SURFACE SOLAR RADIATION IN A TROPICAL AREA ESTIMATED FROM DIFFERENT MODELS

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ABSTRACT: The Global Solar Radiation (Rg) is considered the main power that regulates the biophysical processes in the surface-atmosphere interface. Many regions of the Earth do not have those kinds of measures available, consequently approach models that allow confident estimates are necessary. Thus, the objective of this article was to evaluate the performance of the models proposed by Angström-Prescost (1940), Hay (1979) and Bristow & Campbell (1984) to estimate the solar global radiation in the city of Sinop, Mato Grosso state. Therefore, data from the National Institute of Meteorology (INMET), from 2006 to 2012, have been used to validate and parameterize the models. Solar Global Radiation (Rg), air temperature (maxima and minima), clarity atmospheric index (kt) and the relative sunshine duration (Ir) have shown seasonality under influence of the cloudiness. The Rg estimates for the models did not differ significantly from the Rg measurements and all the models showed a Pearson’s correlation coefficient classified as "strong", except for those estimated by the Bristow & Campbell’s model. However, the Hay’s model presented smaller errors and higher coefficients of correlation and accuracy than the other models. The results indicate that all the models are applicable to the region, but additional surveys with other models covering all of Mato Grosso are necessary to improve the adjustment of the Rg estimation models

KEYWORDS: solar radiation modelling, sunshine duration, inter-comparison modelling.
nebulosidade. As estimativas de Rg pelos modelos não diferiram significativamente das medidas de Rg e todos os modelos apresentaram coeficiente de correlação de Pearson classificado como “forte”, exceto para aqueles estimados pelo modelo de Bristow & Campbell. No entanto, o modelo de Hay apresentou menores erros e maiores coeficientes de correlação e exatidão que os demais modelos. Os resultados indicam que todos os modelos são aplicáveis à região, mas pesquisas adicionais com outros modelos e abrangendo todo Mato Grosso são necessárias para melhorar o ajuste dos modelos de estimativa de Rg.

PALAVRAS-CHAVE: Precipitation, IAC, ODP, ENOS, Mundaú

1. INTRODUCTION

Brazil is one of the biggest grain producers of the world with annual production estimated in about 207 millions of tons throughout the 2014/2016 crop. Among all Brazilian states that produce grains, Mato Grosso comes as the second biggest agricultural producer, which represents 24% of the entire Brazilian production (CONAB 2014, 2016). The available region to produce grains in Mato Grosso covers about 8% of the total area of the state, mainly the cities of Sapezal, Campo Novo do Parecis, Nova Mutum, Primavera do Leste and Sinop (IBGE 2015).

The municipality of Sinop was created from the colonization processes of the Amazonia and Central-Northern region of Brazil throughout 1970 decade and was fostered by the Programa de Integração Nacional – PIN (National Integration Program) of the federal government (BOTELHO and SECCHI, 2014). This program had as main purpose to give support to immigrants who were looking for fertilized lands and economic prosperity. Sinop gained view in the regional context from the moment its population increased and the social-economical investments diversified through the agriculture expansion (BOTELHO and SECCHI, 2014). However, the agricultural expansion has been related to the intense deforestation that causes many regional climate impacts (SHEIL and MURDIYARSO, 2009), such as alteration on the energy balance pattern, evapotranspiration reduction, increasing of the surface albedo and temperature, and reduction of the surface roughness (BIUDES et al., 2015).

Solar radiation is the main font of energy to all energetic fluxes in the soil-plant-atmosphere system (BORGES et al., 2011). All kind of radiation originated from the Sun that reaches the Earth surface is called global solar radiation (Rg) (QUERINO et al., 2006). The knowledge of Rg and its components is crucial to understand the total available energy into the soil-plant-atmosphere system, and consequently, how the physical, chemical and biological processes, such as photosynthesis and thunderstorm development, happen in the surface-atmosphere interface (SOUZA et al., 2005; QUERINO et al., 2011).

