



*Supplement of*

## **Enhanced net CO<sub>2</sub> exchange of a semideciduous forest in the southern Amazon due to diffuse radiation from biomass burning**

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## S1 Supplementary Results

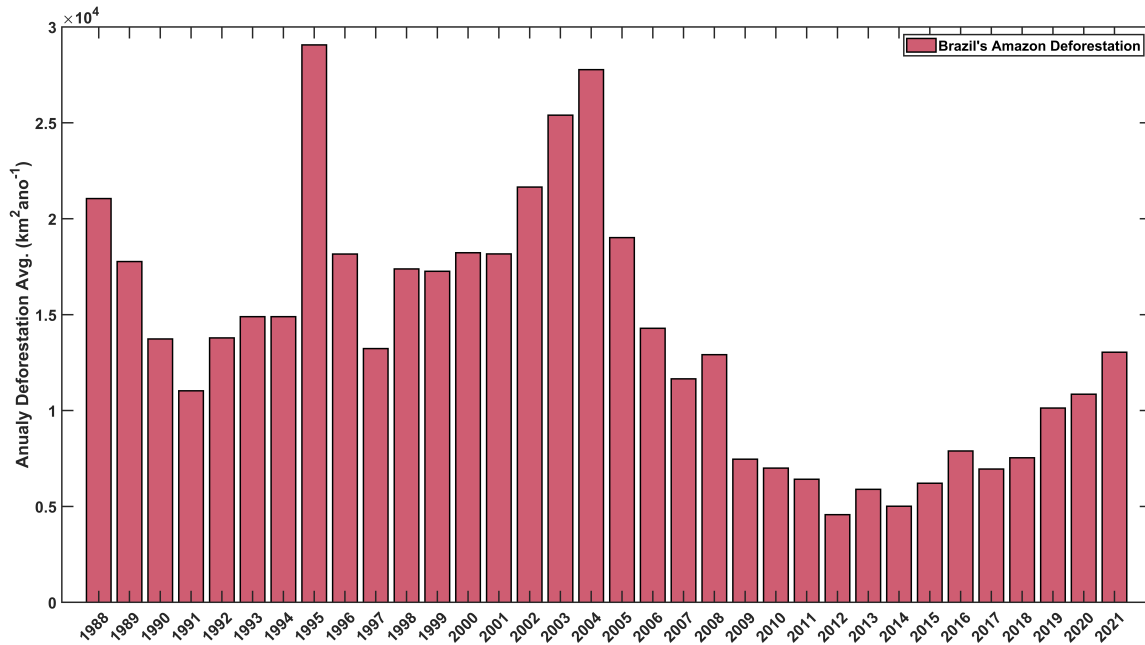


Fig. S1: Time series of clear-cut deforestation in the Legal Amazon (1988 to 2021). Data originated from the PRODES project. The data were collected at the website from The National Institute for Space Research (INPE), link in Table S5.

