Technical note: An innovative monitoring approach to measure spatio-temporal throughfall patterns in forests

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Abstract. Throughfall in forests is spatially highly heterogeneous creating distinct patterns that persist over time and propagate ~~with infiltration~~ into the soil. Despite its importance for forest ecohydrological processes, experimentally derived high-quality datasets describing spatio-temporal throughfall dynamics at fine temporal and spatial resolution are still scarce. The majority of studies were unable to measure throughfall at high temporal and/or spatial resolution because of extensive sampling efforts, especially in forests with complex structures. We present a nove~~w~~l, innovative and modular throughfall monitoring ~~approach~~system for continuous, automated measurement of throughfall either as isolated canopy throughfall and as integrated throughfall (total throughfall reduced by litter interception). Without removing the water~~ for infiltration that~~, the system allows to quantify the spatio-temporal throughfall variability at both intra-event and intra-stand levels. The network captures spatial throughfall patterns and their temporal persistence across rainfall events of varying size during leafed and non-leafed periods. The throughfall monitoring network features 60 self-built, cost effective throughfall samplers, with four throughfall collection compartments and tipping bucket units each connected to a newly developed microcontroller board enabling fully automated, low-maintenance operation during rainfall events. The network, collecting data since the winter of 2024/2025, is setup in a stratified sampling pattern among four forest plots of Beech, Douglas fir, Silver fir, and mixed trees in a mature temperate forest in Germany. ~~Throughfall~~ Data from a four-week observation period in the spring of 2025 are included in this study to showcase the potential of this approach. The data support the networks' ability to capture small-range spatio-temporal throughfall patterns across the study area.

1. Introduction

Rainfall reaching the tree canopy of a forest partitions into interception, stemflow and throughfall. ~~Throughfall (TF) is the largest water input fraction in most forest ecosystems exhibiting high spatial heterogeneity due to redistribution within the tree canopy (Levia et al., 2011). The emerging spatio-temporal patterns may persist over time and determine the variable throughfall input to the forest floor which is depleted at some points (Holwerda et al., 2006; Keim et al., 2006) and exceeds even open rainfall at other points (Lloyd and Marques F., 1988; Siegert et al., 2016).~~ In complex forest ecosystems interception fluxes comprise of canopy (I_c), understory (I_u) and forest floor interception (I_l). Throughfall components are canopy drip (TF_c), understory throughfall (TF_u), stemflow (SF) and percolation of forest floor litter (J_l) (see Figure 1). Canopy drip throughfall is the largest water input fraction in most forest ecosystems exhibiting high spatial heterogeneity due to redistribution within the tree canopy (Levia et al., 2011). Understory throughfall from secondary vegetation can further enhance this spatial variability. The emerging spatio-temporal patterns result in variable throughfall input to the forest floor which is depleted at some points (Holwerda et al., 2006; Keim et al., 2006) and can exceed even open rainfall at other points (Lloyd and Marques F., 1988; Siegert et al., 2016). Pronounced spatially heterogeneous pattern may persist over time and propagate through forest floor percolation further into the soil (Fischer et al., 2023; Schume et al., 2003; Shachnovich et al., 2008; Zimmermann et al., 2009) influencing the forest water balance - in particular near-surface hydrological processes. Throughfall flux components also act as nutrient pathways from the canopy to the ground (Zimmermann et al., 2008a) and are linked to biogeochemical processes and plant-water availability near the forest floor (Dalsgaard, 2007; Raat et al., 2002).

~~The main controls for the spatio-temporal variability of throughfall are vegetation structure and composition (Staelens et al., 2006b; Zimmermann et al., 2008a), topography (Siegert et al., 2016) and regional meteorology including precipitation (Levia and Frost, 2006; Raat et al., 2002; Siegert et al., 2016). The general mass balance is (adapted from Carlyle-Moses(2004)):~~

~~$$TF = P_g - (I_c + SF) \qquad (1)$$~~

~~with TF = throughfall; P_g = gross precipitation; I_c = canopy interception; SF = stemflow (all in mm). The spatially heterogeneous character of throughfall influences the forest water balance - in particular near-surface hydrological processes - and potentially propagates into the soil (Fischer et al., 2023; Schume et al., 2003; Shachnovich et al., 2008; Zimmermann et al., 2009). In addition, throughfall functions as a nutrient pathway from the canopy to the ground (Zimmermann et al., 2008a) and is linked to biogeochemical processes and plant-water availability near the forest floor (Dalsgaard, 2007; Raat et al., 2002).~~
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$$P_g = I_{tot} + TF_{tot} + SF \qquad (1)$$

with P_g = gross precipitation, I_{tot} = total interception, TF_{tot} = total throughfall and SF = stemflow (all in mm). Total vegetation interception and throughfall are given by

$$I_{tot} = I_c + I_u + I_l \qquad (2)$$

$$TF_{\mathrm{tot}} = TF_{\mathrm{c}} + TF_{\mathrm{u}} \qquad (3)$$

with I_c = canopy interception, I_u = understory interception and I_l = litter interception and TF_{c} = canopy throughfall and TF_{u} = understory throughfall (all in mm). These components combine into the integrated (or net) flux of litter percolation (see coloured grouping in Figure 1), defined as

$$J_l = (TF_{\mathrm{c}} + TF_{\mathrm{u}}) - I_l \qquad (4)$$

with J_l = litter percolation or "integrated throughfall" (in mm). Litter percolation describes the amount of total throughfall that - reduced by vegetation and litter interception losses - percolates from the forest floor litter to infiltrate into the soil (Gerrits et al., 2010) and/or generate biomat flow in the litter layer (Bachmair and Weiler, 2011). Litter percolation is hereafter referred to as integrated throughfall, litter refers to the forest floor uppermost litter layer (see 2.2).

1.1. Experimental throughfall measurement approaches: potentials and limitations

The crucial role of interception storage on the one side and throughfall as the major water flux from the forest canopy to the ~~forest floor~~soil shown in Figure 1 ~~has motivated a broad range of experimental studies dedicated to the challenging task of accurately estimating throughfall volumes and intensities, solute inputs and the spatio-temporal variability of throughfall (Keim et al., 2005; Levia and Frost, 2003, 2006; Link et al., 2004; Pypker et al., 2005). The last three decades have seen a rise in interest as research has begun to examine spatio-temporal throughfall patterns across different ecosystems, vegetation structures, and spatial scales, as well as the intricate relationship to biotic and abiotic controls (Bialkowski and Buttle, 2015; Levia et al., 2019; Staelens et al., 2006b). The principle measurement approach of most studies characterize three components: the collector type, the sampling size and the sampling design (Zimmermann and Zimmermann, 2014). Throughfall samplers are either individual funnels (approx. 500 cm² orifice) or troughs (several meter long and several square meter orifice; both see~~ has motivated a broad range of experimental studies dedicated to the challenging task of accurately estimating throughfall volumes and intensities, solute inputs and the spatio-temporal variability of throughfall (Keim et al., 2005; Levia and Frost, 2003, 2006; Link et al., 2004; Pypker et al., 2005). The last three decades have seen a rise in interest as research has begun to examine spatio-temporal throughfall patterns across different ecosystems, vegetation structures, and spatial scales, as well as the intricate relationship to biotic and abiotic controls (Bialkowski and Buttle, 2015; Levia et al., 2019; Staelens et al., 2006b). The principle measurement approach of most studies characterize three components: the collector type, the sampling size and the sampling design (Zimmermann and Zimmermann, 2014). Throughfall samplers are either individual funnels (approx. 500 cm^2 orifice) or troughs (several meter long and several square meter orifice; both see Figure 1) arranged in a stratified or random design. Sample sizes range from tens to hundreds of samplers supplemented by a nearby rain gauge for gross precipitation measurements. Readings are either continuous or at an event basis or a larger interval.

a. Sample size

~~Substantial effort has been made to obtain the most accurate throughfall estimates (Carlyle-Moses et al., 2014; Crockford and Richardson, 2000; Thimonier, 1998). The frequently used method of Kimmins(1973) uses the coefficient of variation to calculate the number of required collectors (sample size) to measure representative throughfall averages for a given confidence interval and precision (at pre-set small mean TF errors) (Carlyle-Moses, 2004; Lloyd and Marques F., 1988; Rodrigo and Àvila, 2001). The required sample sizes are often large and not feasible for experimental setups. The deployment of roving samplers - contrary to samplers at fixed locations - increases the number of sampling locations through relocation (Kimmins, 1973; Link et al., 2004; Lloyd and Marques F., 1988; Rodrigo and Àvila, 2001; Ziegler et al., 2009) and thereby reduces the sample size, but can only be used to estimate long-term averages.~~

~~b. Sampler type~~

~~The quality of the throughfall measurement is determined by the type of collector (trough or funnel) in addition to the sample size. The collector efficiency depends on spatial throughfall structure: funnel sampler generally offer high spatial resolution (Zimmermann et al., 2010) and may capture very short-range variation (few cm) or throughfall dripping edges. Troughs are quite efficient for small sample sizes reducing random errors, integrating outliers and covering large areas below the canopy (Zimmermann et al., 2010; Zimmermann and Zimmermann, 2014). Often, funnel and trough collectors are combined.~~

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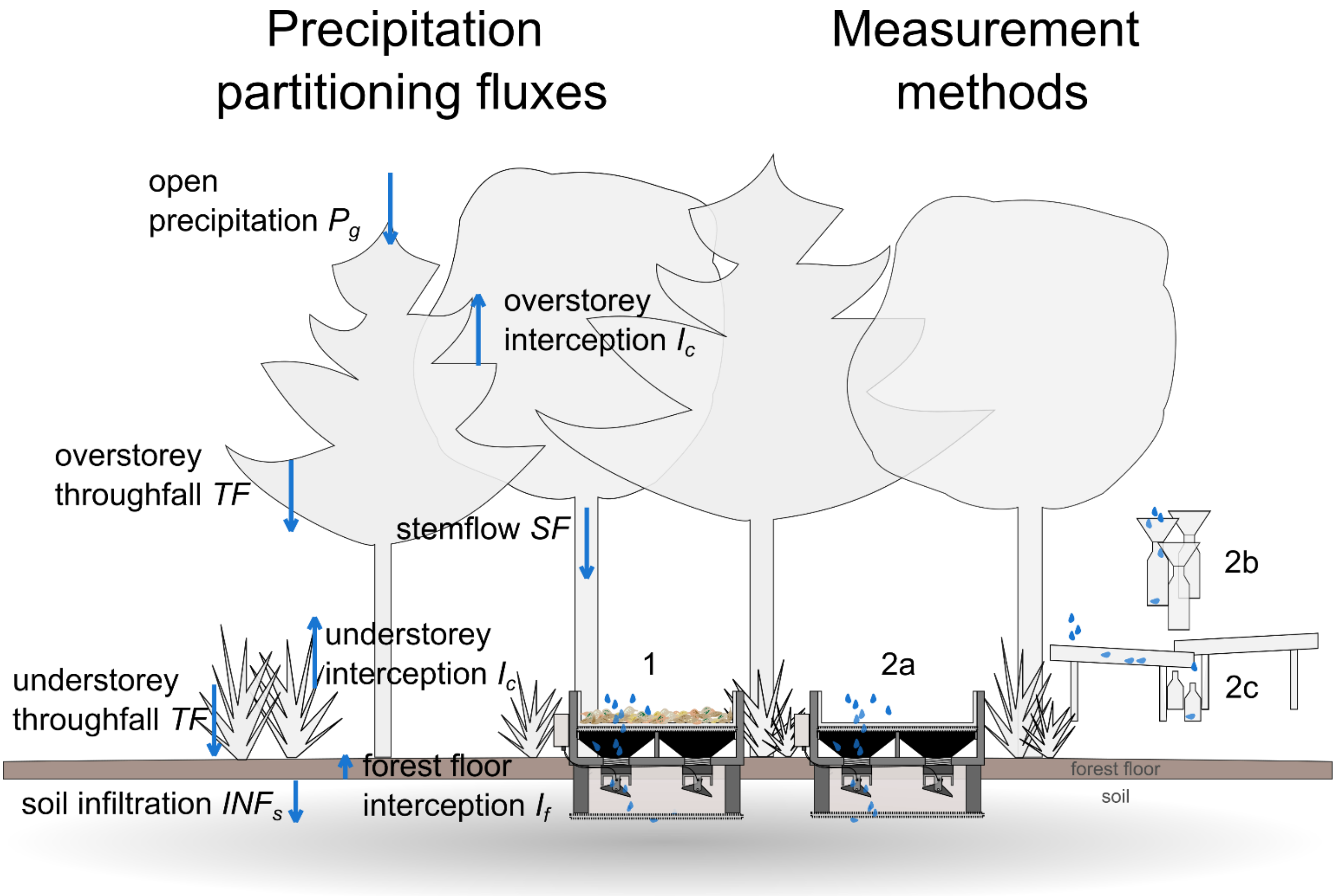
c. Sampling design

