

Author's response – Point to point reply to comments and relevant changes made (highlighted in bold) in the revised manuscript version

Anonymous Referee #1

The paper presents a comparison of mean transit time (MTT) between multiple watersheds in a tropical montane region aiming to evaluate factors controlling stream transit time. The issue addressed in the paper is one of the main concern in understanding hydrological process including sources, flow paths and storages. The paper shows that slope and permeability observed soil-bedrock interface other than land cover are the key factors controlling MTT in the tropical montane area. The paper is well written with significantly new knowledge. I recommend its publication on the Hydrology and Earth System Sciences after properly addressing the following issues:

Reply: Thank you very much for your positive recommendation. Please find our replies to your comments below.

(1) P10989L4, the authors state that short MTT were most strongly related to depth to soil-rock interface particularly for slopes dominated by depth > 2m. The result is derived by a series of divide-points including 0.5m, 1m, 2m. Is there any reason for such classification?

Reply: In the L2-L4 of the discussion paper we stated “a strong positive relation was observed between MTT and depth to soil–bedrock interface (particularly for hillslopes dominated by depths to bedrock > 2 m; Fig. 5f)”, i.e. longer MTTs were related to deeper soil-bedrock interfaces. **We have re-phrased this sentence in the revised manuscript** to make it more clear (**Section 3.2, L400-402**).

The classification was based on the range of soil-bedrock depths observed. Hence, we decided to divide this variable in the following 4 classes: very shallow (<0.5 m), shallow (>50-100 cm), moderate deep (100-200 cm) and relatively deep (>200 cm) soil-bedrock depths, to test their influences on stream MTT. **We have incorporated this information in the revised manuscript (Section 2.5, L279-281).**

(2) In the paper, three functions ($g(\tau)$) have been used to estimate transit time model including Gamma, Exponential and Dispersion. Although model parameters have been listed in Table4, it is perhaps not clear to the readers without specific formulas displayed.

Reply: **We have now made clear in the Results (Section 3.2, L385-386) of the revised manuscript** that the TTD model that we reported for a particular catchment in Table 4 was the one that best fitted the observed baseflow data. **We also have referred to the readers to the supplementary information for the corresponding formulas of the models used by adding a footnote in Table 4.**

(3) Conclusion part, the influencing factors for MTT in variable-area watersheds should be further justified.

Reply: **We have included this information in the revised version (Conclusions, L584-587).**

(4) P10995L16ijNtowards a better understanding “on” the hydrology. . .

Reply: Thank you for highlighting this. **We have changed “on” for “of” in the revised version.**

Anonymous Referee #2

This manuscript aims to assess the control of physical properties on mean stream water transit times in a tropical setting. Although the analyses in the paper are very simple and basic techniques are used, little data from these environments are currently available and this work makes an important contribution in that sense. I would recommend the work for publication, considering the following points below:

Reply: We would like to thank you for your positive recommendation and helpful comments. Please find our replies to your comments below.

The authors comprehensively discuss their results in light of other findings elsewhere. However, there is little discussion on the methods used and the caveats that come with these. There are three major issues that would be good if discussed.

1) Overall, there is a focus on base flows as these were sampled only. However recent work has shown that there can be a strong variability between MTTs of low and high flows. There is no mention of this in the discussion. I think this should be highlighted there and how this relates to potentially holding/breaking down relationships between physiographical properties and MTTs during storms vs base flow.

Reply: In catchments where runoff generation processes are dominated by near-surface or shallow lateral flow pathways, it will be important to carry out stream water samplings during high and low flow conditions to obtain representative estimates of MTTs. However, previous hydrological work at our research site has shown that baseflow is the dominant streamflow component (~ 90%). Furthermore, isotope and chemical-based hydrograph separation for a series of storms and for catchments dominated by different land covers showed that deep subsurface flow (mainly ground water sources) rather than near-surface or shallow lateral flow is the dominant pathway for runoff generation. The high permeability of the volcanic soils (Andisols) and the underlying substrate characterizing this region, as also shown in this paper, promotes vertical soil water percolation and recharge of deeper sources, leading to catchment rainfall-runoff responses dominated by groundwater discharge (Muñoz-Villers and McDonnell, 2012, 2013).

The above supports why our sampling focused on stream baseflow. **We have now made this clear in Methods section (L171-175) in the revised version, citing the above-mentioned references. From this point of the text onwards, we have used consistently the term baseflow MTTs.**

2) One other key point is that the authors found no distinct differences in the isotope signatures of the streams, yet their model results have a large range of implied MTTs. How does this affect the uncertainty in the results? And is this simply a result of the different models which were used (which have different bias)?

Reply: For the calculation of MTTs, input data for the lumped TTD models consist of isotope time series of rainfall, meanwhile the isotope time series of baseflow are used for calibration. Hence, this procedure is different from the statistical method because it uses the temporal variation observed in isotope baseflow data to fit a TTD model and estimate a MTT from it.

In this study, we used the most basic TTD models, which require only one or two distribution parameters to be optimized; note that the used models have been successfully applied in other studies (e.g. McGuire et al., 2005). **We have added this information in the revised manuscript (Section 2.3, L219-221).**

As shown in Table 4, in 7 out of the 12 catchments baseflow MTTs were best estimated using the Exponential model. In two headwater catchments (MAT and SEC), baseflow MTT was best estimated using the Gamma model, which is a variant of the Exponential model. Using the Gamma model, we obtained MTT estimates of 2.6 years for the MAT and 2.7 years for the SEC; using the Exponential model, these estimates would have been 2.9 and 3 years, respectively. Hence, we obtained very similar MTT results using either the Exponential or Gamma model; this is because the optimized values of the α parameter (0.6-0.9) in the Gamma model approximated to 1 (i.e. the gamma distribution with $\alpha = 1$ is equivalent to an exponential distribution (c.f. McGuire et al., 2005). Furthermore, since these nine catchments included those with the lowest and highest MTTs, it is unlikely that the obtained range of MTTs is an artifact of model selection. Finally, MTT estimates for the remaining three catchments as obtained with the Dispersion model were generally close to the overall average MTT.

It should be noted that the TTD model that we reported for a particular catchment was the one that best fitted the observed baseflow data. **We have emphasized this in the revised manuscript (Section 3.2, L385-386).**

Furthermore, the selected models were identified as reliable TTDs because they provided predictions with relatively high fitting efficiencies and small uncertainty ranges. Parameter uncertainty was evaluated through the Nash-Sutcliffe E value (see Section 2.3), and Table 4 reports the values of the model parameters and the uncertainty bounds. In all cases, the obtained E values are indicative of a good predictive skill (> 0.5 to 0.7). Moreover, the overall efficiency of the TTD models was evaluated using the RMSE. In this case, the RMSE of the models was between 0.8 and 1.5 ‰, which indicates a good analytical reproducibility of the observed baseflow isotope data, considering that the isotopic composition range was from - 50.2 to -41.0‰ across all catchments.

3) Many of the sampled subcatchments are nested and only stream samples for isotope data were taken. There is no indication of the relative contributions of runoff from the different parts of the catchment. For example, if the headwaters are contributing more than other subcatchments (could very well be considering the differences in precipitation), then this could bias the results in favour of the processes in those areas.

Reply: Due to accessibility problems and financial constraints, most of the headwaters and subcatchments in this area, as in other tropical regions, remain ungauged. Hence, we followed this approach because it relies primarily on tracer data, and hence is

useful to broadly characterize the hydrological behavior of the studied catchments (McGuire and McDonnell 2006).

Without measuring streamflow in the different catchments, it is difficult to say which ones could potentially contribute more than others. However, most precipitation falls where the first and second-order catchments are located (2,100-2,400 m asl), and it appears that this is where most groundwater recharge occurs, as inferred from previous water balance studies (Muñoz-Villers et al., 2012).

To properly answer this question, more research work and effort should be done to quantify the stream flow contributions from different catchments, and to investigate the connectivity that could exist among sites.

The terrain analysis comprises of many standard techniques, hence I think section 2.4 can be condensed significantly. I would recommend to use either 'watershed' or 'catchment' consistently throughout the manuscript.

Reply: Despite some of the methods used to describe the terrain analysis are standard, we consider it important to explain to the readers how we defined the terrain parameters in terms of datasets and resolutions used, GIS software and processing. Although this may have made section 2.4 a bit longer, we think its length is still sufficiently short (19 lines).

Following the reviewer's recommendation, **we have used the term 'catchment' consistently throughout the revised manuscript.**

Some other minor comments:

10976, L7: change to 'and related these to catchment'

Reply: We have made this change in the revised version (Abstract, L23).

10976, L20-22L This sentence not clear. It suggests there is an effect of scale, while the results have shown there isn't. Greatest difference between scales, or within the two groups?

Reply: We have re-phrased this sentence to make its meaning clear in the revised manuscript (Abstract, L36-37).

10977, L17: needs rephrased. Change 'first' to 'for the first time'?

Reply: We have made this change (Introduction, L63).

10978, L8-9: are these of the same order of magnitude as Munoz-Villers and McDonnell 2012?

Reply: They were larger catchments from ~2 to 77 km². We have added this information in the revised version (Introduction, L81).

10979, L11: change to 'the majority of monitored headwaters are located'

Reply: We have made this change (Section 2.1, L113).

10980, L3: ET not known as this elevation?

Reply: Indeed, no data on ET are available for this elevation.

10981, L13: What kind of correlation? Can you show what the measure of fit is?

Reply: It was a linear regression with an r^2 of 0.95. **We have added this information in the revised version (Section 2.2.2, L170).**

10982, L4: Explain why this is the best method?! e.g. why not nearest gauge, something along the lines of elevation corrected Thiessen polygons

Reply: We actually followed the McGuire et al. (2005) approach, matching the average isotope signature of baseflow with that of rainfall, and so to identify the elevation at which most recharge occurs. **This information has been added in the revised version (Section 2.3, L190-192).**

Furthermore, this approach was supported by the fact that both the TG and SECP sites are located at those elevations in the LG watershed where probably most groundwater recharge occurs, as determined by previous water balance studies (Muñoz-Villers et al., 2012). This information was already mentioned in the text (L198-200).

10982, L12: Why this approach when many others around? Does it fit your data best?

Reply: MTT studies often suffer from short isotope time series of rainfall, which yield nonunique model calibrations or uncertainty in the parameter estimates of the transit time distribution. To deal with this, the majority of the investigations extend their input time series using different methods (i.e. temperature records, seasonality in rainfall, or data from nearby long-term stations). At our site, precipitation shows a marked seasonal pattern (Holwerda et al., 2010; Muñoz-Villers et al., 2012). Rainfall isotope signatures also show a strong and consistent variation with rainfall amount (Goldsmith et al., 2012). **We have added this information in the revised version (Section 2.3, L201-205).** Thus, we decided to generate a warm-up period required for the model simulations, repeating our measured 2-year rainfall time series 15 times following the approach of Hrachowitz et al. (2009, 2010); note that we already tested this method in previous MTT work at our site (cf. Muñoz-Villers and McDonnell, 2012).

10984, L6: not 'created' but 'derived'

Reply: We have made this change (Section 2.5, L255).

10984, L14: which criteria were used for selection?

Reply: This information is provided in Section 2.5, L257-259.

