

**Fracture attribute scaling and connectivity in the Devonian Orcadian  
Basin with implications for geologically equivalent sub-surface  
fractured reservoirs**

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**Abstract:** Fracture attribute scaling and connectivity datasets from analogue systems are widely used to inform sub-surface fractured reservoir models in a range of geological settings. However, significant uncertainties are associated with the determination of reliable scaling parameters in surface outcrops. This has limited our ability to upscale key parameters that control fluid-flow at reservoir to basin scales. In this study, we present nine 1D-transect (scanline) fault and fracture attribute datasets from Middle Devonian sandstones in Caithness (Scotland) that are used as an onshore analogue for nearby sub-surface reservoirs such as the Clair Field, West of Shetland. By taking account of truncation and censoring effects in individual datasets, our multi-scale analysis show a preference for power-law scaling of fracture length over 8 orders of magnitude ( $10^{-4}$  to  $10^4$ ) and kinematic aperture over 4 orders of magnitude ( $10^{-6}$  to  $10^{-2}$ ). Our assessment of the spatial organisation (clustering and topology) provides a new basis for up-scaling fracture attributes collected in outcrop- to regional-scale analogues. We show how these relationships may inform knowledge of geologically equivalent sub-surface fractured reservoirs.

*Keywords:* fault, fracture attribute, multi-scale, Clair Field, Devonian

## 1 Introduction

Fractures, used in this paper as a general term to include faults, joints and veins, fundamentally control the fluid-flow and mechanical properties of many crustal rocks, including many sub-surface reservoirs holding oil, gas or water (e.g. Nelson, 1985; Sibson, 1996; Adler & Thøvert, 1999; Odling et al., 1999) or potential subsurface repositories (De Dreuzy et al., 2012). Establishing the size, spatial organisation, connectivity, scaling and fracture-fill properties of fluid conductive structures is crucial to understanding the performance of sub-surface reservoirs in a range of low porosity/permeability rock types (see review by Laubach et al., 2019). In sub-surface reservoirs, fracture description is typically performed on image logs and drillcores that provide high resolution ( $10^{-4}$  to  $10^0$  m), but highly censored (size limited by borehole diameter), spatially limited and biased 1D samples (e.g. Odling et al., 1999; Zeeb et al., 2013). Accurately characterizing 3D fracture network properties using just borehole and cores is particularly challenging (e.g. Berkowitz & Adler, 1998) hence reservoir analogues can give access to fracture datasets across many scales ( $10^{-2}$  to  $10^6$  m scales) and in 1, 2 and 3 dimensions for use in reservoir models (Jones et al., 2008). Statistical analysis of fracture attributes from appropriate outcrop analogues can provide reliable and robust geological (conceptual models) and quantitative (attribute and scaling) information to inform the planning of exploratory and development drilling, and design and conditioning of reservoir simulation models (Mäkel, 2007).

Fractures can be described by: 1) their *size* (displacement, length and aperture – for opening mode structures). Previous studies have demonstrated that size attributes, in particular, have in many cases scale-invariant properties (power-law distribution) from microns to hundreds of kilometres (e.g. Sanderson et al., 1994; Cowie et al., 1996; Marrett et al., 1999; Bonnet et al., 2001); 2) their *spatial* attributes such as orientation, intensity/density, arrangement, clustering, connectivity and continuity (Laubach et al., 2018 and references therein). Clustering of small faults (fractures with  $< 1$  m displacement) and joints (fractures with no shear displacement) may occur or as part of a damage zone of larger displacement ( $> 10$  s m) faults (e.g. Schultz and Fossen, 2008; Peacock et al., 2016) but also as sub-parallel fracture swarms or corridors (Marrett et al., 2018; Wang et al., 2019). Fracture connectivity can be measured using topological methods (e.g. Sanderson and Nixon, 2015); and 3) their *chemical/cement* attributes (e.g. Laubach et al., 2003; 2019) which describe fracture fill characteristics (e.g. Holdsworth et al., 2019; 2020).

