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Tree Sway Time Series of 7 Amazon Tree Species (July 2015–May 2016)

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1. INTRODUCTION

Trees are a crucial part of ecosystems through their important influence on the water and carbon cycles (Reichstein et al., 2013; Schlesinger and Jasechko, 2014; Patton et al., 2016). They play a key role in hydrological and ecological systems, and land-surface interactions. Ground measurements of tree properties and tree-related fluxes can quantify, for example, tree mass variations, CO₂ uptake, transpiration, rainfall interception and canopy drag, which are crucial parameters for understanding and modeling tree behavior and their roles in ecosystems. Unfortunately, tree measurement sensors are often based on invasive techniques (e.g. sap flow sensors or dendrometers), unable to withstand challenging field conditions (e.g. weathering or power/electronic failure due to climate), or cannot measure all parameters of interest.

Accelerometers offer an economic and robust way for obtaining data series of tree motion, which has been used to infer tree properties and tree-related fluxes. Mounted on a tree trunk, they measure tree sway, which is determined by its mass, elasticity, wood density and drag coefficient. These parameters can in turn be related to water-related mass changes (transpiration and water uptake) (Llamas et al., 2013), biomass-related mass changes (growth, leaf fall and leaf flush) (Selker et al., 2011), tree-atmosphere interactions (drag coefficient and momentum transfer) (van Emmerik et al., 2018a). Additionally, tree sway has shown to be related to tree throw failure mechanisms, and thereby help to reduce storm risks of trees and forests (Flesch and Wilson, 1999a,b; Wilson and Flesch, 1999), by assessing the influence of forest cutblocks.

The relation between tree sway and tree properties and tree-related fluxes can be derived from (1) Newton's second law, and (2) a momentum balance. First, it is often assumed that trees behave like damped harmonic oscillators (Gardiner, 1995; Peltola, 1996). In that case, the tree's natural frequency is linked to mass and elasticity through:

$$\omega_0 = 2\pi f_0 = \sqrt{\frac{k}{m}} \quad (1)$$

With natural frequency in [rad/s] ω_0 , natural frequency in [Hz] f_0 , mass m and elasticity k . From the frequency spectrum of tree acceleration the natural frequency can be determined. Second, following the momentum balance the frequency spectrum of tree sway $P_y(f)$ as a function of frequency can be expressed as (Ammann, 1985; Mayer, 1987):

$$P_y(f) = H_m(f)^2 \rho_a^2 C_d^2 u^2 H_a(f) P_u(f) \quad (2)$$

With mechanical transfer function $H_m(f)$, air density ρ_a , drag coefficient C_d , wind speed u , aerodynamic transfer function $H_a(f)$ and wind load frequency spectrum $P_u(f)$.

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These relations have been used in previous tree-related research. Moore and Maguire (2004) showed that factors such as branch removal and snow loading influences tree natural frequency. Mayer (1987) used tree sway data to demonstrate that the primary sway (i.e., the system as a whole) is related to tree throw. Flesch and Wilson (1999a,b) and Wilson and Flesch (1999) used tilt sensors to analyze tree sway, and assess the influence of management techniques on the possible reduction on wind throw. Later, Moore and Maguire (2005) empirically found a relationship between tree height and tree natural frequency. These, and many more, studies already demonstrated the use of tree sway data. In particular the derived estimation of tree natural frequency is useful for risk assessment of wind damage/throw, and quantification of tree properties (e.g., mass, elasticity) and responses (e.g., intercepted precipitation, mass changed in response to water stress).

Data in previous studies were mainly obtained for several hours, or in rare cases days. Our dataset presents a unique opportunity to use long-term tree sway data of 19 individual trees. This allows analysis of diurnal and seasonal changes, as well as comparison between individuals of the same species, and between different species as a whole. In this paper we describe the details of the obtained dataset, and discuss current and potential applications of the dataset.

2. DATA COLLECTION

2.1. Measurement Location

All data were collected around the K34 observatory (130 m above Mean Sea Level) in the Amazon rainforest (2.6085°S, 60.2093°W), 60km Northwest of Manaus, Brazil (see van Emmerik et al., 2017). A map of the measurement location, including the measured trees and the K34 observatory is presented in **Figure 1A**. The study area is characterized by a wet tropical climate. Additional meteorological, hydrological and plant physiological data may be obtained from the K34 observatory, managed by the National Institute of Amazonian Research (INPA).

2.2. Plant Material

Accelerometers were mounted on 19 individual trees, covering seven different species with one to four individuals per species. The trees were selected to cover a broad range of estimated wood density, tree height, and diameter at breast height (DBH). **Table 1** presents an overview of the measured trees. Tree species were classified by a taxonomist on site. Wood density was estimated using the Global Wood Density Database (Chave et al., 2009; Zanne et al., 2009). Tree height was measured using measurement tape during installation of the accelerometers.