10984, L16: change 'description' to 'descriptions'

Reply: We have made this change (Section 2.5, L264).

10984, L17: n = 3 in total or 3 at each site?

Reply: Three undisturbed soil core samples at each site. We have clarified this in the revised version (Section 2.5, L266).

10985, L15-16: How does this relate to long term (or even 2006-2010) data? Could there be bias in your data as a result?

Reply: The average of 3,476 mm measured at the SECP site (2,100 m) is somewhat higher (9 %) than the average of $3,183 \pm 306$ (SD) mm measured at that same site between November 2005 and November 2009 (Goldsmith et al., 2002). No other data than those given here are available for the other sites. Nevertheless, the rather small

difference with the 2005-2009 data suggests that the two years of data used in this analysis are representative of the longer term pattern.

We have added this information in the revised manuscript (Section 3.1, L298-303).

10985, L24: p value results of which test?

Reply: Anova statistical test. **We have complemented and re-phrased the information regarding the statistical methods in the revised version (Section 2.5, L287-293).**

10986, L5: lagged by how much?

Reply: **We have omitted this text in the revised manuscript.**

10986, L8: not statistically significant at all sites? And again, which test results?

Reply: Yes, not statistically significant at all sites. **We have made this clear in the revised version (Section 3.1, L318).**

For the second question, please see our reply to the previous comment.

10986, L10: But many of your sites are nested. Have you considered that this could be a result e.g. of the headwaters contributing the majority of water? Do you have any indication of how much water is coming from which parts of the catchments spatially?

Reply: Please see our reply to your similar comment above.

10987, L10: the highest proportion of what?

Reply: Of the above mentioned slope categories. **We have made this clear in the revised version (Section 3.3, L351).**

10987, L14-15: This sentence should be moved to methods - and explain why this is important?

Reply: **We have moved this sentence to the Methods section (Section 2.4, L245).** The explanation given was to inform the readers what the measures/numbers indicate.

10991, top sentence needs rephrased

Reply: We were not sure which sentence the reviewer is referring to.

10992, L17: change 'has' to 'have'

Reply: **We have made this change.**

Figure 1: This Figure is currently all a bit busy and it is difficult to make out the subcatchments. Why not focus on the study area only as the lower part of the catchment is not relevant for the study here. Also, what do the elevation labels refer to in A?

Reply: We included the lower part of the catchment to show the entire extent of the LG watershed. The elevation labels in Figure A are presented to give information about the elevation from the bottom to the top of the catchment. In response to the comment, **we have increased the size of the figures and labels, and we have also re-arranged its content to make it easier to read.**

Figure2-3: Swap order of Fig 2 and 3 - in the text you first refer to Fig 3, then 2.

Reply: We have made this change.

Short comment by G. Mosquera

Dear authors, I think it would be quite interesting to discuss the selection of the best TTD for each catchment. Table 4 shows what the reader should assume as the best TTDs for each of them, but a thorough discussion about why different TTDs were selected for different catchments is lacking. Perhaps it would also be interesting to show the parameter space of all or at least the best TTD/s at least for one catchment in order to show why one or another TTD was selected as the best (e.g., Figure 6 in Hrachowitz et al., 2009). Additionally, perhaps the analysis of the probability and cumulative density functions would help to better understand the hydrology of the system as a whole (e.g., why one of the smallest catchments shows the longest MTTs) (See for example Figures 8, 9, 14, and 15 in Timbe et al., 2014).

Reply: We thank you for your comment. As we mentioned in our reply to Reviewer #2, the TTD model that we used for a particular catchment was the one that best fitted the observed baseflow data. **We have made this clear in the revised manuscript.**

Note that the criteria used to select the TTDs are described in Section 2.3. Also note that in the majority of cases we used the Exponential model or the variant (Gamma model) (please see our Reply to reviewer #2). With regard to include a thorough discussion about the selected TTDs, we think that this is out of the scope of this paper. As our objective and research questions stated, we quantified stream MTTs and related these to catchment characteristics to understand and identify the key factors controlling baseflow MTTs across the different catchments. Please note that (probable) answers to the question 'why one of the smallest catchments shows the longest MTTs' are provided in Results (Section 3.2 and 3.3), Discussion (Section 4.2) and Conclusions.

Comments by the Editor

1. Please consider adding the base flow in the title as the focus is the base flow MTT, see the comment and your reply to Referee #2.

Reply: We appreciate the recommendation; however, we feel it is justified that we keep the title as it is, since in the manuscript we informed the readers that our streams are almost completely dominated by baseflow and for that reason samples for isotope analysis were taken during baseflow conditions.

2. The conclusion is too much similar to abstract. Please add more details to the conclusion part, like how the identified physiological properties control the MTT

Reply: We have followed the suggestion. **We have incorporated more details of our results in the Conclusions. We also re-phrased and re-arranged this section in the revised version (L571-589).**

References

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1 **Manuscript title**

2 Factors influencing stream water transit times in tropical montane watersheds

3

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17 40-89.

18 **Abstract**

19 Stream water mean transit time (MTT) is a fundamental hydrologic parameter that
20 integrates the distribution of sources, flow paths and storages present in catchments. However, in
21 the tropics little MTT work has been carried out, despite its usefulness for providing important
22 information on watershed functioning at different spatial scales in (largely) ungauged basins. In
23 particular, very few studies have quantified stream MTTs and related [these](#) to catchment
24 characteristics in tropical montane regions. Here we examined topographic, land use/cover and
25 soil hydraulic controls on baseflow transit times for nested [catchments](#) (0.1–34 km²) within a
26 humid mountainous region, underlain by volcanic soil (Andisols) in central Veracruz (eastern
27 Mexico). We used a 2 year record of bi-weekly isotopic composition of precipitation and stream
28 baseflow to estimate MTT. Land use/cover and topographic parameters (catchment area and
29 form, drainage density, slope gradient and length) were derived from GIS analysis. Soil water
30 retention characteristics, and depth and permeability of the soil-bedrock interface were obtained
31 from intensive field measurements and laboratory analysis. Results showed that baseflow MTT
32 ranged between 1.2 and 2.7 years across the 12 study catchments. Overall, MTTs across scales
33 were mainly controlled by catchment slope and the permeability observed at the soil-bedrock
34 interface. In association with topography, catchment form and the depth to the soil-bedrock
35 interface were also identified as important features influencing baseflow MTTs. The greatest
36 differences in MTTs were found [both within groups of](#) small (0.1–1.5 km²) and large (14–34
37 km²) [catchments](#). Interestingly, longest stream MTTs were found in the headwater cloud forest
38 catchments.

39
40 *Key words:* stable isotopes, baseflow, catchment transit times, topography, land cover, Andisols,
41 Mexico

43 **1. Introduction**

44 The demand for fresh water is rapidly increasing in the humid tropics due to population
45 growth. Nevertheless, in these regions —particularly the montane tropics— relative little
46 process-based hydrological studies have been performed to quantify the states, stocks, flows and
47 residence times of water. These areas are especially prone to land degradation and deforestation
48 for conversion to agricultural and pasture lands (Asner et al., 2009). Notably, tropical montane

49 cloud forests (TMCF) are unique and hydrologically important ecosystems (Bruijnzeel, 2004),
50 but are among the world's most threatened terrestrial ecosystems (Cayuela et al., 2006; Hamilton
51 et al., 1995; Pope et al., 2015). Yet the hydrological impacts associated with [land degradation](#)
52 [and forest conversion](#) at different scales remain poorly understood, thus hampering the
53 development of effective local and regional strategies for water resources protection and
54 management.

55 Stream water mean transit time (MTT) is an important hydrologic metric that integrates
56 the variety of flow paths, storages and runoff sources [potentially existing in catchments](#). In
57 humid temperate environments, MTTs estimated from stable isotopes have been used to broadly
58 characterize the hydrological and biogeochemical behavior of catchments (McDonnell et al.,
59 2010), providing important information on [catchment](#) resistance and resilience to climate change
60 scenarios (Carey et al., 2010). In these same environments, significant progress has been made in
61 exploring the linkages between baseflow MTTs and catchment characteristics, and the dominant
62 factors controlling stream MTT variability across scales and regions. For example, McGuire et
63 al. (2005) showed [for the first time](#), the dependence of stream water mean residence time on
64 catchment topographic indices (hillslope length and gradient) for multiple nested watersheds in
65 Western Oregon, USA. Further, Broxton et al. (2009) found that stream water isotope variability
66 and estimated MTTs were both related to watershed aspect and slope in the Valles Caldera
67 watershed, New Mexico, USA. In Central Japan, Asano and Uchida (2012) showed that base
68 flow MTT was mainly controlled by the depth of the hydrologically active layer (i.e. depth of the
69 soil-bedrock interface), which was not necessarily related to catchment topography. Perhaps the
70 most extensive work to date has [been done](#) in North East Scotland, where several studies have
71 identified soil properties (soil type and permeability) as the main control on stream MTTs (Geris
72 et al., 2015; Rodgers et al., 2005; Soulsby et al., 2006; Tetzlaff et al., 2009a). With the exception
73 of the investigations carried out by McGlynn et al. (2003) in the Maimai watersheds in New
74 Zealand, and more recently by Hale and McDonnell et al. (2016) in the Alsea Watershed Study
75 in the Oregon Coast Range, USA, which both showed strong positive relations between MTTs
76 and catchment area, most studies to date have shown that landscape evolution and organization
77 dictates rainfall-runoff processes in humid temperate environments.

78 In the humid tropics, isotope-inferred stream MTT studies have provided insights into the
79 hydrological functioning of small forested catchments ($\leq 0.3 \text{ km}^2$; Muñoz-Villers and

80 | McDonnell, 2012), and their sensitivity to land use conversion ([< 1.8 km²](#); Roa-García and
81 | Weiler, 2010). At larger scales ([> 2 to 77 km²](#)), the studies carried out in Ecuador by Crespo et al.
82 | (2012) and Timbe et al. (2014) have reported MTT values for various flowing water bodies
83 | (springs, creeks, tributaries and rivers), but as yet, the factors controlling the stream water transit
84 | times in this and other montane regions of the humid tropics remain to be explored.

85 | Here we build upon previous isotope work at our site in central Veracruz, Mexico, where
86 | large water storage capacities have been estimated (~ 3 yr) for an old-growth TMCF upland
87 | catchment based on baseflow MTT (Muñoz-Villers and McDonnell, 2012). The present study is
88 | the first in the humid tropics that we are aware of that explores the relationship between stream
89 | water MTT and landscape characteristics across catchments ranging [different size areas \(from](#)
90 | [0.1 to 34 km²](#)). Our tropical montane watersheds are underlain by volcanic soil substrates
91 | (Andisols). MTT was determined using a 2 year record of rainfall and stream water isotope data.
92 | We used metrics such as land cover, topographic parameters and hydrologic properties of the
93 | soil-bedrock profile to identify the factors controlling stream MTTs in this environment.
94 | Specifically we addressed the following research questions:

- 95 |
- 96 | 1. What are the [stream](#) mean transit times across catchment scales?
 - 97 | 2. How do catchment area, topography and subsurface hydrologic properties relate to
98 | stream [transit times](#)?
 - 99 | 3. Does land cover have an effect on stream MTT [variability](#)?
 - 100 | 4. Is there a dominant factor controlling stream water transit times in this mesoscale
101 | [catchment](#)?