Using one dimensional sampling methods (e.g. scan lines, transects etc), fracture attributes have been investigated in different tectonic contexts and lithologies (e.g. Baecher, 1983; Gillespie et al., 1993;

McCaffrey and Johnston, 1996; Knott et al., 1996; Odling et al., 1999; Bour et al., 2002; Manzocchi, 2002; Olson, 2003; Kim and Sanderson, 2005; Gomez and Laubach, 2006; Schultz et al., 2008; Hooker et al., 2009; Torabi and Berg, 2011). In these studies, the statistically best constrained data tends to be acquired at a single scale for example an outcrop or a well core. In order to better constrain attribute scaling, it is desirable to extend the range of sampling to larger (or smaller) scales. This *multi-scale* sampling generally involves combining data collected at different observational scales (e.g. Walsh and Watterson, 1988; Marrett et al., 1999; Guerriero et al., 2010a, b; Torabi and Berg, 2011; Bertrand et al., 2015). Examples include datasets collected at *regional scale* (seismic reflection, remote sensed image interpretations), *macro scale* (outcrops, drill core, image logs) and *micro-scale* (thin-sections). Marrett et al. (1999) combined data collected at two scales for faults and extension fractures to reduce uncertainties in the scaling of fracture aperture and fault displacement.

In this study, we use an integrated multi-scale sampling approach to describe the scaling fractures formed in Middle Devonian sandstones of the Orcadian Basin, North Scotland. The Orcadian basin exposures are widely viewed as being an appropriate analogue for the fractured Devonian siliciclastic reservoirs that form the giant Clair Field, West of Shetland (Allen and Mange-Rajetzky, 1992; Coney et al., 1993; Barr et al., 2007), one of the largest remaining oilfields in the UKCS (ca. 7 billion barrels of Stock Tank oil initially in place, Robertson et al., 2020). For the Orcadian Basin, we collected datasets from a high-resolution bathymetric map (sub-regional scale), aerial photographs, coastal exposures and, a thin-section made from hand samples. Importantly, we carried out a multi-scale analysis of both *size* and *spatial attributes* of the fracture populations. We use the results to suggest how the determination of multi-scale fracture attribute scaling in 1D and 2D can form a useful input for building realistic static geological models at reservoir scale. These models serve as starting points for simulations of fluid storage, migration processes and production in sub-surface reservoirs.

## 2 Geological Setting

### 2.1 Location and regional structure

The studied siliciclastic strata are Devonian Old Red Sandstone (ORS) of the Orcadian Basin exposed in the Caithness region, North Scotland. The Orcadian Basin covers a large area of onshore and offshore northern Scotland forming part of a regionally linked system of basins extending northwards into western Norway and East Greenland (Seranne, 1992; Duncan and Buxton, 1995) (**Fig. 1a**). The great majority of the onshore sedimentary rocks of the Orcadian Basin in Caithness belongs to the

97 Middle Devonian and sits unconformably on top of eroded Precambrian (Moine Supergroup)  
98 basement. These sedimentary rocks and the fractures they contain have long been used as an onshore  
99 analogue for parts of the Devonian to Carboniferous Clair Group sequence that hosts the Clair oilfield  
100 west of Shetland (**Fig. 1**; Allan and Mange-Rajetzki, 1992, Duncan and Buxton, 1995). It should be  
101 noted that strictly speaking, the Clair Group formed in an adjacent basin, in a somewhat different  
102 tectonic setting (Dichiarante, 2017; Dichiarante et al 2020).

103 Recent fieldwork has shown that the onshore Devonian sedimentary rocks of the Orcadian  
104 Basin in Caithness host significant localized zones of fracturing, faulting and some folding on all  
105 scales. Field and microstructural analyses reveal three regionally recognised groups (sets) of  
106 structures based on orientation, kinematics and infill (Dichiarante et al., 2016; 2020; Dichiarante,  
107 2017). In summary, these are as follows:

108 **Group 1 faults** trend mainly N-S and NW-SE and display predominantly sinistral strike-slip to dip-  
109 slip extensional movements. They form the dominant structures in the eastern regions of Caithness  
110 closest to the offshore trace of the Great Glen Fault (GGF) (**Fig. 1a-b**). Deformation bands, gouges  
111 and breccias associated with these faults display little or no mineralization or veining. It is suggested  
112 that these structures are related to Devonian ENE-WSW transtension associated with sinistral shear  
113 along the Great Glen Fault during formation of the Orcadian and proto-West Orkney basins (Wilson  
114 et al., 2010; Dichiarante et al 2020).