The amount of Rg data available is limited, once there is a low number of weather stations that register this variable due to the high cost and frequently maintenance of the sensors (THORNTON and RUNNING, 1999; WEISS et al., 2001). Therefore, many math models have been suggested mainly based on empiric relation to estimate Rg by using meteorological variable such as relative humidity (YANG and KOIKE, 2002), rainfall (LIU and SCOTT 2001; RIVINGTON et al., 2005), sunshine fraction (TRNKA et al., 2005; CHEN et al., 2006), satellite data (MEHARRAR and BACHARI, 2014) and others. Among the models to estimate Rg, Angström-Prescott (1940) and Hay (1979), both based on the sunshine duration (n), show up as the most popular. The Rg estimated by these
models are driven from the coefficients a and b, which are determined by using linear regression between the atmospheric clearness index (kt) and the relative sunshine duration. Nevertheless, Hay’s model is different from Angström-Prescott’s because it considers the surface reflection factor in its structure. The Bristow and Campbell's model (BRISTOW and CAMPBELL, 1984) assumes that the incoming solar radiation is a function of the local thermal amplitude, as well as of the solar radiation on the top of the atmosphere (Ro).

There are many models to estimate Rg but most of the studies use two types; temperature-based and sunshine-based models. The most famous sunshine-based model is Angström-Prescott’s (LI et al., 2013). This model is commonly used because it suggests a linear relationship between the ratio of average daily Rg and sunshine ratio (DUZEN and AYDIN, 2012). According to the same authors, Angström-Prescott’s performs better than temperature-based or cloud-based, because those last ones must to be calibrated to local parameters rather than others model. However, the limitations of those models are due to the simplifying assumption applied by them (MEHARRAR and BACHARI, 2014).

The estimation of the Rg varies according to each period of the year, which modifies the complexity of the estimation and the adjusting coefficients to reach the best result (BURIOL et al., 2012). Therefore, due to the importance of the agriculture potential of the Sinop city as well as the necessity of understanding the global solar radiation on the region, it is necessary to find alternative ways to estimate Rg. Thus, the objective of this paper was to evaluate the performance of Angström-Prescott (1940), Hay (1979) and Bristow and Campbell's (1984) models, and to adjust their respective coefficients, to estimate Rg in Sinop, MT.

2. MATERIAL AND METHODS

2.1. STUDY AREA

The study was carried out on the municipality of Sinop (Figure 1), which is located in the Central-Northern region of the Mato Grosso state on the margin of the Cuiabá – Santarem highway (BR 163). The site is placed under coordinates 12°07'53” S and 55°35'57” W and is 500 km far away from Cuiabá (capital city of Mato Grosso state). With a total area of 3.942.231 km² and an estimated population of about 126.000 inhabitants, Sinop is considered the biggest urban center of the Northern region of Mato Grosso state (IBGE, 2015).

The climatic condition of the city is hot and humidity with averaged annual temperature around 24ºC. The rainfall pattern is equatorial which is characterized by a dry period throughout the austral winter and a wet season during the summer. The annual amount of rainfall is about 2091.6 mm year⁻¹ and the highest precipitation happens during the months of January, February and March, while the lowest amount of rainfall is observed in June, July and August (BIUDES et al., 2014).
2.2. ACQUIREMENT AND TREATMENT OF DATA

The data were acquired from the National Institute of Meteorology (INMET) between 2006 and 2012. The sunshine duration (n) was measured by a heliograph installed in a weather station placed in an occupation area near Sinop denominated Gleba Celeste (12.00°S and 56.50°W). The global solar radiation (Rg) was measured by the pyranometer LI200SA (LI-COR, Inc) mounted in the automatic weather station of the Sinop under coordinates 11.98°S and 55.57°W. To ensure the quality of the data processes, all negative values of Rg or higher than solar constant were eliminated.

2.3. ESTIMATIVE MODELS OF THE GLOBAL SOLAR RADIATION

Due to the absence of historical solar radiation data in Brazil, especially in the central and northern regions, three models were parameterized: two sunshine-based models and a temperature-based model. The choice of these models considered the most common data available on Brazilian meteorological stations, as well as the simplicity and precision of the models.

