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Development of a Model for Estimation of Sunshine Hour Data for Different Regions of Uganda

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A B S T R A C T

The present study is concerned with the development, estimation and validation of sunshine hours models (SHM) in Uganda. The SHM is based on geographical (latitude) and climatological (clearness index) indices. The meteosat data (1984-2018) acquired from the National Aeronautics and Space Administration were used to compute the coefficients of the models which, yielded a coefficient of determination close to unity, signifying a good association between the sunshine hours (SH) and the associated indices. The models become distributed by introducing a longitudinal function of clearness index into the primary SHM developed. Moreover, the models were subjected to statistical validation using; mean absolute relative error (MARE), root mean square error (RMSE) and mean absolute percentage difference (APD). Consequently, the primary SHM showed strong agreement with the measured SH data in the three regions with the exception of the northern region with flawed on-station data. Also, validation of the models by; {MARE, RMSE, APD} for Eastern, Central and Western regions, yielding the following results; {0.0788,0.5441,7.8778},{0.0390,0.1453,3.9013} and {0.0124,0.0528,1.2436}, respectively. The following maximum SH; 11.16, 7.87, 9.52, 8.86 and 6.06 h were recorded for Non-regional, Northern, Eastern, Central and Western regions, respectively. Further, comparative validation with redeveloped global SHM showed that the present model stands in all the regions, whereas the global models validated only in the Eastern region. This is attributed to the synergy of geographical and climatological indices against the global models only based on climatological index. The model results show the order of regional SH distribution; eastern>northern>central>western region. These results could be employed in solar power, exploitation and agrometeorology development. This study further recommends for adoption of the present model to non-equatorial regions upon redevelopment as a meaningful extension of this work.

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1. INTRODUCTION

Sunshine Hour (SH) is an indicator of solar potential of a location and plays a central role in defining global solar radiation, clearness index, and relative sunshine hours for various applications in solar technologies [1-7]. Sunshine hour has diverse applications in agriculture, hydrology, and meteorology researches with encompassing applications, in atmospheric and building thermal balance [7], evaluation of agricultural resources [7-10], and in the development of global solar radiation models [11-13]. Meteorologically, the sunshine hour is measured using recorders (Campbell-Stokes), sensors (Kipp and Zonen CSD3), pyranometer, and pyrheliometer [7, 14, 15]. The accuracy of the efficiency of these sensors and recorders is influenced by the susceptibility to genera of clouds: cumulus, altostratus, cirrostratus, cirrus, and nimbus [16]. The reliability of sunshine hour recorder depends on a regular maintenance and calibration; if neglected for a long

time, it would culminate in the failure of the equipment as observed in Lira district of Uganda, which tracked a series of negative sunshine hour values for the district against the expected positive sunshine hours, when compared to the other regions in Uganda with positive values. The negative sunshine hour values observed could be attributed to the poor maintenance and calibration of the meteorological equipment in the region [5, 17, 18]. In order to avail a practical sunshine hour data for the region, the present work is geared towards developing a sunshine hour model based on both geographical and climatological indices using satellite data from [19] to generate direct sunshine hour models for the nation. However, the present work tends to convert the indirect sunshine hour models [5, 20-33] to suit the present application. Furthermore, the redeveloped sunshine model for the region will be compared with the present work through empirical validation [34-37] and statistical validation [38-41] of the sunshine hour models. For the purpose of distribution or estimation of the sunshine hours over the regions, the present work has delved into estimation of the developed sunshine hour models by substituting the clearness index component of the developed

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model with a longitudinal function, which complements the latitude in the primarily developed sunshine hour model to make it sensitive to the geographical coordinates for the regions [32, 42]. Therefore, the sunshine hour model can be used to generate sunshine hour distribution for all the locations in Uganda. The previous studies in Uganda [43, 44] show that eastern Uganda experiences the highest sunshine hours with large expanse of latitude and high longitude, followed by the northern region with small expanse of latitude and wide range of longitude. However, the central region has lower sunshine hours relative to Northern and Eastern Regions, whereas the Western Region has the lowest sunshine hour at extreme latitude and low longitude compared to the rest of the region. From all indications, the dominantly hilly and mountainous terrains have high cloud cover, whereas plain or less hilly and mountainous regions have less cloud cover, which impacts on the relative sunshine hours and sunshine hour values [40, 45-47].