115 **Group 2 structures** are closely associated systems of metre- to kilometre-scale N-S trending folds  
116 and thrusts related to a highly heterogeneous regional inversion event recognized locally throughout  
117 Caithness. Once again, fault rocks associated with these structures display little or no mineralization  
118 or veining. Group 2 features are likely due to late Carboniferous – early Permian E-W shortening  
119 related to dextral reactivation of the Great Glen Fault (Coward et al., 1989; Seranne, 1992;  
120 Dichiarante et al 2020).

121 **Group 3 structures** are the dominant fracture sets seen in the main coastal section west of St. John's  
122 Point (SJ in **Fig. 1b**). They comprise dextral oblique NE-SW trending faults and sinistral E-W  
123 trending faults with widespread syn-deformational low temperature hydrothermal carbonate  
124 mineralisation ( $\pm$  base metal sulphides and bitumen) both along faults and in associated mineral veins  
125 (Dichiarante et al., 2016). Hydrocarbons are widespread in fractures in small volumes and are locally  
126 sourced from organic-rich fish beds within the Devonian sequences of the Orcadian Basin (Parnell,  
127 1985; Marshall et al., 1985). Re-Os model ages of syn-deformational fault-hosted pyrite in Caithness

yield Permian ages (ca. 267 Ma; Dichiarante et al., 2016). This is consistent with the field observation that Group 3 deformation fractures and mineralization are synchronous with the emplacement of ENE-trending lamprophyre dykes east of Thurso (ca. 268-249 based on K-Ar dating; Baxter and Mitchell, 1984). Stress inversion of fault slickenline data associated with the carbonate-pyrite-bitumen mineralization implies NW-SE regional rifting (Dichiarante et al., 2016), an episode also recognized farther west in the Caledonian basement of Sutherland (Wilson et al., 2010). Thus from St. John's Point to Cape Wrath (CW in **Fig. 1b**), Permian-age faults are the dominant brittle structures developed along the north coast of Scotland, forming part of a regional-scale North Coast Transfer Zone translating extension from the offshore West Orkney Basin westwards into the North Minch Basin (see Dichiarante et al., 2016; 2020).

## 2.2 Group 3 structures – analogue for Clair reservoir

The Group 3 structures are the only set widely associated with syn-faulting mineralization and bitumen and have therefore clearly acted as fluid channel-ways in the past. There is also good evidence for the preservation of open fractures and vuggy cavities consistent with these fractures continuing to be good potential fluid-flow pathways at the present day. No such features are associated with Group 1 or Group 2 structures. Most of the Group 3 fractures measured during the onshore study in the Orcadian Basin in Caithness are partially to completely filled with either fault rocks, minerals or bitumen; a range of filling morphologies are preserved that have been described by Dichiarante et al. (2016; 2020) (**Fig. 2a-e**). It is reasonable to assume that wholly bitumen-filled fractures can be viewed as being equivalent to open fractures in a sub-surface reservoir (**Fig. 2a, b**), whilst other veins may be completely filled with minerals/fault rock (lacking bitumen) or partially filled with hydrocarbon held either in vuggy cavities (**Fig. 2c**), fractured mineral fills (**Fig. 2**) and/or porous sediment fills (**Fig. 2e**). There are many examples of partly or fully open fractures in the surface coastal exposures of the Orcadian Basin, but it is difficult to prove whether or not surface weathering and seawater washing of coastal outcrops has not removed pre-existing fracture fills. This is supported by the observation that fracture-hosted bitumen fills are most widely preserved in recently exposed quarry or excavation sites inland (see Dichiarante et al., 2016). These authors presented textural evidence showing that fracture-hosted calcite, sulphides and oil fills are broadly contemporaneous. They suggest that open vugs and fractures are almost certainly only preserved due to hydrocarbon flooding which shuts down the further precipitation of carbonate and sulphide in open or partially open fractures/veins (e.g. **Fig. 2b-e**).

