Soil CO₂ Dynamics in a Tree Island Soil of the Pantanal: The Role of Soil Water Potential

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Abstract

The Pantanal is a biodiversity hotspot comprised of a mosaic of landforms that differ in vegetative assemblages and flooding dynamics. Tree islands provide refuge for terrestrial fauna during the flooding period and are particularly important to the regional ecosystem structure. Little soil CO₂ research has been conducted in this region. We evaluated soil CO₂ dynamics in relation to primary controlling environmental parameters (soil temperature and soil water). Soil respiration was computed using the gradient method using in situ infrared gas analyzers to directly measure CO₂ concentration within the soil profile. Due to the cost of the sensors and associated equipment, this study was unreplicated. Rather, we focus on the temporal relationships between soil CO₂ efflux and related environmental parameters. Soil CO₂ efflux during the study averaged 3.53 μmol CO₂ m⁻² s⁻¹, and was equivalent to an annual soil respiration of 1220 g C m⁻² y⁻¹. This efflux value, integrated over a year, is comparable to soil C stocks for 0–20 cm. Soil water potential was the measured parameter most strongly associated with soil CO₂ concentrations, with high CO₂ values observed only once soil water potential at the 10 cm depth approached zero. This relationship was exhibited across a spectrum of timescales and was found to be significant at a daily timescale across all seasons using conditional nonparametric spectral Granger causality analysis. Hydrology plays a significant role in controlling CO₂ efflux from the tree island soil, with soil CO₂ dynamics differing by wetting mechanism. During the wet-up period, direct precipitation infiltrates soil from above and results in pulses of CO₂ efflux from soil. The annual flood arrives later, and saturates soil from below. While CO₂ concentrations in soil grew very high under both wetting mechanisms, the change in soil CO₂ efflux was only significant when soils were wet from above.

Introduction

The Pantanal tropical wetland ecosystem is a low-relief landscape situated in the broad depression of central South America. Covering 160,000 km², the Pantanal is among the world’s largest wetlands, and is a major priority for conservation [1]. It is comprised of seasonally-flooded savannas and grasslands, permanently saturated depressions and forested terra-firme topographic rises. Across these diverse landforms, soils are strongly influenced by hydromorphism, with soil profiles most fully developed on the terra-firme rises. Locally known as conílhias, these rises occur on broad paleolevées (e.g. residual depositional riverbanks remaining on the landscape following migration of the river channel), which are important refuges for terrestrial animals during flood periods [2]. Globally, forested topographic rises within wetlands are referred to as “tree islands”, and are recognized as hotspots for both biogeochemistry and biodiversity [3].

Tree islands occur as patches within wetland complexes and have distinctive hydrologic, edaphic and biological functioning relative to their surroundings [4], which increases the ecological complexity in the landscape [3]. Tree islands are a known biogeochemical hotspot in the Florida Everglades [6]. Changes to regional groundwater flow patterns due to construction of canals and levees in the Everglades have resulted in the complete loss of more than 50% of tree islands in recent decades with remaining tree islands suffering degradation with ecological complexity declining as a result [6]. Tree islands in other Neotropical wetlands including the Pantanal are generally less well studied than those of the Everglades.

Biological processes in the Pantanal are strongly moderated by hypersessional environmental conditions, e.g. conditions within an annual cycle that are characterized by two contrasting stressors [7]. Within the Pantanal, this hypersesonality is experienced as a pronounced flooding period with more than 2 m of standing water in many areas [8,9], followed by an extensive dry-season typically lasting six-months. During the resulting low-water period,
spatially-extensive land-use activities are common throughout the Pantanal ecoregion, including cattle grazing and agriculture. Tree island soils are a distinct niche within tropical wetland complexes in that they only briefly experience standing water conditions, if at all. Subsoil saturation of the tree island soil can result from both infiltration of precipitation as well as in response to regional water table dynamics due to floodwaters arriving in the vicinity of tree islands from contributing areas within the wetland’s regional watershed.

Maia et al. [10] found that during the 1970–2002 period, the Pantanal suffered the greatest loss of soil organic carbon (SOC) of any ecoregotype within the Brazilian “Legal Amazon” (e.g. the entire area of all Brazilian states that contain a portion of the Amazon basin, and thus includes parts of ecoregions that are outside of the Amazon basin including the portion of the Pantanal located in the Brazilian state of Mato Grosso which also includes a portion of Amazon forest). This loss was the greatest of any ecoregotype considered in the Maia et al. [10] study, both on a per hectare basis as well as when considered as a rate of change (e.g. 1985–2002 relative to 1970–1985). Comparing the 1985–2002 period to the 1970–1983 period, annual SOC loss in the Pantanal increased from 0.94 Mg C ha\(^{-1}\) yr\(^{-1}\) to 1.16 Mg C ha\(^{-1}\) yr\(^{-1}\). This change in Pantanal SOC stocks is attributed primarily to degradation of the grassland ecosystem related to extensive cattle ranching activities [10], noting that artificial drainage is not practiced in the Pantanal. These dynamics point towards the need for a better understanding of biogeochemical processes in this highly biodiverse environment.

The primary loss pathway for SOC is via decomposition, which results in soil respiration losses of CO\(_2\) to the atmosphere, as well as translocation of terrestrial respiration products to the hydrosphere with percolating soil water [11] and transport of SOC via erosion and deposition [12]. At the global scale, CO\(_2\) emissions from wetlands are lower than emissions from non-wetland soils due to the impact of soil saturation, and global wetland emissions tend to correlate with temperature and not with precipitation [13].