2.3.1. ANGSTRÖM-PRESCOTT’S MODEL

The Angström-Prescott’s equation estimates Rg from the relative sunshine duration (Ir) and from the solar radiation on the top of the atmosphere (Equation 1) (DORNEILAS et al., 2006).

\[ R_g = R_o \left[ a + b \left( \frac{n}{N} \right) \right] \] (1)

where Rg is the daily global solar radiation in (MJ m\(^{-2}\) d\(^{-1}\)), n is the effective number of hours that the solar disc was exposed throughout the day (h
d\(^{-1}\) (sunshine duration), \(N\) is the potential sunshine duration (h d\(^{-1}\)) determined by the Equation (2), \(a\) and \(b\) are the linear and the angular coefficients respectively and \(Ro\) is the solar radiation incident on the top of the atmosphere (MJ m\(^{-2}\) d\(^{-1}\)) estimated by Equation (5).

\[
N = \frac{2hp}{15} \quad (2)
\]

where \(hp\) is the hourly angle to the sunset (Equation 3), calculated from the local latitude (\(\varphi\)) and from the solar declination (\(\delta\)) (Equation 4).

\[
hp = \cos^{-1}(\frac{\tan \varphi \tan \delta}{\sin 85^\circ}) \quad (3)
\]

\[
\delta = 23.45 \sin \left(\frac{360(JD + 284)}{365}\right) \quad (4)
\]

\[
Ro = 37.6 \ dr(\sin \varphi \sin \delta + \cos \varphi \cos \delta \sin hp) \quad (5)
\]

Julian day for JD and \(dr\) is the correction of the orbit eccentricity of the Earth (Equation 6).

\[
dr = 1 + 0.033 \cos \left(\frac{360JD}{365}\right) \quad (6)
\]

### 2.3.2. HAY’S MODEL

The Hay’s model is given by the Equation (7), and estimates \(R_g\) from the \(Ro\), from the isolation rates, and also considers the multiple reflections of the atmosphere.

\[
R_g = \frac{Ro}{A} \left[ a' + b' \left(\frac{n}{N'}\right) \right] \quad (7)
\]

where \(a'\) and \(b'\) are linear and angular coefficients respectively, and \(A\) is the adjustment factor associated to the multiple reflections (Equation 8).

\[
A = 1 - \alpha \left[ \beta \frac{n}{N'} + \alpha_c \left(1 - \frac{n}{N'}\right) \right] \quad (8)
\]

where \(\alpha\) is the albedo of the grass (0.20) defined by the Word Meteorological Organization (WMO) as the standard surface where a weather station should be installed. \(\alpha_c\) is the albedo of the cloud bases, usually equals to 0.60, \(\beta\) the scattering coefficient of a clear atmosphere (0.25), \(N'\) is the potential sunshine duration for a specific day, and in this case, considering that the heliograph only registers when the Sun high is upper than 5\(^\circ\) (Brooks and Brooks 1947), is estimated by Equation (9).

\[
N' = \frac{2 \cos^{-1}\left(\frac{\cos 85^\circ - \sin \varphi \sin \delta}{\cos \varphi \cos \delta}\right)}{\pi} \quad (9)
\]
2.3.3. BRISTOW AND CAMPBELL’S MODEL

The Bristow and Campbell’s Model estimates the daily global solar radiation ($R_g$, MJ m\(^{-2}\) d\(^{-1}\)) as function of the daily solar radiation incident on the top of the Earth atmosphere ($R_o$, MJ m\(^{-2}\) d\(^{-1}\)) and also as function of the daily thermal amplitude, which is the difference between the maximum and minimum register of the daily temperature ($\Delta T, ^\circ C$) (Equation 10).

$$R_g = R_o A \left[ 1 - e^{(B \Delta T)} \right]$$  \hspace{1cm} (10)

The empiric constants A, B and C have a physical meaning, once that the A coefficient represents the maxima expected solar radiation for a certain day under clear sky condition and B and C control the variations of A when increasing of the temperature difference happen. The original values of the coefficients are $A = 0.7$, $B$ = between 0.004 and 0.010, and $C$=2.4 (QUEIROZ et al., 2000). The Bristow-Campbell’s model was parameterized by using daily values and we expect that new values of the coefficients can be obtained with the application of the model by using the monthly data.