161 Previous works, for example Barr et al. (2007), suggested that outcrops in the Orcadian Basin show  
162 similar features to the Clair Field, in particular, they highlighted similar faults, open fractures,  
163 granulation seams, cemented fractures and, in particular, linear zones of fracturing as equivalent to  
164 linear zones of disaggregated core in the sub-surface. Our observations from the Clair Field cores  
165 reveal similar associations between fractures filled, or partially filled, with similar hydrothermal  
166 minerals, younger porous sediment and hydrocarbons (see example in Supplementary datafile  
167 extracted from Dichiarante, 2017). This suggests that despite differences in source rocks (local  
168 Devonian onshore versus more distant Jurassic offshore), the Orcadian Basin Group 3 fracture fills  
169 and apertures are a good analogue for the fractured rocks of the Clair Group. Barr et al. (2007) noted  
170 the presence of dispersed joints, in outcrop, which they attributed to exhumation features as being  
171 much rarer in core. In this study, we carried out fracture attribute analyses in areas where Group 3  
172 structures predominate, or in locations where there is good field evidence that pre-existing Group 1  
173 faults have undergone significant later reactivation synchronous with Group 3 age deformation  
174 (Dichiarante et al 2020). We did not include obvious early (Group 1 & 2) or late jointing in our  
175 fracture datasets.

176

177 Downie (1998) reported that sandstones of the middle ORS in the North Sea have poor-reservoir  
178 quality due to widespread cementation comprised of calcite, dolomite, quartz overgrowths and clay  
179 minerals. This author also reported open fractures in all discoveries in ‘tight-matrix’ sandstones in  
180 the Orcadian Basin (the Buchan, Stirling and West Brae fields) and that these features are present in  
181 the Clair Field. He also referred to fracture-fill cements are present which are similar to the Group 3  
182 structures of the Orcadian Basin. With reference to the three criteria for an analogue to be considered  
183 appropriate, as recently suggested by Ukar et al. (2019), the Orcadian Basin outcrops show: 1) a  
184 similar structural setting and lithofacies, 2) the sandstone host rocks were in a similar state of  
185 diagenesis during the deformation, and 3) the fracture cements show similar textures and formed  
186 under similar conditions to the producing structures in the Clair Field. Thus, the Group 3 structures  
187 of the Orcadian Basin clearly formed in the sub-surface and we argue that they are the best direct  
188 analogue for the oil-bearing fracture systems that occur in the Clair Group reservoir.

189

## 190 3 Methodology

### 191 3.1 Sampling of fractures and fracture network attributes

192 The most common data acquisition methodologies use: (i) scanlines (or transects); (ii) window  
193 sampling; (iii) circular scanline windows; and (iv) box counting (**Fig. 3a**), which collectively provide  
194 access to different attributes as shown in **Table 1**. Scanlines (1D method) allow a relatively simple  
195 characterization of individual fracture sizes and spacing (**Fig. 3b**) and act as a good proxy for the  
196 borehole data that typically serve as starting points for building reservoir models (Priest and Hudson,  
197 1981; Baecher, 1983; Gillespie et al., 1993; McCaffrey and Johnston, 1996; Knott et al., 1996; Ortega  
198 and Marrett, 2000; Ortega et al., 2006; Bonnet et al., 2001; Odling et al., 1999). Window sampling  
199 and circular scanline windows (both 2D methods) provide further information on the spatial  
200 relationships within the fractured system (Mauldon, 1994; Mauldon et al., 2001; Rohrbaugh et al.,  
201 2002; Manzocchi, 2002; Zeeb et al., 2013; Watkins et al., 2015; Sanderson and Nixon, 2015; Rizzo  
202 et al., 2017) and importantly provide access to connectivity estimates for the fracture array, which is  
203 a key input when modelling fluid-flow.

