Spatial and Temporal Variability of Throughfall under a Natural Fagus orientalis Stand in the Hyrcanian Forests, North of Iran

A. Dezhban1, P. Attarod1*, G. Zahedi Amiri 1, T. G. Pypker 2, and K. Nanko3

ABSTRACT

The spatial and temporal heterogeneity of throughfall (TF) have important ecological impacts in forest ecosystems. The aim of this study was to quantify spatio-temporal variability of TF and to evaluate the effects of canopy traits and gross rainfall (GR) characteristics on TF at the event scale. Event-based measurements were carried out from September 2015 to October 2017 during the leafed-out period in a natural uneven-aged beech (Fagus orientalis L.) stand located in the Hyrcanian forest of Iran. Leaf area index (LAI) and canopy openness of the stand were 6 m²·m⁻² and 6.2%, respectively. Tree density in the studied plot was 188 tree ha⁻¹ and the basal area (BA) was 51 m²·ha⁻¹. During the measurement period, 25 rainfall events occurred (total rainfall= 784.8 mm). We observed variability of TF under the beech trees canopy in different GR classes (< 15, 15-30, 30-50 and > 50 mm). Increases in rainfall depth and intensity were associated with an increase in TF depth and decrease in TF variability. We found that rainfall depth along with the intensity were the most influential factors on the TF depth, spatial variability as well as time stability. Knowledge of the spatial persistence and variability of TF would help managers to optimize the management of these stands in terms of soil water and nutrition availability.

Keywords: Beech stand, Canopy trait, Ecohydrology, Gross rainfall, Tree density.

INTRODUCTION

Considerable quantities of precipitation transfer from the canopy to the ground through throughfall (TF) (Levia and Frost, 2006, Carlyle-Moses et al., 2004). The redistribution of gross rainfall GR as a result of interactions with the forest canopy creates a high spatial and temporal input of TF (Johnson and Lehmann, 2006, Zimmermann et al., 2008, Muzylo et al., 2009, Macinnis-Ng et al., 2012). TF spatial variability subsequently affects soil moisture variability and the spatial pattern of solute concentration and deposition on the forest floor (Zimmermann et al., 2007). Hence, understanding the heterogeneous patterns of TF beneath forests is critically important to ecological processes of canopy, ecohydrology, and biogeochemistry cycling in forest ecosystems (Keim et al., 2005).

Primary factors affecting the quantity of TF are species composition (Bouten et al., 1992, Lilienfein and Wilcke, 2004), meteorological conditions (Irome and Huber, 2002), season (Henderson et al., 1977), and canopy cover (Zirlewagen and Von Wilpert, 2001, Morris et al., 2003, Pypker et al., 2011). The impact of canopy cover on TF distribution in forest stands varies between stand types (Kimmins, 1973, Carlyle-Moses et al., 2004, Holwerda et al., 2006). In addition, rainfall magnitude,
duration, intensity as well as wind speed and direction, can affect spatial and temporal variability of TF at the forest floor (Vrugt et al., 2003, Levia and Frost, 2006, Staelens et al., 2006). Although numerous studies have focused on the spatial variability of TF in coniferous, broadleaved or mixed stands and tropical rain forests (e.g. Raat et al., 2002, Loescher et al., 2002, Konishi et al., 2006, Zimmermann et al., 2008, Huitao et al., 2012, Nanko et al., 2011, Klos et al., 2014, He et al., 2014, Fan et al., 2015), little information is available about TF patterns in deciduous broadleaved forests (Carlyle-Moses et al., 2004, Staelens et al., 2006, André et al., 2011, Kowalska et al., 2016).