Systematically, the proposed sunshine hour models which are based on climatological and geographical indicators have to be subjected to validation using on-station data (measured data) in order to ascertain the validity of the developed sunshine hour models by means of statistical and comparative validations of the models adapted from the other parts of the world. Besides the statistical validation, Root Mean Square Error (RMSE), Mean Bias Error (MBE), Mean Absolute Percentage Error (MAPE), Mean Absolute Error (MAE), Mean Percentage Error (MPE), Mean Absolute Relative Error (MARE), Absolute Bias Error (ABE), Absolute Mean Percentage Difference (APD) with insignificant deviations buttress the validity and application of developed sunshine hour models, whereas significant deviation suggests that the developed model is invalid and should not be implemented [48-51]. The present work will employ a more appropriate statistical tool (MARE, RMSE, and APD) and comparative validation to distinguish the present models. Therefore, the present work is poised to develop the sunshine hour models, to navigate the models in order to generate sunshine hour distribution in Uganda, and to validate the developed sunshine hour models. Consequently, these objectives will be realized through appropriate material and methods. In this regard, the navigable sunshine hour models and model validations are integrated. Subsequently, the presentation and discussion of results are given besides conclusions of findings and recommendation for further research.

2. METHODOLOGY

The sunshine hour models (combined and uncombined fifth order) were proceeded by the acquisition of a quadragenary satellite data from NASA POWER [19] for all the 122 districts in Uganda (2018) followed by the on-station data acquired from the four meteorological stations: Lira (Northern Region), Tororo (Eastern Region), Kampala (Central Region), and Mbarara (Western Region). The sunshine hour models were developed on latitude and clearness index, representing geographical and climatological indices, respectively. The acquired data were filtered and organized to suit the structure of the proposed sunshine hour models. Then, the sunshine hours data were regressed in OriginLab environment to generate the coefficients of the proposed models and the coefficient of determination (R²) of the developed model, which unveiled the degree of association between the model variables. Additionally, the developed models were modified by introduction of longitudinal function of clearness index to make it distributed in all the regions of Uganda. Also, the models were validated statistically using MARE, RMSE, and APD and were equally compared with global sunshine hour models, ascertaining their validity and effectiveness.

2.1. Formulation of sunshine hour models

SH model is proposed to be uncombined and combined variable regression (non-mechanistic) fifth-order model in Equation (1) in order to build a strong association between SH and the indicators for the different categories: Non-Regional (NRL), Northern Region (NR), Eastern Region (ER), Central Region (CR), and Western Region (WR).

$$\begin{split} SH_{NR,j} &= a_{i,j} + a_{i+1,j} \varphi_j + a_{i+2,j} k_{T,j} + a_{i+3,j} k_{T,j} \varphi_j + a_{i+4,j} \varphi_j^2 \\ &\quad + a_{i+5,j} k_{T,j}^2 + a_{i+6,j} k_{T,j} \varphi_j^2 + a_{i+7,j} k_{T,j}^2 \varphi_j \\ &\quad + a_{i+8,j} \varphi_j^3 + a_{i+9,j} k_{T,j}^3 + a_{i+10,j} k_{T,j} \varphi_j^3 \\ &\quad + a_{i+11,j} k_{T,j}^3 \varphi_j + a_{i+12,j} \varphi_j^4 + a_{i+13,j} k_{T,j}^4 \\ &\quad + a_{i+14,j} k_{T,j} \varphi_j^4 + a_{i+15,j} k_{T,j}^4 \varphi_j + a_{i+16,j} \varphi_j^5 \\ &\quad + a_{i+17,j} k_{T,j}^5 \quad (\text{hour}) \quad \exists \quad i = 0; \quad j \\ &\quad = \{1,2,3,4,5\} \equiv \{\text{NRL}, \text{NR}, \text{ER}, \text{CR}, \text{WR}\} \end{split}$$

where a_i , i=0, 1, 2, ..., 17 are the coefficients of Equation (1), φ_j (°) is the latitude, and $k_{T,j}$ (–) is the clearness index.

