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Evapotranspiration with the Moretti-Jerszurki-Silva model for the Brazilian sub-tropical climate

Aline Aparecida dos Santos , Jorge Luiz Moretti de Souza and Stefanie Lais Kreutz Rosa

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ABSTRACT

The objective of this study was to evaluate the performance of the reference evapotranspiration at the hourly and daily scale, using the Moretti-Jerszurki-Silva alternative approach (MJS; $ET_{O_{MJS}}$), for the Brazilian sub-tropical climate. Data from 25 automatic weather stations were used for calibration and validation analyses. In the linear correlation between $ET_{O_{MJS,h}}$ and ET_{O_h} (ASCE-PM), at the hourly scale, it was found that: (i) values of ET_{O_h} were higher than those of $ET_{O_{MJS,h}}$ in the daytime, while the opposite occurred at night-time; (ii) hourly $ET_{O_{MJS,h}}$ and ET_{O_h} curves had an average two-hour delay; and (iii) the delay correction improves the correlation between $ET_{O_{MJS,h}}$ and ET_{O_h} . Statistically, there was better efficiency between ET_{O_h} and $ET_{O_{MJS,h}}$ in the summer for *Cfa* climate and in the spring for *Cfb* climate. The MJS showed better efficiency concerning the Hargreaves and Samani, modified parametric and Penman-Monteith temperature models, being the best alternative methodology to estimate ET_o at the daily scale in the sub-tropical region of Brazil.

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1 Introduction

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Evapotranspiration (ET) consists of water loss to the atmosphere through soil evaporation, the ground surface and plant transpiration. It is one of the main components of the hydrological cycle, being fundamental in water planning and management in drainage basins and agricultural crops. For a better understanding of trends and interactions between climatic variables in ET , the term reference evapotranspiration (ET_o) was idealized, considering a hypothetical grass reference crop, with uniform and fixed cultivation height (0.12 m for grass and 0.50 m for alfalfa), fixed surface resistance of 70 s m^{-1} and albedo of 0.23. The reference surface resembles an extensive grassy surface, without water restriction, experiencing active growth and completely shading the ground (Allen *et al.* 1998).

ET_o is a water component that is difficult to measure directly, due to the costs for equipment such as evapotranspirometers or lysimeters, as well as the requirement for qualified labour to operate and maintain the equipment. For this reason, numerous indirect ET_o estimation methods have been developed based on meteorological variables, and can be found in the literature (Alves Sobrinho *et al.* 2011, Moura *et al.* 2013, Tegos *et al.* 2013, 2015, 2017, Fenner *et al.* 2019).

Among the indirect methods, the Penman model and its derivatives are widely studied due to their physical basis. In the current literature, the American Society of Civil Engineers (ASCE-PM) model, derived from the Food and Agriculture Organization (FAO) Penman-Monteith method, is considered to be the most suitable, recognized for presenting good precision and approximation with lysimeter data. However, the ASCE-PM model is complex and requires a large amount of

meteorological data (air temperature, relative humidity, wind speed and solar radiation), which is not always available, or may not be available in sufficient quantities or quality for the activity to be performed (Moura *et al.* 2013, Maina *et al.* 2014, Nolz and Rodný 2019).

Due to the difficulty of using highly complex methods of calculation that require many input parameters, simpler alternative methods based on fewer input parameters and climatic variables have been formulated (Owusu-Sekyere *et al.* 2017). Among the models recommended in the literature, the Hargreaves and Samani (HS; Hargreaves and Samani 1985), Penman-Monteith temperature (PMT; Raziei and Pereira 2013, Paredes *et al.* 2020b) and modified parametric (PET; Tegos *et al.* 2017) models stand out due to the simplicity and precision in obtaining daily values of evapotranspiration. The PMT approach uses the Penman-Monteith equation as a base, and the input variables are estimated with equations that consider air temperature. The wind speed is considered using default or regional average values (Raziei and Pereira 2013, Paredes *et al.* 2020b). Tegos *et al.* (2017) introduced an innovative approach, the PET model for estimating potential evapotranspiration. The authors observed high efficiency of the model in relation to other important models for estimating evapotranspiration in different climatic regimes worldwide.

The ASCE-PM method (ASCE-EWRI 2005) allows the estimation of hourly evapotranspiration (ET_{O_h}), including night-time periods. The sum over a 24-h period for ET_{O_h} integrates the values of daily evapotranspiration (ET_{O_d}) (Alves Sobrinho *et al.* 2011, Treder and Klamkowski 2017). Yildirim *et al.* (2004) emphasize the importance of ET_o analysis at the hourly scale, allowing estimates with higher precision and flexibility for agricultural

management. In addition, it has an aspect focused on the physical understanding of this phenomenon. However, research with alternative methodologies that estimate ET_o at the hourly scale is still in the initial phase for regions in Brazil.

Jerszurki *et al.* (2017) and Oliveira (2018) carried out interesting studies with an alternative method for estimating ET_o , which considers the atmospheric water potential (Ψ_{air}) as an input. Oliveira (2018) conducted preliminary studies indicating very satisfactory results in adapting the Moretti-Jerszurki-Silva method (MJS) developed by Jerszurki *et al.* (2017) to estimate ET_o at the hourly scale. The Ψ_{air} calculation requires only the temperature and relative humidity. An interesting aspect is that the measurement of the two variables is easy and makes it possible to estimate ET_o at night-time, something more complicated to be solved by the alternative methods that consider solar radiation. The MJS model considers the atmospheric water potential (Ψ_{air}) as the most sensitive and active component for the occurrence of ET_o (Jerszurki *et al.* 2017). The Ψ_{air} calculation is based on the first and second laws of thermodynamics (Philip 1964, Hillel 1971).

Given the context presented, this study aims to evaluate the performance of the reference evapotranspiration at the hourly and daily scale, using the MJS approach (ET_{oMJS}), for the sub-tropical climate in Paraná State, Southern Brazil.

2 Material and methods

2.1 Study location

The present study was carried out for the Paraná State (Fig. 1), sub-tropical region of Southern Brazil, with an area of 199 307 922 km² and predominance *Cfa* and *Cfb* climate types,

according to Maack (2012). The *Cfa* sub-tropical climate has a good rainfall distribution, with an average 1500 mm year⁻¹, and an average annual temperature of 19°C. The *Cfb* sub-tropical climate presents rainfall of more than 1200 mm year⁻¹, well distributed throughout the year, and a temperate summer, with an annual average temperature of 17°C (Alvares *et al.* 2013).

2.2 Weather data used

Data series from 25 automatic weather stations (Fig. 1) were used, obtained from the National Institute of Meteorology (INMET), covering the period from 1 December 2016 to 8 November 2018.

The following climate data were required to estimate the hourly and daily ET_o with the ASCE-PM (ET_{o_h} and ET_{o_d} , respectively) model: maximum and minimum air temperatures (T ; °C); maximum and minimum relative humidity (RH ; %); incident solar radiation (R_s ; MJ m⁻² h⁻¹); and wind speed at 10 m height (u_{10} ; m s⁻¹), which was later converted to 2 m height (u_2 ; m s⁻¹) at an hourly scale (Allen *et al.* 1998).

The total number of hours analysed is 424 800, or 2 548 800 data points in total (6 variables × 424 800 h = 2 548 800 data), for the 25 stations analysed. However, it was decided to exclude periods with climatic input variable that presented failure on readings data to estimate ET_o , as well as out-of-normal values or outliers. With this data correction, 331 344 hours or 1 988 064 data were effectively used in the ET_{o_h} and $ET_{oMJS,h}$ calculations, representing a 22% reduction in the total hours.

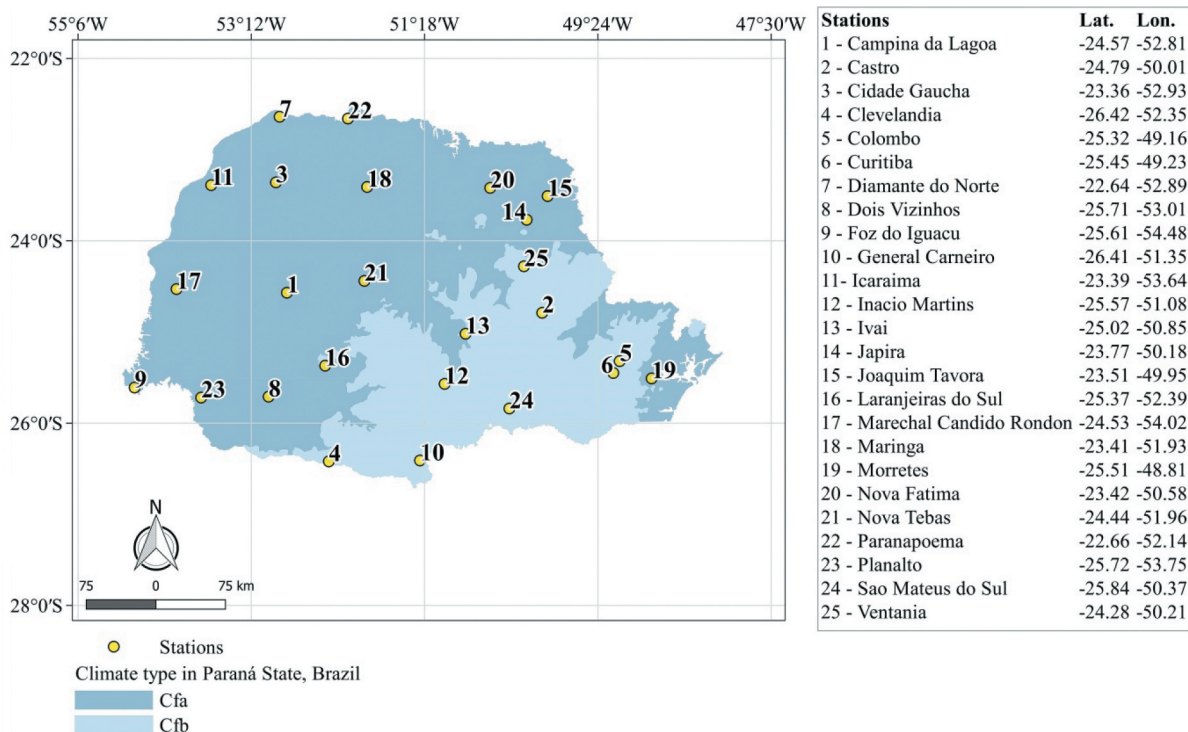


Figure 1. Predominant climate types in Paraná State and location of weather stations. Source: Adapted from Paraná Agronomic Institute (IAPAR 2019); adapted from de Brazilian Institute of Geography and Statistics (IBGE 2010).