The global average surface temperature has increased over the 20th century by about 0.6°C (Houghton et al., 1990). Over the next 20 years, climate change is predicted to reduce rainfall and increase the severity of droughts in the north and west of Iran (Alvankar et al., 2017). Oriental beech (Fagus orientalis Lipsky) forests are the dominant tree between 700 and 1,500 m elevation in Hyrcanian forests in the north of Iran with standing volumes of up to 800 m³ ha⁻¹ (Oladi et al., 2017). Air humidity is high and fog is frequent in this elevation band (Sagheb Talebi et al., 2014, Oladi et al., 2017). Climate change will likely alter GR characteristics and the phenology of trees thereby changing TF variability. To our knowledge, there has been no research quantifying the spatial variability of TF in oriental beech stands. To optimize the management of these natural beech stands, a better understanding of the mechanisms that control the spatial TF variability is required. In addition, a better understanding would help future investigators design better TF sampling strategies based on different GR characteristics and canopy cover behaviors. We hypothesized that: (i) The spatio-temporal variability of TF in natural beech stand is relatively high and, (ii) TF variability patterns are affected by canopy and rainfall characteristics. The objective of this study was: (i) To understand and quantify the variability of spatial TF under a natural F. orientalis stand in the Caspian forests of northern Iran, (ii) To examine the temporal stability of spatial TF patterns, and, (iii) To determine which measured canopy and rainfall characteristics are best correlated to TF variability.

**MATERIALS AND METHODS**

**Study Site**

The study was conducted in the Kheyrud forest research station of the University of Tehran, located in the central part of the Caspian forests, Mazandaran province, Iran (36° 35’ N, 51° 37’ E) (Figure 1). Due to lack of meteorological stations inside the Caspian forests, we reported the meteorological data recorded by Kojur (36° 38’ N, 51° 73’ E, elevation 1,550 m asl) and Nowshahr (36° 39’ N, 51° 30’ E, elevation -21 m asl) synoptic meteorological stations. Based on Kojur records, from 2006 to 2017, mean annual precipitation was 325 mm and mean annual relative humidity and air temperature were 60% and 12.8°C, respectively. Minimum and maximum monthly air temperatures were 1.5 (January) and 22°C (August). The prevailing wind direction is from the north (42%) and northeast (37%). When from the north, the 3-hr-average wind speed is 4.5 m s⁻¹. Moreover, Nowshahr Station, located approximately 22 km away from the study plot, reports mean yearly precipitation±SD is 1,291±184 mm and mean air temperature is 16.4°C (1977-2015). The region has a temperate mountainous climate that is affected by changing elevation (1400 to 3000 m) and distance from the Caspian Sea. Fog plays a important role in providing moisture to vegetation in this region (Garstecki, 2017). The study area experiences a short, warm, temperate and extremely humid summer that is marked by heavy fog for 4-8 hours per day during leaf-out seasons (Authors’ observations). The winter is long, cold with freezing temperatures.
Field measurements were carried out inside a 4,900 m² plot (36° 29′ N, 51° 39′ E, 1,476 m asl). Mean angle of the west-facing slope of the site ranged between 20 and 25%. The plot was dominated by an uneven aged stand of oriental beech trees with loamy nutrient rich soil. The stands are comprised of virgin forests that have not been harvested. About 10% of the understory was covered by *Ilex spinigera*, *Euphorbia helioscopia*, *Asperula odorata*, *Primula heterochroma*, *Sanicula europaea*, and *Viola odorata*, while the remaining area was bare ground. Tree density was 188 tree ha⁻¹ with a basal area BA of 51 m² ha⁻¹. Mean tree height and diameter at breast height DBH were 29.5 m and 51.2 cm, respectively. Maximum and minimum height were 52.1 and 5.2 m, and for DBH, 138.4 and 10.5 cm, respectively. Prevailing wind direction during the measurement period was south and southwest.

**Data Collection and Sampling Frame**

Event-based GR and throughfall (TF) measurements were carried out continuously during the leafed-out seasons (from May to November) when the leaves are fully developed, for two years (September 2015 to October 2017). In this study, GR is defined as the rainfall entering the top of the canopy, TF is the portion that reaches 35 cm above the ground through canopy gaps without colliding with the vegetation (free TF) or via water dripping from the leaves, branches and stems (released TF) (Dunkerley, 2000, Levia and Frost, 2003, Levia et al., 2017).