For the purpose of representing the SH on geographical coordinates, the present work has further proposed developing clearness index (k_T) as a function of latitude and longitude in Equation (2).

$$\begin{split} k_{T,j} &= b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j + b_{l+3,j} \varphi_j \lambda_j & \exists \ i = 0; \\ &= \{1, 2, 3, 4, 5\} \equiv \{\text{NRL, NR, ER, CR, WR}\} \end{split} \tag{2}$$

where b_1 , l = 0, 1, 2, 3 are the coefficients of Equation (2). Substituting Equation (2) into (1) makes SH dependent on latitude and longitude in Equation (3).

$$\begin{split} \mathrm{SH}_{NR,j} &= a_{i,j} + a_{i+1,j} \varphi_j \\ &\quad + a_{i+2,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j + b_{l+3,j} \varphi_j \lambda_j \Big) \\ &\quad + a_{i+3,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j \\ &\quad + b_{l+3,j} \varphi_j \lambda_j \Big) \varphi_j + a_{i+4,j} \varphi_j^2 \\ &\quad + a_{i+5,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j \\ &\quad + b_{l+3,j} \varphi_j \lambda_j \Big)^2 \\ &\quad + a_{i+6,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j \\ &\quad + b_{l+3,j} \varphi_j \lambda_j \Big) \varphi_j^2 \\ &\quad + a_{i+7,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j \\ &\quad + b_{l+3,j} \varphi_j \lambda_j \Big)^3 \varphi_j + a_{i+8,j} \varphi_j^3 \\ &\quad + a_{i+9,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j \\ &\quad + b_{l+3,j} \varphi_j \lambda_j \Big) \varphi_j^3 \\ &\quad + a_{i+10,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j \\ &\quad + b_{l+3,j} \varphi_j \lambda_j \Big) \varphi_j^3 \\ &\quad + a_{i+11,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j \\ &\quad + b_{l+3,j} \varphi_j \lambda_j \Big)^4 \\ &\quad + a_{i+14,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j \\ &\quad + b_{l+3,j} \varphi_j \lambda_j \Big) \varphi_j^4 \\ &\quad + a_{i+15,j} \Big(b_{l,j} + b_{l+1,j} \varphi_j + b_{l+2,j} \lambda_j \\ &\quad + b_{l+3,j} \varphi_j \lambda_j \Big)^5 \Big(hr \Big) \\ \exists \qquad i = 0; \ l = 0; \ j = \{1,2,3,4,5\} \equiv \{ NRL, NR, ER, CR, WR \} \end{split}$$

Equation (3) for categories {NRL, NR, ER, CR, WR} is subjected to the following constraints in Table 1. Globally, used

clearness index (K_T) according to the following researchers [52-56] is algebraically given in Equation (4).

$$K_T = a(RSH)^2 + b(RSH) + c$$
 (4)

Sequel to the redevelopment of Equation (4) with the local data to obtain coefficients a, b, c and further rearrangement of Equation (4) yielded the corresponding RSH model in Equation (5).

$$\begin{aligned} \text{RSH} &= -0.171875 + (2.232143 \text{K}_{\text{T}} \\ &- 0.61331613)^{0.5} \quad \exists \quad \text{K}_{\text{T}} \geq 0.275 \end{aligned} \tag{5}$$

Expressing RSH in its components, i.e., sunshine hours and daylength, Equation (5) is expounded in Equation (6) as follows:

$$\begin{split} \text{RSH} &= \frac{n}{N} = -0.171875 \\ &\quad + (2.232143 \text{K}_{\text{T}} \\ &\quad - 0.61331613)^{0.5} \quad \exists \quad \text{K}_{\text{T}} \geq 0.275 \end{split} \label{eq:RSH}$$

Equation (7) gives an explicit expression for sunshine hours for comparative validation.

$$\Rightarrow SH = n = N(-0.171875 + (2.232143K_T - 0.61331613)^{0.5}) \quad \exists K_T \ge 0.275$$
 (7)

The daylength, N is defined in Equation (8) as in [57].

$$N = \frac{2}{15}\cos^{-1}(-\tan\delta\tan\phi); \ \delta = 23.45\sin\left[360\left(\frac{284+n}{365}\right)\right] \eqno(8)$$

where n is the number of days from January first and δ is the declination angle.