204 In this study, fracture orientations, trace lengths and apertures, together with composition and  
205 texture of fracture infills and fracture terminations for all Group 3 structures were recorded. The start  
206 and end point of each transect was recorded using a hand-held GPS unit. Most fractures in the  
207 Orcadian Basin are filled with minerals (calcite or pyrite) or, locally, oil, and, following Ortega et al.  
208 (2006), the apertures measured in this study are the orthogonal distance between the fracture walls  
209 and include the fill, i.e. the '*kinematic aperture*'. Most Group 3 fracture sets are made up of fracture  
210 meshes (sensu Hill 1977, Sibson 1996) formed by closely interlinked sets of contemporaneous shear  
211 fractures and tensile veins (Dichiarante et al. 2016, 2020). Thus, in each sample, all fractures  
212 considered to belong to an individual fracture set (in this case Group 3) were included in the analysis  
213 regardless of opening mode. Thus in our view it is not possible to separate brittle structures into  
214 separate sets of simple tensile and shear fractures. This practical approach ensures comparability with  
215 subsurface structures in Clair cover sequences and related fractured basement studies where similar  
216 interlinked mesh systems are dominant (see McCaffrey et al 2020). One reason for the development  
217 of such mesh networks is that many of Group 3 structures reactivate earlier (Group 1 and 2) brittle  
218 structures and therefore display a variety of hybrid opening modes (Dempsey et al. 2014).

219 When it was not possible to measure the transect orthogonally to the main fault because of outcrop  
220 exposure limitations (e.g. at the sub-regional scale), the measured attributes were adjusted using the



221 Terzaghi's Correction (Terzaghi, 1965). To more precisely measure the aperture attributes, an  
222 engineering feeler gauge in conjunction with a hand lens ( $10^{-5}$  to  $10^{-4}$  m) was used in the field to in  
223 order to ensure a larger range of recorded apertures, thereby reducing censoring effects.

224 To extend the analysis to other scales, the mentioned above scanline method was adapted and  
225 applied to both aerial photographs (regional scale) to quantify trace length, and to thin-section  
226 (microscale) to quantify trace length and aperture. Fracture lengths mapped as continuous at regional  
227 scale are likely to comprise segments which may not be resolvable at the scale of observation.  
228 However in terms of fluid flow, segmented faults may be connected as single structures at depth so  
229 our lineaments may represent an interconnected length in the sense of Olson (2003).

230

### 231 3.1.1 *Fracture intensity/frequency plots (1D)*

232 The fracture intensity/frequency distribution for 1D datasets can be visualised by plotting sorted  
233 attribute values (e.g. fracture length) versus cumulative frequency. This enables assessment of the  
234 distribution, spatial and scaling properties of the fracture sample (i.e. the ratio of short to long  
235 fractures for given sample line length). Fracture attribute distributions are thought to display three  
236 main types of statistical distribution (**Fig. 4**, Bonnet et al., 2001; Gillespie et al., 1993; Zeeb et al.,  
237 2013): (a) *Exponential, random or Poisson distributions* are characteristic of a system with one  
238 randomised variable (Gillespie et al., 1993); (b) *Log-normal distributions* are generally produced by  
239 systems with a characteristic length scale where mechanical stratigraphic boundaries control joint  
240 spacing for example (Narr, 1991 and Olson, 2007); (c) *Power-law distributions* lack a characteristic  
241 length scale in the fracture growth process (Zeeb et al., 2013) and the data exhibit scale-invariant  
242 fractal geometries (**Fig. 4c** bottom). For a power-law distribution, the relative number of small versus  
243 large elements remains the same at all scales between the upper and lower fractal limits (Barton,  
244 1995). Ideally, the best-fit in a power-law distribution should be consistent over several orders of  
245 magnitude length scale (Walsh and Watterson, 1993; McCaffrey and Johnston, 1996). Limits to the  
246 fractal behaviour can be related to both spatial and temporal influence, e.g., lithological boundaries  
247 across which fracture characteristics change, changes in stress orientation, respectively or diagenetic  
248 effects (Hooker et al., 2014). However, it is generally accepted that power-law distributions and  
249 fractal geometry provide a widely applicable descriptive tool for fracture size attributes such as  
250 aperture and length (e.g. Bonnet et al., 2001; Olson 2007; Hooker et al., 2014; McCaffrey et al.,  
251 2020).

