Resiliency of Silica Export Signatures When Low Order Streams Are Subject to Storm Events

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Abstract Silicon stable isotope ratios ($\delta^{30}$Si) of over 150 stream water samples were measured during seven storm events in six small critical zone observatory (CZO) catchments spanning a wide range in climate (sub-humid to wet, tropical) and lithology (granite, volcanic, and mixed sedimentary). Here we report a cross-site analysis of this dataset to gain insight into stream $\delta^{30}$Si variability across low-order catchments and to identify potential climate (i.e., runoff), hydrologic, lithologic, and biogeochemical controls on observed stream Si chemical and isotopic signatures. Event-based $\delta^{30}$Si exhibit variability both within and across sites ($-0.22\%$e to $+2.27\%e$) on the scale of what is observed globally in both small catchments and large rivers. Notably, each site shows distinct $\delta^{30}$Si signatures that are preserved even after normalization for bedrock composition. Successful characterization of observed cross-site behavior requires the merging of two distinct frameworks in a novel combined model describing both non-uniform fluid transit time distributions and multiple fractionating pathways in application to low-order catchments. The combined model reveals that site-specific architecture (i.e., biogeochemical reaction pathways and hydrologic routing) regulates stream silicon export signatures even when subject to extreme precipitation events.

Plain Language Summary Periods of intense infiltration in the form of rainfall (storm events) or snowmelt push large volumes of dilute fluid through catchments, resulting in evacuation of dissolved solutes derived from weathering reactions. These events occur over short (daily or weekly) timescales, but these pulses of solutes released into small streams can represent a majority of total annual mass flux (from 40% to 70%). Despite their importance, storm events remain largely underrepresented in studies of weathering in watersheds due to challenges in obtaining high frequency measurements. Yet extreme events are expected to become more frequent in the future in response to a changing climate. In this study, we evaluate the dependence of weathering-derived solute export on hydrology and biogeochemical cycling through measurements of dissolved silicon and silicon stable isotope ratios of stream waters collected during seven significant infiltration events in six small catchments spanning different bedrock lithology and climates. Each catchment was found to have unique silicon chemical and isotopic export signatures reflecting site-specific subsurface water routing and combination of biogeochemical reactions. These signatures were found to be preserved despite the perturbation exerted by such enhanced infiltration and discharge. In this sense, catchments exhibit resilience, bending but not ceding to strong hydrologic forcing.

1. Introduction
Silicate weathering is a critical facet of Earth's surface dynamics, transforming and mobilizing mass to shape landscapes, supplying essential nutrients to a diversity of ecosystems, and regulating Earth's climate over geologic timescales through the drawdown of atmospheric CO$_2$. A principal objective of modern Critical Zone (CZ) science is to better constrain the ensemble of (bio)geochemical, climate (temperature, runoff, precipitation), and lithologic controls driving silicate weathering rates. Evaluating these factors, notably those that are climate-related (such as runoff and precipitation), has proven to be a difficult task due to strong coupling between physical and chemical erosion rates (Millot et al., 2002; Riebe et al., 2001, 2003; West et al., 2005) and first-order lithologic controls (Amiotte Suchet et al., 2003; Bluth & Kump, 1994; Garrels & MacKenzie, 1971; Meybeck, 1987). In
this respect, the study of small, upland catchments hosting low-order streams offers an important opportunity to identify climate controls on silicate weathering. Low-order catchments are considered important silicate weathering “hotspots” (Bluth & Kump, 1994; Drever & Clow, 1995; Drever & Zobrist, 1992; McClain et al., 2003; Millot et al., 2002; Oliva et al., 2003; Stallard, 1985; Stallard & Edmond, 1983, 1987; White & Blum, 1995; White et al., 1999), while their size (<100 km²), generally homogenous lithology, and less diverse hydrologic characteristics makes them relatively simple systems (relative to higher order watersheds). Further, high physical erosion rates typically associated with small catchments (Dixon et al., 2009; Riebe et al., 2001; West et al., 2005) ensure a steady supply of fresh weatherable material accessible to meteoric fluids in the weathering zone, promoting and sustaining silicate weathering. In such a kinetically limited geomorphic regime (Riebe et al., 2017), (bio)geochemical reactivity and climate factors are the main drivers of silicate weathering rates, regulating both water delivery to the weathering zone and the extent of weathering reactions as dictated by kinetic and thermodynamic limitations (Maher, 2011).

Strong dependence on climate and hydrology implies that transient hydrologic events (“hot moments”, “storm events” or periods of high variability in discharge; McClain et al., 2003) should play a major role in small catchment silicate weathering fluxes. Indeed, storm events have been shown to disproportionately affect solute generation and transport dynamics at the catchment scale, constituting 40% to as high as 80% of the total annual rainfall and solute export fluxes (Larsen & Simon, 1993; Moatar et al., 2020; Raymond & Saiers, 2010; Yoon & Raymond, 2012). However, storm events are notoriously underrepresented in the geochemical literature due to challenges in high-frequency sampling; an area that has only recently seen significant progress with technological advancements like the River Lab (ExtraLab®. Floury et al., 2017). Such scarcity of high-resolution event data may have substantial consequences for our estimations of silicate weathering rates, which are determined primarily from long-term, low frequency (annual or seasonal) samples (Oliva et al., 2003; White & Blum, 1995). For example, a study of silica cycling dynamics in Siberian watersheds found spring flooding to account for 60%–80% of annual silica export fluxes (Pokrovsky et al., 2013). Missing such hot moments could result in a significant under-estimation of observed silicate weathering rates. Further, while tracer evidence shows that the biogeochemical pathways that produce weathering-derived solutes can change during these hydrologic events, there is still much to learn about how these processes change with varying hydrologic pathways and transit times (Arora et al., 2020; Benetti, Bailey, et al., 2017; Cenki-Tok et al., 2009; Derry et al., 2005; Kurtz et al., 2011; Rose et al., 2018). Finally, the large fraction of the annual dissolved solute riverine budget represented in storm events strongly links these moments to long-term silicate weathering behavior, a connection that has been implied in recent studies (Knapp et al., 2020; Minaudo et al., 2019; Rose et al., 2018).

From this basis it becomes clear that storm events provide a unique opportunity to evaluate the relationship between silicate weathering and catchment hydrology. Activation of quiescent water stores and reaction pathways during these intensive drainage events offers a unique glimpse into silicate weathering dynamics within a catchment and the interplay between various water stores, biological cycling, and residence times contributing to observed stream chemistry (Dwivedi et al., 2018; McIntosh et al., 2017; White et al., 2019; Zapata-Rios et al., 2015). Stream solute concentration (C)-discharge (Q) relationships serve as an effective tool to decipher these internal catchment solute dynamics. However, the seemingly infinite combinations of hydrologic and (bio)geochemical pathways, mixing and storage mechanisms that can be invoked to generate equivalent observations complicates interpretation of stream chemical signals when total element concentrations are solely relied upon.

This is consistent with longstanding observations and models illustrating low variance in geogenic solute concentrations across a large range of discharge rates (e.g., Anderson et al., 1997; Davis, 1964; Godsey et al., 2009; Kennedy, 1971; Scanlon et al., 2001). Such challenges can be alleviated through the complementary use of geochemical tools such as stable isotope ratios of weathering-derived elements (i.e., Li, Mg, Ca, B, Si, Fe, Cu, and Zn) that offer precise source signatures as well as indicators of the type and extent of biogeochemical reactions actively involved in solute cycling (Wiederhold, 2015). Silicon stable isotope ratios (denoted here as Δ²⁸Si) offer such diagnostics for the second most abundant element in Earth’s crust and are used frequently to infer and quantify solute fluxes associated with silicate weathering (Cardinal et al., 2010; Georg et al., 2006a, 2007, 2009; Hughes et al., 2011, 2012, 2013; Opfergelt & Delmelle, 2012; Pokrovsky et al., 2013; Riotte et al., 2018, etc.). Silicon stable isotopes can be fractionated by biological uptake, secondary mineral precipitation and adsorption reactions, thereby providing constraints on the relative contributions of these secondary processes to overall solute fluxes. Silicon stable isotopic shifts (Δε=10⁸(α−1), Coplen, 2011) associated with these Critical Zone processes typically vary on the order of –3.5‰ (amorphous silica precipitation, Fernandez et al., 2019; Roerdink...
et al., 2015) to ~0.33‰ (uptake by vascular plants, Frick et al., 2020). Likewise, globally observed river δ³⁰Si values exhibit variability of ~5‰ (Fringts et al., 2016) and are typically enriched in ³⁰Si relative to the bedrock and primary minerals comprising the surrounding lithology (De La Rocha et al., 2000; Ding et al., 2004; Georg et al., 2006a).

Those studies that have focused on temporal variation in riverine δ³⁰Si generally observe a decrease in dissolved Si and associated δ³⁰Si in rivers during periods of high discharge (Engström et al., 2010; Georg et al., 2006a; Pokrovsky et al., 2013; Röttte et al., 2018; Steinhoefel et al., 2017). Of these studies, none specifically targeted storm events and only two generated temporally resolved datasets to monitor seasonal variations in river δ³⁰Si (Engström et al., 2010; Pokrovsky et al., 2013). Seasonal stream δ³⁰Si variability from these efforts agreed with those observed globally, from 0.8‰ (Engström et al., 2010) to 1.0‰–1.5‰ (Pokrovsky et al., 2013). To explain observed stream δ³⁰Si behavior as a function of discharge several of these studies (Engström et al., 2010; Georg et al., 2006a; Pokrovsky et al., 2013) have invoked a combination of rapid weathering reactions, notably in situ dissolution of low δ³⁰Si clay fractions in suspended sediments within streams, and two-component mixing between a high δ³⁰Si, enriched dissolved Si soil/groundwater reservoir and a low δ³⁰Si, dissolved Si depleted surface runoff. Other studies such as Ziegler, Chadwick, White, et al. (2005) and Steinhoefel et al. (2017) have proposed alternative explanations based on the seasonally driven interplay between shallow/soil water and deep, steady-state groundwater (baseflow) contributions to observed stream chemistry. In this framework, activation of soil water stores and associated reaction pathways occur as a result of increasing water to rock ratios during high-discharge periods, ultimately driving stream δ³⁰Si behavior. The systematically observed decrease in stream δ³⁰Si with increasing discharge is interpreted in this context to reflect clay or phylolith dissolution (a source of “light” ²⁸Si) promoted by high dissolved organic carbon (DOC) exports derived from extensive biological cycling during dry summer periods.

Each of the mechanisms described above were developed to explain individual catchments. Extension of such site-specific conceptual models to serve as a global framework for short-term Si catchment dynamics remains an open question, limited by a lack of any cross-site analyses for δ³⁰Si of small catchments on the scale of what has been done for solute concentrations (Godsey et al., 2009, 2019; Oliva et al., 2003; White & Blum, 1995). Cross-site comparison of Si concentrations and mass fluxes suggest the dominance of site-specific catchment processes such as hydrologic function and permeability structure (Ameli et al., 2017; Wlostowski et al., 2021), soil characteristics (Bluth & Kump, 1994; Oliva et al., 2003), and weathering intensity (Riebe et al., 2001), which together serve as a fundamental driver of cross-site variability (Godsey et al., 2019; Hunsaker & Johnson, 2017). However, cross-site comparisons using high-frequency storm event data are rare (Wlostowski et al., 2021) and virtually non-existent in the silicate weathering and silicon stable isotope literature. Thus, we are left with a critical knowledge gap regarding how stream Si fluxes and δ³⁰Si compositions vary between periods of quiescence and these high-discharge events, as well as across catchments. Ultimately, this gap in knowledge obscures our capacity to identify site-specific effects versus universal controls on silicate weathering fluxes.

In the present study, we present a novel high frequency dataset of aqueous Si concentrations and stable isotopes (δ²⁸Si) for multiple storm events in multiple low-order streams, providing a first of its kind cross-site comparison of Si solute and isotope dynamics. We complement this dataset with new analyses of the germanium to silicon ratios (Ge/Si) promoted by high dissolved organic carbon (DOC) exports derived from extensive biological cycling during dry summer periods.

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1. How widely do storm event stream δ³⁰Si signatures vary across headwater catchments featuring distinct rock types, climate, and ecosystems?
2. Can storm events serve as proxies for long-term silicon isotope signatures?
3. What are the principal hydrological and (bio)geochemical controls on stream δ³⁰Si compositions?

In this pursuit, we focus principally on the analysis of these novel datasets between sites. Detailed analysis of Si–Q and δ³⁰Si–Q across each storm hydrograph presents a significant and compelling opportunity which will appear in subsequent publications. As a consequence, here we do not focus on variability of Si and δ³⁰Si within
individual sites over the course of a given storm event. As will be seen, the present cross-site analysis offers a substantial contribution within which future site-specific analysis may be contextualized.

2. Site Description

We present datasets across two continents (North America and Europe) and three countries (United States, France, and Canada). Of the six sites investigated, two are located on islands (Guadeloupe—a Caribbean Island and Calvert Island off the coast of British Columbia). The catchments are all small (0.08–17 km²) with first-to-third-order drainage networks spanning a range of climates (tropical to sub-humid), lithologies (volcanic, granitic, and shale), and ecological biomes (Caribbean tropical to coastal temperate rainforest to temperate mixed conifer forests). The data analyzed derive from a distinct period of high fluid infiltration and drainage within the respective catchment. In five of these catchments, we report rainfall-driven events and in one a seasonal snowmelt event (La Jara Creek). A summary of catchment characteristics is presented in Table 1 and Figure 1. A brief description of individual sites and the respective hydrological events that were sampled for this study (Table 2) are discussed in the following subsections.

2.1. La Jara Creek (35°50′N–36°00′N, 106°24′W–106°37′W)

La Jara Creek Catchment (basin area = 3.3 km²) is a part of the Jemez Critical Zone Observatory (National Research Council, 2001; T. White et al., 2015) located in the Valles Caldera National Preserve (VCNP) in the Jemez Mountains within northern New Mexico (USA). The creek is part of the larger East Fork of the Jemez River basin draining a 21 km wide caldera formed ~1.25 Ma with an elevation of 2664–2934 m ( McIntosh et al., 2017; Zapata-Rios et al., 2015). The climate is sub-humid with a bimodal annual precipitation pattern totaling 710 mm, where snowfall accounts for ~44% of the total precipitation ( McIntosh et al., 2017). The vegetation is characterized as a mixed conifer forest at higher elevations containing Spuce-fir ( Picea pungens ), ponderosa pine ( Pinus ponderosa ), aspen ( Populus tremuloides ), and gambel oak ( Quercus gambelii ) and forest meadows and grasslands in the valley ( Coop & Givnish, 2007; Muldavin et al., 2006).
Situated within a resurgent dome, the lithology of the La Jara Creek catchment (hereafter referred to as La Jara) is the most complex of all sites investigated in this study. In general, the lithology is largely dominated by Pleistocene aged Bandelier Tuff and fine-grained rhyodacite with primary mineral assemblages composed of Ca-clino-philite, alkali feldspar, cristobalite, albite, quartz and some minor accessory minerals such as apatite, hematite, zircon, titanite, ilmenite, and faujasite (Zapata-Rios et al., 2015). A deep regolith (42–27 m from ridge to base; Olyphant et al., 2016) and a relatively uniform (∼1 m thick) soil layer overlies the mixed rhyolitic-sandstone bedrock (Moravec et al., 2020).