2.3 Estimation of reference evapotranspiration (ET_o) at the hourly scale

The estimation of hourly ET_o (ET_{o_h}; standard) was performed with the standardized Penman-Monteith equation (Equation (1)), presented by the American Society of Civil Engineers (ASCE-EWRI 2005), using a short crop height of 0.12 m.

$$ET_{o_h} = \frac{0.408 \cdot (R_n - G) + \gamma \cdot \frac{C_n}{(T+273)} \cdot u_2 \cdot (e_s - e_a)}{\Delta + \gamma \cdot (1 + C_d \cdot u_2)} \quad (1)$$

where ET_{o_h} is the reference evapotranspiration at each *i* hour (mm h⁻¹); Δ is the slope of the saturated water–vapour–pressure curve to the air temperature in the period considered (kPa °C⁻¹); 0.408 is the inverse value of the latent heat of vaporization (λ = 2.45 MJ kg⁻¹); R_n is the net radiation balance in the period considered (MJ m⁻² h⁻¹); G is the soil heat flux in the period considered (MJ m⁻² h⁻¹); γ is the psychrometric constant (kPa °C⁻¹); C_n is the constant related to the type of vegetation and time scale considered (C_{n_{hourly}} = 37 K mm s³ Mg⁻¹ h⁻¹ for soil cover with short grass); T is the average air temperature in the period considered (°C); u₂ is the wind speed at 2 m height in the period considered (m s⁻¹); e_s is the saturation vapour pressure in the period considered (kPa); e_a is the actual vapour pressure in the period considered (kPa); and C_d is the constant related to the type of vegetation and time scale (considered C_{d_{daytime}} = 0.24 s m⁻¹ for daytime period and short grass, or C_{d_{nighttime}} = 0.96 s m⁻¹ for nighttime period and short grass).

The hourly ET_{o_{MJS,h}} was calculated using the MJS model, which considers only the atmospheric water potential Ψ_{air} (ET_{o_{MJS}} as a function of Ψ_{air}; Equations (2) and (3)):

$$ET_{o_{MJS,h}} = a + b \cdot \Psi_{air_h} \quad (2)$$

$$\Psi_{air_h} = \frac{R \cdot T}{M_v} \cdot \ln\left(\frac{e_a}{e_s}\right) \quad (3)$$

where ET_{o_{MJS,h}} is the reference evapotranspiration estimated with the atmospheric water potential (mm h⁻¹); *a* is the linear coefficient obtained from the linear regression equation, resulting from the relation between Ψ_{air_h} and ET_{o_h} (mm h⁻¹); *b* is the angular coefficient obtained in the linear regression equation, resulting from the relation between Ψ_{air_h} and ET_{o_h} (mm h⁻¹ MPa⁻¹; MPa is megapascal); Ψ_{air_h} is the atmospheric water potential at each *i* hour (MPa); R is the gas constant (8.314 J mol⁻¹ K⁻¹); T is the average air temperature in the period considered (K); M_v is the partial molar volume of water (18.10⁻⁶ m³ mol⁻¹); e_a is the actual vapour pressure in the period considered (MPa); and e_s is the saturation vapour pressure in the period considered (MPa).

The analysis with the model that estimates ET_{o_{MJS,h}} was carried out in two stages:

(i) The first stage, according to Jerszurki *et al.* (2017), consisted in calculating the values for the Ψ_{air_h} (Equation (3)) and ET_{o_h} (Equation (1)) series. Then the calibration was performed by a simple linear regression analysis between Ψ_{air_h} and ET_{o_h}, to obtain the *a* and *b* coefficients to use in Equation (2) to estimate the ET_{o_{MJS,h}}. Calibration was performed for 25

weather stations analysed in Paraná State, considering the climate data for the period from 1 December 2016 to 1 December 2017.

(ii) The second stage consisted in analysing the performance of the method that estimates ET_{o_{MJS,h}} (Equation (2)), performing a correlation between ET_{o_{MJS,h}} and ET_{o_h}. Validation analyses were performed for 25 weather stations tested in Paraná State, considering the climate data for the period from 2 December 2017 to 8 November 2018.

2.4 Estimation of reference evapotranspiration (ET_o) at the daily scale

The reference evapotranspiration estimates at the daily scale with the MJS, HS (Hargreaves and Samani 1985), PET (Tegos *et al.* 2017) and PMT methods were compared with the estimates performed by the ASCE-PM method. The estimation was performed for 24 weather stations in the Paraná State, considering the climate data for the period from 2 December 2017 to 8 November 2018.

The methods ASCE-PM (ET_{o_h}; Equation (1)) and MJS (ET_{o_{MJS,h}}; Equations (2) and (3)), calculated at the hourly scale, had the ET_o of the 24 hours of the day added to compose the daily values: ET_{o_d} and ET_{o_{MJS,d}}, respectively. The HS and PET methods were calculated using Equations (4) and (5). The PMT method was calculated with Equation (1), using as input, for each location, the maximum and minimum daily temperatures and the average normal wind speed at 2 m height (Raziei and Pereira 2013, Paredes *et al.* 2020a, 2020b). As the areas analysed have a humid climate (C_{fa} and C_{fb}), the T_{dew} was considered equal to [T_{mean} - 2°C] (Paredes *et al.* 2020b), and the incident solar radiation (R_s) was estimated with the Hargreaves and Samani equation (Equation (6)), adopting the coefficient K_{R_s} = 0.16°C⁻⁵, characteristic of inland sites proposed in FAO 56 (Allen *et al.* 1998).

$$ET_{o_{HS,d}} = 0.0023 \cdot \frac{Ra}{\lambda} \cdot \sqrt{(T_{max} - T_{min})} \cdot (T + 17.8) \quad (4)$$

$$PET_d = \frac{a' \cdot Ra}{1 - c' \cdot \frac{T_{min} + T_{max}}{2}} \quad (5)$$

$$R_{s_{HS}} = k_{R_s} \cdot (T_{max} - T_{min})^2 \cdot Ra \quad (6)$$

where ET_{o_{HS,d}} is the evapotranspiration at each *i* day, estimated with the Hargreaves and Samani equation (mm day⁻¹); PET_d is the potential evapotranspiration at each *i* day, estimated with the modified parametric model (mm day⁻¹; Tegos *et al.* 2017); R_{s_{HS}} is the daily shortwave solar radiation (MJ m⁻² d⁻¹) directly expressed as influenced by K_{R_s} (°C⁻⁵), required in the calculation of the PMT; λ is the latent heat of vaporization (2.45 MJ kg⁻¹); Ra is the extraterrestrial radiation (MJ m⁻² day⁻¹); T is the temperature (°C), which can be maximum, minimum or medium; and *a'* and *c'* are the parameters of PET equation. The extraterrestrial radiation (Ra) was calculated according to Allen *et al.* (1998).

The parameters a' and c' of the PET equation were obtained using the inverse distance weighted (IDW) method, in the R software. Values of a' and c' obtained from Tegós *et al.* (2017) for the Paraná State (Maringá: $a' = 0.0000875$ and $c' = 0.0037$; Curitiba: $a' = 0.0000570$ and $c' = 0.0153$; and Ponta Grossa: $a' = 0.0000547$ and $c' = 0.0166$) were interpolated and extrapolated to the entire state.

2.5 Statistical analysis for comparison

The values of hourly and daily ET_o obtained with the MJS ($ET_{oMJS,h}$ and $ET_{oMJS,d}$), HS ($ET_{oHS,d}$), PET ($ET_{oPET,d}$), PMT ($ET_{oPMT,d}$) and ASCE-PM (ET_{o_h} and ET_{o_d}) methods were compared and verified in linear regression analysis, as well as the main error (root mean square error), index (Willmott's " d " agreement) and coefficient (Pearson's r correlation) recommended in the literature (Jacovides and Kontoyiannis 1995) and the model efficiency (Nash and Sutcliffe 1970):

$$RMSE = \sqrt{\frac{1}{n} \cdot \sum_{i=1}^n (ET_{o_{p_i}} - ET_{o_{a_i}})^2}$$

$$r = \frac{\sum_{i=1}^n [(ET_{o_{p_i}} - \overline{ET_{o_p}}) \cdot (ET_{o_{a_i}} - \overline{ET_{o_a}})]}{\sqrt{\sum_{i=1}^n (ET_{o_{p_i}} - \overline{ET_{o_p}})^2 \cdot \sum_{i=1}^n (ET_{o_{a_i}} - \overline{ET_{o_a}})^2}}$$

$$d = 1 - \frac{\sum_{i=1}^n (ET_{o_{a_i}} - ET_{o_{p_i}})^2}{\sqrt{\sum_{i=1}^n (|ET_{o_{a_i}} - \overline{ET_{o_p}}| |ET_{o_{p_i}} - \overline{ET_{o_p}}|)^2}}$$

$$NSE = 1 - \frac{\sum_{i=1}^n (ET_{o_{p_i}} - ET_{o_{a_i}})^2}{\sum_{i=1}^n (ET_{o_{p_i}} - \overline{ET_{o_p}})^2}$$

where $RMSE$ is the root mean square error (mm period^{-1}); r is the Pearson correlation coefficient (dimensionless); d is the agreement index of Willmott (1982) (dimensionless); NSE is

the Nash-Sutcliffe model efficiency (dimensionless); $ET_{o_{p_i}}$ is the reference evapotranspiration estimated with the standard ASCE-PM method at each i period (mm period^{-1}); $ET_{o_{a_i}}$ is the reference evapotranspiration estimated with the alternative approach (MJS, HS, PET, PMT) at each i period (mm period^{-1}); n is the number of days or hours analysed (dimensionless); $\overline{ET_{o_p}}$ is the mean reference evapotranspiration estimated with the standard ASCE-PM method (mm period^{-1}); and $\overline{ET_{o_a}}$ is the mean reference evapotranspiration estimated with the alternative approach (MJS, HS, PET or PMT; mm period^{-1}). The period considered varies according to the time scale (hourly or daily).

