC
LGO747	Ikeja	Lagos Mainland	3.3680552	6.6195586	ELGBS08
LGO749	Ikeja	Lagos Mainland	3.3256087	6.6481078	ELGBS11
LGO788	Lagos Island	Lagos Island	3.3906	6.45455	ELGBS17
LGO789	Lagos Island	Lagos Island	3.39282	6.45889	ELGBS17

contributions were in large part, moderated by those from the 'basic median' pathloss, which varied from about -153% close to the transmitter, to about 90.45%, 1km away. The transmitter antenna height gain factor contributed 43.77% close to the transmitter and about 29.8%, 1km away. The least (but nonetheless significant) contributions came from 'receiver antenna height gain factor' whose contributions varied from 7.01% at the near end, to about 5.2%, at  $d=1$ km. Corresponding contributions by the component parameters of basic large city model are 'free space attenuation', 64.98% to 73.94%, basic median pathloss', 4.9% to 17.47%, 'transmitter antenna height gain factor', 29.71% to 8.06%; and 'receiver antenna height gain factor', from 6.8% to about 0.55%. A similar trend was exhibited by the calibrated 'medium city' model, though the actual numerical values were significantly different from those of the 'large city' model, for all components. From Fig. (4), and using the example of the SUI profiles, computational results reveal that the 'free space attenuation' component of the calibrated model contributed 63% close to the transmitter, and about 13.17%, at the far-end. Contributions from the 'pathloss exponent' component varied from -36% at the near end to about 20%, 1km away from the transmitter. The 'frequency correction' factor contributed -0.4%, close to the transmitter, and about 0.23% at the far end. Interestingly, the 'receiver antenna' height correction factor for the calibrated model contributed about 50% of the net,

close to the transmitter and close to 28% at the far end. Contributions by the 'shadowing correction factor' varied from 63% at the near end to about 36%, 1km away from the transmitter. The effects of QMM calibration are reflected by the corresponding computational results for the basic model, which include (104%, 46%), for the 'free space attenuation component, (-60%, 28%), for the 'pathloss exponent' component, (-0.3% -0.1%), for frequency correction', (58%, 20.57%), for 'receiver antenna height correction', and (1.4%, 5.6%), for 'shadowing correction.

### 3.3 Ikeja and Lagos Island

QMM model calibration for the examples described in this section, was effected with use of samples of data obtained from field measurements by the authors, for four BTS' located at Lagos Mainland and Lagos Island, in Nigeria. Two of these transceivers operated at 900MHz, and the other pair, at 1800MHz. Identification particulars of the BTS' involved are described in Table 2.

In addition to a TOSHIBA (EQUUM) laptop, field measurement equipment included a SONY-Ericsson (C72) compatible mobile station, a USB GPS instrument, TEMS dongle, handheld spectrum analyzer (SA2650-TEKTRONIX), and a GARMIN handheld GPS device.

Essentially, measurements were carried out using the 'Drive

through' approach involving continuous measurements and multiple recording of data samples from a principal cell and at least six neighboring cells, as explained elsewhere, [10]. Thereafter, selective monitoring of each cell was conducted as a means of validating the measurements. Data utilized for calibration are averages, taken over several measurements.

### 3.3.1 Pathloss profiles for Ikeja

For the computational results described in this section, operating frequency is 900MHz, transmitter antenna height is 30m, and the receiver antenna height is 1.5m. The corresponding solutions to the QMM calibration problem (for the four candidate models) were obtained, for BTS1 (LG0747) as

$$\begin{aligned} (c_{m \text{ SUI}})_{m=0} &= [-4.8465, 0.2189, 25.8193, 7.5108, 32.0665] \\ (c_{m \text{ Eric}})_{m=0} &= [3.3846, 0.3479, 7.5532, 0.4204, -10.3100, 0.7262] \\ (c_{m \text{ ECC-L}})_{m=0} &= \begin{bmatrix} 0.8732, 0.9306, 6.4257, 1.6319, 1.8595, -1.0629, \\ -2.8937, 6.3974, 2.7900, -1.8615 \end{bmatrix} \\ (c_{m \text{ ECC-M}})_{m=0} &= \begin{bmatrix} 0.5134, 8.5538, 5.2101, 2.8465, -13.6543, -9.3080, \\ -8.5517, 6.3974, 5.6041, -1.7772 \end{bmatrix} \end{aligned} \quad (14)$$