2.3. Sensor Description

All accelerometers used were of the type Acceleration Logger Model AL100 (Oregon Research Electronics, Tangent, OR, USA), which is specifically designed to be physically robust and water proof. Robustness is achieved through a casing that prevents water and organisms to reach the sensor. The sensor has internal batteries and data storage, so no (wire) connections are required

during measurements. Consequently, its dimensions (14.5 × 9.2 × 5.5 cm, 0.4 kg) are larger than other available accelerometers. The accelerometers can measure acceleration in three axis with a frequency up to 25 Hz. Depending on the sampling rate and environmental conditions, it can log for several months on two C-size cell batteries. Data are stored on regular microSD cards. For example, with a 8GB data card it can store up to 320 days of data measured with a frequency of 10 Hz. Data are written to a newly created file each day to minimize data loss in case of empty batteries, full data storage, or other potential failures. The casing has space to include several silica bags to absorb moisture when installed in humid environments.

2.4. Measurement Setup

The accelerometers were placed on the tree trunk, right below the point of main branching. This guaranteed the largest excitation of the sensor, and therefore largest acceleration values, that were still representative for the whole tree. Installing accelerometers on the primary or secondary branches comes with the risk of mainly measuring higher order effects. The accelerometers were mounted using a spring around the trunk. This allows some expansion of the tree when growing, and also assures a rigid attachment to the tree. All measurements were done with a sampling frequency of 10 Hz.

3. DATA DESCRIPTION

The available dataset includes raw acceleration values for all three dimensions as measured by the sensor with a frequency of 10 Hz. In **Table 1** it can be found which accelerometer (serial numbers SN1001-SN1019) was mounted on which specific tree. For each sensor, data are available per day. This reduces computational needs, as each daily file can be read and processed individually.

As most applications require transformation of the data from the time to frequency domain, an additional dataset of the acceleration data in the frequency domain can be downloaded. For each sensor, files are available that contain time series of the acceleration in the frequency domain. These were produced by applying a Fast Fourier Transform (FFT) on 30 min of horizontal (y-axis) acceleration with a moving window of 10 min. **Figure 1B** illustrates a typical frequency spectrum estimated using tree acceleration. It can be seen that the natural frequency around 0.2 Hz can be estimated well. In this example, even higher order resonance peaks are visible.

When applying the FFT to the whole dataset, a time series of the frequency spectrum is created **Figures 1C–E** show a typical time series of the frequency spectrum for two trees. It can be seen that when wind speed is higher than a certain threshold, the resonance peaks are activated. It can also be seen that there are slight differences in timing and magnitude of the frequency spectra, indicating that tree specific information can be deduced. Note that this figure only shows a subset of the total available data. The complete measurement period stretches from July 2015 to May 2016.

For all applications so far (van Emmerik et al., 2017, 2018a), data have been processed using a FFT. We show that using a simple FFT the natural frequency can already be determined.