Statistically, Equations (1) to (3) are to be validated with RMSE, MARE, and APD. The RMSE is defined in Equation (4) as follows:

$$\begin{split} \text{RMSE}_{j} &= \left[\frac{1}{n} \sum_{i=1}^{n} \left(\text{RSH}_{m,j,i} - \text{RSH}_{s,j,i} \right)^{2} \right]^{0.5} \\ \exists \quad j &= \{1,2,3,4,5\} \equiv \{ \text{NRL}, \text{NR}, \text{ER}, \text{CR}, \text{WR} \} \end{split}$$

Table 1 shows the boundaries of the sunshine hour models developed.

	_	able 10 Boundary conditions (constraints) in Eq.	(aution (b)
Category	Operation	Latitude (degree)	

Category	Operation	Latitude (degree)	Longitude (degree)		
j	j, j	Lautude (degree)	Longitude (degree)		
NRL	NRL, NRL	$-0.750000 \le \phi_{NRL} \le 3.649997$	$29.683331 \le \lambda_{NRL} \le 34.94999722$		
NR	NR, NR	$1.633330556 \le \varphi_{NR} \le 3.64999$	$30.899997 \le \lambda_{NR} \le 34.9499972$		
ER	ER, ER	$0.283333 \le \phi_{ER} \le 2.03333$	$33.050000 \le \lambda_{ER} \le 34.73333$		
CR	CR, CR	$0.066667 \le \phi_{CR} \le 1.333333$	$29.783331 \le \lambda_{CR} \le 33.266667$		
WR	WR, WR	$-0.750000 \le \varphi_{WR} \le 2.183333$	$29.683331 \le \lambda_{WR} \le 32.00000$		

Table 1 Roundary conditions (constraints) in Equation (3)

The MARE is expressed in Equation (5) as follows:

$$MARE_{j} = \frac{1}{n} \sum_{i=1}^{n} \left| \frac{\left(SH_{m,j,i} - SH_{s,j,i}\right)}{SH_{m,j,i}} \right|$$
 (5)

$$\exists$$
 $j = \{1, 2, 3, 4, 5\} \equiv \{NRL, NR, ER, CR, WR\}$

The mean Absolute Percentage Difference (APD) is given in Equation (6) as follows:

$$APD_{j} = \frac{1}{n} \sum_{i=1}^{n} \left| \frac{\left(SH_{m,j,i} - SH_{s,j,i}\right)}{SH_{m,j,i}} \right| \times 100\%$$
(6)

$$\exists$$
 $j = \{1, 2, 3, 4, 5\} \equiv \{NRL, NR, ER, CR, WR\}$

3. RESULTS AND DISCUSSION

This section presents the results and discussion of the study in Subsections 4.1 and 4.2, respectively.

3.1. Presentation of results

Firstly, the regression coefficients generated for the fifth-order model in Equation 1 which is both uncombined (K_T, ϕ) and combined $(K_T \phi)$ variable sunshine hours model is represented in Table 2 for NRL, NR, ER, CR, and WR. Similarly, Table 3 gives the clearness index coefficients of Equation 2 (K_T = $f(\phi, \lambda)$ for non-regional and the different regions of Uganda. Furthermore, Table 3 provides the goodness of fit or coefficient of determination (R²) which shows the level of association between the variables (latitude and longitude) and clearness index. The R2 values reported for the mentioned categories tend to unity, showing a close association between the variables. All the categories have quadratic coefficients except the northern region, which shows a good association with a linear model. Lastly, Table 4 gives the results of statistical tools (RMSE, MARE, and APD) employed in the validation of the sunshine hour models in Uganda.