whilst the results for BTS2 (LG0749) were obtained as

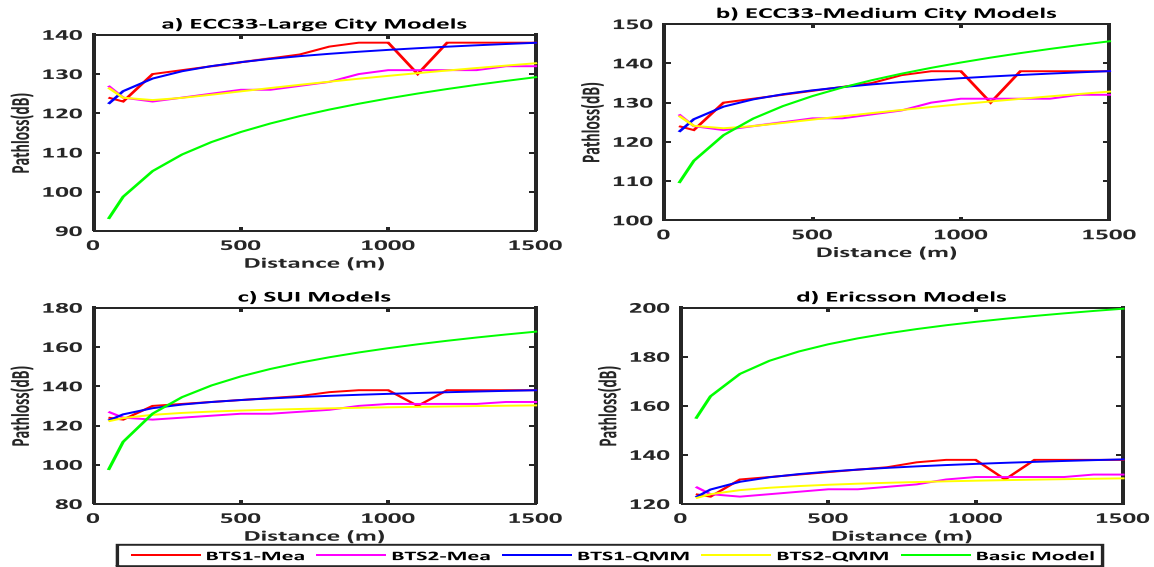


Figure 5. Profiles of pathloss predicted by QMM and basic models for LAG0747 (BTS1) and LAG0749(BTS2).

Table 3. RMSE and MPE metrics for QMM-calibrated models-Ikeja

Site	Ikeja BTS1 (900MHz)							
Model	Ericsson		SUI		ECC33-L		ECC3-M	
	Basic	QMM	Basic	QMM	Basic	QMM	Basic	QMM
MPE	-89.759	0.943	-55.004	-0.008	-24.076	-0.004	-42.190	0.120
RMSE	89.822	2.047	55.573	1.795	24.184	0.906	42.252	0.914
Site	Ikeja BTS2 (900MHz)							
Model	Ericsson		SUI		ECC33-L		ECC3-M	
	Basic	QMM	Basic	QMM	Basic	QMM	Basic	QMM
MPE	-72.226	-1.714	-37.471	-0.009	-6.542	-0.005	-24.656	0.141
RMSE	72.341	3.138	38.404	2.683	7.221	1.352	24.845	1.360