252

Fracture sampling issues (e.g. censoring and truncation in **Fig. 4c**) are commonly encountered and can result in difficulty in ascribing the best-fit distribution. For instance, when long fractures are incompletely sampled (e.g. censoring in **Fig. 4c**), it is difficult to determine between log-normal and power-law fits to distributions. These sampling issues (due to resolution effects) may mean that, while a log-normal distribution is the best-fit to a dataset, a power-law distribution can also show a good fit (Corral and González, 2019) and may be preferred because of its greater physical significance and practical applicability (Bonnet et al., 2001). These assumptions need to be examined closely in any analysis of scaling (see Clauset et al., 2009) and power-law behaviour should not be assumed. The maximum likelihood estimator (MLE) is a statistical technique that determines which distribution model is most likely to describe the data and it returns governing parameters of the fitting equations (see Supplementary Data File). The Kolmogorov-Smirnoff (KS) test is then used to evaluate the difference between the data and synthetic data generated using the governing parameter derived from the MLE (Clauset et al., 2009). We use these statistical methods and adapted the methodology proposed by Rizzo et al. (2017) and used in FracPaQ (Healy et al., 2017) to calculate the MLE on progressively truncated populations for power-law, exponential and log-normal distributions.

### 3.1.2 2D topology analysis

Whilst 1D analyses provide information about fractures as single entities and their distribution per unit length of sample, 2D analyses measure fracture network properties and provide estimates of fracture connectivity and self-similarity. The 2D analysis used here was carried out on fractures at mesoscale using outcrop pavement photographs and at a larger scale using an offshore bathymetric data. Circular scanline windows and box counting methods were performed using the Corel Draw Graphic Suite™, ArcGis™ and MATLAB™ to produce small-scale fracture density maps (**Fig. 4d**), ternary plots of connection types (**Fig. 4e**) and box counting dimension (**Fig. 4f**). To understand fracture topology, we follow Sanderson and Nixon (2015) in considering that fracture arrays are typically composed of nodes and branches. Nodes are points where a fracture terminates (I-type), abuts against another fracture (Y-type) or intersects another fracture (X-type) and branches are the portions of a fracture confined between two nodes. These branches are defined as I-I type (isolated) if delimited by two I-nodes, I-C type (singly connected) if delimited by an I-node and Y- or X-node and C-C type (multiply connected) if delimited by Y- and X-nodes.

The number of branches and nodes for a given fracture network is strictly related meaning that, by knowing one of the two elements for the fracture network, it is possible to quantify all its

components.  $N_I$ ,  $N_Y$  and  $N_X$  can be defined as the number of I-, Y- and X-type nodes and  $P_I$ ,  $P_Y$  and  $P_X$  their relative proportions. Once the number of nodes and/or branches making up a fracture array is known, the fracture trace connectivity can be visualized using the ternary plot of the component proportions (see e.g. **Fig. 4e**) or can be quantified by calculating the number of connections existing in the 2D map. In general, X- and Y-type nodes provide respectively 4 and 3 times more connectivity than I-type nodes (Nixon, 2013). This forms the basis for creating 2D density maps (see **Fig. 4d**). An array dominated by I-nodes is isolated, while arrays dominated by Y- and X-type nodes are increasingly more connected. Connectivity can be quantified by measuring the number of connections per line  $n_{C/L}$  and the number of connections per branch  $n_{C/B}$  (see Sanderson & Nixon 2015 for details).

#### 4 Locations and orientation data from the 1D scanlines

In the present study, 1D scanlines were performed at different scales in the Caithness area resulting in datasets from regional- (km-scale, **Fig. 1b**) and sub-regional- ( $10^2$  m- to m-scale, **Fig. 1c**), meso- (m- to cm-scale, **Fig. 1d**) and micro-scales ( $\mu$ m-scale, **Fig. 1e**).

##### 4.1 Regional- and sub-regional scale

Scanline data have been collected at a regional scale (km-scale) using a tectonic lineament interpretation map created by Wilson et al. (2010). In their study, the lineament analysis was conducted at 1:100k scale extending from Lewisian basement outcrops in western Sutherland eastwards into the Devonian rocks of Caithness (**Fig. 1b**). We performed two scanlines (WTr1 and WTr2) trending orthogonally to the Brough-Risa Fault, the major N-S trending basin-scale fault in Caithness (**Fig. 1b**; Dichiarante et al. 2016). Scanline WTr1 intersects mainly NE-SW and NW-SE trending lineaments, while scanline WTr2 intersects mainly N-S and a few NE-SW trending lineaments (**Fig. 1a**). Although, datasets with few data points generally give poorly defined distributions on graphical presentations, it will be shown that the data from these two transects are of value in the multiscale approach adopted here.