However, little is known about soil respiration within tropical wetland complexes [14,15]. Since temperature regimes in tropical wetlands do not vary substantially over the course of a year, we hypothesized that soil moisture dynamics would play a stronger role than temperature in controlling soil respiration. Since water is involved in both the production of CO\(_2\) within soil, as well as limiting CO\(_2\) and O\(_2\) diffusion within soil and between soil and the atmosphere [16], a dynamic consideration of the relationships involved is needed. Several studies have investigated the dual role that the infiltration front exerts on soil respiration dynamics following rain events plays by increasing CO\(_2\) concentrations while decreasing diffusivity [17,18]. In particular, the gradient method for soil respiration provides the opportunity to explore relationships between soil water dynamics and subsurface soil CO\(_2\) dynamics in conjunction with other soil parameters [19].

In this study, we sought to quantify soil respiration for a terra-firme rise within the Pantanal wetland complex, and to identify the key controls on soil CO\(_2\) efflux and soil CO\(_2\) dynamics within the soil profile. Focusing the study on the topographic rise of the tree island enabled consideration of the full range of soil moisture conditions and flooding dynamics (e.g. as regionally controlled via water table dynamics vs. locally controlled by direct precipitation). In particular, the study was designed to evaluate the roles of and interactions between soil moisture content and soil water potential (e.g. soil tension) on soil CO\(_2\) processes. Relatively few studies to date have explicitly considered the influence that changes in soil water potential can have on soil respiration [20].

### Materials and Methods

#### Site Description

Research was conducted in the northern portion of the Pantanal wetland (56.28\(^{\circ}\)W, 16.57\(^{\circ}\)S) within the long-term ecological research (LTER) station known as SESC-Pantanal near Poconé, Mato Grosso, Brazil (Figure 1). SESC-Pantanal is included in the Ramsar Convention list of Wetlands of International Importance, and is managed by the Brazilian Social Service of Commerce (SESC) [21]. Soils in the region show a high degree of hydromorphism [22]. The soil profile studied in this investigation was classified as a Haplic Planosol in the FAO Classification, with sand and clay contents of 90% and 7% respectively throughout the upper 85 cm, with the clay content increasing to 40% only below 1 m depth. Soils are acidic (pH 4.6 in 1:2.5 slurry of soilwater), with very low (<1%) organic carbon (C) contents. Precipitation averages about 1250 mm yr\(^{-1}\), with an extensive dry season lasting from May through September. Due to the hyperseasonality of the study area, the soil moisture regime for the tree island soil is classified in the U.S. soil taxonomy [23] as Ustic in the dry season and Aquic in the wet season.

The overstory canopy is dominated by Curatella americana L. (known locally as lixeira) and Diperyx alata (known locally as cumbari), with an understory dominated by Scheelea phalerata (Mart.) [9], which is locally referred to as acuri palm. The site has a leaf area index of 3.5 m\(^{2}\) m\(^{-2}\) and a stand density of 1130 trees ha\(^{-1}\), and a basal area of 0.2 m\(^{2}\) ha\(^{-1}\) [9]. Vourliitis et al. [9] measured the soil surface litter pool at 1.25 kg m\(^{-2}\), which is equivalent to 600 g C m\(^{-2}\) assuming a carbon density of 0.48 g C per g dry litterfall [24]. Litterfall is highly seasonal for this semi-deciduous forest type, peaking during the dry season [25].

#### Field measurements

Field measurements were carried out from late November 2008 through March 2010. Environmental sensors were installed within a soil profile in clusters at 10 and 30 cm depths within the soil profile. The location of the soil profile was chosen to characterize the tree island by selecting a central location on the island that was neither a topographic high nor low. The 10 cm and 30 cm sensor clusters were staggered laterally (e.g. offset horizontally) by 1.5 m such that the installation would not impede nor enhance the movement of water, heat or soil gases between measurement depths. Each cluster consisted of sensors to measure temperature, soil water content, soil water potential (e.g. soil tension), and the carbon dioxide concentration of the soil air. Within each cluster, sensors were separated laterally by 10 cm to avoid interferences. Volumetric soil water content was measured using probes that utilize a capacitance/frequency domain approach (model EC-5, Decagon Instruments, Pullman, Washington, USA). Soil water potential was analyzed using dielectric water potential sensors (model MPS-1, Decagon Instruments, Pullman, Washington, USA). Soil CO\(_2\) concentrations were determined \textit{in situ} using infrared gas analyzers (model GMM221, Vaisala Inc., Helsinki, Finland) [26,27].

Soil water parameters (soil water content (\(\Theta\) in cm\(^3\) cm\(^{-3}\)) and soil matrix potential (e.g. soil water potential, \(\Psi\) in kPa)), barometric pressure, logger panel temperature and battery voltage were measured at 30 second intervals with averages recorded every thirty minutes. The data logger (Campbell Scientific model CR-1000, Campbell Scientific International, Logan Utah) controlled the soil water sensor excitation and the on-off cycling of the CO\(_2\) sensors. The CO\(_2\) sensors were powered up for five minutes during each half-hour due to power consumption of 4W per sensor [28]. The first three minutes of each power-on period corre-
sponded to the sensor warm-up period, with readings made each 30 seconds during the subsequent two minutes of the power-on period. The resulting soil CO2 readings were then averaged and recorded on a half-hourly basis.

