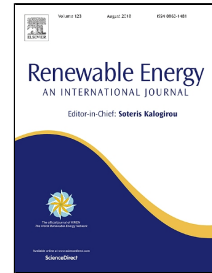


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E. Kaplani, S. Kaplanis, S. Mondal



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# 1 A spatiotemporal universal model for the prediction of the global solar 2 radiation based on Fourier series and the site altitude

3 E. Kaplani<sup>a,\*</sup>, S. Kaplanis<sup>b</sup>, S. Mondal<sup>b</sup>

4 <sup>a</sup>Engineering Division, Faculty of Science, University of East Anglia, Norwich, NR4 7TJ,  
5 UK

6 <sup>b</sup>Renewable Energy Systems Lab, Technological Educational Institute of Western Greece,  
7 Meg. Alexandrou 1, Patra 26334, Greece

8 \*Corresponding author. Email address: e.kaplani@uea.ac.uk  
9

## 10 Abstract

11 This paper presents the development, testing and validation of a novel generic type universal  
12 model consisting of a set of sine and cosine harmonics in the temporal and spatial domain  
13 suitably parameterized for the prediction of the mean expected global solar radiation  $H(n,\varphi)$   
14 on the horizontal for a day,  $n$ , at any latitude  $\varphi$ . Its prediction power is further enhanced with  
15 the introduction of a correction term for the site altitude taking into account the  $\varphi$  dependent  
16 atmospheric height. Solar radiation data from 53 stations around the earth were obtained from  
17 GEBA database to train the model.  $H(n,\varphi)$  is expressed by a Fourier series of compact form  
18 with the zero frequency component dependent on  $\varphi$  providing the main spatial dependence  
19 and two  $n$  dependent harmonics in the form of cosine functions giving the time dependence.  
20 The  $\varphi$  dependent model parameters follow symmetry rules and are expressed by Fourier  
21 series up to the 3<sup>rd</sup> order harmonic. The 3D spatiotemporal profile of the model is in  
22 agreement to the extraterrestrial one. The model was validated using GEBA data from  
23 additional 28 sites and compared with NASA, PVGIS and SoDa data, showing the  
24 robustness, reliability and prediction accuracy of the proposed model.

25

26 **Keywords:** solar radiation prediction; universal model; Fourier series; site altitude;  
27 atmospheric height

28

## 29 1. Introduction

30 For the sizing of Renewable Energy Systems (RES) configurations it is necessary to provide  
31 as input the values of the daily global solar radiation  $H(n,\varphi)$  kWh/m<sup>2</sup>/day on the horizontal in  
32 any place with latitude  $\varphi$ , for any day  $n$ , while in more detailed dynamic simulation models  
33 the values of the intensity of the global solar radiation,  $I(h;n)$ , at a site in any hour  $h$  of a day  
34  $n$ , are required, [1-4]. Solar radiation is monitored in many stations around the world and data  
35 are processed and stored in international databases as in [5-8]. A large number of research  
36 studies outline models which provide for  $H(n)$  and/or  $I(h;n)$  estimates for various sites. Those  
37 models are categorized as semi-empirical, ASHRAE [9] and Iqbal [10] models, providing  
38 elaborated expressions based on theoretical approaches with regard to the solar light optics  
39 such as transmission, reflection and scattering, as well as the atmospheric pressure versus  
40 altitude and the ambient temperature for the site and the time period concerned. Both models  
41 predict the beam, incident and diffuse components of the global solar radiation in a site

42 enabling the estimation of the mean expected daily global solar radiation based on  
43 expressions as below.

$$44 \quad I_n = C_n A_{\text{ext}} e^{-B \left( \frac{P}{P_o} \right) \sec(\theta_z)} \quad (1)$$

$$45 \quad I_n = 0.9751 E_o I_{sc} \tau_r \tau_o \tau_g \tau_w \tau_a \quad (2)$$

46 where  $A_{\text{ext}}$  ( $\text{W}/\text{m}^2$ ) is the apparent extraterrestrial irradiance given in tables [9].  $I_n$  is the direct  
47 normal irradiance ( $\text{W}/\text{m}^2$ ),  $C_n$  is the ratio of the direct normal irradiance calculated with the  
48 local mean clear-day water-vapour over the direct normal irradiance calculated with water-  
49 vapour according to the basic atmosphere.  $P$  (mbar) is the actual local-air pressure and  $P_o$  is  
50 the standard pressure (1013.25 mbar). In eq.(2),  $I_{sc}$  is the solar constant taken as  $1367 \text{ W}/\text{m}^2$ .  
51  $E_o$  (dimensionless) is the eccentricity correction-factor of the Earth's orbit. Finally,  $\tau_r, \tau_o, \tau_g, \tau_w$   
52 and  $\tau_a$  are the Rayleigh, ozone, gas, water and aerosols scattering transmittances,  
53 (dimensionless), respectively.

54 A second group comprises of empirical models which provide the daily global solar radiation  
55 based on the Angström-PreScott model [11-13] and use various regression based expressions  
56 outlined in [14,15]. The parametric values of those expressions are generally valid for the  
57 geographical sites or regions they have been determined for. Values for these parameters  
58 applicable at any site have been proposed in [12]. The variable in these models is the ratio of  
59 the actual sunshine hours,  $S$ , over the maximum possible sunshine hours,  $S_o$ , in a day  $n$  in the  
60 site of concern. A third group of empirical solar radiation models correlate further  $H(n)$  with  
61 the  $T_{\text{min}}$  and  $T_{\text{max}}$  air temperature, the relative humidity,  $RH$ , and other meteorological  
62 parameters, such as the cloud coefficient,  $C$  and the precipitation,  $R$ , [16-20], as in the general  
63 form of eq.(3) with one or more of the above quantities included.

$$64 \quad \frac{H(n)}{H_{\text{ext}}(n)} = f \left( \frac{S}{S_o}, (T_{\text{max}} - T_{\text{min}}), RH, R, C \right) \quad (3)$$

65 where  $H_{\text{ext}}(n)$  is the daily solar radiation at a site at the top of the earth's atmosphere.

66 More elaborated models proposed and applied in several projects are the ones in a fourth  
67 group which use artificial neural networks (ANN) to provide for  $H(n)$ , in solar energy  
68 systems [21-23]. Finally, there is a group of empirical models which determine  $H(n)$  in a site  
69 with parameter the number of the day,  $n$ , in the year [24-27]. A sub-group uses simple or  
70 more complex sine or cosine expressions of Fourier series [28-33],

$$71 \quad H(n) = A + B \cos \left( \frac{2\pi}{365} n + C \right) \quad (4)$$

$$72 \quad H(n) = a + b \cos(z) + c \sin(z) + d \cos(2z) + e \sin(2z) \quad (5)$$

73 where,  $z = (2\pi/365) \cdot n$ , and the parameters  $A, B, C$ , and  $a, b, c, d, e$  depend on the site and are  
74 determined by regression analysis. The predictive performance of those models was shown to  
75 be reliable for the region of study. However, the above parameters were determined for the  
76 specific region and are not universally applicable. The mean expected hourly global solar  
77 radiation on the horizontal may then be determined by the models outlined in [26, 34-36]  
78 using the  $H(n)$  predicted above as input to satisfy boundary conditions, such as the model  
79 analyzed in [28] using the expression,

$$I_{m,exp}(h;n) = A + B \exp(-\mu(n)(x(h) - x(12))) \cos\left(\frac{2\pi h}{24}\right) \quad (6)$$

81 where  $\mu(n)$  is the solar radiation attenuation coefficient through the atmosphere determined  
 82 using the predicted  $H(n)$  based on eq.(4),  $x(h)$  is the solar radiation path in the atmosphere  
 83 dependent on the hour  $h$  in a day  $n$  at a site and  $x(12)$  is the corresponding path for the solar  
 84 noon.

85 Since the parametric expressions of the above models derived through regression analysis  
 86 were valid only for the regions of concern, a model of universal validity was proposed and  
 87 tested using 2 cosine functions to predict the global solar radiation in a day at any site [37].  
 88 That model was shown to give good  $H(n)$  predictions for sites both in the N. and S.  
 89 Hemispheres. However, the parameters of that model as determined do not guarantee that the  
 90  $H(n)$  function is continuous when  $n$  changes from 365 to 1, i.e. from the end of December to  
 91 beginning of January the following year.