2.4. ATMOSPHERIC CLEARNESS INDEX (KT) AND ISOLATION RATES (IR)

The Atmospheric Clearness Index (kt) is the ratio between $R_g$ and $R_o$ (Equation 11) (RENSHENG et al., 2004). The relative sunshine duration (Ir) is the ratio between sunshine duration (n) and potential sunshine duration (N) (Equation 12).

$$Kt = \frac{R_g}{R_o}$$  \hspace{1cm} (11)

$$Ir = \frac{n}{N}$$  \hspace{1cm} (12)

2.5. STATISTICS ANALYSIS

The monthly, seasonal and annual averages in a confident interval of ± 95% of the meteorological variables and measured and estimated $R_g$ have been calculated by the 1000 iteration of the bootstrapping of the aleatory resampling with substitution (EFRON e TIBSHIRANI, 1993). The values of $R_g$ estimated by the models were confronted with $R_g$ measured in the weather station by Wilmott’s accordance index $d$ (Equation 13) (WILLMOTT et al., 1985), Pearson’s correlation coefficient $r$ (Equation 14), the Root Mean Square Error (RMSE) (Equation 15) and the Mean Absolute Error (MAE) (Equation 16).

$$d = 1 - \frac{\sum_{i=1}^{n} (P_i - O_i)^2}{\sum_{i=1}^{n} (|P_i - O| + |O_i - O|)^2}$$  \hspace{1cm} (13)

$$r = \frac{\sum_{i=1}^{n} (P_i - P)(O_i - O)}{\sqrt{\sum_{i=1}^{n} (P_i - P)^2 \sum_{i=1}^{n} (O_i - O)^2}}$$  \hspace{1cm} (14)
\[
RMSE = \sqrt{\frac{\sum_{i=1}^{n}(P_i - O_i)^2}{n}} \tag{15}
\]

\[
MAE = \frac{\sum_{i=1}^{n}|P_i - O_i|}{n} \tag{16}
\]

where \( P_i \) are the estimated values of \( R_g \), \( P \) is the averaged value of the estimated \( R_g \), \( O_i \) is the value of the measured \( R_g \), \( O \) is the averaged value of the measured \( R_g \) and \( n \) is the number of observations. The Willmott’s index shows its results based on different levels of performance established on the distance between the estimated and measured values. Its value varies from 0 (no accordance) to 1 (perfect accordance). The Pearson’s correlation coefficient indicates a measure of the level and the signal of the correlation between two variables and it is placed from -1 to 1. The MSRE shows the fail of the model when it estimates variability of the measurement related to the average and gives the variation of the values estimated by the measured values, and the MAE indicates the absolute distance (deviation) of the averages. The minimum limit of the MSRE and of the MAE is 0, which represents a perfect approach between the real data and the model’s estimative. The index proposed by Devore (2006) (Table 1) has been used to classify the correlation between measured \( R_g \) and the one estimated by the models.

**Table 1** - Classification of the Pearson’s correlation coefficient (r) to the \( R_g \) estimative methods (DEVORE, 2006).

| \( r \)       | Definition                        |
|--------------|-----------------------------------|
| 0,00 a 0,19  | Extremely weak correlation        |
| 0,20 a 0,39  | Weak Correlation                  |
| 0,40 a 0,69  | Moderate correlation              |
| 0,70 a 0,89  | Strong correlation                |
| 0,90 a 1,00  | Extremely strong correlation       |

**3. RESULTS AND DISCUSSION**

**3.1. ANALYSIS OF THE METEOROLOGICAL VARIABLES**

The major accumulation of precipitation happened during the seven months of the wet season (October to April) which corresponds to 95% of the annual total amount of the rain, meanwhile the lowest register of rainfall was observed during the five months of the dry season (May to September) equivalent to 5% of the annual total (Table 2 and Figure 2a). The month with the highest sum of rainfall was January, corresponding to 19% of the total precipitation while lowest was registered in July, corroborating to Biudes et al., (2012; 2015). The rainfall regime on the study region is governed mainly by large-scale phenomenon such as the acting of the Bolivia high pressure and South Atlantic Convergence Zone (SACZ). The SACZ is characterized by a huge nebulosity cover that extends from the South of Amazon to the Southeast of Brazil, influencing on the amount of rainfall in the Central-Western region (ESCOBAR, 2014). The Bolivia high pressure acts during the summer and
creates a meridional flow that helps to create instability zones in the central area of the country, inducing the rain formation (VIANELLO and MAIA, 1986; CARVALHO and JONES, 2009). The non-actuation of these phenomena, throughout the winter season, results in the lowest amount of rainfall on the study area (ESCOBAR, 2014).