Barometric pressure and soil temperature measurements (BPS sensor and soil thermister, Apogee Instruments, Logan, Utah, USA) were used to correct soil CO2 readings due to pressure and temperature dependencies of sensor output related to the ideal gas law [28]. Previously published studies [27] presented a temperature correction equation for use with Vaisala CO2 sensors that results in an erroneous temperature correction factor when CO2 concentrations are very high, as is the case in the present study. This is due to the inclusion of raw CO2 concentration as a variable in the third-order polynomial temperature correction equation of Tang et al. [27]. Rather, we employed temperature and pressure correction terms that were presented in the more recent equipment manual [28].

The data logger was connected to a large (85 Ah) capacity 12V DC battery and a 20W solar panel and charge regulator. The solar panel was deployed in a clear-sky area, allowing the system power to be maintained by the solar panel. The large capacity of the battery ensured continuous power to the system during nights and extended cloudy periods of reduced solar radiation. At no point during the study did the battery voltage drop below 12V.

The soil CO2 sensors were calibrated with a measurement range of 0–100,000 ppm CO2 (0–10% CO2), with a maximum reading of 115,000 ppm CO2. This range was selected based on previous experience in the Brazilian Amazon [11]. Measurements exceeding the sensor range were gap-filled using the maximum sensor reading of 115,000 ppm in order to produce a continuous record for time-series analysis. About 5.1% of CO2 observations at 30 cm were gap-filled, which increased the annual flux estimate by 8% compared to the non-gap filled data. At the 10 cm depth, the CO2 readings were always within the measurement range of the sensor.

Calculating soil respiration CO2 flux

CO2 efflux from the soil surface (e.g. soil respiration) was calculated at each 30 minute time step using the flux gradient method [17,27]. This approach is based on Fick's fist law of diffusion, where the CO2 fluxes within the soil profile are calculated at two or more depths as:

$$F_z = -D_z \frac{dC}{dz}$$

(1)

where the $F_z$ is the flux ($\mu$mol m$^{-2}$ s$^{-1}$) at depth $z$ (m) determined based on $dC$, the change in CO2 mole concentration ($\mu$mol m$^{-3}$) over the depth interval $dz$ (m), and the diffusivity of CO2 in soil, $D_z$ (m$^2$ s$^{-1}$), at depth $z$. As the Vaisala sensors produce concentrations as volume fractions ($\mu$mol mol$^{-3}$), the output must be transformed to mole concentrations based on the ideal gas law as described in Tang et al. [27].

Soil respiration ($F_0$) is computed after Vargas et al. (2010) from the fluxes within the soil profile as:
where the soil CO₂ fluxes at depths $z = i$ and $i + 1$ are determined from equation (1) and the soil respiration efflux ($F_0$) corresponds to $z = 0$. In the present study, CO₂ fluxes within the soil profile were calculated for $z_i = 20$ cm and $z_{i+1} = 5$ cm and using CO₂ concentrations at 10 cm and 30 cm for $F_20$ and soil CO₂ at 10 cm and atmospheric CO₂ for $F_a$. Atmospheric CO₂ concentration values ($C_{air}$) were obtained at 2 m above the soil surface using a Li-cor LI-6400XT infrared gas analyzer mounted below the forest canopy on a flux tower located within 1 km of the tree island where the soil CO₂ sensors were installed (M. Biudes, unpublished data).

Soil diffusivity, $D_s$, was computed at each timestep based on an empirically derived relationship between volumetric soil water content ($Θ$) and CO₂ diffusivity in soil determined in the laboratory. For diffusivity measurements, large rings (10 cm diameter and 10 cm height) were used to collect undisturbed soil samples from 5–15 cm ($n = 3$) and 25–35 cm ($n = 2$) corresponding to IRGA installation depths. The samples were brought from the field to the lab at UFMT where soil diffusivity measurements were conducted according to Jassal et al. [29] for water contents ranging from 0.04–0.38 cm$^3$ cm$^{-3}$. Here, a gas mix of CO₂ in air (5% CO₂) was introduced into the base of a chamber that housed an additional Vaisala CO₂ sensor. The chamber base also provided a support for the soil sample. A barometric pressure sensor was placed in the lower chamber to ensure pressure in the chamber remained at atmospheric. The chamber was equipped with an unobstructed exit port to allow free flow of calibration gas out of the chamber while a portion of the calibration gas diffused upwards into the soil sample. The calibration gas was vented from the exit port to an outside window to avoid build up of CO₂ in the laboratory environment.

We used an additional field sample to determine the length of time required to reach steady-state conditions (e.g. constant $F_0$) by making repeated measurements of $F_0$ at the sample surface using a Li-cor soil CO₂ flux chamber coupled to a Li-cor LI-6400XT gas analyzer. Steady-state conditions were achieved after allowing the calibration gas to diffuse for an hour. From there, $D_s$ was calculated after Jassal et al. (2005) as:

$$D_s = L \left( \frac{F_0 + F_L}{2} \right) \frac{1}{C_L - C_0}$$

(3)

where $L$ is the length of the soil sample, $F_0$ is soil respiration at steady-state, $F_L$ is the CO₂ flux at the bottom of the soil sample, and $C_0$ and $C_L$ are the CO₂ concentrations at the top and bottom of the soil sample at steady state, respectively. $C_0$ was measured with the Li-cor LI-6400XT, and $C_L$ was measured with the Vaisala sensor. $F_L$ resolves algebraically to $F_0 - F_w$, where $F_w$ is the flux of CO₂ generated within the soil sample (e.g. the background soil CO₂ production rate). $F_w$ was measured on each soil sample prior to each diffusivity measurement by capping the bottom of the sample and measuring soil efflux with the Li-cor analyzer prior to placing the sample on the diffusivity measurement chamber. $F_w$ is then substituted into equation (3) as $F_L = F_0 - F_w$ [29].