102 |

103 | **2. Materials and Methods**

104 | **2.1 Study site**

105 | The 5th-order Los Gavilanes (LG) river [catchment](#) (41 km²; 19° 28' N - 97° 01' W) is
106 | located on the eastern (windward) slopes of the Cofre de Perote mountain. It is the main stream
107 | water supply for the city of Coatepec and surroundings (~80,000 inhabitants). The landscape of
108 | this region is complex and strongly dissected by perennial streams draining catchments of
109 | different sizes. For this study, 12 catchments were selected, ranging in area from 0.1 to 34 km²

110 and located between 1300 and 3000 m a.s.l. (Figure 1a). Table 1 summarizes the physical
111 characteristics of the study catchments.

112 The mid and upper parts of the LG [catchment](#) (1800–3000 m a.s.l.), where the majority
113 of [monitored](#) headwaters are located, are characterized by short steep hillslopes covered mostly
114 by pine-oak forest, mature and secondary tropical montane cloud forest, and pasture (Figure
115 1b,c). The lower portions of the LG [catchment](#) (1300–1800 m a.s.l.) are characterized by more
116 gentle terrain covered by pasture, fragments of cloud forests on the steeper slopes and shaded
117 coffee plantations below 1400 m.

118 The general climate in the LG [catchment](#) is temperate humid with abundant summer rains
119 (Garcia, 1988). [About](#) 80 % of [the annual rainfall](#) falls as convective storms during the wet
120 season (May–October), [when](#) the region is under the influence of the easterly trade wind flow.
121 Maximum ground water recharge and catchment runoff [also](#) occur during [this](#) season (cf.
122 Muñoz-Villers and McDonnell, 2013). The relatively dry season (November–April) is
123 characterized by light rains and/or fog and drizzle associated with the passage of cold fronts
124 (Holwerda et al., 2010). Fog interception occurs exclusively during this time of year, and
125 accounts for ≤ 2 % of the annual rainfall for the upper part of the LG [catchment](#) (Holwerda et al.,
126 2010; Muñoz-Villers et al., [2012](#), 2015).

127 The local climate varies markedly with elevation. At 1210 m a.s.l. (lower part of the LG
128 [catchment](#)), the annual mean daily temperature is 19 °C. Corresponding mean annual rainfall and
129 reference evapotranspiration (ET_0) are 1385 and 1120 mm, respectively (Holwerda et al., 2013).
130 At 2100 m a.s.l. (middle part), the annual mean daily temperature is 15 °C, [and](#) mean annual
131 rainfall and ET_0 are [3185](#) and 855 mm, respectively ([Goldsmith et al., 2012](#); Muñoz-Villers et
132 al., 2012). Finally, at 3000 m a.s.l. (upper part), mean annual temperatures range between 5 and
133 10 °C, and mean annual rainfall is 1900 mm (SMN, 2014).

134 Andisols derived from volcanic ashes are the dominant type of soil across the LG
135 [catchment](#). These soils are characterized by low bulk density, high permeability, high water
136 retention capacity and high organic matter content (Gomez-Tagle et al., 2011). Soil profiles are
137 usually deep, well developed and multilayered (A, A/B, Bw, Bw/C), with silt loam and silty clay
138 loam as the dominant textures ([Gómez-Tagle et al., 2011](#)). The parental material is permeable,
139 consisting of moderately weathered andesitic breccias, underlain, in turn, by semi-permeable

140 saprolite that has been weathered from fractured andesitic-basaltic bedrock (cf. Muñoz-Villers
141 and McDonnell, 2012).

142

143 Figure 1

144 Table 1

145

146 **2.2 Field data collection and analysis**

147 **2.2.1 Rainfall measurements**

148 To quantify daily precipitation and its spatial variation along the altitudinal gradient,
149 rainfall was measured at 1560, 2100 and 2400 m a.s.l. (Figure 1a). For the sites at 1560 m
150 (hereafter RA) and 2100 m (SECP), stand-alone tipping bucket rain gauges of the type RG2M
151 (Onset) and Casella CEL, respectively, were used (both with a resolution of 0.2 mm). For the site
152 at 2400 m, rainfall was measured with an ARG100 tipping bucket rain gauge (Environmental
153 Measurements; 0.2 mm) as part of a meteorological station (TG; Figure 1a). The signals from the
154 stand-alone gauges were stored in an HOBO pendant event logger (Onset), whereas for the
155 gauge in the weather station a CR1000 data logger (Campbell Scientific) [was used](#).
156 Measurements at SECP were made continuously from July 2006 to November 2010, whereas
157 measurements at RA and TG covered the period of isotope sampling (see below).

158

159 **2.2.2 Collection and analysis of rain and stream water samples**

160 To establish the records of $\delta^2\text{H}$ and $\delta^{18}\text{O}$ isotope composition of precipitation and
161 streamflow, samplings during non-storm conditions were carried out over two hydrological years
162 (May 2008 – April 2010). Paired with the tipping bucket rain gauges, samples of bulk rainfall
163 were collected using a sampler consisting of a 95 mm diameter funnel assembled to a 40 mm
164 diameter and 400 mm long transparent collection tube. The tube contained a float to minimize
165 evaporation. In addition, the rain water collector was inserted into a 75 mm diameter PVC pipe
166 wrapped with bubble foil insulation to protect the collected water against sunlight and minimize
167 temperature variations. Rainwater sampling intervals ranged between 1 and 25 days, depending
168 on rainfall amount and frequency. For logistical reasons, rainwater collection at the RA site was
169 only possible from March 2009 to April 2010. The missing isotope data (10 months) were
170 completed using a [linear regression](#) with data from the [SECP](#) site ($r^2 = 0.95$).

171 [Previous hydrological work at our research site has shown that baseflow is the dominant](#)
172 [streamflow component \(~ 90%\). Furthermore, isotope and chemical-based hydrograph](#)
173 [separation for a series of storms and for catchments dominated by different land covers showed](#)
174 [that rainfall-runoff responses are mainly dominated by ground water sources \(Muñoz-Villers and](#)
175 [McDonnell, 2012, 2013\). For these reasons, we focused our stream sampling on baseflow.](#) Grab

176 samples of base flow were collected every 2 weeks at the outlets of the 12 study catchments.
177 These included nine sampling points representing headwaters up to 4th-order streams (numbers
178 1-9; Figure 1a), two main tributaries of the LG river (10 and 11) and the main river (12).

179 All water samples were collected in 30 ml borosilicate glass vials with polycone sealing
180 cap to prevent evaporation. The samples were analyzed for $\delta^2\text{H}$ and $\delta^{18}\text{O}$ on a laser liquid-water
181 isotope spectrometer (Version 2, Los Gatos Research, Inc.) in the Hillslope and Watershed
182 Hydrology Lab at Oregon State University, USA. The isotope values are expressed in permil
183 (‰) relative to Vienna Standard Mean Ocean Water (VSMOW). The precision of $\delta^2\text{H}$ and $\delta^{18}\text{O}$
184 measurements was 0.3 and 0.1‰, respectively.

186 **2.3 Transit time model**

187 Biweekly $\delta^2\text{H}$ signatures of stream water and rainfall were used to estimate base flow
188 mean transit time (MTT) and transit time distribution (TTD) for each of the study catchments.
189 First, 2-week volume-weighted means (VWMs) of rainfall isotope composition were calculated
190 for each of the three sampling sites. [Second, we followed the McGuire et al. \(2005\) approach to](#)
191 [compare the average isotope signature of baseflow for each study catchment with that of rainfall,](#)
192 [and so to identify the elevation at which most recharge occurs. For this,](#) we calculated 2-yr
193 averages, and determined for each catchment which of the rainfall time series had its average
194 $\delta^2\text{H}$ value closest to the average $\delta^2\text{H}$ baseflow value. The overall mean $\delta^2\text{H}$ base flow value was
195 -44.9 ‰ (range: -50.2 to -41.0 ‰ across all catchments), whereas rainfall at TG, SECP and RA
196 had volume-weighted mean $\delta^2\text{H}$ values of -43.0, -37.6 and -33.5 ‰, respectively. Hence, for all
197 of the study catchments, MTT simulations were carried out using the rain isotope data from
198 either TG or SECP. [Further, this approach was supported by the fact that both the TG and SECP](#)
199 [sites are located at those elevations in the LG catchment where most groundwater recharge](#)
200 occurs, as determined by previous water balance studies (Muñoz-Villers et al., 2012).

201 [The \$\delta^2\text{H}\$ precipitation data collected over 2 hydrological years \(May 2008 – April 2010\)](#)
202 [may be too short to properly constrain stream base flow MTT estimates \(cf. Hrachowitz et al.,](#)
203 [2009\). However, precipitation at our site shows a marked seasonal pattern \(Holwerda et al.,](#)
204 [2010; Muñoz-Villers et al., 2012\). In addition, rainfall isotope signatures shows a strong and](#)
205 [consistent variation with rainfall amount \(Goldsmith et al., 2012\).](#) Therefore, to generate an
206 artificial warm-up period required for the MTT model simulations, we followed the approach of
207 Hrachowitz et al. (2009) and repeated our measured 2-year rainfall time series 15 times (cf.
208 Muñoz-Villers and McDonnell, 2012). We then used a lumped parameter convolution model to
209 predict the $\delta^2\text{H}$ output for the stream water as a weighted sum of its respective past $\delta^2\text{H}$
210 measured input in precipitation (Maloszewski and Zuber, 1993). Mathematically, the stream
211 water outflow composition at any time, $\delta_{\text{out}}(t)$, consisted of past inputs lagged $\delta_{\text{in}}(t-\tau)$ and
212 weighted by the transfer function $g(\tau)$, representing its lumped TTD (Maloszewski and Zuber,
213 1982):

$$214 \delta_{\text{out}}(t) = \int_0^{\infty} g(\tau) \delta_{\text{in}}(t - \tau) d\tau \quad (1)$$

216 where τ are the lagged times between the input and output tracer composition. The weighting
217 function or TTD describes the travel time of the water from the ground surface to an outflow
218 location (i.e. the catchment outlet) (McGuire and McDonnell, 2010).

219 [In this study, we used the most basic TTD models \(Exponential, Gamma and Dispersion\),](#)
220 [which require only one or two distribution parameters to be optimized and have been](#)
221 [successfully applied in other studies \(e.g. McGuire et al., 2005\).](#) The performance of different
222 TTD functions for each of the 12 study catchments [was evaluated](#) using the transfer function
223 hydrograph separation model TRANSEP (McGuire and McDonnell, 2010; Weiler et al., 2003).
224 This model utilizes the Generalized Likelihood Uncertainty Estimation (GLUE) methodology
225 (Freer et al., 1996) based on Monte Carlo simulations to determine the identifiability of the
226 individual parameters. Our Monte Carlo analysis of each TTD consisted of 10,000 runs. Model
227 performance was assessed using the Nash–Sutcliffe efficiency E (Nash and Sutcliffe, 1970),
228 based on the best agreement parameter value, where a value of 1 would indicate a perfect fit.
229 Parameter uncertainty was defined as the range between 10th and 90th percentile value for the
230 best 20% performing parameter sets based on E (McGuire and McDonnell, 2010; Seibert and

231 McDonnell, 2010). The overall performance of the TTD models was evaluated using the root
232 mean square error (RMSE).