Table 2 - Monthly, seasonal and annual total of the precipitation and average (± 95% confidence interval) of the solar radiation on the top of the atmosphere (Ro; MJ m⁻² d⁻¹), of the global solar radiation (Rg; MJ m⁻² d⁻¹), averaged (Tavg; °C), maximum (Tmax; °C), and minimum temperature (Tmin; °C), atmospheric clearness index (Kt) and isolation rates (Ir) between the year of 2006 and 2012 in Sinop, Mato Grosso, Brazil.

| Month | Ppt  | Ro     | Rg     | Tavg | Tmax | Tmin | Kt    | Ir    |
|-------|------|--------|--------|------|------|------|-------|-------|
| Jan   | 376.8| 40.1±0.4| 19.4±0.8| 24.4±0.1| 29.9±0.2| 21.4±0.1| 0.49±0.0| 0.39±0.0|
| Feb   | 343.2| 39.5±0.1| 20.0±0.7| 24.7±0.1| 30.6±0.3| 21.4±0.1| 0.51±0.0| 0.41±0.0|
| Mar   | 197.3| 37.4±0.2| 19.9±0.7| 25.1±0.1| 31.3±0.2| 21.5±0.1| 0.53±0.0| 0.47±0.0|
| Apr   | 156.2| 33.8±0.2| 18.7±0.6| 25.3±0.1| 31.5±0.2| 21.1±0.1| 0.55±0.0| 0.56±0.0|
| May   | 29.3 | 30.1±0.1| 18.6±0.4| 25.2±0.3| 31.7±0.3| 19.9±0.4| 0.62±0.0| 0.67±0.0|
| Jun   | 11.2 | 28.2±0.0| 18.4±0.3| 25.6±0.4| 32.9±0.2| 19.0±0.6| 0.65±0.0| 0.77±0.0|
| Jul   | 5.4  | 29.1±0.2| 19.2±0.3| 26.0±0.4| 33.6±0.2| 18.3±0.6| 0.66±0.0| 0.79±0.0|
| Aug   | 6.8  | 32.4±0.2| 20.3±0.3| 26.6±0.3| 35.0±0.2| 17.9±0.5| 0.63±0.0| 0.75±0.0|
| Sep   | 39.8 | 36.1±0.2| 19.9±0.5| 27.4±0.3| 35.6±0.4| 19.8±0.3| 0.55±0.0| 0.58±0.0|
| Oct   | 204.7| 38.7±0.1| 20.3±0.6| 26.2±0.2| 33.2±0.4| 21.4±0.1| 0.52±0.0| 0.48±0.0|
| Nov   | 329.1| 39.7±0.0| 20.1±0.7| 25.3±0.1| 31.4±0.3| 21.6±0.1| 0.51±0.0| 0.42±0.0|
| Dec   | 361.7| 40.0±0.0| 19.8±0.7| 24.5±0.1| 30.0±0.2| 21.4±0.1| 0.48±0.0| 0.37±0.0|
| Wet   | 1969.0| 38.5±0.1| 19.6±0.7| 25.1±0.1| 31.1±0.3| 21.4±0.1| 0.51±0.0| 0.44±0.0|
| Dry   | 92.5 | 31.2±0.1| 19.2±0.4| 26.2±0.3| 33.8±0.3| 19.0±0.5| 0.62±0.0| 0.71±0.0|
| Annual| 2061.6| 35.4±0.1| 19.5±0.6| 25.5±0.2| 32.2±0.3| 20.4±0.3| 0.56±0.0| 0.56±0.0|

The maximum Ro has happened in January and the lowest in July (Table 2 and Figure 2b). The interval with the lowest monthly average of Ro was from April to August and the highest averages from September to March. These intervals, respectively, correspond to the period that the Sun, on its apparently position, is located in the North Hemisphere (autumn and winter) and South Hemisphere (spring and summer), then, lower and higher solar radiation will occur on the top of the atmosphere (DALLACORT et al., 2004).

In a general, Rg has shown less accentuated variations when compared with the Ro’s temporal dynamic. The period of the maximum averaged of Rg was observed between October and April (rainy season), when the Rg was nearly 2% higher than the values registered between May and September (dry season) (Table 2 and Figure 2d). Nevertheless, the maximum monthly averaged
of Rg was noticed in August (dry season) when the cloud cover is lower in the region (BIUDES et al., 2015).