Each precipitation event was measured eight hours after the end of rainfall,
however, nighttime rainfalls were measured in the morning after sunrise (Carlyle-Moses et al., 2004). When the rain occurred over consecutive days, the rainfall was considered as an individual rainfall event (He et al., 2014). The duration of each rainfall event was calculated as the time between the first and last recorded measurement of an event. $GR$ was measured using 10 manual rain-collectors located in a clearing area less than 100 m from the border of the study plot. $GR$ manual rain-collectors were set 130 cm above the ground to prevent water drop splash from the ground and were wrapped in aluminum foil to minimize evaporation from increased radiation absorption. Manual $TF$ rain-collectors ($n=122$) were established under the canopy in a 70x70 m square plot (Figure 2). All low shrubs, grass, ferns, and herbs within 70 cm of a collector were removed. Manual $TF$ rain-collectors were set 35 cm above the ground to reduce entry of splash water and $TF$ rain-collectors were emptied prior to the rainfall event. The rain-collectors used for measuring both $GR$ and $TF$ had an orifice of 132.7 cm$^2$ (13 cm in diameter) and a height of 20 cm. $TF$ sampling points were distributed in a stratified simple random sampling design under the canopy to assess spatial distribution of $TF$. This method is more advantageous than grid and simple random sampling designs for spatial predictions because the stratified simple random sampling design provides more even distribution of sampling locations and produces unbiased estimates of the spatial mean (Zimmermann and Zimmermann, 2014). The 4,900 m$^2$ study plot was divided into 49 subplots, 10×10 m and two $TF$ sampling points were then selected randomly in each subplot (Figure 2). In addition, we randomly sampled an additional 24 points within the study plot. These 24 sampling points were placed in a random direction, 0.5 m away from the $TF$ sampling points already established in the subplot (Figure 2).

**Leaf Area Index (LAI) and Canopy Openness**

Leaf area index $LAI$ and canopy openness ($\%$) of the study plot were measured by Digital Hemispherical Photography (DHP, Canon EOS 6D digital camera with a 180°

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Figure 2. Throughfall $TF$ sampling layout. The study plot was divided into 49 subplots; open circles in each subplot indicate $TF$ sampling point. Crosses show the additional 24 sampling points.
Spatial and Temporal Variability of Throughfall

The variability of throughfall (TF) was indicated by the spatial coefficient of variation (CV, %) described by a series of studies (Seiler and Matzner, 1995; Ratt et al., 2002, Staelens et al., 2006). The most appropriate correlation analysis (Pearson or Spearman) based on data distribution was performed to test the relationship between TF amount, TF%, CV%TF and rainfall characteristics. Before statistical calculations, normal distribution was assessed using skewness and kurtosis, and the Kolmogorov–Smirnov statistic was used to test the normality of data. The points with TF depth more than 100 percent (hotspot points) for each GR class were detected using Getis Ord Hot-spot analysis in Arc GIS 10.2 at three significant levels (α= 0.10, 0.05 and 0.01) and using an Inverse-Distance spatial relationship concept (Getis and Ord, 1992). Data statistical analyses were carried out using statistical software IBM SPSS Statistics Version 25.0 and Microsoft Excel 2013.

RESULTS

Rainfall Characteristics

From September 2015 to October 2017, 25 rainfall events were measured with a cumulative rainfall of 784.8 mm (CV= 63.2%). The rainfall events were divided into four GR classes: < 15, 15-30, 30-50 and > 50 mm. There were 3, 12, 5, and 5 rainfall events at each rainfall class, respectively, which accounted for 2.9, 34.1, 21.7, and 41.3% of total rainfall events, respectively (Figure 3). Fifteen rainfall events (37.6%) were less than 30 mm (295.4 mm cumulatively) and the remaining (62.5%) exceeded 30 mm (490.6 mm cumulatively). Rainfall events ranging from 15 to 30 mm were the most frequent (48% of total events) and accounted for 34.1% of the cumulative rainfall during the study period. Small rainfall events (< 15 mm) were infrequent (12% frequency) and contributed only 2.9% of the total cumulative rainfall. Five large storms (> 50mm) were recorded but they accounted for 41.3% of the cumulative rainfall (Figure 3). In addition, rainfall