233

234 **2.4 Terrain analysis**

235 To evaluate whether landscape characteristics had an influence on base flow MTT,
236 several metrics describing catchment topographic and morphometric features were calculated in
237 ILWIS 3.3, a raster and vector GIS system. Catchment area was obtained by delineating and
238 extracting each catchment boundary using a digital contour elevation map (10 x 10 m
239 resolution). Land cover was obtained from a regional land cover/use raster map (20 x 20 m)
240 elaborated by Muñoz-Villers and López-Blanco (2008), using satellite images and ground truth
241 verification data. For vegetation cover, each catchment was classified in one of the [following](#) 4
242 categories: (1) > 90 % covered by TMCF; (2) > 60 % covered by any type of forest (pine-oak,
243 [TMCF](#)); (3) > 90 % covered by pasture; and (4) even mixture ([~](#)50-50 %) of pasture and any
244 type of forest.

245 Catchment form factor ([a measure of catchment shape](#)), drainage density, slopes and
246 hillslope length were calculated using topographic maps (scale 1: 20,000) and a 10 x 10 m digital
247 elevation model (DEM). Catchment form factor (R_f) and drainage density (D_d) were calculated
248 [following](#) Horton (1932). Hillslope length was obtained as the average distance between
249 catchment ridge top and valley bottom. Horizontal and vertical gradients of each pixel in the
250 DEM were used to calculate the mean and the percentage distribution of slopes in each
251 catchment, using for the latter the following six classes: [0-5](#); 5-10; 10-20; 20-30; 30-45; > 45°.

252

253 **2.5 Soil sampling and analysis**

254 Field surveys, soil samplings and subsequent laboratory analysis were conducted from
255 May 2011 to May 2012. First, hillslope forms (ridge top, mid and valley bottom) were [derived](#) in
256 GIS using topographic analysis algorithms (Jenness, 2006) and then overlaid with catchment
257 boundaries. From the intersection of the polygon units thirty-two soil toposequences were
258 selected, distributed mostly in the mid and lower portions of the LG [catchment](#) because access to
259 the upper part was very difficult. At each toposequence, soil auger-holes up to 2.2 m deep were
260 performed from ridge top to valley bottom to determine the organization of soil layers along the

261 hillslope. Soil penetration resistance was also measured down to 2 m using a dynamic cone
262 penetrometer, following the design and method of Herrick and Jones (2002).

263 | At selected toposequences, soil profile pits of approximately 1.5 [m](#) x 1.5 [m](#) x 2 m (length,
264 | width and depth, respectively; 43 in total) were excavated for detailed soil descriptions following
265 | the method of Schoeneberger et al. (2002). In addition, undisturbed soil core samples ($n = 3$) at
266 | the soil-bedrock interface [in each soil profile pit](#) were taken to determine saturated hydraulic
267 | conductivity (K_s) in [the](#) laboratory using the constant-head method. Further, a pedotransfer
268 | function, correlating the observed K_s and penetration resistance values, was used to extrapolate
269 | K_s of the least permeable layer to the catchment scale.

270 | In each soil pit, soil samples from the A and B-horizons (solum) were collected for
271 | laboratory analysis. Bulk density was determined from samples taken with cylindrical stainless
272 | steel cores of 100 cm³ in each horizon ($n = 3$), and dried at 105 °C until constant weight. For soil
273 | moisture content at field capacity, undisturbed samples from 5 x 1 cm rings (diameter and height;
274 | $n = 3$) were collected, then weighed after reaching saturation and equilibration (normally within
275 | 48 hours), and placed in a pressure-plate apparatus at 30 kPa. From water retention at field
276 | capacity and bulk density values, the amount of water held in the solum (in mm) was calculated.
277 | All laboratory analyses were performed in the Soil Laboratory of the Instituto de Ecología A.C.,
278 | Xalapa, Veracruz.

279 | [Based on the observed range of depths to soil-bedrock interface \(DSBI\), this variable was](#)
280 | [divided in 4 classes: very shallow \(< 50 cm depth\), shallow \(> 50-100 cm\), moderate deep \(100-](#)
281 | [200 cm\) and relatively deep \(> 200 cm\).](#) Soil K_s and water retention (WR) capacity categories
282 | were defined for each site and hillslope sequence. The K_s classes were obtained from the Soil
283 | Hydrology Group of the National Engineering Handbook, Part 630 (NRCS-USDA, 2007), and
284 | partly modified based on the HOST classification system (Boorman et al., 1995). The WR
285 | capacity classes were defined on *ad hoc* ranges. Based on relationships between the soil
286 | hydrologic properties and geoforms, the data was extrapolated to the entire LG [catchment](#).

287 | [Finally, to evaluate differences in the isotopic composition and deuterium excess \(\$d\$ -](#)
288 | [excess = \$\delta^2\text{H} - 8 \cdot \delta^{18}\text{O}\$; Dansgaard, 1964\) values for rainfall and stream water across sites, we](#)
289 | [used ANOVA t-tests.](#) Statistical relationships between [baseflow](#) MTT, [depth and permeability at](#)
290 | [the soil-bedrock interface](#), soil [water retention](#) and landscape characteristics (land cover and
291 | topographic variables) were [tested](#) using Spearman's rank order correlations. [All statistics were](#)

292 | [evaluated at the 0.05 confidence level and conducted in Sigma plot \(version 12, Systat Software](#)
293 | [Inc.\).](#)

294

295 | **3. Results**

296 | **3.1 Isotopic composition of rainfall and stream water**

297 | From May 2008 to April 2010, mean annual precipitation varied from 2670 mm at RA
298 | (1560 m), 3476 mm at SECP (2100 m) to 3264 mm at TG (2400 m). [The average of 3,476 mm](#)
299 | [measured at the SECP is somewhat higher \(9 %\) than the average of \$3,185 \pm 305\$ \(SD\) mm](#)
300 | [measured at that same site between November 2005 and November 2009 \(see above\).](#)
301 | [Nevertheless, the rather small difference suggests that the two years of data used in this study are](#)
302 | [representative of the longer rainfall pattern. For the other sites, no other data than those given](#)
303 | [here are available.](#)

304 | Rainfall showed a clear seasonal pattern, with 80% on average falling during the wet
305 | season (May–October). During the same period, a wide range of variation in the biweekly
306 | rainfall isotope values was found across elevations. The largest variation (118 ‰ for $\delta^2\text{H}$ and 17
307 | ‰ for $\delta^{18}\text{O}$) and most negative (depleted) values (-110 ‰ for $\delta^2\text{H}$ and -16 ‰ for $\delta^{18}\text{O}$) were
308 | observed at the highest altitude (2400 m). With decreasing altitude, rainfall isotope values
309 | became more positive (enriched) and their range of variation smaller (Figure 2a). However,
310 | differences in rainfall isotopic composition among elevations were only suggested for $\delta^{18}\text{O}$ ($p =$
311 | 0.031). Mean annual deuterium excess (d -excess) values of rainfall increased from 15 to 17 ‰
312 | with elevation, but differences among sites were not significant ($p \geq 0.05$).

313 | Across all sites, $\delta^2\text{H}$ and $\delta^{18}\text{O}$ rainfall signatures in the wet season (-32.3 ± 25.7 ‰ and -
314 | 5.9 ± 3.2 ‰ SD, respectively) were significantly depleted as compared to the dry season ($-15.6 \pm$
315 | 13.6 ‰ and -4.3 ± 1.7 ‰; $p \leq 0.05$) (Figure 3). Stream water isotopic composition followed the
316 | seasonal pattern as observed for precipitation; however values were damped (range: -49.4 to -
317 | 41.8 ‰ for $\delta^2\text{H}$ and -8.1 to -7.2 ‰ for $\delta^{18}\text{O}$; Figure 3) as compared to rainfall (range: -106 to 10
318 | ‰ for $\delta^2\text{H}$ and -15 to -1 ‰ for $\delta^{18}\text{O}$, respectively). For [all](#) stream water [sites](#), differences in
319 | isotope composition between the dry and wet seasons were not statistically significant ($p \geq 0.05$).

320 | Unlike rainfall, there were no distinct differences in the isotope signatures of the streams,
321 | despite the altitude difference of more than 1000 m between the upland headwaters and the
322 | downstream LG river tributaries (Figure 2b). During the wet season, d -excess values in the

323 streams were rather constant across all sites, whereas during the dry season values were slightly
324 enriched at lower elevations as compared to upper elevations (Table 2).

325 | Figure 2c shows that all samples of precipitation and stream water fall along the local
326 meteoric water line (LMWL), with *d*-excess values consistently above 10 ‰ (range of rainfall:
327 10.7 to 24.2 ‰, and streams: 10.4 to 22.5 ‰). The fact that the rainfall and stream baseflow
328 samples had very similar *d*-excess ranges indicates that the catchment water input-output was
329 little affected by evaporation.

330

331 Figure 2

332 Figure 3

333 Table 2

334

335 **3.3 Land cover, topography and soil hydraulic properties**

336 | Our GIS analysis showed that approximately 70 % of the LG [catchment](#) was covered by
337 some type of forest. Eight out of the 12 study catchments were located in the middle and upper
338 portions of the LG [catchment](#), which are characterized by moderate to steep terrain. These eight
339 catchments ranged in size between 0.1 and 14 km², and were covered predominantly by mature
340 and secondary TMCF (≥ 50 %; Table 3). Forest was also the dominant land cover in one of the
341 two tributary catchments (Huehueyapan) and in the LG [catchment](#) itself. In [these two](#)
342 [catchments](#), the upper part was dominated by pine-oak (≥ 2500 m), the middle part by TMCF and
343 pasture, meanwhile coffee plantations and forest fragments characterized the lower part (≤ 1400
344 m). Two out of the 12 catchments were dominated by pasture (having areas of 0.1 and 1.5 km²),
345 and only one catchment (1.9 km² of area) was covered by even portions of forest and pasture.

346 Hillslope lengths were shortest (113 m on average) in the smallest catchments (0.1–1.5
347 km²), and longest (273 m) in the largest ones (14–34 km²; Table 3). Slopes of intermediate
348 length (217 m) were found in the 4–9 km² catchments. Mean slope was $32 \pm 5^\circ$ across all
349 catchments. The dominant categories of slopes were 10-20° and 20-30°. Within these groups, the
350 headwater mature and secondary cloud forest catchments (MAT and SEC; ≤ 0.25 km²) showed
351 the highest proportions [of the above mentioned slope categories](#). The pasture headwater
352 catchment (PAS; 0.1 km²) had the highest percentage (46 %) of gentle slopes (0-10°), meanwhile

353 the 20 km² Huehueyapan tributary catchment showed the highest proportion (33 %) of very steep
354 slopes ($\geq 30^\circ$).

355 The [catchment](#) form factor (R_f) ranged between 0.071 (CATM1 and CATM5) and 0.231
356 (SEC). Drainage density (D_d) ranged from 1.3 to 8.0 km km⁻². Low D_d values (2.4 ± 0.4 km km⁻²)
357 were found at the larger catchments (14-34 km²) whereas higher D_d values (5.3 ± 2.4 km km⁻²)
358 characterized the smaller catchments (0.1-9 km²; Table 3).