Figure 2 - Total precipitation (a), Solar Radiation on the top of the Atmosphere (Ro) (b), average, maxima and minima temperature of the air (c), Global Solar Radiation (Rg) (d), relative sunshine duration (Ir) (e) and Atmospheric Clearness Index (kt) (f) to the period from 2006 through 2012 in Sinop – MT.

The highest averages of Rg were observed in the wet season and are related with high intensity of Ro throughout that period. Thus, even though a reduction of Rg in function of the cloudiness during the rainy time is noticed, the intensity of Ro overcomes the averages values of Rg when compared to the dry period. Comparing values of Rg and Ro, it was observed a relation of 49% between Rg and Ro in January (maximum monthly average of Ro), and 65% in
June (minimum monthly average of Ro). These results suggest that the oscillation of Rg in Sinop does not depend only on the intensity of Ro, but also depends on the local atmospheric clearness (kt), which determines what is the portion of Ro that is attenuated by the atmosphere (QUERINO et al., 2011).

The mean air temperature during the study period oscillated between 24 and 27°C with values throughout the dry season 4% higher than the rainy period (Table 2 and Figure 2c). Between October and April (rainy season) it was observed the lowest variations among maxima and minima air temperature, with thermal amplitude varying from 8 to 12°C, and from May to September (dry period) the thermal amplitude oscillated from 12 to 17°C.

The lowest and uniform thermal amplitudes during the wet time are in function of the atmospheric moisture. The water acts as an air temperature moderated factor due to its elevated specific heating, inhibiting an abrupt increasing and decreasing of the air temperature. However, large daily amplitudes on the dry period happen due to the low quantity of clouds (and consequently water vapor) in the atmosphere. Despite of a high warming during the diurnal time, most of the infrared radiation, emitted by the surface, is rapidly released to the atmosphere and escapes to the space, causing cooling of the air during the nocturnal time. Another cause that can influence the thermal amplitude is the phenomenon called “friagens”, common in the region and that normally generate significant decreasing in the air temperature (MARENGO and NOBRE, 2009; BIUDES et al., 2012).

The maximum value of Ir has happened in July and the minimum in December (Table 2 and figure 2e). The occurrence interval of the lowest values of Ir was in the period from October to April (rainy period) and the highest from May to September (dry period). The reason is probably because the sky presents less nebulosity (high insolation n) throughout the dry season, it presents also the lowest potential sunshine duration (N) (autumn and winter) and, consequently, Ir tends to be higher.

The kt has shown similar patterns of the Ir (Table 2 and Figure 2f) and its monthly maximum was in July while the minimum in December. That means a reduction of 28% of the incoming solar radiation reaching the surface when compared the months of maximum and minimum kt. The months with highest values of kt were concentrated from May to September, and the lowest from October to April, dry and wet seasons respectively.

We could observe an inversely proportional relation among the lowest values of kt with the months of high amount of precipitation, due to the major concentration of clouds during this period. The nebulosity, defined as the cloud cover in a certain place, acts as the main attenuation factor of the incoming solar radiation over surface (MENEZES and DANTAS, 2002). Another element that should be considered in our analysis is that during the rainy season, the sun is placed in the South Hemisphere, and consequently, highest values of Ro. Nevertheless, despite of biggest radiative fluxes on the top of the Earth atmosphere, just a small portion of the solar radiation will reach the surface, due to the concentration of clouds that tends to scatter the radiation, and reduce the atmospheric transmissivity (QUERINO et al., 2011).
3.2. APPLICATIONS OF THE MODELS

The lowest values of the coefficient a to the Angström-Prescott’s model occurred in January, April and December, and the highest in July, and have shown smooth seasonal variations during the period (Table 3). On the other hand, the coefficient b has shown the highest values in December and the lowest in July with low seasonal variation. The a coefficient indicates a low kt in a day totally cloudy, what explain the occurrence of the lowest values of a in the rainy period (CAMPELO, 1998; DALLACORT et al., 2004). The sum of the coefficients a and b is related with the maximum kt when the Ir tends to 1. Hence, considering the annual coefficients to a total situation of nebulosity, it was verified for a kt equals to 0.32, with an Ir tending to 1, we obtain kt equals to 0.75.