Data Processing

Time stability of TF spatial patterns was evaluated by standardizing the TF rain-collector amount for each GR event using the method proposed by Keim et al., (2005). The method was a modified equation of Raat et al. (2002), and Vachaud et al. (1985), so that the variance at the sampling point quantified the stability of high or low TF areas:

$$\delta_{i,E} = (\delta_{i,E} - \overline{\delta}_E)(s_\delta)_E^{-1}, \quad (1)$$

Where, $\delta_{i,E}$ and $\delta_{i,E}$ are the normalized variable and variable at sampling point $i$ of rainfall event $E$, respectively, $\overline{\delta}_E$ is the mean and $s_\delta_E$ is the standard deviation of the variables for all sampling points in the GR event. We used Equation (2) to analyze the time stability of TF spatial patterns but replaced $\overline{\delta}_E$ with the median of $\delta_E$, $M\delta_E$, and $s_\delta_E$ with the Median absolute deviation (MAD) of $\delta_E$, $MAD\delta_E$ as Equation (2) (Zimmermann et al., 2007):

$$\overline{\delta}_{i,E} = (\delta_{i,E} - M\delta_E)(MAD\delta_E)^{-1}, \quad (2)$$

The median and MAD were used because TF often does not exhibit a normal distribution at the rainfall event scale, and to control the influence of outliers on the dataset.

Statistical Analysis

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Figure 3. Relative frequency of rainfall events for four rainfall classes (white bars) and relative contribution of each rainfall class to the total rainfall amount (grey bars) during the study period from September 2015 to October 2017.

Canopy Characteristics (LAI and Canopy Openness)

The mean LAI and canopy openness of the study plot measured vertically above each TF rain-collector during the leaf-out period on summer were 6 m² m⁻² (CV% = 13.7) and 6.2% (CV% = 20.2), respectively. Spring had the mean LAI and canopy openness of 5 m² m⁻² (CV% = 14.5) and 7.6% (CV% = 17.7), respectively. The canopy parameters value in the fall were 4.5 m² m⁻² (LAI) (CV% = 14.1) and 7.9% (canopy openness) (CV% = 18.2). There was significant difference between LAI during the summer and both the spring and fall (independent t-test, t = 9.65, P < 0.001).

Spatial Variability of TF in Relation to Rainfall Characteristics

The mean TF depth derived from 122 rain-collectors was strongly correlated with GR (r² = 0.98, P < 0.01) (Figure 4-a) and significantly correlated with event-based mean rainfall intensity (P < 0.01), although this correlation was weak (r² = 0.3) (Figure 4-d). In contrast, the correlation with rainfall duration was weak and not significant (P = 0.08) (Figure 4-g). Generally, TF depth increased with increasing GR depth and rainfall intensity (Figures 4-a, -d, and -g). The relationship between TF% and GR, rainfall intensity and rainfall duration was not significant (P > 0.05) (Figures 4-b, -e, and -h). There was a significant correlation between spatial variability of TF (CV%TF) with GR depth (P < 0.01) (Figure 4-c) and rainfall intensity (P < 0.01) (Figure 4-f), whereas the correlation with rainfall duration was weak and not significant (P = 0.6) (Figure 4-i). Generally, CV%TF decreased and then stabilized with increasing GR depth and rainfall intensity (Figures 4-c and -f).
Spatial and Temporal Variability of Throughfall

Figure 4. The correlation between mean values of throughfall (TF), TF%, CV%TF with (a, b, c) gross rainfall (GR) amount, (d, e, f) rainfall intensity, (g, h, i) and rainfall duration for individual rainfall events. Bars indicate ± standard error.

CV%TF for each GR class was highest (48.1%) when the GR amount was lower than 15 mm (< 15 mm), and decreased to 33, 22.2 and 19.8% for GR classes of 15-30, 30-50 and > 50 mm, respectively (Table 1). The points with TF depth more than 100 percent (hotspot points) were highest for GR classes 15-30 and 30-50 mm and lowest for GR classes < 15 and > 50 mm (Table 1) that were more concentrated on the center of plot for all GR classes (Figure 5) where the large DBH trees were more frequent than the other parts of the study plot.