359 Soil depth and water retention (WR) capacity of the solum were greatest in hillslopes
360 located in the middle portion of the LG [catchment](#); maximum WR values were observed in the
361 headwater mature (MAT) and secondary (SEC) forest catchments, and in other small catchments
362 ≤ 0.5 km² dominated by TMCF (Category 15; Figure 4a). Catchments with areas of
363 approximately 2 km² were dominated (≥ 50 %) by soil depths and WR capacities ranging
364 between 1.0 and 1.5 m and 580 and 850 mm, respectively (Category 14). Shallower soil depths
365 (from 0.5 to 1 m) and reduced WR values (from 310 to 510 mm; Category 13) characterized the
366 slope areas (46 % on average) of the larger catchments (9-34 km²). CATM5 showed the highest
367 proportion of area (33 %) covered by very shallow soils and relatively low water retentions
368 (Category 12).

369 Across [all](#) sites, the depth to the soil-bedrock interface ([DSBI](#)) ranged from 0.5 to more
370 than 2 m, and soil saturated hydraulic conductivity (K_s) at the interface ranged from 1 to nearly
371 40 mm hr⁻¹. However, for the majority of the catchments, the [DSBI](#) was between 1.0 and 1.5 m
372 (~ 65 % of the catchment area on average), with corresponding K_s values between 1 and 15 mm
373 hr⁻¹ (Category 2C; Figure 4b). Notably, the SEC was dominated by [DSBIs](#) between 1 and 2 m
374 (70 % of the catchment area); at some locations [in this catchment DSBI](#) was greater than 2 m,
375 with permeabilities at the soil-bedrock interface higher than 36 mm hr⁻¹ (Categories 2A and 1A).
376 In contrast, the Huehueyapan catchment showed the highest percentage of area (30 %) covered
377 by very low [DSBI](#) values (0.5 – 1.0 m on average) of all catchments, but K_s ranged from 4 to 36
378 mm hr⁻¹ (Categories 4C and 3B).

379

380 Table 3

381 Figure 4

382

383 **3.2 Stream baseflow MTTs and their relationship with catchments characteristics**

384 Estimated baseflow mean transit time (MTT) ranged between 1.2 and 2.7 years across the
385 12 [study](#) catchments. [Note the TTD model that we reported for a particular catchment was the](#)
386 [one that best fitted the observed baseflow data \(Table 4\).](#) The root mean square error (RMSE)
387 and Nash-Sutcliffe efficiency value (E) for these model results ranged from 0.8 to 1.5 ‰ (δ^2H)
388 and 0.42 to 0.69, respectively. Table 4 provides further details on the values of the [TTD](#)
389 parameters and the uncertainty bounds.

390 Catchment form, slope, land cover and depth to soil-bedrock interface explained each
391 about 50 % of the variance of baseflow MTT across catchments (Figure 5). The positive
392 correlation found between form factor (R_f) and baseflow MTT suggests that catchments with
393 narrow and elongated shapes lead to shorter transit times (Table 5; Figure 5a). Long MTTs were
394 positively correlated with moderately steep catchments (particularly where slopes between 20
395 and 30° predominated; Figure 5c). Conversely, short MTTs were most strongly related to
396 catchments with high proportions of gently slopes (between 5 and 10°). Interestingly, catchments
397 covered by areas with very steep slopes (> 30°) showed very poor correlations with MTTs. Weak
398 correlations were also obtained with catchment drainage density and mean slope length.

399 Soil WR categories determined along the hillslope transects did not explain much of the
400 variation in baseflow MTTs. Instead, a strong [and](#) positive relation was observed between MTT
401 and [DSBI; specifically longer stream transit times were related to catchment hillslopes where](#)
402 [deep \(> 2 m\) soil-bedrock interfaces dominated](#) (Figure 5f). Conversely, low and negative
403 correlations were obtained with shallower depths (< 1 m; Table 5). Regardless of the [DSBI](#)
404 classes, observed K_s values remained generally high across all sites (5–30 mm hr⁻¹; on average).

405 Land cover explained a significant variation of the [baseflow](#) MTT (Table 5; Figure 5e);
406 catchments covered by more than 60 % of forests (Categories 1 and 2) had on average longest
407 MTTs (1.9 ± 0.4 (SD) yr) compared to catchments dominated by ≥ 90 % of pasture or evenly
408 mixed covers with pasture and forest (1.5 ± 0.2 yr; Categories 3 and 4).

409 Baseflow MTT showed no relation to catchment area (Table 5; Figure 5b). However, at
410 the smallest scale (< 0.3 km²), major differences in the MTT were found (1.5 to 2.7 yr). At the
411 intermediate scale (4–9 km²), differences in MTTs (1.4 to 1.9 yr) were small among catchments.
412 At the larger scale (≥ 14 km²), some more variation in the [baseflow](#) MTTs was observed [\(1.2 to 2](#)
413 [yr\)](#). The 20 km² Huehueyapan showed the shortest baseflow transit times (1.2 yr) compared to

414 other large catchments examined—this was also the lowest MTT estimated across all the study
415 catchments.

416

417 Table 4

418 Table 5

419 Figure 5

420

421 **4. Discussion**

422 **4.1 How do our baseflow MTTs compare to those found in other tropical montane streams?**

423 Our stable isotope data showed that wet season rainfall is the main catchment stream
424 water source in this tropical montane region. This is similar to findings in other humid tropical
425 environments (Crespo et al., 2012; Roa García and Weiler, 2010; Scholl and Murphy, 2014), but
426 | contrasts with results for temperate regions where seasonality in flow regime is usually [much](#)
427 | more pronounced and, consequently, stream water tends to reflect input sources from different
428 seasons (Brooks et al., 2012; Mueller et al., 2013; Peralta-Tapia et al., 2015). Our estimates of
429 base flow transit times ranged between 1.2 and 2.7 years across the 12 study catchments. These
430 rather long transit times suggest deep, and presumably long subsurface flow paths, contributing
431 to sustain catchment baseflows across scales (0.1 to 34 km²) and seasons.

432 Comparing our results with those obtained by Roa-García and Weiler (2010) for three
433 adjacent headwater catchments differing in size (0.6–1.8 km²) and land cover (forest versus
434 | pasture) in central-western Colombia, our [baseflow](#) MTTs for the cloud forest catchments (~ 2.7
435 | yr; 0.1–0.3 km²) are almost twice the value obtained for their forest-dominated catchment (1.4
436 yr). Further, for two pasture-dominated catchments, these authors obtained MTTs that differed
437 considerably (0.1 and 1.4 yr), which they attributed to differences in soil permeability.
438 Furthermore, the relatively short stream MTTs in the Andean catchments were attributed to the
439 relatively low hydraulic conductivities that characterize the volcanic soils (Acrodoxic
440 Hapludans) of that region, limiting rain water percolation and promoting near-surface flow (Roa-
441 García and Weiler, 2010). This contrasts with our sites, where deep subsurface flow rather than
442 shallow lateral flow is the dominant flowpath for runoff generation (Muñoz-Villers and
443 | McDonnell, 2012, 2013).

444 In southern Ecuador, Crespo et al. (2012) used a simple sine-wave approach to estimate
445 the [baseflow](#) MTTs for a 74 km² nested mesoscale watershed (the San Francisco river basin),
446 underlain mostly by Histosols. They found MTTs on the order of 0.7–0.9 yr for nine cloud forest
447 catchments (1.3–74 km²). Further, for a 0.8 km² pasture catchment, they reported a MTT of 0.8
448 yr. Shallow lateral subsurface flow and high catchment runoff ratios (76–81 %) due to relatively
449 low topsoil and subsurface permeabilities (14–166 mm hr⁻¹) characterized the hydrology of that
450 montane area (Crespo et al., 2012). In contrast, soil hydraulic conductivities at our site were
451 higher (~ 400 mm hr⁻¹ on average across land covers; Muñoz-Villers et al., [2015](#)), leading to
452 lower (annual) rainfall-runoff ratios (35–50 %), and hydrological responses mainly driven by
453 groundwater sources, which likely explain [the](#) much larger catchment water storage capacities of
454 our systems.

455 For eight of the catchments in the San Francisco river basin previously investigated by
456 Crespo et al. (2012), Timbe et al. (2014) obtained much higher MTTs values by fitting several
457 TTD models. For seven cloud forest dominated catchments (1.3–77 km²), they reported an
458 average MTT value of 2.1 yr, while for a pasture catchment they obtained a MTT value (3.9 yr)
459 that was twice the average value for the forests. However, the authors did not provide an
460 explanation of why they found longer MTTs and contradictory results (i.e. higher MTT in the
461 pasture than in the forests) compared to the earlier work by Crespo et al. (2012).

462

463 **4.2 Factors determining baseflow MTTs in this tropical montane watershed**

464 It is well known that topography plays an important role in the transit time of water
465 through catchments (Tetzlaff et al., 2009a), particularly in montane environments (cf. McGuire
466 et al., 2005). Our findings are consistent with previous work and show that longest baseflow
467 [MTTs](#) are related to rounded shapes of catchment (0.19–0.23), where moderate slope gradients
468 (20–30°) predominate. In contrast, catchments with elongated forms—regardless of [their](#) internal
469 slope assemblages—produced the shortest [MTT](#) estimates. Our interpretation is that in narrow
470 forms, the hydrological connectivity between hillslopes and the stream is higher than [in](#)
471 catchments with more rounded shapes. This in turn would increase the frequency of water table
472 formation and response to precipitation leading to shorter water travel times. Related work on
473 this was carried out by Hrachowitz et al. (2009) in the Scottish Highlands, who evaluated the
474 influence of topography on stream MTT. In their study, form factor ratios and drainage densities

475 were computed for 20 different catchments (< 1 to 35 km²). Their work showed that elongated
476 forms of catchments were roughly distinguished from rounded shapes. Drainage density,
477 however, characterized much better the catchments topography of that region showing a strong
478 and inverse relationship with stream MTTs. They found high drainage density values associated
479 to high percentages of responsive soil cover (peat soils) as rapid water routed via overland flow
480 enhances connectivity between hillslopes and stream channel network. In our study site,
481 drainage density was inversely related to slope length (data not shown) and showed no
482 correlation with soil type as Andisols dominate entirely the hillslopes of our [catchments](#).

483 We also explored the influences of land cover on [baseflow](#) MTT. Our findings showed
484 that catchments covered predominantly by forests had longer MTT estimates compared to
485 catchments dominated by pasture. We attributed this to topographic differences among sites
486 more than land cover effects since most forested areas are themselves located on steep terrain.
487 This is supported by results obtained by Muñoz-Villers and McDonnell (2013), who investigated
488 the streamflow dynamics at the mature and secondary TMCF and pasture headwater catchments.
489 They found on average 50 % lower baseflow in the pasture at the end of the dry season compared
490 to forests, explained by lower recharge of subsurface water storages due to [smaller](#) catchment
491 gradients (cf. Sayama et al., 2011; Tetzlaff et al., 2009b) in the pasture, and lower surface soil
492 infiltration capacities caused by animal grazing compaction ($30 \pm 14 \text{ mm hr}^{-1}$ versus 696 ± 810
493 mm hr^{-1}). Thus, the fact that forested catchments have [steeper](#) slopes and higher topsoil
494 infiltration capacities might be a more likely explanation for their higher subsurface water
495 storage capacities.