$$\begin{aligned} (c_{m\ SUI})_{m=0}^4 &= [-4.8465, 0.2189, 25.8193, 7.5108, 32.0665] \\ (c_{m\ Eric})_{m=0}^5 &= [-0.5600, 0.2371, -1.6802, -11.3452, -13.3023, 0.4172] \\ (c_{m\ ECC-L})_{m=0}^9 &= \begin{bmatrix} 2.4257, -3.3167, -0.6886, -1.4178, 8.4121, -8.6891, \\ -4.3964, 2.2573, 1.3430, -9.1756 \end{bmatrix} \\ (c_{m\ ECC-M})_{m=0}^9 &= \begin{bmatrix} 1.5770, -1.5299, 2.6286, 2.2780, 4.7714, -3.1994, \\ -0.8150, 2.2577, -2.9036, 5.3219 \end{bmatrix} \end{aligned} \quad (15)$$

Predicted pathloss profiles associated with these results are displayed in Figs (5) below.

RMSE and MPE performance metrics for the profiles of Fig. (5) are shown in Table 3.

From the metrics in Table 3, it is readily observed that in terms of both RMSE and MPE, the ECC-33 (large city) model is the best performing of the four models, for BTS1 in Ikeja. Although the SUI model for this BTS recorded a better MPE value than the ECC-33 (medium city) model, the latter recorded a significantly better RMSE value. These observations also generally apply for the statistical metrics recorded for the Ikeja BTS2.

### 3.3.2 Pathloss profiles for Lagos Island

Computational results obtained for the QMM-calibrated candidate models for the two BTS' selected for Lagos Island utilized the following operational parameters: frequency  $f = 1800\text{MHz}$ , transmitter antenna height = 45m; receiver antenna height = 1.5m. The corresponding solution to the QMM-calibration problem, when measurement data for Lagos Island BTS1 (LAG0788) is utilized is obtained as given in Eqn. (16):

$$\begin{aligned} (c_{m\ SUI})_{m=0}^4 &= [2.7819, 0.4797, 0.9613, -3.1244, -3.1148] \\ (c_{m\ Eric})_{m=0}^5 &= [4.8616, 0.7044, 5.4463, -2.5931, -1.4528, -0.3564] \\ (c_{m\ ECC-L})_{m=0}^9 &= \begin{bmatrix} 0.2008, 0.8747, -2.6695, 1.6195, 1.4123, 5.5568, \\ 5.2875, 3.5695, -3.1807, 0.5461 \end{bmatrix} \\ (c_{m\ ECC-M})_{m=0}^9 &= \begin{bmatrix} 0.3710, 0.5422, -0.7125, 1.7829, 2.0843, 5.5825, \\ 3.0640, 3.5692, -0.3007, 0.5295 \end{bmatrix} \end{aligned} \quad (16)$$

With the use of measurement data for Lagos Island BTS2 (LAG0789), the QMM-calibration process is characterized by

$$\begin{aligned} (c_{m\ SUI})_{m=0}^4 &= [3.1514, 0.4615, 1.6051, -2.9639, -4.9611] \\ (c_{m\ Eric})_{m=0}^5 &= [-0.7007, 0.7019, -1.8770, -1.9478, -0.5840, 0.4983] \\ (c_{m\ ECC-L})_{m=0}^9 &= \begin{bmatrix} 0.2259, 0.8206, -3.6633, 0.5862, 1.8856, 5.0566, \\ 8.7586, 5.1925, -8.6370, 3.1535 \end{bmatrix} \\ (c_{m\ ECC-M})_{m=0}^9 &= \begin{bmatrix} 0.5685, -0.7172, 0.5634, -0.2668, 5.0090, 4.8136, \\ 4.3768, 5.1917, 1.0706, 0.9875 \end{bmatrix} \end{aligned} \quad (17)$$

Pathloss profiles corresponding to the solutions prescribed by Eqns. (16) and (17) are displayed in Fig. (6) below.

The relative performances of the QMM-calibrated models are described by the statistical performance metrics displayed on Table 4.