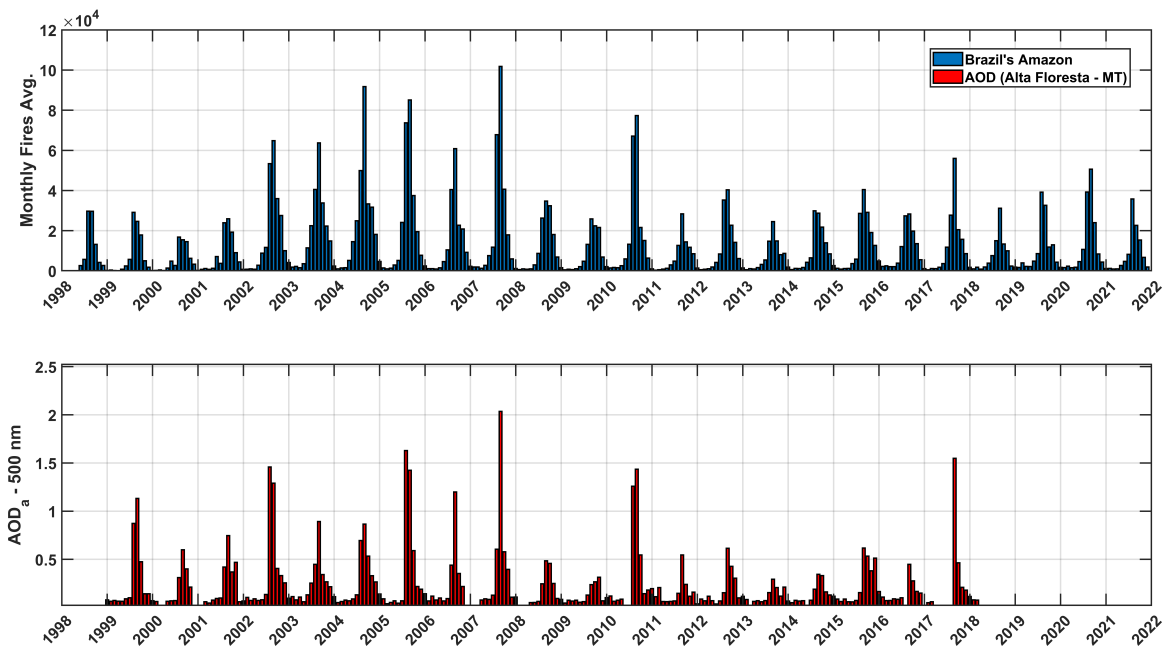


Fig. S2: Shows the time series of fires in the Legal Amazon from 1998 to 2022; data originated from the PRODES project (top panel). At the bottom panel, we show the time series of AOD (500 nm) for the Legal Amazon (1999 to 2017), originating from the AERONET solar photometer (on the ground), remote measurements managed by NASA. All data can be found in the links disposed in the Table S5.

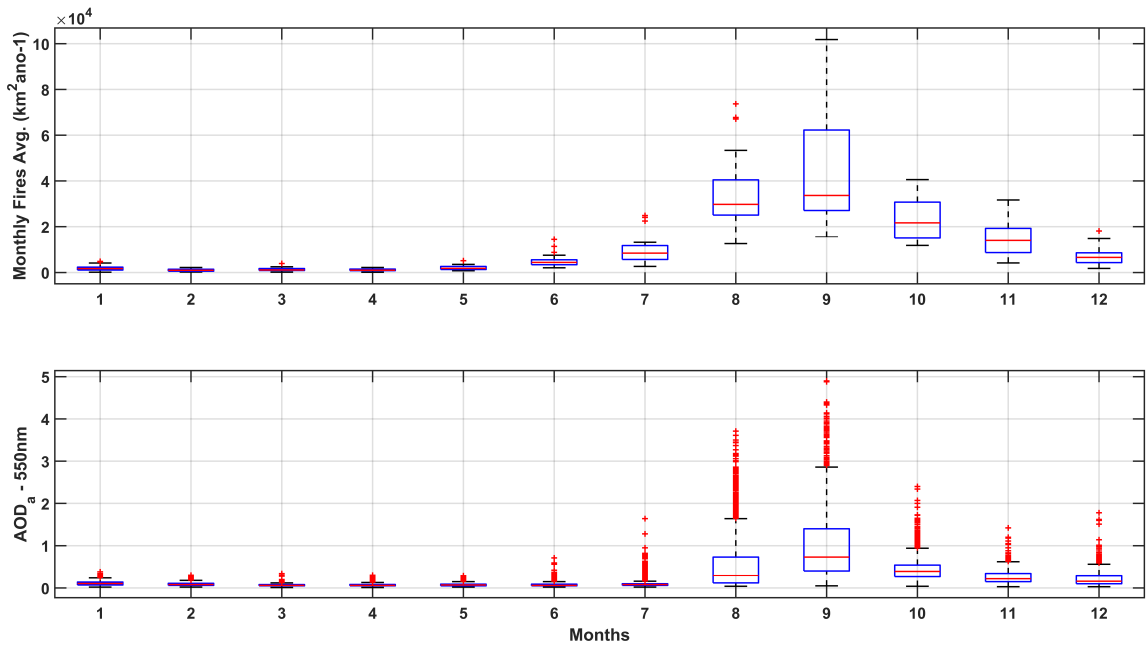


Fig. S3: In the image above, the boxplot shows the statistical behavior of the time series of monthly data on fires that occurred in the Legal Amazon (1998 to 2022). We observed an expressive seasonality from August to November, with a high peak occurrence during September. In the image below, the AOD at 550 nm (AQUA /TERRA) boxplot tracks the fire behavior for the same period. Additional results for fire numbers in the wet season and transitions are presented in Fig. S5.