496 In general, very few studies have investigated the effect of land cover on catchment
497 stream MTTs. Mueller et al. (2013) studied the influence of shrub cover area on MTTs in four
498 micro catchments in the Swiss Alps. They found that soil and bedrock hydraulic characteristics
499 had a stronger control on stream transit times rather than land cover. [High](#) subsurface flow
500 promoted by fast soil water percolation through fractured [bedrock, which can contain](#) karstic
501 rock [in deeper layers](#), dominated the catchment water storage, mixing and release in this alpine
502 environment. More recently, Geris et al. (2015) investigated the relative influence of soil [type](#)
503 and vegetation cover on storage and transmission processes in a headwater catchment (3.2 km²)
504 in northeast Scotland. Forested and non-forested sites were compared on poorly drained
505 Histosols in riparian zones and freely draining Podzols on steeper hillslopes. Their results

506 | showed that soil permeability properties linked to soil type [rather than vegetation influences](#)
507 | were dominant features on water storage dynamics at the plot and catchment scales.

508 | Our study determined the depth [and permeability](#) of the soil-bedrock [interface](#) through
509 | intensive [and extensive](#) measurements [in the subsurface over numerous hillslope transects](#) across
510 | the LG [catchment](#). This is rather unlike most studies that have derived flow path depths and
511 | source contributing areas to stream discharge from surface topography based on digital terrain
512 | models (Hrachowitz et al., 2010; McGuire et al., 2005; Tetzlaff et al., 2009b) or from
513 | geochemical tracers such as SiO₂ (Asano and Uchida, 2012). Our approach showed that
514 | hillslopes with deeper soils along with high hydraulic conductivities at the soil-bedrock interface
515 | allowed more subsurface water transmission and storage, leading to longer catchment baseflow
516 | transit times. In this case, longest stream MTTs (ca. 3 yr) were obtained in the mature and
517 | secondary TMCF headwater catchments, associated to their highest percentage of area covered
518 | by deep soil-bedrock profiles related in turn to their moderate steep relief, and greatest
519 | subsurface permeabilities. Previous work at these sites showed that the very high permeability of
520 | the Andisols (1000 mm hr⁻¹ at 0.1 m to 4 mm hr⁻¹ at 1.5 m depth; Karlsen, 2010) and underlying
521 | volcanic substrate promote vertical and fast soil water percolation and recharge of deeper
522 | sources, as the preferred flow path mechanism controlling catchment water storage and storm
523 | runoff responses (Muñoz-Villers and McDonnell, 2013).

524 | Across all catchments, the observed range of saturated hydraulic conductivities at the
525 | soil-bedrock interface was from 5 to 30 mm hr⁻¹, suggesting little impedance for water to
526 | continue percolating vertically below the soil profile and to recharge ground water reservoirs.
527 | This could explain the generally long MTTs found across sites (1.8 yr on average). Further, we
528 | observed greatest depths to bedrock at mid and ridge top hillslope positions (data not shown).
529 | Thus these topographic features seem to be the main contributing areas to subsurface [water](#)
530 | recharge. While soil water retention capacities were also greatest at mid and ridge top slope
531 | positions, they did not explain much of the variation in baseflow MTTs.

532 | [These](#) findings are partly consistent with those obtained by Asano and Uchida (2012) in
533 | central Japan, who examined the baseflow MTT spatial variation for a 4.3 km² forested montane
534 | watershed underlain by granitic soils. They used the dampening of the isotopic signal as a proxy
535 | for the relative difference in MTTs among locations. They also used dissolved silica as a tracer
536 | to identify the contributing depth of the flow path to stream discharge. Their work showed that

537 the depth of hydrologically active soil-bedrock layer was the main factor determining catchment
538 water storage. Longer baseflow MTT were associated to increased flow path contributions
539 related in turn to hillslope length and topography. McGuire et al. (2005) also showed strong
540 correlations between catchment terrain indices (flow path length) and mean stream residence
541 times for seven catchments (0.085–62.4 km²) in the western Cascade Mountains of Oregon,
542 USA, showing that landscape organization was the main factor controlling catchment-scale water
543 transport.

544 While some investigations have reported that catchment area controls the variation in
545 stream MTT (i.e. Hale and McDonnell, 2016; McGlynn et al., 2003), the majority of the work
546 published to date has shown no relation between MTT and catchment size for [catchments](#)
547 ranging between 0.1 and 200 km² (Crespo et al., 2012; McGuire et al., 2005; Mueller et al.,
548 2013; Rodgers et al., 2005; Soulsby et al., 2006). Our findings support these latter studies and
549 show that increasing catchment area does not lead to longer mean stream travel times.

550 We [also](#) found that baseflow MTTs were more variable in smaller catchments (0.1-1.5
551 km² sizes) where topography imposed its strongest effect (cf. Hrachowitz et al., 2010; Tetzlaff et
552 al., 2009b). Further, longer MTT were found at the forest-dominated headwater catchments (\leq
553 [0.25 km²](#); ~ 3 yr). This is similar to the findings obtained by Timbe et al. (2014) in a tropical
554 montane cloud forest watershed underlain by Histosols in southern Ecuador, who reported longer
555 and larger variation of MTTs in small streams (0.1 – 5 km²; $\sim 3 \pm 1.09$ yr) in comparison to
556 downstream tributaries and main river channels (10 – 77 km²; $\sim 2 \pm 0.08$ yr). At the intermediate
557 scale (4 – 9 km²), [our](#) differences in MTTs were small and associated probably to catchment
558 topography. Unexpectedly, MTTs showed a slight convergent pattern at this scale (Figure 5b)
559 (cf. Hrachowitz et al., 2010; Timbe et al., 2014). Beyond this scale (≥ 9 km²), catchment
560 processes determining streamflow behavior seems to be different. For instance, the Huehueyapan
561 watershed (20 km²) showed the shortest MTT (1.2 yr) across all catchments investigated. [We](#)
562 [attributed](#) its low water storage capacity to its narrow form, combination of gentle and [very](#) steep
563 slope [areas](#) (the latter limiting the development of soil on its hillslopes), [and shallow depths to](#)
564 [bedrock](#). Alternatively, the main outlet of the [LG catchment](#) (35 km²) has an MTT of 2.2 yr that
565 nearly doubled the MTT value of its main tributary (Huehueyapan [catchment](#)). This might
566 suggest that runoff processes of smaller catchments does not necessary combine to define MTT

567 at larger scales ($> 14 \text{ km}^2$)(cf. Shaman et al., 2004), probably due to changes in geomorphology,
568 related in turn to past landscape formation of this volcanic area.

569

570 **5. Conclusions**

571 This study provides an important first step towards a better understanding [of the](#)
572 hydrology of tropical montane regions and the factors influencing stream water transit times in
573 these environments. Our estimates of baseflow MTT ranged between 1.2 and 2.7 years across 12
574 catchments (0.1 to 34 km^2) in central Veracruz, Mexico, suggesting deep and presumably long
575 subsurface flow paths contributing to sustain baseflows, particularly during dry periods. Our
576 findings showed that catchment slope and the permeability observed at the soil-bedrock interface
577 are the key factors controlling baseflow MTT in this tropical montane region. [Longest stream](#)
578 [MTTs were found in the cloud forest headwater catchments, related to their moderate steep](#)
579 [slopes, and greater transmissivity at the soil-bedrock interface. Conversely, the MTT was](#)
580 [shortest in one tributary of the main river outlet, which was](#) [mainly](#) [attributed to its](#) [high](#)
581 [proportions of](#) [both](#) [gentle and very steep slopes. In association with topography, catchment form](#)
582 [and the depth to the soil-bedrock interface were also identified as important features influencing](#)
583 [baseflow MTT variability across scales. More specifically, longer baseflow MTTs appeared to be](#)
584 [related to rounded shapes of catchments and deeper soil-bedrock interfaces. Greatest depths to](#)
585 [bedrock were particularly observed in the](#) [mid](#) [and ridge top hillslope positions, thus these](#)
586 [topographic locations](#) seemed to be the main contributing areas for catchment subsurface [water](#)
587 recharge. Major differences in MTTs were found [both within groups of](#) [small](#) (0.1 – 1.5 km^2) and
588 [large](#) (14 – 34 km^2) [catchments](#), related mostly to catchment slope and morphology, and to much
589 lesser extent [to](#) land cover.

590

591 **Author contribution**

592 L.E.M. and J.J.M. developed the idea of this research. L.E.M., D.R.G. and F.H. collected data.
593 L.E.M. and D.R.G analyzed and interpreted data. L.E.M. wrote the first draft of the manuscript.
594 D.R.G. and F.H. edited and commented on this first draft. D.R.G. and J.J.M. edited and
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596

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Table 1. Topographic characteristics of the 12 catchments (0.1–34 km²) investigated

#ID	Catchment	Area (km ²)	Stream order	Mean elevation (m a.s.l.)	Elevation range (m a.s.l.)
1	MAT	0.25	1	2160	2020-2300
2	SEC	0.12	2	2130	2040-2220
3	PAS	0.10	1	2400	2320-2480
4	CATM1	0.46	2	2230	1980-2480
5	CATM2	0.62	2	2230	1980-2480
6	CATM3	1.9	3	2380	2000-2760
7	CATM4	1.5	2	2240	1860-2620
8	CATM5	4.1	2	2050	1340-2760
9	CATM6	8.9	4	1980	1340-2620
10	PUENTE ZARAGOZA	13.5	4	2030	1300-2760
11	HUEHUEYAPAN	19.7	4	2120	1300-2940
12	LOS GAVILANES	33.5	5	2120	1300-2940

Table 2. Annual, and wet and dry season means of the isotopic composition of rainfall (3 sites) and stream water (12 sampling locations) plus corresponding values of *d*-excess, as calculated from 2 years of data (April 2008 – May 2010)

Rainfall	VWM Annual			VWM Wet season			VWM Dry season		
	$\delta^2\text{H}$	$\delta^{18}\text{O}$	<i>d</i> -excess	$\delta^2\text{H}$	$\delta^{18}\text{O}$	<i>d</i> -excess	$\delta^2\text{H}$	$\delta^{18}\text{O}$	<i>d</i> -excess
TG (2400 m)	-43.0	-7.5	17.0	-48.2	-8.0	15.8	-23.7	-5.5	20.3
SECP (2100 m)	-37.6	-6.7	16.0	-43.6	-7.4	15.6	-18.9	-4.6	17.9
RA (1560 m)	-33.4	-6.1	15.4	-44.0	-7.4	15.2	-12.2	-3.7	17.4
Catchments	Mean annual			Mean wet season			Mean dry season		
	$\delta^2\text{H}$	$\delta^{18}\text{O}$	<i>d</i> -excess	$\delta^2\text{H}$	$\delta^{18}\text{O}$	<i>d</i> -excess	$\delta^2\text{H}$	$\delta^{18}\text{O}$	<i>d</i> -excess
MAT	-42.5	-7.3	15.9	-43.1	-7.4	16.1	-41.8	-7.2	15.8
SEC	-41.8	-7.2	15.8	-42.5	-7.3	15.9	-40.9	-7.0	15.1
PAS	-47.7	-7.9	15.5	-47.7	-7.9	15.5	-47.6	-7.8	14.8
CATM1	-49.4	-8.1	15.4	-48.9	-8.1	15.9	-50.1	-8.1	14.7
CATM2	-47.2	-7.8	15.2	-47.0	-7.8	15.4	-47.4	-7.8	15.0
CATM3	-46.8	-7.8	15.6	-46.4	-7.7	15.2	-47.4	-7.9	15.8
CATM4	-44.8	-7.5	15.2	-42.5	-7.6	18.3	-44.3	-7.4	14.9
CATM5	-42.5	-7.3	15.9	-42.8	-7.3	15.6	-42.1	-7.3	16.3
CATM6	-41.8	-7.2	15.8	-42.2	-7.2	15.4	-41.3	-7.2	16.3
PUENTE ZARAGOZA	-42.1	-7.3	16.3	-42.5	-7.3	15.9	-41.7	-7.2	15.9
HUEHUEYAPAN	-46.1	-7.7	15.5	-46.6	-7.8	15.8	-45.4	-7.7	16.2
LOS GAVILANES	-43.3	-7.4	15.9	-43.7	-7.4	15.5	-43.0	-7.4	16.2