Features of the results in Table (4) differ significantly from the corresponding results in Table (3) for the Ikeja BTS'. First, the best performing calibrated model is the ECC-33 medium city model, for which, as can be seen from the Table, MPE and RMSE values are 0.060dB and 2.067dB, respectively, in the case of BTS1. One remarkable feature of these metrics is that the calibrated Ericsson model (MPE =

0.061dB, RMSE = 2.061dB) not only performed better than its ECC-33 (large city) counterpart, but also virtually recorded the same metrics as the best performing ECC-33 (medium city) model.

However, for BTS2 of Lagos Island, whereas the calibrated Ericsson model recorded virtually the same MPE metric (-0.06dB) as the ECC-33 (medium city) model as best result, its RMSE metric (2.012dB) was significantly poorer than those of the ECC-33 models – 0.608dB (large city) and 0.595dB (medium city).

Contributions to net pathloss (as percentages of the net) by the component parameters of the calibrated models and their basic versions are described by the profiles of figure (7)-for the ECC-33 models, and figure (8), for SUI and Ericsson models both for BTS2 only.

Prior to QMM calibration, the contribution of the 'free space attenuation' component of the ECC-33 (medium city) basic component varied between 63% close to the transmitter, and about 66% at the far end. After calibration, this component's contributions to net predicted pathloss varied between 66% close to the transmitter and about 40.9%, 1km away from the transmitter. On the other hand, in the case of the ECC-33 (large city) model, was between 74.33% and 74.4% for the basic model, compared to the corresponding post-calibration values of -13.3% and 3.91%. The 'basic median pathloss' component recorded contributions of (10.47%, 5.76%) and (12.29%, 18.22%) for the basic medium and large city models, respectively. And the corresponding respective values of QMM calibrated versions of these models were recorded as (53.3%, 31.17%) and (923% and 61.83%). The profiles of Fig. (8) and the associated simulation results reveal that the predicted pathloss for the basic SUI model is dominated by the free space attenuation and pathloss exponent components. However, after calibration, the dominance of 'free space attenuation' component shot up from (64.88%, 44.89%) for the basic model, to (240%, 193%) for the QMM-model. Moderation for this component, in the QMM model, essentially came from MS height correction factor (-98%, -77%), and the shadowing correction component (-41%, -33%). Profiles in the lower half of Fig. (8) are for contributions to total predicted pathloss by the basic and QMM-calibrated Ericsson models. As can be readily observed from the profiles, in the basic model, contributions from the components 'g(f)', 'logd', and 'a0', in that order, dominate. On the other hand, in the QMM-calibrated version, the dominant contributions are from the; logsd', 'g(f)' and 'loghte' (in that order) components. As a matter of fact, contributions from the 'a0' component fell from (21.8%, 17.9%) in the basic component, to (-24.3%, -19.8%) in the QMM-calibrated model.

## 4. CONCLUDING REMARKS

This concludes the paper's analytical investigation of the characteristic features of basic radiowave propagation pathloss models, subjected to calibration in a Quasi-Moment-Method (QMM) algorithm. After a concise formulation of the QMM least-square-approximation problem and its solution in a matrix format similar to Harrington's method of moments, four basic models (Ericsson, SUI, and the ECC-33 large city and medium city) were selected as candidate basic models. The basic models were subjected to QMM-calibration, using field measurement data for Faysala and 'Industrial Zone', available from Mawjoud [7]; and for Ikeja and Lagos Island, available from Adelabu's Ph.D. thesis [10]. Performance evaluation, using computational results for Root Mean Square

Error (RMSE) and Mean Prediction Error (MPE) revealed that in general, the two calibrated ECC-33 performed better than the SUI and Ericsson models. Exceptions to this remark include MPE metrics for ‘Industrial Zone’, for which the calibrated SUI model performed best, as well as the Ikeja BTS’, for which the calibrated SUI model recorded better MPE metrics than the QMM calibrated ECC-33 medium city model. The calibrated Ericsson model’s performance for Lagos Island BTS’ compared favorably with those of the two

calibrated ECC-33 models.