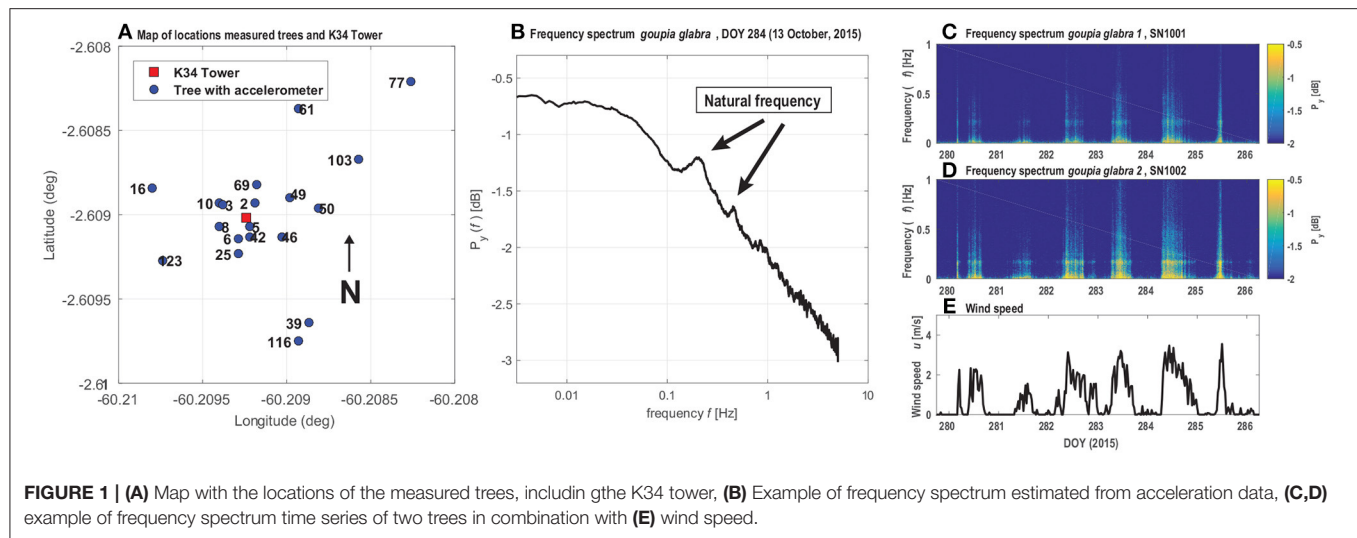


TABLE 1 | List of measured tree individuals, including acceleration sensor number, estimated wood density, estimated tree height and diameter at breast height (DBH).

Accelerometer	Species	Estimated wood density		Estimated height [m]	DBH [cm]
		[10 ³ kg/m ³]	High - Low		
SN1001	<i>Goupia glabra</i>	0.7	High	25	242.5
SN1002	<i>Goupia glabra</i>	0.7	High	32	208
SN1003	<i>Goupia glabra</i>	0.7	High	32	135
SN1004	<i>Lecythis prancei</i>	0.875	Intermediate	24	112
SN1005	<i>Lecythis prancei</i>	0.875	Intermediate	35	116.5
SN1006	<i>Lecythis prancei</i>	0.875	Intermediate	24	108.4
SN1007	<i>Scleronema micranthum</i>	0.5 - 0.7	Low	38	127
SN1008	<i>Scleronema micranthum</i>	0.5 - 0.7	Low	28	189.5
SN1019	<i>Scleronema micranthum</i>	0.5 - 0.7	Low	26	81
SN1010	<i>Eschweilera</i> sp.	0.8	Intermediate	20	93.5
SN1009	<i>Eschweilera coriacea</i>	0.8	Intermediate	21	268
SN1011	<i>Eschweilera coriacea</i>	0.8	Intermediate	18	92.4
SN1012	<i>Eschweilera coriacea</i>	0.8	Intermediate	27	213.5
SN1013	<i>Dipteryx odorata</i>	1.1	High	35	177
SN1014	<i>Dipteryx odorata</i>	1.1	High	32	219.5
SN1015	<i>Pouteria anomala</i>	0.7	Low	22	111
SN1016	<i>Maquira sclerophylla</i>	0.5	Low	30	264
SN1017	<i>Maquira scherophylla</i>	0.5	Low	18	201
SN1018	<i>Maquira scherophylla</i>	0.5	Low	35	90.6
SN1019	<i>Pouteria</i> sp.	0.7	Low	23	117.5

However, we also encourage further exploration with other spectrum estimation methods, such as Welch's method or the MUSIC algorithm.

3.1. Data Availability

All acceleration data are available on the 4.TU Repository (doi: 10.4121/uuid:c9974180-aa9b-40b4-8dbb06d5b1fce693) (van Emmerik et al., 2018b). All files are named “/YYYY/SN10XX/ALOGzzz.csv,” with year YYYY, sensor number XX and day after start of logging zzz. Data are available from July 2015 to May 2016. We invite the community to explore the dataset and combine it with for example higher resolution wind data, hydro-meteorological measurements, tree physiological data and remote sensing observations to explore

potential additional applications of long-term tree acceleration data.

4. EXAMPLES OF APPLICATIONS

4.1. Intercepted Precipitation

During rainfall events precipitation is intercepted by leaves of a tree. This increases the total tree mass. Selker et al. (2011) compared the natural frequency of a tree before and after a rainfall event and found a shift in frequency. A more detailed study by van Emmerik et al. (2017) found different relations between intercepted rainfall and shift in frequency of trees. For some trees, an almost linear relation was found between the

frequency shift and precipitation. For other trees, the relation was less obvious. This was hypothesized to be caused by (1) the location of the tree in the canopy (understory trees are less likely to intercept rainfall) and (2) by splashing of water drops during high intensity rainfall events. Accelerometers were demonstrated to be a promising new method for interception measurements. Current measurement techniques mainly focus on measuring throughfall, which requires a more extensive measurement setup with multiple rainfall sensors. However, in future measurements additional throughfall measurements might give additional insights in the accuracy of interception estimations using accelerometers. Moore and Maguire (2005) also demonstrated that tree sway measurements can be used to monitor snow cover of trees. This is not applicable to the dataset presented in this paper, but might be of interest to researcher who focus on snowy ecosystems.