92 This paper proposes a reliable and self-consistent generic model of universal applicability  
 93 composed of a complete set of spatiotemporal terms based on Fourier series satisfying the  
 94 above requirement. The parameters of the model display symmetries with regard to the N.  
 95 and S. Hemisphere. Additionally, the proposed model includes a correction for the site  
 96 altitude and the atmospheric height appropriately parameterized. The model is outlined in  
 97 Sections 2 and 3 and provides directly the mean expected daily global solar radiation at  
 98 horizontal  $H(n,\varphi)$  at any site with altitude  $h_s$ . The validation of the model is presented and  
 99 discussed in Sections 4 and 5 where results are given in comparison with the measured data  
 100 from GEBA and other databases.

101

## 102 2. The Generic Universal Model

103 The proposed generic model predicts  $H(n,\varphi)$  for any day  $n$  and site with latitude  $\varphi$ , and takes  
 104 also into consideration the site altitude and the atmospheric height. A double harmonic  
 105 analysis was applied to solar radiation data obtained from GEBA database from a grid of 53  
 106 stations around the earth with altitude less than 500m. This process resulted in a set of  
 107 harmonic spatiotemporal terms whose coefficients are functions of the site latitude.

108 The model proposed to predict  $H(n,\varphi)$  is expressed through a Fourier series of compact form  
 109 and is presented in eq.(7).

$$H(n,\varphi) = A(\varphi) + B_1(\varphi) \cos\left(l_1(\varphi) \frac{2\pi}{365} n + C_1(\varphi)\right) + B_2(\varphi) \cos\left(l_2(\varphi) \frac{2\pi}{365} n + C_2(\varphi)\right) \quad (7)$$

112 The key requirements and conditions set are:

113 a. the cyclicity and continuity in the behaviour of the  $H(n,\varphi)$  profiles and especially their rate  
 114 of change at the end of December and beginning of January to take the same value

115 b. applicability to both N. and S. Hemispheres

116 c. the model's coefficients corresponding to the spatial domain  $[-\pi/2, \pi/2]$  to be expressed  
 117 with the same order Fourier series

118 d. the model's coefficients to follow symmetry rules with respect to N. and S. Hemisphere

119 e. the altitude,  $h_s$ , of the site and the variable atmospheric height to be taken into account in  
 120 the determination of  $H(n,\varphi)$

121 The sites chosen were distributed in both S. and N. Hemisphere from East to West as shown  
 122 in Fig.1. Time series of monthly average global solar irradiance were obtained from GEBA  
 123 database [5]. The monthly average daily global solar radiation was estimated and averaged  
 124 over the years data were recorded for each site (in most cases these were more than 10 and in  
 125 some cases more than 50 years). The estimated monthly averages of daily  $H(n)$  values were  
 126 deployed along 2 consecutive years so that the model coefficients satisfy the requirement (a)  
 127 above. The monthly averages were mapped to the representative day of each month.

128 The proposed model based on the compact Fourier series of eq.(7) was fitted on the estimated  
 129 monthly averages of the daily  $H(n)$  values for each of the above sites. A nonlinear regression  
 130 analysis was applied based on the proposed compact Fourier series model using the nonlinear  
 131 least squares method. The fundamental frequency is  $2\pi/365$ .

132 The model coefficients  $A$ ,  $B_1$ ,  $B_2$ , and the associated parameters  $C_1, C_2$ ,  $l_1$ ,  $l_2$  functions of  $\varphi$ ,  
 133 were derived by nonlinear regression analysis for each one of the 53 sites, satisfying the  
 134 requirements described above. The coefficient of determination  $R^2$ , for any latitude and  
 135 longitude were between 0.97-0.99 for 96% of the sites, while the NRMSE values were  
 136 between 0.09-0.34 for all sites.

137 The frequency parameters  $l_1$ ,  $l_2$  are  $\varphi$  dependent take integer values [1, 2] corresponding to  
 138 the 1st and 2nd harmonic. To secure symmetry, parameters  $C_1$ ,  $C_2$  were normalised based on  
 139 the well known cosine function properties:

$$140 \cos(x + (C + 2\lambda\pi)) = \cos(x + C), \quad \cos(x + C) = -\cos(x + (C + \pi)) \quad (8)$$

141 where,  $\lambda$  is an integer.

142 In a second stage a Fourier analysis was performed on the model coefficients  $A$ ,  $B_1$ ,  $B_2$ , and  
 143 their associated parameters  $C_1, C_2$  as described in Section 3. This resulted in a self-consistent  
 144 prediction model for  $H(n,\varphi)$ .

145



146

147 Fig.1. The 53 sites (drop-shaped) used to train the model and the 28 sites (circles) used for  
148 the model validation as illustrated on a Google map.

149

### 150 3. The model coefficients as functions of $\varphi$ analysed in Fourier series

151 The Fourier analysis of the coefficients  $A(\varphi)$ ,  $B_1(\varphi)$ ,  $B_2(\varphi)$  and the parameters  $C_1(\varphi)$ ,  $C_2(\varphi)$ ,  
152 determined by nonlinear regression analysis, showed that they may be optimally represented  
153 by Fourier series of up to the 3rd order harmonic, providing for the spatial profile of the  
154 model expressed through the latitude  $\varphi$ , with the general expression of eq. (9).

$$155 f(\varphi) = \alpha_0 + \sum_{i=1}^3 (\alpha_i \cos(i\omega_0\varphi) + b_i \sin(i\omega_0\varphi)) \quad (9)$$

156 The fundamental frequency  $\omega_0$  was set equal to 2 to satisfy the condition that  $\varphi$  takes values  
157 in  $[-\pi/2, \pi/2]$ .

158 The Fourier coefficients  $\alpha_i$ ,  $b_i$  and the zero frequency component  $\alpha_0$  for the  $A(\varphi)$ ,  $B_1(\varphi)$ ,  
159  $B_2(\varphi)$ ,  $C_1(\varphi)$  and  $C_2(\varphi)$ , obtained through harmonic regression using the nonlinear least  
160 squares method taking into account the aforementioned fundamental frequency, are provided  
161 in Table 1.

162

163 Table 1. The Fourier coefficients of the up to 3rd order harmonics of the  $H(n, \varphi)$  model  
164 parameters.

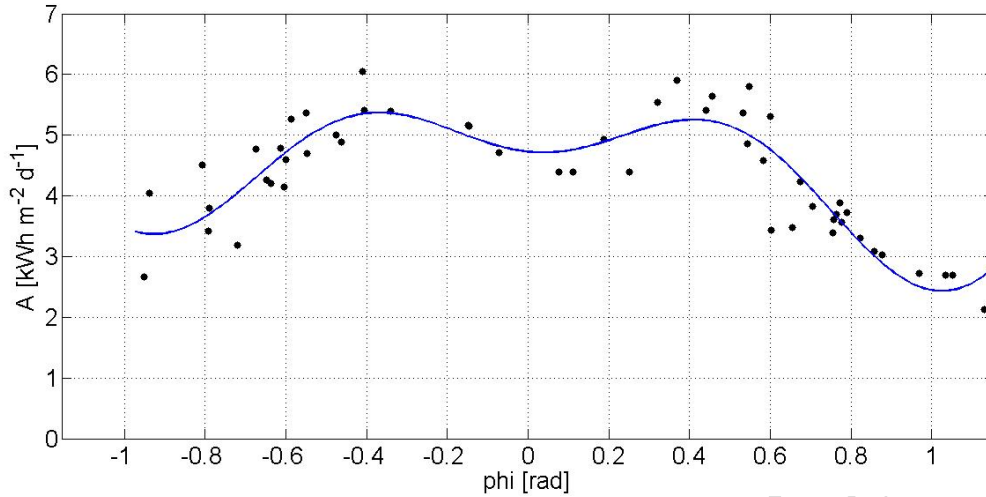
Fourier coefficients	$H(n, \varphi)$ model parameters				
	$A(\varphi)$	$B_1(\varphi)$	$B_2(\varphi)$	$C_1(\varphi)$	$C_2(\varphi)$
$\alpha_0$	4.5180	1.3040	-1.2020	1.9160	1.9160
$\alpha_1$	0.2055	-0.9208	0.9841	-2.1840	-1.8300
$b_1$	-0.3439	-1.6650	-1.1560	2.3150	-2.4560
$\alpha_2$	0.9144	-0.3445	-0.1021	-0.4498	-0.8647
$b_2$	0.3526	0.3413	0.3068	-1.7250	1.6520
$\alpha_3$	-0.9101	0.0143	0.2973	0.3847	0.5499
$b_3$	-0.2346	-0.2715	0.1411	0.1346	0.1396

165

#### 166 3.1. On the $\varphi$ dependence of the model coefficient A

167 The zero frequency model coefficient  $A(\varphi)$  in eq.(7) is presented in Fig. 2 and exhibits  
168 symmetry with respect to the y-axis at  $\varphi=0^\circ$ . It provides for the baseline spatial dependence  
169 of the proposed model  $H(n, \varphi)$ .