The paper, as part of its investigations, also examined the effects of QMM calibration on contributions to net predicted pathloss, by component parameters of the basic and corresponding QMM-calibrated models. Simulation results revealed that the calibration process significantly modified these contributions, in a manner informed by the pathloss profiles of associated field measurements.

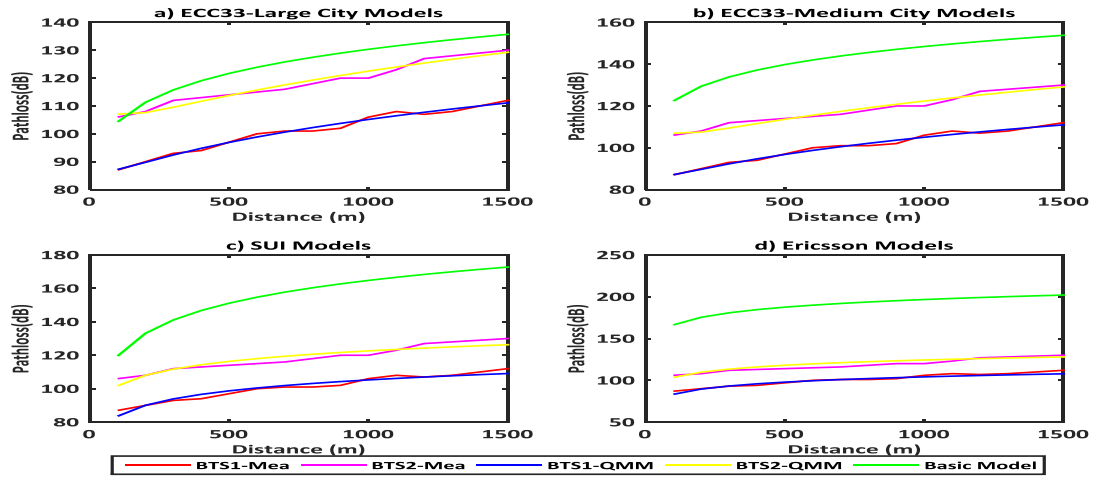


Figure 6. Profiles of pathloss predicted by QMM and basic models for LAG0788 (BTS1) and LAG0789(BTS2).

Table 4. RMSE and MPE metrics for QMM-calibrated Lagos-Island

Site	Lagos Island BTS1 (1800MHz)							
Model	Ericsson		SUI		ECC33-L		ECC3-M	
	Basic	QMM	Basic	QMM	Basic	QMM	Basic	QMM
MPE	-52.766	-0.061	-13.369	0.158	16.452	0.145	-0.012	0.060
RMSE	53.451	2.061	20.573	2.067	17.645	2.066	6.473	2.061
Site	Lagos Island BTS2 (1800MHz)							
Model	Ericsson		SUI		ECC33-L		ECC3-M	
	Basic	QMM	Basic	QMM	Basic	QMM	Basic	QMM
MPE	-58.328	-0.060	-18.931	0.153	10.858	0.135	-5.575	0.061
RMSE	59.272	2.012	25.902	2.017	13.617	0.608	9.936	0.595