Table 2. Coefficients of Equation (1) for sunshine hours

Category	a_0	a ₁	a ₂	a_3	a ₄	a ₅	a ₆	a ₇	a ₈	a ₉
Non-regional	-238	167	1530	-1070	2.05	-3620	-3.87	2560	-0.758	3830
Northern	23.443	83.063	-0.186	-626.653	-4.178	-472.751	10.031	1807.904	0.939	1267.508
Eastern	-126.793	3618.276	-3.44584	-17558.84	74.3881	1187.509	-125.277	28128.73	-19.5588	-1348.8
Central	-7.92	206	1.27	-1530	14.9	236	-27.9	4190	-16.7	-745
Western	-0.101	2.894	0.151	-13.957	-0.048	-1.687	0.089	21.582	0.047	86.417
					•					
Category	a ₁₀	a ₁₁	a ₁₂	a ₁₃	a ₁₄	a ₁₅	a ₁₆	a ₁₇	R	2
Non-regional	1.53	-2700	0.0778	-1490	-0.199	1050	0.00395	0.11	0.99	926

Northern	-2.736	-2341.6	-0.035	-997.856	0.277	1122.785	-0.010	143.798	0.9985
Eastern	32.953	-14854	_	_	_	_	_	_	0.9985
Central	30.8	-5080	5.85	992	-10.4	2310	-0.132	-484	0.9993
Western	-0.090	-10.465	0.0006	-80.937	_	_	_	_	0.9999

Table 3. Clearness index coefficients of Equation (2)

Category	b ₀	b ₁	$\mathbf{b_2}$	R ²
Non-regional	0.1752	0.0051	0.0002	0.9683
Northern	0.2054	0.0116	_	0.8638
Eastern	- 17.036	1.0165	- 0.0146	0.8765
Central	10.366	-0.6447	0.0106	0.9411
Western	11.135	-0.7101	0.0119	0.9507

Table 4. Statistical validation of sunshine hour in Uganda

Region	MARE	RMSE	APD (%)
Northern	_	-	-
Eastern	0.0788	0.5441	7.8778
Central	0.0390	0.1453	3.9013
Western	0.0124	0.0528	1.2436

The figures of results are presented from Figures 1–11. Figure 1 presents the non-validated sunshine hours for the northern region. This non-validation is ascribed to the poor quality of maintenance of the instrument and possibly due to accumulation of dust and vapor on the sensitive spotting surface of the instrument [7, 14, 16]. This is usually taken care of by recalibration of the instruments and regular maintenance. More so, Figures 2–4 illustrate the validation of measured and simulated sunshine hours for ER, CR, and WR, respectively. Additionally, Figures 5–9 present the estimation of sunshine hours based on the latitude (ϕ) and longitude (λ) of the locations: NRL and regional basis for NR, ER, CR, and WR, respectively. Moreover, Figures 10 and 11 give the comparative validation of the global and present sunshine hour models for eastern and central regions, respectively.