Soil samples were weighed prior to each diffusivity measurement, and soil water content was determined for each measurement based on the difference between measurement weight (mass of wet soil minus ring weight) and the oven-dry soil mass. The latter was determined after concluding diffusivity measurements. For soil collected from 5–15 cm depth (e.g. centered on the CO₂ sensor installed at 10 cm depth), the relationship between soil volumetric water content (VWC, cm$^3$ cm$^{-3}$) and diffusivity of CO₂ in soil (m$^2$ s$^{-1}$) was found to be:

$$D_s = -1.96 \times 10^{-6} [\ln(VWC)] - 1.84 \times 10^{-6}$$

(4)

$$(R = 0.96, n = 9)$$

For the 30 cm depth (e.g. soil samples collected from 25–35 cm), the following relationship was found:

$$D_s = -2.33 \times 10^{-6} [\ln(VWC)] - 2.47 \times 10^{-6}$$

(5)

$$(R = 0.80, n = 6)$$

Soil efflux computed from the gradient method [17] was compared against field-based soil efflux measurements using Li-cor soil CO₂ flux chamber kit (model 6400–19) coupled to a Li-cor LI-6400XT gas analyzer on six different measurement days. There was a significant relationship between the gradient-calculated soil CO₂ efflux and that measured in the field:

$$CO_2\text{-gradient} = 0.95[CO_2\text{-field}] + 0.42 (\text{slope} = 0.95, p < 0.05, R^2 = 0.69)$$

(6)

Statistical analyses

Initial data exploration was conducted via factor analysis in order to identify interdependencies between measured parameters (e.g. common factors), and to identify temporal clusters within the reduced parameter space. Factor analysis techniques have frequently been used successfully with soil data [30]. We applied factor analysis using a varimax rotation that included all measured variables (soil CO₂ concentration, soil temperature, soil moisture content and soil water potential), treating parameters at each depth as independent variables. The varimax method rotates the significant axes of resulting factors orthogonally in order to force the loadings of the original components of each factor to be either as large as possible, or near zero [31]. This procedure has the advantage of simplifying the interpretation of the resulting factors, and has been utilized in research related to soil, water and meteorology [32,33]. The factor analysis was run in SPSS v17 (IBM Corp., New York).

Linear correlations were determined between all measured parameters. For water potential, correlations were determined for both the sensor output in kPa as well as in units of pF calculated as:

$$pF = \log (-10 \times kPa)$$

(7)

where kPa is expressed as tension (e.g. negative values). Soil water potential is often expressed in its logarithmic form (pF) due to the nonlinear relationship between soil water potential and soil moisture content, as well as due to the log-linear relationship between soil water potential and soil respiration [20].

The relationships between soil CO₂ efflux and its principal controlling variables (soil water and soil temperature) were evaluated for the full study period using wavelet coherency analysis [34]. Here, we determined the coherence of the variance between two variables in the frequency domain, analyzing independently for soil CO₂ efflux vs. soil water potential, and
again for soil CO₂ efflux vs. soil temperature. Soil water potential was utilized in the time series analyses as it was measured independently from soil moisture content, and was not used in the calculation of soil CO₂ efflux. Wavelet coherency analysis was conducted in the R software environment for statistical computing and graphics (R Version 2.15.3) [35] using the R package biwavelet (Version 0.13) [36]. For wavelet coherency analysis, we used a total time series length of 270 days, which encompassed 90 days for each of the temporal clusters identified by factor analysis.

We then analyzed each of the temporal clusters independently using conditional nonparametric spectral Granger causality analysis [37,38]. In this approach, causal relationships between the dependent variable (here, soil CO₂ efflux) and independent variables are evaluated in the frequency domain for a multivariate system based on established principles known as Granger causality, or G-causality [39,40,41]. We analyzed soil CO₂ efflux in relation to soil water potential and soil temperature for a 90-day period for each temporal cluster after porting data from R to Matlab using the R package R.matlab [42]. We used the Granger causality Matlab toolbox available for download [37], which we ran in Matlab R2011b.

Briefly, Granger causality is based upon the logical assumption that causes must precede effects. For Granger causality analysis, a series of t-tests and F-tests are employed on lagged time series data to quantify if there is information in the presumed causal variables series that causes must precede effects. For Granger causality analysis, a univariate model of X2 on Y is first evaluated for G-causality in the frequency domain, and is then reevaluated while controlling for the other independent variable (X1). If the G-causality between X1 and Y is significant on X2 (with and without controlling for X1). Complete details on the method have been presented previously in the literature [37,38].

Permissions

No specific permits were required for the described field studies, which took place within the SESC Pantanal Reserve. Research within the reserve is coordinated as part of the Brazilian Long-term Ecological Research (LTER) network by the Federal University of Mato Grosso. The field studies did not involve any endangered or protected species.

Results

Factor analysis and identification of seasons

Factor analysis resulted in two principal factors. Soil water parameters (soil water potential and soil moisture content) and soil CO₂ from both 10 cm and 30 cm depths comprise Factor 1, which explained slightly more than 60% of the total variance in the parameter set (Figure 2). Factor 2 was limited to soil temperature, which explained an additional 17.8% of the total variance. Results from the factor analysis were aggregated and plotted within the two-factor space by month, which resulted in three distinct clusters of data (Figure 2). The largest cluster was found to consist of wet season months, with the other two clusters split between cooler and warmer months during the dry season.