La Jara is a perennial stream where streamflow derives principally from shallow and deep groundwater flow paths rather than overland flow (Bales et al., 2006; Henderson & Shuman, 2010; McIntosh et al., 2017; Zapata-Rios et al., 2015). The hydrology is typical for snowmelt driven systems, with peak discharge occurring in late March–early April. Mean annual runoff for the 2017 year was 1.25 mm day$^{-1}$ (Troch et al., 2021; Table 1). In this study, we analyzed a 2017 spring snowmelt event spanning 3 months from early March to Late May (Olshansky et al., 2018). Peak discharge during this event (0.15 m$^3$ s$^{-1}$) was roughly three times higher than the largest snowmelt event recorded over the previous 15 years (McIntosh et al., 2017).

### 2.2. Providence Creek, P301 (37°65'2"N, 119°20'55"W)

Located near Fresno, California (USA) in the Southern Sierra Nevada Mountains, Providence Creek is a second-order perennial stream draining the catchment designated P303 (basin area = 7.8 km$^2$) as part of the Kings River Experimental Watersheds (Hunsaker, 2013) within the Southern Sierra CZO network (National Research Council, 2001; White et al., 2015). This catchment resides at an elevation of 1731–2025 m, placing it within the rain–snow transition zone. The climate is Mediterranean with over 90% of the mean annual precipitation (1015 mm; O’Geen et al., 2018) taking place between October and April (Hunsaker & Johnson, 2017). Vegetation is predominantly mixed conifer forest (Bales et al., 2011; Hunsaker et al., 2012) dominated by white fir (Abies concolor) and ponderosa pine (Pinus ponderosa). Providence Creek (hereafter referred to as Providence)
### Table 2

#### Storm Event Hydrology Characteristics

<table>
<thead>
<tr>
<th>Site</th>
<th>Duration (MM/DD, YYYY)</th>
<th>Seasonal context (wet/dry)</th>
<th>$V_{\text{event}}$</th>
<th>$V_{\text{annual}}$</th>
<th>$F_{\text{event}}$</th>
<th>$Q_{\text{max}}$</th>
<th>$Q_{\text{avg}}$</th>
<th>$Q_{0.05}$</th>
<th>$Q_{0.50}$</th>
<th>$Q_{0.90}$</th>
<th>$q_{\text{avg}}$</th>
<th>$q_{\text{max}}$</th>
<th>$q_{\text{max}}^*$</th>
<th>(q_{\text{max}}/MAR)</th>
</tr>
</thead>
<tbody>
<tr>
<td>La Jara</td>
<td>3 months (03/13–05/18, 2017)</td>
<td>Snowmelt event</td>
<td>547,425</td>
<td>88,800</td>
<td>61.6</td>
<td>0.125</td>
<td>0.026</td>
<td>0.077</td>
<td>0.03</td>
<td>2.23</td>
<td>21.44</td>
<td>0.02</td>
<td>2.03</td>
<td>3.26</td>
</tr>
<tr>
<td>Providence</td>
<td>19 h (01/08–01/09, 2018)</td>
<td>Wet season</td>
<td>1451</td>
<td>167,783</td>
<td>0.86</td>
<td>0.35</td>
<td>0.0023</td>
<td>0.0071</td>
<td>1.8 x 10^{-4}</td>
<td>3.89 x 10^{-3}</td>
<td>0.052</td>
<td>0.01</td>
<td>1.34</td>
<td>2.22</td>
</tr>
<tr>
<td>Elder</td>
<td>7 days (01/05–01/12, 2017)</td>
<td>Wet season</td>
<td>4,358,012</td>
<td>30,515,599</td>
<td>14.3</td>
<td>33.70</td>
<td>1.34</td>
<td>6.31</td>
<td>0.51</td>
<td>11.44</td>
<td>45.75</td>
<td>7.94</td>
<td>37.2</td>
<td>163</td>
</tr>
<tr>
<td>Sapine</td>
<td>5 days (10/12–10/17, 2016)</td>
<td>Dry → wet</td>
<td>8516</td>
<td>326,934</td>
<td>2.60</td>
<td>0.054</td>
<td>4.26 x 10^{-4}</td>
<td>0.02</td>
<td>5.0 x 10^{-8}</td>
<td>1.0 x 10^{-5}</td>
<td>5.0 x 10^{-5}</td>
<td>0.01</td>
<td>3.23</td>
<td>8.35</td>
</tr>
<tr>
<td>Hakai (Watershed 708)</td>
<td>2 days (09/16–09/19, 2017)</td>
<td>Dry → wet</td>
<td>57,750</td>
<td>16,243,021</td>
<td>0.36</td>
<td>0.61</td>
<td>0.18</td>
<td>0.33</td>
<td>3.0 x 10^{-3}</td>
<td>0.28</td>
<td>1.74</td>
<td>0.13</td>
<td>3.68</td>
<td>6.76</td>
</tr>
<tr>
<td>Hakai (Watershed 708)</td>
<td>7 days (10/14–10/21, 2017)</td>
<td>Wet season</td>
<td>1,250,198</td>
<td>16,243,021</td>
<td>7.70</td>
<td>5.34</td>
<td>0.15</td>
<td>1.47</td>
<td>3.0 x 10^{-3}</td>
<td>0.28</td>
<td>1.74</td>
<td>1.08</td>
<td>16.2</td>
<td>59.2</td>
</tr>
<tr>
<td>Quiock</td>
<td>20 days (10/06–10/26, 2015)</td>
<td>&quot;Wet&quot; tropical storm season</td>
<td>1447</td>
<td>27,594</td>
<td>5.20</td>
<td>0.030</td>
<td>1.50 x 10^{-4}</td>
<td>0.0028</td>
<td>3.04 x 10^{-6}</td>
<td>1.20 x 10^{-3}</td>
<td>0.01</td>
<td>0.005</td>
<td>2.08</td>
<td>26.6</td>
</tr>
</tbody>
</table>

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*Estimated through trapezoid integration method on event and long-term stream discharge hydrographs. 
1Fraction of the total volume of water exported annually ($V_{\text{annual}}$) released during the storm event ($V_{\text{event}}$). 2Low-flow conditions for given year where flood event took place; $Q_{0.05} = $ low flow discharge exceeded 95% of the time. 3Median flow conditions for given year where flood event took place; $Q_{0.50} = $ median flow discharge exceeded 50% of the time. 4High-flow conditions for given year where flood event took place; $Q_{0.90} = $ high flow discharge exceeded 5% of the time. 5Average runoff for the given storm event calculated by normalizing the average discharge to the respective basin area (see Table 1). 6Maximum runoff of the storm event determined by normalizing maximum discharge to the basin area (see Table 1). 7Average runoff for the respective storm event ($q_{\text{avg}}$) normalized to the mean annual runoff (MAR). Normalization requires converting MAR to mm/day (see Table 1).
is underlain by granodiorite overlain by a relatively thick, partially weathered regolith (35 m for catchment P303; Holbrook et al., 2014) and thin soils (1.0–1.5 m).

Mean annual runoff is approximately 281 mm day⁻¹ (Hunsaker, 2019; Hunsaker et al., 2012; Table 1). Snow constitutes roughly 35%–60% of annual precipitation (Bales et al., 2018; Liu et al., 2013). Streamflow in Providence is largely supplied by deep (50%–75%) and shallow (~32%) subsurface flow with small contributions from overland flow (3%), mainly in the form of snowmelt (Bales et al., 2011). The storm event analyzed for this site took place in the winter of 2018 during the wet season. This event corresponded to a high precipitation period lasting 2 days (from Jan. 8–9, 2018).

2.3. Elder Creek (40°00'48"N, 122°33'00"W)

Elder Creek is a perennial, third-order stream that flows for 16 km within the Angelo Coast Range Reserve as part of the Eel River watershed and Critical Zone Observatory (ER-CZO, National Research Council, 2001; White et al., 2015) in Northern Mendocino County, California (US). Samples were taken from Elder Creek (hereafter referred to as Elder) next to the heavily monitored Rivendell site (Rempe & Dietrich, 2014; Salve et al., 2012), a thin, soil-mantled hillslope (average slope angle of 30°). Elder resides in a Mediterranean climate marked by warm, dry summers and cold, wet winters where average temperatures range from −8°C to 32°C. The mean annual precipitation is 1900 mm with rainfall concentrated in the months of October-April (Salve et al., 2012). The site is characterized by a densely vegetated coniferous forest (Hahm et al., 2017) of Douglas fir (Pseudotsuga menziesii) and Pacific madrone (Arbutus menziesii).

The lithology characterizing this site consists of tectonically folded mudstone dominated by turbidite sequences and interbedded lenses of sandstones and conglomerates as part of the Coastal Belt in the Franciscan Formation (McLaughlin et al., 2000). The bedrock primary mineral assemblages is a clay-rich shale where plagioclase, Fe-rich chlorite, illite, montmorillonite, smectite, and kaolinite are the main constituents (Gu et al., 2020; Kim et al., 2014). Between the shallow soil layer (0.5–0.75 m) and coherent bedrock lies a thick section of fractured, partially weathered regolith (ranging from 25 to 5 m deep from ridge to base of the hillslope; Salve et al., 2012). The deepest section of the weathered regolith is fully saturated with a water table that fluctuates by several meters in height throughout the year. Above the saturated zone the regolith is partially saturated, rich in reactive gases such as O₂ and CO₂ and hosts a deep root network (Tune et al., 2020).

Elder is supplied principally by subsurface drainage with no consistent observations of overland flow (Kim et al., 2014; Salve et al., 2012). Mean annual runoff for the year 2016–2017 was 4.78 mm day⁻¹ based on streamflow data from Dendra (formerly the Berkeley Sensor Database https://dendra.science/orgs/erczo; Table 1). The event reported in this study was an exceptional atmospheric storm that spanned a 10-day period in the middle of the wet season from Jan. 5–13, 2017, achieving runoff values 7.8 to ~34 greater than the mean annual runoff (Table 2).

2.4. Sapine Creek (44°37'N 3°82'W)

Located ~80 km from the Mediterranean Sea at an elevation between 1150 and 1350 m (Cognard-Plancq et al., 2001), Sapine Creek (hereafter referred to as Sapine) drains a small granitic catchment (basin area = 0.54 km²) of the southern flank of the Mont-Lozère and resides within the National Park of Cevennes in Southern France. This catchment has been continuously monitored since the 1980s by the Observatoire Hydrométéorologique Méditerranéen Cévennes-Vivarais (OHMCV), now part of the French Critical Zone research network, OZCAR (Observatoires de la Zone Critique: Applications et Recherche; Gaillardet et al., 2018).

The Mont-Lozère region experiences a Mediterranean climate with average temperatures between −5°C and 17°C and a mean annual rainfall of 2000 mm. Autumn and spring are characterized by periods of intense rain called “événements cévenol” that consist in flash flood events that can be particular dangerous to nearby villages located downstream. Beech (Fagus sylvatica) forest make up the vegetation cover of the catchment (Küßner, 2018). The lithology is composed of porphyritic granodiorite and monzogranite (part of the Pont-de-Montvert-Borne complex) where large (up to 10 cm) orthoclase megacrysts and fine-grained andesine, potassium feldspar, biotite, and quartz make up its overall primary mineral assemblage (Küßner, 2018). A thin, highly
permeable soil layer (~0.6 m Cognard-Plancq et al., 2001; Marc et al., 2001), overlies a relatively deep, heterogeneous regolith layer (10 m thick).

Sapine is a perennial, first order stream hydrologically characterized by quick responses in discharge and stage height to rainfall (Martin et al., 2003). Mean annual runoff for the 2015–2016 years was 3.07 mm day$^{-1}$ (Küßner, 2018; Table 1). Notably, no development of overland flow from infiltration excess is observed during storm events at this site (Martin et al., 2003). The storm event analyzed in this study was the first of significance following the dry season, occurring over a one-week period from Oct. 11–18 in 2016. Long-term sampling was also analyzed for the purpose of this study, taken on a roughly bi-monthly basis over a 2 year period from 7 July 2013 to 31 March 2014 ($n = 8$).

2.5. Hakai (51°39’12.9672″N, 128°7’48.4032″W)

One of seven watersheds comprising the Hakai Institute’s Kwakshua Watersheds Observatory recently integrated into the Critical Zone Exploration Network (CZEN, https://www.czen.org/), Kwakshua Watershed 708 is located in the northern section of Calvert Island off the central coast of British Columbia, Canada (Giesbrecht et al., 2021). Hereon referred to as “Hakai”, this third-order, perennial stream drains a basin area of 7.8 km$^2$. Elevation ranges from sea level to 385 m (inland). Calvert Island is located in a perhumid coastal temperate rainforest. The ocean-moderated climate generally consists of cool and mild winters with mean annual precipitation ~3228 mm yr$^{-1}$ and mean annual temperature of ~8.5°C. Wet and dry seasons are less pronounced than those found in the other catchments investigated in this study, but generally span from September through April and May through August, respectively (Oliver et al., 2017). Short statured bog forests and wetlands make up the main vegetation cover (Thompson et al., 2016). Western red cedar (Thuja plicata), yellow cedars (Chamaecyparis nootkatensis), western hemlock (Tsuga heterophylla), and shore pine (Pinus contorta) are the dominant trees. Several shrubs, herbs, sedges, and bryophytes (e.g., Sphagnum spp.), are also present (Levy-Booth et al., 2019).

A major feature of Kwakshua watersheds are that they represent a global “hotspot” for dissolved organic carbon yields to the ocean reaching 24.1 Mg C km$^{-2}$ in watershed 708 (Oliver et al., 2017). While yields of organic matter associated nutrients are high, inorganic nutrient yields are low enough (e.g., 59.6–72.6 kg Si(OH)$^+_2$ km$^{-2}$ yr$^{-1}$) to cause dilution in estuaries (St. Pierre et al., 2021).

Quartz diorite is the principal lithology on Calvert Island (Roddick, 1996), overlain by relatively thin (<1 m), organic-rich soils and a weakly developed regolith (Oliver et al., 2017). Mean annual runoff for Watershed 708 during the 2017 years was 5.71 mm day$^{-1}$ based on streamflow data available in the Hakai Metadata Catalog (Korver et al., 2020; Table 1). For this site, two storm events are reported which took place early in the wet season from September 16–19, 2017 and October 14–21, 2017, respectively.