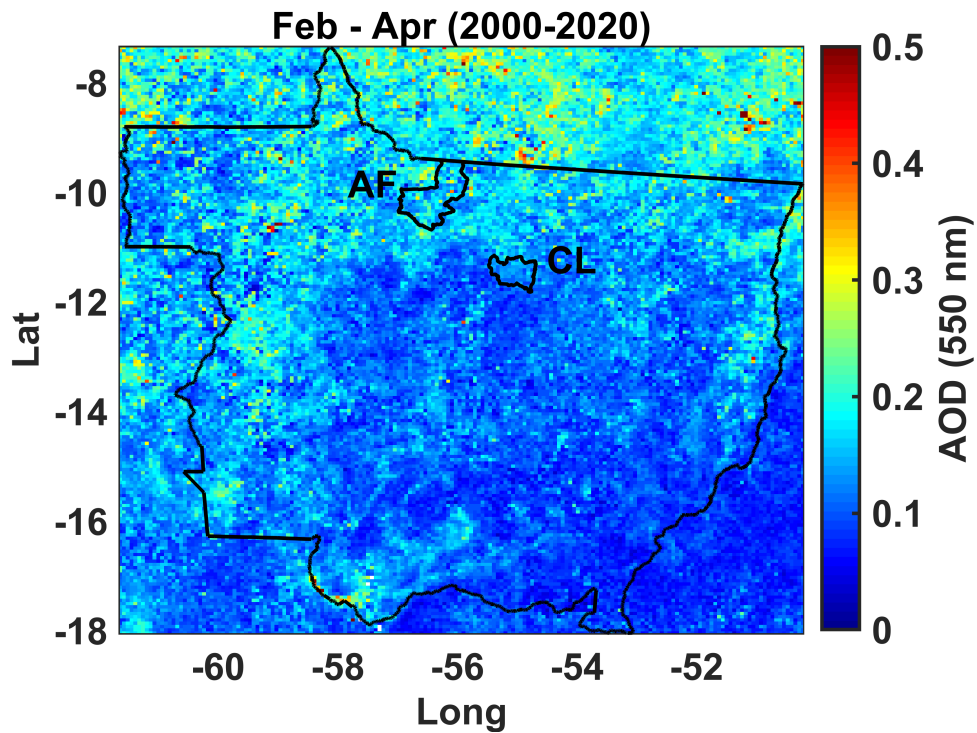


Fig. S4: Spatial behavior of aerosol optical depth (AOD) over the State of Mato Grosso (intense area of the arc of deforestation). In the image above is the behavior of AOD during the wet season (February to April) for the data period from 2000 to 2020, obtained from the AQUA and TERRA satellites.

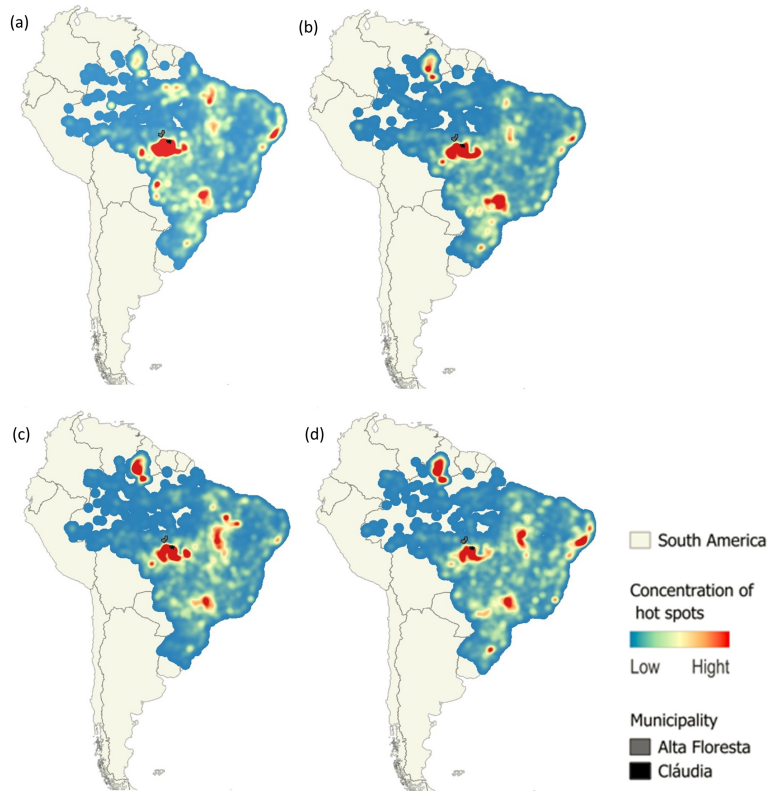


Fig. S5: Accumulated fire focus in Brazil for non-dry months. (a) Jan-Jun/2005, (b) Jan-Jun/2006, (c) Jan-Jun/2007 and (d) Jan-Jun/2008. Reference Satellite: MODIS/Aqua+Terra Thermal Anomalies/Fire locations 1 km FIRMS V0061 NRT (Vector data). Table S4 details the computed fire numbers to the wet season and transitions.