These clusters correspond to the hydrologic periods of the northern Pantanal, which is at maximum flood during February, with lowest water in August [2]. The clusters, while primarily descriptive, were useful for distinguishing seasonal behavior within the system. We also used them to identify episodes of similar magnitude occurring in distinct seasons, which we present in subsequent section of the paper.

A correlation matrix of variables utilized in the factor analysis is presented in Table 1. For CO₂ concentration in the near-surface layer (10 cm), the strongest correlation obtained among other near-surface (10 cm) parameters was for soil water potential, followed by soil moisture (Table 1). At 30 cm depth, CO₂ concentration was most highly correlated with 30 cm measurements of soil moisture followed by soil water potential. Soil temperature was weakly correlated with soil CO₂ concentrations at both depths. The depth-dependent differences in the relationships between soil water potential, soil moisture and soil CO₂ concentrations are explored further in a subsequent section.

Soil CO₂ dynamics

Mean soil CO₂ concentrations during the period of study were 4940 ppm at 10 cm and 27630 ppm at 30 cm. Soil CO₂ concentrations and soil respiration were strongly seasonal (Figure 3), with highest values during the wet season and lowest during the dry season. This broad seasonal trend is overlain by soil CO₂ responses to wetting episodes in both wet and dry seasons, with soil CO₂ at 10 cm (Figure 3c) exhibiting a more dynamic response than soil CO₂ at 30 cm (Figure 3c). Soil efflux during the study averaged 3.53 μmol CO₂ m⁻² s⁻¹. Annual soil respiration was 1220 g C m⁻² y⁻¹. As the study encompassed November 2008 through March 2010, we averaged the values for the days of year that occurred more than once in the time series to avoid seasonal biases (e.g., values for February 2, 2009 and February 2, 2010 were averaged before computing annual means).

In order to evaluate the sensitivity of calculating soil CO₂ efflux via the gradient approach using CO₂ values for Caw that were measured 2 m above the soil surface within 1 km of the location of the soil measurements, we recalculated soil CO₂ efflux using the average global CO₂ concentration during the period of study for Caw (386 ppm) [43]. The resulting value was within 0.7% of the value determined when using the tower-measured CO₂ concentrations. Computing soil CO₂ efflux with the gradient approach using a constant value for Caw introduces minor systematic errors at diurnal and seasonal time scales, although these errors tend to cancel over the course of a year [44]. In general, since soil CO₂ at 10 cm was typically one to two orders of magnitude greater than Caw, variations in Caw have little impact on the steep gradient between soil CO₂ and Caw [44].

Soil water dynamics

Soil moisture and soil water potential also exhibited broad seasonal trends, which were interspersed with episodic wetting events during the dry season, and episodic drying events during the wet season (Figure 3d and e). During the wet season, soils were generally at or near saturation with periodic drying events observed during which soil CO₂ concentrations declined rapidly. During the dry season, soil water was typically at minimal levels for soil water potential and soil moisture content measurements, with periodic wetting events observed that corresponded to rapid increases in soil CO₂ concentrations. Generally, soil CO₂ concentrations were highest following extended periods with soil water potential values near zero (Figure 3).

The relationships between soil water parameters and soil CO₂ concentrations differed for soil moisture as compared to soil water
potential. Soil water potential had a bimodal behavior in this coarse-textured soil, dropping rapidly to very negative values when soils drained, and climbing rapidly to near-zero values following precipitation and infiltration events (Figure 4a and 4c). High CO₂ values were observed only under moist conditions when soil water potential approached zero. Below a threshold soil water potential of approximately $-0.15 \text{kPa}$, changes in soil CO₂ were fairly limited, suggesting that physiological moisture stress of roots or soil microbes were not the dominant drivers of soil CO₂ throughout the year. At the 30 cm depth soil CO₂ increased exponentially with moisture to the saturation point ($0.3 \text{ cm}^3 \text{ cm}^{-2}$). The relationship between CO₂ concentration and soil diffusivity at 30 cm was strongly linear ($r^2 = 0.53$, $p < 0.001$), suggesting that diffusivity restricts CO₂ transport at high soil moisture contents, which leads to transient storage of CO₂ in soil. At the more porous 10 cm depth, soil CO₂ increased with moisture to an intermediate water content ($\approx 0.25 \text{ cm}^3 \text{ cm}^{-2}$), but decreased at higher water contents (Figure 4b). The declining values in soil CO₂ at higher levels of soil moisture suggests diffusive limitations on gas transport that could reduce CO₂ efflux or O₂ influx. We also observed respiration pulses at both 10 cm and 30 cm depths for soils that were initially quite dry (e.g., soil water dynamics in Figure 4a and 4c relative to CO₂ efflux in Figure 3b). This is an example of discrete events leading to dynamics that deviate from the gross seasonal grouping identified in the factor analysis.

