Determining reliable histories of slip on normal faults with bedrock scarps using cosmogenic exposure data

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Abstract

Cosmogenic exposure data can be used to calculate time-varying fault slip rates on normal faults with exposed bedrock scarps. However, the method relies on assumptions related to how the scarp is preserved, which should be consistent at multiple locations along the same fault. Previous work commonly relied on cosmogenic data from a single sample locality to determine the slip rate of a fault. Here we show that by applying strict sampling criteria and using geologically informed modelling parameters in a Bayesian-inference Markov chain Monte Carlo method, similar patterns of slip rate changes can be modelled at multiple sites on the same fault. Consequently, cosmogenic data can be used to resolve along-strike fault activity. We present cosmogenic 36Cl concentrations from seven sites on two faults in the Italian Apennines. The average slip rate varies between sites on the Campo Felice Fault (0.84 0.23 to 1.61 0.27 mm yr $^{-1}$), and all sites experienced a period of higher than average slip rate between 0.5 and 2 ky and a period of lower than average slip rate before 3 ky. On the Roccapreturo fault, slip rate in the centre of the fault is 0.550.11 and 0.350.05 mm yr $^{-1}$ at the fault tip near a relay. The estimated time since the last earthquake is the same at each site along the same fault. These results highlight the potential for cosmogenic exposure data to reveal the detailed millennial history of earthquake slip on active normal faults.

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Key Points:

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11	•	Fault slip rates constrained with cosmogenic isotopes from multiple sites on the
12		same fault agree.
13	•	Fault slip rate varies during the Holocene, and pulses of rapid slip rate are tem-
14		porally correlated along strike.
15	•	Cosmogenic isotopes on bedrock fault scarps provide a detailed record of activ-
16		ity, provided the sample locations are carefully selected.

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17 Abstract

Cosmogenic exposure data can be used to calculate time-varying fault slip rates on normal 18 faults with exposed bedrock scarps. However, the method relies on assumptions related to 19 how the scarp is preserved, which should be consistent at multiple locations along the same 20 fault. Previous work commonly relied on cosmogenic data from a single sample locality to 21 determine the slip rate of a fault. Here we show that by applying strict sampling criteria 22 and using geologically informed modelling parameters in a Bayesian-inference Markov chain 23 Monte Carlo method, similar patterns of slip rate changes can be modelled at multiple sites 24 on the same fault. Consequently, cosmogenic data can be used to resolve along-strike fault 25 activity. We present cosmogenic ³⁶Cl concentrations from seven sites on two faults in the 26 Italian Apennines. The average slip rate varies between sites on the Campo Felice Fault 27 $(0.84 \pm 0.23$ to 1.61 ± 0.27 mm yr⁻¹), and all sites experienced a period of higher than 28 average slip rate between 0.5 and 2 ky and a period of lower than average slip rate before 3 29 ky. On the Roccapreturo fault, slip rate in the centre of the fault is 0.55 ± 0.11 and 0.35 ± 0.05 30 mm yr $^{-1}$ at the fault tip near a relay. The estimated time since the last earthquake is the 31 same at each site along the same fault. These results highlight the potential for cosmogenic 32 exposure data to reveal the detailed millennial history of earthquake slip on active normal 33 faults. 34

35 1 Introduction

Fault activity constrained over multiple earthquake cycles and across fault systems helps 36 to address fundamental questions of how faults interact (Nixon et al., 2016; Wedmore et al., 37 2017; Cowie et al., 2017; Mueller, 2017), how tectonic strain accumulates and is released 38 on brittle faults (Hergert & Heibach, 2010; Ferry et al., 2011), and how fault slip varies in 39 time and space (Nicol et al., 2010; Dolan et al., 2016). Fault slip rates can be measured or 40 inferred using a variety of tools, including geodesy (Bendick et al., 2000; Walters et al., 2013; 41 Hussain et al., 2016), palaeoseismology and historical records (Pantosti et al., 1996; Galli 42 et al., 2008; Cinti et al., 2019), and dating of offset geological, geomorphological, and man-43 made features (Phillips et al., 2004; Zechar & Frankel, 2009; Wang et al., 2011; Gregory et 44 al., 2014; R. D. Gold et al., 2017; Mechernich et al., 2018). Each method has different spatial 45 and temporal coverage and resolution, and as a whole provide insight into tectonic processes 46 occurring over a range of scales. Despite the range of techniques, there are still discrepancies 47 between long-term average slip rates and geodetic strain rates, which in part may be due to 48 methodological uncertainties and problems related to the preservation of earthquake surface 49 deformation in the geological or geomorphological record (R. D. Gold et al., 2009; Schmidt 50 et al., 2011; Searle et al., 2011). Individual earthquakes often have incomplete or variable 51 surface displacements along strike (and at depth; Bull et al., 2006; Wesnousky, 2008; 52 Rockwell & Klinger, 2013; P. O. Gold et al., 2013; Ando et al., 2017; Walters et al., 2018), 53 and if patterns of variable displacement persist over multiple earthquake cycles, cumulative 54 Quaternary displacement and slip rate will be different along the fault. 55

By sampling multiple locations along a single fault, it is possible to test the influence of 56 along-strike variation in earthquake slip and preservation on Quaternary slip rate. Bedrock 57 normal fault scarps are excellent targets for investigating along-strike slip variation because 58 they record a more temporally detailed history of progressive fault exposure compared to 59 displaced landforms (Benedetti et al., 2002; Schlagenhauf et al., 2010; Akçar et al., 2012; 60 Cowie et al., 2017: Mechernich et al., 2018). In the Mediterranean, scarps in limestone have 61 been preserved since the Last Glacial Maximum (LGM; Armijo et al., 1992; Giraudi, 1995; 62 Tucker et al., 2011), accruing slip over multiple earthquake cycles. The rate of exhumation 63 of the fault plane by earthquakes can be determined with measurements of the cosmogenic 64 isotope chlorine-36 (36 Cl), which primarily accumulates in the scarp as a result of progressive 65 exposure to cosmic radiation and production from abundant calcium (Ca) present in the 66 limestone footwall (Gosse & Phillips, 2001). A forward model is required to determined 67 fault slip rates and the pattern of exhumation through time. Normal fault scarps have a 68

complex exposure history that starts to develop when the fault is buried several metres
below the surface, and the same profile of ³⁶Cl can result from different earthquake time
and displacement histories as a result of partial exposure to cosmic radiation whilst buried.
Though many previous studies suggest that the timing of individual earthquakes can be
determined by this technique (Benedetti et al., 2002; Schlagenhauf et al., 2010; Akçar et al.,
2012; Benedetti et al., 2013; Tesson et al., 2016; Tesson & Benedetti, 2019), our primary
aim is to determine fault slip rates and slip rate variations.

Cumulative fault slip can vary along strike on an individual fault as a result of (1) the 76 77 natural along-strike displacement profile (Cowie & Shipton, 1998); (2) complexity of fault structure such as overlapping segments (Peacock & Sanderson, 1991); or (3) due to problems 78 in the long-term preservation of displacement as a result of slope instability. In the case of 79 (1) or (2), we would expect the total displacement to vary at different localities, but the 80 timing of major slip rate changes should be temporally correlated along-strike if earthquakes 81 typically rupture the length of the fault. Only one study has attempted to document the 82 synchronicity of along-strike fault slip using cosmogenic isotopes on bedrock fault scarps 83 (Schlagenhauf et al., 2011). They were able to model the data from multiple sites with a similar earthquake history, but only by changing the total amount of time that the scarp 85 had been partially exposed at each site by several thousand years (2.5 ka vs 13.0 ka, termed 86 'pre-exposure'). If this parameter is kept constant between the sites, the data from different 87 sites cannot be modelled with a temporally correlated exposure history, suggesting that 88 the preservation of their sampling sites has been modified (supporting information Figure 89 S1–S2). 90

To demonstrate the reliability of bedrock scarps for preserving earthquake and tectonic 91 process, we present five new ³⁶Cl datasets from the Italian Apennines: three localities on 92 the Campo Felice fault and two on the Roccapreturo fault (Figure 1). We focus on the 93 central Italian Apennines because limestone fault scarps are common in the region and the 94 faults are excellently exposed, well mapped, and easily accessible. There are 19 published 95 ³⁶Cl sample sites in the region (Schlagenhauf et al., 2010; Benedetti et al., 2013; Tesson et 96 al., 2016; Cowie et al., 2017, Figure 1). We also remodel data published by Benedetti et al. 97 (2013) from a site on the Campo Felice fault and data published by Schlagenhauf (2009) 98 from a site on the Roccapreturo fault, in order to directly compare with our new data on the 99 same faults. Our sites were selected on the basis of extensive field reconnaissance, mapping, 100 terrestrial laser scanning (TSL), and remote sensing surveys, in order to ensure that the 101 slip preserved at the surface is the result of earthquake displacements and not affected by 102 hillslope processes. We use a Markov chain Monte Carlo (MCMC) method to model the 103 data at each site, which constrains the timing of slip rate changes to facilitate comparison 104 of sample localities along strike. Our modelling approach incorporates uniform parameters 105 related to early ³⁶Cl production at different sites from the same fault, using the timing of 106 global climatic change as a constraint on how long the fault scarp has been preserved at 107 each of our sampling locations. We show how our approach can be used to determine spatial 108 and temporal variation in earthquake displacement on normal faults. 109