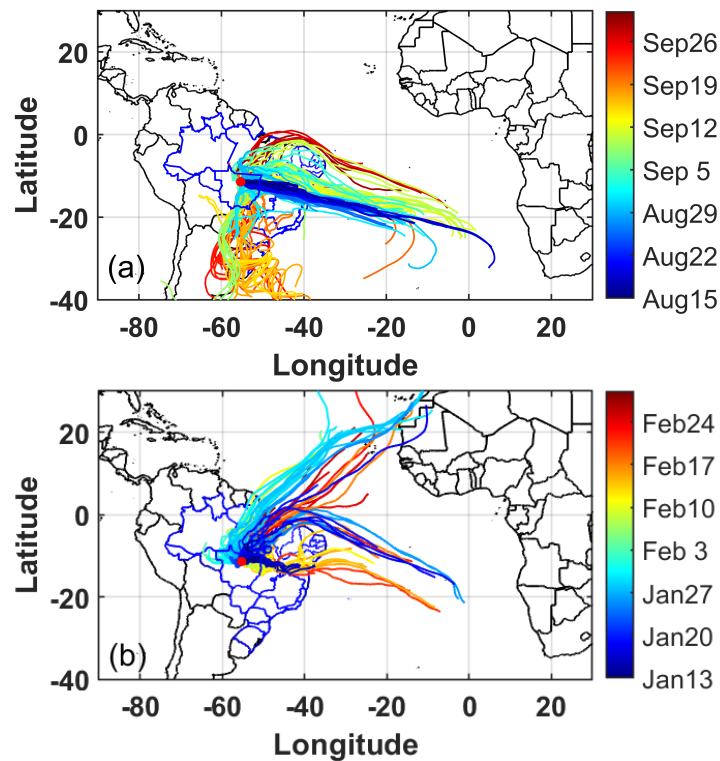


Fig. S6: Simulation of 14-day air mass backtrajectories using the HYSPLIT system during (a) Aug-Sep/2008 and (b) Jan-Feb/2008 (Claudia, Mato Grosso). Vertical air dispersion limited to 100 m.

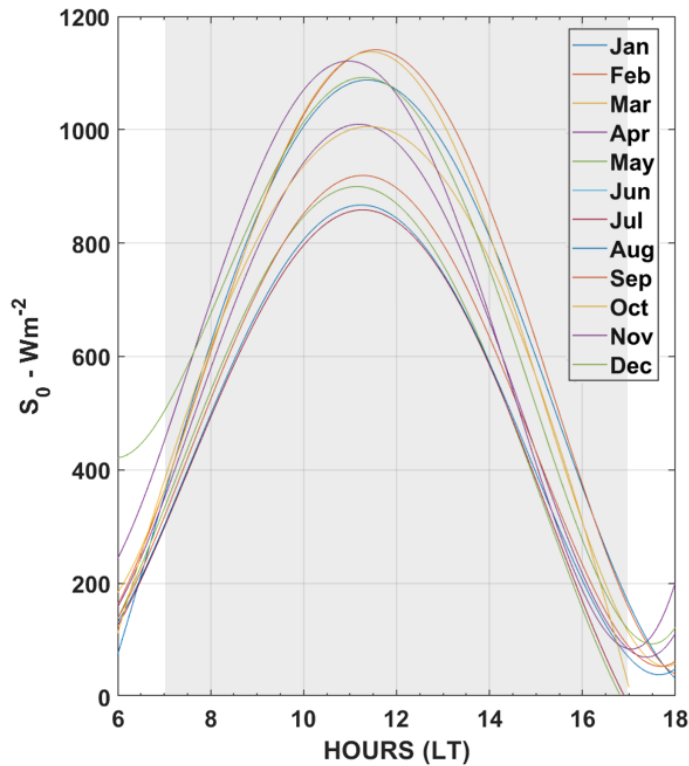


Fig. S7: Curves containing the average hourly behavior (monthly) of incident solar radiation values measured by AERONET in Alta Floresta-MT. The analyzed hourly values are between 07h and 17h (LT) from July 2005 to June 2008. We excluded from the data set all the values before 07h (LT) and after 17h (LT). Only the daylight hours are considered. Curve fits are listed in Table S1.