**Time series analysis of soil CO₂ efflux relative to soil water potential and soil temperature**

The wavelet coherence spectra between soil CO₂ efflux and soil temperature exhibited significant coherency at the daily timescale across most of the study period (Figure 5a). The coherence between soil CO₂ efflux and soil water potential showed greater
### Table 1. Correlation matrix of measured parameters.

| Soil CO2 efflux (10 cm) | Soil CO2 efflux (30 cm) | Soil CO2 (10 cm) | Soil CO2 (30 cm) | Soil temp. (10 cm) | Soil temp. (30 cm) | Soil water (cm⁻³ cm⁻³) | Soil water (cm⁻³ cm⁻³) | Soil tension (pF, 10 cm) | Soil tension (pF, 30 cm) |
|-------------------------|-------------------------|------------------|------------------|-------------------|-------------------|----------------------|----------------------|-------------------------|-------------------------|
| Soil CO2 efflux (10 cm) | 1                       | 1.03             | 0.36             | 0.14              | 0.26              | 0.57                 | 0.73                 | 0.65                    | 0.79                    |
| Soil CO2 efflux (30 cm) | 1.03                    | 1                | 0.12             | 0.22              | 0.26              | 0.57                 | 0.81                 | 0.81                    | 0.80                    |
| Soil CO2 (10 cm)        | 0.36                    | 1                | 0.27             | 0.12              | 0.81              | 0.22                 | 0.99                 | 0.98                    | 0.98                    |
| Soil CO2 (30 cm)        | 0.14                    | 0.12             | 1                | 0.35              | 0.81              | 0.81                 | 1                    | 1                       | 1                       |
| Soil temp. (10 cm)      | 0.26                    | 0.22             | 0.35             | 1                 | 0.57              | 0.22                 | 0.22                 | 0.57                    | 0.22                    |
| Soil temp. (30 cm)      | 0.26                    | 0.22             | 0.35             | 0.81              | 1                 | 0.57                 | 0.22                 | 0.57                    | 1                       |
| Soil water (cm⁻³ cm⁻³)  | 0.57                    | 0.73             | 0.65             | 0.81              | 0.22              | 1                   | 0.57                 | 0.22                    | 0.57                    |
| Soil water (cm⁻³ cm⁻³)  | 0.73                    | 0.65             | 0.81             | 0.81              | 0.22              | 1                   | 0.81                 | 0.22                    | 0.81                    |
| Soil tension (pF, 10 cm) | 0.65                    | 0.79             | 1                | 1                 | 0.81              | 0.57                 | 1                    | 1                       | 1                       |
| Soil tension (pF, 30 cm) | 0.79                    | 1                | 1                | 1                 | 1                 | 1                   | 1                    | 1                       | 1                       |

Values correspond to Pearson’s correlation coefficient (R). *P < 0.05 for all correlations in the matrix.*

Coupled seasonal dynamics between soil respiration and soil hydrology

The strongly seasonal nature of climatic conditions in the Pantanal results in a significant coupling between carbon and water cycling. At the start of the study period, the soil was dry with low soil CO₂ concentrations and minimal soil respiration. During the wet-up period, the soil quickly reached saturated and near saturated conditions (Figure 3). This resulted in the accumulation of a substantial amount of CO₂, particularly at the 30 cm depth where values in excess of 80,000 ppm were recorded for several weeks. Despite the high concentration at depth, the efflux of CO₂ from the soil remained low due to the high water content and hence low diffusivity preventing movement of CO₂ from lower in the profile to the soil surface.

During the wet-up, there were a few excursions from saturation in the surface soil layer. For these periods when the surface soil was less than saturated, precipitation resulted in CO₂ efflux events due to CO₂ production in the surface soil. Only once the soil CO₂ concentration in the surface soil was sustained at a high level did the soil CO₂ efflux reach maximum (e.g. January 2009 in Figure 3b). Drying conditions observed in the soil water content data (e.g. February 2009 in Figure 3e) resulted in declines in soil CO₂ concentrations and soil CO₂ efflux. However, only once soil water potential showed clear evidence of drying did soil CO₂ efflux exhibit drastic declines. This drying event during the wet season is explored further in the next section.
Episodic behavior of soil respiration

Soil CO\textsubscript{2} concentrations and soil efflux responded rapidly to precipitation events (Figure 7). We selected three time slices to illustrate wetting and drying processes, with one illustration provided for each seasonal cluster described in Figure 2. During the wet season, numerous discrete precipitation events are observed at 10 cm depth (Figure 7 upper left panel). The rapid responses of soil CO\textsubscript{2} and soil respiration were dampened at 30 cm depth. During the dry season, the more discrete nature of wetting events resulted in very pronounced responses in soil CO\textsubscript{2} concentrations at 10 cm and 30 cm, although the response at 30 cm lags that of 10 cm by 24 hours (Figure 7, upper panel of the center column). A period without rainfall during the wet season allowed for consideration of soil CO\textsubscript{2} dynamics during a “drying episode”. Here (Figure 7, upper right panel), soil CO\textsubscript{2} clearly demonstrates diurnal periodicity which is stronger at 30 cm than at 10 cm. Hysteresis patterns in CO\textsubscript{2} versus soil water parameters of the lower two rows of panels in Figure 7 are considered later in the paper.

Discussion

Carbon Fluxes in a Tropical Hyperseasonal Wetland Soils

Soil respiration was calculated as 1220 g C m\textsuperscript{-2} y\textsuperscript{-1} efflux of CO\textsubscript{2} from soil during the study period. These are the first known values of soil respiration for any Pantanal soil. Further, there are no data at present for any non-carbon accumulating (e.g. non-peat) tropical wetland soils in the Soil Respiration Database (SRDB) described in Bond-Lamberty & Thomson [15] (version 20120510a was consulted during the writing of this paper).