3 Results and discussion

3.1 Estimation of reference evapotranspiration (ET_o) at the hourly scale

Twenty-five weather stations were analysed between 1 December 2016 and 8 November 2018, 15 in the Cfa climate type and 10 in Cfb . Air temperature (T), relative humidity (RH), incident solar radiation (R_s) and wind speed at 2 m height (u_2) in general showed very similar trends among the predominant climates of Paraná (Fig. 2). It was observed that: (i) T was higher in the spring (approximately 20°C for Cfa climate and 18°C for Cfb) and summer (24°C for Cfa and 19°C for Cfb); (ii) RH showed no high seasonal variations for either climate, being between 66% and 80% throughout the year, with winter displaying the lowest RH for Cfa (66%) and summer for Cfb (74.9%); (iii) R_s showed a similar trend to T , with higher R_s periods in the spring ($0.84 \text{ MJ m}^{-2} \text{ h}^{-1}$ for Cfa and $0.75 \text{ MJ m}^{-2} \text{ h}^{-1}$ for Cfb) and summer ($0.95 \text{ MJ m}^{-2} \text{ h}^{-1}$ for Cfa and $0.94 \text{ MJ m}^{-2} \text{ h}^{-1}$ for Cfb); and (iv) u_2 showed a similar trend to RH , with little seasonal variation, being between 0.95 and 1.45 m s^{-1} , with the highest values observed during the autumn period (1.45 m s^{-1} for Cfa and 1.36 m s^{-1} for Cfb).

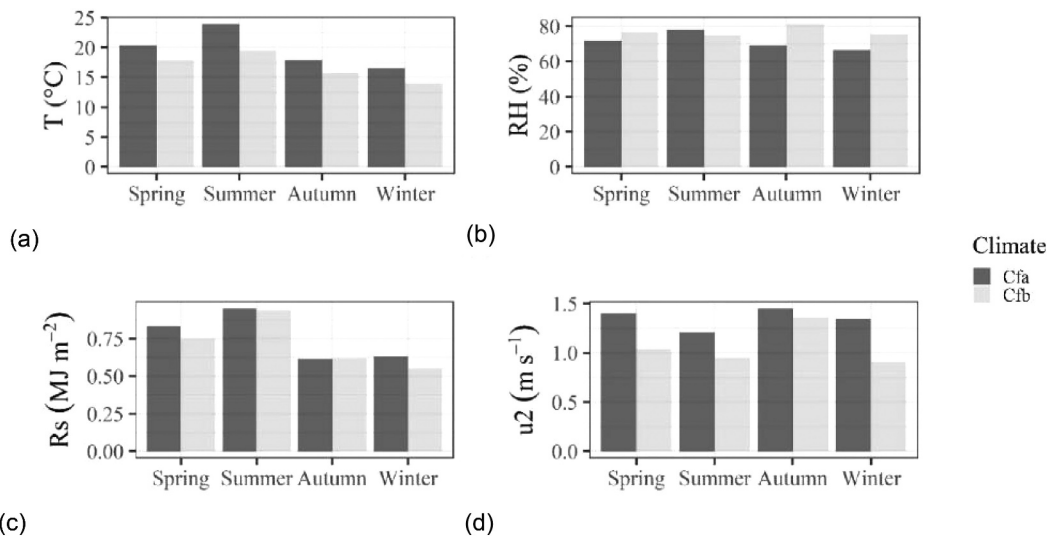


Figure 2. Seasonal average variation of the climatic variables of 15 and 10 weather stations in Cfa and Cfb climate types, respectively, in Paraná State, for the period 1 December 2016 to 8 November 2018: (a) air temperature (T ; $^\circ\text{C}$); (b) relative humidity (RH ; %); (c) incident solar radiation (R_s ; $\text{MJ m}^{-2} \text{ h}^{-1}$); (d) wind speed at 2 m height (u_2 ; m s^{-1}).

3.2 Calibration of the MJS model to obtain “a” and “b” parameters

As the hourly values of ET_o are low, the “a” parameters (linear coefficient of the MJS model) were between -0.0079 and 0.0328 mm h^{-1} for Cfa climate, and between -0.024 and 0.026 mm h^{-1} for Cfb climate (Table 1). The “b” parameters (angular coefficient of the MJS model) were between $-6.96\text{E-}09$ and $-2.61\text{E-}09 \text{ mm h}^{-1} \text{ MPa}^{-1}$ for locations with Cfa climate, and between $-5.59\text{E-}09$ and $-2.9\text{E-}09 \text{ mm h}^{-1} \text{ MPa}^{-1}$ for Cfb . The coefficients obtained in the correlation between Ψ_{air_h} and ET_{o_h} were not as narrow as those obtained by Oliveira (2018), at the hourly scale, for Cfa ($0.81 \leq r \leq 0.96$) and Cfb ($0.84 \leq r \leq 0.90$) climates. However, Oliveira (2018) performed linear and quadratic adjustments in their analyses. The range of correlation coefficients in this study is wider and lower than those obtained by Oliveira (2018) due to the longer period analysed here. In the present study, in addition to making only linear adjustments, a one-year hourly data period was used, whereas Oliveira (2018) used only two months of hourly data in their calibration analyses. Thus, removing the Campina da Lagoa station ($r = 0.37$), the correlation coefficients were $0.51 \leq r \leq 0.77$ for locations with Cfa climate, and $0.58 \leq r \leq 0.78$ for locations with Cfb climate.

The results of calibration performed with the MJS model in the present study are consistent with those obtained by Jerszurki *et al.* (2017) at the daily scale, and with Oliveira (2018) at daily and hourly scale. There is evidence that the

Table 1. Calibration of the MJS model for 25 weather stations located in Cfa and Cfb climate types, in Paraná State, for the period 1 December 2016 to 1 December 2017: linear coefficient (“a”); angular coefficient (“b”); correlation coefficient (r); and number of data in the analyses (n).

Weather station	a (mm h^{-1})	b (mm h^{-1} MPa^{-1})	r (dimensionless)	n
<i>Cfa</i> climate				
Campina da Lagoa	0.0328	-3.1307E-09	0.37	7487
Cidade Gaúcha	-0.0079	-2.9987E-09	0.57	7093
Diamante do Norte	0.0162	-3.0102E-09	0.55	8509
Dois Vizinhos	0.0278	-2.4630E-09	0.55	6524
Foz do Iguaçu	0.0098	-3.9421E-09	0.68	6422
Icaraíma	0.0197	-2.6140E-09	0.51	7651
Japirá	0.0153	-2.6234E-09	0.51	7496
Joaquim Távora	-0.0024	-3.2096E-09	0.58	8389
Marechal Cândido Rondon	0.0020	-3.5451E-09	0.69	7993
Maringá	0.0034	-2.7126E-09	0.56	8371
Morretes	0.0008	-6.9641E-09	0.77	4045
Nova Fátima	0.0085	-4.1062E-09	0.73	6209
Nova Tebas	0.0151	-3.0851E-09	0.57	8417
Paranapoema	-0.0133	-4.0179E-09	0.68	7182
Planalto	0.0208	-3.1454E-09	0.61	7462
Average for <i>Cfa</i> climate	0.0099	-3.4379E-09	0.60	7283
<i>Cfb</i> climate				
Castro	-0.011	-3.486E-09	0.71	8614
Clevelândia	-0.004	-2.990E-09	0.69	4629
Colombo	-0.009	-4.580E-09	0.76	8378
Curitiba	-0.024	-3.595E-09	0.66	5697
General Carneiro	-0.012	-4.610E-09	0.78	8161
Inácio Martins	-0.002	-5.535E-09	0.73	5404
Ivaí	-0.011	-3.674E-09	0.68	8053
Laranjeiras do Sul	-0.009	-4.139E-09	0.73	7461
São Mateus do Sul	-0.001	-3.868E-09	0.73	7405
Ventania	0.026	-3.165E-09	0.58	7042
Average for <i>Cfb</i> climate	-0.006	-3.964E-09	0.71	7084

MJS model has less sensitivity in humid sub-tropical climates, due to the small range between the lowest and highest value of the atmospheric water potential (Ψ_{air}). The lower range, or amplitude, is due to the lower temperatures and higher RH values (Jerszurki *et al.* 2017). All of these aspects contributed to the fact that the correlation coefficient (r) values were not so narrow in the correlation between Ψ_{air_h} and ET_{o_h} , as was verified in arid and semi-arid climates (Jerszurki *et al.* 2017, Oliveira 2018). In addition, the number of data used in the regression analysis for each location was very high ($4045 \leq n \leq 8614$; Table 1), which increases variability and generally tends to reduce the values of r when working with climate data.

With the values of the parameters “a” and “b” of the MJS model for each one of the 25 weather stations analysed in the Paraná State, the results were interpolated and a map of the parameters values was developed (Fig. 3). Once established, the coefficients “a” and “b” from each type of climate can be extrapolated to locations that present the same climatic characteristics, without requiring new calibrations.

3.3 Validation of the MJS model: variation and correlation between $ET_{o_{MJS,h}}$ and ET_{o_h} in Paraná State

The Cfa and Cfb climates showed similar trends between the $ET_{o_{MJS,h}}$ and ET_{o_h} methods. There was no considerable variation in the values of ET_o ($ET_{o_{MJS,h}}$ and ET_{o_h}) between seasons. In the autumn period there were many input data errors (lack of data), and therefore, this seasonal period was not evaluated. Similarly, the months of February and March showed many failures in input data and, for this reason, these months were also not analysed (Table 2 and Fig. 4).