The lowest values of a on the Hay’s model happened in January, April and December, while the major ones were noticed in July (Table 3). The maximum coefficient b was observed in December and the lowest in July. The angular coefficients determined by Hay’s equations are similar to that one obtained with Angström-Prescott’s equation. This similarity is explained because of the results of the compensatory effect of the differences presented between these two models, where the introduction of the insolation rates that effectively can reach 1, would have the opposite effect to the insertion of parameters that consider the multiple reflection of the short wave (CAMPELO, 1998).

The A coefficient to the Bristow & Campbell’s equation was fixed in all equations, varying only the coefficients B and C (Table 3). The lowest values of B happened in January, June, July, August and October, and the highest in May. The C coefficient has shown the lowest value in May and the highest in January.

Table 3 - Annual, Seasonal and Monthly coefficients (± confidence interval) to the Angstrom-Prescott, Hay and Bristow & Campbell models to Sinop - MT.

| Period          | Angström-Prescott | Hay | Bristow & Campbell |
|-----------------|-------------------|-----|---------------------|
|                 | a     | b     | R²   | a'    | b'   | R²   | A   | B   | C   | R²   |
| January         | 0.28±0.03 | 0.51±0.07 | 0.61 | 0.25±0.03 | 0.50±0.06 | 0.65 | 0.7 | 0.004 | 2.55 | 0.67 |
| February        | 0.31±0.03 | 0.47±0.06 | 0.64 | 0.27±0.03 | 0.47±0.06 | 0.66 | 0.7 | 0.022 | 1.74 | 0.69 |
| March           | 0.31±0.03 | 0.48±0.06 | 0.65 | 0.27±0.03 | 0.47±0.05 | 0.70 | 0.7 | 0.010 | 2.01 | 0.30 |
| April           | 0.28±0.04 | 0.48±0.06 | 0.65 | 0.24±0.03 | 0.48±0.05 | 0.69 | 0.7 | 0.025 | 1.65 | 0.26 |
| May             | 0.36±0.04 | 0.37±0.05 | 0.57 | 0.31±0.03 | 0.39±0.05 | 0.63 | 0.7 | 0.063 | 1.28 | 0.38 |
| June            | 0.31±0.04 | 0.45±0.05 | 0.72 | 0.26±0.03 | 0.46±0.04 | 0.76 | 0.7 | 0.004 | 2.41 | 0.70 |
| July            | 0.40±0.07 | 0.32±0.09 | 0.44 | 0.35±0.07 | 0.35±0.08 | 0.51 | 0.7 | 0.004 | 2.50 | 0.31 |
| August          | 0.29±0.04 | 0.45±0.05 | 0.70 | 0.25±0.04 | 0.46±0.05 | 0.74 | 0.7 | 0.004 | 2.39 | 0.24 |
| September       | 0.35±0.03 | 0.34±0.05 | 0.61 | 0.30±0.03 | 0.36±0.05 | 0.66 | 0.7 | 0.007 | 2.18 | 0.31 |
| October         | 0.33±0.03 | 0.40±0.06 | 0.58 | 0.29±0.03 | 0.40±0.06 | 0.61 | 0.7 | 0.004 | 2.41 | 0.37 |
| November        | 0.32±0.03 | 0.45±0.07 | 0.57 | 0.27±0.03 | 0.44±0.06 | 0.58 | 0.7 | 0.006 | 2.33 | 0.72 |
| December        | 0.28±0.03 | 0.52±0.07 | 0.62 | 0.24±0.02 | 0.51±0.06 | 0.66 | 0.7 | 0.011 | 2.08 | 0.68 |
| Dry Period      | 0.33±0.02 | 0.40±0.02 | 0.68 | 0.28±0.02 | 0.42±0.02 | 0.72 | 0.7 | 0.004 | 2.36 | 0.46 |
| Wet Period      | 0.30±0.01 | 0.47±0.02 | 0.62 | 0.26±0.01 | 0.46±0.02 | 0.67 | 0.7 | 0.046 | 1.45 | 0.35 |
| Annual          | 0.33±0.02 | 0.40±0.02 | 0.68 | 0.27±0.01 | 0.44±0.01 | 0.74 | 0.7 | 0.012 | 2.11 | 0.43 |
The Rg estimated by all models was not significantly different from the Rg measured in the study area, except the Bristow & Campbell model in the monthly and annual survey, which underestimated Rg by 3.1% and overestimated Rg by 5.5%, respectively (Table 4). The Angström-Prescott and Hay's models have presented good estimates to all study period (monthly, seasonal and annual), with accordance index (d) oscillating from 0.86 and 0.87, and correlation coefficient (r) varying between 0.76 and 0.79, and then, been classified as strong correlation (Figure 3 and Table 4).