2.6. Quiock Creek (16°17’N, 61°70’W)

Situated on Basse Terre Island as part of the Guadeloupe archipelago in the French West Indies, Quiock Creek is a small perennial, first-order tributary (basin area = 0.08 km$^2$) of the Bras-David River. This site is part of the ObsERA CZO (OBSevatoire de l’Eau et de l’éroSion aux Antilles (Dessert et al., 2015; Lloret et al., 2011); and of the French National CZO network OZCAR (Gaillardet et al., 2018). Quiock Creek (hereafter referred to as Quiock) sits at an elevation of 200–350 m and is characterized by a wet tropical climate composed of a dry season from January to June and a wet, cyclonic season from July to December. Mean annual air temperature and rainfall are 5°C and 3500 mm, respectively. Vegetation consists of a dense, humid tropical forest categorized by a variety of tree and plant species such as Gommier (Dacryodes excelsa), Ironwood (Sloanea spp.), Cordia reticulata, Amanoa caribaea, Tapura latifolia, and sedges (cyperaceae) lined with sharp silica edges (Imbert et al., 1998).

The main lithology is Pleistocene andesitic pyroclastic deposits (Boudon et al., 1989; Clergue et al., 2015) overlain by a thick, strongly depleted soil (>15 m; Colmet-Dauge & Bernard, 1979) profile with secondary minerals constituting roughly 95 wt.% of its bulk composition (Buss et al., 2010). The proximity of Quiock (~65 km) to the active Montserrat Volcano (located in the southern portion of Basse-Terre Island) and situated directly along the transatlantic path for Saharan dust plumes leads to significant input from volcanic ash and Saharan Dust, representing major sources of quartz and clays to the soil (Clergue et al., 2015; Dessert et al., 2020).
Hydrologically, Quiock is characterized by high evapotranspiration (60%–70%), a typical feature for tropical rainforests, and receives significant contributions from tropical storms and hurricanes, which can (as was the case for hurricane Rafael in October 2012) account for upwards of 5% of the total annual rainfall (Guérin et al., 2019; Zahibo et al., 2007). Mean annual runoff is 1130 mm (Fries et al., 2019; Guérin et al., 2019; Table 1). The event reported in this study took place during the wet season over a 3-week period from October 6–26, 2015 (Fries et al., 2019). The year 2015 was drier than the preceding years, with a mean annual throughfall input of ~2500 mm (Fries et al., 2019). The wet season for which the event took place was marked by major tropical storm, Erika, that touched down in late August and lead to extensive flooding and mudslides, causing major road damage on the Basse-Terre Island. This storm was followed by a category 1 hurricane Kate that struck Guadeloupe in early November, just after the sampling period in October.

3. Materials and Methods

3.1. Hydrometeorological Data

Annual and storm event rainfall intensity (mm h\(^{-1}\)) and stream discharge time series were collected from a range of meta-databases available via the institutions maintaining the respective CZO networks of each site. The French OZCAR network manages the Quiock (Guadeloupe) and Sapine (Mont-Lozère) sites, affording high-frequency hydro-meteorological data (5- and 30-min measurements, respectively) collected and distributed by the Observatoire de l’Eau et de l’Érosion aux Antilles (ObsERA: https://webobsera.ipgp.fr) and by the Observatoire Hydrométéorologique Méditerranéen Cévennes-Vivarais (OHMCV: https://ohmcv.osug.fr), respectively. The Hakai Institute provides 5-min stream discharge data through their Metadata Catalog (https://hectate.hakai.org/geonetworks/srv/fre/catalog.search#/home). For the U.S Critical Zone networks, high-frequency 15-min stream discharge and rainfall intensity data were obtained through the USGS National Water Information System (NWIS: https://waterdata.usgs.gov/nwis), CUAHSI Hydrologic Information System (HIS: https://data.cuahsi.org), and the respective CZO data repositories such as Dendra (formerly Berkeley Sensor Database) for the Eel River CZO (https://dendra.science/orgs/erczo) and the Climate and Hydrology Database Projects (CLIMDB/HYDROB: https://climhy.internet.edu) sponsored through the NSF-LTER (Long Term Ecological Research) network, USGS, and US Forest Service. Further information regarding data collection and methodology for the respective sites can be found through the references and links provided above.

3.2. Elemental Analyses

Non-acidified, filtered (0.22 µm) stream water samples were analyzed by three laboratories for dissolved Si, major and minor cation (Na, K, Ca, Mg, Al, Sr) and anion (Cl\(^-\), SO\(_4^{2-}\)) concentrations for all storm events. For La Jara, Hakai, and Providence creek chemical analyses were performed by the Arizona Laboratory for Emerging Contaminants (ALEC) using an Agilent 7700 quadrupole inductively coupled plasma spectrometer (Q-ICP-MS) at the University of Arizona, Tuscon, AZ (USA). Analytical QA/QC protocols were adapted from US EPA Method 200.8. Calibration standards were prepared from multi-element stock solutions (SPEX Certiprep, Metuchen, NJ). Calibration curves include at least seven data points with correlation coefficients >0.995. Quality control sample solutions are from an independent source, NIST SRM 1643e—Trace elements in water, or from High Purity Standards (Charleston, S.C.). Acceptable QC responses must be between 90% and 110% of the certified value (precision of ±10%).

Sapine and Quiock, anion concentrations were determined through ion chromatography analyses at both ALEC and IPGP (Institut de Physique du Globe de Paris) using a Dionex ICS-6000 and Dionex DX120, respectively. Quiock major cation and dissolved silicon concentrations were measured by inductively coupled plasma-source mass spectrometry (ICP-MS; Thermo X-Series II) at the National Oceanographic Centre Southampton (NOCS). Synthetic multi-element standards prepared gravimetrically from high purity single element standard solutions were used for calibration. Certified reference material SLR-6 (river water) and IAPSO seawater were analyzed during the sample runs multiple times (n = 8) to determine the analytical precision (±5% of certified values). Sapine major cation and dissolved silicon analyses were performed at IPGP by ICPMS (Agilent 7900 Q-ICP-MS) and UV-VIS spectrophotometry (Dionex UVD170U), respectively. For both analyses, instrumental precision (±5%) was determined and accuracy was checked using the certified river water reference standard SLRS-6.
3.3. Germanium Analyses

For Elder, Sapine, and Hakai storm events and the La Jara snowmelt, germanium (Ge) concentrations were analyzed at Cornell University using hydride-generation ICP mass spectrometry (HG-ICP-MS; Aguirre et al., 2017; Mortlock & Froelich, 1996) that included a custom HG sample introduction system, and a Thermo-Finnigan Element two ICP-MS. Natural water samples were prepared using an isotope dilution method where enriched \(^{70}\text{Ge}\) spike solution is first added and then allowed to equilibrate for at least 48 hr at room temperature. Samples were spiked to a target \(^{70}\text{Ge}/^{74}\text{Ge}\) ratio of 10, inferred from dissolved Si concentrations. For Ge concentration measurements, prepared samples were introduced into the hydride generation system along with a 4% NaBH\(_4\) solution to reduce aequous Ge(OH)\(_4\) to volatile GeH\(_4\). This volatile Ge phase is then passed through a gas-liquid-separation filter along with Ar as a carrier gas into the ICP-MS. Isotope dilution was used to quantify germanium using the \(^{70}\text{Ge}/^{74}\text{Ge}\) ratio. Mass bias and signal drift were corrected via standard sample bracketing. Ge standards were measured at 5, 20, 50, 100, and 200 ng L\(^{-1}\). Isotope dilution calculations were cross-checked using standard response curves generated at \(m/z\ 74\) for each Ge analysis. The contribution from the spike to \(m/z\ 74\), with \(^{70}\text{Ge}/^{74}\text{Ge}\) equal to 161.4, at \(m/z\ 74\) is insignificant (≈0.1%). Analytical uncertainty on Ge concentration measurements at the Cornell University facilities is approximately 3% RSD (Aguirre et al., 2017; Derry et al., 2005; Kurtz et al., 2002, 2011; Lugolobi et al., 2010).

3.4. Silicon Stable Isotope Analyses

Stream samples were prepared for Si stable isotope analyses using a column chemistry protocol adapted from Van Den Boorn et al. (2006), Georg et al. (2006b), and Pringle et al. (2014). Si purification was conducted by loading 10–20 µg dissolved Si onto a pre-cleaned 1.8 mL BioRad AG50W-X12 cation-exchange resin bed in BioRad columns. Si was eluted by addition of 5 mL MQ-e water for a target concentration of 1–2 ppm (~9–18 V). Approximately 70% of silicon is recuperated in the first elution step (2 mL MQ-e) with the remaining 30% recovered in the final elution step (3 mL MQ-e). The parent material for Elder Creek was also analyzed for the purpose of this study and was prepared using a modified alkali fusion technique developed by Georg et al. (2006a) described in detail by Fernandez et al. (2019).

The chemistry protocol outlined above involving a cation-exchange resin provides excellent Si yield, but fails to remove anionic species. Iso-baric anion matrix effects (i.e., \(\text{SO}_4^{2-}\), DOC, \(\text{NO}_3^-\), \(\text{PO}_4^{3-}\)) are known to introduce instrument mass bias in silicon isotope measurements of natural water samples (Dauphas et al., 2009; Hughes et al., 2011; Oelze et al., 2016). Rigorous evaluation of anion residual mass effects conducted by Oelze et al. (2016) places the [anion]/[Si] molar ratio limit at 1, above which notable mass bias would be observed in Si isotope measurements. For the stream samples measured in this study all [anion]/[Si] ratios are <1 with an average of 0.1 for [SO\(_4^{2-}\)]/[Si]. A notable exception are the Hakai samples where dissolved organic carbon content regularly exceeds 10 ppm. To remove this potentially significant mass bias, Hakai samples were subjected to alkali fusion following an adapted version of the protocol developed for the preparation of solid material for silicon isotope analyses (Georg et al., 2006a). The procedure involves an initial evaporation step of pre-weighed fluid consisting of a target 10–40 µg of Si that is dried down on a hot plate at 110°C in 15 mL Teflon vials until a target volume of ~1 mL is reached. Samples were then transferred with a pipette to cleaned Ag (silver) crucibles and dried down completely at 110°C. Half a tablet (i.e., 50 mg) of analytical grade NaOH was added to the Ag crucibles and subsequently fluxed at 720°C for 12 min in a muffle furnace. The fused product was then allowed to dissolve in PTFE vials containing 5 mL of ultrapure water for a target dissolved Si concentration of 2 ppm Si and then acidified with ultrapure HNO\(_3\) to reach 2% HNO\(_3\) concentration.

All silicon stable isotope analyses were performed on a Thermo Finnigan Neptune Plus MC-ICP-MS at the Institut de Physique du Globe de Paris. Sample introduction was in the form of a wet plasma using a SIS (Standard Introduction System), double Scott/cyclonic spray chamber and a 100 µL min\(^{-1}\) Teflon\(^{®}\) nebulizer. Measurements were run at medium resolution mode with a resolving power M/ΔM ~ 3300 to reduce polyatomic interferences (notably the NO\(^+\) interference on the \(^{30}\text{Si}\) signal). To correct for magnet drift, a high-resolution peak centering was performed every hour to maintain a stable signal. Stable isotopic compositions are reported in standard per mil notation (\(\delta^{30}\text{Si}, \delta^{29}\text{Si}\)) based on blank-corrected ratios calculated through a standard bracketing method (Equation 1) using the silica sand internal standard (NBS-28, RM 8546):
\[
\delta^{30}\text{Si} = \left[ \frac{\left(^{30}\text{Si} / ^{28}\text{Si}\right)_{\text{sample}}}{\text{AVG } \left(^{30}\text{Si} / ^{28}\text{Si}\right)_{\text{std-I}}, \left(^{30}\text{Si} / ^{28}\text{Si}\right)_{\text{std-II}}} - 1 \right] \times 1000 \tag{1a}
\]

\[
\delta^{29}\text{Si} = \left[ \frac{\left(^{29}\text{Si} / ^{28}\text{Si}\right)_{\text{sample}}}{\text{AVG } \left(^{29}\text{Si} / ^{28}\text{Si}\right)_{\text{std-I}}, \left(^{29}\text{Si} / ^{28}\text{Si}\right)_{\text{std-II}}} - 1 \right] \times 1000 \tag{1b}
\]

where std-I and std-II correspond to the NBS-28 standard measured directly before and after the sample measurement. Repeated evaluation \((n = 400)\) of the certified basalt standard BHVO-2 resulted in an appropriate value with long term precision of \(-0.31 \pm 0.12\%\) (2SD). These results are consistent with previous measurements conducted at IPGP facilities \((-0.28 \pm 0.09\%\) (2SD), Pringle et al., 2014; Savage et al., 2013\) and with reported values from the literature \((-0.30 \pm 0.14\%\) \((n = 133)\), GeoReM Database: Jochum et al., 2016; Oelze et al., 2016\).

A three-isotope plot confirms mass-dependent fractionation for the samples analyzed in this study (Figure S1; \(\delta^{29}\text{Si} = 0.54 \times \delta^{30}\text{Si}, R^2 = 0.98\), such that in the following only \(\delta^{30}\text{Si}\) values are reported and discussed with a precision of 2SD derived from repeated measurements of individual samples \((n = 4)\).

### 4. Results

#### 4.1. Storm Event Characteristics and Relative Intensities

A majority of the storm events took place within and at the start of the wet season with the exception of La Jara, which was snowmelt driven. Mean runoff for all events was higher than the corresponding mean annual runoff \((\text{MAR; Table 1})\) for these streams. Three storm events \((\text{La Jara, Elder, and Quiock})\) displayed a maximum discharge higher than the 5% exceedence criterion defined from long-term discharge data \((\text{i.e., flow equaled or exceeded only 5% of the time (Table 2)})\). The storm event analyzed at Elder was the most extreme with regards to the historical record, corresponding to 14% of the total volume of water released during the year of sampling \((\text{Table 2})\). Mean and maximum discharge values for this event were \(\sim 7.8\) and 34 times greater than mean annual flow, respectively. The first Hakai event was the most muted with the mean and maximum runoff barely exceeding the average annual runoff. The cumulative volume of water released during all sampled storm events represents on average 13% of the total annual export volume for the given catchments, with La Jara's snowmelt event being the largest \((62\%)\) and Hakai's first storm event \((0.36\%)\) being the smallest.