Soil organic carbon (SOC) stocks in the upper soil are low, and approximately equivalent in magnitude to the carbon efflux from soils. Of the 1400 g C m\textsuperscript{-2} for the 0–20 cm of soil [45], the upper 10 cm of soil contained 940 g C m\textsuperscript{-2}, with the 10–20 cm depth containing an additional 460 g C m\textsuperscript{-2}, which is consistent with other studies of SOC in the Pantanal [9,46]. In a separate study, the carbon stock in surface litter was found to be about 600 g C m\textsuperscript{-2} during the dry season [9].

Despite the recurrent saturation at or near the soil surface and occasional shallow inundation, the hyperseasonality of the Pantanal inhibits the accumulation of carbon in soils. This non-accumulating nature of soil carbon in the Pantanal directly differs

Figure 3. Precipitation (A), soil respiration (B), soil CO\textsubscript{2} concentrations at 10 cm and 30 cm depths (C), soil water potential (D) and soil water content (E) during the December 2008 -December 2009 study period. Vertical boxes indicate the seasonal periods identified by factor analysis.


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Figure 4. Soil water parameters vs. CO\textsubscript{2} concentrations at 10 cm and 30 cm depths, plotted as daily averages to reduce overplotting.

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from the general characteristics of wetland soils with large carbon stocks and low fluxes, including tropical and semi-tropical wetland systems such as the Florida Everglades and tropical peat wetlands in Indonesia [47].

For much of the study period, the soil moisture is seen to respond to local precipitation, which is expressed in the rapid rise and slow recession in the soil water content time series (Figure 3E). The exception occurred during May 2009 when regional flooding led to a rising water table that saturated the soil from below, resulting in high CO₂ concentrations but low soil CO₂ effluxes. Regional controls on soil moisture via flooding would then be expected to limit turnover of soil carbon, while local controls on soil moisture via precipitation infiltration leads to rapid turnover of soil carbon.

Soil water potential as primary control on soil CO₂ dynamics in the Pantanal tree island soil

Respiration increases following wetting are known as the “Birch effect”, and are well documented in the literature [48,49]. Kim et al. [50] recently surveyed the literature related to soil gas responses to rewetting events, and noted a number of biological and physical mechanisms potentially responsible. Biological priming mechanisms identified include the accumulation of substrate during dry periods for subsequent microbial metabolism, and enhanced root exudation following rewetting that primes microbial metabolism [50]. Physical mechanisms included disruption of soil aggregates, and reduced diffusivity following rewetting [50]. Transient storage of CO₂ also occurs in soil when diffusivity limits soil CO₂ efflux [51,52].

The overwhelming majority of soil respiration studies evaluating soil water controls on CO₂ efflux have evaluated soil water content without independently considering soil water potential [53]. This is largely due to methodological factors, as soil moisture content has long been a simpler measurement relative to soil water potential [20]. While some studies have computed soil water potential from measurements of soil moisture content in conjunction with the soil water characteristic [55], this approach does not permit consideration of hysteresis in the relationship between soil water potential and soil moisture content.

Hysteresis as it refers to soil water behavior is the non-monotonic behavior of the relationship between soil moisture content and soil water potential [56]. That is, the relationship between soil moisture content and soil water potential differs significantly between the wetting and drying phases, which also affects soil respiration. For example, clockwise hysteresis between soil moisture content and soil water potential (Figure 8) was observed during the wetting event that occurred during the dry season depicted in Figure 7 (center column). Soil water content varied by 50% during the event at −20 kPa soil water potential (Figure 8),

Figure 5. Wavelet coherence between soil CO₂ efflux (F₀) and soil temperature (Tₛ) (A, upper panel), and between F₀ and soil water potential (pF) (B, lower panel). The thick black lines outline the periods that are statistically significant at the 0.05 level. The variance in the soil CO₂ efflux and soil temperature time series exhibited statistically significant coherence at the 24-hour timescale, with coherence between soil CO₂ efflux and soil water potential strongest for bi-weekly timescales.

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indicating the importance of independent measurements of the two soil water properties.

Previous studies in temperate zone soils have found hysteresis between soil respiration and soil temperature [57,58]. However, due to the minor variance in soil temperature in the Pantanal study area, soil temperature did not demonstrate any consistent hysteresis patterns with soil respiration. Rather, hysteresis was observed between soil respiration and soil water measurements (Figure 7). For wetting events, soil CO₂ increased rapidly once soil water reached a critical level, and then declined slowly as soil dried.

**Figure 6. Granger-causality spectra between soil water potential (pF) and soil temperature (Tₛ) on soil CO₂ efflux (Fₒ).** Results are presented for each temporal cluster identified in the factor analysis, with 90-day time series used for each cluster.

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(e.g. counter-clockwise hysteresis). The one exception was for soil CO₂ vs. soil water potential at 10 cm during the dry-season wetting episode. There, soil CO₂ built up gradually during the wetting phase and then declined rapidly as the soil dried, leading to clockwise hysteresis. This feature could be related to a priming of microbial respiration [18] that is lagged by root respiration in the upper soil during the dry season [59], which is beyond the scope of the present study.