The averages of $ET_{o_{MJS,h}}$ and ET_{o_h} for the seasonal and monthly periods were very close (Fig. 4(a) and (b)). The highest amplitudes between $ET_{o_{MJS,h}}$ and ET_{o_h} occurred in the winter, a period in which the RH remains high and the values of T were lower. Jerszurki *et al.* (2017) and Oliveira (2018), analysing the MJS model as a function of only Ψ_{air} , for several Brazilian climate types, also found that the model performance worsened in these conditions – in other words, for colder and wetter climates or periods. The highest discrepancies occurred in Dois Vizinhos and Nova Tebas cities for Cfa climate, and in Clevelândia for Cfb climate.

In the validation process of the 25 locations in Paraná State, excluding the autumn period (all locations) and Clevelândia and Nova Tebas stations, due to problems with the climatic data, the following statistical indicators of the MJS model were found in the correlation between $ET_{o_{MJS,h}}$ and ET_{o_h} (Table 3):

- The correlation coefficients (r) were lower in winter for Cfa ($0.53 \leq r \leq 0.82$) and Cfb ($0.81 \leq r \leq 0.84$) climates, and higher in summer and spring for Cfa ($0.69 \leq r \leq 0.93$) and Cfb ($0.91 \leq r \leq 0.93$);
- The NSE was higher in the spring ($-2.29 \leq NSE \leq 0.82$) and lower in the winter ($-13.90 \leq NSE \leq 0.49$) for Cfa climate. Disregarding autumn, due to the lack of data, it was found in the Cfb climate that the mean value observed was a better predictor than the simulated value.

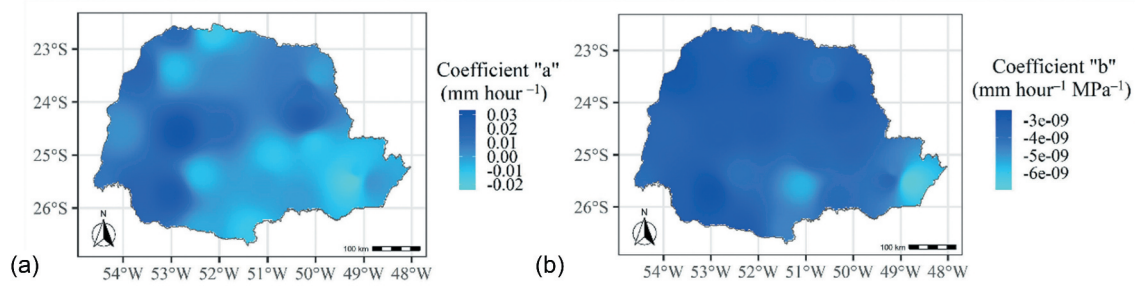


Figure 3. Spatialization of “a” and “b” coefficients of the Moretti-Jerszurki-Silva model, in Paraná State: (a) linear coefficient (“a”; mm h^{-1}); (b) angular coefficient (“b”; $\text{mm h}^{-1} \text{MPa}^{-1}$).

Table 2. Seasonal averages^a of $ET_{O_{MJS,h}}$ and ET_{O_h} (mm h^{-1}) of 25 weather stations in Paraná State, for the period 2 December 2017 to 8 November 2018.

Weather station	Spring		Summer		Autumn		Winter		Annual average	
	$ET_{O_{MJS,h}}$	ET_{O_h}	$ET_{O_{MJS,h}}$	ET_{O_h}	$ET_{O_{MJS,h}}$	ET_{O_h}	$ET_{O_{MJS,h}}$	ET_{O_h}	$ET_{O_{MJS,h}}$	ET_{O_h}
	(mm h^{-1})									
	<i>Cfa</i> climate									
Campina da Lagoa	0.10	0.10	0.10	0.12	–	–	0.10	0.07	0.10	0.10
Cidade Gaúcha	0.15	0.14	0.09	0.14	–	–	0.20	0.11	0.15	0.13
Diamante do Norte	0.10	0.11	0.09	0.12	–	–	0.10	0.07	0.10	0.10
Dois Vizinhos	0.10	0.11	0.09	0.12	–	–	0.10	0.07	0.10	0.10
Foz do Iguaçu	0.12	0.11	0.11	0.12	–	–	0.12	0.07	0.12	0.10
Icaraíma	0.08	0.09	0.09	0.13	–	–	0.09	0.07	0.09	0.10
Japirá	0.09	0.11	0.08	0.12	–	–	0.09	0.07	0.09	0.10
Joaquim Távora	0.09	0.11	0.08	0.12	–	–	0.09	0.07	0.08	0.10
Marechal Cândido Rondon	0.10	0.11	0.09	0.12	–	–	0.10	0.07	0.10	0.10
Maringá	0.08	0.11	0.07	0.12	–	–	0.08	0.07	0.08	0.10
Morretes	0.19	0.11	0.17	0.12	–	–	0.19	0.07	0.19	0.10
Nova Fátima	0.12	0.11	0.11	0.12	–	–	0.12	0.07	0.12	0.10
Nova Tebas	0.12	0.11	0.11	0.12	–	–	0.12	0.07	0.12	0.10
Paranapoema	0.10	0.11	0.09	0.12	–	–	0.10	0.07	0.09	0.10
Planalto	0.20	0.22	0.17	0.22	–	–	0.25	0.17	0.20	0.20
Average for <i>Cfa</i> climate	0.12	0.11	0.10	0.13	–	–	0.12	0.08	0.11	0.11
	<i>Cfb</i> climate									
Castro	0.11	0.12	0.09	0.14	0.08	0.07	0.11	0.08	0.10	0.11
Clevelândia	0.08	0.02	0.07	0.03	–	–	0.08	0.02	0.08	0.02
Colombo	0.12	0.11	0.11	0.12	–	–	0.12	0.07	0.11	0.10
Curitiba	0.08	0.11	0.07	0.12	–	–	0.08	0.07	0.07	0.10
General Carneiro	0.12	0.11	0.11	0.12	–	–	0.12	0.07	0.11	0.10
Inácio Martins	0.15	0.11	0.14	0.12	–	–	0.15	0.07	0.14	0.10
Ivaí	0.12	0.11	0.11	0.12	–	–	0.12	0.07	0.11	0.10
Laranjeiras do Sul	0.11	0.11	0.10	0.12	–	–	0.11	0.07	0.10	0.10
São Mateus do Sul	0.11	0.11	0.10	0.13	–	–	0.11	0.08	0.10	0.10
Ventania	0.11	0.11	0.11	0.12	–	–	0.11	0.07	0.11	0.10
Average for <i>Cfb</i> climate	0.11	0.10	0.10	0.11	0.08	0.07	0.11	0.07	0.11	0.09

^aThe seasons were considered to occur in the following periods: summer begins on 21 December and ends on 20 March; autumn begins on 21 March and ends on 20 June; winter begins on 21 June and ends on 22 September; and spring begins on 23 September and ends on 20 December.

- The “*d*” index, which measures the proximity of the associated values in relation to the 1:1 line, also indicated lower values in winter for both *Cfa* ($0.38 \leq d \leq 0.89$) and *Cfb* ($0.54 \leq d \leq 0.89$) climates, and higher values in summer and spring for *Cfa* ($0.5 \leq d \leq 0.95$) and *Cfb* ($0.73 \leq d \leq 0.97$).
- On average, the *RMSE* values obtained were also low. By season, winter also showed the highest errors for *Cfa* ($0.02 \leq RMSE \leq 0.13$) and *Cfb* ($0.02 \leq RMSE \leq 0.9$) climates, whereas the error was lower in summer and spring for *Cfa* ($0.02 \leq RMSE \leq 0.14$) and *Cfb* ($0.02 \leq RMSE \leq 0.06$).

In general, it was found that $ET_{O_{MJS,h}}$ presented higher values on average than ET_{O_h} at night-time (Fig. 5(a) and (c)). With the sunrise, ET_{O_h} became progressively higher than the

$ET_{O_{MJS,h}}$. Considering the input variables in both models (MJS and ASCE-PM), the results are consistent. *T* and *RH* are considered directly or indirectly in both methodologies. The ASCE-PM method considers *Rs* and u_2 while the MJS method does not. Thus, at night-time there are no *Rs* and the u_2 is low, providing lower values of *ET_o*. During the day, the opposite occurs, considering that *Rs* and u_2 are higher, favouring higher values of *ET_o*.

In checks of all daily and average trends (Fig. 5(a) and (c)) of the hourly *ET_o*, for 25 weather stations in Paraná State (*Cfa* and *Cfb* climates), we found the existence of a delay on the maximum *ET_o* point obtained with the two methodologies (MJS and ASCE-PM). For the ASCE-PM model, the maximum ET_{O_h} occurred at between 12:00 and 14:00 hours (except for Clevelândia station), the time when the highest values of incident solar radiation are generally observed. For the MJS model,