Table 3. Slope length and gradient, drainage density, form factor and land cover of the study catchments

Catchment	Mean slope length (m)	Drainage density (km km ⁻²)	Form factor (-)	% of cover per slope class						Land cover category and % of vegetation coverage
				0-5°	5-10°	10-20°	20-30°	30-45°	≥ 45°	
MAT	123	4.8	0.222	6	14	35	33	11	1	[1]: 100% mature TMCF
SEC	105	5.8	0.231	6	14	33	31	15	1	[1]: 100% secondary TMCF
PAS	68	7.0	0.164	7	39	36	13	5	0	[3]: 90% pasture; 10% shrubs
CATM1	80	7.9	0.071	6	21	28	26	18	1	[2]: 67% TMCF; 31% pasture; 2% pine-oak forest
CATM2	77	8.0	0.093	6	21	29	26	17	1	[2]: 66% TMCF; 33% pasture; 1% pine-oak forest
CATM3	190	2.6	0.122	7	30	34	17	11	1	[4]: 49% pasture; 36% TMCF; 15% pine-oak forest
CATM4	150	3.2	0.087	6	30	36	17	10	1	[4]: 55% pasture; 34% TMCF
CATM5	208	6.8	0.071	9	23	29	20	17	2	[2]: 55% TMCF; 35% pasture; 6% pine-oak forest
CATM6	225	1.3	0.131	5	15	25	25	25	5	[2]: 65% TMCF; 27% pasture; 5% pine-oak forest
PUENTE ZARAGOZA	235	2.9	0.187	7	17	26	24	22	4	[2]: 62% TMCF; 29% pasture; 5% pine-oak forest
HUEHUEYAPAN	300	2.0	0.134	13	16	18	21	25	8	[2]: 43% TMCF; 29% pine-oak forest; 21% pasture
LOS GAVILANES	285	2.4	0.220	9	16	21	22	20	7	[2]: 51% TMCF; 24% pasture; 20% pine-oak forest

Table 4. Stream baseflow MTTs, and corresponding model parameters and model efficiencies

Catchments	MTT (days)	Model ^a	Model parameters	Model efficiency	
MAT	958	Gamma (α, β)	α, β 0.74 (0.70, 0.85), 1299 (524, 1137)	E 0.53	RMSE ($\delta^2\text{H}$, ‰) 1.5
SEC	975	Gamma (α, β)	0.74 (0.59, 0.93), 1326 (484, 2329)	0.68	1.4
PAS	548	Exponential (τ_m)	τ_m 548 (493, 609)	0.57	1.0
CATM1	531	Exponential (τ_m)	τ_m 531(514, 550)	0.58	1.0
CATM2	636	Dispersion (τ_m, D_p)	τ_m, D_p 636 (463,824) 0.66 (0.44, 0.89)	0.66	1.1
CATM3	624	Dispersion (τ_m, D_p)	τ_m, D_p 624 (536,734) 0.85(0.68, 0.96)	0.45	1.0
CATM4	522	Dispersion (τ_m, D_p)	τ_m, D_p 522 (451,571) 2.2 (1.4, 3.0)	0.53	1.4
CATM5	710	Exponential (τ_m)	τ_m 710 (555, 859)	0.63	0.8
CATM6	702	Exponential (τ_m)	τ_m 702 (550, 856)	0.64	0.9

PUENTE ZARAGOZA	633	Exponential (τ_m)	τ_m 633 (520, 751)	0.64	0.9
HUEHUEYAPAN	424	Exponential (τ_m)	τ_m 424 (371, 482)	0.63	1.2
LOS GAVILANES	788	Exponential (τ_m)	τ_m 788 (646, 935)	0.42	1.0

MTT is the mean transit time, E is the Nash-Sutcliffe efficiency and RMSE is the root mean square error.

Numbers in parenthesis are the 10th and 90th percentile values of the MTT estimates and the model parameters.

^a [Refer to the supplementary information for the corresponding formulas of the TTD models.](#)

Table 5. Spearman's rank correlation coefficients (r_s) between baseflow MTT and land cover, catchment area, topographic characteristics and subsurface hydrologic properties for the study [catchments](#).

	r_s
Land cover	-0.74
Area (km ²)	-0.09
Form factor (-)	0.56
Drainage density (km km ⁻²)	0.05
Mean slope length (m)	-0.13
Slope 0-5°	-0.22
Slope 5-10°	-0.63
Slope 10-20°	-0.01
Slope 20-30°	0.57
Slope 30-45°	0.04
Slope > 45°	0.06
DSBI > 200 cm	0.48
100 < DSBI ≤ 200 cm	-0.28
50 < DSBI ≤ 100 cm	-0.15
DSBI ≤ 50 cm	-0.08
Soil WR per category	
11	-0.08
12	0.24
13	-0.18
14	0.30
15	-0.25

Figures and captions

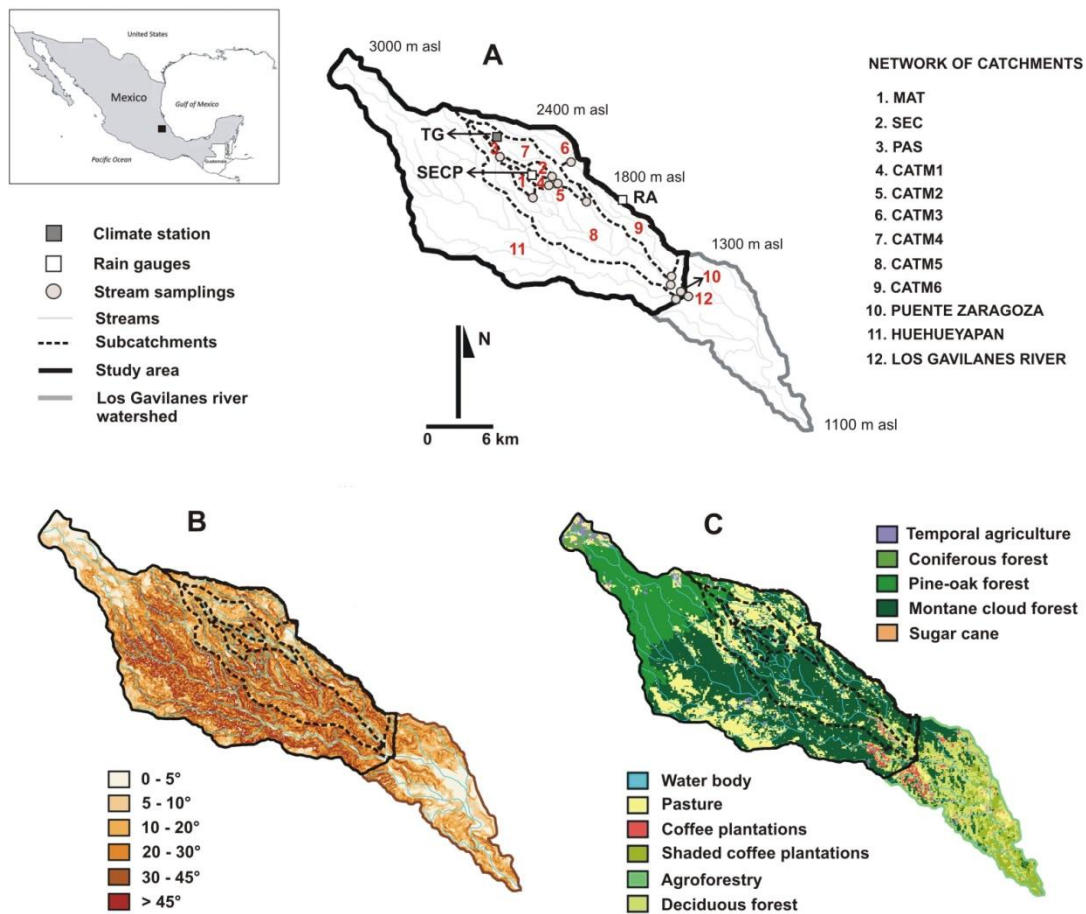


Figure 1. Location of the study site in central Veracruz, eastern Mexico, and maps of the Los Gavilanes catchment showing (A) the stream and rain water collection points; (B) slopes; and (C) land covers (see text for further explanation).