~~According to Thimonier(1998) and Zimmermann et al.(2010), the sampling strategy or design is more important than the sample size or type of sampler. For example a random distribution can avoid clustering, assure representative sampling and support geostatistical analysis approaches (Metzger et al., 2017; Zimmermann et al., 2009). In terms of measurement~~

frequency, automated or manual readings define the temporal and spatial resolution of the data and consequently its scope of application. The absence of a consensus on reference standard devices and schemes for throughfall measurements is emphasized by the substantial body of research with varying experimental and analytical approaches (Levia and Frost, 2006; Llorens and Domingo, 2007; Zimmermann and Zimmermann, 2014). Different choices of spatial scales, measurement setups and ecosystems under study limit the comparability and transferability of study findings (Crockford and Richardson, 2000; Lloyd and Marques F., 1988). Zimmermann et al.(2016) showed that existing measurement schemes are frequently not optimally matched to the system under study, omitting relevant factors such as the study area extent. Mismatches between required theory and practical feasibility (Thimonier, 1998; Zimmermann and Zimmermann, 2014) can produce large errors in the data (Kimmins, 1973; Thimonier, 1998; Zimmermann et al., 2010). According to Thimonier (1998) and Zimmermann et al. (2010), the sampling strategy or design is more important than the sample size or type of sampler. For example a random distribution can avoid clustering, assure representative sampling and support geostatistical analysis approaches (Metzger et al., 2017; Zimmermann et al., 2009). In terms of measurement frequency, automated or manual readings define the temporal and spatial resolution of the data and consequently its scope of application. The absence of a consensus on reference standard devices and schemes for throughfall measurements is emphasized by the substantial body of research with varying experimental and analytical approaches (Levia and Frost, 2006; Llorens and Domingo, 2007; Zimmermann and Zimmermann, 2014). Different choices of spatial scales, measurement setups and ecosystems under study limit the comparability and transferability of study findings (Crockford and Richardson, 2000; Lloyd and Marques F., 1988). Zimmermann et al. (2016) showed that existing measurement schemes are frequently not optimally matched to the system under study, omitting relevant factors such as the study area extent. Mismatches between required theory and practical feasibility (Thimonier, 1998; Zimmermann and Zimmermann, 2014) can produce large errors in the data (Kimmins, 1973; Thimonier, 1998; Zimmermann et al., 2010).

Throughfall is most variable at small scales (Wullaert et al., 2009) and varies as function of canopy complexity, tree density and precipitation magnitude (Rodrigo and Àvila, 2001; Staelens et al., 2006a). Measurements at the event or even intra-event scale and with a large sample size are recommended in order to investigate throughfall variability (Staelens et al., 2006b) and temporal stability of throughfall patterns (Fischer et al., 2023; Keim et al., 2005; Staelens et al., 2006a; Zimmermann et al., 2010). However, individual rain events are rarely observed at high temporal resolution Throughfall is most variable at small scales (Wullaert et al., 2009) and varies as function of canopy complexity, tree density and precipitation magnitude (Rodrigo and Àvila, 2001; Staelens et al., 2006a). Measurements at the event or even intra-event scale and with a large sample size are recommended in order to investigate throughfall variability (Staelens et al., 2006b) and temporal stability of throughfall patterns (Fischer et al., 2023; Keim et al., 2005; Staelens et al., 2006a; Zimmermann et al., 2010). However, individual rain events are rarely observed at high temporal resolution (Staelens et al., 2008) and spatio-temporal analysis of throughfall typically rely on measurement schemes of a low number of events (Raat et al., 2002; Staelens et al., 2006a) typically missing intra-event variability. and spatio-temporal analysis of throughfall typically rely on measurement schemes of a low number of events (Raat et al., 2002; Staelens et al., 2006a) typically missing intra-event variability.



Precipitation partitioning fluxes
Measurement methods
open precipitation Pg
overstorey interception Ic
overstorey throughfall TF
stemflow SF
understorey interception Ic
understorey throughfall TF
forest floor interception If
soil infiltration INFs
1
2a
2b
2c
forest floor
soil

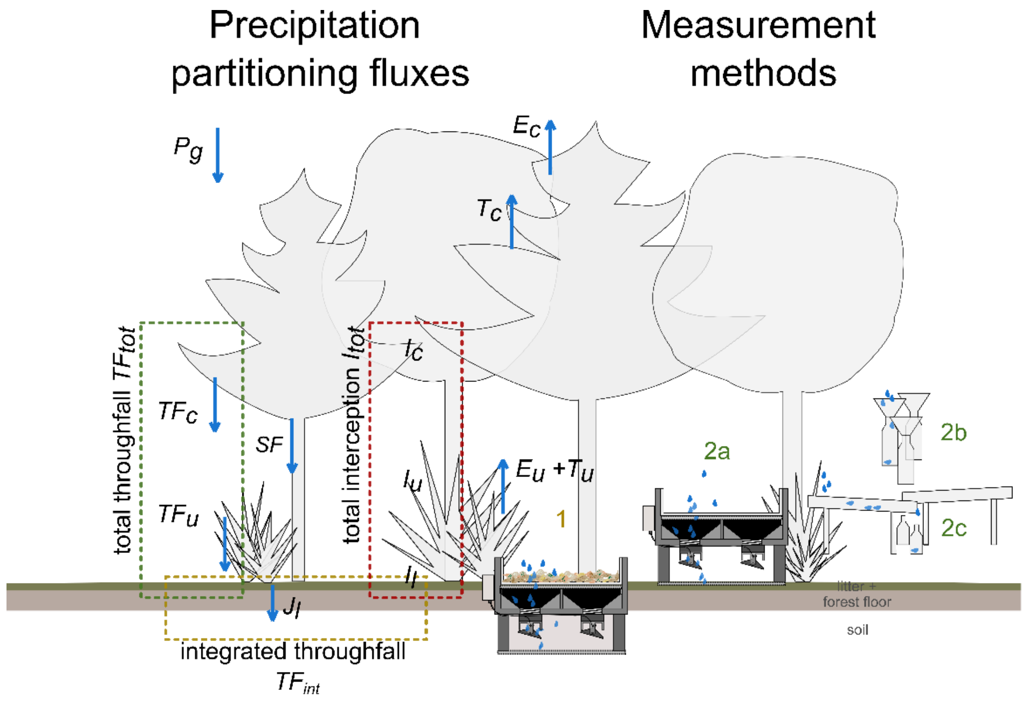


Figure 1: Schematic overview of precipitation partitioning in a forest (left) and throughfall measurement methods (right). (1) shows the ~~throughfall~~FluxIT sampler with ~~forest floor~~litter inlay measuring ~~soil infiltration~~integrated throughfall J_l, (2a-c) show methods for canopy throughfall measurements TF_c: the ~~throughfall~~FluxIT sampler without ~~forest floor~~litter inlay (2a), funnel (2b) and trough (2c) collectors.

1.2. Experimental throughfall measurement approaches: new directions

~~To address the outlined challenges, high quality throughfall data are needed, collected with appropriate measurement designs and devices (Germer et al., 2006; Levia and Frost, 2006; Zimmermann et al., 2009; Zimmermann and Zimmermann, 2014). In order to observe small-scale variability at the individual tree level and to representatively cover entire stand vegetation structures at high spatial resolution, these schemes should ideally have large number of samplers with adequately sized samplers at fixed locations (Fischer et al., 2023; Metzger et al., 2017; Zimmermann et al., 2010). Only measurements at high temporal resolution on intra-event scale allow to investigate canopy storage, interception changes and temporal stability of spatial throughfall patterns (Keim et al., 2005; Zimmermann and Zimmermann, 2014). Ideally, these schemes integrate automated and continuous measurement devices that collect long-term high quality data which meet the requirements for the analysis of seasonality effects (Zimmermann et al., 2008b) and long-term changes of throughfall spatio-temporal patterns (Link et al., 2004). Together, these schemes would then contribute to a deeper understanding of throughfall controls like rainfall regimes, relationship to near-surface hydrological processes and canopy interception variability in forest stands (Blume et al., 2022; Zimmermann et al., 2008a; Zimmermann et al., 2008b). Furthermore, it might improve water flux modelling of watersheds providing better insights into water losses and yields of forested areas (Crockford and Richardson, 2000; Zimmermann et al., 2008b).~~

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In this technical note, we present a novel ~~throughfall measurement~~ monitoring system for throughfall flux components with flexible application options that allow throughfall to be measured either in isolation or in an integrated manner across e.g. forest ecosystems of varying vegetation complexity. The presented system was implemented in forest plots of pure Beech, pure Douglas fir, pure Silver fir and mixed Beech and Douglas fir stands. The presented system comprises a network of 60 ground-level tipping bucket samplers with 240 individual collection compartments ~~for~~, enabling quantification of isolated or integrated throughfall spatio-temporal variability on an intra-event and intra-stand scale. ~~The systems were set up in forest plots of pure Beech, pure Douglas fir, pure Silver fir and a mixed Beech-Douglas fir plot. The~~ The automated, continuous ~~measurement produces~~measurements generate long-term time series at a high spatio-temporal resolution with moderate maintenance ~~effort that can be used to investigate temporal persistence, long-term changes in throughfall, and~~requirements and minimal disturbance to near-surface hydrological processes. These data support investigations of spatial species-specific throughfall patterns~~. In addition, it causes minimal disruption to near-surface hydrological processes.~~ , their temporal persistence and long-term changes.

~~In order to gather data on throughfall spatio-temporal variability,~~ The study aims are to (i) design a throughfall sampler with flexible application options that ~~allows~~enables continuous, automated and minimally invasive ~~data collection~~measurement of integrated or isolated throughfall, (ii) develop a sampling scheme ~~using a combination of samplers~~ that accurately ~~measures~~captures e.g. the variability of integrated throughfall ~~variability~~ across plots of different tree compositions; and (iii) implement the sampling scheme at the study site of a temperate, mature forest stand.