¹¹⁰ 2 Geological background

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2.1 Quaternary faulting in the central Apennines

GNSS measurements indicate that the central Italian Apennines is extending at a rate 112 of 2.7 ± 0.2 mm yr⁻¹ in a NE-SW direction, active since 2–3 Ma (D'Agostino et al., 2011; 113 Roberts & Michetti, 2004, Figure 1). This extension has produced a series of NW-SE 114 trending normal faults that host $>M_w$ 6 surface rupturing earthquakes, which are recorded 115 in both the instrumental and historical records (Rovida et al., 2019; Walters et al., 2018). 116 Average extension rate estimates for the region based on the offset of postglacial slopes by 117 active faults since the Last Glacial Maximum (LGM) are 3.1 ± 0.7 mm yr⁻¹ (Roberts & 118 Michetti, 2004; Faure Walker et al., 2010), in agreement with GNSS rates. The Quaternary 119

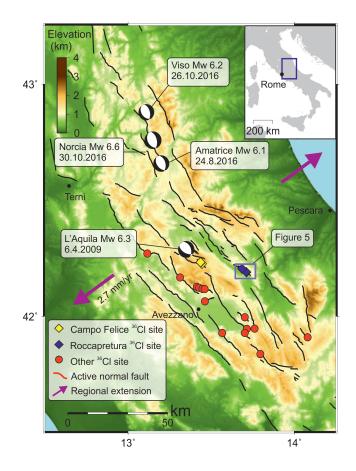


Figure 1. An overview of the central Italian Apennines showing fault scarp sample localities (yellow and blue diamonds). Additional site locations are from Palumbo et al. (2004); Schlagenhauf (2009); Schlagenhauf et al. (2011); Benedetti et al. (2013); Tesson et al. (2016); Cowie et al. (2017), earthquake moment tensors are from www.globalcmt.org, and the fault map is modified from Roberts and Michetti (2004). The regional extension direction indicated is based on D'Agostino et al. (2011). The DEM elevations are from 1 arcsecond (30 m) SRTM (Satellite Radar Topography Mission) data.

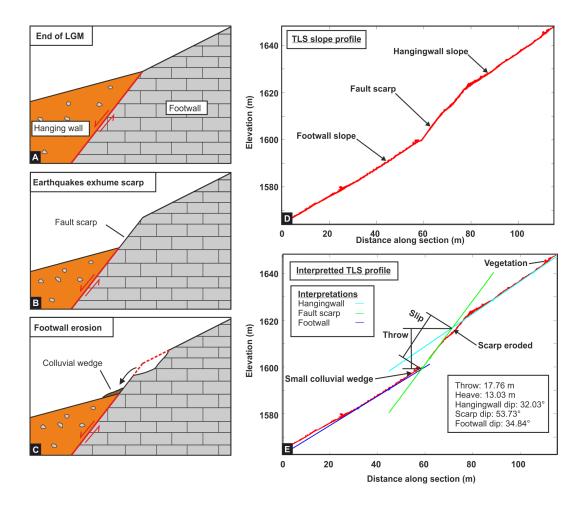


Figure 2. (A) Schematic model of fault scarp development when exhumation rate is slower than erosion rate, as was the case during the LGM in the Central Italian Apennines. (B) and (C) shows a schematic model of scarp evolution since the end of the LGM, when exhumation rate outpaces erosion rate. Panel (D) shows a typical fault-perpendicular profile, from the Campo Felice fault, through a point cloud generated from TLS data and (E) shows one profile interpretation. Samples are not collected from the eroded section of the fault scarp due to the uncertainty in the timing of erosion; only the planar lower section is sampled.

and geodetic extension rates are similar to average extension rates calculated using offset stratigraphy, indicating that the total extension rate across the region may have remained constant for the last 0.75 Ma (Roberts & Michetti, 2004). Time variable fault slip rates and spatio-temporal earthquake clusters have been inferred in the region based on models of ³⁶Cl cosmogenic data (Schlagenhauf et al., 2011; Benedetti et al., 2013; Cowie et al., 2017), and several spatially correlated (along-strike) sequences of large earthquakes have occurred in the modern record.

Planar limestone bedrock fault scarps have been preserved along normal faults since the demise of the LGM (10–20 ka) when the bedrock exhumation rate, normally as a result of fault displacement during earthquakes, exceeded the erosion rate of the fault scarp (Figure 2; e.g. Giraudi, 1995; Tucker et al., 2011; Bubeck et al., 2015; Giraudi et al., 2011). In the central Italian Apennines, fault scarps are observed in Mesozoic limestone, but scarps are poorly preserved where the faults pass into other lithologies, such as siliciclastic turbidite deposits. The preferential formation and preservation of fault scarps is due to the strong

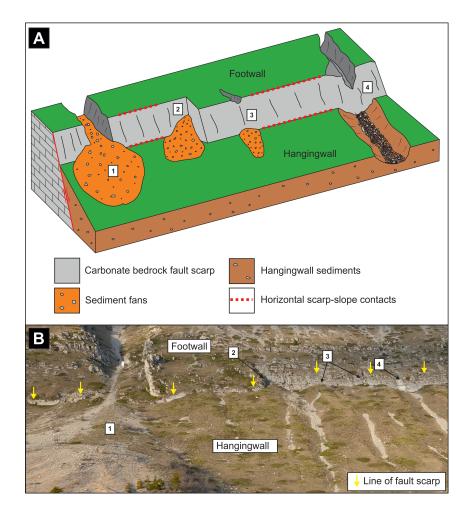


Figure 3. (A) Schematic diagram of slope processes that can lead to non-tectonic exhumation and burial of active limestone fault scarps. Labels 1-4 indicate areas of hangingwall erosion or deposition that are inappropriate sample locations. Ideal sites are located tens of metres away from areas affected by depositional and erosional slope processes, where the scarp-slope contacts are horizontal, after Bubeck et al. (2015). (B) The Campo Felice fault with features from (A) indicated. Photo taken from (42.2308°N, 13.4343°E), view northeast. The horizontal scale is approximately 320 m across image at the height of scarp.

erosional resistance of limestone fault surfaces, and is also well documented in Greece and
western Turkey, which host lithologies similar to central Italy (Goldsworthy & Jackson,
2000; Akçar et al., 2012). Exhumation of bedrock fault scarps in the central Apennines is
not always only due to fault slip in earthquakes. In many areas the footwall and hangingwall
are subject to erosional and depositional processes that are currently active or have been
active since the demise of the LGM. Removal or deposition of material on the hangingwall
and footwall can contribute to the exhumation history of the scarp (Bubeck et al., 2015).

2.2 Fault geomorphology and site descriptions

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We compare slip histories from the multiple sites on two faults: the Campo Felice and Roccapreturo faults (Figures 1 and 4). We sampled three new sites on the Campo Felice fault and two new sites on the Roccapreturo fault, and we also make use of previously published data from one site on each fault (Schlagenhauf, 2009; Benedetti et al., 2013). We describe how the sites were selected, the background literature, and geomorphology relating
 to both faults and all sample sites.

2.3 Sample locality selection

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Sample localities are selected to minimise the impact of post-LGM depositional or 149 aggradational processes acting to expose or bury the planar fault scarp (Figure 3). We 150 follow the criteria set out by Bubeck et al. (2015) and Cowie et al. (2017) to identify suitable 151 cosmogenic sampling sites that have a stable hanging wall and footwall slope. The new sites 152 presented in this study fulfill the following five criteria: 1) the footwall and hangingwall 153 slopes are intact, planar, and show no evidence of incision; 2) the hanging wall slope is free 154 of post LGM sediments (typically associated with actively agrading alluvial fans, colluvial 155 wedges, sloping footwall-hanging wall contacts, and the edge of major drainages); 3) the site 156 is located away from fault relay zones (site RP2 is an exception); 4) the fault plane surface 157 is well preserved; and 5) the contact between the free-face (fault plane) and the hangingwall 158 slope is horizontal, ruling out along-strike mass movement. 159

We identify areas that conform to the first three of these criteria by investigating the 160 contacts between the footwall, the fault scarp, and the hangingwall. Horizontal contacts 161 at a consistent height over a distance of 10 meters or more indicate a lack of significant 162 erosion or deposition since the demise of the LGM (Figure 3). The footwall slope should 163 be smooth and uninterrupted by major drainages in the vicinity of the sample locality. 164 We identify appropriate areas for sampling using a combination of satellite image analysis 165 (Google Earth), interpretation of Terrestrial Laser Scanning (TLS)-derived point clouds, 166 and fieldwork. 167