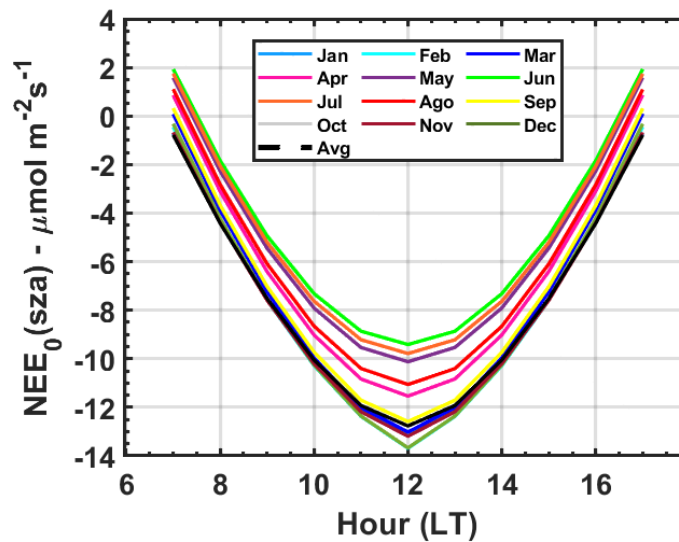


Fig. S8: Shows hourly cycle changes on the  $NEE_0(sza)$ . Fit curves under clear sky conditions ( $f \sim 1.0$ ) for each month of the year between Jul/2005-Jun/2008, 07-17h (LT). Curve fits are listed in Table S2.

## S2 Additional Methods and Analyses

### S2.1 Leaf Canopy Temperature -

Here, we call Leaf Canopy Temperature ( $LC_{Ts}$ ) as a method using the Stefan-Boltzmann equation from the shortwave solar radiation measurements and longwave and albedo estimates as follows, Equation S1:

$$LC_{Ts} = \sqrt[4]{\left(\frac{SWi(1 - \alpha + LWa - Rn)}{\epsilon\sigma}\right)} \quad (S1)$$

Where:  $Rn$  is the net radiation measured in the experimental area ( $Wm^{-2}$ );  $\alpha$  is the mean albedo of the leaf canopy during the dry and wet seasons, respectively equal to 0.079 and 0.126 [Marques et al., 2017];  $LW_a$  is the atmospheric longwave radiation in  $Wm^{-2}$  [Idso and Jackson, 1969], adjusted for the polluted atmospheric conditions of Mato Grosso during the dry season [Von Randow, 2006];  $\epsilon$  is the emissivity, assumed to be 0.98 [Monteith and Unsworth, 1990, Marques et al., 2017] and,  $\sigma$ , the Stefan-Boltzmann constant ( $5.67 \times 10^{-8} Wm^{-2} K^{-4}$ ). A similar method has been widely used, lacking direct leaf temperature measurements in the Amazon (Doughty et al., 2010; Cirino et al., 2014; Andrade et al., 2021).

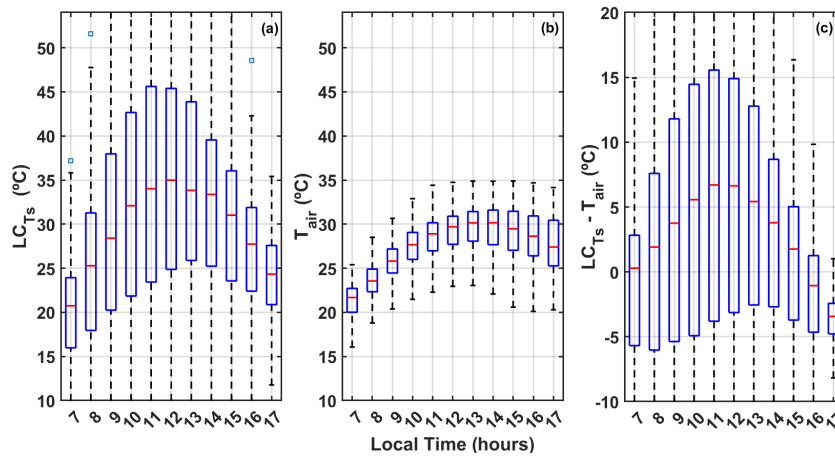


Fig. S9: Shows (a) the hourly cycle of leaf canopy temperature estimated from (Equation 1); (b) the hourly cycle of air temperature; and (c) the difference found between  $LC_{Ts}$  and  $T_{air}$  ( $^{\circ}C$ ), illustrating the warmest leaf canopy during daylight, between 07–17h (LT).