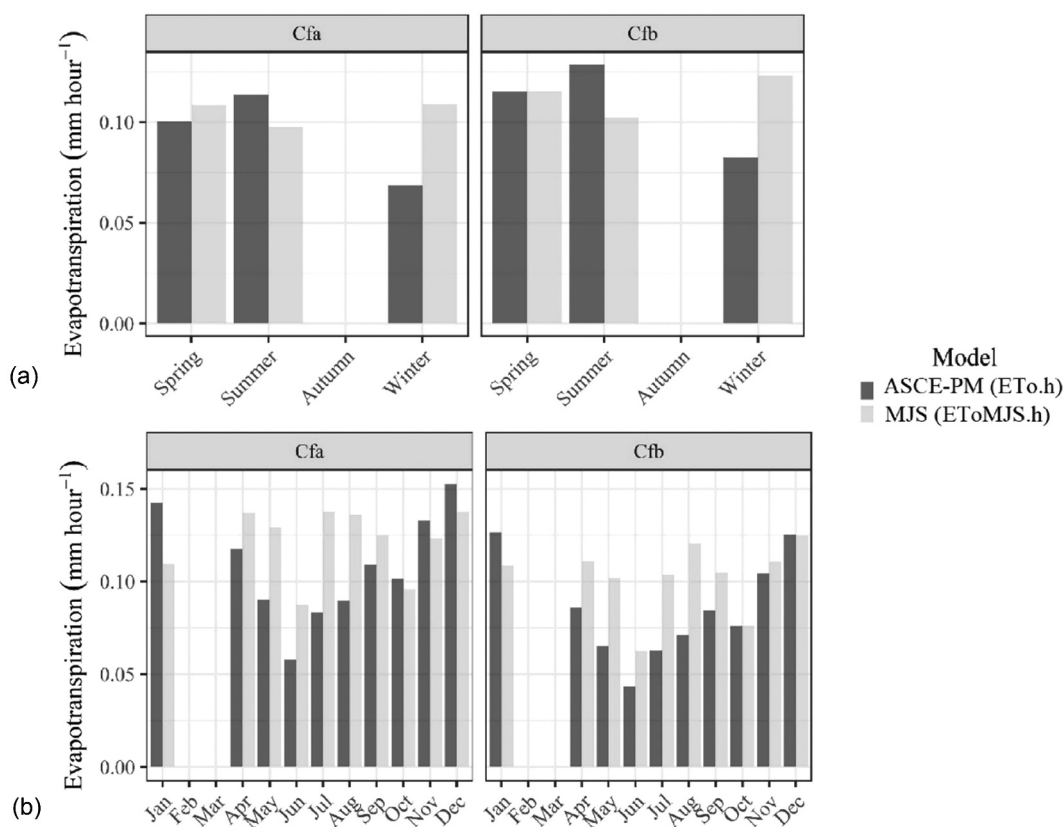


Figure 4. Averages of $ET_{0,MJS,h}$ and $ET_{0,h}$, at the hourly scale, from weather stations according to the *Cfa* and *Cfb* climate types, in Paraná State, for the period 2 December 2017 to 8 November 2018: (a) seasonal average; (b) monthly average.

the maximum $ET_{0,MJS,h}$ occurred between 14:00 and 16:00 hours (Fig. 5(a) and (c)), a period in which the highest temperatures of the day and the lowest relative humidity generally occur. This aspect is interesting and deserves to be investigated. On average, there was a delay of approximately two hours delay in the curve of the hourly $ET_{0,h}$ value throughout the day, estimated using both methodologies (Fig. 5(a) and (c)).

Although the statistical results shown in Table 3 are very promising, the delay in the values of $ET_{0,h}$ estimated with both methodologies (MJS and ASCE-PM) certainly limited the statistical measures used to verify the validation process of the correlation between $ET_{0,MJS,h}$ and $ET_{0,h}$ (Tables 2 and 3). Thus, correcting the hours and analysing the effect of the delay on $ET_{0,h}$ estimates can lead to better and more promising statistical indicators for an alternative method as simple as the MJS, to estimate hourly $ET_{0,h}$.

The correction of the two-hour delay between $ET_{0,MJS,h}$ and $ET_{0,h}$ on all days, at the 25 stations analysed in Paraná State, provided an average trend as shown in Fig. 5(b) and (d). With the correction, the occurrence time of maximum $ET_{0,h}$ estimated with the MJS model started to coincide with the ASCE-PM methodology. With the adjustment made, there was an improvement in the correlation between $ET_{0,MJS,h}$ and $ET_{0,h}$ (Fig. 6).

The existence of a delay between the values of $ET_{0,MJS,h}$ and $ET_{0,h}$ for *Cfa* and *Cfb* climates, in Paraná State, generated uncertainty regarding the existence of a similar trend for the

$ET_{0,h}$ estimated with the two methodologies for the main climates in Brazil (*Af*, *Am*, *Aw*, *BSh*; *Cfa*, *Cfb*, *Cwa* and *Cwb*). This is an important aspect and will need to be investigated in more detail later, considering the findings obtained in the present study and the conclusions of Oliveira (2018). It would be interesting to verify the magnitude and time of occurrence of the highest and lowest hourly values of $ET_{0,h}$ estimated with the ASCE-PM and MJS methodologies, as well as the existence and cause of delays in the trends of the estimated values for $ET_{0,MJS,h}$ and $ET_{0,h}$.

3.4 Estimation of reference evapotranspiration ($ET_{0,h}$) at the daily scale

In the analysed period (2 December 2017 to 8 November 2018), not all alternative approaches tested (MJS, HS, PET and PMT) showed average values of daily $ET_{0,h}$ very close to those obtained with the standard ASCE-PM model (Table 4). The average values of daily $ET_{0,h}$ estimated for the 24 weather stations in the Paraná State were interpolated and presented on maps (Fig. 7). The location of Clevelândia was excluded from the analyses as it had a lot of missing data.

The indexes, errors and statistical coefficients indicated better performance of the MJS model in the 24 locations and climates (*Cfa* and *Cfb*) analysed in the Paraná State (Table 5). Although the values of $ET_{0,MJS,d}$ resulted from the 24-hour sum, as in the ASCE-PM ($ET_{0,d}$), it is believed that the calibration of the “a” and “b” coefficients of the model contributed substantially to improving

Table 3. Seasonal^a and annual values of Nash-Sutcliffe efficiency (*NSE*; dimensionless), “*d*” index (dimensionless), root mean square error (*RMSE*; mm h⁻¹) and correlation coefficient (*r*; dimensionless) between *ETo_{MJS,h}* and *ETo_h*, of 15 and 10 weather stations in *Cfa* and *Cfb* climate types, respectively, in Paraná State, for the period 2 December 2017 to 8 November 2018.

Weather station	Spring				Summer				Autumn				Winter				Annual average			
	<i>NSE</i>	<i>d</i>	<i>RMSE</i>	<i>r</i>	<i>NSE</i>	<i>d</i>	<i>RMSE</i>	<i>r</i>	<i>NSE</i>	<i>d</i>	<i>RMSE</i>	<i>r</i>	<i>NSE</i>	<i>d</i>	<i>RMSE</i>	<i>r</i>	<i>NSE</i>	<i>d</i>	<i>RMSE</i>	<i>r</i>
<i>Cfa</i> climate																				
Campina da Lagoa	0.73	0.90	0.03	0.88	0.52	0.81	0.04	0.89	–	–	–	–	–0.45	0.67	0.04	0.71	0.27	0.79	0.04	0.83
Cidade Gaúcha	0.49	0.89	0.05	0.81	0.11	0.76	0.07	0.84	–	–	–	–	–5.40	0.53	0.11	0.73	–1.60	0.73	0.08	0.73
Dois Vizinhos	0.24	0.77	0.06	0.72	–0.67	0.66	0.07	0.76	–	–	–	–	–8.87	0.38	0.06	0.74	–3.10	0.60	0.06	0.68
Diamante do Norte	0.82	0.94	0.02	0.93	0.55	0.85	0.04	0.92	–	–	–	–	0.06	0.79	0.03	0.81	0.48	0.86	0.03	0.83
Foz do Iguaçu	0.80	0.95	0.02	0.93	0.79	0.94	0.03	0.92	–	–	–	–	–1.31	0.68	0.05	0.81	0.09	0.86	0.03	0.83
Icaraíma	0.76	0.91	0.02	0.92	0.47	0.81	0.05	0.93	–	–	–	–	0.38	0.83	0.03	0.81	0.54	0.85	0.03	0.83
Joaquim Távora	0.72	0.92	0.03	0.93	0.33	0.80	0.05	0.92	–	–	–	–	0.44	0.87	0.03	0.81	0.50	0.86	0.04	0.83
Japirá	0.68	0.89	0.03	0.93	0.29	0.77	0.05	0.92	–	–	–	–	0.49	0.85	0.02	0.81	0.49	0.84	0.03	0.83
Marechal Cândido Rondon	0.84	0.95	0.02	0.92	0.56	0.87	0.04	0.92	–	–	–	–	–0.06	0.80	0.04	0.81	0.45	0.87	0.03	0.83
Maringá	0.55	0.86	0.04	0.93	0.06	0.73	0.06	0.92	–	–	–	–	0.06	0.89	0.02	0.81	0.23	0.83	0.04	0.83
Morretes	–2.29	0.68	0.10	0.93	0.42	0.80	0.07	0.92	–	–	–	–	–13.90	0.40	0.13	0.82	–5.26	0.63	0.10	0.84
Nova Fátima	0.75	0.94	0.03	0.93	0.83	0.95	0.02	0.92	–	–	–	–	–1.72	0.65	0.06	0.81	–0.05	0.85	0.04	0.83
Nova Tebas	0.21	0.71	0.14	0.71	–0.01	0.74	0.07	0.81	–	–	–	–	0.19	0.65	0.11	0.53	0.13	0.70	0.11	0.60
Paranapoema	0.80	0.95	0.02	0.92	0.40	0.85	0.04	0.89	–	–	–	–	–0.13	0.80	0.04	0.80	0.36	0.87	0.03	0.82
Planalto	0.68	0.91	0.05	0.86	–2.26	0.50	0.06	0.69	–	–	–	–	–1.51	0.67	0.10	0.79	–1.03	0.69	0.07	0.73
Average for <i>Cfa</i> climate	0.45	0.88	0.04	0.88	0.16	0.79	0.05	0.88	–	–	–	–	–2.12	0.70	0.06	0.77	–0.50	0.79	0.05	0.79
<i>Cfb</i> climate																				
Castro	0.74	0.94	0.03	0.9	–0.07	0.73	0.06	0.91	0.71	0.03	0.03	0.65	–0.29	0.81	0.05	0.84	0.27	0.82	0.04	0.83
Clevalândia	–15.54	0.09	0.08	–0.34	–15.88	0.08	0.06	–0.37	–	–	–	–	–12.29	0.10	0.08	–0.16	–14.57	0.09	0.07	–0.29
Colombo	0.77	0.95	0.03	0.93	0.79	0.94	0.03	0.92	–	–	–	–	–1.42	0.69	0.05	0.81	0.05	0.86	0.04	0.89
Curitiba	0.55	0.88	0.04	0.93	0.02	0.76	0.06	0.92	–	–	–	–	0.49	0.89	0.02	0.81	0.36	0.84	0.04	0.89
General C.	0.76	0.95	0.03	0.93	0.8	0.95	0.03	0.93	–	–	–	–	–1.47	0.69	0.05	0.81	0.03	0.86	0.04	0.89
Inácio Martins	0.13	0.86	0.05	0.93	0.74	0.94	0.03	0.93	–	–	–	–	–5.22	0.54	0.09	0.83	–1.45	0.78	0.06	0.90
Ivaí	0.77	0.94	0.03	0.93	0.40	0.84	0.05	0.92	–	–	–	–	0.24	0.85	0.03	0.81	0.47	0.88	0.04	0.89
Laranjeiras do Sul	0.83	0.96	0.02	0.92	0.70	0.91	0.03	0.92	–	–	–	–	–0.05	0.76	0.04	0.81	0.49	0.88	0.03	0.88
São Mateus do Sul	0.86	0.97	0.02	0.93	0.63	0.90	0.04	0.92	–	–	–	–	–0.43	0.76	0.04	0.81	0.35	0.88	0.03	0.89
Ventania	0.84	0.95	0.02	0.93	0.71	0.90	0.03	0.91	–	–	–	–	–0.80	0.69	0.05	0.81	0.25	0.85	0.03	0.88
Average for <i>Cfb</i> climate	–0.93	0.85	0.04	0.87	–1.12	0.80	0.04	0.79	0.71	0.03	0.03	0.65	–2.12	0.68	0.05	0.72	–1.40	0.77	0.04	0.71