168 2.3.1 Campo Felice

The Campo Felice fault has a total length of ~ 15 km. It is composed of two overlapping 169 segments, an ~ 6 km southern section striking on average 130° and an ~ 8 km northern 170 section with an average strike of 120° . The Campo Felice fault and the 8 by 3 km basin 171 it bounds have been the focus of several studies (Giaccio et al., 2003; Giraudi et al., 2011; 172 Giraudi, 2012; Benedetti et al., 2013; Wilkinson et al., 2015). The Campo Felice fault may 173 have ruptured during an event in 1300 AD (\sim 720 ybp) based on paleoseismic data (Salvi 174 et al., 2003), but this is not certain as the data are from a paleoseismic trench on a fault 175 segment 5 km north of the main structure, which Salvi et al. (2003) propose is linked to 176 the Campo Felice fault. Pantosti et al. (1996) undertook a paleoseismic study along the 177 Ovindoli-Pezza fault, which is 5–6 km south east of the Campo Felice fault and could be a 178 related structure (at depth). They suggest that earthquakes of M 6.5–7.0 occurred sometime 179 between 700–1140 years ago, likely around 3900 years ago, and between 5300–7000 years ago 180 on the Ovindoli-Pezza fault, and they estimated an average slip rate of $0.9-2.5 \text{ mm yr}^{-1}$. A 181 terrestrial laser scanning (TLS) dataset was collected along the length of the Campo Felice 182 fault by Wilkinson et al. (2015) to investigate the Quaternary activity and geomorphology, 183 and we used these data paired with field reconnaissance to select appropriate sampling 184 localities. 185

The footwall of the Campo Felice fault is characterized by a slope that is affected by 186 the bedding of Upper Jurassic to Upper Cretaceous carbonates. The bedding dips sub-187 perpendicular to the slope dip, and sometimes forms prominent steps in the landscape, 188 but the slope formed during glacial periods is distinct and dips towards the hanging wall 189 basin. The footwall slope has active drainage channels and gullies between $\sim 1-100$ m wide 190 that feed debris fans and gullies in the hanging wall slope. Away from active drainage, 191 the hanging wall and footwall slopes form smooth planar surfaces (Figure 2), similar to the 192 idealised model shown in Figure 3a. The hangingwall slope is composed of an apron of well-193 cemented colluvium, typical of faults in the region. The bedrock fault scarp is generally well 194 exposed and has a morphology typical of normal fault scarps in the region, and the trace of 195

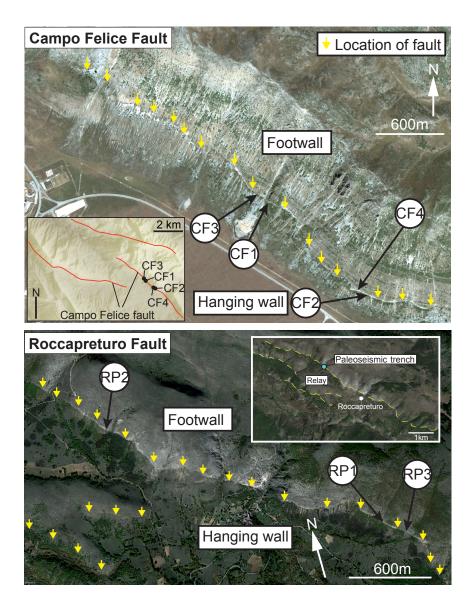


Figure 4. Location of sites used in this study, inset in the top panel shows the length of the Campo Felice fault. Campo Felice sites CF1, CF2, and CF3 were sampled during this study, and site CF4 was sampled and processed by (Benedetti et al., 2013). On the Roccapreturo fault (lower panel), sites RP1 and RP2 were sampled during this study, and site RP3 was sampled and processed by Schlagenhauf (2009). The inset panel shows the relay where two strands of the RP fault overlap. Imagery from Google Earth, 2018.

the scarp is located ~60–200 m above the basin floor. Preservation of the planar fault scarp
varies along strike, becoming more degraded near the fault tips. Further details of each site
location and characterisation can be found in the supporting information, Figures S3–S8;
Tables S1–S5.

There is one pre-existing ³⁶Cl sample site on the Campo Felice fault that was published 200 by Benedetti et al. (2013), to which we will compare our new data and refer to as site CF4. 201 The geomorphology at this sample site is stable and the scarp is well-preserved, as the site 202 characteristics satisfy the criteria used for site selection outlined in our methodology. In this 203 study we present results from three additional Campo Felice sites. Site CF1 was sampled 204 in 2014 and Sites CF2 and CF3 were sampled in 2017. All sites are located on the southern 205 segment of the fault along a ~ 1 km long section (Figure 4). The distribution of sample sites 206 was dictated by the geomorphology of the fault - we can only sample where the site criteria 207 is acceptable and the bedrock scarp is well preserved. 208

2.3.2 Roccapreturo fault

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The Roccapreturo fault is part of the Middle Aterno Valley Fault system (MAVF), which has a total length of 21 km (Galadini & Galli, 2000). The fault is composed of two segments: the southern segment is ~8 km long, and the northern segment is ~3 km long. A 1 km long relay zone separates the two segments, with the distance between the segments varying between 400–900 m (Figure 4 inset). The footwall is characterised by planar slopes incised by gullies up to ~300 m wide. The hangingwall slope is composed of forested colluvium and the bedrock footwall has low density bushy vegetation.

A paleoseismological study of the Roccapreturo fault identified two events based on 217 the offset of stratigraphic layers dated with radiocarbon techniques (Falcucci et al., 2015). 218 The trenches of this study are located $\sim 400-500$ m northwest along strike from site RP2 219 (Figure 4). The most recent event occurred between 1879–2009 BP and 3787–6055 BP 220 and the penultimate event occurred between 3787–4055 BP and 7329–7499 BP (reported 221 by Falcucci et al., 2015, as 2σ age ranges). Falcucci et al. (2015) used the offset of early 222 Pleistocene breccias to calculate a slip rate on the Roccapreturo fault of between 0.23–0.34 223 mm yr^{-1} . The fault has been seismically inactive during the time period covered by the 224 historical record (approximately the past 700 years, Galadini & Galli, 2000). Schlagenhauf 225 (2009) sampled and modeled one ³⁶Cl site (herein referred to as RP3) on the Roccapreturo 226 fault. They find that the scarp did not form in one event, but multiple events of unknown 227 number and magnitude. They suggest that the most recent event occurred approximately 228 2.0-3.0 ka and that the entire scarp was exhumed between 2 ka and 6 ka BP. They report 229 the total offset in the plane of the scarp as 10.2 m and have calculated the average slip rate 230 during the period of exhumation to be 1.7 mm yr^{-1} . The geomorphology of this site does 231 not meet the necessary site selection criteria described in this paper and Cowie et al. (2017), 232 because it is located close to a gully that appears to have contributed to exhumation of the 233 fault scarp (supporting information, Figure S9). We remodel their data using our approach 234 in order to demonstrate the effect of enhanced erosion on cosmogenic data from bedrock 235 fault scarps, comparing it with new data from localities with acceptable morphological 236 characteristics. 237

We present data from two additional sites on the Roccapreture fault, the first (RP1) 238 located 180 m northwest along strike from site RP3 and the second (RP2) a further 2.5 km 239 northwest along strike (Figure 4). Site RP2 is located within the relay zone of the two 240 strands of the fault, and because some deformation may be shared between the overlapping 241 parts of the fault we expect the resulting slip rates to be slower than the central portion 242 of the main strand. However, the timing of slip rates changes (if there are changes) should 243 coincide, as we assume large earthquakes rupture both strands of the fault. Details of site 244 location and characterisation can be found in the supporting information, Figures S10–S13; 245 Tables S1, S6–S7. 246

²⁴⁷ 3 Methods

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3.1 Sample collection and preparation

Limestone fault scarps are composed of fractured limestones with an increase in frac-249 ture density into the fault core where an indurated carbonate fault gouge is present. Where 250 unaffected by erosion, the limestone scarps have planar surfaces with slickensides and stri-251 ations commonly visible on the surface. We use these indicators to identify areas where 252 the fault plane is well preserved, because erosion will destroy fault surface features. We 253 avoid areas of fault plane that are intensely fractured or where the scarp is eroded, as well 254 as areas with obvious secondary precipitation of calcite. We avoid fractures and secondary 255 calcite in an attempt to sample fault rocks that are not contaminated with vadose carbonate 256 cements that might contain cosmogenic chlorine produced in the atmosphere and circulated 257 in groundwater (Dunai, 2010). 258

Sampling involves excavating a trench in the hanging wall against the fault scarp to a 259 depth of 1-2 m. At most sites the density of the excavated colluvium is measured using a 260 simplified version of the method outlined by Muller and Hamilton (1992), because colluvial 261 density is a shielding parameter in the cosmogenic modelling. Discrete samples that are 15 262 cm wide x 5 cm high and 2.5 cm thick are cut from the exposed fault plane using a handheld 263 angle grinder, along a line parallel to the slip vector on the fault (parallel to dip direction at 264 all sites). Some samples are horizontally offset from the main vertical sample line to avoid 265 eroded parts of the fault plane. Photos of each site including the location of the samples on 266 the scarp are shown in the supplementary materials. We collect a 3D point cloud dataset 267 using TLS at each sampling site and extract the geometry of the slip parallel profile of the 268 slope using the Matlab[®] code *crossint* (Figure 2; Wilkinson et al., 2015; Cowie et al., 2017). 269