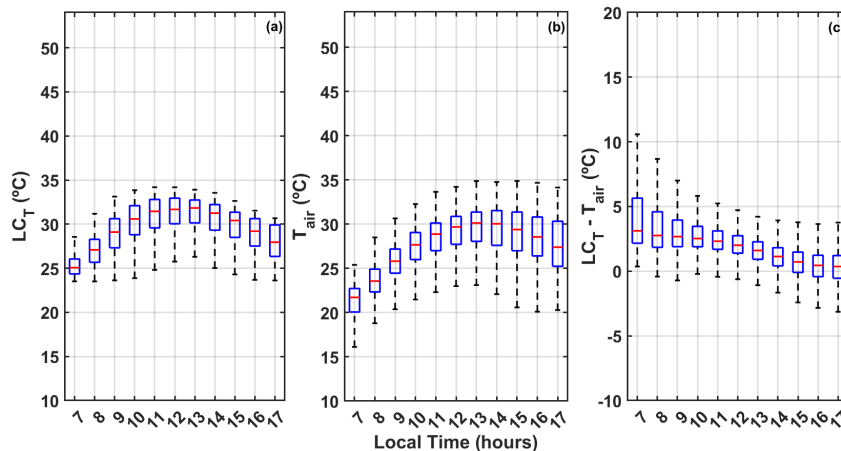


Fig. S10: (a) hourly cycle of leaf canopy temperature estimated by the  $LC_T$  model, Equation 9 (Tribus, 2005); (b) the hourly cycle of air temperature; and (c) the difference found between  $LC_T$  and  $T_{air}$  ( $^{\circ}C$ ), indicating the warmest leaf canopy than the air between 07–17h (LT).

## S2.2 Gross Primary Productivity (*GPP*) -

To compute *GPP*, we used the detailed procedures of [Wutzler et al., 2018] via the REddyProc package. The source code is available via the git revisioning system (Max Planck Institute - BGI). The code takes input from meteorological and flux variables such as  $SW_i$ ,  $T_{air}$ ,  $VPD$ ,  $NEE$ ,  $H$ ,  $LE$ , and  $u^*$  from which are applied filtering methods  $u^* > 0.28$  [Papale et al., 2006, Barr et al., 2013] to correct underestimates at the nighttime  $R_{eco}$ . Moreover, the system estimates the threshold distribution for  $u^*$  to gap filling on the input data and then partitions the fluxes into  $NEE$ ,  $GPP$ , and  $R_{eco}$ . The Look-up Tables – LUT, Mean Diurnal Course – MDC, and Marginal Distribution Sampling – MDS techniques are applied to the partitioning fluxes based on the meteorological conditions and carbon dioxide and water fluxes. REddyProc system takes advantage of these measurements to estimate the uncertainty of the hourly fluxes of our database. From Equation S2, we evaluated the *GPP* and, finally, the  $R_{eco}$ , calculated according to Equation S3. For the partitioning of fluxes, three methods are considered: the Look-Up Table (LUT), Mean Diurnal Course (MDC), and the Marginal Distribution Sampling (MDS) gap-filling method, which is currently the most commonly employed approach [Wutzler et al., 2018].

$$GPP = NEE - R_{eco} \quad (S2)$$

Where: *GPP* and  $R_{eco}$  are the Gross Primary Productivity and the ecosystem respiration ( $\mu\text{mol m}^{-2} \text{s}^{-1}$ ), respectively (Wutzler et al., 2018).

$$R_{eco} = R_{ref} \cdot \exp \left[ E_0 \left( \frac{1}{T_{ref} - T_0} - \frac{1}{T - T_0} \right) \right] \quad (S3)$$

Where:  $R_{eco}$  is given as a function of temperature sensitivity ( $E_0$ ),  $T_0$  is kept constant at  $-46.02$  °C, and the reference temperature  $T_{ref}$  at  $15$  °C. The respiration at the reference temperature ( $R_{ref}$ ) is recalculated based on nighttime data using the annual  $E_0$  temperature sensitivity estimate [Reichstein et al., 2005]. This estimation is performed within 7-day windows that are sequentially shifted over a span of 4 days (Wutzler et al., 2018). Gap-filling hourly *NEE* data is imperative to obtain a complete time series for calculating daily averages or balances, such as monthly or seasonal sums. The REddyProc system implements the following three gap-filling methods.

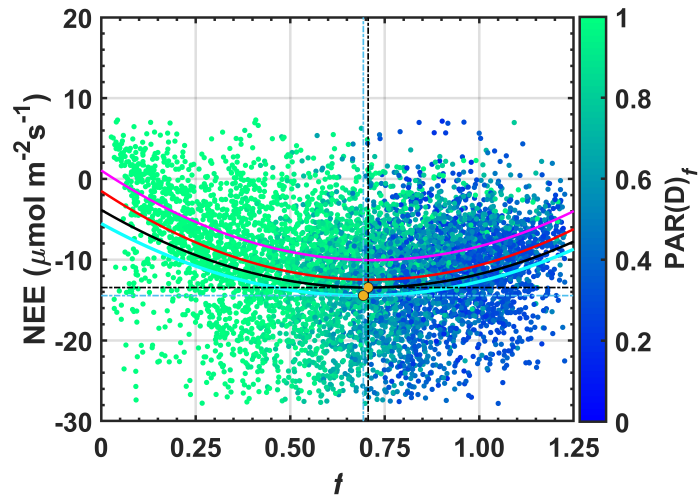


Fig. S11: Variability on the  $NEE$  for various SZA ranges (deg) during all the year (Jan-Dec/2005-2008), at Claudia, Mato Grosso: 0-20 (cyan), 20-40 (black), 40-60 (magenta), and 0-60 (red).