2. Study area and methods

~~2.1. Throughfall measurement sampler~~

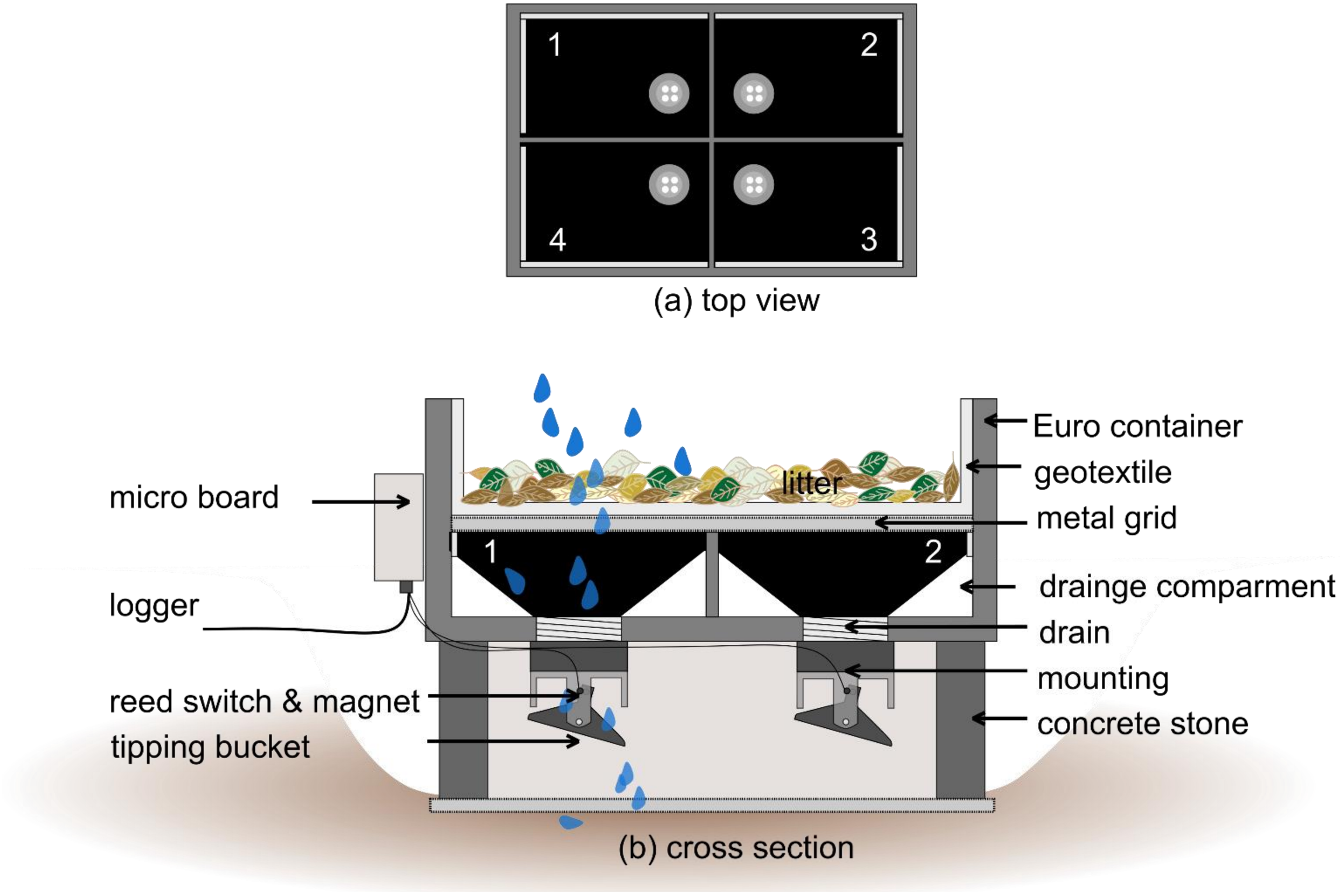
2.1. Throughfall monitoring system

The sampler and tipping bucket unit presented in this Section constitute the principle design of the throughfall monitoring system named "**FluxIT sampler**" (**Flux** of **I**ntegrated **T**hroughfall **sampler**). The FluxIT sampler setup is modifiable in various ways to measure throughfall fluxes either in isolated or integrated manner according to respective research objectives and ecosystem under study. In this section, the sampler and tipping bucket unit designs are presented (2.1.1 & 0). A detailed overview of application options and corresponding sampler modifications (2.1.3) is followed by a description of the samplers set up in a throughfall measurement network (2.3).

2.1.1. Sampler design

The throughfall sampler depicted in Figure 2~~Figure 2~~ uses a rectangular, solid plastic container (60 x 40 x 20 cm, Euro container) divided into four equal trough drainage compartments. Each compartment, with a 600 cm^2 collection area, collects and funnels throughfall water to a separate tipping bucket unit for water flux measurement. The container's inner walls are lined with water and UV-resistant pond liner, clamped by aluminium ledgers to create sufficient slope for water drainage. Additionally, the aluminium ledgers serve as mounts for a detachable metal grid and geotextile ~~upon which forest floor litter can be placed or will fall naturally (~~(Figure 2~~Figure 2). Hence, the metal grid in combination with the geotextile~~) that filters the throughfall to prevent clogging of the tipping bucket. The pond liner is attached to an opening at the bottom exit of the container with a commercial sink drain in each compartment. A nylon mesh filter keeps small particles out of the drain. The tipping bucket units (see 0) are fastened to the drain treads underneath the container~~, allowing them to tip freely into the open area below (Figure 2). The whole throughfall containers are levelled on porous concrete stones (25 x 25 x 20 cm) with a hollow centre.~~. Prior to installation, topsoil of 30 cm depth was removed at each sampler location. Metal grids and ~~a~~porous concrete stone ~~together~~frames (25 x 25 x 20 cm) with ~~a metal grid~~hollow centres were ~~inserted.~~placed into the excavations to serve as sampler mountings. The throughfall samplers are positioned on top of these levelled stone frames. This setup creates a flat, sturdy mounting that prevents ~~tipping of the container~~animal intrusion and allows ~~infiltration of~~ the tipping buckets to tip freely (Figure 2) releasing the quantified water fluxes ~~in~~to the soil below. Although excavation of a 25 x 25 cm area inevitably introduces local soil disturbance, this intervention is performed only once and provides a stable, levelled, and long-term secure mounting for the sampler and its tipping units. Routine maintenance (e.g. exchange of tipping buckets) thereafter requires only lifting or tilting the sampler sideways. Unlike the typical large troughs, this design is minimally invasive to near-surface hydrological processes and tree water supply~~.~~, as water is not redirected in a larger funnel. It also enables direct measurement of other soil processes ~~in the soil~~ like ~~soil~~ water content~~. In addition, the water could also be sampled underneath for water quality measurements. The stone base also makes maintenance easier, like exchanging tipping buckets, and prevents animal intrusion.~~ or matric potential. Positioning the containers at ground level minimizes measurement inaccuracies ~~caused~~

~~by~~e.g. wind~~ generating typical under-catch of rainfall gauges.~~ -induced undercatch. Alternative installation approaches with reduced soil disturbance are addressed in Section 2.1.3.


1
2
4
3
(a) top view
micro board
logger
reed switch & magnet
tipping bucket
litter
1
2
Euro container
geotextile
metal grid
drainge comparment
drain
mounting
concrete stone
(b) cross section

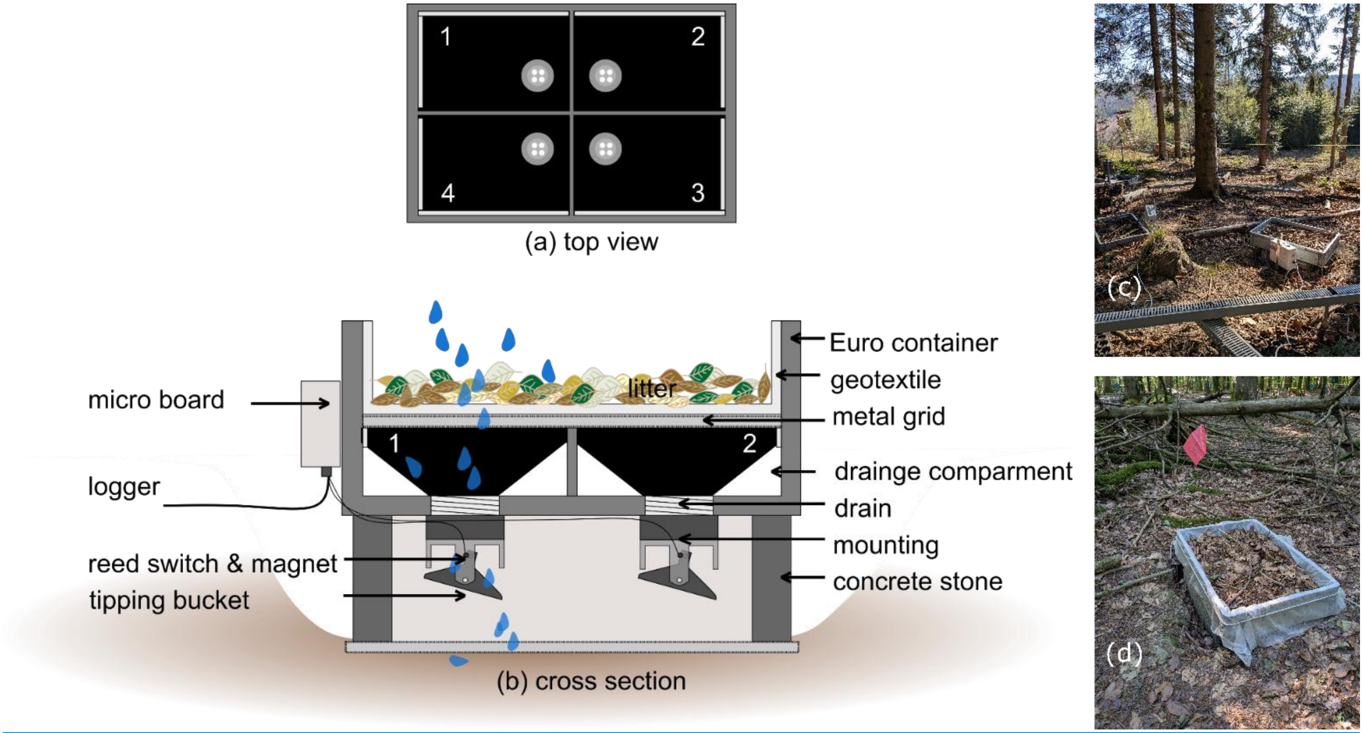


Figure 2: Design of ~~throughfall~~FluxIT sampler with four equal drainage compartments (top view), each equipped with single tipping bucket units for measurement (cross section). Photographs (c, d) show samplers installed at two locations in the field, at a moment of configuration II for measurements of integrated throughfall with litter in the samplers

~~2.1.2. Forest floor inlay~~

~~The effect of forest floor interception (Gerrits et al., 2007; Paulsen and Weiler, 2025) that reduces throughfall to the fraction infiltrating into the mineral soil, can be measured by adding unfragmented forest floor litter to the throughfall sampler (Figure 2 & Figure 2). In case this is not of interest, the throughfall sampler can be deployed without forest floor litter directly obtaining throughfall data above the forest floor (see (2a) in Figure 1). For our study, we decided to include the interception of the forest floor in our measurements (Gerrits et al., 2007; Paulsen and Weiler, 2025) and, to directly investigate the propagation of the overall throughfall patterns into near-surface soil moisture (Fischer et al., 2023; Metzger et al., 2017; Raat et al., 2002; Schume et al., 2003). Adding litter into the samplers has the practical benefit of lowering measurement errors since the organic material prevents splash out of the sampler. As for this study, samplers with a forest floor addition were used, the throughfall samplers are referred to as infiltration samplers thereafter.~~

~~2.1.3.~~2.1.2. Tipping bucket units

The 3D-printed tipping buckets consist of two parts: a mounting with a reed switch and threaded top, and a tipping bucket with a magnet on top (Figure 2~~Figure 2~~). The tipping bucket is attached to the mounting by a stainless-steel pin. Both components are made from durable and chemically resistant PETG, with the inner surfaces of the tipping bucket units ironed reducing water adhesion and rewetting effects. For further information on the design and implementation of the tipping bucket unit see Paulsen and Weiler (2025). A ~~4~~3 mm diameter pipe inside the mounting channels water from the drain to the tipping bucket, guaranteeing a maximum flow rate that cannot be exceeded. During very high intensity rainfalls, the drainage compartments can store and buffer up to 1 L of excessive water. Tipping buckets with a defined volume of 2.7 mL alternately fill up with water, tip to empty and thereby move the magnet past the reed switch. This mechanism triggers the reed switch to sense out short electric pulses. The pulses are recorded by the micro board as tip counts per defined time interval. Counted tips are converted to water volumes and divided by collection area of the ~~infiltration~~integrated throughfall sampler compartments to calculate the ~~infiltration~~integrated throughfall depth (mm) for every compartment.

The micro boards were designed as a cost-effective solution for continuous, automated measurement of ~~infiltration~~integrated throughfall water volumes. They each accommodate 4 pulse count inputs and an extension for four additional pulse count inputs. A micro controller (AVR SAMD21) is used to realize the pulse inputs as interrupts, guaranteeing that no pulses are missed. Two steps are involved in debouncing the tipping bucket units: each reed switch has a resistance capacitor filter linked to reduce bouncing. Laboratory tests with debouncing tipping bucket units produced 500–800 ms as ideal timeout settings preventing erroneous double tips" while still capturing all pulses for tested maximum rainfall intensity of less than 90 mm h^{-1}. The micro board is connected to a Campbell Scientific CR350 data logger via SDI-12 data communication, logging at 15 min intervals. Any other SDI-12 logger could be used as well. A mini-USB port offers direct connection to a computer. For detailed descriptions on the micro board hardware and programming please refer to Paulsen and Weiler (2025).

2.1.3. Application options of throughfall sampler

The FluxIT sampler setup is modular and can be adapted to respective research objectives and ecosystems under study. Two primary configurations are distinguished:

Configuration I: Alternative to traditional methods measuring isolated throughfall fluxes e.g. canopy throughfall (TF_c) with or without understory throughfall (TF_u)

Configuration II: Novel method measuring integrated throughfall fluxes e.g. litter percolation (J_l)

The height of the sampler is adjustable to enable these configurations. Mounted on a simple stainless-steel frame above understory vegetation and without any inlay, the sampler measures canopy throughfall (TF_c) (see (2a) in Figure 1) in a manner comparable to traditional trough- or funnel systems (configuration I). Installed at ground level, the sampler captures integrated canopy and understory throughfall ($TF_c + TF_u$). With a litter layer- defined here as the uppermost forest floor layer of unfragmented leaves and needles (definition see 2.2)- placed or naturally accumulating onto the sampler's metal grid and geotextile (Figure 2), the sampler operates as a novel device for measuring litter percolation at high spatio-temporal resolution. This application allows to investigate the less frequently observed flux of litter percolation and the effect of litter interception that reduces throughfall to the fraction of water percolating towards deeper layers of forest floor and soil (Gerrits et al., 2007; Paulsen and Weiler, 2025).