Sample preparation and measurement is undertaken following standard methods de-270 scribed by Cowie et al. (2017). Chemical sample preparation is conducted at the Leeds 271 University Cosmogenic Isotope Laboratory and prepared samples are measured with the 272 accelerator mass spectrometer (AMS) at the Scottish Universities Environmental Research 273 Centre (SUERC). We report the ³⁶Cl concentration in atoms g^{-1} . Reported 1σ uncertain-274 ties are AMS analytical errors and include propagation of uncertainty based on procedural 275 blanks and standard material measurements. Bulk rock chemistry is constrained by induc-276 tively coupled plasma optical emission spectrometry (ICP-OES) at the University of Leeds. 277 Notably, sample aliquots for Ca weight % measurements must be diluted to ~ 1 ppm Ca for 278 accurate and repeatable measurements. A more detailed description of the sampling and 279 laboratory processes, alongside full results tables enabling recalcuation of ³⁶Cl concentra-280 tions, can be found in the supporting information, Text S1. 281

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3.2 Modelling of the data

Cowie et al. (2017) show that the relationship between ³⁶Cl and height on a fault scarp 283 should approximately scale with the average fault slip rate, such that faster faults have a 284 steeper slope in ³⁶Cl concentration versus height and slower faults have a shallow slope. 285 The concentrations from two sites with a similar slip history (or rate) should approximately 286 overlap, though this may not be precisely true if there is a difference in site geometries or 287 target element abundance (e.g. calcium). ³⁶Cl concentrations from a bedrock scarp do not 288 represent direct exposure ages, because a portion of the total ³⁶Cl in each sample is accu-289 mulated while the fault is partially buried in the shallow sub-surface. In order to model slip 290 rate and the pattern of exhumation through time, we use a modified version of the Bayesian 291 Markov chain Monte Carlo (MCMC) approach developed by Cowie et al. (2017) to explore 292 the age-slip relationships that adequately explain the observed ³⁶Cl measurements within 293 uncertainties (further described later in this section, the supplement, and available online, 294 github.com/lcgregory/SimpleSlips). Bayesian statistical methods are widely applied in 295 earth science and geochronology in order to incorporate prior information about a set of 296 parameters and calculate the posterior distribution for a set of parameters given quantita-297

tive measurements, using a mathematical model (Bronk Ramsey, 2009; Montova-Noguera 298 & Wang, 2017). Bayesian inversions can also be transdimensional, meaning that the num-299 ber of model parameters ('unknowns') for which we solve is allowed to vary, increasing or 300 decreasing the complexity of the model depending on what is required by the data (Green, 301 1995; Sambridge et al., 2006; Bodin & Sambridge, 2009; Dettmer et al., 2010; Amey et al., 302 2019). We use a hyperparameter to vary the number of slip rate changes, which change in 303 time and height, and can be added or removed (varying the number of model parameters), 304 limited by a reversible jump algorithm that favors simple solutions (Sambridge et al., 2006). 305 Bayesian techniques are often applied to deal with uncertainty associated with limited data 306 (Bronk Ramsey, 2009; Montova-Noguera & Wang, 2017; Amey et al., 2019). Several differ-307 ent Bayesian MCMC approaches have been developed for modelling cosmogenic data from 308 fault scarps (Beck et al., 2018; Tesson & Benedetti, 2019; Tikhomirov et al., 2011). In this 309 study, we prefer the modified version of the approach in Cowie et al. (2017) because our 310 primary aim is to identify and compare first-order variations in fault slip rate. Whilst the 311 code used here does not change some of the factors affecting production (attenuation depth 312 and colluvial density), this code has fewer parameters than other available codes, and does 313 not attempt to identify individual earthquakes, which fits within the limitations of our data. 314

The MCMC code relies on the modified version of the Matlab[®] code from Schlagenhauf 315 et al. (2010) to forward model the ³⁶Cl concentration. The forward model simulates ex-316 humation of a normal fault plane and calculates the resulting ³⁶Cl concentrations including 317 corrections for parameters such as site geometry, sample composition, and cosmogenic par-318 ticle flux. We employ a time varying cosmogenic particle flux derived for each site using the 319 most recent cosmogenic calculator CRONUS-2 described by Marrero et al. (2016). Further 320 details on site-specific production rate scaling are included in the supporting information, 321 Table S1. Previous studies use a constant value for colluvium density at each site (Cowie et 322 al., 2017), and, given the poorly quantified uncertainties associated with determining mean 323 colluviual density in the field, and the agreement of our measured values with average values 324 previously determined, we also use a mean value of 1.5 g cm^{-3} . 325

At each sampling locality, the height of the preserved fault scarp is known (from TLS 326 observations, Figure 2). There is a gap between the highest sample and the top of the scarp, 327 because the top part has been subject to weathering processes for longer than the base and 328 is poorly preserved. The exposure history of the unsampled portion of the scarp is modelled 329 by assuming that the scarp has been preserved since the demise of the LGM. The MCMC 330 algorithm explores a trans-dimensional parameter space, solving for both slip rate and the 331 number and timing of changes in slip rate. A slip history is generated with parameters 332 conditioned on the prior probability, to calculate a forward model of 36 Cl values for this 333 slip history. The likelihood of the proposed slip history is calculated given the comparison 334 between the modelled ³⁶Cl values relative to the measured data. The algorithm then varies 335 one of the parameters used to define the slip history and runs the forward model again. 336 The new slip history is accepted if it has a higher likelihood than the previous model or if 337 the ratio of new/current likelihood is higher than a random number drawn from a uniform 338 distribution between 0 and 1, otherwise the new model is rejected, as per the Metropolis 339 Hastings algorithm (Metropolis et al., 1953; Hastings, 1970). We run this for 500k iterations. 340

The model parameters for which we solve to define a slip history are: 1) scarp age (SA, 341 time of the first event that produced preserved fault scarp, with a 1σ normally distributed 342 prior of 15 ± 3 kyr), 2) elapsed time (ET, time since last earthquake, no prior unless something 343 is known about the most recent earthquake), 3) timing of change points (timing of change in 344 slip rate), 4) height on fault scarp of a change point and 5) a hyper-parameter, the number 345 of change points. The slip rate between change points is kept constant. The actual number 346 of parameters can vary between each iteration, dependent on how many change points are 347 defined. In each iteration we make a small change to one of the parameters. Further details, 348 including synthetic tests and examples from other faults, can be found in Cowie et al. (2017) 349 and online. We use the flexible change point method of Cowie et al. (2017) rather than the 350

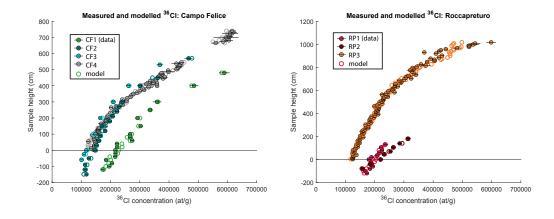


Figure 5. ³⁶Cl concentration versus sample height in the plane of slip. Each filled point represents a different sample, and each suite of coloured points represents data from one site as indicated. Open circles show the modelled ³⁶Cl from the MAP (maxium a posterior estimation). The colour scheme for each site is kept uniform in the figures that follow (CF4 is navy in following figures). Data from Campo Felice Site CF4 are from (Benedetti et al., 2013), and data from Roccapreturo Site RP3 are from Schlagenhauf (2009). Analytical 1 σ uncertainties are plotted as black lines. These data are not corrected for calcium concentration, and some variation in ³⁶Cl between sites and noise in sample data is related to different production rates resulting from variable Ca.

fixed change point model (where the change point height up the fault scarp is fixed) because 351 we have no additional data such as fault roughness to fix the height of the changes in slip 352 rate up the scarp. The flexible change point model allows timing and number of changes in 353 slip rate to vary between iterations, whilst the reversible-jump transdimensional algorithm 354 naturally favours simpler models with fewer change points, potentially resolving the issue 355 of over-fitting the data (Sambridge et al., 2006). The Bayesian MCMC algorithm results 356 in a distribution of possible slip histories and their likelihood and misfit to the data. We 357 then calculate the posterior probability by multipling the likelihood by the prior. We use a 358 constant slip size of approximately 1 m to exhume the scarp incrementally in our modelling 359 as we find that using a smaller constant slip size has little effect on the overall model 360 results but does make the inversion process more computationally expensive (supporting 361 information Figure S14). If we were attempting to model individual events then the choice 362 of slip size would have to be further considered. We also run a suite of models at different 363 constant slip rates, to determine whether a simple exposure history can adequately explain 364 the data and fit the LGM hypothesis (supporting information, figures S15–S16). 365

366 4 Results

³⁶Cl data are plotted as cosmogenic isotope concentration versus sample height on 367 Figure 5 for the Campo Felice and Roccapreturo faults. At each site, the ³⁶Cl concentration 368 increases gradually with increasing height, due to higher parts of the scarp being exposed 369 for longer. In general, data from different sites on each fault overlap, and the slope of 370 36 Cl concentration versus height is related to the slip rate on the fault, such that a steeper 371 gradient indicates faster slip rates. The concentration at the height of zero is indicative 372 of both the fault slip rate, and the time elapsed since the last earthquake, because the 373 buried portion of the scarp has accumulated ³⁶Cl during the time since the last earthquake, 374 such that a greater 36 Cl concentration at height 0 and in the trench can indicate a longer 375 elapsed time. ³⁶Cl concentration profiles are similar for sites CF2, CF3, and CF4 whilst 376