^aThe seasons were considered to occur in the following periods: summer begins on 21 December and ends on 20 March; autumn begins on 21 March and ends on 20 June; winter begins on 21 June and ends on 22 September; and spring begins on 23 September and ends on 20 December.

the performance. The result is interesting, since Jerszurki *et al.* (2017) considered that the MJS model presents good results at the daily scale, with lower performance in regions with higher *RH* and low temperatures, as occurs in *Cfa* and *Cfb* climates. Therefore, there is an expectation that the spatialization of *ETo_{MJS,d}* in hot and dry regions will be even better.

The statistical results of the HS method (Table 5) did not indicate good performance in some locations, mainly in Dois Vizinhos, Diamante do Norte, Icaraíma, Japirá and Colombo. The HS and PMT models underestimated the values of daily *ETo* in relation to the ASCE-PM model for *Cfa* climate. For *Cfb*, the HS model had the highest underestimation in relation to all tested models (Fig. 7 and Table 4). Several locations presented *NSE* < 0, indicating that the average values of *ETo_d* (ASCE-PM) result in better prediction than the HS model (*ETo_{HS,d}*). The opposite results were observed by Todorovic *et al.* (2013), which in Mediterranean climates observed overestimation of *ETo* with the HS method in relation to the standard ASCE-PM. As the HS method does not consider the relative humidity, which has high value for *Cfa* and *Cfb* climates, the method was less accurate in humid climates, with low performance in relation to the other analysed methods.

There was an overestimation of the daily values of *ETo* with the PET model in relation to the ASCE-PM model (Fig. 7 and Table 4), in *Cfa* and *Cfb* climates, indicating less efficiency for the sub-tropical region of Brazil. In addition, 16 locations presented *NSE* < 0. The lowest efficiencies (*NSE*) occurred in Morretes, Joaquim Távora, Ventania, Ivaí, São Mateus do Sul and Inácio Martins (Table 5). The few parameters *a'* and *c'* available for IDW extrapolation may have contributed to the model's low performance. However, Tegos *et al.* (2017), evaluating evapotranspiration with the PET model for 4088 stations worldwide, also found less precision with the model in the equatorial regions of South America, Africa, Indonesia and the Indian Peninsula. The authors considered that the poor performance was probably because the model does not account for the relative humidity and wind speed. The two variables are not considered in the PET model, but are very active in the evapotranspiration process in the mentioned areas, influencing the net solar radiation and the evaporation demand.

The PMT model underestimated the *ETo_d* presenting results similar to the HS model for the *Cfa* climate. However, in the *Cfb* climate, the PMT was not the model that had the lowest underestimations in relation to the ASCE-PM (Fig. 7; Table 4). Considering the 1:1 line, PMT and HS (Fig. 8(b) and (d)) also

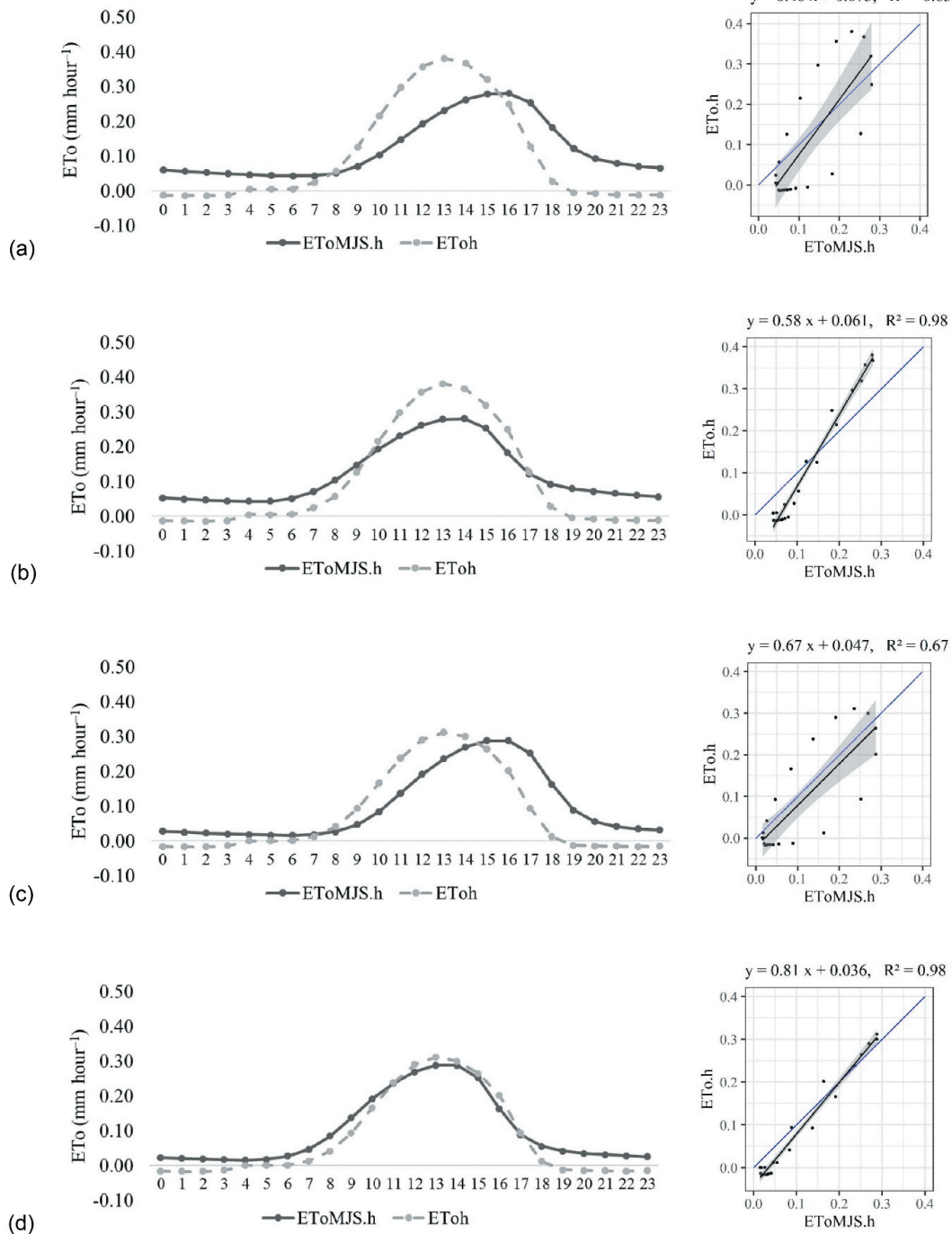


Figure 5. Variation and correlation between $ET_{MJS,h}$ and $ET_{o,h}$, at the hourly scale, of 15 and 10 weather stations in *Cfa* and *Cfb* climate types, respectively, in Paraná State, for the period 2 December 2017 to 8 November 2018: (a) hourly ET_o without delay adjustment, in *Cfa* climate; (b) hourly ET_o with 2-h delay adjusted, in *Cfa* climate; (c) hourly ET_o without delay adjustment, in *Cfb* climate; (d) hourly ET_o with 2-h delay adjusted, in *Cfb* climate.

produced similar results, presenting the lowest adjustment and performance in the analyses (Table 5) for the sub-tropical region of Brazil.

4 Conclusions

In the calibration, the “*a*” coefficients of the MJS model were between -0.0133 and 0.0328 mm h⁻¹ for *Cfa* climate, and between -0.024 and 0.026 mm h⁻¹ for *Cfb* climate. The “*b*”

parameters were between $-6.96E-0.9$ and $-2.46E-0.9$ MPa⁻¹ for *Cfa* climate, and between $-5.53E-0.9$ and $-2.99E-0.9$ mm h⁻¹ MPa⁻¹ for *Cfb* climate.

The values of $ET_{MJS,h}$ are higher than $ET_{o,h}$ at night-time. With the sunrise, the opposite occurs, and the $ET_{o,h}$ becomes progressively higher than $ET_{MJS,h}$.