#### 4.2. Dissolved Silicon, \(\delta^{30}\text{Si}\), and Ge/Si Event Signatures

Dissolved Si concentrations and associated \(\delta^{30}\text{Si}\) for all six catchments \((n = 155)\) are summarized in Table 3 and illustrated in Figure 2 \((\text{see Supplementary for full dataset})\). Available long-term stream Si data for Sapine \((\text{this study})\), Elder \((\text{USGS NWIS, 2016; station 11475560})\), and La Jara \((\text{McIntosh et al., 2020})\) along with published data for Quiock \((\text{Gaspard et al., 2021})\) and Providence \((\text{Frings et al., 2021})\) are presented alongside the events for comparison. Stream Si concentrations \((\text{Figure 2a})\) during these high discharge events span roughly two orders of magnitude across sites from a minimum of 25 \(\mu\text{mol L}^{-1}\) measured at Hakai to a maximum of 1035 \(\mu\text{mol L}^{-1}\) measured at La Jara. Sapine and Hakai exhibited the tightest range in dissolved Si concentrations, reflecting a change of \(\sim 30\) and 22 \(\mu\text{mol L}^{-1}\) throughout the course of the storm events, respectively. La Jara and Providence displayed the largest variability in dissolved Si concentrations with value ranges of \(\pm 747\) and \(\pm 206 \mu\text{mol L}^{-1}\), respectively. Measured stream \(\delta^{30}\text{Si}\) across all catchments \((\text{Figure 2b})\) shows a mean value of \(+0.71 \pm 0.54\%\) where Quiock holds the lowest measured values \((-0.22 \pm 0.12\%)\) and Providence the highest measured values \(+2.02 \pm 0.12\%). All stream \(\delta^{30}\text{Si}\) are elevated with respect to an average silicate bedrock value \((-0.28 \pm 0.15\%, \text{Savage et al., 2013})\). Within individual storm events, Quiock expressed the widest range in stream \(\delta^{30}\text{Si} (-0.22 \pm 0.21\%\) to +0.45 \pm 0.14\%) whereas Hakai Event 1 exhibited the smallest variations \((+0.38 \pm 0.09\%\) to +0.60 \pm 0.13\%). Figure 2 displays the sites ordered along a general climate gradient from sub-humid \((\text{La Jara})\) to wet, tropical \((\text{Quiock})\), illustrating a general trend toward more dilute stream dissolved Si and less overall variability. In contrast, stream \(\delta^{30}\text{Si}\) lack any coherent trend along this climate gradient.

The La Jara, Providence, Elder, and Sapine storm events are found to mimic long-term variability in dissolved Si concentrations of these watersheds. Of these three sites La Jara exhibits the greatest variability \((288–1035 \mu\text{mol L}^{-1})\)
<table>
<thead>
<tr>
<th>Dissolved Si (µmol/L)</th>
<th>La Jara</th>
<th>Providence</th>
<th>Elder</th>
<th>Sapine</th>
<th>Event 1</th>
<th>Event 2</th>
<th>Both Events</th>
<th>Quiuck</th>
<th>All sites</th>
</tr>
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<tbody>
<tr>
<td>μ</td>
<td>476.5</td>
<td>458.4</td>
<td>327.9</td>
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<td>24.7</td>
<td>26.2</td>
<td>24.7</td>
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<tr>
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<td>38.9</td>
<td>38.9</td>
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<tr>
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<th>Elder</th>
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<th>Event 1</th>
<th>Event 2</th>
<th>Both Events</th>
<th>Quiuck</th>
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<th>Providence</th>
<th>Elder</th>
<th>Sapine</th>
<th>Event 1</th>
<th>Event 2</th>
<th>Both Events</th>
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<th>All sites</th>
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<th>Δ²⁸Si diss (bedrock corrected δ²⁸Si) [‰]</th>
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<th>Providence</th>
<th>Elder</th>
<th>Sapine</th>
<th>Event 1</th>
<th>Event 2</th>
<th>Both Events</th>
<th>Quiuck</th>
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<tr>
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<tr>
<td>Median (50%)</td>
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<td>0.68</td>
<td>0.70</td>
<td>0.45</td>
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with respect to its long-term record (146–565 µmol L$^{-1}$, 2010–2013, McIntosh et al., 2017), which may be related to the unique characteristics of this spring snowmelt event. Elder, on the other hand, exhibits slightly less dissolved Si variability at the event-scale (260–364 µmol L$^{-1}$) relative to its long-term data (184–300 µmol L$^{-1}$, USGS NWIS, 2016; station 11475560). Event based $\delta^{30}$Si data for Providence and Quiock are consistent with published data (Frings et al., 2021; Gaspard et al., 2021). For Sapine, stream $\delta^{30}$Si variability from the event (+0.02‰ to +0.55‰) is greater than observed long-term values (+0.17 ± 0.05‰ to +0.30 ± 0.06‰, this study, Figure 2b). General agreement between event and historical stream dSi data indicates that storm events are informative proxies of long-term Si weathering signatures.

Four of the five Ge/Si datasets reported here were measured on the same fluid samples for the same storm events. These include Sapine, Elder, La Jara, and Hakai (Table S1). The La Jara snowmelt event displayed low fluid Ge/Si ratios ranging from 0.07 (minimum) to 0.15 (maximum) µmol mol$^{-1}$ and an average corresponding to 0.11 µmol mol$^{-1}$ (n = 23). The Elder event fluid Ge/Si ratios remained relatively constant throughout the storm event with an average value of 0.28 ± 0.03 µmol mol$^{-1}$ (n = 13). Average fluid Ge/Si values for the Sapine and Hakai events correspond to 0.81 ± 0.25 µmol mol$^{-1}$ (n = 30) and 1.63 ± 0.37 µmol mol$^{-1}$ (n = 63, both events), respectively. A fifth dataset is also included for Providence based on 39 samples collected over approximately one year (Aguirre, 2019), yielding a flux-averaged fluid Ge/Si of 0.06 ± 0.02 µmol mol$^{-1}$. These samples are clearly distinct from the storm events reported thus far, but we show below that (a) these catchments appear to cluster closely among themselves (see Section 4.3.2) and (b) storm events display close agreement with longer-term datasets (Figure 2).

Discharge or q-weighted averages are calculated and presented in the subsequent figures (Figure 3 onwards) alongside the respective datasets to facilitate comparison across sites and storm events (e.g., Pokrovsky et al., 2013; Prokushkin et al., 2011). These discharge-weighted averages adjust the raw mean values to reflect their relative contributions to the total mass discharged during the event (Equation 2; Table 3):

$$[\text{DSi}] = \frac{\sum q_t \cdot [\text{DSi}]_t}{\sum q_t}$$

$$\bar{\delta^{30}Si} = \frac{\sum (q_t \cdot [\text{DSi}]_t \cdot \delta^{30}Si)_t}{\sum q_t \cdot [\text{DSi}]_t}$$

where $q_t$ is the measured stream runoff as a function of time (t) throughout the event duration, [DSi] and $\bar{\delta^{30}Si}$ represent the discharge-weighted mean for dissolved Si concentrations and stream Si stable isotope ratios, respectively. It is important to note that $\bar{\delta^{30}Si}$ includes an additional dependence on the measured dissolved Si ([DSi]) as well. This approach essentially accounts for the relative intensities in the solute fluxes of the respective events. The overall discharge-weighted mean of stream $\delta^{30}$Si values across all catchments is 0.70 ± 0.61‰. Across storm events, La Jara exhibits the highest [DSi] (464 µmol L$^{-1}$) and Hakai shows the lowest (33.8 µmol L$^{-1}$). Compar-

<table>
<thead>
<tr>
<th>1st quartile (25%)$^a$</th>
<th>3rd quartile (75%)$^b$</th>
<th>At peak Q$^c$</th>
<th>At min Q$^d$</th>
<th>Q-weighted avg.</th>
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<tr>
<td>0.62</td>
<td>1.01</td>
<td>-</td>
<td>-</td>
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</tr>
</tbody>
</table>

$^a$First quartile indicates 25% of the observations lie below this value. $^b$Third quartile indicates that 75% of the of the observations lie below this value. $^c$Value of DSi (µM) or $\delta^{30}$Si (‰) at maximum discharge during either the storm event at the respective catchment or across all sites (noted here as "all sites"). $^d$Value of DSi (µM) or $\delta^{30}$Si (‰) at minimum discharge during either the storm event at the respective catchment or across all sites (noted here as All sites). $^e$Dissolved silicon concentration discharge-weighted average calculated as defined in Equation 2a. Note all $\delta^{30}$Si values report a precision of 2SD. $^f$Dissolved Si stable isotope discharge-weighted average calculated as defined in Equation 2b. $^g$Fraction of Si remaining in dissolved phase; mainland catchments calculated using Equation 4 with bedrock Si/Na ratios equal to 17.96 (Sapine; Küßner, 2018), 17.98 (Elder; Gu et al., 2020), 19.94 (Providence Creek; Hahn et al., 2014), and 12.38 (La Jara; Supplemental Materials, Section S.3). For Hakai we used Si/Ca ratios and normalization to bedrock following equivalent computation to Equation 4 based on bedrock Si/Ca value of 7.68 (Roddick, 1983, 1996). Quiock was assigned a value of 1.0 (see Section 4.3.2 for further details).
ing individual measurements and discharge-weighted averages of stream $\delta^{30}$Si against dissolved Si concentrations for all events (Figure 3) yields a weak positive correlation ($R^2 = 0.46$) bounded by a low dissolved Si ($<200$ µmol L$^{-1}$), low $\delta^{30}$Si ($<+0.5\%$) end member describing Hakai, Quiock, and Sapine events and an elevated dissolved Si and $\delta^{30}$Si end member consistent with Providence. With the exception of La Jara, which has several notable outliers, data points for individual events tend to cluster around their respective discharge-weighted averages with little overlap across sites. In the case of Hakai where two consecutive storm events were measured for the same

![Graph showing dissolved Si concentrations and $\delta^{30}$Si values for different catchments.](image1)

**Figure 2.** Box-and-whisker plots of stream dissolved Si concentrations (a) and $\delta^{30}$Si (b) for each given site organized from left to right according to their respective climates from sub-humid (La Jara) to wet, tropical (Quiock). Box lengths represent the overall variability in dissolved Si and $\delta^{30}$Si where the mean is depicted as a solid line within the box interior. Outliers are depicted as solid black diamonds. Published Si data (yellow) is compared against storm events (blue) for sites where such datasets are available La Jara (McIntosh et al., 2020), Providence (Frings et al., 2021), Elder (USGS NWIS, 2016; station 11475560), and Quiock (Gaspard et al., 2021). Sapine is an exception in this comparison and reflects long-term Si data measured for the purpose of this study.

![Graph showing stream $\delta^{30}$Si plotted against dissolved Si concentrations for all sites.](image2)

**Figure 3.** Stream $\delta^{30}$Si plotted against dissolved Si concentrations for all sites. Individual measurements are shown as shaded circles whereas discharge-weighted averages (Equation 2) are represented by solid diamonds. Marine catchments (Quiock and Hakai) tend to plot at the lower $\delta^{30}$Si and dissolved Si domain. Mainland catchments besides Sapine have higher $\delta^{30}$Si and dissolved Si with Providence representing the upper bound. No definitive cross-site relationships are found between dissolved Si and associated $\delta^{30}$Si values for the respective storm events. Error bars correspond to 2SD.
year, stream $\Delta$Si and $\Delta^{30}$Si fall in the same range, together generating overall flux-weighted [DSi] and $\delta^{30}$Si signatures for Hakai of $34 \pm 5 \mu$mol L$^{-1}$ and $+0.42 \pm 0.10\%e$, respectively.

Stream dissolved Si and $\delta^{30}$Si show no systematic cross-site trend as a function of discharge (Figure 4). In Si–Q space, dissolved Si concentrations appear to remain relatively constant for a given site despite as much as two orders of magnitude variations in discharge (Figure 4a). There is no apparent relationship between Si and storm discharge. For example, Elder illustrates one of the highest discharges, yet produces one of the largest Si concentrations. In $\delta^{30}$Si-Q space, we note a similar lack of correlation. Further, the lowest storm discharge is associated with the largest $\delta^{30}$Si.

4.3. Data Adjustment

Two corrections were applied to our storm event dataset prior to further analysis. These were made to address the influence of two external factors: (1) marine aerosols accompanying rainwater and (2) catchment lithology. Here we outline the respective data treatment methods implemented in this study and provide context relevant to the following discussion.

4.3.1. Atmospheric Corrections

Atmospherically-derived marine (sea salt) inputs can be a major source of dissolved cations to catchments situated close to oceans and seas (Chadwick et al., 1999; Derry & Chadwick, 2007; Stallard & Edmond, 1981). Four sites in our compilation are located in such areas. Hakai and Quiock are island catchments in the Pacific Ocean and Caribbean Sea, respectively, while Sapine and Elder reside ∼80 km from the Mediterranean Sea in Southeastern France and ∼15–20 km from the Pacific Ocean in Northern California, respectively. To remove potential rainwater aerosol contributions, a commonly employed method was implemented where stream Cl$^-$ concentrations serve as a metric for atmospheric input (Equation 3; Gaillardet et al., 1999):

$$[X]^* = [X]_{\text{meas}} - \left( \frac{X}{\text{Cl}^-} \right)_{\text{rain}} \cdot [\text{Cl}^-]_{\text{meas}}$$

Figure 4. Dissolved Si and stream $\delta^{30}$Si as a function of storm average runoff for each storm event. Shaded circles and diamonds correspond to individual sampling points and discharge-weighted averages (Equation 2), respectively. Across all sites, dissolved Si concentrations exhibit little sensitivity to runoff despite considerable variability in the latter, on the order of ∼2 orders of magnitude (a). Silicon stable isotopes appear to capture intra-site variability, as shown for Elder with $\delta^{30}$Si trending toward lower values with increasing runoff (b). For both dissolved Si and $\delta^{30}$Si, across-site trends as a function of runoff are negligible. Error bars correspond to 2SD.
Here elemental seawater ratios (\(\langle X/Cl\rangle_{\text{wat}}\); Gaillardet et al., 1999; Riley and Tongudai, 1967; Stallard 1980) are used alongside stream Cl\(^-\) concentrations to infer the atmospheric contribution to measured stream water concentrations (where \(X\) corresponds to major cations Na\(^+\), Mg\(^{2+}\), K\(^+\), Ca\(^{2+}\) and anion SO\(_4^{2-}\)). For island sites Hakai and Quiock, routine observations of Ca/Cl rainwater ratios exceeding that of seawater (Keresztesi et al., 2020; Derry & Chadwick, 2007; NADP (NRSP-3), 2020; Vet et al., 2014) are accommodated in further calculations (Section 4.3.2) by implementing a Ca/Cl ratio 1.5 times that of seawater. No such correction is required for Si and Ge, given the scarcity of these elements in wet deposition. Application of this method to stream dissolved cation concentrations would be compromised in the event that Ca-rich evaporites are present in the catchment, or in the case of hydrothermal inputs. However, prior geochemical and mineralogical characterization shows no such evidence for the sites included in this study (Buss et al., 2010; Frings et al., 2021; Gu et al., 2020; Küßner, 2018; Liu et al., 2008; Uhlig et al., 2017; Vázquez-Ortega et al., 2015; Zapata-Rios et al., 2015). Correcting major cation stream chemistry for marine aerosol input accounts for most of the dissolved cation flux for the island catchments, Quiock and Hakai, with Na and Mg the most heavily impacted (~86%–100%; Table S1). In these settings, Ca appears to be the least impacted by atmospheric contributions accounting for 48% (Quiock) and 20% (Hakai) of the stream flux. Because of the extreme effect on Na in these settings, an alternative parent lithology correction was implemented for the island catchments (discussed further below, Section 4.3.2). Sapine is similarly impacted by aerosol input, although to a lesser extent. For Na and Mg, atmospheric contributions constituted on average 61% and 30% of the total dissolved stream budget, respectively. Despite being situated relatively close to the Northwestern California coastline, Elder is more comparable with the remaining sites further inland (La Jara, Providence), displaying only minor effects from atmospheric inputs (1%–26%).