170



171

172 Fig.2 The zero frequency coefficient  $A$  of the generic model in  $\text{kWh}/\text{m}^2\text{day}$  vs  $\phi$  (rad), as  
 173 obtained from the nonlinear regression analysis for each of the 53 sites. The fitted curve is a  
 174 Fourier series of up to 3rd order harmonics whose coefficients are given in Table 1.

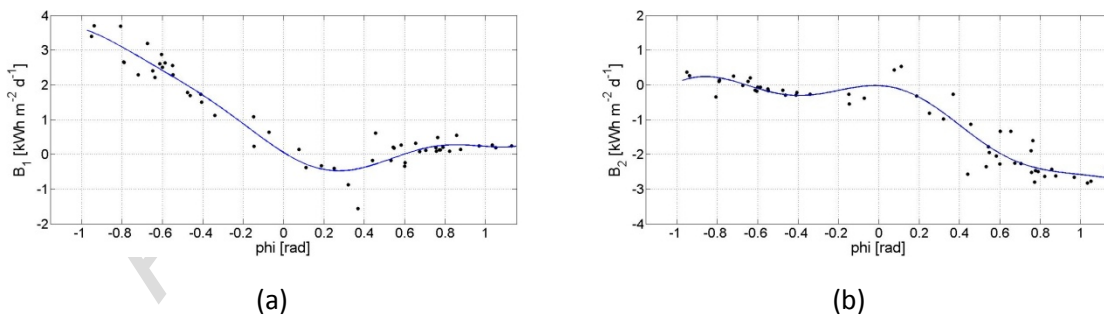
175

### 176 3.2 On the $\phi$ dependence of the model coefficients $B_1$ and $B_2$

177 The model coefficients  $B_1(\phi)$ ,  $B_2(\phi)$  presented in Figs. 3(a)-(b) appear to be anti-symmetric to  
 178 one another, with  $B_1(\phi)$  corresponding mainly to the S. Hemisphere and  $B_2(\phi)$  to the N.  
 179 Hemisphere.  $B_1(\phi)$  and  $B_2(\phi)$  take values close to zero for  $\phi > 0.5$  rad in Fig.3(a) and  $\phi < -0.5$   
 180 rad in Fig.3(b), respectively. This implies that the two time domain harmonics of the model  
 181 with amplitude  $B_1(\phi)$  and  $B_2(\phi)$ , eq.(7), converge to the one cosine model, in sites satisfying  
 182 the above latitude range in either of the Hemispheres. In the region  $-0.5 \text{ rad} \leq \phi \leq 0.5 \text{ rad}$  there  
 183 is contribution from both  $B_1(\phi)$  and  $B_2(\phi)$  in the model with the two time domain harmonics  
 184 differing in frequency (see Section 3.4) and in phase (see Section 3.3).

185  $B_1(\phi)$  and  $B_2(\phi)$  are expressed through a Fourier series of up to 3<sup>rd</sup> order harmonics, whose  
 186 coefficients are provided in Table 1.

187



188

189

190 Fig.3 (a)  $B_1$  ( $\text{kWh}/\text{m}^2\text{day}$ ) and (b)  $B_2$  ( $\text{kWh}/\text{m}^2\text{day}$ ) vs  $\phi$  (rad). The fitted curves are Fourier  
 191 series of up to 3rd order harmonics.

192

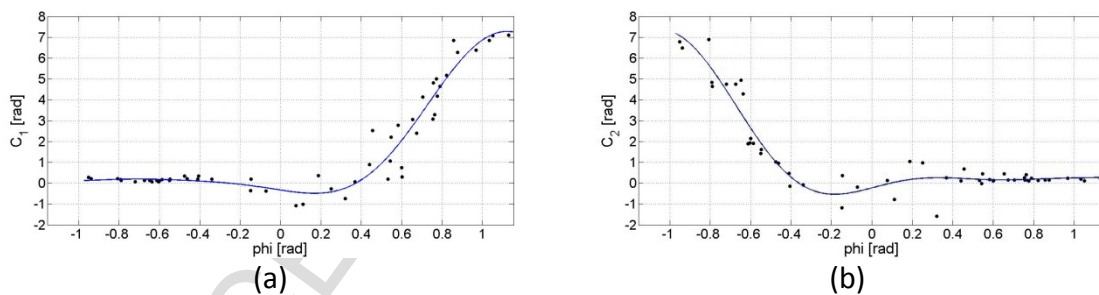
193

### 194 3.3 On the $\varphi$ dependence of the model parameters $C_1$ and $C_2$

195 Parameters  $C_1(\varphi)$  and  $C_2(\varphi)$  correspond to the phase shift in the two time domain harmonics  
 196 of the model. Eqs.(8) were applied in the values of  $C_2$  obtained from the nonlinear regression  
 197 analysis of the  $H(n)$  values from the 53 sites, in order to secure symmetry of the function with  
 198 respect to  $C_1$ . In this process the sign of  $B_2$  was adjusted accordingly. The  $C_1(\varphi)$  and  $C_2(\varphi)$   
 199 shown in Figs.4(a)-(b) are expressed through a Fourier series of up to 3<sup>rd</sup> order harmonics,  
 200 whose coefficients are provided in Table 1.  $C_1(\varphi)$  and  $C_2(\varphi)$  appear symmetric to one another  
 201 with respect to the y-axis at  $\varphi=0^\circ$ . This symmetry is also reflected in the Fourier coefficients  
 202 of  $C_1(\varphi)$  and  $C_2(\varphi)$  shown in Table 1, where generally the respective  $\alpha_i$  coefficients which  
 203 correspond to the cosine (even function) have the same sign, whereas the  $b_i$  coefficients  
 204 which correspond to the sine (odd function) have the opposite sign, reinforcing a mirror  
 205 symmetry between  $C_1(\varphi)$  and  $C_2(\varphi)$  on the y axis at  $\varphi=0^\circ$ .

206 It may be observed that  $C_1(\varphi)$  and  $C_2(\varphi)$  take values close to zero for  $\varphi < -0.5$  rad in Fig.4(a)  
 207 and  $\varphi > 0.5$  rad in Fig.4(b) respectively. This indicates that when the two time domain  
 208 harmonics of the model converge to one cosine model, in either of the Hemispheres, with  
 209  $B_1(\varphi)$  or  $B_2(\varphi)$  prevailing, the corresponding phase shift  $C_1(\varphi)$  or  $C_2(\varphi)$  respectively is zero.  
 210 This reduces the model to one cosine model with zero phase shift. The values of  $C_1(\varphi)$  or  
 211  $C_2(\varphi)$  are larger mainly when the contribution of  $B_1(\varphi)$  or  $B_2(\varphi)$  respectively is small, in  
 212 which case this emphasizes the effects of seasonality in these regions with  $|\varphi| > 0.5$  rad. In the  
 213 tropical and extra-tropical regions with  $-0.5 \text{ rad} \leq \varphi \leq 0.5 \text{ rad}$ , where both the two time domain  
 214 harmonics of the model contribute, the values of  $C_1(\varphi)$  and  $C_2(\varphi)$  reveal larger seasonal  
 215 effects, leading for example in the N. Hemisphere the daily solar radiation to be slightly  
 216 higher in Spring than in Autumn. This effect is illustrated in the 2D and 3D representation of  
 217 the model in Section 3.5.