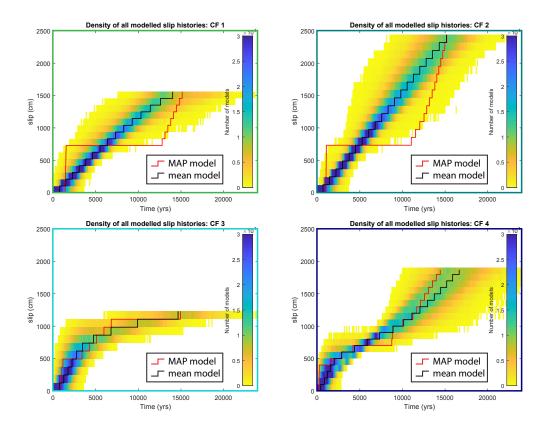


Figure 6. Modelling results from the Campo Felice fault ³⁶Cl data. Each sub-figure includes 500k iterations, minus the burn in of 50,000 iterations, as a 2D histogram showing the distribution of accepted slip histories in time-slip space. The distribution reflects the density of overlapping models, but does not capture the pattern of any individual slip history. The red line is the maximum a posteriori probability (MAP) estimation model, which is the maximum likelihood multiplied by the prior probability based on scarp age. The black line represents the mean model - which is the mean time for each slip step.

concentrations are greater at site CF1 for samples from the same height. All profiles show 377 a change in gradient at ~ 3 m on the Campo Felice fault. On the Roccapreturo fault, site 378 RP3 has lower ³⁶Cl concentrations for the same height than at sites RP1 and RP2, and 379 has a steeper gradient at the base of the scarp compared to sites RP1 and RP2. The 380 gradient at site RP3 gradually reduces with height. Sites RP1 and RP2 have similar ³⁶Cl 381 concentrations, but with minor differences in gradient and the concentration at height 0. 382 Site RP3 samples a section of preserved scarp that has an offset of 10.2 m, compared to 383 sites RP1 and RP2 which have offsets of 7.2 m and 4.7 m respectively; this difference in 384 heights is discussed in the context of the site geomorphology and strain partitioning in the 385 discussion. Here we present the modelling results from each cosmogenic sampling site in the 386 context of each fault and how the results can be compared between sites on the same fault. 387

388 4.1 Campo Felice fault

Each site was modelled with 500k iterations using the Bayesian inference MCMC code described in the methods section, with the site characteristics listed in Table S1, and modelling parameters in Table S8 (Cowie et al., 2017, https://github.com/lcgregory/

SimpleSlips). 2 dimensional histograms of all accepted exhumation models for each site 392 with the burn in removed are shown in Figure 6. A burn-in of 50,000 iterations is removed 393 from each set of results, because the initial models are affected by the starting parameter 394 values. The histograms show the modelled distribution of time at which the fault surface 395 was first exposed to the surface. The slip is modeled in approximately 1 meter increments 396 in the slip direction and is binned into 200 year intervals in the histograms. In order to 397 compare between sites, we plot the 95 percentile range of these same exhumation histo-398 ries for all sites (Figure 7a). The models are poorly constrained above 7-10 m due to the 399 lack of samples on the degraded part of the scarp, demonstrated by the increased variance 400 between model results higher on the scarp. The 36 Cl concentration in each sample does 401 reflect exposure of the fault surface at least 2 m above the sample due to being partially 402 exposed to cosmic radiation whilst residing below the ground surface. As such, exposure of 403 the un-sampled portion of the fault is somewhat constrained by the cosmogenic data at the 404 top of the sampled portion, as well as by the independent prior in our modelling dictating 405 that the top part of the scarp was preserved following the demise of the LGM (15 kyr with 406 a 1 σ standard deviation of 3 ka). However, the older portions of the slip models have more 407 variability in exposure time, primarily due to the range in predicted ages for the demise of 408 the LGM and preservation of the scarp (Figure 6). We rank the models by the posterior 409 probability, which is the likelihood multiplied by the scarp age prior probability. We then 410 select the top 10% most probable models to have the same distribution of the number of 411 change points in the full distribution. The top 10% most probable models are used to cal-412 culate the average slip rate over time (Figure 7e), and the fit of CF1 models is shown in 413 Figure 8a with the corresponding exposure histories. We also show the probability of events 414 over time for the full model distribution in Figure 7c. Figure S17 shows the fit to the data 415 of the full model distribution for each site on the Campo Felice fault. In general, the range 416 of accepted models fit the data well (and all fit within the standard deviation of the data), 417 though the top samples are poorly fit to the analytical errors by the least likely models. 418

For each site, we calculate the time-varying fault slip rate in mm yr^{-1} for the models 419 that are the top 10% most probable (Figure 7e). Each model is a relatively simple time-slip 420 vector, with a constant slip rate between the elapsed time, each change point, and the scarp 421 age. The slip rate through time is the slope of each portion of each model at any given 422 time. In order to average the variability across the accepted models, mean slip rates are 423 calculated in 1 year increments (Figure 7e). Because one of the modeled parameters is the 424 time elapsed since the last earthquake, each model has a period of time between the present 425 day and the last proposed earthquake during which the incremental slip rate is zero. If 426 another earthquake occurred today, the mean slip rate between the present day and what 427 would then be the penultimate earthquake would change to accommodate the 'new' slip, 428 but modeled slip rates previous to the penultimate event would remain the same. Therefore, 429 the apparent drop to zero mm yr^{-1} in our slip rate calculations reflects the modeled elapsed 430 time, and does not imply that the fault is inactive – an important consideration if time-431 varying fault slip rates are to be incorporated into earthquake hazard assessment. Because 432 slip rate is calculated as mean of all of the models, we only show the rate up to 10 ka; older 433 than 10 ka the slip rate is poorly constrained where the models 'end' at the scarp age and 434 the mean is not representative. 435

In general, the modelling results show agreement in exposure histories between the 436 sites (Figures 6 and 7a,c,e). Sites CF1 and CF3 are located within 150 m of each other, 437 as are CF2 and CF4, and there is approximately 1 km between both sets of sites (Figure 438 4). There is a difference in scarp height between the group of northwestern sites (CF1 and 439 CF3, average height 14 m) and the southeastern sites (CF2 and CF4, average height 21.5 440 m), but the height change does not lead to a significant difference in slip rate and the timing 441 of change in slip rate during the past 0-7 ka, which is the best constrained time interval, 442 because the ³⁶Cl concentrations at each site are all fit by an increase in slip rate during the 443 same time interval. Models of sites CF1–CF3 have peak slip rates of 1.5-3 mm yr⁻¹ between 444 500 and 2000 years, with a reduced slip rate of <1.5 mm yr⁻¹ before approximately 3–4 ka 445

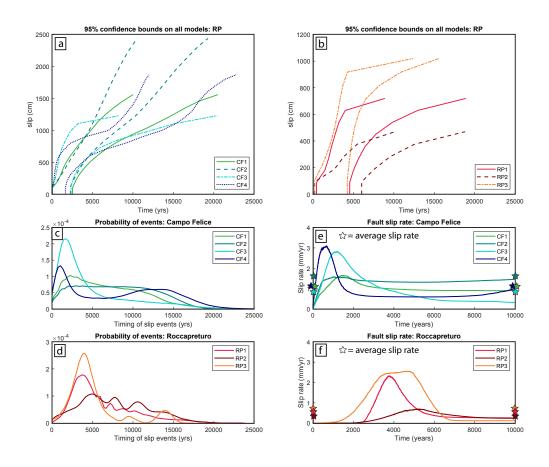


Figure 7. Summary of results from the Campo Felice and Roccapreturo faults. Panels (a) and (b) show the 95% confidence bounds on models from each site minus the burn in. Panels (c) and (d) show the probability distribution of slip events over time (for the full model results). Panels (e) and (f) show the average slip rate for the top 10% most probable models over time, in 1 year bins, and stars indicate the average slip rate calculated based on the median scarp age and scarp height.

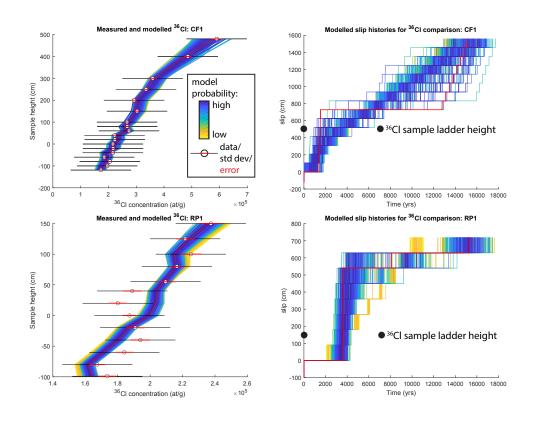


Figure 8. Figures (a) and (c) show the fit to data of models used to calculate average fault slip rates, for sites CF1 and RP1. The circles and error bars represent the ³⁶Cl measurements and the standard deviation of the data (used in the likelihood calculation); each colored line represents a model, with dark to light colours representing highest to lowest probability models, regulated by the scarp age probability and the number of change points. We present 400 models for each site ranging from the highest (dark blue) to lowest (yellow) probability at equal intervals (100) through the distribution. Figures (b) and (d) represent the corresponding model slip histories. The full distributions from the inversion for each site can be found in the Supplementary Materials.