Silicon is generally not included in these corrections due to its naturally low abundance in seawater (58 ± 49 \(\mu\)M, De La Rocha et al., 2000) compared to the major ions (Na\(^+\), Ca\(^{2+}\), Mg\(^{2+}\), K\(^+\), Cl\(^-\), SO\(_4^{2-}\)). Nevertheless, this does not exclude the possibility of atmospheric contributions to dissolved silicon budgets in catchments. Rather, other potentially prevalent atmospheric sources through dry deposition may be considered, such as North African (Engelstaedter et al., 2006) or Asian (Kurtz et al., 2001) dust inputs, which can be significant. For example, Quiock receives total dust fluxes of 100 kg ha\(^{-1}\) yr\(^{-1}\) owing to its position along the westward trajectory of the Saharan dust plume (Clergue et al., 2015).

Geochemical tools such as radiogenic Sr isotopes (\(^{87}\)Sr/\(^{86}\)Sr) have been used frequently to constrain dust inputs (Dessert et al., 2020; Dia et al., 2006; Kennedy et al., 1998; Pett-Ridge, 2009; Stewart et al., 2001; Van der Hoven & Quade, 2002). However, applying dust corrections to stream chemistry presents a unique set of challenges due to limited information on dust dissolution kinetics and spatiotemporal variability in dust deposition and accumulation within the critical zone. From the handful of studies available on this subject, laboratory-based dissolution experiments suggest that dust contributions to the dissolved load can range between 1% and 10% depending on particle size, pH, and chemical affinity (Guerzoni et al., 1999). Without sufficient information to properly correct for dust inputs, we proceed under the assumption that this external pathway exerts a minor effect on our dissolved Si data.

4.3.2. Normalization of Stream Si Abundance and Parent Lithology Corrections

Dissolved Si concentrations alone do not represent relevant information in terms of Si release to and scavenging from solution as these raw concentrations are also influenced by dilution and/or evaporative effects. For this reason, the concentration of a solute of interest is often normalized to that of another element presenting conservative behavior, that is, one that is insensitive to processes that scavenge solutes from solutions. Here we use Na as a conservative element (Gaillardet et al., 1999) and examine stream dissolved Si/Na ratios to inform Si mass transfers.

Stream dissolved Si/Na ratios are also strongly influenced by source rock composition, which can obscure the effects of other relevant weathering parameters such as temperature, precipitation, and runoff (Edmond et al., 1995; Gaillardet et al., 1999; Huh et al., 1998). Application of a correction for the mostly mono-lithological small catchments presented in this work permits us to evaluate source rock controls across sites and to distinguish these differences from those caused by hydrology (runoff, fluid transit times) and reactivity (fractionating biogeochemical processes). Consequently, we further correct stream dissolved Si/Na molar ratios from bedrock effects by normalizing them to the Si/Na molar ratio of the corresponding bedrock, leading to the definition of \(f_{\text{Si}}\) (Equation 4, Brichau et al., 2008; Georg et al., 2006a; Gíslason et al., 1996; Stallard, 1980):

\[
f_{\text{Si}} = \frac{(\text{Si}/\text{Na})_{\text{stream}}}{(\text{Si}/\text{Na})_{\text{bedrock}}}
\] (4)
where Na* refers to Na concentrations corrected for atmospheric contributions (Section 4.3.1). Following approaches from prior work (Hughes et al., 2013), a quartz correction was performed on bedrock Si concentrations to account for the non-stoichiometric dissolution of the primary mineral assemblage where Si is preferentially released from more reactive plagioclase and feldspars whereas Na is relatively inert. A full description of the bedrock compositions used for each site and the quartz correction procedure is provided in Supplemental Materials (Table S2; Equation S.1). An $f_{\text{diss}}^{\text{Si}}$ value of one indicates congruent release of Si and Na from rock to streams. Conversely, an $f_{\text{diss}}^{\text{Si}} < 1$ indicates that dissolved Si has been scavenged by secondary processes, such as clay formation (Ziegler, Chadwick, White, et al., 2005) or biologic uptake (Opfergelt et al., 2010); $f_{\text{diss}}^{\text{Si}}$ values above 1 might indicate unaccounted-for Si supply by external sources, like dust, anthropogenic effects, or hydrothermal input. As explained in Section 4.3.1, correction for atmospheric sources of Na at the two island sites of Hakai and Quiock leads to Na* values very close to 0, thereby resulting in large uncertainties on $f_{\text{diss}}^{\text{Si}}$ at these sites when using Equation 4. As a consequence, at these sites we estimate $f_{\text{diss}}^{\text{Si}}$ based on atmospheric input-corrected Ca rather than Na. Though Na is generally preferred, we note in Section 4.3.1 that Ca appeared to be less impacted by atmospheric inputs at these two sites.

The quartz-corrected (Si/Na)$_{\text{bedrock-qtz}}$ ratio for each of the sites was determined based on the extent of available information for each respective bedrock chemical composition. A value of 17.98 was applied for Elder (Gu et al., 2020), 14.94 for Providence (Hahn et al., 2014), 1844 for Quiock (Buss et al., 2010), and 12.38 for La Jara (see Supplementary Material, Table S2; Broxton et al., 2007; Wilcock et al., 2013). A value of 17.96 was used for Sapine (Künnner, 2018). For Hakai, the quartz-corrected (Si/Ca)$_{\text{bedrock-qtz}}$ ratio is 7.68 (Roddick, 1983, 1996). We note that in any given catchment, heterogeneity of bedrock mineralogical composition could produce a wide range of $f_{\text{diss}}^{\text{Si}}$ values, which we have encapsulated by using an average bedrock Si/Na in Equation 4. Fortunately, the drainage networks feeding these streams are already averaging over lithological heterogeneity across the catchment. Hence, the temporal variability in $f_{\text{diss}}^{\text{Si}}$ observed over the course of a storm event may (in part) reflect the diversity of lithological contributions, such that this variability can be used as an estimate of uncertainty in $f_{\text{diss}}^{\text{Si}}$.

This normalization of our storm events generated $f_{\text{diss}}^{\text{Si}}$ values that were all substantially less than 1 in the four non-island catchments, indicating that approximately 46% (La Jara) to 82% (Elder) of Si solubilized from bedrock is missing from the dissolved load. For Hakai, the alternative definition for $f_{\text{diss}}^{\text{Si}}$ produces an average value of 0.90 and appears to reflect a reasonable representation of this organic-rich, high-weathering intensity system in which the majority of solubilized Si is expected to be exported in solution. Based on the same modified version of Equation 4 using Ca instead of Na, Quiock returns values consistently >1.0, with an average of 1.9 using a seawater Ca/Cl ratio in Equation 4, and even higher values if we use 1.5× the seawater ratio. We attribute these unrealistically high $f_{\text{diss}}^{\text{Si}}$ values to bias inherent to the method of calculation (e.g., rock source composition, possible incorporation of Ca into the biomass), and take this estimation as confirmation that this system is exporting the entirety of dissolved Si released by silicate weathering. We resolve the issue by only reporting the flux-average values for Quiock in the remainder of our analyses using an $f_{\text{diss}}^{\text{Si}} = 1.0$ as a representative of congruent export of dissolved Si. Furthermore, we note that these $f_{\text{diss}}^{\text{Si}}$ values above 1, although possibly deriving from uncertainty inherent to our approach, can also to some extent reflect the actual release of Si from critical zone compartments (regolith and/or biota) during the sampled storm events. Regardless, the principal observation that can be drawn from our $f_{\text{diss}}^{\text{Si}}$ estimates at these two island sites is a near-congruent release of dissolved Si at the catchment scale.

Stream Si isotopes ratios are also partly controlled by source rock composition, and similarly require correction leading to the parameter $\Delta^{30}\text{Si}_{\text{diss-rock}} = \delta^{30}\text{Si}_{\text{diss}} - \delta^{30}\text{Si}_{\text{bedrock}}$. A value of $\Delta^{30}\text{Si}_{\text{diss-rock}} = 0$ suggests isotopically congruent primary mineral weathering as a primary control on stream $\delta^{30}\text{Si}$ while $\Delta^{30}\text{Si}_{\text{diss-rock}} > 0$ implies enrichment most likely driven by fractionating pathways such as secondary weathering or biological cycling. Granitoids and Upper Continental Crust (UCC) $\delta^{30}\text{Si}$ values systematically lie within a tight range (Savage et al., 2013), implying that primary mineral compositions are relatively uniform. Thus, bulk bedrock values are good approximations for the $\delta^{30}\text{Si}$ of congruently dissolving minerals. Local granitoid bedrock $\delta^{30}\text{Si}$ values were approximated by the average granitic ($-0.23 \pm 0.15\%$, Savage et al., 2012) and rhyolitic ($-0.16 \pm 0.04\%$, Savage et al., 2011) compositions for Sapine, Hakai, Providence, and La Jara (Table 1). The mixed-shale lithology characterizing Elder was measured directly for this study ($-0.39 \pm 0.06\%$). The $\Delta^{30}\text{Si}$ composition chosen for Quiock’s andesite lithology ($-0.23 \pm 0.09\%$) derives from Opfergelt et al. (2012). Observed $\Delta^{30}\text{Si}_{\text{diss-rock}}$ for all reported storm events are greater than 0 (Table 3) and display wide variability. Across catchments these values range from
+0.01 ± 0.23‰ (Quiock) to +2.26 ± 0.19‰ (Providence) and within individual events (e.g., +0.01 ± 0.23‰ to +0.68 ± 0.17‰ for the Quiock storm event). In order to better appraise Ge versus Si enrichment in stream compared to the initial composition provided by rock dissolution, we also normalize dissolved Ge/Si ratios to the source rock Ge/Si ratio ((Ge/Si)$_{\text{diss}}$/ (Ge/Si)$_{\text{rock}}$). In the absence of further constraints on local bedrock Ge content, we use an invariant (Ge/Si)$_{\text{rock}}$ ratio of 1.5 µmol mol$^{-1}$, consistent with values reported in Baronas et al. (2020), Edmond et al. (1995), and Rudnick & Gao (2013). All $(\text{Ge/Si})_{\text{diss}}$ and $(\Delta \delta^{30}\text{Si})_{\text{diss-rock}}$ values are summarized for each site in Table 3 (compiled results, including (Ge/Si)$_{\text{diss}}$/ (Ge/Si)$_{\text{rock}}$ values, can be found in the Supplementary).

5. Discussion

We observe no universal relationship between dissolved Si or $\delta^{30}\text{Si}$ as a function of runoff across our seven storm events and six low-order catchments. We now seek to determine why this is the case and to identify the common variable(s) responsible for observed intersite variability in stream Si chemistry. Toward this objective, we present a rigorous evaluation of possible lithologic, climate, hydrologic, kinetic and biologic controls. We explore how the inherent critical zone architecture and functionality of a given site can generate unique chemical and isotopic signals, and highlight the utility of high frequency, storm event datasets compared to those more commonly reported over annual or monthly resolutions (Bluth & Kump, 1994; Dessert et al., 2003; Gaillardet et al., 1999; Godsey et al., 2019, 2009; Oliva et al., 2003; West et al., 2005; White & Blum, 1995).

5.1. Absence of a Direct Control by Storm Runoff on Si Chemistry

Accounting for catchment lithology (see Section 4.3.2, Table 3) creates a pronounced distinction between the trends observed in raw (Figure 4a) and bedrock-corrected (Figure 5a) stream Si concentrations as a function of runoff. Uncorrected dissolved Si concentrations are highly scattered as a function of runoff with Providence, La Jara, and Elder grouping at the upper bound of dissolved Si concentrations while the remaining sites (Sapine, Hakai, and Quiock) group toward the lower bound. This observed scatter changes when the bedrock correction is applied, revealing a rather stark grouping between mainland and island catchments. Within the mainland
catchments (Providence, La Jara, Sapine, and Elder), lower flux-weighted \( f_{\text{diss}}^{\text{Si}} \) are generally associated with higher runoff rates, though individual data points display significant scattering about this apparent trend. In contrast, the island catchments (Hakai and Quiock) export nearly the entirety of dissolved Si expected to be released from bedrock \( (\Delta^{30}\text{Si}_{\text{diss}} \sim 1) \) and Hakai shows little sensitivity to runoff. Bedrock-corrected \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) values (Figure 5b) illustrate virtually no distinction from uncorrected values (Figure 4b) and remain relatively invariant as a function of discharge.

Storm events from different sites do not overlap in \( f_{\text{diss}}^{\text{Si}} \) versus \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) space (Figure 6). Rather, the dataset is principally distinguished by individual catchments even after correction for atmospheric inputs and lithology. Hence, individual catchment properties clearly produce distinct solute and stable isotope signatures. Further, we find no unique relationship across catchments between \( f_{\text{diss}}^{\text{Si}} \) and \( \Delta^{30}\text{Si}_{\text{diss-rock}} \): a range of catchments display similar \( f_{\text{diss}}^{\text{Si}} \) values of 0.2–0.3, with widely varying (by \( \sim 2\%e \)) \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) values (Providence, Sapine, and Elder), while catchments with similarly low \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) values (<1‰) are characterized by very different \( f_{\text{diss}}^{\text{Si}} \) (Sapine, La Jara, Hakai, and Quiock).

We note the existence of Si isotope fractionation at Hakai \( (\Delta^{30}\text{Si}_{\text{diss-rock}} \sim 0.7\%e) \) despite apparent congruent release of Si from bedrock \( (f_{\text{diss}}^{\text{Si}} \sim 1) \) and Hakai shows little sensitivity to runoff. Bedrock-corrected \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) values (Figure 5b) illustrate virtually no distinction from uncorrected values (Figure 4b) and remain relatively invariant as a function of discharge.