218



219

220

221 Fig.4 (a)  $C_1$ (rad) and (b)  $C_2$  (rad) vs  $\varphi$  (rad). The fitted curves are Fourier series of up to 3<sup>rd</sup>  
 222 order harmonics.

223

### 224 3.4 On the $\varphi$ dependence of the model parameters $l_1$ and $l_2$

225 The nonlinear regression analysis of the measured  $H(n)$  data for the 53 sites showed  $l_1(\varphi)$  and  
 226  $l_2(\varphi)$  to take values 1 or 2 as presented in Figs. 5(a)-(b), exhibiting a mirror symmetry with  
 227 respect to the y axis at  $\varphi=0^\circ$ . The values of  $l_1(\varphi)$  and  $l_2(\varphi)$  reflect the condition that these are  
 228 multipliers of the fundamental frequency  $2\pi/365$  in the two cosine day-dependent terms of  
 229 eq.(7). Therefore, in the proposed model  $l_1(\varphi)$  and  $l_2(\varphi)$  are provided by the following  
 230 equations.



$$231 \quad l_1(\varphi) = \begin{cases} 1, & \varphi < 0^\circ \\ 2, & \varphi \geq 0^\circ \end{cases} \quad (10a)$$

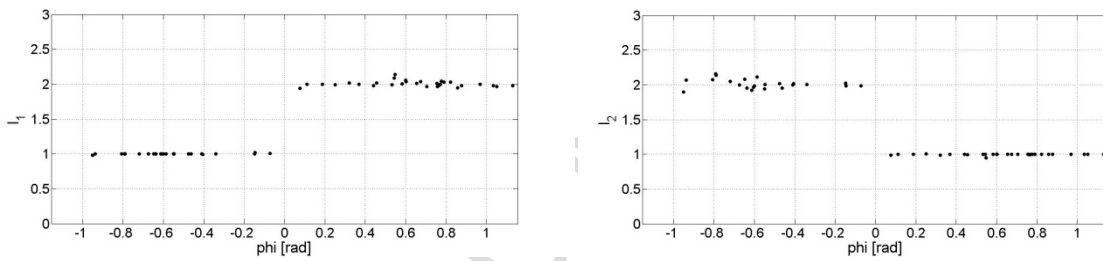
$$232 \quad l_2(\varphi) = \begin{cases} 1, & \varphi \geq 0^\circ \\ 2, & \varphi < 0^\circ \end{cases} \quad (10b)$$

233

234 This indicates that when the two time domain harmonics of the model converge to one cosine  
 235 model, as for example in latitudes with  $\varphi > 0.5$  rad in the N. Hemisphere then the amplitude of  
 236 the cosine  $B_1(\varphi)$  tends to zero and  $B_2(\varphi)$  prevails with an  $l_2(\varphi)$  frequency multiplier equal to  
 237 1, i.e. the frequency is the fundamental  $2\pi/365$ . This agrees with the one cosine model of  
 238 eq.(4). Similar analysis holds for latitudes with  $\varphi < -0.5$  rad in the S. Hemisphere, where  $B_2(\varphi)$   
 239 tends to zero and  $B_1(\varphi)$  prevails with  $l_1(\varphi)$  equal to 1.

240 In regions with latitudes  $-0.5\text{rad} \leq \varphi \leq 0.5\text{rad}$ , the amplitudes  $B_1(\varphi)$  and  $B_2(\varphi)$  are comparable  
 241 with a contribution from both cosines of eq.(7) where one of the  $l_1(\varphi)$  or  $l_2(\varphi)$  is 1 and the  
 242 other 2 as shown in Fig.5. This is reflected in the two peaks of the  $H(n, \varphi)$  profile whose time  
 243 distance depends on the phase shift  $C_1(\varphi)$  and  $C_2(\varphi)$ , as shown in Section 4.

244



245

(a)

(b)

246

247 Fig.5(a) Parameter  $l_1$  and (b)  $l_2$  vs  $\varphi$  (rad).

248

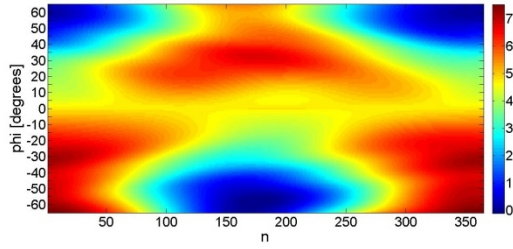
249 **3.5 Model representation**

250 The proposed universal model given by eq.(7) with the coefficients  $A(\varphi)$ ,  $B_1(\varphi)$ ,  $B_2(\varphi)$  and  
 251 the parameters  $C_1(\varphi)$ ,  $C_2(\varphi)$  determined by eq.(9) and  $l_1(\varphi)$ ,  $l_2(\varphi)$  by eq.(10) was executed for  
 252 latitudes from  $-65^\circ$  to  $+65^\circ$  and all days of the year ( $1 \leq n \leq 365$ ). The resulting 2D and 3D  
 253 image representations are shown in Figs.6(a)-(b). For comparison reasons, the 2D and 3D  
 254 image representation of the extraterrestrial solar radiation  $H_{\text{ext}}$  are shown in Figs.6(c)-(d).  
 255 The general spatiotemporal profile of the proposed model is in agreement to the  
 256 extraterrestrial one. Features such as the higher solar radiation received in the S. Hemisphere  
 257 in December compared to the solar radiation received in the N. Hemisphere in June as a  
 258 result of the Earth's orbit, are preserved as shown in Fig.6, where additionally these are also  
 259 higher to that in the equator.

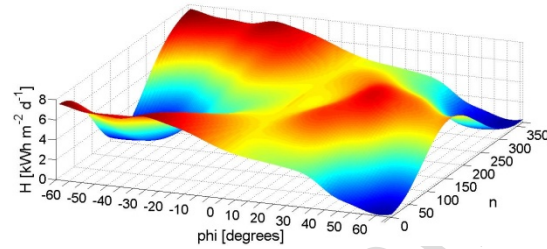
260 Seasonal variations of the global solar radiations may be observed in the 2D and 3D image  
 261 representations of the model. In the N. Hemisphere at the tropical region, where the daily  
 262 solar radiation profile is expressed by two peaks, i.e. both  $B_1(\varphi)$  and  $B_2(\varphi)$  are contributing, a  
 263 higher peak is observed during the spring months than during the autumn months. This is a

264 result of the difference in phase between the two harmonic terms and is generally in  
 265 agreement with solar radiation from databases.

266



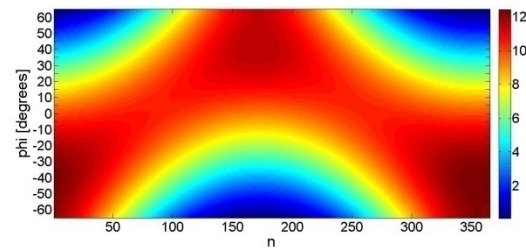
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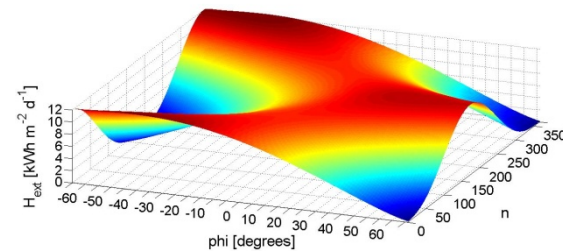
268

(a)

(b)



269



270

(c)

(d)

271 Fig.6 (a) 2D and (b) 3D image representation of the proposed universal model, (c) 2D and (d)  
 272 3D image representation of the extraterrestrial solar radiation respectively, for latitudes from  
 273  $-65^\circ$  to  $+65^\circ$  and the number of day in the year  $n$ . The color map displays the solar radiation  
 274 in  $\text{kWh m}^{-2} \text{d}^{-1}$ .