(Figure 7e). Models of site C4 have a higher peak slip rate of just over 3 mm yr^{-1} occurring 446 more recently than at sites 1-3 (around 0.5-1 ka), reflecting the lower ³⁶Cl concentration 447 and steeper ³⁶Cl vs height at this site, requiring a faster slip rate more recently. Models of 448 sites CF2 and CF4 have a second longer period of increased slip rate between the demise of 449 the LGM and ~ 8 ka, though this part of the exposure history is not well resolved, based on 450 the spread of model results at the top of the scarp in Figure 6a-d. The results from all sites 451 on the Campo Felice fault (Figure 7c) indicate that the fault was relatively active between 452 1-4 ka and relatively less active between 4-8 ka. The fault at sites CF1 and CF3 likely has 453 less total slip in each event, compared to at sites CF2 and CF4, because the total scarp 454 height is lower. Despite having less total slip, the timing of peak slip rate and rate change 455 is correlated between all sites (Figure 7c,e). 456

4.2 Roccapreturo Fault

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The modelling results from the Roccapreturo fault, including 2D histograms of slip 458 versus time and mean slip rate through time, are shown in Figure 9, and the fits to the data 459 of the 10% most probable models are shown in Figure 8. Site RP1 is located in the centre of 460 the main fault strand, and RP2 is located on a relay between the main fault and a northern 461 strand (Figure 4). Site RP3 was sampled by Schlagenhauf (2009) and is located near the 462 southeastern tip of the main strand, on the edge of an active gulley (supporting information, 463 Figure S9). The fault slip rates in Figure 7f are calculated for the 10% most probable models, 464 and the probability of events over time is calculated for the full model distribution (Figure 465 7d). The general pattern of exhumation is characterised by relatively faster slip rate between 466 2-6 kyr, and slow to zero slip between 6 kyr and the demise of the LGM, and between the 467 present to 2 kyr, which implies a long elapsed time. The maximum slip rates are 2.2, 0.7, and 468 2.5 mm yr^{-1} at sites RP1, RP2, and RP3 respectively. The difference in average slip rate 469 between sites is primarily due to the difference in scarp height, as a larger scarp requires a 470 faster rate of exhumation averaged over the time period. The decrease from fast to slow slip 471 rate occurs at the same time at sites RP1 and RP2 (2.5–3 kyr), but RP3 has younger and 472 more rapid peak in slip that lasts until 1.5-2 kyr, required by the much lower overall 36 Cl 473 concentration and higher scarp height at the site (Figures 5 and 9). The modelling provides 474 a good fit to the data at sites RP2 and RP3, but data just above the ground surface at RP1 475 are poorly fit (Figures 8 and S18). The pattern of these outlier data is not systematic, and 476 suggests that there is additional noise that is not accounted for in some samples. 477

478 5 Discussion

Accurate fault slip rates derived from cosmogenic isotopes measured on bedrock fault 479 scarps can contribute to our understanding of fault behavior over multiple earthquake cycles 480 and should be considered when estimating seismic hazard (Benedetti et al., 2002; Schlagen-481 hauf et al., 2010; Akçar et al., 2012; Cowie et al., 2017; Beck et al., 2018). However, until 482 now it has not been demonstrated that results are consistent at different sites along strike on 483 the same fault. Here we show that the timing of slip rate changes is similar at different sites 484 along strike on the Campo Felice and Roccapreturo faults, but there are some differences in 485 slip rate and total displacement between sites due to multiple factors. We discuss how the 486 modelling parameters can be compared between sites, and highlight the assumptions and 487 limitations of the ³⁶Cl method. We outline how results from sites with acceptable indicators 488 of morphological preservation can be used to infer that both spatial (e.g. along strike on the 489 fault) and temporal (changes in slip rate) variability is resolved on millennial timescales. We 490 also compare our results with those from paleoseismic trenching on the same fault, which 491 further supports the ability of cosmogenic isotopes measured on bedrock fault scarps to 492 provide a reliable measure of fault activity. 493

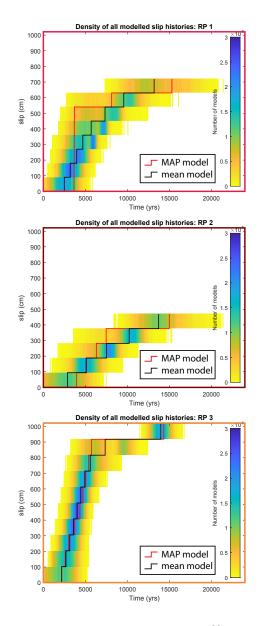


Figure 9. Modelling results from the Roccapreturo fault ³⁶Cl data. Each sub-figure includes 500k iterations, minus the burn in of 50,000 iterations, as a 2D histogram showing the distribution of accepted slip histories in time-slip space. The distribution reflects the density of overlapping models, but does not capture the pattern of any individual slip history. The red line is the maximum a posteriori probability (MAP) estimation model, which is the maximum likelihood multiplied by the prior probability based on scarp age. The black line represents the mean model - or the mean time for each slip step.

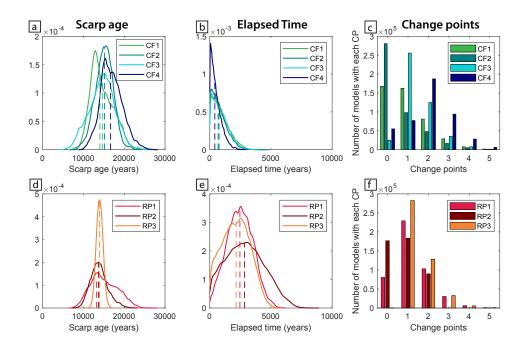


Figure 10. Posterior probability distribution functions for scarp age (a, c) and elapsed time parameters (b, e), for each fault with each site plotted as different coloured lines. The vertical dashed lines correspond to the median of each distribution. The number of change points (number of times the slip rate changes in each model) for each site is shown in c and f.

5.1 Along-strike comparison of fault activity

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The magnitude of surface displacement in individual earthquakes can vary along-strike 495 (Wesnousky, 2008; Rockwell & Klinger, 2013; P. O. Gold et al., 2013; Ando et al., 2017), 496 and as the result the pattern of cumulative slip on a fault may be temporally synchronous 497 but spatially variable in slip magnitude. Cosmogenic analyses on bedrock scarps provide 498 constraints on both the time and cumulative displacement, so data from multiple sites can 499 be used to isolate the spatial variation in slip along-strike over multiple earthquake cycles. 500 Previous ³⁶Cl studies on bedrock normal scarps have concluded that significant temporal 501 slip rate variations occur on thousand year time scales (Cowie et al., 2017; Schlagenhauf 502 et al., 2010). Faults are demonstrated to have intervals of relatively fast slip rate or short 503 earthquake recurrence intervals, interspersed with periods of relative quiescence. Changes 504 in slip rate are probably linked to elastic interactions or strain partitioning processes that 505 are larger in scale than a single fault (Cowie et al., 2012; Dolan et al., 2016) and, therefore, 506 the timing of significant slip rate changes are likely to be temporally correlated along one 507 fault. 508

We compare the posterior probabilities of the time of scarp preservation and the time 509 since the most recent earthquake, the number of change points in each model run, and the 510 average fault slip rate and timing of 'events' between the different sites on the same fault 511 (Figure 7 and 10). One of the greatest uncertainties in modelling the cosmogenic data is the 512 timing of preservation of the fault scarps, associated with the demise of the LGM and the 513 transition from relatively fast to slow erosion of the scarp. We assign a wide Gaussian prior 514 in our modelling to account for the uncertainty in how long it takes for fault activity to 515 outpace erosion, and that this transition may be different for different faults. Figures 10a,d 516 show the similarity in the posterior probability for scarp age at each site. We modelled the 517

results from the Schlagenhauf et al. (2011) study using the same approach, and the scarp age posterior probabilities do not overlap (Figure S2), suggesting that morphological factors not associated with the LGM have affected the development of scarps at those sites.