In total our observations (Figure 6) are at odds with a unique fractionation mechanism and mass balance model for Si isotopes across our six catchments. We now turn to consideration of a variety of hydrologic and chemical factors that could underly these observations.

5.2. Hydrologic and Chemical Controls

Thus far we have demonstrated that these headwater storm events are representative of overall solute export dynamics in these streams (Figure 2). We now explore a series of simple scenarios to explain the cross-site
variability in this export record. First, we ask if this variability is controlled by binary mixing as suggested in previous studies of Si isotope fractionation at individual sites. Next, we consider the distinction among variable hydrologic transit time distributions, and finally the possibility of multiple fractionating processes.

5.2.1. End Member Mixing

Mixing controls are frequently invoked in the silicate literature to explain δ³⁰Si behavior as a function of discharge in a variety of locations (Engström et al., 2010; Georg et al., 2006a; Riotte et al., 2018; Steinhoefel et al., 2017). In this framework, stream δ³⁰Si variability is driven by the interplay between a constant deep groundwater (baseflow) supply and varying contributions from soil waters. During dry periods, when discharge rates are low, stream δ³⁰Si is driven by baseflow. This end member experiences long contact times with fresh weatherable primary minerals, generating groundwater δ³⁰Si close to, but slightly elevated with respect to the bedrock composition due to the balance between primary mineral dissolution and the precipitation of secondary minerals more favorably formed at near-surface conditions. Soil solutions during these periods are typically elevated (δ³⁰Si > groundwater) due to low water/rock ratios favoring the formation of secondary clays and the additional effects of biological activity characterized by extensive plant cycling and uptake of dissolved Si. Following this framework, the first flush coming out of the dry season should drive stream δ³⁰Si to more elevated values with increasing discharge as a result of the mobilization of these enriched δ³⁰Si pools in the soil. This flushing should theoretically be associated with dilution in dissolved Si as a result of lower soil water Si concentrations. In contrast, the wet season is marked by more frequent high discharge events, increased water/rock ratios in the soil triggering clay and phytolith (or litter) dissolution and releasing light ³⁰Si back into soil waters (Steinhoefel et al., 2017). The result is lower soil water δ³⁰Si, such that increased contributions from soil water to the stream should drive δ³⁰Si to lower values during periods of high discharge. Clay dissolution taking place within the stream has also been put forth as a potential low δ³⁰Si source in the binary mixing schematic, the extent of which depends on the suspended sediment load and clay content (Georg et al., 2006a).

Extrapolation of this end member framework to the global scale of our δ³⁰Si observations may not necessarily be expected to succeed. However, given the commonality of this approach in application to riverine δ³⁰Si analysis, we begin by considering whether mixing of two end members, common to all sites, could underlie our observed cross-site δ³⁰Si behavior. Binary mixing is tested by comparing Δ³⁰Si_diss-rock against 1/ƒ_diss (Figure 7a; Equation 4). Indeed, given that 1/ƒ_diss scales with (Na/Si)_diss, binary mixing in the Δ³⁰Si_diss-rock = 1/ƒ_diss space should result in a straight line. No clear cross-catchment relationship is observed in this mixing space. Intra-site trends are difficult to parse, particularly among the four continental catchments. This observation is not necessarily in contradiction to the typical expectation from prior site-specific studies (described above). However, we find that there is no universal emergence of (1) a high δ³⁰Si, low dissolved Si (or high Δ³⁰Si_diss-rock, low ƒ_diss) shallow soil water pool and (2) a deep, low δ³⁰Si, high dissolved Si (low Δ³⁰Si_diss-rock, high ƒ_diss) groundwater pool that can explain our data across several sites. Altogether, mixing between internal δ³⁰Si stores is anticipated to play a role in observed site-specific variability, but lack of a systematic relationship across sites discredits the possibility of globally consistent δ³⁰Si soil and groundwater end members governing cross-site behavior.

5.2.2. Hydrologic Routing

We next consider the possibility that the fluid solute signatures measured in these six catchments all reflect varying degrees of a common fractionating reaction which removes Si from solution and enriches the remaining Δ³⁰Si_diss-rock, as is typical of secondary mineral formation (Fernandez et al., 2019). We employ a simplifying assumption that this reaction proceeds via an isotopic distillation process, as has been previously applied to common isotopic signatures (e.g., δ³⁰Si, δ¹⁰Li, δ¹⁸Nd, Mg) of weathering (Baronas et al., 2018; Dellinger et al., 2015; Georg et al., 2007; Tipper et al., 2012). The relationship between Si lost from solution due to vegetation cycling and/or secondary mineral accumulation (ƒ_diss) and the resulting observable fractionation of Δ³⁰Si_diss-rock from some nominal starting value of fluid near equilibrium with bedrock is then given by a simple Rayleigh equation:

\[
\Delta^{30}\text{Si}_{\text{diss-rock}} = 1000 \cdot (\alpha_s - 1) \cdot \ln \left( \frac{f_{\text{diss}}^{\text{Si}}}{f_{\text{diss}}^{\text{Si}_{\text{diss-rock}}}} \right)
\]  

(5)

where ƒ_diss corresponds to the same definition as above (Equations 4 and 5) and α_s corresponds to the fractionation factor associated with either secondary mineral precipitation (α_{sec}) or plant uptake (α_{plant}). This model creates one trajectory of coupled fluid ƒ_diss and Δ³⁰Si_diss-rock evolution for all water exported from a given catchment (Figure 7b). In this application we use α_s values derived from experiments rather than from field data. Alternative
frameworks, such as open-flow through models (Bouchez et al., 2013) utilize the difference in isotope signatures between clays (or organic material) and waters, as measured in the field, as an estimate of the fractionation factor. However, in the Rayleigh model, the difference between solution and the bulk of neo-formed solids (i.e., the cumulative product) is not an accurate estimate of the isotope fractionation factor. Rather, the fractionation factor in a distillation model manifests as the difference in isotopic ratio between dissolved Si (the reactant) and an instantaneously formed solid, which is often challenging to measure in the field due to differences in the timescales of formation of solid and solute isotope signatures. We return later to comparison between the Rayleigh approach and other widely applied models, as well as to the various ways isotope fractionation factors can be estimated for such models.

Simply applying this basic Rayleigh distillation model (Equation 5) to our data carries the assumption that all fluid spends an equivalent amount of time traveling through the watershed and arriving in the stream, for
example, a “plug” or “piston” flow system. Deviation from this basic model among our six catchments is then introduced through differences in the distribution of fluid travel times through a given watershed. These non-uniform travel time distributions (TTDs) offer a collective representation of the variety of pathways water may take in passing through the critical zone into the drainage network of a given landscape. In practice it is commonly observed that most of the fluid discharging into these fluvial drainage networks took relatively short (or fast) pathways through the watershed, while some smaller amount took much longer (Hornberger et al., 2001; Maloszewski & Zuber, 1996; McGuire & McDonnell, 2006). This leads to TTD functional forms that are typically weighted toward relatively young water but feature long tails which integrate a small portion of much longer transit times. Common examples include exponential and gamma functions (e.g., Gabrielli et al., 2018; McGuire & McDonnell, 2010).

Convolution of these non-uniform TTDs with the chemical reactions that supply geogenic solutes from water-rock interactions results in relatively “muted” or dilute concentrations relative to a piston-flow system (Maher, 2010, 2011). This is due to the relatively large proportion of young (i.e., less reacted) water sampled by these TTD functions. In relation to the present analysis, exponential and gamma TTDs would produce a relatively larger \( f_{\text{diss}}^\text{Si} \) (i.e., less evolution from the bedrock-equilibrated concentration, Equation 4) than the corresponding piston-flow model for a given mean fluid travel time. Similarly, these TTDs may be applied to an isotopically fractionating reaction. Convolution of an exponential TTD with a fractionating first-order irreversible reaction produces a closed-form expression which is essentially a modification to the Rayleigh model (Druhan & Maher, 2017), given here in terms relevant to the present silicon isotope analysis:

\[
\Delta^{30}\text{Si}_{\text{diss-rock}} = 1000 \cdot \left[ \frac{1}{\alpha_s + f_{\text{diss}}^\text{Si}} \cdot (1 - \alpha_s) - 1 \right] \tag{6}
\]

This particular function is based on a convolution integral combining a first-order irreversible reaction with an exponential TTD. Those familiar with common isotope fractionation models may note that Equation 6 is functionally similar to the evolution of a closed system, such that the reactant and product pools are always offset by \( \varepsilon = 1000 \cdot (\alpha_s - 1) \). We caution that this is simply a result of the use of a reaction rate law that is first order and irreversible. One should not expect that all fractionating reactions would convolve with an exponential TTD to produce this functional form. Similarly, convolution of the same first-order irreversible reaction with a different choice of TTD produces a different solution. In the case of the gamma TTDs, a closed form expression is not achievable, and the integral is calculated numerically (Druhan & Maher, 2017).

For the present application, we select a value of \( \alpha_s = 0.9986 \) (\( \varepsilon_s = -1.38 \% \)) that represents average contributions from both secondary weathering and biological cycling (discussed further Section 5.2.3). The results (Figure 7b) are a suite of \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) versus \( f_{\text{diss}}^\text{Si} \) trajectories that distinguish among the six catchments. Within this framework, Providence closely adheres to a piston-flow model (Equation 5), which would suggest a relatively uniform fluid travel time through this system. The Elder storm was the most extreme event captured in our

\[\text{Figure 7.} \text{ Evaluation of end member mixing, hydrologic routing, and fractionation pathway controls across storm events. Shaded circles and diamonds correspond to individual sampling points and discharge-weighted averages (Equation 2), respectively. For Quoock, given the congruent export of dissolved Si from the system suggested by our analysis, we set } f_{\text{diss}}^\text{Si} \text{ to 1.0. The presence of a global end member mixing (Section 5.3) is refuted as a possible explanation for cross-site differences (a). Application of various travel time distributions (b) ranging from “plug” or “piston” flow (analogous to Rayleigh distillation along a given flow path, Equation 5) to exponential (Equation 6) to gamma (Section 5.2.2) accurately capture observed stream } \Delta^{30}\text{Si}_{\text{diss-rock}} \text{ for the mainland catchments, whereas release of dissolved Si at the two island catchments are too congruent to be accurately evaluated through such an approach. Reactivity (c) was evaluated through the application of “steady-state open-flow through” (solid lines, Equation 7) and “Rayleigh” (dashed lines, Equation 5) isotope mass balance models. Here we show three possible scenarios: (1) 100% net secondary mineral formation (orange dashed and solid lines) with a fractionation factor (\( \varepsilon_{\text{diss}} \)) equal to laboratory derived values of \(-2.00\% \) (Fernandez et al., 2019; Geilert et al., 2014; Roerdink et al., 2015; Yanhe et al., 1995; Ziegler, Chadwick, Brzezinski, et al., 2005); (2) 100% net biological uptake (green dashed and solid lines) with \( \varepsilon_{\text{bio}} \) reflecting the average of estimates determined by prior studies \((-0.76 \pm 0.42\%_c \), Ding et al., 2008; Frick et al., 2020; Opfergelt et al., 2010; Sun et al., 2008, 2016); and (3) a 50/50 scenario where net secondary clay formation and biological uptake have equal contributions (black dashed and solid lines), which is adjusted in the steady-state open-flow through model by setting } f_{\text{diss}}^\text{Si} = f_{\text{bio}}^\text{Si} \text{ (Equation 7) while in the Rayleigh-model this is represented by an effective fractionation factor that represents an average between the two (\( \varepsilon_{\text{diss}} = -1.38 \pm 0.52 \)). Providence is the only storm event that is compatible with piston-flow behavior coupled with a Rayleigh model, through a scenario in which only secondary clays are forming. The rest of the storm events appear to display more steady-state open flow through behavior, but this is less established. Error bars correspond to 2SD. An independent metric, (Ge/Si)$_{\text{bio}}$/((Ge/Si)$_{\text{diss}}$ (Section 4.3.2), (d) is afforded by germanium (Ge) measurements from Sapine, Elder, La Jara and Hakai along with prior measurements for Providence (Aguirre, 2019). Both flux-averaged and individual measurements are shown for La Jara, Sapine, Elder, and Hakai. Only the flux-averaged value is given for Providence since the Ge/Si ratios do not directly correspond to the same storm event. A dashed vertical line indicates the direction of Ge/Si partitioning in the fluid phase when dissolved Si is scavenged from solution due to secondary mineral precipitation and biological uptake.}]

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dataset (Section 2.3) and would be anticipated to source a substantial portion of discharge from very short fluid travel times. This is consistent with the exponential TTD (Equation 6), which strongly weights these short residence times and shows good agreement with the Elder $\Delta^{30}\text{Si}_{\text{diss-rock}}$ versus $f_{\text{Si}}^{\text{diss}}$ values for the same $\alpha_\text{f} = 0.9986$. Finally, two common choices of gamma distributions are provided (Maher, 2011) distinguished by their shape factor, $\alpha$. As the value of the shape factor decreases, the tail of the TTD becomes more pronounced and thus the range over which fluid travel times are sampled becomes larger. In this application the La Jara system appears to agree closely with a gamma TTD subject to a shape factor of $\alpha = 0.5$, notably including the anomalous high $f_{\text{Si}}^{\text{diss}}$ values that appeared at the end of the storm event. Sapin appears to fall closer to gamma TTD with a shape factor of $\alpha = 0.25$, suggesting an overall broader distribution in fluid residence times. Since the two island catchments have high $f_{\text{Si}}^{\text{diss}}$ values that essentially indicate congruent release of Si, meaningful evaluation of TTD effects are impeded. However, we do note that these two catchments, particularly Quiock, produce $f_{\text{Si}}^{\text{diss}}$ and $\Delta^{30}\text{Si}_{\text{diss-rock}}$ values that essentially mark the starting point of unfractionated fluid in the Rayleigh model space.