By considering both soil water potential and soil moisture contents relative to soil CO₂ concentration, we are able to see the influence of soil water on biological limitations to respiration. The saturated and near saturated soil water potential values form a near vertical line on the right hand side of the water potential graphs in Figures 4a and 4c. Variation in soil CO₂ values for wetter soil water potential values can only be appreciated by also considering the soil moisture content graphs in Figures 4b and 4d. That is, for soil water at or near saturation, soil oxygen becomes limited, and declines in aerobic biological activities can be inferred from lower CO₂ concentrations at these higher soil water contents. Whether this decline in aerobic biological activity is accompanied by increased anaerobic activity and increases in methane (CH₄) concentrations requires further investigation.

After the wet-up resulting from direct precipitation on the tree island (with the wetting front moving downwards into the soil profile from the soil surface) followed by drier conditions in late March and early April (Figure 3d and e), there is evidence of wetting of the soil profile from below due to the arrival of floodwaters. The flooding period in the northern Pantanal is characterized by slow movement of large amounts of water draining from the surrounding contributing area, which slowly move down-gradient. These floodwaters arrived during late April and early May of the study period. This can be seen in the shape of the soil water content response at the 10 cm depth, which is much less abrupt than precipitation infiltration events where the soil moisture exhibits a sharp rise. The CO₂ concentrations reached very high levels while soil CO₂ efflux remained rather low due to water filled pore spaces slowing gas transfer from soil to the atmosphere.

During the dry season, precipitation controlled soil water dynamics and soil CO₂ efflux. In late June and early July for example, soil respiration events accompanied changes in the soil water potential at the 10 cm depth. For one of the respiration events, no response was noted in soil water potential at 30 cm depth, with also little response in soil water content at either depth (Figure 3b, d and e). Soil respiration events triggered by precipitation that only caused changes in soil water potential at 10 cm depth were quite brief (e.g. July 16), whereas larger events that affected all soil water measurements lasted several days (e.g. 130 mm on July 23, which is explored greater detail in the following section). Overall, soil respiration patterns were more tightly coupled with soil water potential, particularly at the 10 cm depth. No respiration events were observed without a soil water

Figure 7. Episodic behavior of soil respiration in response to soil water potential and soil moisture content at 10 and 30 cm depths. The arrows in the lower panels indicate the temporal direction of hysteresis loops. doi:10.1371/journal.pone.0064874.g007
soil CO2 efflux (F0) across almost the entire spectrum, and was a wet period, soil water potential (pF) was a significant control on G-causality across a spectrum of timescales. During the warm-wet period, soil respiration is significantly related to soil water potential for timescales throughout the study period (Figure 5b), suggesting that clusters. Soil water potential exhibited strong coherence at longer timescales during the study (Figure 5a), although there were brief episodes of potential response at 10 cm, though numerous CO2 respiration events are observed without any concomitant responses from soil water content.

Time series analysis of soil CO2 efflux relative to primary controls

Wavelet coherency analysis and the conditional nonparametric spectral Granger causality analysis were used to elucidate dynamics of soil CO2 efflux in response to primary controls of soil water potential and soil temperature. Soil temperature was most consistently related to soil CO2 efflux at the daily timescale during the study (Figure 5a), although there were brief episodes of coherence at longer timescales during each of the temporal clusters. Soil water potential exhibited strong coherence at longer timescales throughout the study period (Figure 5b), suggesting that soil respiration is significantly related to soil water potential for days to weeks following precipitation events. However, the Granger causality analysis enables the causal variables to be evaluated in relation to the outcome variable independently, and while controlling for another causal variable.

Performing these analyses for each season allows an explanation of G-causality across a spectrum of timescales. During the warm-wet period, soil water potential (pF) was a significant control on soil CO2 efflux (F0) across almost the entire spectrum, and was a stronger control than soil temperature (Ts) except for at shorter timescales (Figure 6a and 6b). The relationship between F0 and Ts found for longer timescales apparent in the wavelet coherency during the cool-dry period (Figure 5a) also showed up in the Granger causality analysis (Figure 6d). When controlling for Ts, F0 was found to be more strongly G-caused by pF at the daily timescale during the cool-dry and warm-dry temporal clusters, and G-caused at a similar level for the warm wet period (0.1 for pF conditioned by Ts, and 0.13 for Ts conditioned by pF, Figures 6a and 6b, respectively). pF as a control on F0 when conditioned by Ts remained significant at longer timescales in the warm-wet and cool-dry period, and was significant at the daily and shorter timescales during the warm-dry period.

Conclusions

As the Pantanal is located within the Brazilian “agricultural frontier”, further land-use change is likely in the region, including expansion of sugarcane production for biofuels despite the relatively low agricultural potential of the region [21]. Land use practices in other areas of the Pantanal has resulted in major losses of SOC [10]. We found that the annual soil CO2 efflux from a tree-island environment in the Pantanal was approximately equivalent to the C stock in the upper 20 cm of soil, which suggests that SOC stocks are extremely susceptible to loss. We caution, however, that this study was unreplicated.

The frequent wetting/drying cycles result in high rates of soil carbon turnover, contributing to the non-accumulating nature of soil carbon in the Pantanal, which directly differs from the general characteristic of wetland soils with large carbon stocks and low fluxes. Soil water potential was found to be a significant control (e.g. exhibited significant G-causality) on soil CO2 efflux across a spectrum of timescales during the warm-wet period, at daily and longer timescales during the cool-dry period, and at daily and shorter timescales during the warm-dry period.

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Author Contributions

Conceived and designed the experiments: MSJ EGC. Performed the experiments: MSJ OBP MSB. Analyzed the data: MSJ EGC OBP. Contributed reagents/materials/analysis tools: MSJ EGC OBPM RSSA IM MSB. Wrote the paper: MSJ.

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