Temporal Persistence of Spatial Variability of TF

Time stability normalized TF plots indicate persistence of drier and wetter sampling points in the study area (Figure 6). Time stability plots using normalized TF were divided into four GR classes based on 20 mm thresholds. Each of these thresholds contained the same number of rainfall events. Mean normalized TF for GR classes < 20, 20-30, 30-50, > 50 mm were significantly different from zero for 48, 42, 22 and 19% of the TF rain-collectors, respectively (t-test; α= 0.05), thus TF at individual sampling positions were not randomly distributed over time. With greater GR, the water flow paths in the canopy changed. For example, collector 103 had smaller TF with smaller GR but it received greater TF when the GR was larger (Figure 6). In contrast, collector 3 had greater TF when GR < 20 mm but it had much smaller TF in GR classes 20-30, 30-50 and > mm 50 (Figure 6).
Table 1. Mean, Standard deviation SD, Coefficient of variation of throughfall (CV%TF) and rainfall intensity for 122 TF rain-collectors for different gross rainfall GR classes.

<table>
<thead>
<tr>
<th>GR classes (mm)</th>
<th>&lt; 15</th>
<th>15-30</th>
<th>30-50</th>
<th>&gt;50</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mean (mm)</td>
<td>7.6</td>
<td>22.3</td>
<td>34</td>
<td>64.9</td>
</tr>
<tr>
<td>SD (mm)</td>
<td>2.4</td>
<td>7.1</td>
<td>7.5</td>
<td>13.2</td>
</tr>
<tr>
<td>CV%TF</td>
<td>48.1</td>
<td>33</td>
<td>22.2</td>
<td>19.8</td>
</tr>
<tr>
<td>Rainfall intensity (mm h^{-1})</td>
<td>0.4</td>
<td>0.7</td>
<td>1.1</td>
<td>2.1</td>
</tr>
<tr>
<td>No of Hotspot point</td>
<td>5</td>
<td>12</td>
<td>9</td>
<td>5</td>
</tr>
</tbody>
</table>

Figure 5. Spatial distribution of hotspot points under the oriental beech canopy for GR classes. Red, green and blue squares are hotspot points at three significant levels.
Figure 6. Time stability plots for normalized throughfall (TF) water amount to zero median in gross rainfall (GR) classes (a) < 20 mm (events 7, 8, 12, 13, 15, 17, 18, 19), (b) 20-30 mm (events 4, 5, 6, 16, 20, 21, 25), (c) 30-50 mm (events 1, 2, 9, 10, 11), and (d) > 50 mm (events 3, 14, 22, 23, 24). Each dot represents one TF rain-collector plotted along the horizontal axis and ranked by their means. Bars indicate 95% confidence intervals. Bars of for each data point indicate the stability of TF rain-collectors. Data points with bars that do not overlap with the median line (horizontal line at zero) are extreme persistent whereas data points with bars that intersect the median line are general persistent (Zimmermann et al., 2008)
DISCUSSION

Spatial Variability of TF

The increase of TF production with increasing rainfall depth may result from longer periods when the canopy is saturated, thereby increasing the amount of rainwater falling to the ground as TF (Carlyle-Moses et al., 2004). For smaller rainfall events, the interaction of rainwater with forest canopy likely restricts TF to free TF (without contacting the canopy). Released TF (dripping from leaves and branches) becomes dominant when GR is larger than the canopy saturation point (Zhang et al., 2016, Gash, 1979). Similar to past research (Loustau et al., 1992, Zhan et al., 2007, Huitao et al., 2012, Fathizadeh et al., 2014, Fan et al., 2015), the variability of TF (CV%TF) decreased with increasing GR (Figure 4-c).

Past work on the relationship between spatial variability of TF (as CV%) and cumulative rainfall suggests that larger rainfall events would have a more homogenous redistribution of rainfall (e.g., Loustau et al., 1992, Huitao et al., 2012, Li et al., 2013, Carlyle-Moses et al., 2004, Fan et al., 2015). In addition, the lower variability of TF for large GR may be attributed, in part, to a greater GR intensity that saturates the canopy in a short time (Zhan et al., 2007). Typically, as rainfall intensity increases, the CV of TF initially declines and then stabilizes (Zhang et al., 2016).