275

### 276 3.6 Correction for the site altitude

277 The impact of the site altitude to the  $H(n, \varphi)$  prediction was investigated. For sites with  
 278 altitude  $h_s$ , the predicted  $H(n, \varphi)$  values corresponding to the sea level were corrected  
 279 according to eq.(11). The term  $\exp(h_s/h_{\text{atm}}(\varphi))$  is in conformity to other correlated  
 280 atmospheric quantities which affect the solar radiation transmission, such as pressure and air  
 281 density versus altitude [38]. The  $H(n, \varphi, h_s)$  taking into account the site altitude is provided by:

$$282 \quad H(n, \varphi, h_s) = H(n, \varphi) \cdot e^{\frac{h_s}{h_{\text{atm}}(\varphi)}} \quad (11)$$

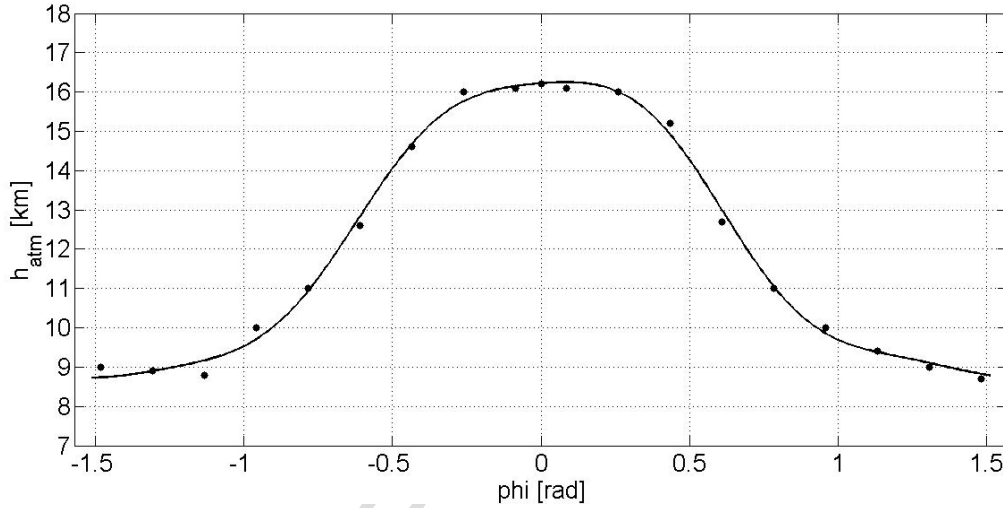
283 where  $h_{\text{atm}}$  is the height of the atmospheric layer for a site,  $\varphi$ . In this paper  $h_{\text{atm}}$  is taken to be  
 284 that of the Tropopause, which includes more than 80% of the air mass. To estimate the  
 285 Tropopause height the vertical profile of the atmospheric temperature is required to determine  
 286 the rate of temperature decrease versus altitude, i.e. the Lapse Rate Tropopause (LRT). The  
 287 tropopause altitude is the lowest level at which the LRT decreases to  $2^\circ\text{C}/\text{km}$  or less provided  
 288 that the LRT in the upper levels does not exceed  $2^\circ\text{C}/\text{km}$ . The LRT is experimentally proven  
 289 to be dependent on the latitude and temperature [39-41], as there is a tropopause warming and  
 290 this causes increase in the tropopause height seasonally [42]. This phenomenon brings a very  
 291 important issue into the proposed model which is the seasonal variations most important in  
 292 regions outside the tropic zones. In general, the altitude of the first LRT decreases from 16.2

293 km in the equator to 8.5 km near the polar regions. Between 20° -50° either N or S there is a  
 294 strong gradient of Tropopause Layer (TL) with  $\varphi$ . The input of the atmospheric height,  
 295  $h_{atm}(\varphi)$  into this proposed model was analyzed in Fourier series with up to 4th order  
 296 harmonics shown in eq.(12), which fitted very well to the LRT data provided in [41] and  
 297 agree with the profiles in [40-43]. This is shown in Fig.7.

$$298 \quad h_{atm}(\varphi) = \alpha_0 + \sum_{i=1}^4 (\alpha_i \cos(i\omega_0\varphi) + b_i \sin(i\omega_0\varphi)) \quad (12)$$

299 where  $\varphi$  is given in radians and the fundamental frequency  $\omega_0=2$  due to the period in  $\varphi$  equal  
 300 to  $\pi$ . The Fourier coefficients  $\alpha_i$ ,  $b_i$  and the zero frequency component  $\alpha_0$  are shown in Table  
 301 2. For more accurate predictions it is important to introduce to the model the seasonal  
 302 variations to  $h_{atm}(\varphi)$  which depend on the latitude and the month.

303



304

305 Fig. 7 The curve of the altitude (km) of the first LRT.

306

307

308

309 Table 2: The Fourier coefficients of the  $h_{atm}$  Fourier series up to the 4th order harmonics.

Fourier coefficients								
$\alpha_0$	$\alpha_1$	$b_1$	$\alpha_2$	$b_2$	$\alpha_3$	$b_3$	$\alpha_4$	$b_4$
11.95	3.971	0.1123	0.7537	0.00892	-0.2332	0.05556	-0.2204	0.00086

309

310

#### 311 4. Results and Model Validation

312 The proposed spatiotemporal model, as expressed by eqs.(7)-(12), was validated with 28  
 313 extra sites with GEBA stations widespread from tropical and extra-tropical, to temperate and  
 314 cold climates. Those sites shown in Fig.1 are independent from the set of sites used for the  
 315 training of the proposed model. The validation was performed against the estimated monthly  
 316 average daily global solar radiation averaged over the years data were recorded in GEBA for

317 each of these sites. A comparison between the predicted  $H(n,\phi)$  monthly mean daily values  
 318 and the GEBA data is given in Figs.8(a)-(f) and Figs.9(a)-(f), where the corresponding data  
 319 profiles from NASA, SoDa and PVGIS databases are also shown, to provide for a complete  
 320 picture of the inherent deviations between the various databases and the predicted  $H(n,\phi)$   
 321 values by this model. The statistical analysis between predicted and measured (GEBA) values  
 322 for the 28 sites is given in Table 3, where the correlation coefficient  $R$ , the Normalized Mean  
 323 Bias Error (NMBE), the Normalized Root Mean Square Error (NRMSE) and  $t$ -statistic results  
 324 are provided. In general, the values predicted by this model follow the profile of the GEBA  
 325 values and in most cases the  $\text{abs}(\text{NMBE})$  is smaller than 0.2, the NRMSE is smaller than  
 326 0.25, the correlation coefficient is higher than 0.90, and the  $t$ -statistic is below or close to the  
 327  $t$  critical value 3.106 for  $\alpha=0.01$ . However there are cases where one or more of the statistics  
 328 fall outside this range and these are discussed below to disclose any factors of deviation. In  
 329 these cases it is very important to examine how the corresponding data from the other  
 330 databases behave and get a complete picture of the proposed model.

331 The predicted  $H(n,\phi)$  profiles shown in Figs.8(a)-(d) for the regions with latitude  $40.67^\circ\text{S}$ ,  
 332  $34.95^\circ\text{S}$ ,  $26.57^\circ\text{S}$ , and  $19.12^\circ\text{S}$  compared to GEBA data have a very good correlation  
 333 coefficient generally higher than 0.99 and low NMBE and NRMSE but relatively high  $t$ -  
 334 statistic especially for the site in Fig.8(c) where this model provides lower than the GEBA  
 335 values, however similar to SoDa. In Fig.8(d), the predicted  $H(n,\phi)$  profile for September-  
 336 December-April is very close to GEBA compared to the other database results. In general, the  
 337 model performs very well and the corrected to the site altitude  $H(n,\phi,h_s)$  values are closer to  
 338 GEBA in most of the periods of the year, see Fig.8(d).