The curve of hourly ET_o value, estimated using the ASCE-PM and MJS methodologies, presents, on average, a two-hour delay between the maximum values of hourly

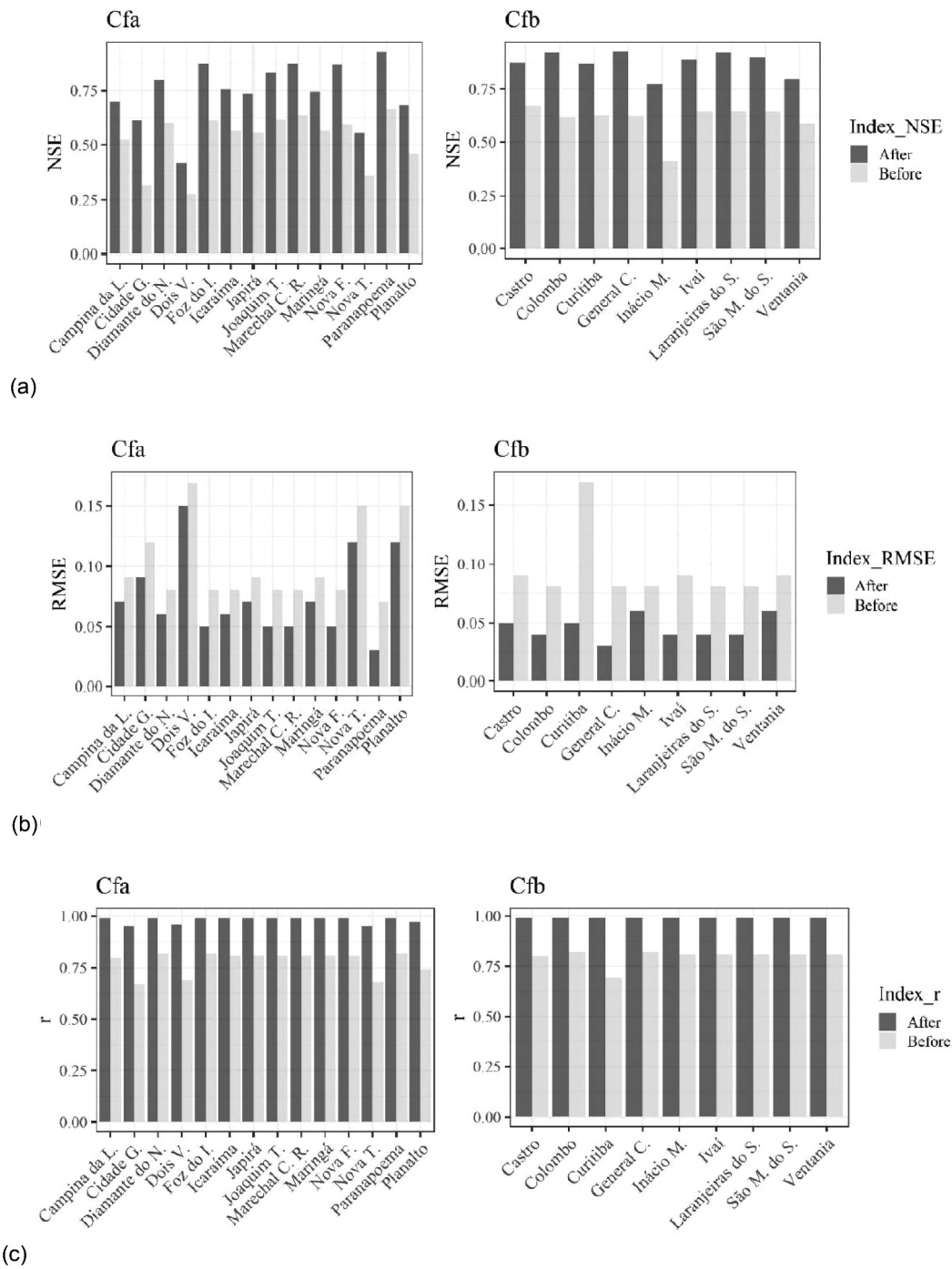


Figure 6. Indexes and errors before and after the correction of the delay observed in the correlation between $ET_{O_{MJS,h}}$ and ET_{O_h} at the hourly scale, of 25 weather stations, 15 and 10 in climate types *Cfa* and *Cfb*, respectively, in Paraná State, for the period 2 December 2017 to 8 November 2018: (a) Nash-Sutcliffe efficiency (*NSE*; dimensionless); (b) root mean square error (*RMSE*; mm day^{-1}); (c) correlation (*r*; dimensionless).

ET_o throughout the day. The correction of the two-hour delay improved the $ET_{O_{MJS,h}}$ estimates in relation to ET_{O_h} for *Cfa* and *Cfb* climates, in Paraná State.

On average, the values of $ET_{O_{MJS,h}}$ and ET_{O_h} were close and well associated statistically in Paraná State. The highest amplitudes and less narrow correlations occurred in the winter season, a period when the *RH* remains high and the values of

T are lower. Summer and spring had equivalent values of $ET_{O_{MJS,h}}$ and ET_{O_h} , with smaller amplitudes and closer correlations.

The Moretti-Jerszurki-Silva alternative approach showed better efficiency in relation to the Hargreaves and Samani, modified parametric and Penman-Monteith temperature models, being the best alternative

Table 4. Average daily values of reference evapotranspiration (ET_0 ; mm day^{-1}) of 24 weather stations in Paraná State, for the period 2 December 2017 to 8 November 2018.

Weather station	ET_0 (mm day^{-1})				
	ASCE-PM	MJS	HS	PET	PMT
<i>Cfa</i> climate					
Campina da Lagoa	2.38	2.14	2.08	2.58	2.10
Cidade Gaúcha	3.22	3.09	2.90	2.10	2.49
Diamante do Norte	2.48	2.21	1.23	1.97	1.36
Dois Vizinhos	4.23	3.77	1.44	4.12	1.49
Foz do Iguaçu	2.43	2.50	1.20	3.20	1.37
Icaraíma	2.27	2.09	1.15	1.89	1.40
Japirá	2.51	2.37	1.22	3.57	1.26
Joaquim Távora	2.49	2.14	1.22	4.55	1.34
Marechal Cândido Rondon	2.46	2.21	1.21	2.91	1.39
Maringá	2.42	2.27	2.14	1.30	2.32
Morretes	2.30	2.40	2.07	4.49	1.86
Nova Fátima	2.40	2.45	2.11	3.58	2.24
Nova Tebas	3.52	3.25	2.71	3.65	2.32
Paranapoema	2.36	2.12	2.11	1.31	2.38
Planalto	3.97	3.76	3.14	3.44	2.55
Average for <i>Cfa</i> climate	2.76	2.58	1.86	2.98	1.86
<i>Cfb</i> climate					
Castro	2.59	2.19	2.22	3.23	1.94
Colombo	2.59	2.80	1.22	3.85	1.94
Curitiba	2.59	2.30	1.23	3.53	1.73
General Carneiro	1.84	1.92	0.97	3.75	1.18
Inácio Martins	2.41	2.67	1.20	4.49	1.20
Ivaí	2.50	2.71	1.21	5.15	1.17
Laranjeiras do Sul	2.44	2.27	1.19	2.24	1.28
São Mateus do Sul	2.38	2.31	2.06	4.47	1.29
Ventania	2.31	2.39	2.08	5.18	1.85
Average for <i>Cfb</i> climate	2.41	2.40	1.49	3.99	1.51

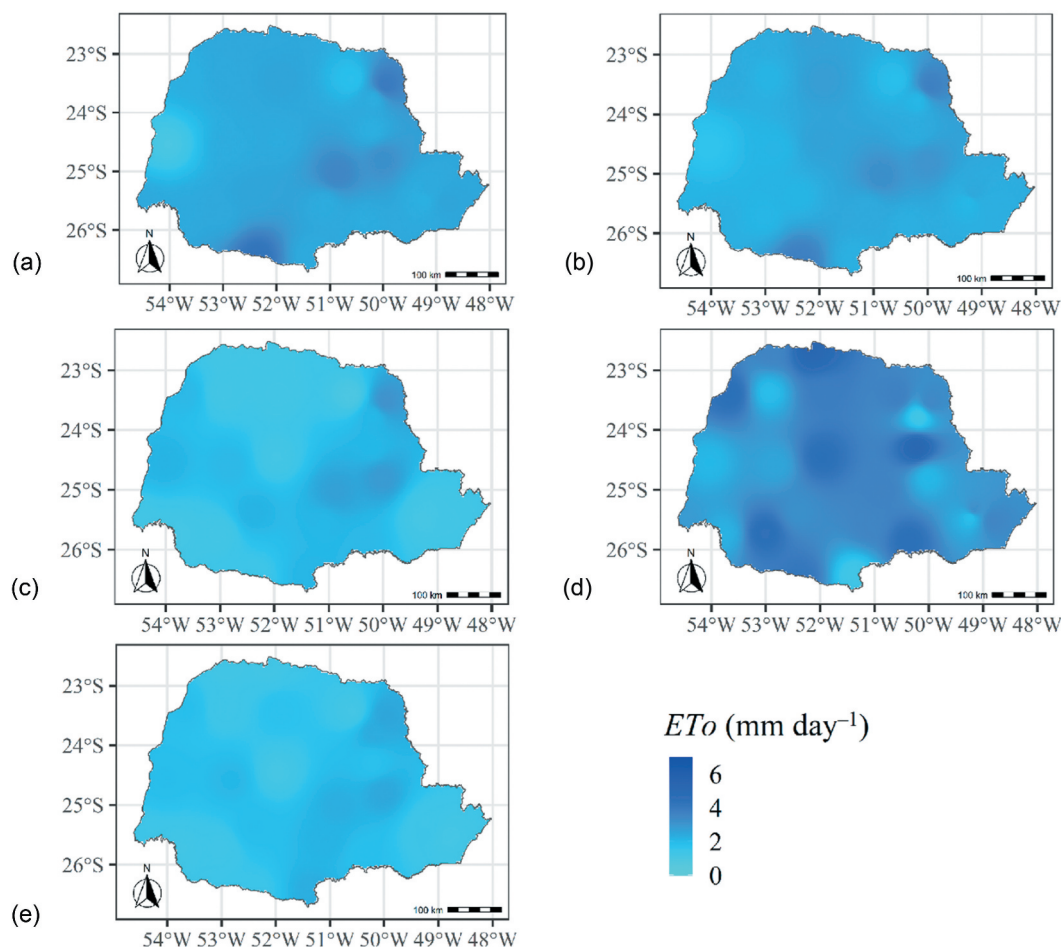

Figure 7. Daily reference evapotranspiration (ET_0 ; mm day^{-1}) of 24 weather stations in Paraná State, obtained with the inverse distance weighting (IDW) method, for the following models: (a) ASCE-PM ($ET_{0,d}$); (b) Moretti-Jerszurki-Silva ($ET_{0,MJS,d}$); (c) Hargreaves and Samani ($ET_{0,HS,d}$); (d) modified parametric ($ET_{0,PET,d}$); and (e) Penman-Monteith temperature ($ET_{0,PMT,d}$).