In the present study, the CV%TF of GR events < 15 mm was highest and for GR events > 50 mm was lowest. Both GR events > 50 and < 15 mm were more frequent in fall season with fairly similar LAI. It shows that the impact of canopy characteristics (e.g. spatial variability of canopy storage) is more diverse among different rainfall regimes. When the canopy is fully saturated for higher GR, both free TF and released TF participate in TF production with lower variability.

Input points producing hotspots were more common for GR classes 15-30 and 30-50 mm, whereas they decreased in GR classes < 15 and > 50 mm. Our results showed that 67% of rainfall event from 15 to 30 mm and 60% of rainfall event from 30 to 50 mm occurred in summer. On the other hand, 100% of rainfall events larger than 50 mm and 67% of rainfall smaller than 15 mm occurred in the fall. Therefore, the hotspots were likely related to both changes in rainfall and canopy characteristics. Yousefi et al. (2018) reported that hotspots were more frequent in leaf-out period compared with leaf-off condition. Hotspots in forest ecosystem that are dependent on TF tend to enhance drainage to the forest floor (Coenders-Gerrits et al., 2013, Hopp and McDonnell, 2011). Hence, the whole leaves and branches of the canopy may direct rainfall downward to allow for easy drainage of the TF during fully leafed conditions.

Time Stability of TF Spatial Variability

Mean normalized TF for 55% of the TF rain-collectors in the present study were significantly different from zero, thereby indicating high TF heterogeneity during the leafed period. Zimmermann et al. (2007) reported 68% TF heterogeneity in a tropical montane forest in Ecuador. Keim et al. (2005) showed 31-46% of collectors represented mean normalized TF different than zero in two conifer stands and one deciduous stand at leafed and leafless periods in the Pacific Northwest, USA. We found that the percent of TF rain-collectors with mean normalized TF different from zero was highest for GR class < 20 mm (48%) and lowest for GR class > 50 mm (19%), indicating TF heterogeneity decreases with increase in rainfall, supporting the claim of lower TF variability (CV%TF) for large rainfall events. Gómez et al. (2002) reported that the distribution of TF for large GR events were consistent, while there was no consistency of TF distribution in the smaller GR events under
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individual olive trees. Time stability plots indicated the switching of wet and dry collectors with changing GR class (e.g. collectors 3 and 103) might result from a change in the TF flow path. Therefore, changing GR under climate change or different seasons can affect TF flow path under the canopy.

Our results indicate that there was high temporal stability and low spatial variability of TF for high GR classes. The decrease in temporal stability and lower spatial variability suggests that canopy characteristics are less important as GR increases. Robson et al. (1994) suggested the effect of event size was possibly as important as variation in canopy structure characteristics on the stability of the TF spatial pattern. However, we found that fully developed canopy cover were more effective in creating hotspots of TF than different GR depth.

Changing climate has significantly altered GR characteristics around the globe (Huntington, 2006). Climate-change-induced effects on TF variability are especially important to forest ecosystems in the north of Iran. These forests have steep mountain slopes that are sensitive to soil erosion when water is locally in excess and forest regeneration partially depends on TF inputs (Akbarzadeh et al., 2016). Changing precipitation patterns may affect the spatial variability of TF and, therefore, affect ecological processes throughout a forest stand (Manderscheid and Matzner, 1995, Raat et al., 2002). Indeed, the spatial heterogeneity of TF may improve the calculations of percolation fluxes and the downward movement of water through the soil, in these beech stands (Beier, 1993). Generally, studies on the spatial and temporal variability of TF and its associated driving factors will increase our understanding of hydrologic and biogeochemical cycles of forest ecosystems that will permit better stewardship of forest and water resources (Levia and Frost, 2006).

TF is an important water source for trees. A changing climate and/or tree harvesting may lead to changing TF distributions, ultimately altering the spatial distribution of water input to the soil. Therefore, fully understanding the spatial distribution of rainfall within forests under different GR depth helps the managers to optimize the management of these stands in terms of soil water and nutrition availability. Also, this study would help future investigators to design sampling strategies of TF based on different rainfall regimes by taking a look at canopy in these natural beech stands under similar climate conditions.