Using the total displacement measured at each site and the median value for each scarp 521 age posterior probability, we compute the Holocene average slip rate for each site (Table 522 1). The median represents the midpoint of the distributions, and is not affected as much 523 by a small number of very large or small outliers as the mean value. We report the median 524 and standard deviation of all sites based on the median of the combined posterior values 525 for scarp age (divided by scarp height) of all sites. The median slip rate over the time since 526 the demise of the LGM is $1.15 \pm 0.36 \text{ mm yr}^{-1}$ at Campo Felice based on all four sites and 527 $0.42 \pm 0.14 \text{ mm yr}^{-1}$ at Roccapreturo, based on sites RP1 and RP2. 528

The elapsed time parameter only has a positivity prior value assigned in the modelling, 529 because there are no paleoseismic trench sites within 2 km of our sites on the Campo Felice 530 fault and there is no historical seismicity associated with either fault. The Campo Felice 531 sites have a similar posterior probability distribution favouring an elapsed time of less than 532 800 years, with a non-normal distribution that is skewed towards younger values (median 533 values range between 411–771 yrs, Figure 10b, Table 1). Paleoseismic data from trenches 534 north and south of our site on adjacent fault strands indicate that the most recent surface 535 rupturing event on the Campo Felice fault was 720 years (north segment) and between 536 700–1140 years (southeast segment - the Ovindoli-Pezza fault; Salvi et al., 2003; Pantosti 537 et al., 1996). These results are in agreement with our estimated elapsed time, and they 538 suggest that large earthquakes on the Campo Felice fault may involve multiple strands in 539 the same earthquake, or sequences of events that occur over a relatively short time period, 540 similar to several modern sequences in the region such as the 2016 central Italian sequence 541 (Chiaraluce et al., 2017; Villani et al., 2018; Walters et al., 2018). Based on the interpreted 542 surface displacements of 2-3 m, Pantosti et al. (1996) suggest that the causative event was 543 a M 6.5–7.0, which is reasonable for a combined fault length of 35 km if all of the sampled 544 segments were involved in one event. Pantosti et al. (1996) estimated that the average slip 545 rate of the Campo Felice fault is 0.9-2.5 mm yr⁻¹, on the basis of multiple events occurring 546 over the past 5300–7000 years, which also fits well with our long-term average fault slip rate 547 $(1.15 \pm 0.36 \text{ mm yr}^{-1}).$ 548

The Roccapreture sites have a broad distribution of elapsed time values, with the 549 median values at RP1 and RP2 in agreement (median = 2.6 ± 1.4 kyr, Figure 10e, Table 550 1), and a younger preferred value for site RP3 (median = 2.1 ± 1.1 kyr). These results 551 agree with paleoseismic data that suggest the most recent event on the Roccapreturo fault 552 was between 2–6 ka, with another large event occurring between 3.8–7.5 ka (Falcucci et al., 553 2015). These dates agree with the rapid slip rate between 2–7 ka at site RP2, which is located 554 approximately 500 m from the paleoseismic trenches (Figures 4 and 7d,f). Whilst traditional 555 paleoseismic data have been compared to ³⁶Cl slip histories in previous studies, these have 556 either been on different fault strands at distances of >5 km (Tesson et al., 2016) or have 557 suggested disparate results (Benedetti et al., 2003; Kokkalas et al., 2007). The agreement 558 we find between the two techniques, which have been applied in such close proximity on 559 the Roccapreturo fault, provides further evidence for the reliability of slip histories derived 560 from modelling of ³⁶Cl on bedrock fault scarps, and the potential for these two techniques 561 to be combined for more informed seismic hazard analysis. 562

There are many assumptions that must be taken into account when interpreting cosmo-563 genic data from bedrock fault scarps. Because the top of the fault scarp is not well preserved 564 and cannot be sampled, exposure histories older than $\sim 8-10$ ka are poorly resolved and the 565 566 modelling is reliant on estimates of how the scarp is preserved through the demise of the LGM (Figures 6 and 9 Schlagenhauf et al., 2011; Benedetti et al., 2013; Mechernich et al., 567 2018; Beck et al., 2018; Tesson & Benedetti, 2019). The trans-dimensional nature of the 568 Bayesian inversion favors simple slip histories with the lowest number of changes in slip rate 569 and we do not apply any weighting to the data other than the standard deviation of each 570

Site name	Elapsed time (ET, yrs)	$\begin{array}{c} \text{ET std}^a \\ \text{(yrs)} \end{array}$	Scarp age (SA, yrs)	SA std (yrs)	$\begin{array}{c} \text{Slip rate}^b \\ (\text{mm yr}^{-1}) \end{array}$	$\frac{\text{SR MAD}}{(\text{mm yr}^{-1})}$
Campo Felice Fe	ault					
CF1	771	684	14024	2841	1.11	0.21
CF2	678	619	15155	2238	1.61	0.27
CF3	476	634	14656	3139	0.84	0.23
CF4	411	445	16650	2688	1.12	0.18
$\mathbf{CF} \ \mathbf{median}^c$	631	620	15209	2909	1.15	0.36
Roccapreturo Fa	ult					
RP1	2461	1065	13162	2714	0.55	0.11
RP2	2891	1565	13856	2159	0.35	0.05
RP3	2140	1119	13960	825	0.73	0.04
$\mathbf{RP} \ \mathbf{median}^{c,d}$	2603	1355	13500	2460	0.42	0.14

Table 1. Mean scarp age, elapsed time, and fault slip rates

 a Standard deviation.

 b Holocene average slip rate calculated based on the height of the scarp divided by the scarp age pdf.

^cTotal median slip rate is calculated using the average scarp height divided by the pdf of scarp age at each site, and stacking the slip rate pdf. d Calculated using only sites RP1 and RP2.

571 data set. The sampling bias is a challenge for calculating fault slip rates, because of the higher sample density at the base of the scarp. Consequently, the inversion favors simple slip 572 histories that fit the data well in the bottom section of the scarp where there is a higher den-573 sity of data, and that may fit less densely sampled data further up the scarp poorly. Models 574 can fit the data with a more simple slip history by not fitting the top few data points as well 575 as the data at the base of the scarp. If the oldest part of the exposure history can be better 576 quantified, perhaps by incorporating more sophisticated geomorphological models and data 577 constraining the timing of the LGM (e.g. Tucker et al., 2020), the entire slip history may 578 be better determined. 579

At some sites no models fit the data to within the analytical uncertainties because they 580 have outliers or noisy data that are not fit by any model. Applying site averaged calcium 581 values at the two sites where we did not collect the data ourselves reduces the ability 582 of models to fit the data because small variations in Ca concentration has a significant 583 effect on the production rate of 36 Cl in each sample. One challenge in interpreting the 584 output of MCMC Bayesian modelling is that, whilst there is a single best fit or most likely 585 model, there are commonly hundreds or thousands of models that fit the data almost as 586 well (Figure 8). After removing the burn-in, all of the models fit within the standard 587 deviation of each data set, and can be incorporated when calculating average slip rates 588 and making broad interpretations. Identifying higher frequency variations or individual slip 589 events (earthquakes) is challenging because the data can be fit with a range of models, 590 and is not possible using the data and modelling methods in this study. However, the first 591 order variations in slip rate including pulses of rapid slip rate, which may represent temporal 592 clustering of earthquakes, are consistent features in results from multiple sites along the same 593 fault. The Bayesian MCMC approach with minimal parameters ensures that cosmogenic 594 data are not overfit, and the result is an acceptable range of exposure histories, rather than 595 a non-unique earthquake history. 596

Based on the results presented here and the large time and financial costs associated with sample processing and ³⁶Cl measurements, future studies may benefit from sampling

multiple sites with discrete spaced samples rather than a continuous sample ladder at one 599 sample site. The multi-site sampling approach also allows more information to be gained 600 on along strike variability of slip rates. The geomorphology of each sample site should be 601 carefully understood and documented to demonstrate the tectonic origin of bedrock fault 602 scarps. Sampling at regular intervals on the fault scarp limits sampling bias and can reduce 603 the complexity of interpreting modeled slip rates. While the prior that scarps are preserved 604 only since the demise of the LGM is strongly supported in the Central Italian Apennines 605 (Galadini et al., 2003; Tucker et al., 2011), application of the method to other regions will 606 require equally robust evidence to define the scarp age prior distribution. Combining other 607 data sources with the ³⁶Cl data, such as historical records and estimates from other dating 608 techniques, helps to support results from cosmogenic data. 609

5.2 Temporal slip rate variability

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Temporal slip rate variability is observed at all of the sites on the Campo Felice and 611 Roccapreturo faults (Figure 10c,f). Both faults experience pulses of relatively fast slip 612 rate (over thousands of years), peaking at 3 mm yr⁻¹ at Campo Felice and 2 mm yr⁻¹ at 613 Roccapreturo, separated by intervals relative quiescence with slower average slip rate. Fault 614 slip rate variability or discrepancies between geodetic, Quaternary, and geological slip rates 615 is observed on many faults in various tectonic settings (Papanikolaou et al., 2005; Oskin et 616 al., 2007; Faure Walker et al., 2010; Ferry et al., 2011; Dolan et al., 2016; Zinke et al., 2017), 617 with several mechanisms invoked to explain the variability. Orogen-scale changes in erosion 618 patterns or the kinematics, growth, and localisation of faulting may affect the comparison 619 of geological rates $(> 10^5)$ with geodetic and Quaternary rates (Hoth et al., 2006; Nicol et 620 al., 2010). In the Italian Apennines, Cowie et al. (2017) suggest that time variable slip rates 621 are primarily caused by large scale interaction across the whole fault network, in order to 622 minimize the work done by faults. In this geodynamic model, different regions of faults are 623 active at different times as a result of the change in gravitational potential energy acting 624 on the uplifted footwall, inducing flexural bending of the normal fault footwall and time 625 varying fault strength. Coulomb stress changes due to earthquakes are suggested to play a 626 role in causing clustering of earthquakes and variable slip rates (Dolan et al., 2016; Wedmore 627 et al., 2017). Dolan and Meade (2017) indicate that there is not yet a single mechanism 628 that can explain this behavior across different faults, and suggest that it is caused by the 629 complex interaction of processes that may be controlled by properties of a particular fault 630 as well as the fault system as a whole. We observe peak slip rates at different times on 631 the Campo Felice and Roccapreturo faults (Figure 10c,e), suggesting that fault activity 632 migrates spatially over time between the two relatively close structures, possibly related to 633 stress interaction (Figure 1). 634