Clearly many factors complicate the evaluation of these TTDs as distinguishing features governing $\Delta^{30}\text{Si}_{\text{diss-rock}}$ signatures among our six catchments. These storm events are by definition transient periods of heightened discharge, and thus the underlying rates of fluid transport are time-variable. Recent efforts have focused on Storage Selection (SAS) functions and related frameworks that allow for this unsteady behavior (Benettin, Soulsby, et al., 2017; Harman, 2015; Van Der Velde et al., 2012), but parameterization of these models requires calibration against extensive datasets of tracer concentrations. Furthermore, it is highly unlikely that the $\Delta^{30}\text{Si}_{\text{diss-rock}}$ signatures of these catchments result from a single fractionating pathway (discussed further, Section 5.2.3). Nevertheless, our results show good agreement between the watershed datasets and modeled isotope signatures based on a single underlying fractionation factor and a range of common TTDs often applied to describe watershed transit times. Upon this basis, we cannot rule out an explanation for the intra-site variability observed in our lithology-corrected datasets originating from distinctions in the hydrologic routing of fluid through these catchments. As more constraints on hydrological flowpaths become available for these locations thanks to future research at the Critical Zone Observatories, convolution with fractionating reactions to model isotopic datasets such as ours presents a compelling opportunity to test this hypothesis. Importantly, until now this analysis of our cross-site Si isotope dataset in terms of variability in hydrological routing has relied on an invariant isotope fractionation factor thought to reflect equivalent roles of clay formation and biological uptake at all sites. However, and as discussed in the next section, our data clearly indicate that the role of these different fractionation pathways differs from site to site.

5.2.3. Multiple Fractionating Pathways

We next turn to the role of multiple fractionating pathways in producing our observed stream Si fluxes and $\delta^{30}\text{Si}$ ratios across these six catchments and seven storm events. In doing so, we effectively hold the influence of fluid TTDs fixed (Section 5.2.2) and explore variability in the fractionation factor(s) and fractionating process(es) that may impact site-specific behavior. In addition to the Rayleigh model, we utilize a steady-state open flow through model (Bouchez et al., 2013) where stable isotope ratios are predicted based on input and output solute fluxes between three biogeochemical processes characterizing the weathering zone: biological cycling, bedrock dissolution, and secondary mineral formation. In this framework, $\delta^{30}\text{Si}$ of the dissolved load exiting a given catchment can be described based on the mass transfer of Si between congruent dissolution of primary minerals, the formation of secondary minerals, or uptake via vascular plants. The latter two processes (secondary minerals and biological uptake) are the principal fractionating mechanisms for $\delta^{30}\text{Si}$ where their respective fractionation factors dictate the magnitude of dissolved Si isotopic partitioning. The original Bouchez et al. (2013) formulation is expressed here for Si stable isotopes:

$$\Delta^{30}\text{Si}_{\text{diss-rock}} = f_{\text{sec}}^{\text{Si}} \cdot \epsilon_{\text{sec}} + f_{\text{bio}}^{\text{Si}} \cdot \epsilon_{\text{upt}}$$  \hspace{2cm} (7a)

$$1 - f_{\text{diss}}^{\text{Si}} = f_{\text{sec}}^{\text{Si}} + f_{\text{bio}}^{\text{Si}}$$  \hspace{2cm} (7b)

where $\Delta^{30}\text{Si}_{\text{diss-rock}}$ corresponds to bedrock-corrected stream $\delta^{30}\text{Si}$ values and $\epsilon_{\text{sec}}$ and $\epsilon_{\text{upt}}$ correspond to the associated fractionation factors (given in epsilon notation) for secondary mineral precipitation and plant uptake by the biosphere, respectively. The relative proportions of secondary mineral precipitation and biological uptake to the overall amount of Si removed from the dissolved load $(1 - f_{\text{diss}}^{\text{Si}})$ is represented by $f_{\text{sec}}^{\text{Si}}$ and $f_{\text{bio}}^{\text{Si}}$, respectively. Note $f_{\text{diss}}^{\text{Si}}$ still remains consistent with Equations 4 and 5 as the bedrock-normalized dissolved Si concentration,
serving as a proxy for the amount of Si remaining in the fluid from some nominal bedrock-equilibrated starting point. Importantly, within this framework $\Delta^{30}\text{Si}_{\text{diss-rock}}$ depends linearly on the removal terms $f_{\text{app}}^{\text{sec}}$ and $f_{\text{app}}^{\text{bio}}$, as it would in the so-called "closed-system" model. Nevertheless, we stress that this linear dependency is obtained from a very different set of assumptions regarding the transfer of elements within the modeled system, which are those of an open-flow through system rather than those of a closed system lacking any export of matter out of the Critical Zone, for example, by erosion. One merit of this modeling approach is that it explicitly considers the possibility for mineral dissolution (both primary and secondary) to occur along flowpaths in the weathering systems, where the Rayleigh model assumes that mineral dissolution occurs only where the fluid acquires its "starting" composition. However, the open-flow through model neglects any spatial separation between these different processes, and considers that the entirety of the fluid contained in the weathering system is well mixed at all times. Therefore, for the sake of their simplicity, both model types feature assumptions that somewhat contradict field observations, such that that their capability to explain data needs to be systematically examined, when possible, as done in the present study. Regardless of their different sets of underlying assumptions, the basic principle in both the Rayleigh (Equation 5) and open flow through (Equation 7) approaches is that each sample taken during a given storm reflects progression of a (set of) reaction(s) subject to mass balance. In all these approaches, storm flow is assumed to nominally evacuate water, the composition of which is acquired from a relatively consistent set of weathering reactions over a time scale that is much longer than that of the flood itself.

Three fractionation scenarios are considered (1) 100% secondary mineral formation, (2) 100% plant uptake, and (3) 50/50 contributions from both secondary weathering and biological cycling. The fractionation factors implemented for these scenarios correspond to the average of previously published laboratory-derived values yielding $\epsilon_{\text{sec}}$ of $-2.00\%e$ (Fernandez et al., 2019; Geilert et al., 2014; Roerdink et al., 2015; Yanhe et al., 1995; Ziegler, Chadwick, Brzezinski, et al., 2005) and an $\epsilon_{\text{bio}}$ of $-0.76\%e$ (Ding et al., 2008; Frick et al., 2020; Opfergelt et al., 2010; Sun et al., 2008; Sun et al., 2016). As for the application of the Rayleigh model above, here these parameter values are restricted to fractionation factors determined through controlled laboratory experiments. Such an approach is necessary in multi-component isotope-enabled numerical reactive transport models (Druhan et al., 2019 and citations therein) and has been recently leveraged to successfully describe $\delta^{13}\text{C}$ weathering dynamics in unit hillslopes (Golla et al., 2021). The combined apparent fractionation factor ($\epsilon_{\text{app}} = -1.38\%e$) is determined from taking the average of the entirety of the known values for both fractionating pathways and is consistent with the value used in the Rayleigh (Equation 5) and TTD (e.g., Equation 6) models presented in Section 5.2.2.

Application of the open-flow through and Rayleigh models to the storm event data is presented in $\Delta^{30}\text{Si}_{\text{diss-rock}}$ versus log($f_{\text{app}}^{\text{Si}}$) space (Figure 7c), where Rayleigh distillation now appears as a straight line. All three fractionation factors are considered, and generally appear to distinguish between the cross-site $\Delta^{30}\text{Si}_{\text{diss-rock}}$ versus $f_{\text{app}}^{\text{Si}}$ data-sets. Providence again exhibits Rayleigh-distillation behavior that is closely bounded by the cumulative apparent fractionation factor ($\epsilon_{\text{app}} = -1.38\%e$) and the secondary mineral formation fractionation factor ($\epsilon_{\text{sec}} = -2.00\%e$). The remaining mainland catchments (La Jara, Elder, and Sapine) all display behavior which is more consistent with the Bouchez et al. (2013) compartmental model. Elder and La Jara fall within the envelope between $\epsilon_{\text{app}}$ and $\epsilon_{\text{sec}}$, whereas Sapine lies closer to the plant uptake value ($\epsilon_{\text{app}} = -0.76\%e$). The two island catchment storm events (Hakai, Quiock) are again exporting dissolved Si consistent with the net input to these systems and reflect correspondingly minimal fractionation, thus obscuring interpretation based on these model frameworks. However, we note that the combination of high $f_{\text{app}}^{\text{Si}}$ and low $\Delta^{30}\text{Si}_{\text{diss-rock}}$ values at these two sites is consistent with the inferences made by Bouchez et al. (2013) from the open-flow through model that the preservation (or erosion) of fractionated secondary materials such as clays or litter, leading to lower $f_{\text{app}}^{\text{Si}}$ values, is a pre-requisite for the expression of isotope fractionation in streams.

In comparison with the prior TTD analysis (Section 5.2.2), this similarly reasonable agreement poses a basic question: is it more realistic to expect that six catchments hosting low-order streams could differ in their overall $\delta^{30}\text{Si}$ fractionation factors by $>1.0\%e$ or differ in the functional form of their fluid TTDs? Plausible arguments can be made for both, or some combination of the two. We note that the La Jara system is sub-humid and the event corresponds to spring snowmelt, suggesting conditions that would not normally support biological uptake as the primary fractionating pathway. Conversely, biological uptake could be anticipated to affect Si isotope fractionation at the two Island catchments, Hakai and Quiock. The former is characterized by limited secondary mineral formation most likely impacted by high organic-rich (i.e., DOC) interstitial waters that can hamper clay formation.
(Oliver et al., 2017), while the latter is located in a highly productive tropical forest developed on a strongly cation-depleted soil (Buss et al., 2010; Clergue et al., 2015; Dessert et al., 2020). However, the data from both of these systems appears to indicate only minor net effects of vegetative cycling, which are not extensive enough to impact \( f^{\text{Si}}_{\text{diss}} \), maybe due to complete recycling of Si from litter (Schuessler et al., 2018). In the \( \delta^{30}\text{Si} \)-\( f^{\text{Si}}_{\text{diss}} \) space the Sapine data fall below even the lowest models presented for the assessment of multiple fractionating pathways, lending support to our earlier inference of hydrologic routing as a primary control, as indicated by the close agreement of the gamma TTDs to the La Jara and Sapine storm events (Section 5.2.2, Figure 7b). We further note the important role of hydrologic routing in the extremity of the Elder event, where very large discharge volumes and rapid fluid transit could potentially overprint the evacuation of geochemically evolved fluid to the stream.

Further support for inter-site differences in fractionating and hydrologic pathways as the dominant control on observed Si isotope values is provided by the ratio of germanium (Ge) to silicon, for which data are available in five of our catchments (four of which have been measured in this study, Section 4.2). Germanium acts as a pseudo-isotope for silicon and the Ge/Si ratio partitions between pools in a similar fashion to \( \delta^{30}\text{Si} \) when Si is scavenged from solution due to secondary mineral formation. However, the elemental equivalent of the isotope fractionation factor called the partition coefficient \( K_D = \frac{c_{\text{Ge,secondary}}}{c_{\text{Si,primary}}} \) is in this case inverted, leading to lower fluid Ge/Si ratios and enriched secondary mineral Ge/Si ratios as the reaction proceeds (Aguirre et al., 2017; Ameijeiras-Mariño et al., 2017; Baronas et al., 2020, 2018; Gaspard et al., 2021; Kurtz et al., 2002; Lugolobi et al., 2010; S. Opfergelt et al., 2010; Perez-Fodich & Derry, 2020; Qi et al., 2019; Scribner et al., 2006).

In contrast, evidence based primarily on phytoliths suggests Si uptake by vegetation strongly discriminates against Ge, causing the remaining fluid Ge/Si ratio to increase (Blecker et al., 2016). However, we note a recent study reporting whole plant Ge/Si ratios which suggests the capacity for no discrimination, and in some cases even preferential Ge uptake (Frings et al., 2021). Partitioning of Ge relative to Si across fluids, clays and vegetation creates quite pronounced signatures, and critically, should adhere to the same underlying fractionating effects and be subjected to the same TTDs as the \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) data.

From this basis, a direct comparison between stream \((\text{Ge/Si})_{\text{diss}}\) versus \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) offers an independent opportunity to assess the role of multiple fractionation pathways (Figure 7d). Hakai’s \((\text{Ge/Si})_{\text{diss}}\) versus \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) ratio of \( \sim1.0 \) is in agreement with the \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) and \( f^{\text{Si}}_{\text{diss}} \) results and further supports our inference that this system is exporting the entirety of Si expected to be released from bedrock. We note that a recent limited dataset of \((\text{Ge/Si})_{\text{diss}}\) for Quiock would also translate to stream \((\text{Ge/Si})_{\text{diss}}\) versus \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) values of \( \sim1.0 \) (Gaspard et al., 2021), consistent with a congruent weathering signature. The remaining four catchments (Providence, Elder, and Sapine and La Jara) all exhibit \((\text{Ge/Si})_{\text{diss}}\) versus \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) ratios that require secondary clay formation as a dominant process impacting Si exports at these sites (Figure 7d). For Providence and Elder, a dominant effect of clay formation is consistent with the \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) data and do not offer any further distinction between the effects of TTDs (Section 5.2.2) and variable fractionation models (Section 5.2.3). In Sapine and La Jara, stream \((\text{Ge/Si})_{\text{diss}}\) versus \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) ratios are far too low to be describable by a system in which Si is predominantly cycled by vegetation which exhibits Ge discrimination, lending support to the model laid out in Figure 7b for gamma TTDs (Section 5.2.2) rather than strong vegetation cycling (Section 5.2.3).

In total, the Ge/Si ratios of these samples offer an independent proxy for the pathways by which silica is mobilized and retained in low order catchments. These data indicate that our interpretations based on Si isotopes alone are generally robust. Both Hakai and Quiock reflect \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) and \((\text{Ge/Si})_{\text{diss}}\) ratios typical of (near-) congruent silica exports. This is consistent with other work suggesting congruent weathering in an independent catchment at the Hakai site (Nelson et al., 2021). The remaining catchments appear to be dominated by secondary clay formation. However, the inclusion of Ge/Si as an independent check cannot definitively rule out either variation in fluid TTDs or variability in the fractionation factor as an underlying contributor to the \( \Delta^{30}\text{Si}_{\text{diss-rock}} \) signatures of these low-order streams. Rather, it appears both models are required in order to fully explain our observations.

5.3. A Combined Model

Our data-model comparisons suggest that a combination of unique TTDs and fractionation factors influence the distribution of storm event silica solute signatures across these six catchments. We thus suggest that a combined
modeling approach offers a more realistic and comprehensive means of encapsulating the effects of hydrology and biogeochemical reactions reflected in our global event-based dataset.

The two models (Sections 5.2.2 and 5.2.3) are combined by defining solution spaces for each of the four TTDs using upper (secondary mineral precipitation) and lower (vegetation) bounds for the corresponding fractionation factor (Figure 8a). Within the range of $f^{Si}_{diss}$ and $\Delta^{30}Si_{diss-rock}$ delineated by our datasets this combined model produces clear distinctions in the range of values that each TTD envelope can encapsulate. Both Elder and La Jara fall within a range defined by the exponential TTD (Equation 6), suggesting that the Elder system is strongly impacted by secondary mineral formation relative to vegetation. Sapine is quite muted in isotopic enrichment, resulting in close agreement with the gamma TTDs. For the two shape factors tested, Sapine also appears to correspond to strong secondary mineral precipitation effects given a shape factor of 0.25, though other combinations of gamma TTD could be expected to suggest more vegetation influence. The two island catchments reflect $f^{Si}_{diss}$ values so close to 1.0 that it remains difficult for any TTD to distinguish their behavior. As noted earlier, Providence remains strongly associated with a large fractionation factor and a piston-flow fluid transit time. There, the flux-weighted values of $f^{Si}_{diss}$ and $\Delta^{30}Si_{diss-rock}$ cannot be reasonably achieved with any of the other models.