Only the loose, uppermost, unfragmented fresh litter is placed in the sampler initially; as litter fragments and accumulates over time, percolation time may increase. The samplers can be emptied e.g. annually or sub-annually. In all configurations, the quantified water can also be sampled beneath the tipping buckets for water quality analysis. Adding litter into the samplers has the practical benefit of lowering measurement errors since the organic material prevents splash out of the sampler and functioning as an additional litter trap (Figure 2). For less invasive field installation, the porous concrete stone frames can be replaced by PCV pipes, requiring only an excavation of 4 times 10 cm in diameter and 20 cm depth. Freely swinging, automatically levelled tipping buckets are also feasible and would reduce the requirement of a precisely levelled sampler mounting. Both container and tipping bucket unit sizes are flexible and adaptable to the study purpose and monitored system. Euro-containers are available in a wide range of dimensions.

2.2. Site description

The ~~infiltration~~integrated throughfall samplers are part of an ecohydrological sensor network located at the ECOSENSE forest research site (Figure 3~~) near Ettenheim (48.2685° N, 7.8782° E), between the Upper Rhine Valley and the Black Forest in south western Germany.~~) near Ettenheim (48.2685° N, 7.8782° E), between the Upper Rhine Valley and the Black Forest in south western Germany. The ECOSENSE forest research site monitors forest ecosystem dynamics such as carbon and water flux dynamics in space and time with a wide range of established measurement technologies and newly developed sensors (Tesch et al., 2025). The elevation of the research site ranges from 480 to 500 m a.s.l., the climate is temperate-humid (Köppen *Cfb*) with mean annual precipitation of 1120 mm and mean annual temperature 9.5° C in reference period 1961–1990 (Landesanstalt für Umwelt Baden-Württemberg, 2024). Occasional summer drought can occur (year 2003 & 2018). The northern part of the research site is located on Cambisols over Triassic Buntsandstein, while the southern part features Pseudogley soils over Triassic Muschelkalk (Werner et al., 2024). The forest floor, defined as the upper most forest soil layer of fragmented and unfragmented organic material above the mineral soil layer (LOf/Oh), averages 10 cm in thickness and ~~has been~~is classified as moder (Landesamt für Geologie, Rohstoffe und Bergbau Baden-Württemberg, 2024). The litter layer (L) of < 2 cm thickness consists of loose, uppermost, unfragmented fresh litter including leaves and needles.

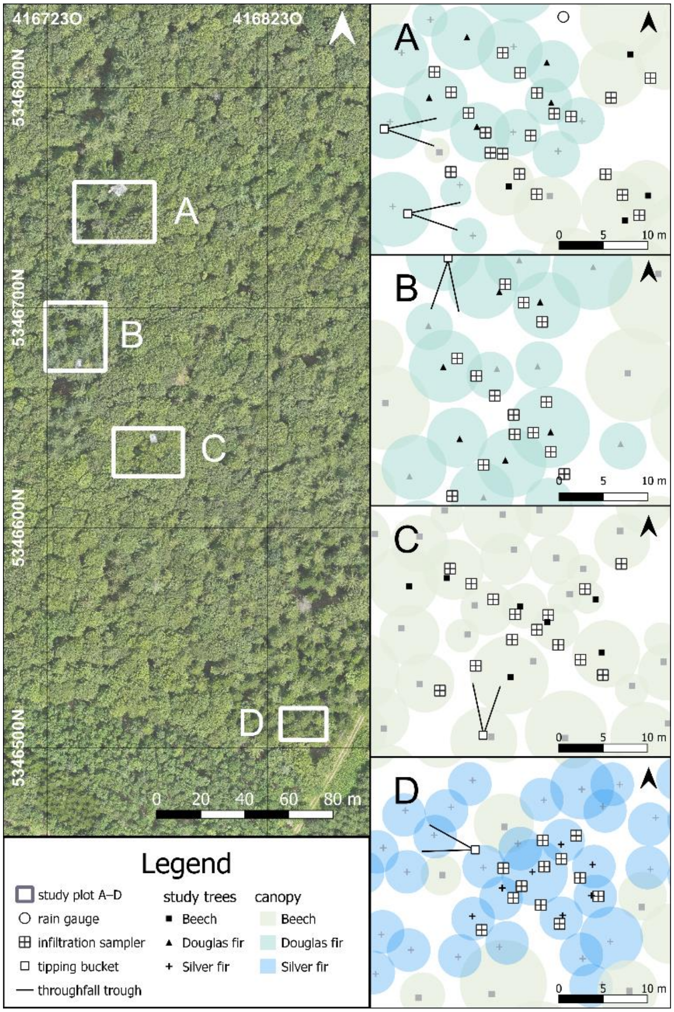


Figure 3: ECOSENSE forest research site near Ettenheim, Germany. To the right the ~~infiltration~~integrated throughfall measurement network of each study plot A–D with mixed Beech and Douglas fir (A), pure Douglas fir (B), pure Beech (C) and pure Silver fir (D). Image: M. Gassilloud

The research site is covered by a European Beech-dominated mixed forest managed for timber production. In addition to Beech trees (*Fagus sylvatica*), the stand is composed by patches of Silver fir (*Abies alba)* and Douglas fir (*Pseudotsuga menziesii*), aged approximately 80, 60 and 50 years, respectively (Landesamt für Geoinformation und Landentwicklung Baden-Württemberg, 2025). Individual spruce and oak trees (Werner et al., 2024) complement the average tree density of 500 trees ha^{-1}. With the exception of a few younger 10–20 m tall Beech and Silver fir trees, there is no understorey vegetation, and the overstorey reaches heights of 25–35 m. The study area shown in Figure 3 was divided into four subplots: a mixed Beech & Douglas fir plot (A), pure Douglas fir (B) and pure Beech (C) in direct vicinity on a plateau and a Silver fir plot (D) located 200 m southeast on a gentle southeastern slope (Figure 3). Each subplot is approx. 400 m^2 large and includes 7–10 selected "measurement trees" of comparable age and height.

2.3. Throughfall measurement network

The throughfall measurement network of 60 ~~infiltration~~integrated throughfall samplers is installed in a stratified design of perpendicular transects across the four plots (Figure 3~~ A–D). This distribution supports measurements below various canopy positions, including stem closeness, canopy centre, and canopy edge, and prevents clustering while guaranteeing coverage of the variance of the forest structure (Holwerda et al., 2006; Link et al., 2004). The network design combines a model-based and a design-based component as suggested by Zimmermann and Zimmermann(2014) to account for objective assessment of uncertainty together with purposive measurement such as for interest in distribution across the entire area. The sampler design of four neighbouring trough and tipping bucket units adds measurement points in direct vicinity (model-based component). This supports outlier detection and is favourable for geostatistical analysis approaches improving variogram estimation (Voss et al., 2016; Zimmermann et al., 2009). The distribution of the samplers on stratified transects with 3 and 5 m spacing in NW-SE and NE-SW orientation (design-based) ensures an unbiased estimate of spatial mean throughfall. Furthermore, variably spaced transects also support the variogram development and enable the estimation of short- and long range throughfall variability including anisotropy (Schume et al., 2003). Beginning in December 2024, continuous infiltration measurements of the first installed infiltration samplers were conducted. Regular maintenance has been performed on all samplers.~~
A–D). This distribution supports measurements below various canopy positions, including stem closeness, canopy centre, and canopy edge, and prevents clustering while guaranteeing coverage of the variance of the forest structure (Holwerda et al., 2006; Link et al., 2004). The network design combines a model-based and a design-based component as suggested by Zimmermann and Zimmermann (2014) to account for objective assessment of uncertainty together with purposive measurement such as for interest in distribution across the entire area. The sampler design of four neighbouring trough and tipping bucket units adds measurement points in direct vicinity (model-based component). This supports outlier detection and is favourable for geostatistical analysis approaches improving variogram estimation (Voss et al., 2016; Zimmermann et al., 2009). The distribution of the samplers on transects with 3 and 5 m spacing in NW-SE and NE-SW orientation (design-based) ensures an unbiased estimate of spatial mean throughfall by covering the entire plot. Consequently, the sampler range in random distances of 0.5–5 meters to the nearest stems of surrounding measurement tree depicting throughfall at diverse canopy densities/positions. Furthermore, variably spaced transects also support to apply geostatistic approaches like variograms and enable the estimation of short- and long range throughfall variability including anisotropy (Schume et al., 2003). Beginning in December 2024, continuous integrated throughfall measurements of the first installed integrated throughfall samplers were conducted. Regular maintenance has been performed on all samplers.

Figure 3 also depicts complementary measurements of the network: five traditional throughfall trough collectors~~, which are~~ positioned at random ~~throughout~~within each study plot, and two rain gauges for gross precipitation complement the setup. The trough collectors consist of two V-shaped stainless-steel troughs of 0.3 m^2 receiving area and a length of 3 m installed 1 m above ground at an 45° angle relative to each other. The troughs channel collected water into a tipping bucket rain gauge (Rain collector II 7852, Davis Instruments, resolution 0.2 mm) covered by a nylon mesh. Gross precipitation is measured with tipping bucket rain gauges (YOUNG Tipping bucket rain gauge 52202, resolution 0.1 mm). They were placed at on opening located at 200 m distance from plots mounted 1 m above ground and above the canopy at 47 m height installed on a measurement tower located at the centre of the mixed plot (Figure 3). Precipitation data were wind-corrected according to Kochendorfer et al. (2017). LAI values are derived from hemispherical photographs and validated by litter trap. They are available for two dates immediately before and after the measurement period for the Mixed, Beech and Douglas fir plots from Lotz et al. (2025).

~~**3. Results**~~

~~**3.1. Calibration of the tipping bucket unit**~~

3. ~~**Every individual tipping bucket unit was calibrated in the laboratory to determine single tip volumes. 3D-printed inflow regulation plugs were screwed onto PET laboratory bottles of 100 mL and plugged upside down into the drains of each compartment dripping at constant rate of around 1 mL s^{-1} (Bialkowski and Buttle, 2015; Levia et al., 2019; Staelens et al., 2006b). For all tipping buckets the average tipping bucket volume is 2.87 mL (SD = ±0.28) resulting in a resolution of 0.05 mm tip^{-1} and an accuracy of 6.4 % calculated as mean percent error of measurements based on actual tip volume (2.7 mL). The tip resolution and accuracy are comparable with other commercial tipping bucket rain gauges (e.g. HOBO 0.2 mm resolution with 1% accuracy; YOUNG 0.1 mm resolution with 2% accuracy; DAVIS 0.2 mm with 4% accuracy). Four different 3D printers (Bamboo Lab P1P, Original Prusa i3 MK2, Raise3D Pro2, Ultimaker4) were used to manufacture the 240 tipping buckets in order to speed up the manufacturing process. The calibration results in**~~ **Results**

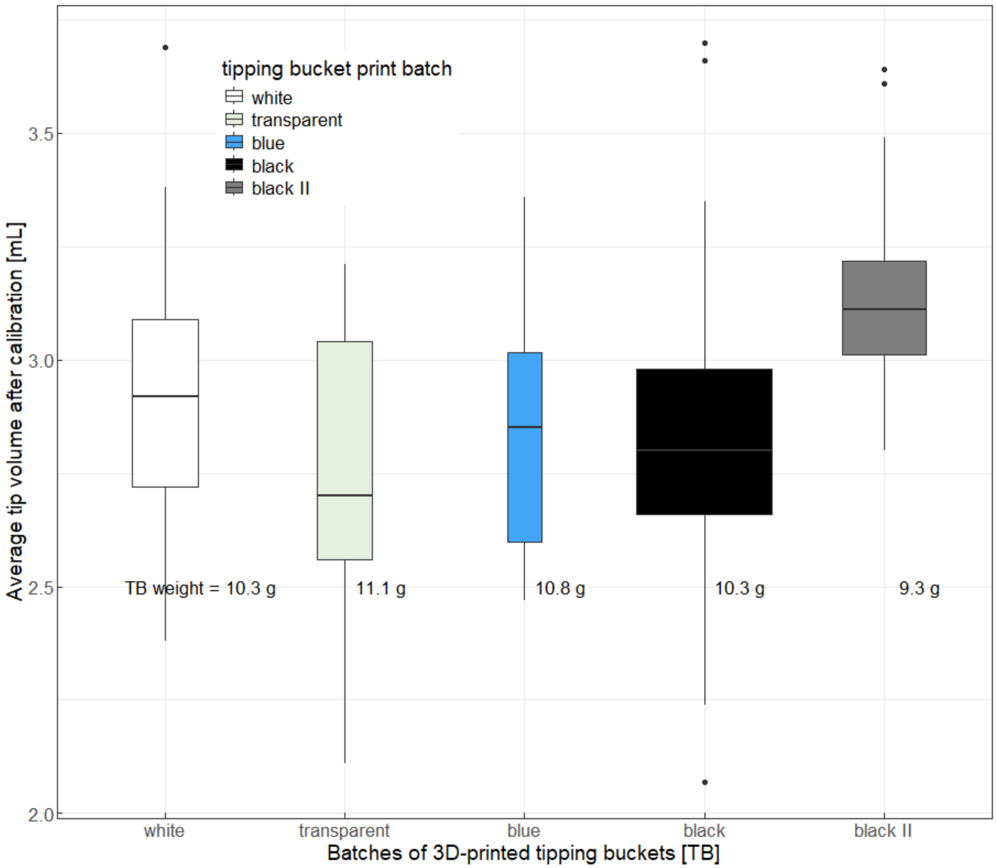
3.1. Calibration of the tipping bucket unit