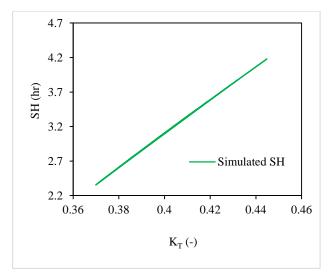


Figure 1. Non-validated sunshine hours for northern region

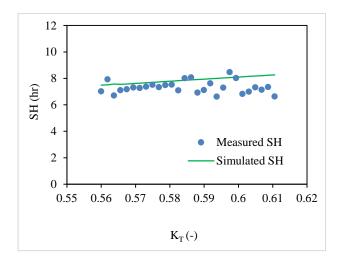


Figure 2. Validation of sunshine hours for eastern region

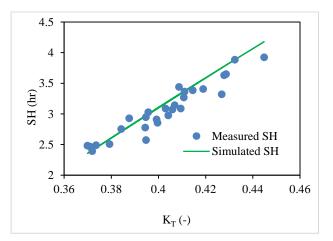


Figure 3. Validation of sunshine hours for central region

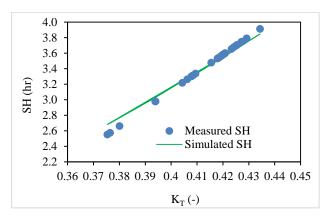


Figure 4. Validation of sunshine hours for western region

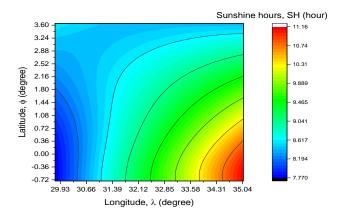


Figure 5. Non-regional estimation of sunshine hours in Uganda

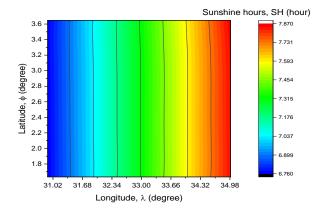


Figure 6. Regional estimation of sunshine hours for northern region

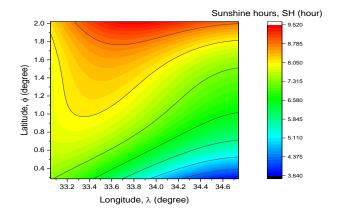


Figure 7. Regional estimation of sunshine hours for eastern region

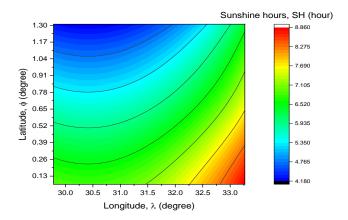


Figure 8. Regional estimation of sunshine hours for central region

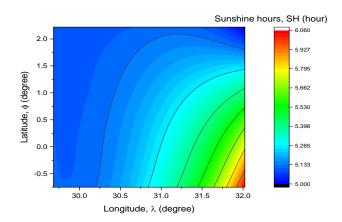


Figure 9. Regional estimation of sunshine hours for western region

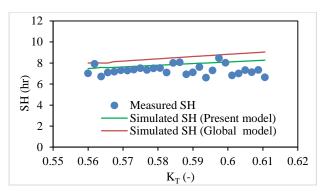


Figure 10. Comparative validation of global and present sunshine hours models for eastern region

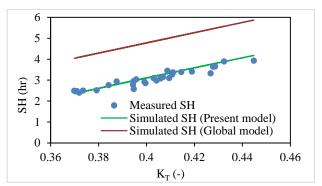


Figure 11. Comparative validation of global and present sunshine hours models for central region

3.2. Discussion

Approximately, Table 2 presents the coefficients of Equation (1) and the coefficients of determination (R^2) , which are approaching unity (1). The R^2 values support that there is a strong association between the dependent variable (sunshine hours) and the independent variables (the latitude and clearness index), representing the geographic and climatological indices for the sunshine hour models.

The sunshine hour models in Table 2 are bound in Table 1, which displays the latitude and longitude of the entire Uganda and the regions. Thus, the sunshine hour models developed define the sunshine hours with certainty in as much as the constraints delineated in Table 1 are not traversed; otherwise, the model may fail if not redeveloped with data outside the boundaries.

Considering that Equation (1) cannot fulfill the main objective of this work, the clearness index is formulated as a function of longitude to engender the estimation of sunshine hour models in Equation (1) according to the following literature [32, 42].

Pertinently, Table 4 presents the statistical validation (MARE, RMSE, and APD) of sunshine hours in Uganda with minimal deviations, which is in strong agreement with the other published results [38, 40, 41]. The validation results are displayed in Figures 1–4 to substantiate the agreement between the present validations and those in the aforementioned literature.

The distribution of sunshine hours in Uganda and the corresponding regions is displayed in Figures 5-9, respectively. Generally, sunshine hours are abundant at latitudes above 3.0 and longitude below 3.15. Thus, these locations should be earmarked for installation of solar facilities in Uganda indisputably. Figure 6 portrays that the sunshine hours expand all through the longitude of the northern region with a strict restriction on the latitude (3.4-3.7). Therefore, government and developers should apply these findings without bias in the bid to exploit solar resources for maximum in this region. With the target latitude (0.7-1.8°N) and limited longitude (34.6–35.0°E), the sunshine hour is highly concentrated in the eastern region indiscriminately (as shown in Figure 7). Credibly, Tororo plant (32 000 panels; 10 MW; [58]) is within the mapped out area, whereas Soroti plant (32 680 panels; 10 MW; [58]) is slightly off the hot spot. Thus, the hot spots for enormous exploitation of solar resources is mapped out by these geographical locations. The sunshine hour is sparsely distributed in the central region from latitudes (1.2-1.4°N) and longitude (30-30.7°E); thus, government and developers should only place the solar facilities within the mapped out location; otherwise, this present work suggests that the solar facilities to be installed in locations with higher sunshine hours and power generated from such a location to be transmitted to region with low sunshine hours in order to maximize the utility of solar installations in Uganda. Notably, the largest solar installation in Uganda is located in the Gomba district in the Kabulasoke sub-county with a capacity of 68 000 solar panels (20 MW; [58]) located in the mapped out region by the study in Figure 8. Had it been that the installation at Kabulasoke is installed in northern and eastern regions, obviously, more power would have been added to the national grids. Hence, the present work has unveiled the best spots for future localization of solar facilities in order to boost solar power harvests in Uganda.