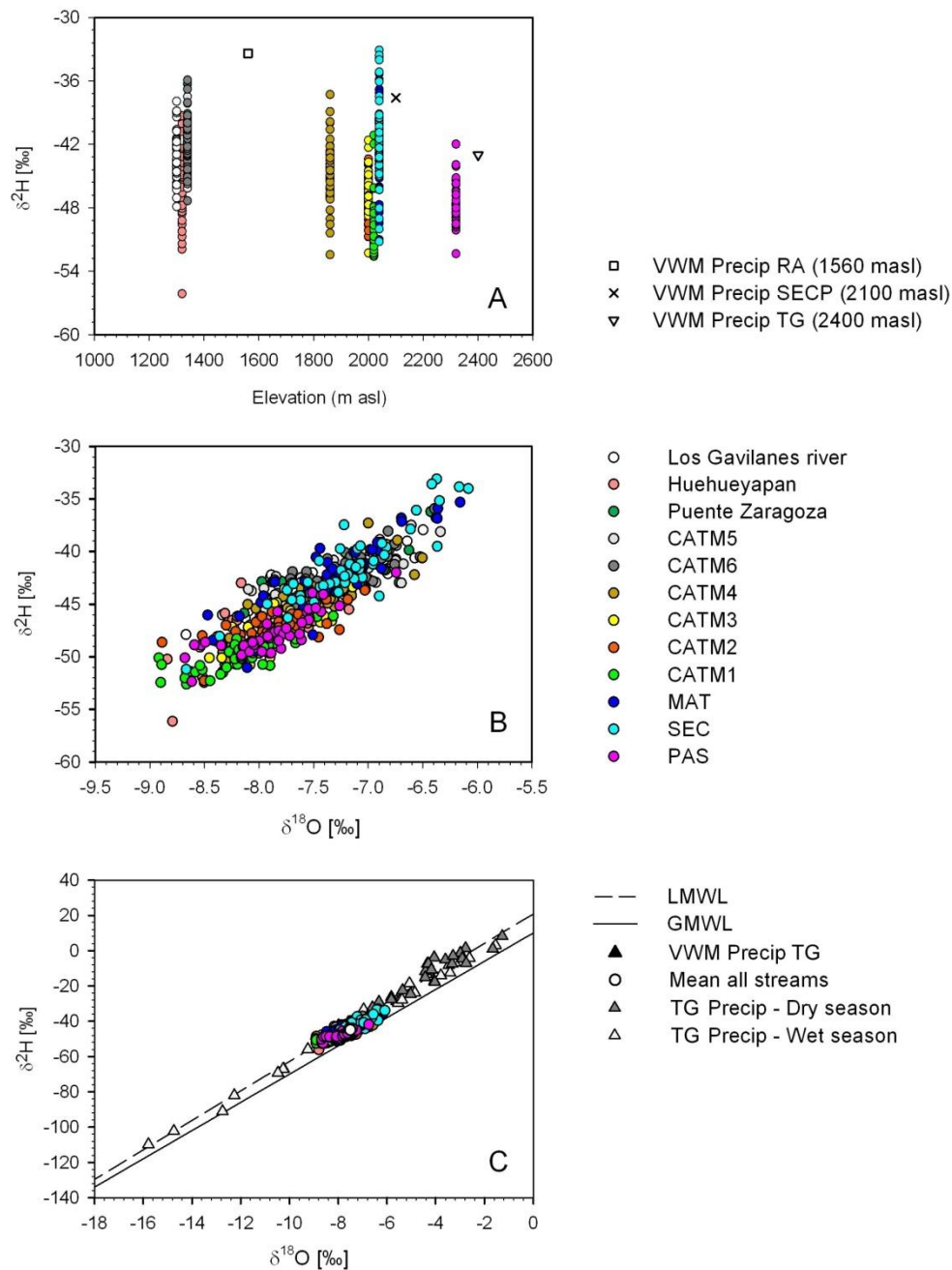


Figure 2. (A) Relationship between $\delta^2\text{H}$ signatures of stream base flow and elevation of the catchment outlets (i.e. the sampling locations), and volume-weighted means (VWMs) of the deuterium composition of rainfall at the three elevations within the Los Gavilanes river [catchment](#); (B) $\delta^{18}\text{O}$ versus $\delta^2\text{H}$ signatures of baseflow from the 12 catchments investigated; and (C) Isotope ($\delta^{18}\text{O}$ versus $\delta^2\text{H}$) signatures of rainfall and stream baseflow. The local meteoric water line (LMWL; dashed line) is based on the 2008-2010 precipitation at TG, and reads: $\delta^2\text{H}=8.36* \delta^{18}\text{O} + 20.37$; the solid line represents the global meteoric water line (GMWL): $\delta^2\text{H}=8* \delta^{18}\text{O} + 10$.

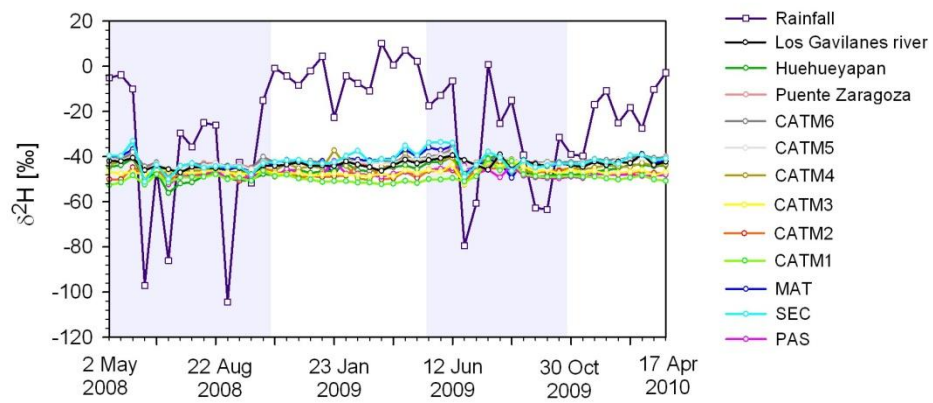


Figure 3. Biweekly values of $\delta^2\text{H}$ composition of stream baseflow for each of the 12 study catchments, and corresponding values of deuterium composition of rainfall at 2400 m (TG) for the period between May 2008 and April 2010. The shaded areas indicate the wet seasons.

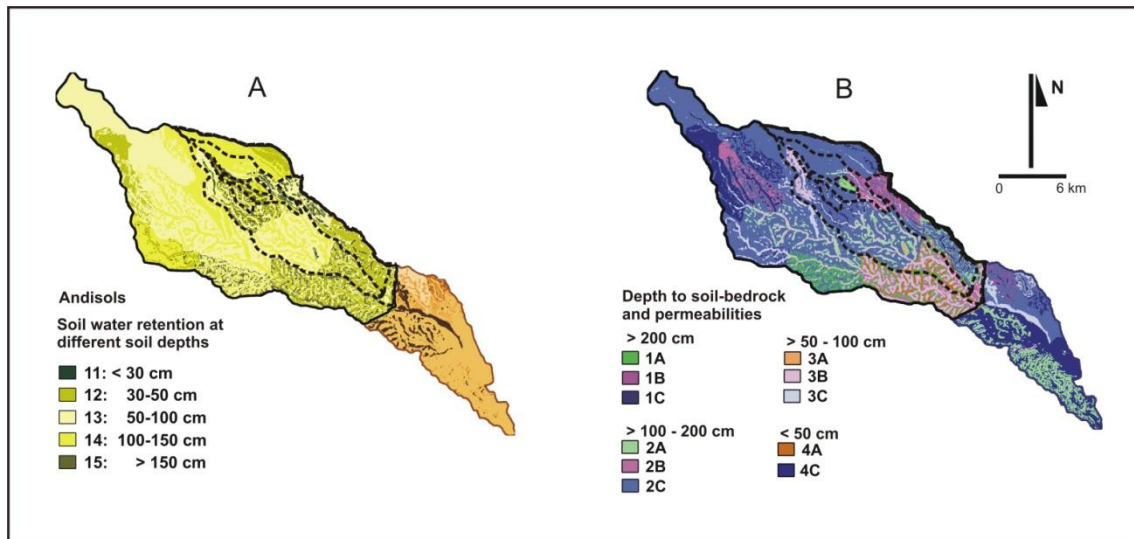


Figure 4. Map of hydro-pedologic properties of the Los Gavilanes river [catchment](#). (A) Soil water retention at field capacity in the solum. Category 11: < 180 mm; Category 12: $\geq 180 \leq 310$ mm; Category 13: $\geq 310 \leq 580$ mm; Category 14: $\geq 580 \leq 850$ mm; and Category 15: ≥ 850 mm. (B) Depth to soil-bedrock interface and corresponding saturated hydraulic conductivities (K_s). For depth ≤ 100 cm, K_s categories A, B and C correspond to: $K_s > 36$; $14 < K_s \leq 36$ and $1 < K_s \leq 14$ mm hr⁻¹, respectively. For depth ≥ 100 cm, A, B and C correspond to $K_s > 144$; $36 < K_s \leq 144$ and $4 < K_s \leq 36$ mm hr⁻¹, respectively.

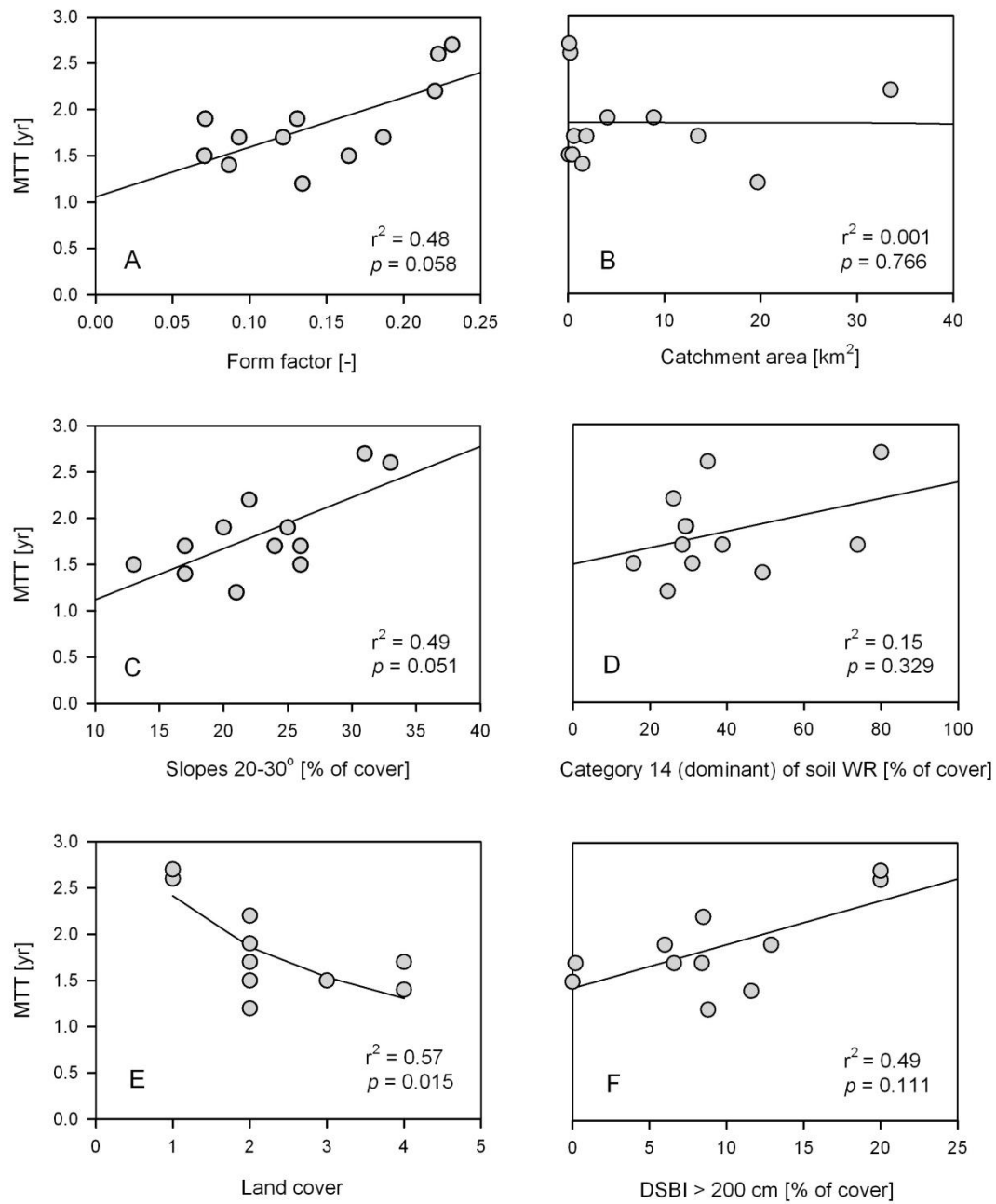


Figure 5. Regressions between stream baseflow MTTs and topographic features, subsurface properties, land cover and catchment area for the study catchments.