Every individual tipping bucket unit was calibrated in the laboratory to determine single tip volumes. 3D-printed inflow regulation plugs were screwed onto PET laboratory bottles of 100 mL and plugged upside down into the drains of each compartment dripping at maximum rate of 0.67 mL s^{-1} (Bialkowski and Buttle, 2015; Levia et al., 2019; Staelens et al., 2006b). For all tipping buckets the average tipping bucket volume is 2.87 mL (SD = ±0.36) resulting in a resolution of 0.05 mm tip^{-1} and an accuracy of 4.4 % calculated as mean relative error of measurements to mean tip volume (2.87 mL). The tip resolution and accuracy are comparable with or even exceeds other commercial tipping bucket rain gauges (e.g. HOBO 0.2 mm resolution with 1% accuracy; YOUNG 0.1 mm resolution with 2% accuracy; DAVIS 0.2 mm with 4% accuracy). Four different 3D printers (Bamboo Lab P1P, Original Prusa i3 MK2, Raise3D Pro2, Ultimaker4) were used to manufacture the 240 tipping buckets in order to speed up the manufacturing process. The calibration results in Figure 4~~Figure 4~~ show differences in the ~~average~~median tip volumes ~~ranging~~ between 2.~~21 to~~71 and 3.~~69~~14 mL and average tipping bucket weights between 9.3 to 11.1 g across the print batches, despite the use of identical CAD input files. These variations start with varying printer accuracies and subsequently propagate into the tipping bucket volume measurements. Because of undercatch effects, the average tip volume from calibration decreases with the weight of the tipping bucket. Distinct boxplot IQRs of the print batches of 0.2 mL (black) to 0.5 mL (transparent) (Figure 4~~Figure 4~~) show the variability of within-batch precision of the individual printers. The calibration results highlight the importance of individual calibration ~~coefficients~~ for each tipping bucket unit and the need for regular field calibration to assure continuous high data quality. ~~Infiltration~~Integrated throughfall volumes and according depths were derived from tip counts of tipping bucket units as following:

$$INF_{depth} = \frac{(x_{pulse} \times c_{calibration})}{A_{collection}} \quad (2)$$

$$J_{depth} = \frac{(x_{pulse} \times c_{calibration})}{A_{collection}} \quad (5)$$

with x_{pulse} = counted tips, the individual calibration coefficient $c_{calibration}$ = 2.21 to 3.69 mL and a collection area of $A_{collection}$ = 600 cm^2.


tipping bucket print batch
white
transparent
blue
black
black II
Average tip volume after calibration [mL]
3.5
3.0
2.5
2.0
TB weight = 10.3 g
11.1 g
10.8 g
10.3 g
9.3 g
white
transparent
blue
black
black II
Batches of 3D-printed tipping buckets [TB]

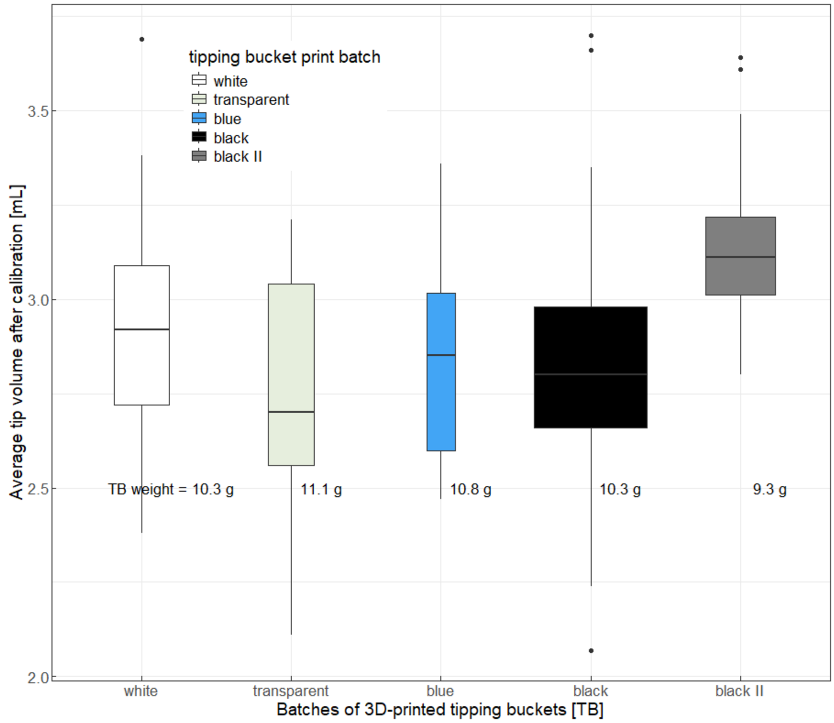


Figure 4: Calibration of 240 tipping bucket units with boxplots showing average tip volumes of tipping bucket units from five 3D-printing batches (batches defined by filament colours white, transparent, blue, black I & II).

3.2. Spatio-temporal ~~infiltration~~ dynamics of integrated throughfall from ~~infiltration network data~~measurement period in spring 2025

3.2.1. Separation of precipitation events and development of event ~~infiltration~~throughfall coefficient

The dataset included in this technical note originates from an early measurement period in which unfragmented litter was retained inside the samplers (configuration II) and the samplers were installed at ground level. Effectively, litter percolation (J_l) was measured as an integrated flux of over- and understory throughfall ($TF_c + TF_u$) reduced by litter interception (I_l). The event separation and development of throughfall coefficient are based on integrated throughfall data (J_l); however, the approach is transferable to datasets derived from isolated throughfall (TF_c) measurements.

Continuous in situ ~~infiltration~~integrated throughfall (J_l) data from all four subplots were available starting April 2nd, 2025. The corresponding time series of wind-corrected (Kochendorfer et al., 2017) hourly gross precipitation (P_g) from the tower station was separated for individual rainfall events and validated against ground station precipitation data. Events were defined as periods of continuous hourly $P_g \geq 0.1$ mm with a maximum of one hour $P_g \leq 0.1$ mm in between and a minimal total event $P_g \geq 2.5$ mm. An event ends when $P_g \leq 0.1$ mm for two consecutive hours or more. When applied to the ~~infiltration~~ integrated throughfall data, the event separation accounts for ~~canopy drip from open~~ throughfall and ~~forest floor~~litter percolation up to two hours post-event. The observed data showed that this ~~timeframe~~time frame is well suited as most ~~infiltration~~integrated throughfall events ended 2 hours after precipitation ~~ceased. ‘~~cessation. “Last ~~drips’ from infiltration samplers~~drips” recorded several hours past the event were not included. ~~Infiltration~~Water amounts from these periods were marginal and ~~can be neglected~~negligible.

~~Based on~~To express event integrated throughfall depths as ratios of event gross precipitation (P_g), throughfall coefficients (c_{tf}) were developed for each of the separated precipitation events ~~infiltration coefficients (C_{inf}) - following the principle of~~ . Analogous to the runoff coefficient (e.g. Blume et al. (2007) or ~~Savenije~~Savenije~~(1996)~~ ~~– were developed for every tipping bucket unit and event. Event C_{inf} of a throughfall bucket unit describes the total event infiltration at a location as a fraction of the total event gross precipitation and thereby facilitate the comparison of event infiltration across the four plots. It indicates how P_g is redistributed by forest vegetation: the percentage of P_g lost due to interception and the percentage of P_g infiltrating into soil.~~(1996)), the event throughfall coefficient c_{tf} describes the total event throughfall depth at a given location (a tipping bucket unit) as a fraction of total event gross precipitation. The metric facilitates comparison of throughfall fluxes across the four plots and indicates how P_g is redistributed by forest vegetation, specifically the percentages of P_g lost to interception and infiltrating into the soil. For a precipitation event *i* at a location *j* (tipping bucket unit), the throughfall coefficient is defined as ~~Infiltration~~

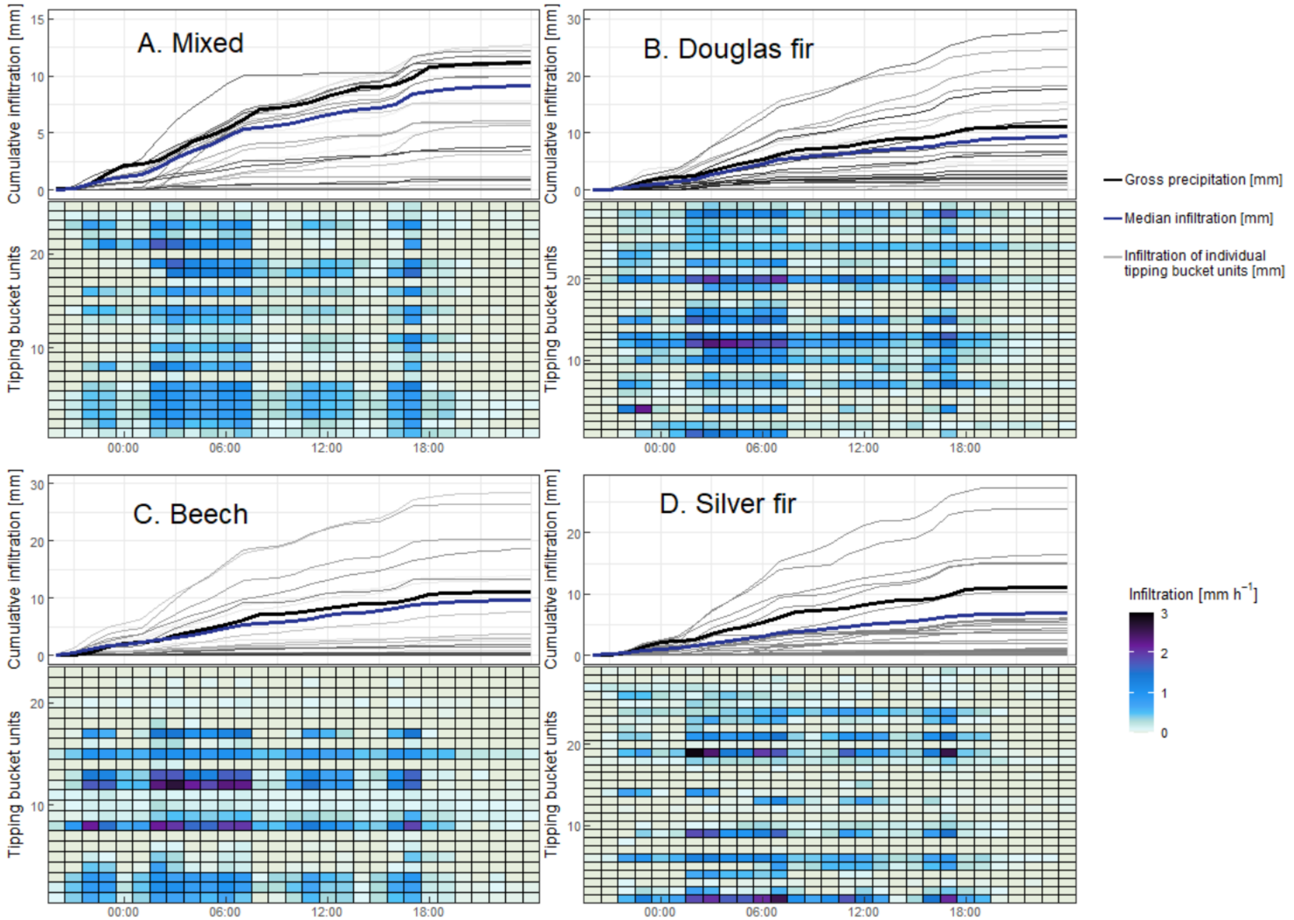
$$c_{tf,ij} = \frac{TF_{i,j}}{P_{g,i}} \qquad (6)$$

with $P_{g,i}$ = event gross precipitation and TF_{ij} = event throughfall (both in mm). Depending on the sampler configuration (configuration I or II), c_{tf} describes either isolated canopy throughfall (TF_c) or integrated throughfall (J_l) expressed as a fraction of gross precipitation.