In terms of maximum sunshine hour distribution, western region (6 hours) is less favored like other regions: northern (8.0 hours), eastern (9.5 hours), and central (9.0 hours). Hence, government and developers should concentrate on solar facilities in the order of increasing sunshine hours: eastern, northern, central, and western regions for maximizing solar power exploitation. Thus, the present work is recommended for generation and transmission of solar power from the other regions (northern, eastern, and central) to the western region. However, the western region with a rifted valley is compensated by geothermal resources, which are not available in other regions. Thus, this work is encouraging for exploration and exploitation of geothermic energy to support the solar power in the sustaining power supply to the grids.

Lastly, Figure 10 shows that the redeveloped global sunshine hour models were in agreement with the present model in the eastern region, whereas it failed in the other regions due to lack of geographical indicators in the model [32, 42]. Representatively, Figure 11 portrays the aforementioned failure in the preceding paragraph. However, the present model comprehensively fitted to the observed data due to the binary indicators (latitude and clearness index). Thus, a robust sunshine hour model should be built or formulated with heterogeneous indicators by incorporating extraterrestrial and terrestrial indices.

4. CONCLUSIONS

This study successfully developed and estimated the sunshine hour models for all the regions of Uganda. Fundamentally, the developed sunshine hour models were uncombined and combined variables fifth fifth-order non-differential models, which were dependent on the clearness index and latitude. These models were distributed by substituting the clearness index with its longitudinal function into the sunshine hour lumped models. Consistently, there was a strong association between the dependent (sunshine hours) and independent variables (clearness index and latitude), as shown by the high values of coefficient of determination ($\mathbb{R}^2 \to 1$).

Furthermore, the developed sunshine hour models were validated statistically using MARE, RMSE, and APD for eastern, central, and western regions, yielding the following results {0.0788, 0.5441, 7.8778}, {0.0390, 0.1453, 3.9013}, and {0.0124, 0.0528, 1.2436}, respectively. The following maximum sunshine hours, i.e., 11.16, 7.87, 9.52, 8.86, and 6.06 h, were recorded for non-regional, northern, eastern, central, and western regions, respectively. Comparatively, the sunshine hour models developed were in strong agreement with the measurement data in all the regions relative to the adapted and redeveloped global models which were in agreement with the eastern measured data and failed in the rest of the regions. This shows the effect of combining the geographical index (latitude) with the climatological index (clearness index) in developing robust sunshine hour models by the present study.

Moreover, the distributed sunshine hour models showed the descending order of sunshine hours in Uganda with eastern > northern > central > western region. Thus, this study suggests that localization and installation of solar power plants should be concentrated in the hotspots in the regions. In addition, the sunshine hour model could be employed in agrometeorology for the biological functioning of plants and in boosting agroproduce.

This study strongly recommends the adaption and redevelopment of the present models outside the equatorial regions as a further study.

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Sunshine hours (hr)

NOMENCLATURE

SH

K_{T}	Clearness index (-)
RSH	Relative sunshine hours (-)
λ	Latitude (°)
ф	Latitude (°)
δ	Solar declination angle (°)
SHM	Sunshine hour models
MARE	Mean absolute relative error
RMSE	Root mean square error
APD	Mean absolute percentage di
\mathbf{P}^2	Coefficient of determination

fference Coefficient of determination

N Davlength NRL Non-regional Northern region NR ER Eastern region Central region CR WR Western region

 $a_i, i = 0, 1, 2, \cdots, 17$ Regression coefficients $b_l, l = 0, 1, 2, 3$ Regression coefficients

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