In order to understand the mechanism behind slip rate variability on a single fault, it 635 may also be informative to constrain the activity of faults in the rest of the network based 636 on observations over multiple timescales. Probabilistic seismic hazard models currently 637 use time averaged constant slip rates on faults (Valentini et al., 2017), and have limited 638 temporal and spatial data coverage due to the sparsity of paleoseismic data sets (Dolan et 639 al., 2016). What can be inferred from ³⁶Cl data on bedrock scarps is also limited in time, 640 but we are able to capture 2 major changes in slip rate at some sites, helping to better 641 understand the variability of earthquake recurrence on timescales that are important for 642 understanding fundamental geological problems and seismic hazard. The method can be 643 widely applied where scarps are preserved to reveal fault interaction on kilo-year timescales 644 and to determine how several faults contribute to the large scale pattern of deformation. 645 Slip rate variability may also be captured by quantifying slip rates using alternative methods 646 647 that have different spatial and temporal coverage and resolution. Faure Walker et al. (2012) show that slip rates averaged over the Holocene (based on fault scarp heights) match the 648 geodetic deformation rates, when averaged over large spatial scales (10^2 km) . Cowie et al. 649 (2013) suggest that the 10^4 year strain rates are representative of long-term geological rates 650 based on the correlation between high strain and high topography, suggesting that faulting is 651

driven by viscous flow on localised shear zones in the lower crust. ³⁶Cl derived slip histories have the potential to fill some of these spatial and temporal gaps and will help to elucidate the timing and mechanisms responsible for earthquake clustering and fault interaction.

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5.3 Spatial fault complexity

The agreement between results from the Campo Felice Fault demonstrates that ${}^{36}Cl$ 656 data from multiple sites spaced ≤ 1 km on one fault can be modeled successfully with similar 657 slip histories. The larger fault scarp and a period additional slip between 7 ka and 15 ka 658 only observed at the 2 southern most sites (sites CF2 and CF4, Figure 7c,e) may suggest 659 that the fault does not always rupture continuously or uniformly along strike, which matches 660 modern observations of faults in the region (Boncio et al., 2004; Walters et al., 2018; Villani 661 et al., 2018). Sites CF1 and CF3 are closer to the overlap between the central and northwest 662 Campo Felice fault strands (Figure 4), and it may be expected that total displacement is 663 less at that location in each earthquake. Although Benedetti et al. (2013) determined an 664 exposure history at site CF4 similar to our results, with 2 earthquakes at 1.1 ka, and events 665 at 3.4, 4.2, 4.4, and 9.4 ka, their solution is non-unique, as there are many other exposure 666 history that fit the data at CF4 equally well (Figure 6). The continuous ladder at CF4 leads 667 to tighter constraint on parameters compared to sites CF1-CF3 (Figure 10). Models at 668 site CF4 also include more change points than CF1-3 (Figure 10c), suggesting reliable more 669 complex models (but not precise earthquake timings) can be generated from densely spaced 670 data. This agrees with synthetic tests in Beck et al. (2018), which show that continuous 671 sampling of the fault scarp does not necessarily resolve better constraints on absolute slip 672 rates and the timing of change in slip rate compared to discrete sampling every 25–50 cm. 673

Modelled slip histories from sites on the Roccapreturo fault are not as similar as results 674 from the Campo Felice fault. Site RP3, sampled by Schlagenhauf (2009), has a significantly 675 larger fault scarp and a longer period of fast slip rate reaching 2 mm yr⁻¹ from the present 676 until 7–8 ka. The larger scarp at site RP3 is most likely a result of fast erosion of an unstable 677 hanging wall on the edge of a major gully that incises the hanging wall and footwall of the 678 fault (Figure 4). The fault scarp has been subject to active net erosive slope processes 679 that likely removed material from the hangingwall slope exposing the fault surface in the 680 gully, resulting in the higher scarp and faster slip rate than at other sites on the fault. The 681 difference in timing of peak slip at site RP2 suggests that site RP1 experienced a more 682 recent or larger slip event, implying that the fault does not always rupture continuously 683 or that there is a variation in surface slip in a single event along the fault. We suggest 684 that the slower average slip rate and shorter scarp height at site RP2 compared to site RP1 685 is because strain is partitioned between the end of the strand sampled (at RP2) and an 686 overlapping fault strand located 1 km west and across strike (Figure 4). Between the two 687 fault strands, there is a ramp in the topography that slopes down towards the southeast, 688 and a step across each fault segment, perpendicular to fault strike, which is typical of a 689 classic relay ramp morphology, where the length of the relay ramp is approximately 3 times 690 the width (Fossen & Rotevatn, 2016, Figure 4). 691

Interaction between closely spaced fault segments can reduce the total displacement 692 across individual faults due to strain partitioning, including at fault splays (McLeod et 693 al., 2000; Cowie & Roberts, 2001; Manighetti et al., 2015). Our analyses at Roccapreturo 694 suggests that over thousand year time scales the overlapping fault segments do not be-695 come completely inactive, but instead overlapping segments have slower average slip rates 696 (or less slip per event) relative to the center of the main fault segment. Quaternary slip 697 rate variation along strike is not typically observed at this scale and temporal resolution, 698 demonstrating that ³⁶Cl provides a unique ability to investigate fault segment interaction 699 and strain partitioning over millennial timescales. Due to the relatively young age of the 700 normal fault network in the central Italian Apennines (2.3-3.3 Ma; Roberts et al., 2002) 701 and low extension rates across the region (2.7 mm yr^{-1} D'Agostino et al., 2011), the fault 702 system is immature, with a complex network of faults in the region that are highly inter-703

active on relatively short timescales (including in earthquake sequences, e.g. Nixon et al.,
 2016). Individual faults in the central Apennines are still growing, and through the process
 of localisation, splays like the ones observed along the Roccapreturo fault may eventually
 become hard linked through to the surface and be capable of larger earthquakes and faster
 slip rates.

On the Campo Felice and Roccapreturo faults, we can observe the cumulative effect 709 of the complexity of earthquake surface ruptures and resulting variation in displacement 710 along strike. Some complexity arising during individual earthquakes may cumulatively can-711 712 cel out over multiple earthquake cycles, if it is localised or random, and may contribute relatively insignificant noise to calculated slip rates. However if patches of high or low slip 713 occur repeatedly in the same location on the fault, the displacement at any one site is not 714 representative of either the fault rupture as a whole during that event, or that particular 715 site over multiple earthquake cycles. We find that the variation in slip is consistent over 716 multiple earthquakes cycles at some sites, such as site RP2 having lower slip than at RP1, 717 and that sites CF2 and CF4 have higher slip than CF1 and CF3. By comparing ruptures 718 from individual events with multiple offsets accumulated over longer timescales, it is possible 719 to better understand along-strike variability (Brozzetti et al., 2019; Cinti et al., 2019). 720

721 6 Conclusions

We present 36 Cl cosmogenic isotope results and modelled exposure histories from 4 722 sites on the Campo Felice fault and 3 sites on the Roccapreturo fault. Unlike previous 723 work, our modelling approach can be uniformly applied to all data in order to test whether 724 they agree, without arbitrarily varying parameters related to the preservation of the fault 725 scarps. Models from different sites on the same fault have the same long-term preservation 726 age and elapsed time since the last earthquake, as well as similar long-term patterns of slip 727 rate variability. The slip histories do not agree where samples are collected from unstable 728 slopes, and there are parts of the faults that have slower average rates (likely caused by 729 lower displacement in cumulative events), due to rupture complexity or strain partitioning 730 between overlapping fault strands. Each fault experiences periods of time with much faster 731 slip rates than the average Holocene rate, likely due to multiple earthquakes occurring over 732 a few thousand years. The average slip rate for the Campo Felice fault is 1.15 ± 0.36 mm 733 yr-1, and 0.42 ± 0.14 mm yr⁻¹ on the Roccapreturo fault, with peak slip rates of 3 mm 734 yr^{-1} and 2 mm yr^{-1} at each fault, respectively. The range of along-strike variability in slip 735 rate means that one location may not represent the typical behaviour or hazard of an entire 736 fault, and while sampling faults multiple times along strike is not always feasible, it can 737 improve confidence in results by elucidating the range of slip rate and the timing of changes 738 in slip rate. 739

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