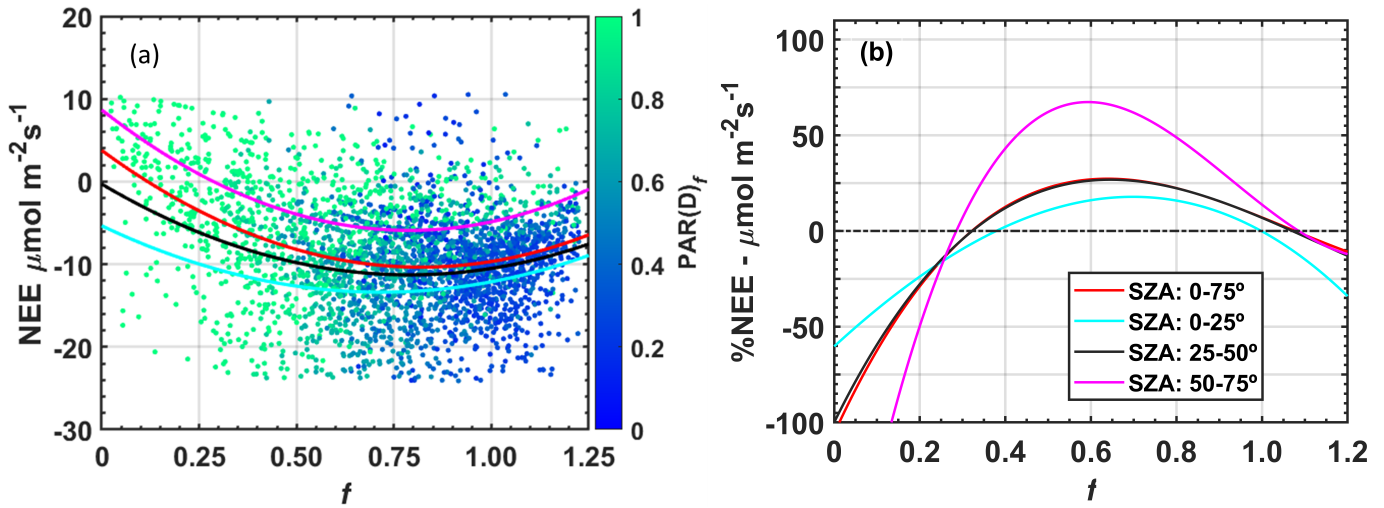


Fig. S12: (a) Variability on the  $NEE$  for various SZA ranges (deg) during the wildfire season (Jul-Nov/2005-2008), at Claudia, Mato Grosso: 0-20 (cyan), 20-40 (black), 40-60 (magenta), and 0-60 (red). Curve fits and tipping points ( $C_p$ ) are listed in Table S3. (b) Shows relative change  $\%NEE$  for the same period (Jul-Nov/2005-2008).



Table S1: Coefficients of the fitted curves of the average hourly (monthly) data of incident solar radiation values measured by AERONET in Alta Floresta-MT. The analyzed hourly values are between 07h and 17h (LT) for the period from July 2005 to June 2008.

<b>4th degree polynomial curves</b>					
<b>Months</b>	$p1$	$p2$	$p3$	$p4$	$p5$
January	+0.38	-17	$+2.4 \times 10^2$	$-1.1 \times 10^3$	$+1.2 \times 10^3$
February	+0.57	-26	$+4.1 \times 10^2$	$-2.4 \times 10^3$	$+4.7 \times 10^3$
March	+0.69	-32	$+4.9 \times 10^2$	$-3.0 \times 10^3$	$+6.2 \times 10^3$
April	+0.64	-29	$+4.3 \times 10^2$	$-2.5 \times 10^3$	$+5.0 \times 10^3$
May	+0.44	-20	$+2.9 \times 10^2$	$-1.6 \times 10^3$	$+2.8 \times 10^3$
June	+0.43	-20	$+3.0 \times 10^2$	$-1.7 \times 10^3$	$+3.1 \times 10^3$
July	+0.49	-22	$+3.3 \times 10^2$	$-1.9 \times 10^3$	$+3.6 \times 10^3$
August	+0.51	-23	$+3.5 \times 10^2$	$-2.0 \times 10^3$	$+3.8 \times 10^3$
September	+0.15	-07	$+8.8 \times 10^1$	$-1.4 \times 10^2$	$-8.7 \times 10^2$
October	+0.77	-34	$+5.1 \times 10^2$	$-2.9 \times 10^3$	$+5.8 \times 10^3$
November	+0.74	-35	$+5.5 \times 10^2$	$-3.5 \times 10^3$	$+8.1 \times 10^3$
December	-0.42	+24	$-5.1 \times 10^2$	$+4.9 \times 10^3$	$-1.6 \times 10^4$

Table S2: Coefficients of the  $NEE_0(asz)$  curves adjusted monthly to clear sky conditions in the municipality of Cláudia-MT between Jun/2005 and Jul/2008.