Table 5. Daily values of Nash-Sutcliffe efficiency (NSE), root mean square error (RMSE) and correlation coefficient (*r*; dimensionless) in the correlations between “*ET*_{MJS,d}” and *ET*_d,” “*ET*_{HS,d} and *ET*_d,” “*ET*_{PET,d} and *ET*_d” and “*ET*_{PMT,d} and *ET*_d” of 15 and nine weather stations in *Cfa* and *Cfb* climate types, respectively, in Paraná State, for the period 2 December 2017 to 8 November 2018.

Weather station	Interaction between the evaluated models (<i>ET</i> ₀ ; mm day ⁻¹)															
	<i>ET</i> _{MJS,d} and <i>ET</i> _d				<i>ET</i> _{HS,d} and <i>ET</i> _d				<i>ET</i> _{PET,d} and <i>ET</i> _d				<i>ET</i> _{PMT,d} and <i>ET</i> _d			
	RMSE	<i>d</i>	<i>r</i>	NSE	RMSE	<i>d</i>	<i>r</i>	NSE	RMSE	<i>d</i>	<i>r</i>	NSE	RMSE	<i>d</i>	<i>r</i>	NSE
	(mm day ⁻¹)	(dimensionless)	(dimensionless)	(dimensionless)	(mm day ⁻¹)	(dimensionless)	(dimensionless)	(dimensionless)	(mm day ⁻¹)	(dimensionless)	(dimensionless)	(dimensionless)	(mm day ⁻¹)	(dimensionless)	(dimensionless)	(dimensionless)
<i>Cfa</i> climate																
Campina da Lagoa	0.64	0.92	0.96	0.50	0.40	0.94	0.97	0.80	0.46	0.91	0.91	0.74	1.23	0.75	0.89	0.27
Cidade Gaúcha	0.78	0.87	0.80	0.62	0.61	0.93	0.92	0.77	1.39	0.71	0.76	-0.20	1.31	0.75	0.86	0.26
Diamante do Norte	0.47	0.92	0.92	0.74	1.38	0.56	0.96	-1.27	0.75	0.74	0.91	0.33	2.63	0.53	0.81	-1.09
Dois Vizinhos	0.77	0.83	0.81	0.46	2.89	0.41	0.88	-6.61	0.71	0.87	0.76	0.53	2.70	0.52	0.78	-1.42
Foz do Iguaçu	0.37	0.95	0.94	0.85	1.36	0.58	0.96	-1.09	0.86	0.80	0.91	0.17	2.40	0.54	0.79	-0.71
Icaraima	0.35	0.92	0.94	0.77	1.20	0.56	0.95	-1.67	0.55	0.81	0.90	0.45	2.78	0.51	0.60	-0.76
Japir	0.40	0.95	0.92	0.81	1.42	0.56	0.96	-1.29	1.14	0.71	0.91	-0.47	2.12	0.53	0.77	-1.06
Joaquim Tvora	0.50	0.91	0.96	0.71	1.40	0.57	0.96	-1.26	2.12	0.53	0.91	-4.20	2.08	0.54	0.78	-1.15
Marechal C. Rondon	0.48	0.92	0.92	0.74	1.38	0.57	0.96	-1.18	0.60	0.87	0.91	0.59	2.45	0.54	0.82	-0.82
Maring	0.45	0.94	0.90	0.78	0.37	0.96	0.98	0.85	1.32	0.56	0.89	-0.88	1.16	0.81	0.90	0.43
Morretes	0.32	0.97	0.94	0.87	0.35	0.96	0.97	0.85	2.24	0.51	0.92	-5.28	0.98	0.78	0.86	0.38
Nova Ftima	0.33	0.96	0.94	0.86	0.38	0.95	0.98	0.82	1.24	0.68	0.91	-0.95	1.35	0.75	0.88	0.36
Nova Tebas	0.44	0.94	0.94	0.79	0.90	0.78	0.94	0.09	0.43	0.94	0.90	0.79	1.54	0.69	0.86	-0.01
Paranapoema	0.43	0.93	0.92	0.76	0.37	0.95	0.97	0.82	1.22	0.57	0.91	-0.95	1.37	0.77	0.9	0.34
Planalto	0.37	0.96	0.95	0.85	0.84	0.83	0.95	0.22	0.61	0.89	0.90	0.59	1.49	0.71	0.87	0.01
Average	0.47	0.93	0.92	0.74	1.02	0.74	0.95	-0.61	1.04	0.74	0.89	-0.58	1.84	0.65	0.82	-0.33
<i>Cfb</i> climate																
Castro	0.78	0.71	0.90	0.39	0.46	0.94	0.97	0.79	0.75	0.58	0.93	0.43	0.79	0.84	0.90	0.56
Colombo	0.57	0.92	0.88	0.74	1.55	0.56	0.96	-0.95	1.36	0.70	0.88	-0.51	1.63	0.55	0.39	0.15
Curitiba	0.48	0.92	0.94	0.76	1.49	0.57	0.97	-1.34	1.01	0.77	0.87	-0.08	1.74	0.56	0.85	-0.74
General Carneiro	0.28	0.98	0.98	0.94	1.37	0.63	0.99	-0.42	1.85	0.69	0.97	-1.58	1.56	0.59	0.89	-0.47
Incio Martins	0.42	0.95	0.94	0.79	1.34	0.58	0.96	-1.06	2.13	0.54	0.91	-4.23	1.67	0.56	0.83	-0.40
Iva	0.43	0.95	0.92	0.80	1.43	0.57	0.96	-1.25	2.71	0.47	0.91	-7.08	1.89	0.55	0.85	-0.95
Laranjeiras do Sul	0.38	0.94	0.93	0.80	1.35	0.57	0.96	-1.50	0.44	0.91	0.92	0.74	1.96	0.56	0.86	-0.72
So Mateus do Sul	0.40	0.95	0.91	0.81	0.42	0.94	0.97	0.79	2.15	0.54	0.92	-4.46	3.26	0.52	0.63	-0.08
Ventania	0.35	0.95	0.93	0.84	0.34	0.96	0.97	0.85	2.92	0.43	0.91	-10.08	1.07	0.76	0.88	0.31
Average	0.45	0.92	0.93	0.76	1.08	0.70	0.97	-0.45	1.70	0.63	0.91	-2.98	1.73	0.61	0.79	-0.26

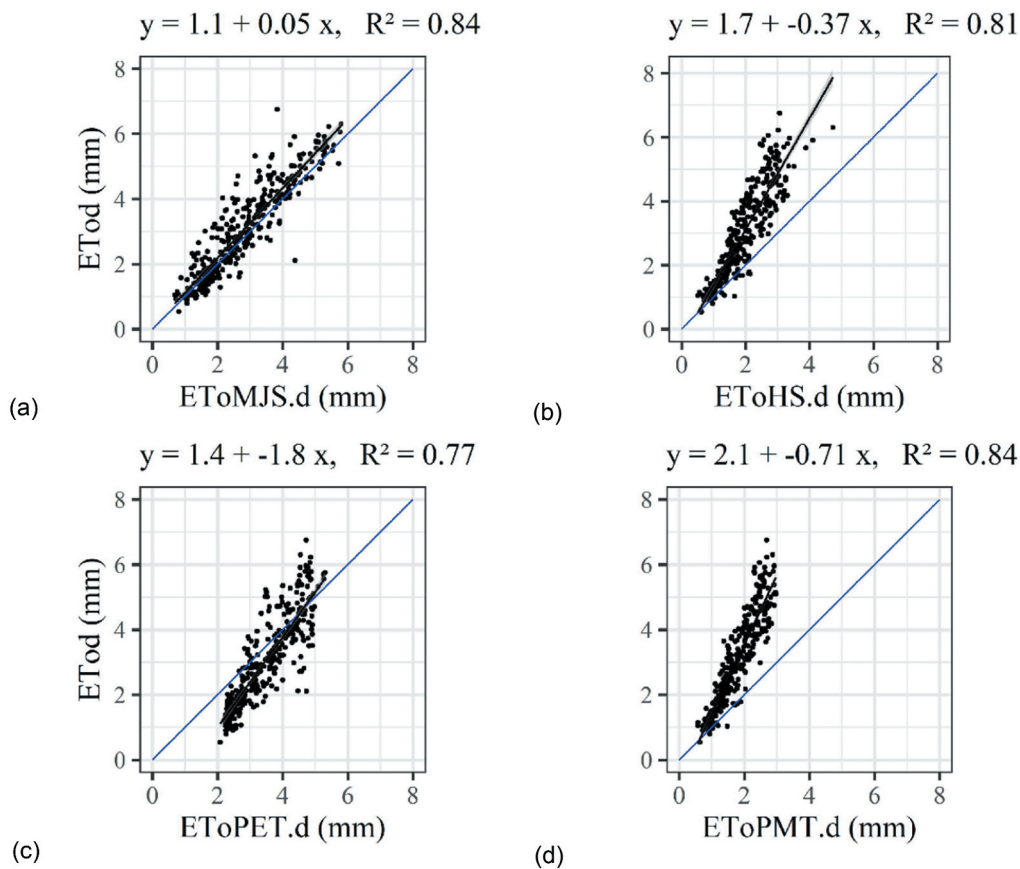


Figure 8. Simple linear regression analysis of the values of daily reference evapotranspiration (*ET*₀; mm day⁻¹) of 24 weather stations in Paraná State, for the period 2 December 2017 to 8 November 2018, for the correlation between: (a) *ET*_{MJS,d} and *ET*_d; (b) *ET*_{HS,d} and *ET*_d; (c) *ET*_{PET,d} and *ET*_d; (d) *ET*_{PMT,d} and *ET*_d.

methodology to produce accurate evapotranspiration estimates at a daily scale, for *Cfa* and *Cfb* climate types, in the sub-tropical region of Brazil.

Disclosure statement

No potential conflict of interest was reported by the author(s).

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