339 Fig.8(e) presents the comparison of the model performance for a site with latitude  $5.08^\circ\text{S}$  in  
 340 the tropic zone in Tanzania with satisfactory statistic results as far as it concerns the NMBE  
 341 and NRMSE, while the correlation coefficient has a low value, and the  $t$  statistic is higher  
 342 than the critical value (site 7 in Table 3). The model results are shown along with the other 5  
 343 databases, GEBA, NASA, SoDa, PVGIS-CMSAF and PVGIS-Helioclim. In this case, it is  
 344 very important to discuss over the low correlation between the predicted and GEBA data and  
 345 investigate on the deviations observed. It is underlined that considerable deviations also exist  
 346 between PVGIS-CMSAF and PVGIS-Helioclim and GEBA. The investigation on the poor  
 347 correlation coefficient,  $R$  for the sites 6, 7, 8 in Table 3, focused on the  $H(n,\phi)$  profiles of  
 348 several sites in Tanzania such as Iringa  $7.67^\circ\text{S}$   $35.75^\circ\text{E}$  (site 6 in Table 3), Arusha  $3.33^\circ\text{S}$   
 349  $36.62^\circ\text{E}$ , Morogoro  $6.83^\circ\text{S}$   $37.65^\circ\text{E}$ , Tabora airport  $5.08^\circ\text{S}$   $32.83^\circ\text{E}$  (site 7 in Table 3) and  
 350 Kilimanzaro airport  $3.42^\circ\text{S}$   $37.07^\circ\text{E}$ . Although these sites differ by  $1^\circ$ - $3^\circ$  in latitude they  
 351 experience largely different profiles. The research study in [44] mapping the Tanzania solar  
 352 resources shows these different profiles which for the case of Morogoro, Arusha and  
 353 Kilimanzaro are similar to the ones predicted by this model as the profile displayed in  
 354 Fig.8(e). The deviations are attributed to the different micro-climatic conditions which  
 355 prevail in the regions in Tanzania with plain and mountainous areas.

356 For the case of Momote  $2.1^\circ\text{S}$ ,  $147.72^\circ\text{E}$  (site 8 in Table 3) shown in Fig.8(f), a large  
 357 deviation is observed in spring and autumn months, and a low correlation coefficient but  
 358 good results for the statistical criteria NRMSE and NMBE and the  $t$ -statistic. It is observed  
 359 that large deviation is also exhibited between the SoDa and GEBA profile for this site. A  
 360 similar performance appears for the site with latitude  $10.33^\circ\text{N}$  in the N. Hemisphere, shown  
 361 in Fig.9(a).

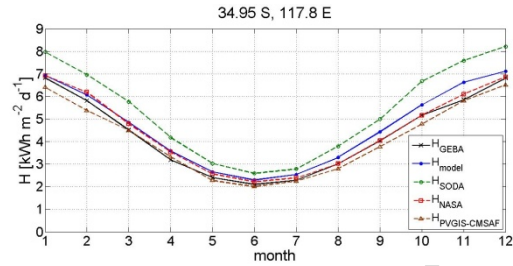
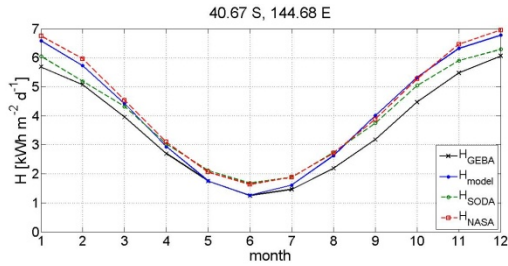
362 A substantial deviation between the predicted and measured  $H(n,\phi)$  values is shown in  
363 Fig.9(b). This is the case of Guangzhou in China, with latitude  $23.13^{\circ}\text{N}$ , and longitude  
364  $113.32^{\circ}\text{E}$ . All four statistics in this case give poor results. An investigation of the large  
365 deviation in this case gave that the clearness index  $K_T$  calculated for this site is significantly  
366 low and correspondingly the fraction of the diffuse to the global solar radiation is  
367 considerably high compared to the other sites of the same latitude, as presented in Table 4.  
368 The reason of the high deviation between predicted and measured is attributed mainly to the  
369 high anthropogenic pollution which prevails in that region rather than to the climatic  
370 conditions. The other databases, and particularly SoDa and PVGIS show also substantial  
371 deviations providing higher values compared to GEBA as shown in Fig.9(b).

372 Fig.9(c) provides the comparison between the results of this model, corrected for the site  
373 altitude, for Lhasa, Tibet ( $29.40^{\circ}\text{N}$ ,  $91.80^{\circ}\text{E}$  and altitude 3.649 km) with the GEBA data and  
374 the other databases. In this case, the correction to the site's altitude was introduced to provide  
375  $H(n,\phi,h_s)$ . The statistical criteria NMBE, NRMSE and R take very good values, but the t  
376 statistic exceeds the critical value. It is noteworthy that the Tibetan Plateau (TP) due to its  
377 large volume and height perturbs the tropopause height especially during the short boreal  
378 Summer when the TP behaves as a heat sink and boosts the Tropopause to higher altitudes by  
379 about 2 km than in the Plain (region in China with the same latitude as TP but different  
380 longitude) while during the boreal Winter the Tropopause altitude in the TP drops even lower  
381 than that in the Plain [45]. The height of the first LRT exhibits considerable seasonal  
382 variations ranging from about 13km during Winter up to about 19km during Summer as  
383 discussed in [45]. These seasonal variations in the first LRT height for the case of Tibet were  
384 introduced directly in the height correction term in the model eq.(11), which resulted in a  
385 considerable improvement of the  $H(n,\phi,h_s)$  prediction. This prediction profile both with the  
386 seasonal  $h_{\text{atm}}$  and with the  $h_{\text{atm}}$  determined by eq.(12) is presented in Fig.9(c). The statistical  
387 results shown in Table 3 correspond to the latter profile. It is interesting to note that the  
388 predicted profile  $H(n,\phi,h_s)$  is in a very good agreement with the results given in [38] for the  
389 TP.

390 Finally, for the sites  $35.05^{\circ}\text{N}$   $106.62^{\circ}\text{W}$  and altitude 1.631km,  $41.7^{\circ}\text{N}$   $87.98^{\circ}\text{W}$  and  $55.35^{\circ}\text{N}$   
391  $131.57^{\circ}\text{W}$  the predicted profiles shown in Figs.9(d)-8(f) are in good agreement with the  
392 measured data expressed also through the statistical criteria where the correlation coefficient  
393 is higher than 0.99 and the NMBE, NRMSE take low values as shown in Table 3. Additional  
394 cases for the model performance are presented for the other sites in Table 3. The agreement  
395 of the predicted solar radiation by the proposed model with the corresponding measured data  
396 from GEBA database for the 28 validation sites is shown in Fig.10 along with the dichotomy.  
397 The resulting correlation coefficient is 0.881 and RMSE  $0.806 \text{ kWhm}^{-2}\text{d}^{-1}$ . The case of  
398 Guangzhou, China and Lhasa, Tibet are shown to have larger deviation.

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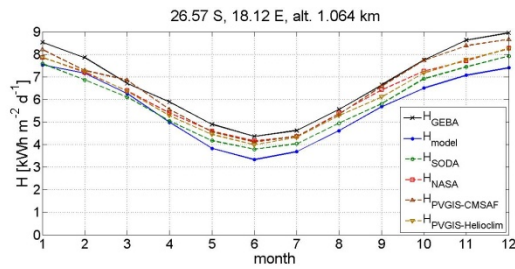
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(a)

(b)

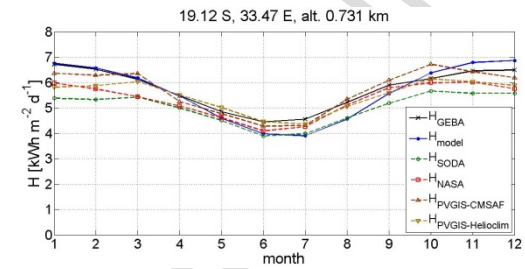


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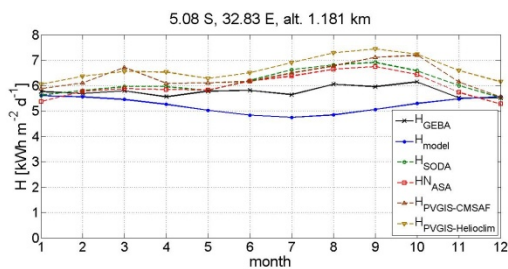
405

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(c)



(d)