3.2.2. Throughfall variability of an example event

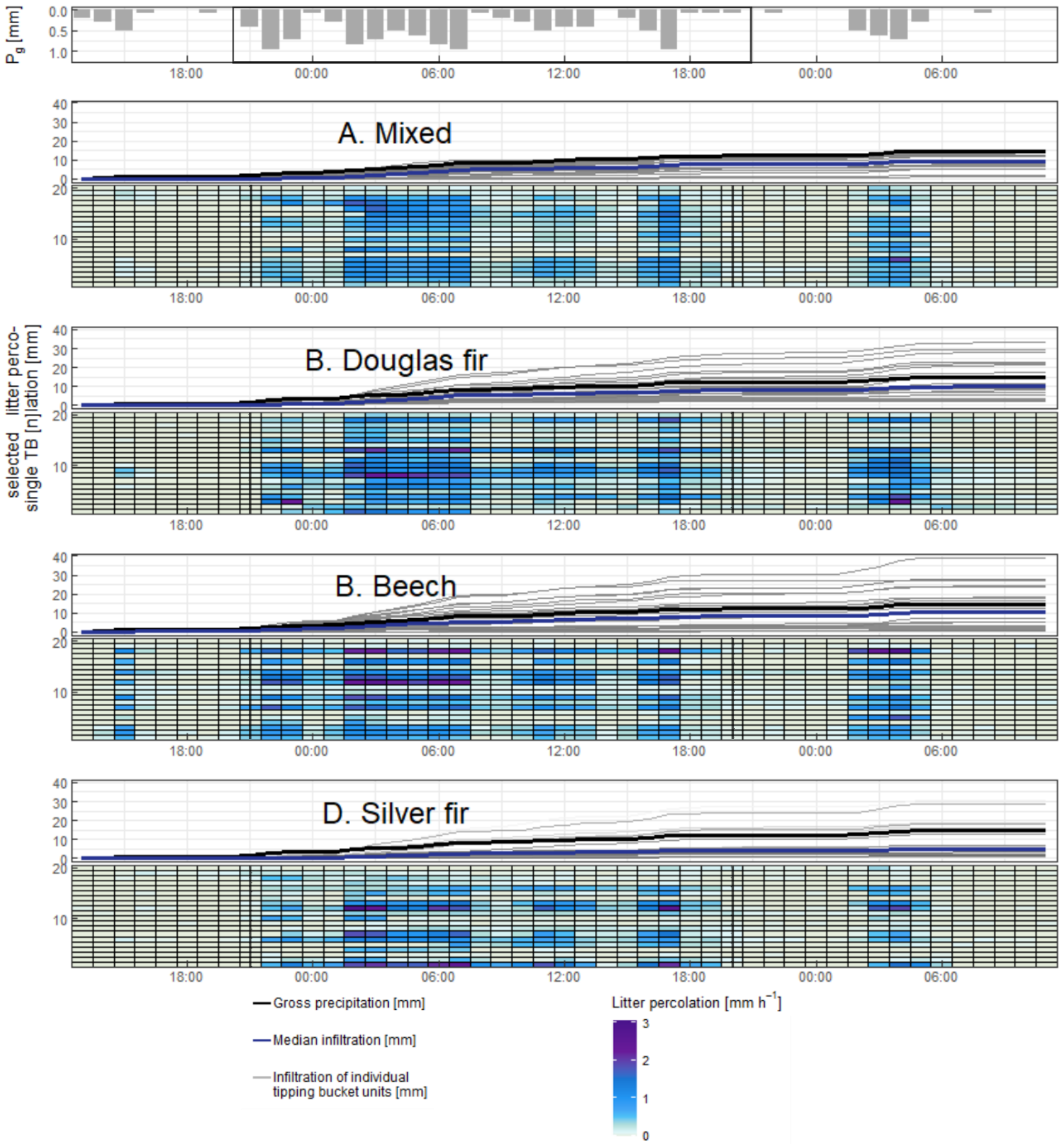
The selected example of a moderate precipitation event occurring April 17th 2025 (total event P_g = 11.2 mm, ~~28~~23 h, max. intensity of 1 mm h^{-1}) ~~presented in~~originates from an early measurement period in which litter was retained inside the samplers (configuration II). Effectively, litter percolation was measured as an integrated flux of throughfall reduced by litter interception. Figure 5 illustrates the spatio-temporal variability of integrated throughfall ~~and resulting event infiltration for~~(J_l) of 80 selected tipping bucket units at the four forest stands of Beech, Douglas fir, Silver fir and mixed trees. The median total event ~~infiltration~~ integrated throughfall (J_l) of the selected tipping bucket units per plot is highest at the Beech plot with ~~INF_{median} = 9.7~~ J_l = 11.1 mm followed by the Douglas fir and mixed plot (~~INF_{median} = 9~~J_l= 10.4 and 9.~~2~~8 mm~~) whereas~~). The Silver fir plot has the lowest median ~~infiltration~~ with ~~INF_{median} = 6~~J_l = 5.9 mm. Throughout the event, the ~~median infiltration and all single sampling locations~~ cumulative integrated throughfall curves of selected tipping bucket units as well as their medians follow the overall trend of gross precipitation (Figure 5~~). Early-stage infiltration is lower than precipitation (approx. 50% P_g) and indicates filling up of canopy and forest floor water storage. The cumulative infiltration~~

~~curves of all recording tipping bucket units~~ , upper panels). Measurement locations (tipping bucket units) across the Beech and Douglas fir plot show a higher spatial heterogeneity in ~~infiltration~~integrated throughfall depth ~~across the Beech and Douglas fir plot,~~ then at the Silver fir and mixed plot. ~~At several sampling locations across all plots infiltration during the event is low but steady with consistent hourly infiltration < 0.3 mm h^{-1} and an event infiltration < 6 mm. The increasing precipitation intensity on April 17th 2025 at 02:00 h to 07:00 h LT translates directly into increasing infiltration.~~



~~**Figure 5: Spatio-temporal infiltration variability during a precipitation event on April 17th 2025 (total event Pg = 11.2 mm, 28h) measured by infiltration samplers at four subplots Mixed (A), Douglas fir (B), Beech (C) and Silver fir (D) at the ECOSENSE field site**~~

Heat maps in Figure 5 (lower panels) visualize the intra-stand spatio-temporal variability of ~~event infiltration is visualized by heat maps of all tipping bucket units recording the selected event (Figure 5 lower panels):~~integrated throughfall: General spatio-temporal ~~infiltration~~ patterns resemble among plots with ~~infiltration~~J_I ranging from 0 to 3 mm h^{-1} across measurement points of few meters' distance. ~~Precipitation~~At several sampling locations across all plots integrated throughfall during the event is low but steady with consistent hourly J_l < 0.3 mm h^{-1} and an event J_l < 6 mm. Early-stage integrated throughfall during the first hours of the event is lower than precipitation (approx. 50% P_g). Increasing precipitation intensity on April 17th 2025 at 02:00 h to 07:00 h LT appears to translate after canopy redistribution (TF_c + TF_u) and litter modulation (I_l) into increasing integrated throughfall intensities during these hours. These periods of $P_g \geq 0.75$ mm h^{-1} (April 17th 2025, 02:00 to 07:00 h & 17:00 h LT) generate ~~peak infiltration~~peaks (≥ 3 mm h^{-1}) at some locations in the Silver, Douglas fir- and Beech plot~~,~~. However, the majority of tipping bucket units in all four plots record ~~infiltration~~integrated throughfall up to 1 mm h^{-1}. In particular, the first 2 and the last 4 hours of the event characterize low ~~infiltration~~percolation depths in line with small gross precipitation depths.



~~Infiltration~~

Figure 5: Spatio-temporal variability of J_l) during an event on April 17th 2025 (Pg = 11.2 mm, 23h) at four subplots Mixed (A), Douglas fir (B), Beech (C) and Silver fir (D) at the ECOSENSE research site. Black lines and box indicate the event (start to end).

3.2.3. Integrated throughfall variability of observation period

~~Figure 6~~Figure 6 compares the measured integrated throughfall (litter percolation, J_l) for selected events (total event $P_g \geq 2.5$ mm) that occurred during a two-week period end of April 2025, ~~in which~~when foliage ~~has been~~was already well developed. The total event precipitation ranged between 3.2 and 11.2 mm, with durations between 4 to ~~28~~23 h. Comparing the events, the median ~~infiltration~~throughfall coefficients (c_{tf}) for J_I increase with total ~~event precipitation~~ P_g for the Douglas fir, Mixed and less pronounced for the Silver fir plot. ~~Infiltration~~In contrast, throughfall coefficients of the Beech plot show no distinct relationship to event precipitation magnitude. Events of duration ≥ 10 h (event 2 & 6) show larger variability in ~~infiltration~~J_I across the single plots than events of duration ≤ 10 h (event 1, 3, 4, 5). This ~~larger~~increased heterogeneity is ~~visible at the boxplot~~ reflected in larger interquartile ranges (IQRs of 0.9 to 1 for ~~C_{inf} (~~c_{tf}) for event 2~~,~~ at Beech ~~&~~and Douglas fir plot~~)~~ ~~together with~~. It is accompanied by several locations with a ~~C_{inf}~~c_{tf} up to ~~2.5~~3.8 (scatter) and a large number of locations with low or no ~~infiltration~~percolation measured.

~~The influence of event duration on spatio-temporal variability becomes more evident comparing event 1 & 6. Both~~For events ~~of~~ 1 & 6 of equal precipitation depths (approx. P_g = 7 mm ~~differ in~~) but different duration (4 to 11 h~~).~~), the median ~~infiltration coefficients~~c_{tf} values are similar for all species groups ~~for the two events, however~~. However, the longer event 6 characterizes a stronger ~~infiltration~~ variability of J_I shown by the large IQR of Beech, Douglas fir and mixed boxplot quartiles expanding from 0 to 0.9 ~~C_{inf}~~c_{tf}. In contrast, the shorter event 1 appears to produce spatially ~~a~~ comparably homogenous ~~infiltration~~J_I (comparably narrow IQRs of < 0.5 ~~C_{inf}~~c_{tf} with shorter upper whiskers and few locations of zero total event ~~infiltration) due to higher~~ throughfall). Event-specific differences in integrated throughfall must also be interpreted in light of antecedent wetness conditions and phenological changes. The mean ~~event precipitation intensities causing fast canopy saturation~~leaf area index (LAI) varied between 11th April 2025 and ~~limited interception loss~~8th May 2025 due to the ongoing growing season in April 2025. Mean LAI increases from 3.79 to 5.2 at the Beech plot and from 4.55 to 4.86 at the Mixed plot, while it decreases from 6.18 to 4.64 at the Douglas fir plot. These temporal and inter-plot differences in LAI likely influenced interception and throughfall redistribution and should be considered when comparing c_{tf} across events and species.

~~4.~~ Discussion

~~4.1.~~4.1 Performance of ~~infiltration~~ integrated throughfall measurement sampler

The ~~throughfall~~ dataset included to this technical note originates an early measurement period in which in falling leaves and the unfragmented uppermost litter were retained inside the samplers (configuration II). This configuration is particularly robust, requires moderate maintenance and provides a novel method to measure the less frequently observed flux of integrated throughfall (J_I) at high spatio-temporal resolution.

Data obtained with this configuration in Figure 5 & Figure 6 demonstrate the capability of the FluxIT sampling network to capture spatio-temporal flux dynamics of throughfall components emerging from precipitation partitioning. The samplers measured ~~infiltration~~continuously for all events of the observed period (Figure 6~~) demonstrating~~), showing a consistent performance. ~~The samplers capture the~~ Pronounced spatial small-scale ~~infiltration~~ heterogeneity of integrated throughfall within ~~the~~and between plots was captured, as ~~the~~ illustrated by heat maps of Figure 5 and event ~~infiltration data in Figure 6 show: event infiltration~~ -based throughfall coefficients rangeing from 0 to 1 ~~C_{inf} across all four plots.~~ c_{tf} (Figure 6). Despite modulation by litter, the distinct patterns of depleted or augmented ~~infiltration~~integrated throughfall positions ~~across the plots~~

become morewere evident observing single sampler positions in the heat maps of Figure 5. Infiltration differencesat the level of individual tipping bucket units (heat maps Figure 5). Differences of J_l among neighbouring collection compartments highlight the small-scale variability and validate the sampler (compartment) support.