The Bristow & Campbell’s model had accordance index below 0.80 in the seasonal analysis and 0.83 in the annual and monthly investigations (Figure 3 and Table 4). The correlation coefficient was below 0.65 to the seasonal evaluations (moderated correlation) and has varied between 0.72 and 0.70 (strong correlation) in the annual and monthly comparisons respectively.

Figure 3 - Annual, seasonal and monthly relations between the Rg measured and estimated with the Angström-Prescott (a, b, c), Hay (d, e, f), and Bristow & Campbell’s model (g, h, i).
Table 4 - Mean ± CI (95% Confidence Interval), Root Mean Square Error (RMSE), Mean Absolute Error (MAE), Willmott’s accordance index (d), Pearson’s correlation coefficient (r) and intercept and slope of linear regression of estimated and measured solar radiation (Rg).

| Models          | Mean±CI  | RMSE | d     | MAE  | r     | Intercept | slope |
|-----------------|----------|------|-------|------|-------|-----------|-------|
| Annual Rg (Measured) | 19.66±0.20 |      |       |      |       |           |       |
| Annual Rg (Prescot)   | 19.77±0.16 | 2.67 | 0.86  | 1.99 | 0.76  | 7.32      | 0.63  |
| Seasonal Rg (Prescot) | 19.40±0.16 | 2.68 | 0.86  | 2.01 | 0.76  | 6.51      | 0.65  |
| Monthly Rg (Prescot)  | 19.56±0.16 | 2.64 | 0.87  | 1.95 | 0.77  | 6.67      | 0.65  |
| Annual Rg (Hay)      | 19.53±0.17 | 2.42 | 0.86  | 1.82 | 0.78  | 7.93      | 0.59  |
| Seasonal Rg (Hay)    | 19.55±0.16 | 2.41 | 0.87  | 1.81 | 0.78  | 7.46      | 0.62  |
| Monthly Rg (Hay)     | 19.56±0.17 | 2.38 | 0.87  | 1.77 | 0.79  | 7.37      | 0.62  |
| Annual Rg (B e C)    | 20.74±0.19 | 3.47 | 0.83  | 2.56 | 0.72  | 6.55      | 0.72  |
| Seasonal Rg (B e C)  | 19.44±0.17 | 3.88 | 0.77  | 2.72 | 0.64  | 9.89      | 0.49  |
| Monthly Rg (B e C)   | 19.05±0.20 | 3.41 | 0.83  | 2.67 | 0.70  | 4.60      | 0.74  |

The difference presented by Bristow & Campbell’s model and the other two models, especially during the winter, is associated to the thermal amplitude. The linear Angström-Prescott’s model tend to be more accurate due to the established relationship between Rg and relative sunshine duration, and the need for calibration of local physical parameters (DUZEN AND AYDIN, 2012). The lower performance of the Bristow & Campbell model may be due to the model’s inability to explain Rg variability and variation in daily thermal amplitude that was not significant over the study period (BELÚCIO et al., 2014).

4. CONCLUSIONS

The variables Rg, temperature (maxima and minima), kt and Ir have presented seasonality, mainly driven by the nebulosity.

Although the Rg estimates for the models did not differ significantly from the Rg measurements and that all models had Pearson’s correlation coefficient classified as "strong", except for those estimated by the Bristow & Campbell´s model. The Hay´s model showed a better performance in relation to the other models studied. Probably, the best performance of the Hay´s model was due to the insertion of the adjustment factor "A", which considers multiple reflections of Rg in the atmosphere.

Even though the Rg estimated by the Bristow & Campbell´s model differed significantly from the Rg measured, on average that difference was around 5%. This means that in the absence of the measure of sunshine duration, the Rg of the study region can be also estimated by the Bristow & Campbell´s model.

Our results suggest that further research should be carried out, such as application of the models to the other areas of the Mato Grosso state to improve the estimative Rg on the study region, as well as on the whole state.
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