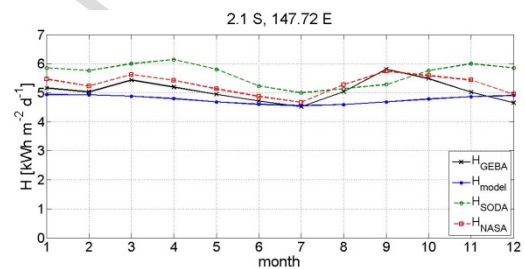


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(e)



(f)

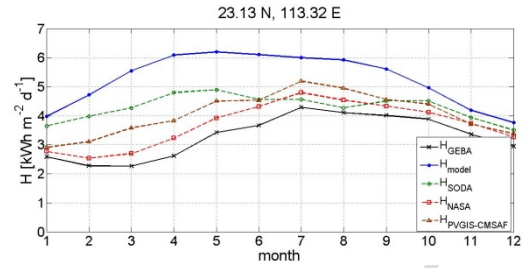
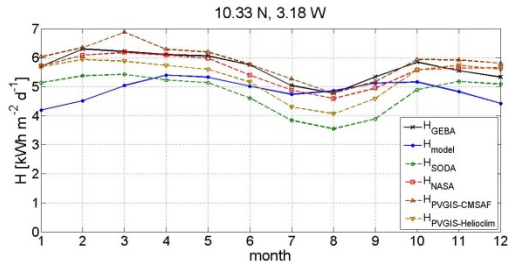
410 Fig 8(a)-(f). Predicted  $H(n,\phi)$  values, corrected to the site's altitude where appropriate, versus  
 411 GEBA, SoDa, NASA, PVGIS-CMSAF and PVGIS-Helioclim available data for different  
 412 sites in the S. Hemisphere. The latitude, longitude and altitude (where appropriate) of the  
 413 sites are shown in the figures' title.

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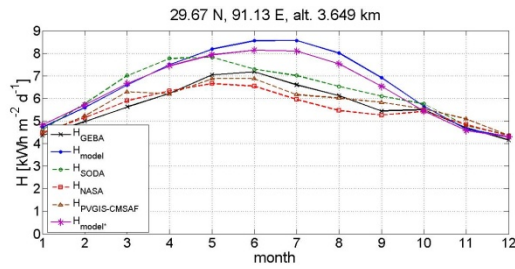
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(a)

(b)

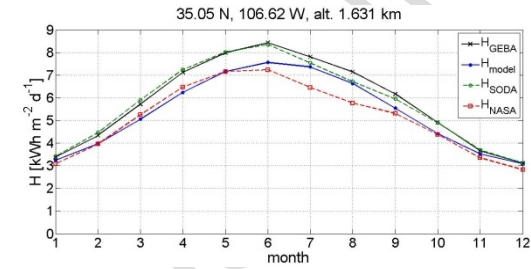


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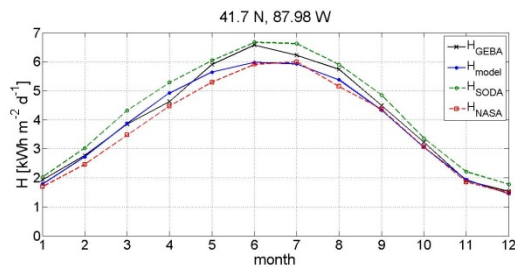
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(c)



(d)

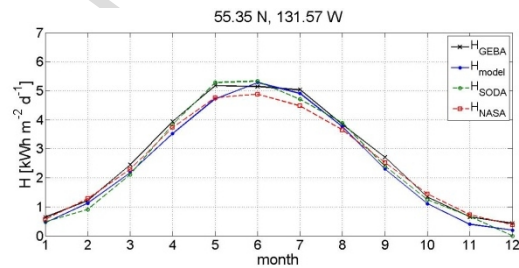


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425

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(e)



(f)

427 Fig. 9(a)-(f). Same as in Figure 8 but for sites in the N. Hemisphere. Note, in 9(c) an  
 428 additional curve  $H_{\text{model}^*}$  is presented in which the height correction uses  $h_{\text{atm}}$  values directly  
 429 from the seasonal altitude of the first LRT for the Tibetan Plateau.

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438 Table 3: Statistics of the comparison between the predicted by this model  $H(n,\varphi)$  and the  
 439 measured values from GEBA database. For the sites with considerable altitude the correction  
 440 for the height based on eqs.(11) -(12) was applied.

441

Site #	Latitude (deg), Longitude (deg), Altitude (km)	NMBE	NRMSE	R	t-statistic
1	-40.67, 144.68	0.140	0.167	0.996	5.165
2	-34.95, 177.8	0.078	0.086	0.996	6.906
3	-26.57, 18.12, 1.064 <sup>a</sup>	-0.153	0.160	0.983	-11.416
4	-19.12, 33.47, 0.731 <sup>a</sup>	-0.019	0.063	0.983	-1.040
5	-17.95, 122.23	-0.150	0.168	0.876	-6.614
6	-7.67, 35.75, 1.426 <sup>a</sup>	-0.148	0.182	-0.240	-4.627
7	-5.08, 32.83, 1.181 <sup>a</sup>	-0.094	0.118	-0.357	-4.326
8	-2.1, 147.72	-0.062	0.093	0.225	-3.014
9	4.4, 18.52	-0.035	0.080	0.711	-1.635
10	10.33, -3.18	-0.138	0.164	0.253	-5.177
11	10.62, -61.35	0.089	0.127	0.425	3.261
12	19.53, 41.05	-0.140	0.147	0.986	-10.553
13	22.65, 88.45	0.154	0.205	0.699	3.760
14	23.07, 72.63	-0.035	0.163	0.568	-0.738
15	23.13, 113.32	0.601	0.657	0.392	7.522
16	23.17, -82.35	0.030	0.062	0.979	1.817
17	29.67, 91.13, 3.649 <sup>a</sup>	0.166	0.204	0.956	4.688
18	32.27, -64.33	0.085	0.103	0.990	4.805
19	35.05, -106.62, 1.631 <sup>a</sup>	-0.086	0.099	0.996	-5.869
20	35.67, 138.62	0.212	0.266	0.942	4.388
21	37.92, 12.52	-0.062	0.114	0.996	-2.141
22	41.7, -87.98	-0.036	0.064	0.994	-2.265
23	49.63, 100.17	-0.196	0.228	0.971	-5.580
24	50.35, 80.25	-0.248	0.265	0.996	-8.779
25	51.32, -108.4	-0.271	0.281	0.996	-12.242
26	51.52, -0.12	0.067	0.137	0.991	1.868
27	55.35, -131.57	-0.080	0.099	0.996	-4.478
28	58.75, -94.07	-0.214	0.257	0.984	-4.959

Note: t critical ( $\alpha=0.01$ ): 3.106

The latitude in the Northern Hemisphere is taken positive and in the Southern Hemisphere negative. The longitude towards East from Greenwich is taken positive and towards West negative.

<sup>a</sup> indicates the corrected to height solar radiation  $H(n,\varphi,h_s)$  for the marked sites with significant altitude

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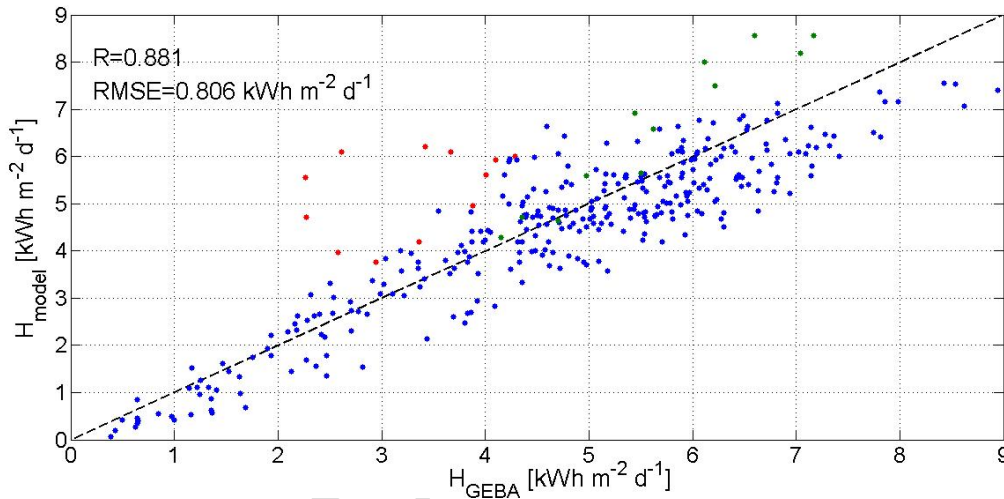
445 Table 4. The ratio of the diffuse over the global solar radiation and the clearness index on an  
 446 annual basis in sites with the same latitude as with Guangzhou and different longitudes.