Sampling of small precipitation events remains an experimental challenge due to high spatial heterogeneity emerging from interception processes and generally small water amounts. Small events require large sampling sizes (Levia and Frost, 2003; Price and Carlyle-Moses, 2003; Rodrigo and Àvila, 2001), prior studies suggested > 200 samplers to measure throughfall from small precipitation events (Zimmermann et al., 2010). Although this technical note includes events of total $P_g > 2.5$ mm (see Figure 6), several smaller events- mostly under pre-wetted conditions - were successfully recorded by a substantial fraction of tipping bucket units, demonstrating the networks' capacity to capture events of $P_g < 2.5$ mm.

Configured for isolated throughfall measurements (configuration I), the monitoring system and derived data are directly comparable to traditional throughfall measurement approaches. The FluxIT system provides a flexible, cost-effective alternative that enables the distribution of a large sampler size (60 samplers, collection area of 11.4 m^2) (Figure 3 right-side panels) comparable to large-scale trough systems (Germer et al., 2006; Mużyło et al., 2012; van Stan et al., 2017) while retaining high spatial resolution (collection area of 600 cm^2). This way, the samplers cover the entire stand variation in a monitoring network across the four forest plots (Figure 3 right-side panels) while retaining high spatial resolution and avoiding to integrate short-range variability with a large collection unit. It is comparable with short-range spatial throughfall variability measurement approaches of e.g. Carlyle-Moses (2004), The cost-effective design allowed the distribution of a large sampler size (60 samplers) (Figure 3 right-side panels) with a total collection area of 11.4 m^2, which aligns with some of the large-scale trough systems (Germer et al., 2006; Mużyło et al., 2012; van Stan et al., 2017). In contrast, the spatial resolution of the infiltration measurement system (collection area of 600 cm^2) compares to sampling approaches of short-range spatial throughfall variability e.g. Carlyle-Moses(2004), Gerrits et al. (2010), Macinnis-Ng et al.Macinnis-Ng et al.(2012) .(2012). This way, the samplers cover the entire stand variation in a monitoring network across the four forest plots (Figure 3 right-side panels) while retaining high spatial resolution and avoiding to integrate short-range variability with a large collection unit.

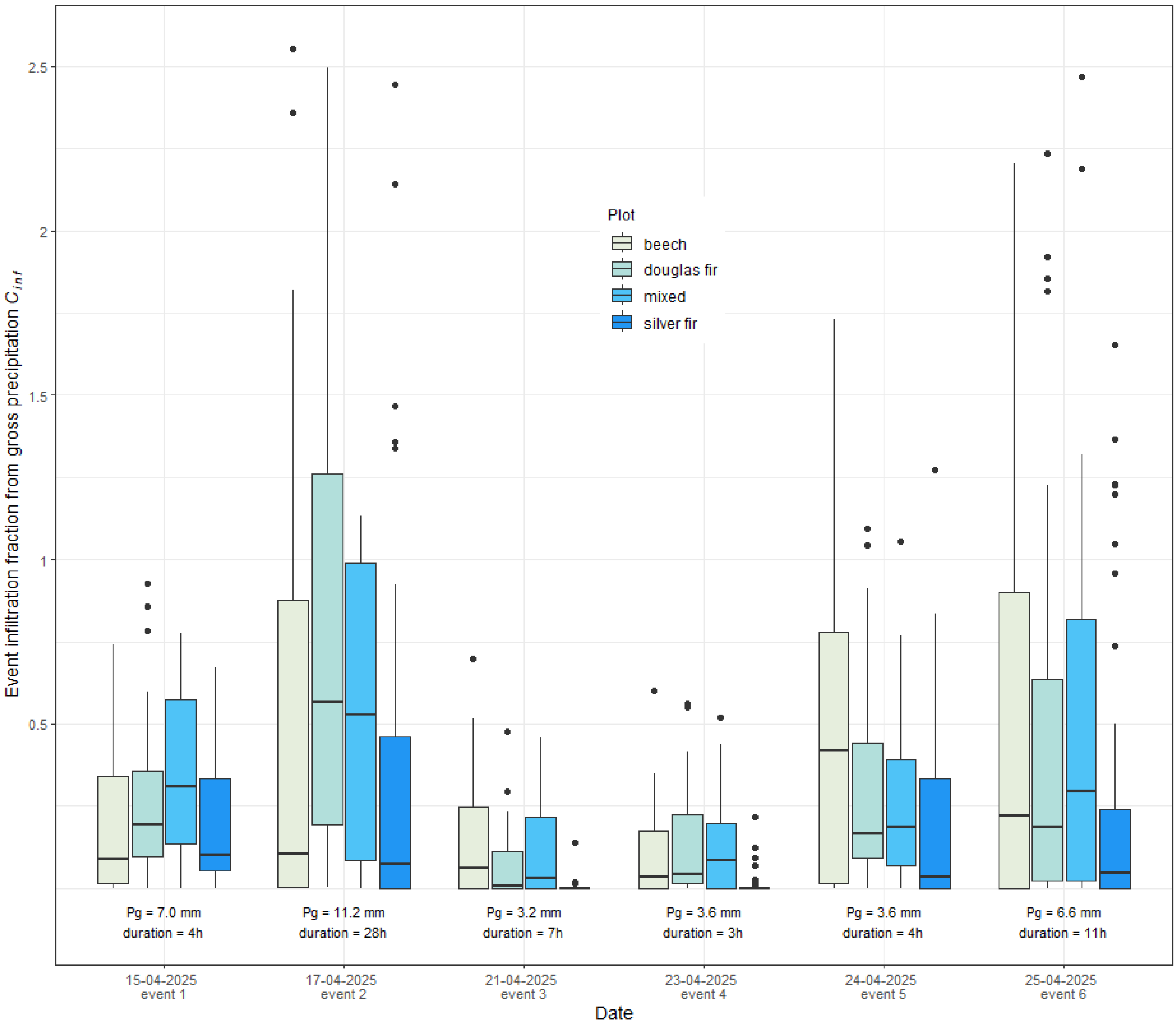


Figure 6: Event infiltration measured by the infiltration sampler network during two weeks in spring 2025; boxplots show total event infiltration at the four subplots in relation to event Pg for all events of total Pg > 2.5 mm

The example event (Figure 5) highlights the samplers' ability to depict the temporal infiltration variability: the continuous measurements of hourly infiltration reflect the rain intensity profile of the event with partly lagged, attenuated or depleted peaks (Dunkerley, 2015). Sampling schemes with automated, continuous readings of throughfall are rare and mostly limited in sampler size (n < 50) (Mużyło et al., 2012; Staelens et al., 2008; van Stan et al., 2017) making them unsuited for investigating spatial throughfall characteristics. After the first months of operation, automated sampling of 240 tipping bucket units proved to considerably reduce the sampling effort compared to manual sampling approaches of similar sampler sizes (Fischer et al., 2023; Metzger et al., 2017; Zimmermann et al., 2009). The samplers are insensible to wind and the measured water infiltrating into the soil enables nearby soil moisture measurements. The reduced logistical and financial costs enables monitoring beyond typical timeframes of several weeks (Keim et al., 2006; Klos et al., 2014; Raat et al., 2002), a vegetation period (Cisneros Vaca et al., 2018; Molina et al., 2019; Su et al., 2019) or one to two years (Holwerda et al., 2006; Mużyło et al., 2012; Siegert et al., 2019).

Infiltration data from the observed period (3.2.3 & The network further benefits from automated, continuous readings of 240 tipping bucket units, a feature rarely realized for a large sampler size (n < 50) (Mużyło et al., 2012; Staelens et al., 2008; van Stan et al., 2017) and suited for investigation small-scale spatio-temporal variability. The automated sampling of 240 tipping bucket units considerably reduce the sampling effort compared to manual sampling approaches of similar sampler sizes (Fischer et al., 2023; Metzger et al., 2017; Zimmermann et al., 2009). The samplers are insensible to wind and the measured water infiltrating into the soil enables nearby soil moisture measurements or even water quality measurements. The reduced logistical and financial costs enables monitoring beyond typical time frames of several weeks (Klos et al., 2014; Raat et al., 2002), vegetation period (Cisneros Vaca et al., 2018; Molina et al., 2019; Su et al., 2019) or one to two years (Holwerda et al., 2006; Mużyło et al., 2012; Siegert et al., 2019).

4.2. Limitation of further research directions

This technical note introduces the FluxIT monitoring network and illustrates selected application options; it does not constitute a comprehensive interception study. The presented dataset is limited in temporal extent and is not directly comparable to traditional throughfall studies based solely on isolated canopy throughfall. A forthcoming, dedicated throughfall study will include an expanded dataset collected during the 2025 vegetation period and measurements under sampler configuration I (isolated throughfall TF_c see 2.1.1 & 2.1.3 combined with lysimeter, stemflow and trough-based throughfall measurements..We than address the here described limitations in a detailed investigation of spatio-temporal patterns of isolated throughfall, their temporal persistence, intra-event dynamics and controls.

Under configuration II, measured integrated throughfall reflects combined effects of canopy and understory redistribution as well as litter interception. Although litter modulates throughfall patterns, the observed spatial variability in J_l (3.2.3 & Figure 6) was pronounced, with some locations recording zero or negligible J_l while others exceeded 250% P_g (Figure 6) reveal that several tipping bucket units recorded none or low infiltration for some events, while other tipping bucket units measured event infiltration up to 250% of P_g (Figure 6). These measurements may reflect strong spatial heterogeneity of throughfall, interception and infiltration as a result of precipitation redistribution and funnelling within canopy and forest floor (Carlyle-Moses and Lishman, 2015; Holwerda et al., 2006; Zimmermann et al., 2009). Records of zero infiltration may inform about a blocked tipping bucket unit, but may also be a valid, valuable information of highly depleted or even absent infiltration location (Levia et al., 2019). Likewise, highly elevated infiltration might correspond to a dripping edge position but could also arise from technical issues e.g. an unlevelled sampler or the displacement of the forest floor inlay (result is direct throughfall measurement without forest floor interception). The field application and manual construction of the samplers introduce potential sources of error e.g. water leakage from the tipping bucket units or material failure. All potential measurement errors can only be resolved with regular maintenance and calibration that was carried out to mitigate these issues, in particular before and after large events or during periods of freeze, snow or heat stress. In addition, the collected data undergo a quality control which includes also precipitation, throughfall and air temperature data and flags suspicious measurements (e.g. infiltration without precipitation or snow melt). As the dataset of collected events expands, we can improve the distinction between valuable information and measurement errors.

Monitoring small to large precipitation events will contribute to understanding the controls of throughfall and infiltration generation at different positions (e.g. species-specific thresholds of canopy and forest floor saturation). In particular the sampling of small precipitation events (Figure 6) remains an experimental challenge due to high spatial heterogeneity emerging from interception processes and generally small water amounts. Small events require large sampling sizes (Levia and Frost, 2003; Price and Carlyle-Moses, 2003; Rodrigo and Àvila, 2001), prior studies suggested > 200 samplers to measure throughfall from small precipitation events (Zimmermann et al., 2010). This study includes events of total P_g > 2.5 mm (see). The spatial variability of throughfall from canopy drip and understory vegetation across the plots appears to be sufficiently strong to (partially) persist the modulation by the thin, homogenous, unfragmented, uppermost litter layer. Consequently, the observed variability in integrated throughfall (litter percolation after litter interception) likely reflects, at least in part, canopy throughfall pattern and thereby underlines the systems' ability to depict small-scale spatio-temporal dynamics.

Even though the heat maps of example event (Figure 5) demonstrates the samplers' capability to depict temporal integrated throughfall variability, its remains to further investigation how the measured integrated throughfall reflect the rain intensity profile of the event with partly lagged, attenuated or depleted peaks (Dunkerley, 2015).