The case of Providence is an interesting one in that two separate mass balance models appear to explain the same stream isotope dataset. Our analysis above shows that a fractionation factor based on experimental studies of secondary silicate formation ($\epsilon_{sec}$ of $-2.00\%e$) together with a combined piston-flow TTD and Rayleigh-type fractionation matches observed storm stream Si isotope signatures at this site. Using a longer-term stream dataset yielding similar $\delta^{30}Si$ values, Frings et al. (2021) proposed that Providence stream Si isotope signatures could instead be explained by an open-flow through model with larger $\epsilon_{sec}$ of $-3\%e$, consistent with the measured difference between clays and waters. Therefore, we are faced with a case where two models can explain the
observations, both in a way that is internally consistent, and each presenting their own caveats regarding the assumptions formulated based on the processes at play in the critical zone. This comparison underlines the need to further test these underlying assumptions when such simple models are the only option; as well as to develop and apply more realistic isotope models that encompass the variety of flowpaths and biogeochemical reactions in the critical zone, along with some description of their spatial distribution (e.g., Golla et al., 2021).

To our knowledge, this is the first attempt to combine a hydrologic model featuring non-uniform fluid transit representations with variability in the effective fractionation factor governing an isotopic signature. Given this novelty, it would seem reasonable to expect some additional testing of the model framework against an independent dataset. In pursuit of this requirement, we return to the Ge/Si ratios and ask whether the combined model can also describe this second metric for silica cycling in these catchments. Importantly, the partition coefficients for Ge/Si are not as tightly constrained as the fractionation factors for δ30Si (Frings et al., 2021). However, in this test, our intent is to make use of the general capacity for Ge to enrich in secondary mineral phases, and thus produce a negative correlation between fluid phase Ge/Si and δ30Si influenced by this reactive pathway. In contrast, the strong discrimination of Ge suggested to occur based on the phytolith composition of vegetation creates a very different correlation where both fluid phase Ge/Si and δ30Si should enrich when subjected to biological uptake (Figure 7d). Such a marked difference in behavior should be observable regardless of the choice of TTD and inaccuracies in the Ge/Si partition coefficients are minimal relative to this fundamental difference between δ30Si and Ge/Si. Once again, we make use of data measured for the same storm event samples in four of our catchments. For Providence, we again use the data reported by Aguirre (2019) and further use their corresponding fSi, values, applying the same atmospheric correction as the event-based dataset (Equation 3 with correction factor 0.183). To apply our combined model, we employ a partition coefficient of 6.0 for secondary mineral formation and discriminate Ge during uptake by vegetation using a value of 0.005 (Baronas et al., 2018, 2020; Frings et al., 2021; Perez-Fodich & Derry, 2020).

The resulting model, now adjusted to Ge/Si partition coefficients and with a (Ge/Si)dis/(Ge/Si)rock ratio of 1.0 as a starting point, is applied to the five catchments using a parameter set based on the Δ30Si_diss-rock versus fSi_diss model (Figure 8b). Hakai falls near the starting fluid composition in this model. Sapine, Providence and Elder present an interesting opportunity in that all three catchments appear to be strongly influenced by secondary clay formation in their Δ30Si_diss-rock values (Figure 8a) but are differentiated by unique TTDs. Extending this inference to their (Ge/Si)dis/(Ge/Si)rock ratios, we consider that secondary mineral precipitation constitutes 100% of the net effective fractionation factor (Figure 8b). The only difference between these three catchments is that Elder is modeled using an exponential TTD, Providence is considered a uniform flow field, and Sapine corresponds to a gamma TTD with shape factor of 0.25. Close agreement is noted for the (Ge/Si)dis/(Ge/Si)rock versus fSi_diss values and the corresponding models across all three sites, creating a solution that is consistent across Δ30Si_diss-rock and (Ge/Si)dis/(Ge/Si)rock using the unified model. Finally, the La Jara dataset presents a distinct relationship between Δ30Si_diss-rock and (Ge/Si)dis/(Ge/Si)rock which was not immediately evident in the cross-plot (Figure 7d). Here, the combined model (Figure 8) highlights the fact that the (Ge/Si)dis/(Ge/Si)rock values for La Jara (~0.09) are much lower than what would be expected given the corresponding Δ30Si_diss-rock values of ~1.0‰, regardless of the chosen TTD. Elder already essentially defines the maximum effect of clay formation on the overall fractionation of (Ge/Si)dis/(Ge/Si)rock for an exponential TTD and a partition coefficient of 6.0. Invoking a gamma distribution with the same strongly clay-dominated effects only further mutes the fractionation, resulting in a solution which would fall between the results shown for Elder and those shown for Sapine (Figure 8b). Thus, the apparent discrepancy between (Ge/Si)dis/(Ge/Si)rock and Δ30Si_diss-rock for La Jara appears to be a result of the choice in fractionating pathways affecting these two (pseudo)isotope systems in this catchment (or differences in the timescale of δ30Si and Ge/Si partitioning; Fernandez et al., 2021), rather than a result of the choice of TTD. In our combined model, either the fractionation factor impacting Δ30Si_diss-rock must be strongly muted at La Jara, or the partition coefficient impacting (Ge/Si)dis/(Ge/Si)rock must be larger than the parameter values describing the rest of the streams analyzed in this study. As is apparent in Figure 8b, the 6.0 partition coefficient used to represent the affinity for secondary clays to preferentially incorporate Ge would require a TTD that is much closer to a uniform or piston flow model. In Δ30Si_diss-rock versus fSi_diss solution space (Figure 8a), this would require a decrease in the Si isotope fractionation factor such that the value for secondary mineral formation would have to become even lower than the value used for uptake by vegetation. Further, as noted above, the fractionation factors for Δ30Si_diss-rock are generally better constrained than the partition coefficients for (Ge/Si)dis/(Ge/Si)rock, making this explanation unlikely. Assuming instead that the inferences on TTD at La Jara made from the Δ30Si_diss-rock versus
The $f_{\text{Si}}$ relationship are robust, the exponential TTD would also reproduce the La Jara $(\text{Ge/Si})_{\text{diss}}/(\text{Ge/Si})_{\text{rock}}$ ratios for a partition coefficient of 15.0, which is large, although plausible, relative to the range reported in prior studies (Aguirre et al., 2017; Ameijeiras-Mariño et al., 2017; Baronas et al., 2018, 2020; Kurtz et al., 2002; Lugolobi et al., 2010; Oplergelt et al., 2010; Perez-Fodich & Derry, 2020; Qi et al., 2019; Scribner et al., 2006). The La Jara regolith is highly enriched in well-crystallized smectite and zeolite, in which Si incorporation could lead to strong, uncharacterized Ge/Si fractionation (Moravec et al., 2020). Alternatively, we note that the La Jara dataset is the only example of a snowmelt-driven hydrograph in our study, which might have led to the export of a specific solute pool during the event at this site. This solute pool could be topsoil waters enriched in species derived from litter remineralization over the months of snowpack presence, which should lead to low Ge/Si ratios in the stream export. Overall, regardless of the cause, the La Jara system appears to have experienced stronger retention of Ge relative to the other catchments during this discharge event.

In total, the combined model is able to accurately distinguish between catchments that exhibit unique TTDs and are subject to unique fractionating pathways while maintaining consistency across $f_{\text{Si}}$, $\Delta^{30}\text{Si}_{\text{diss-rock}}$ and $(\text{Ge/Si})_{\text{diss}}/(\text{Ge/Si})_{\text{rock}}$ values. We emphasize that the results of this unified TTD and variable fractionation model have been applied to the Ge/Si data without any effort to closely fit the models to the measured values. All parameters were consistent across the five sites, apart from the choice of TTDs (Sapine, Elder and Providence). We note the choice of TTD does not constitute a tuning parameter, but rather reflects three widely applied functional forms that are commonly employed to describe fluid transit times in small catchments (Kirchner et al., 2000, 2001; McGuire & McDonnell, 2006). A single exception is noted for La Jara, which appears to require a distinct $(\text{Ge/Si})_{\text{diss}}/(\text{Ge/Si})_{\text{rock}}$ partition coefficient, but at the cost of essentially creating a new tuning parameter. The apparent consistency across (pseudo)isotope metrics for silicon weathering despite the absence of tuning strongly supports the validity of the combined model framework and underscores the intertwined effects of fractionating pathways and hydrologic routing on observed stream silica signatures. Hence, the fundamental constant across our six sites are the combined effects of hydrology and biogeochemistry, which work in concert with one another to produce observed stream Si chemical and isotopic signals. Site-specific differences arise in how these hydrological and biogeochemical pathways are expressed, which is dictated by the distinct critical zone architecture and functionality of each site, including water drainage pathways, chemical weathering intensity and the balance between secondary mineral weathering and vegetation-driven nutrient cycling.

### 5.4. Resiliency of Silicate Solute Export Signatures During Storm Events

These six low-order streams exhibit remarkably distinct silicate weathering signatures even as they are subjected to storm events. Our model indicates that these distinctions derive from the unique architecture that is specific to each of these sites, rather than from systematic variation along a gradient in some overarching environmental variable. Here we take a holistic definition of critical zone “architecture” which encapsulates physical structure (drainage pathways, soil and regolith depth, lithology, spatial heterogeneity) as well as characteristic climate, vegetation, and biogeochemical reactivity (Anderson et al., 2007; Lin, 2010; Rasmussen et al., 2011; Riebe et al., 2017; Wlostowski et al., 2021) that combine to form the basis for ecosystem functioning and chemical weathering of a given catchment (Troch et al., 2015). The apparent significance of individual architecture in regulating silicate solute signatures complicates cross-site comparison. However, our study shows that within a given system, it appears that architectural controls are preserved even under extreme events. The Elder dataset, for example, reflects a period of time in which an "atmospheric river" established off the coast of northern California, leading to a storm that produced 14% of the total annual volumetric discharge of the stream in 7 days. The La Jara dataset is associated with spring snowmelt, which produced >60% of the total annual water exported through this system. Under such circumstances, biologically cycled solutes (e.g., N, P, DOC, K) and redox-sensitive elements (Fe, Mn, S) have been shown to exhibit significant variability and loss of coherent trends (Basu et al., 2011; Godsey et al., 2019; Knapp et al., 2020; Musolff et al., 2015; Olshansky et al., 2018; Raymond & Saiers, 2010; Rose et al., 2018; Thompson et al., 2011).

In all six of our catchments, we observe that $f_{\text{Si}}$ and $\delta^{30}\text{Si}$ solute signatures remain remarkably stable and tightly clustered around their respective flux-weighted average (no overlap between sites). This is despite the fact that TTDs and the fluid flow pathways they describe are known to shift dramatically through the course of infiltration and discharge events, and differ from the TTDs that would describe the same system during periods of quiescence (Benettin, Bailey, et al., 2017; Birkel et al., 2012; Roa-García & Weiler, 2010; Tetzlaff et al., 2014). Further,
silicon is a bio-cycled element, participating in both abiotic weathering and plant uptake pathways (Alexandre et al., 1997; Conley, 2002; Derry et al., 2005; Street-Perrott & Barker, 2008). Thus, Si uptake by plants is considered an active process across sites contributing to observed fractionation (Fring et al., 2021) and we make no suggestion that this effect is negligible. Our model results suggest that in the studied catchments during storm events, the solute signatures exported by these streams are strongly influenced by secondary mineral formation, which appears to be a principal contributor to such tightly bounded $f_{\text{M}}$ and $\delta^{30}\text{Si}$ signatures.

We find that, even under extreme stress, the site-specific functioning of these catchments is preserved in the silicate solutes they export, indicating a resiliency of these signatures to perturbations. The critical zone architecture inherent in these catchments are imprinted into solute export fluxes where TTDs and reaction pathways still exert first-order control even as these systems are perturbed. Solute signatures of catchments influenced by secondary mineral formation (Sapine, Elder, Providence, La Jara) still differentiate from one another regardless of the stress inflicted. Resiliency, thus, is the preservation of CZ architecture even as storm events overprint the system on shorter timescales. At what point resiliency breaks down under the stress of more frequent and intense extreme events driven by a changing climate presents a substantial unknown that should guide future work.

6. Conclusions

Our cross-site analysis of 157 stream water samples representing seven different storm events from six headwater catchments across the international Critical Zone Observatory network demonstrates no global trend in $\delta^{30}\text{Si}$ as a function of discharge or dissolved Si concentrations, even when bedrock and atmospheric inputs are considered. Consistency between event-scale and longer-term datasets in three of our sites suggest storm events can serve as proxies of long-term silica weathering dynamics. This is consistent with recent cross-site work supporting the notion of events as “snapshots” into long-term variability (Knapp et al., 2020; Minaudo et al., 2019; Rose et al., 2018). Our results highlight the advantage of leveraging storm events, which can comprise significant portions of annual solute export fluxes (40%–80%, Larsen & Simon, 1993; Moatar et al., 2020; Raymond & Saiers, 2010; Yoon & Raymond, 2012), and high frequency geochemical sampling approaches.

A set of three model frameworks were evaluated to determine the underlying factors driving observed variability across sites. Mixing between global Si end members was ruled out as a viable explanation. However, models accounting for multiple fractionation pathways (i.e., reactivity) and variable transit time distributions (i.e., hydrological routing) achieved similar agreement with the observed cross-site behavior. Such accordance offered the opportunity to merge these two distinct conceptualizations, yielding a novel framework based on non-uniform travel time distributions and multiple fractionation pathways. This combined approach indicates that site-to-site differences in hydrological routing characteristics and fractionating reactions are the principal drivers of intra-site variability in observed stream Si solute fluxes and $\delta^{30}\text{Si}$ signatures. Site-specific controls invoked in this study advance recent interpretations from cross-site studies such as Hunsaker & Johnson (2017) and Godsey et al. (2019) who notably proposed that, although chemostasis is commonly observed, each catchment’s chemostat is a site-specific signature. This study offers a foundation for further development and application of combined hydrological and biogeochemical models in application to catchment-scale isotopic signatures, suggesting a promising path forward in constraining silicate solute dynamics and weathering rates.

Data Availability Statement

Datasets for this research are hosted on the open source repository CUAHSI HydroShare: http://www.hydroshare.org/resource/89cfde573936487a9ab4ff406638465b. DOI for this published resource is 10.4211/hs.89cf-de573936487a9ab4ff406638465b (shared under the Creative Commons Attribution CC BY).

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