Site	Latitude	Longitude	$H_d/H$	$K_T^*$
Guangzhou, China	23.13°N	113.32°E	0.66 <sup>a</sup> , 0.55 <sup>b</sup>	0.33
Macau, China	22.20°N	113.54°E	0.48 <sup>a</sup> , 0.51 <sup>b</sup>	0.42
Ahmedabad, India	23.07°N	72.63°E	0.37 <sup>a</sup> , 0.36 <sup>b</sup>	0.61
Casa Blanca, Cuba	23.17°N	82.35°W	0.47 <sup>a</sup>	0.52
Tamanrasset, South Algeria	22.47°N	5.31°E	0.25 <sup>a</sup> , 0.34 <sup>b</sup> , 0.28 <sup>c</sup>	0.70

\*calculated from the ratio  $H/H_{ext}$  with the annually average  $H$  obtained from METEONORM and the annually average  $H_{ext}$  calculated for the specific site latitude  
<sup>a</sup> METEONORM  
<sup>b</sup> PVGIS-CMSAF  
<sup>c</sup> PVGIS-Helioclim

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449

450 Fig. 10.  $H(n, \varphi)$  predicted by the model vs the measured values from GEBA database for the  
 451 28 sites and the 12 months. Correction for the site altitude was applied where appropriate. The  
 452 special cases of Guangzhou, China (red dots) and Lhasa, Tibet (green dots) are highlighted.

453

## 454 5. Discussion

455 The proposed universal model was developed by analyzing  $H(n)$  data for a number of years  
 456 obtained from GEBA database for 53 sites uniformly distributed around the world. The data  
 457 analysis process follows a double spatiotemporal harmonic analysis. The analysis showed  
 458 that the model parameter  $A(\varphi)$  is symmetric with respect to the y axis at  $\varphi=0^\circ$ , and similarly  
 459 the parameters  $C_1(\varphi)$  and  $C_2(\varphi)$ ,  $I_1(\varphi)$  and  $I_2(\varphi)$  take symmetric forms in the  $\varphi$  space mirrored  
 460 on the y axis at  $\varphi=0^\circ$ , while  $B_1(\varphi)$  and  $B_2(\varphi)$  appear anti-symmetric inverted at the origin  
 461  $\varphi=0^\circ$ . The latter is also realized through the 3D image representation of the proposed model  
 462 (Fig.6(b)). Obviously, due to the Earth's orbit, an absolute symmetry between N. and S.  
 463 Hemispheres does not exist and therefore an absolute symmetry in the model parameters was  
 464 not expected.

465 The model was validated in 28 additional sites, randomly selected and covering a large  
 466 geographical space extended within  $\text{abs}(\varphi) < 65^\circ$  and any longitude. For the majority of the  
 467 cases examined at least 3 out of the 4 statistical criteria used had values which displayed a  
 468 successful prediction as compared to GEBA measured data. Predicted profiles were also  
 469 compared with the corresponding profiles from NASA, PVGIS and SoDa databases. It is  
 470 underlined that the validation included some abnormal cases such as the region of Guangzhou  
 471 in China where the anthropogenic environmental pollution reaches  $52 \mu\text{g}/\text{m}^3$  compared to the  
 472 national standard  $35 \mu\text{g}/\text{m}^3$  and is the major factor in deviations between the predicted and  
 473 measured data (Figs.9(b),10). Another such case was sites in Tanzania differing in latitude by  
 474  $1^\circ$ - $2^\circ$ , where the microclimate pattern in that region caused the deviations between the  
 475 predicted  $H(n,\varphi)$  and the measured profiles.

476 An important feature of this model is that it converges to the one cosine model for sites with  
 477  $\text{abs}(\varphi) > 0.5\text{rad}$  where the model coefficients  $B_1(\varphi)$  for the S. Hemisphere and  $B_2(\varphi)$  for the N.  
 478 Hemisphere become almost zero. The investigation of the impact of the site's altitude to  
 479  $H(n,\varphi)$  resulted in an effective correction term dependent on  $\varphi$  and incorporated into the  
 480 model. This was shown in several cases with site altitudes ranging from 0.73 to 3.65 km the  
 481 latter corresponding to Lhasa, Tibet. Additionally, it was shown that the variation of the  $h_{\text{atm}}$   
 482 with latitude plays a significant role in the prediction of solar radiation and the incorporation  
 483 of the seasonal variations of the  $h_{\text{atm}}$  in the  $H(n,\varphi,h_s)$  improves the predicted profile as  
 484 compared to the measured data, shown for Lhasa in Fig.9(c).

485 The monthly average daily global solar radiation data used to train the model were averaged  
 486 over multiple years that data were recorded in the GEBA database for each site, which makes  
 487 the proposed model resilient to annual fluctuations in the solar radiation profile and promotes  
 488 the long-term applicability of the model. Nevertheless, long-term trends with decadal changes  
 489 in the global solar radiation have been analysed in [5,46] and attributed among other causes  
 490 to changes at the tropopause, aerosol characteristics and pollution. This highlights the need  
 491 for consideration of the influence of these parameters in a local but also temporal level.

492

## 493 6. Conclusions

494 The development of a spatiotemporal universal model to predict the expected mean daily  
 495 global solar radiation,  $H(n,\varphi)$ , and its validation results were described and argued in this  
 496 paper. The model is based on a Fourier series of compact form with variable the day of the  
 497 year,  $n$ , while its  $\varphi$ -dependent parameters,  $A$ ,  $B_1$ ,  $B_2$ ,  $C_1$ ,  $C_2$  are given by Fourier series of up  
 498 to 3<sup>rd</sup> order harmonics. It is applicable as a generic model which through a set of  
 499 mathematical expressions may predict the mean expected daily solar radiation at the  
 500 horizontal,  $H(n,\varphi)$ , for any site at any day. Further, it may be used in the prediction of the  
 501 solar irradiance at any hour of the day,  $I(h;n;\varphi)$ , with the least required data. The impact of  
 502 the site's altitude was incorporated into the model using an exponential correction term and  
 503 Fourier series up to the 4th order harmonic for the estimation of the  $\varphi$ -dependent atmospheric  
 504 height necessary for the correction. The results obtained using the altitude correction and the  
 505 seasonal variations of the atmospheric height were impressive. The validation process  
 506 showed that the model is reliable and self-consistent. The predicted  $H(n,\varphi)$  values for a very  
 507 large spectrum of latitudes and longitudes show that the model predicts  $H(n,\varphi)$  very close to  
 508 the measured global solar radiation. The model predicts with a good accuracy the cases where

509  $H(n,\varphi)$  exhibits 2 peaks during a year within and near to the tropic zones. For  $\text{abs}(\varphi) > 0.5\text{rad}$   
 510 the model converges to the one cosine model. Finally, the proposed model can be easily  
 511 incorporated into any sizing software for solar energy applications.

512 Further work will focus on integrating in this model the hourly prediction of the global solar  
 513 irradiance  $I(h,n,\varphi)$  at the site altitude and the seasonal variation of the atmospheric height,  
 514 which are not yet considered in such models, providing a complete generic universal model  
 515 for both  $H(n,\varphi)$  and  $I(h,n,\varphi)$ .

516

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**Highlights**

- A spatiotemporal universal model to predict the mean daily global solar radiation
- Generic model based on Fourier series with symmetries in the N. and S. Hemisphere
- Model incorporates the site altitude and the atmospheric height as a function of  $\phi$
- Model trained using GEBA data from 53 sites and validated at extra 28 random sites
- Model predictions compared with GEBA, NASA, PVGIS and SoDA data