The measurements may not only reflect strong spatio-temporal variability of interception fluxes as a result of precipitation redistribution and funnelling within canopy and litter (Carlyle-Moses and Lishman, 2015; Holwerda et al., 2006; Zimmermann et al., 2009). Zero throughfall records may inform about a blocked tipping bucket unit, but may also represent valid, valuable information of strongly depleted or absent throughfall (Levia et al., 2019). Likewise, highly elevated, throughfall might correspond to a dripping edge position but could also arise from technical issues e.g. an unlevelled sampler or the displacement of the litter inlay (result is direct throughfall measurement without litter interception). The field application and manual construction of the samplers introduce potential sources of error e.g. water leakage from the tipping bucket units or material failure. These uncertainties are addressed with regular maintenance and calibration carried out in particular before and after large events or during periods of freeze, snow or heat stress. In addition, the collected data undergo a quality control which includes also precipitation and air temperature data and flags suspicious measurements (e.g. throughfall without precipitation or snow melt). As the dataset expands, the distinction between valuable information and measurement errors will improve.

Integrated throughfall generally ceased within approximately 2 hours after the event. Occasional single tips recorded (approx. 0.05 mm each) 3 hours or later were typically attributed to minor additional rainfall input rather than delayed litter percolation. Laboratory tests using site-specific litter support these observations, showing percolation durations of max. 1.5 hours under both dry and pre-wetted conditions. Nevertheless, future analyses will test different event separation thresholds of 4-8 hours using the extended dataset.

While previous throughfall studies report larger events (> 5 mm) are generally associated with larger throughfall volumes, reduced spatial variability (Raat et al., 2002) and consequently precipitation-controlled, small events (< 5 mm) were found to be more strongly controlled by vegetation properties such as canopy storage capacity (Gerrits et al., 2010). Only the extended dataset comprising diverse precipitation events throughout multiple seasons will clarify controls of throughfall fluxes and their contribution to spatio-temporal variability. Preliminary comparisons (Figure 6~~), however several smaller events – mostly under pre-wetted conditions - were recorded by a substantial fraction of tipping bucket units demonstrating the networks' capability to sample events of P_g < 2.5 mm. Precipitation characteristics control throughfall and consequently infiltration: In comparison of event 1 & 6 in Figure 6, event duration, depth and intensity appear to influence infiltration. While larger events (> 5 mm) are generally associated with large throughfall volumes (and consequently infiltration) and reduced spatial variability (Raat et al., 2002), small events (< 5 mm) were found to be more affected by vegetation parameters such as canopy storage capacity (Gerrits et al., 2010). The largest sampled event (Figure 6 event 2) shows infiltration during long periods of low precipitation intensity, potentially influenced by interception e.g. "within event" evaporation (Crockford and Richardson, 1990) reflecting the uncertain role of precipitation intensity (Raat et al., 2002).~~

~~Further improvements~~ event 1 & 6) suggest that event duration, intensity, and antecedent wetness influence integrated throughfall variability. The role of "within event" evaporation (Crockford and Richardson, 1990), canopy and and litter storage dynamics, and species-specific effects remains to be quantified.

~~4.2.~~ 4.3. Perspective modifications for the ~~infiltration measurement~~FluxIT sampler monitoring network

~~Tipping bucket rain gauges implemented in this setup (2.1.1 & 3.1) are widely used for quantifying precipitation, throughfall or stemflow due to their simple principle of operation and cost-effectiveness. However, they suffer from systematic non-linear measurement errors depending on precipitation intensity (Marsalek, 1981; World Meteorological Organization, 2021). In particular measurements of low and high intensities are prone to errors. A correction function will be considered to address this issue for the infiltration measurement sampler e.g. Colli et al.(2013), Shimizu et al.(2018) and Stagnaro et al.(2016).~~

~~As described for the example event in 3.2.2 the infiltration trend during the event compared to the precipitation profile indicate forest interception processes in form of canopy and forest floor water storage filling in the initial event phase. However, the samplers' potential to monitor canopy and forest floor interception in detail is limited e.g. to comparisons of data from leafed and non-leafed periods or dry and pre-wetted canopy conditions. Following the steps of this research include the operation of the samplers without forest floor inlays to collect direct canopy throughfall (see 2.1.1 & 2.1.2) and a combined analysis of infiltration data with available data of lysimeters, stemflow and throughfall of the same plots. Together, this will enable a distinctive analysis of spatio-temporal throughfall patterns in relation to forest floor percolation. In order to further investigate spatio-temporal infiltration heterogeneity, the presented infiltration measurement samplers (2.1.1~~Beyond the two primary configurations presented (2.1.3), the following modifications of the FluxIT sampler are feasible.

All components – including drainage compartments volume and size, pipe diameter and tipping bucket volumes – can be flexibly adapted using modular Euro-containers and 3D-printing parts. The selected dimensions of this setup were selected to match prevailing rainfall intensities at the study site but can be scaled to accommodate more extreme rainfall regimes. This scalability supports also the application of this monitoring approach to drier or wetter climates or more complexly structured forest ecosystems.

Alternative mounting on stainless-steel frames or PVC pipes proposed in the section 2.1.3 allow flexible positioning to isolate e.g. understory vegetation influence. These options are less invasive than porous concrete stone frames and reduce potential alteration of flow paths and soil disturbance around the samplers. In order to further investigate spatio-temporal throughfall heterogeneity or the less frequently observed litter interception dynamics, the presented throughfall measurement samplers (2.1.1 & 0) can be extended to fully equipped mini-lysimeter (Gerrits et al., 2007; Paulsen and Weiler, 2025). To achieve this, every sampler requires two load cells positioned beneath the short sides of the container on e.g. a stainless-steel frame. Micro boards and software can be updated accordingly, see Paulsen and Weiler (2025).

Tipping bucket rain gauges implemented in this setup (2.1.1 & 3.1) are widely used for quantifying precipitation, throughfall or stemflow due to their simple principle of operation and cost-effectiveness. However, they suffer from systematic non-linear measurement errors depending on precipitation intensity (Marsalek, 1981; World Meteorological Organization, 2021). In particular measurements of low and high intensities are prone to errors. In our case, the sampler design partially mitigates this issue. The pipes connecting the drainage compartments to the tipping bucket units have an inner diameter of 3 mm, which limits the inflow rate. During high-intensity rainfalls incoming water is collected in the sampler, funnelled and temporally stored in the four drainage compartments that function as a buffer emptying at 0.4 L min^{-1}. To evaluate under- and overestimation for lower and higher intensities, we conducted a dynamic calibration on a subset of the tipping bucket units (n = 10). At highest tested rainfall intensity of 120 mm h^{-1}, in average 27.83 tips per 100 mL were recorded. Using the mean tip volume from static calibration (2.87 mL), this indicates an undercatch of approximately 23% at very high intensity rainfalls. During 1.5 years of observation, rainfall intensities at the field site exceeded 120 mm h^{-1} for 11 minutes (0.015% of total rainfall duration in 1.5 years) and 40 mm h^{-1} for 2.4 hours (0.2% of total rain duration). The resulting total underestimation from extreme events is less then 0.05%, which we consider acceptable. The recently collected, larger dataset from the 2025 vegetation period will enable the evaluation of over- and underestimation of the tipping bucket units for rainfall events of varying intensities. Furthermore, correction functions will be considered to address this issue for the throughfall measurement sampler e.g. Colli et al. (2013), Shimizu et al. (2018) and Stagnaro et al. (2016).

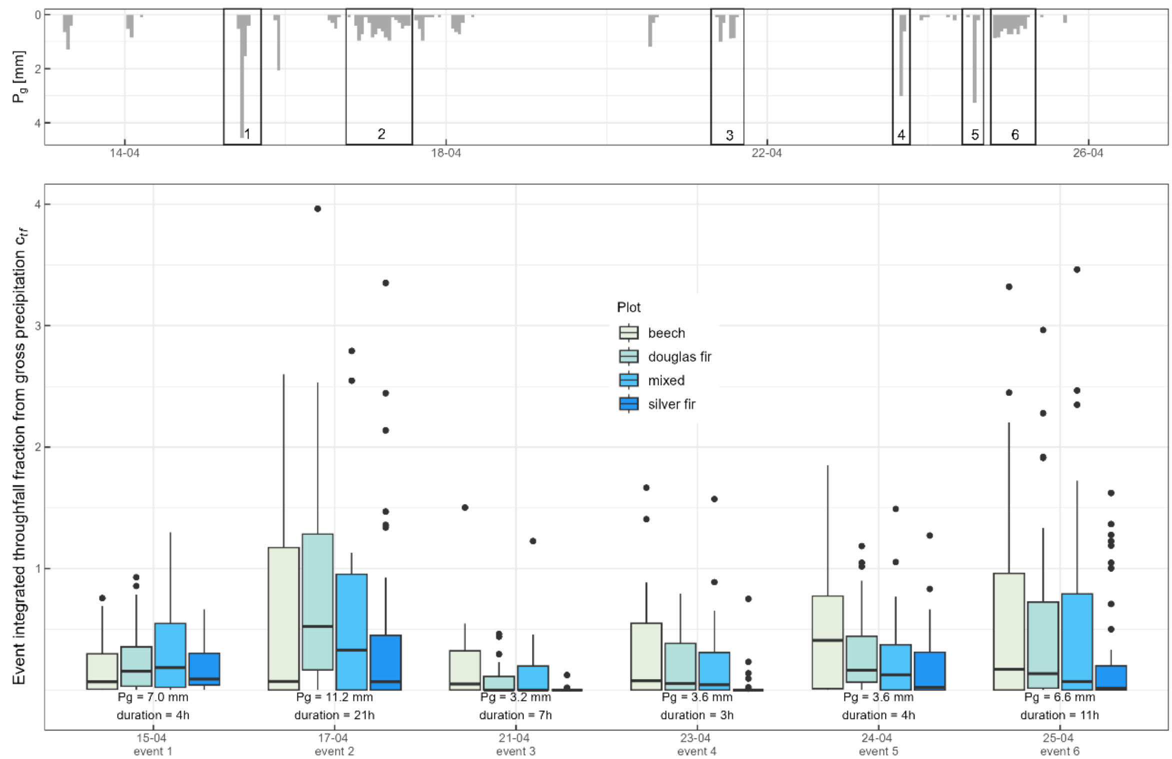


Figure 6: Event integrated throughfall measured by the sampler network during two weeks in spring 2025; boxplots show total event integrated throughfall at the four subplots in relation to event *Pg* for all events *Pg* > 2.5 mm. Black boxes indicate events 1-6

~~5.~~4. Conclusion

The presented ~~throughfall~~FluxIT monitoring network at the ECOSENSE forest research site provides a robust, flexible, and high-resolution approach for quantifying throughfall fluxes in forest ecosystems. It enables automated, continuous measurements of spatio-temporal throughfall flux dynamics across subplots of pure Beech, pure Douglas fir, pure Silver fir and mixed trees. Comprising 60 samplers with a total of 240 collection compartments and tipping bucket units, the network achieves a high spatial resolution at tree-scale while capturing the full range of throughfall variability across the plots through a stratified sampling design.

The presented example data from an observation period in spring 2025 demonstrate the system measures continuously events of all magnitudes with sensitivity to short-range spatial throughfall heterogeneity and temporal trends during events. Furthermore, the samplers are able to capture both integrated throughfall (litter percolation after litter interception) and, in alternative configurations, isolated fluxes such as canopy throughfall small-scale spatio-temporal variability within and across forest plots. The network complements conventional throughfall and stemflow measurements and offers as a flexible, modular system a novel method to monitor less frequently measured fluxes such as litter percolation. With its automated, minimally invasive operation at low maintenance it considerably reduces the sampling effort compared to manual sampling methods making it well-suited for throughfall monitoring over several years. The growing dataset of high-resolution throughfall measurements from plots of coniferous and deciduous trees will enable the detailed analysis of species-specific throughfall dynamics, the within-plot variability and temporal stability of patterns. It will also provide the basis for investigating how spatial throughfall patterns propagate through percolation into soil water content patterns. Data from longer operation periods will furthermore support the investigation of seasonality effects and forest age-related changes (increasing canopy complexity) on throughfall and ~~infiltration.~~integrated throughfall. Ultimately, the network and generated data contribute to our understanding of precipitation partitioning in forests especially the role of vegetation structure shaping throughfall and interception processes.

Code/Data availability. Code ~~will be~~is available on Github~~,~~ together with documentation of sampler design (https://github.com/Lea-Dedden/Infiltration-sampler/tree/main). Data are available upon request.

Author contributions. LD and MW designed the ~~infiltration~~integrated throughfall measurement setup. LD conducted the field set-up and measurements with support of the team of Hydrology named in the Acknowledgements. LD analyzed the data and drafted the first version of this manuscript with contributions from MW.

Competing interests. At least one of the (co-)authors is a member of the editorial board of Hydrology and Earth System Sciences.

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