Months	p1	p2	p3	R <sup>2</sup>	RMSE
January	+0.0014	+0.0826	-13.90	0.6	4.6
February	+0.0054	-0.3876	-04.60	0.4	5.6
March	+0.0011	+0.1145	-14.10	0.3	5.4
April	+0.0033	-0.0370	-13.50	0.5	5.8
May	+0.0057	-0.2150	-10.20	0.4	4.4
June	+0.0091	-0.6277	-01.00	0.2	7.2
July	+0.0039	-0.1793	-08.50	0.1	3.4
August	+0.0058	-0.2657	-08.60	0.4	5.3
September	+0.0018	+0.0045	-11.40	0.3	4.8
October	+0.0015	+0.0762	-14.10	0.4	5.4
November	+0.0026	+0.0527	-14.80	0.6	5.0
December	+0.0117	-0.7177	-03.40	0.8	4.0

Table S3: Polynomial adjustments, coefficients, and statistics between 07-17h (LT) in the micrometeorological tower 50 km from Sinop-MT, in the municipality of Cláudia-MT, between 2005-2008 (Figure S12a).

Settings	Angles	Coeficientes			Statistic
		a	b	c	
Poly fit 2nd	SZA				$C_p(x_v, y_v)$
	0-20°	+19	-26	-5.5	(0.68, -14.39)
NEE of CO <sub>2</sub>	20-40°	+19	-27	-3.8	(0.71, -13.39)
	40-60°	+21	-31	+1.0	(0.73, -10.44)
	0-60°	+22	-31	-1.5	(0.70, -12.42)

Table S4: It lists the number of fire outbreaks recorded by the AQUA-TERRA space platforms during the years between 2005-2008 in the Amazon Basin. Where: 1 – flag to short no-dry-season; 2 – flag long no-dry-season. No-Dry means polluted-wet-season (BBOA), and Rel. Fire is ration or quotient (No-Dry/Dry-Season).

Yrs	Fire Season	Dry Season	No-Dry <sub>1</sub>	No-Dry <sub>2</sub>	Rel. Fire <sub>1</sub>	Rel. Fire <sub>2</sub>
-	Jul-Dec	May-Sep	Jan-Apr & Dec	Jan-Apr & Oct-Dec	-	-
2005	239,814	190,920	16,122	73,074	8%	38%
2006	155,304	117,881	12,710	58,227	11%	49%
2007	239,905	191,645	13,219	71,749	7%	37%
2008	120,156	73,640	10,399	60,853	14%	83%
Total	755,179	574,086	52,450	263,903	9%	46%

Table S5: List with hyperlinks (in blue) and Digital Object Identifiers (DOI) of all data used to figures and tables in this publication.

Open Research - Data Availability Statement				
Secondary Data	Repository	Hyperlink		WebPage
Remote Sensing (Satellites)	Terra-Aqua	<a href="#">Laads Daac</a>	–	NASA Laads Daac
Remote Sensing (Ground)	Aeronet (Lev.2.0)	<a href="#">Goddard-Sfc</a>	–	NASA Goddard
Short Wave Radiation	Aeronet (Lev.1.5)	<a href="#">Goddard-Sfc</a>	–	NASA Goddard
Deforestation and fires	Prodes	<a href="#">Terra-Brasilis</a>	–	INPE Prodes
Weather forecasts Models	Sol-Calculator	<a href="#">Solar-Calculator</a>	–	METEO Exploration
Primary Data	Repository	Hyperlink	DOI	Managers
Eddy Flux (CO <sub>2</sub> )	Mendeley Data	<a href="#">Brazil-EFlux-Stf</a>	<a href="https://doi.org/10.17632/m5h5fw872g.1">10.17632/m5h5fw872g.1</a>	Cirino et al. (2022)
Net Ecosystem Exchange (NEE)	Mendeley Data	<a href="#">Brazil-NFlux-Stf</a>	<a href="https://doi.org/10.17632/m5h5fw872g.1">10.17632/m5h5fw872g.1</a>	Cirino et al. (2022)
Meteorological Data	Mendeley Data	<a href="#">Brazil-AWsta-Stf</a>	<a href="https://doi.org/10.17632/m5h5fw872g.1">10.17632/m5h5fw872g.1</a>	Cirino et al. (2022)

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