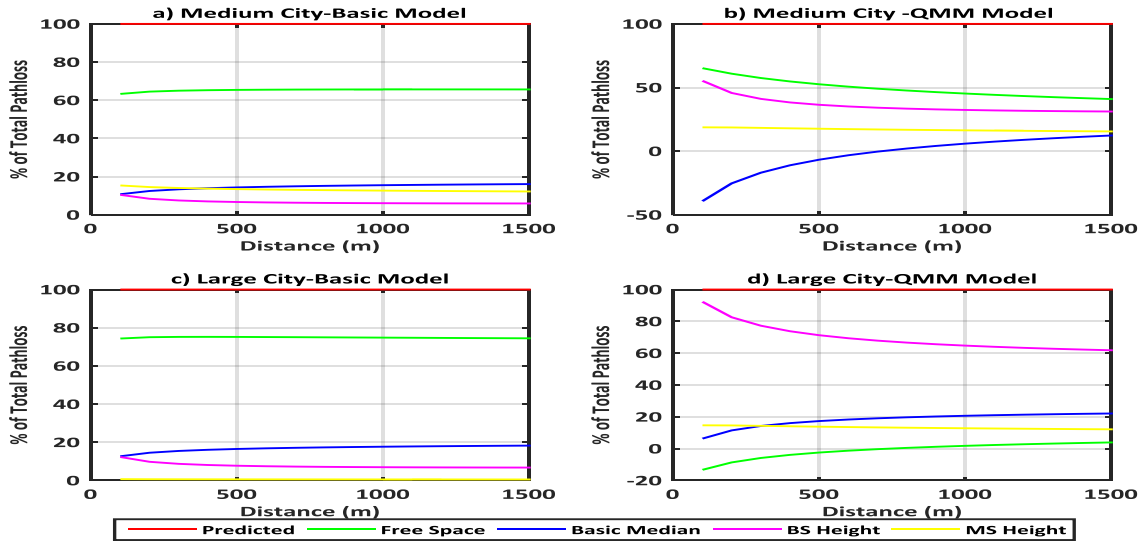


Figure 7 Percentage contributions to net predicted pathloss by calibrated and basic ECC-33 models for LAG0789 (BTS2).

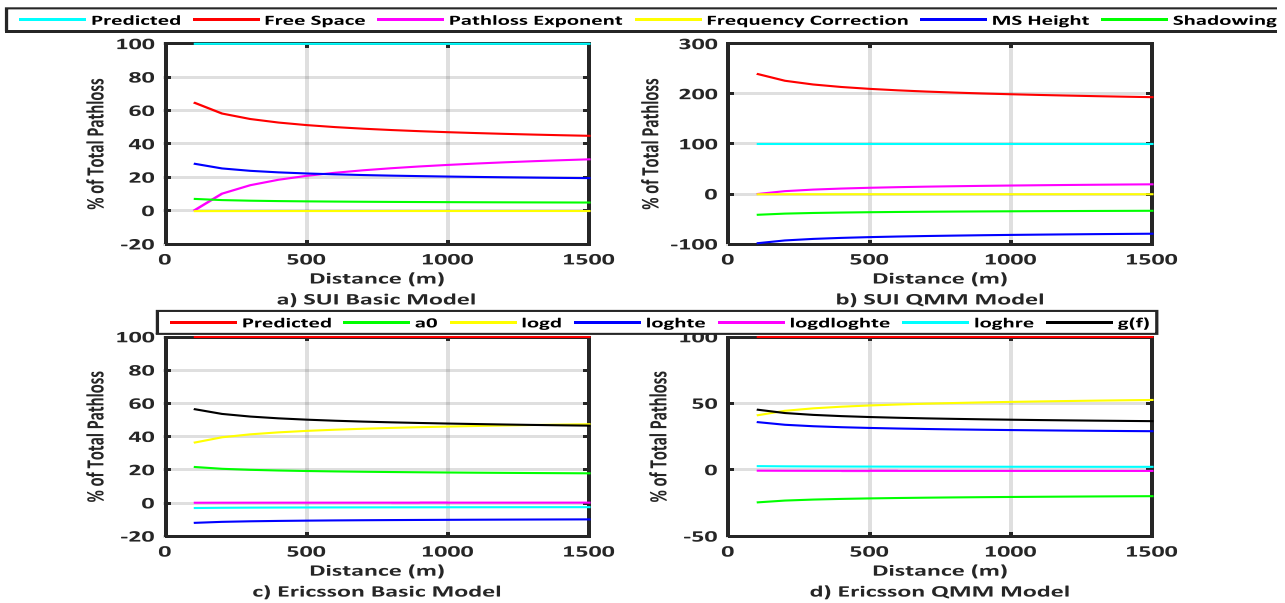


Figure 8 Percentage contributions to net predicted pathloss by calibrated and basic Ericsson and SUI models for LAG0789 (BTS2)

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