CONCLUSIONS

The spatial variability and temporal stability of TF as well as its correlation with rainfall characteristics and canopy traits were analyzed under a natural beech stand. In terms of rainfall characteristics, the spatial variability of TF tended to be high for the GR size class lower than 15 mm and decreased with increasing GR. During larger rainfall events, the canopy was fully saturated, resulting in decreasing of spatial variability of TF. The size and the variability of TF were more correlated to rainfall intensity and the size of GR. Based on time stability plots, the observed spatial variability was temporally stable for each GR class and temporal stability increased with increasing GR. The number of hotspots decreased for the highest and lowest GR classes (> 50 and < 15 mm, respectively) that were more frequent rainfall events in the fall and spring season when the LAI was lower. This indicates that the vertical and horizontal structures of forest canopy when fully developed in the summer season (when GR ranged between 15-50 mm) create stable water flow paths in the canopies that generate more persistently wet and dry areas in this forest. The present study suggests that spatial variability and time stability of TF were more related to rainfall characteristics than canopy traits. The canopy characteristics appears to play an important role in transferring water flow path to the
ground when the canopy is fully developed in the beech stand, creating input points as hotspots of $TF$ in different $GR$ classes. This research could help the forest managers to optimize the management of these stands in terms of soil water and nutrition availability. The outcome of this research would assist future investigators to better sampling design strategies under different water input affected by climate change or in various seasons in these natural beech stands under similar climatic conditions.

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Spatial and Temporal Variability of Throughfall


تغییرات زمانی و مکانی تاج بارش نوده راش در چالش‌های هیرکانی شمال ایران

میانگین شاخص سطح برگ توده (LAI) و درصد روشنه تاج پوشش در فصل برگدار به

چکیده
تغییرات زمانی و مکانی تاج بارش در بومسازگاری‌های چالش‌های جنگلی مهمی به‌شمار می‌آورند. در این پژوهش تغییراتی در مکانی و زمانی تاج بارش و نیز اثرات تایپوشن و خصوصیات بارش بر تغییرات تاج بارش ارزیابی شدند. نمونه‌برداری‌های بارش و تایپوشن از مهر 1394 تا مهر 1396 از زمان رشد کامل برگ درختان تا قبل از شروع خزن برگ‌ها در نوده طبیعی راش (Fagus orientalis) کاملاً برگ درختان تا گل منظور شده و بنابراین، هرگونه تایپوشن در طبیعی راش (Fagus orientalis) واقع در سری‌های بارش جنگل آموزشی و تایپوشن دانشکده منابع طبیعی دانشگاه تهران (جنگل خرود) انجام شد. میانگین شاخص سطح برگ توده (LAI) و درصد روشنه تایپوشن در فصل برگدار به
ترتیب 6 و 7 درصد به آورد شدند. تراکم درختان در توده 188 درخت در هکتار با مقدار نتیجه 784/8 میلیمتر ثابت گردید. تغییرات الگوهای مکانی تاج بارش در چهار کلاسه باران 5/411 درصد برآورد شدند. در طول دوره مطالعه، 25 رخداد باران با مقدار تجمعی 0-05/015، 015/5-015 و 0-05 میلیمتر در طول دوره مطالعه مشاهده شد. مقدار تاج بارش و تغییرات مکانی تاج بارش (ضریب تغییرات تاج بارش (%TF)) با افزایش عمق باران و شدت بارندگی به ترتیب افزایش و کاهش معنی دار داشت (P < 0/01). 

پژوهش حاضر نشان داد عمق باران و شدت بارندگی را می‌توان به عنوان مهم‌ترین فاکتورهای اثرگذار بر مقدار، تغییرات مکانی و ثبات زمینی تاج بارش دانست. شناخت تغییرات مکانی و ثبات زمینی تاج بارش، به مدیران چنگال در اجرای عملیات پرورشی به منظور مدیریت بهینه توده‌های چنگالی یاری خواهد کرد. نتایج بهتری از برآکشن مکانی تاج بارش رسیده به کف چنگال، از لحاظ دسترسی خاک چنگال به آب و مواد غذایی داشته باشند. 

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