4.2. Tree Mass Variations

Besides through rainfall and snow cover, tree mass also varies through other mechanisms, such as tree growth, changes in tree water content, or leaf fall and flush. A first analysis of the data presented in this paper, van Emmerik et al. (2017) demonstrated that, for most tree species, a relation can be found between aboveground tree biomass and natural frequency. Selker et al. (2011) presented a proof of concept that there is a difference in tree natural frequency when comparing the same tree with and without leaves. More recently, Jackson et al. (in review) demonstrated that there is a significant difference between trees during summer (with leaves) and winter (no leaves). van Emmerik (2017) and van Emmerik et al. (2018a) related changes in tree sway to water stress induced mass changes, which was hypothesized to be caused by decreasing tree water content and/or leaf fall. Future efforts might focus on using tree sway data to quantify tree mass changes on even lower time scales. With the current dataset, it might also possible to track diurnal changes in natural frequency, which could be related to changes in water content. Additional measurements of leaf water potential would allow further study of the relation between tree sway and tree mass variations.

4.3. Drag Coefficient and Roughness Length

Tree drag coefficient is a crucial parameter for the momentum transfer from the atmosphere to the trees. Accurate estimates of tree and canopy drag coefficient might allow better representation of the canopy in, for example, land-surface models. An approximation of tree drag coefficient was presented by van Emmerik (2017) and van Emmerik et al. (2018a), based on the assumption that during turbulent conditions, available wind energy can be estimated using Kolmogorov's theory (Kolmogorov, 1991). If actual high frequency wind speed data is available, drag coefficients can be determined under all wind conditions using equation 2. Here it was demonstrated that aggregating tree sway data over different time scales (weekly, monthly, complete monitoring period), can give different degrees of insights. When aggregated over the complete monitoring period, a clear relation between tree sway and tree mass was found. The weekly analysis in turn demonstrated shorter term

mass variations in the individual trees. A similar approach might also allow for estimating the roughness length of a forest canopy if data from multiple sensors are combined, as the variation and evolution of individual tree drag and tree-atmosphere momentum transfer can be quantified.

4.4. Tree Damage

Using relatively short data records, it was already shown that tree sway is related to wind throw (Mayer, 1987). Longer-term records such as those presented in this paper might allow further analysis of tree failing mechanisms. Additionally, relatively unmeasured phenomena such as tree dormancy and mortality might be further investigated.

5. LIMITATIONS

Besides the many potential applications of tree sway data, fundamental investigations of the role of meteorological variables (e.g., temperature, humidity, wind speed and direction, precipitation) and tree properties (e.g., mass, elasticity) on tree sway is also crucial. Many applications assume tree sway can be described as mass-spring system, or through the momentum balance in Equation (2). However, a better fundamental understanding of tree sway might open doors to even more potential applications.

As applications of long-term tree sway data are still limited, further assessments are still very dependent on the availability of other data. With the current auxiliary data we hypothesized several relations between tree sway and other processes. However, more detailed data on e.g., leaf potential, wind speed, water uptake and intercepted precipitation are required to further test our hypotheses.

One of the key assumptions of the tree selection is that the chosen trees are free standing. If they are in contact with other trees, the mechanical system is more complex and the standard approach might not be applicable. The trees in this study were selected based on visual inspection, which might have been inaccurate. Also growth of the trees might have altered the position in respect to other trees.

The optimal installation location might also be different for other tree species. The selected trees were all relatively straight, until the point of branching off. For tree species with more complex architecture and geometry, the optimal location should be reassessed.

The presented dataset only includes tree species in the Brazilian Amazon. For tree species in other climatic regions, different relations might be found between tree sway and e.g., mass (variations), tree-atmosphere interactions and intercepted precipitation. We encourage deployment in other climatic regions to further explore the possibilities of tree sway measurements.

6. CONCLUDING REMARKS

Tree acceleration data have been proven to give new insights in assessing risk of tree failure, intercepted rainfall, mass variations, and tree-atmosphere interactions. To date, only relatively short datasets have been used and published (hours

to days). Our dataset covers tree acceleration for a period of 10 months, including changing seasons, for 19 Amazon trees. We demonstrate that these data offer new opportunities of analyzing tree properties, and tree behavior in response to changing environmental conditions. We demonstrate various current applications. Yet, many potential new applications are to be explored. We encourage the scientific community to use our data to do so.

AUTHOR CONTRIBUTIONS

TvE, SS-D, MG, and RH designed the study, TvE, MG, and RO installed the sensors, all authors contributed to data analysis and writing this manuscript.

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