At the sub-regional scale, scanlines have been performed on lineament maps produced from Google Earth satellite images at 1:1k scale (pixel resolution ca. 10 m). These datasets are limited to well exposed wave-cut platforms on the coast because the flat topography and thick cover of drift has obscured the structures inland. The interpreted lineaments from the images were verified during fieldwork as being faults (large to mesoscale) and joints and not anthropogenic features. The narrow

width of the platform limits the analysis to only one scanline in each locality (DO at Dounreay, SJ at St. John's Point; see **Fig. 1c**). We estimate that we would have recorded 10-20 % of the total number sampled at these localities if the transect line had followed the sub-regional trend rather than the outcrop extent. Fracture spacing measurements were been corrected using the Terzaghi's Correction (see dashed red and blue lines in the rose diagrams in **Fig. 5b-c**).

The scanline at Dounreay (DO) is NE-SW trending and intersected mainly NW-SE and NNE-SSW trending, with a subset of NE-SW trending lineaments (**Fig. 5b**). The scanline at St. John's Point (SJ) intercepts mainly ENE-WSW trending lineaments with subsets of N-S and NW-SE trending lineaments (**Fig. 5c**).

#### 4.2 Mesoscale outcrops

Fracture data along six mesoscale scanlines were collected at three field localities: Brims Ness (BTr1, BTr2; **Fig. 5d, e**), Castletown (CTr1, CTr2; **Fig. 5f, g**) and Thurso (TTr1, TTr2; **Fig. 5h, i**) where there is very good exposure. In each outcrop, the position, direction and length of the scanlines were chosen with reference to the trend of the basin-scale master faults in each area (e.g. ENE-WSW at Castletown and NNE-SSW at Thurso and Brims Ness; **Fig. 5d-i**). At Castletown and Brims Ness, two scanlines were carried out to record the full range of fracture orientations: one parallel and one perpendicular to the master fault set. Scanlines at Thurso differ from the others because they are both measured parallel and next to a fault zone, resulting in higher values of fracture intensity (see TTr1 and TTr2 in **Table 2**). These scanlines are also shorter (< 4m) and record exclusively thin veins. Each locality is characterized by one (e.g. Thurso) or more fracture sets (e.g. Castletown, Brims Ness). Where two sets of fractures are present, they mutually cross-cut each other which enabled us to infer that they were active during the same geological event; hence they are analysed here as single population (Dichiarante, 2017).

Additionally, for each scanline, fracture termination type, kinematics and type of fractures were recorded (**Table 2**). Although fracture terminations are more usefully assessed in a 2D analysis, we recorded the nature of fracture branch terminations for each structure intersecting the transect line. These data are reported using a ternary plot (**Fig. 5j**) which shows that there is no dominant fracture termination type. In general, the transects show intermediate to high connectivity, except for scanline TTr1, which shows a more isolated pattern.

352 An orientation analysis of fracture intersection has been carried out for onshore faults and  
353 fractures data at St. John's Point (**Fig. 5k**, Dichiarante, 2017), based on its proximity and geological  
354 similarity to the area covered by the bathymetric map which lies immediately offshore (see section  
355 7). A similar plot is also shown for all the faults and fractures data collected in Caithness (**Fig. 5l**,  
356 Dichiarante, 2017). Both datasets show consistent best-fit intersections that are sub-vertical to steeply  
357 plunging to the east, 73/084 and 78/098, respectively (yellow diamonds in the stereonet in **Fig. 5k**  
358 and l).

359

#### 360 4.3 Microscale scanline

361 At microscale, one transect was performed on an oriented thin-section taken from sample SK04  
362 (inset in **Fig. 1e**, left). At the scale of a thin-section, only samples from fault zones contain enough  
363 fractures to produce a statistically significant sample. We thus recognise that the results at this scale  
364 are representative of fracture intensities within fault zones and provide an upper limit relative to  
365 background. Our field observations ensured that the age of this fault was the same as the other Group  
366 3 structures analysed at different scales. This fault rock was chosen as it is a typical example of a NE-  
367 trending fault with normal dextral oblique kinematics, filled with carbonate mineralization and red  
368 stained (hematite) sandstone-breccia of inferred Permian age (**Fig. 1e**, see also **Fig. 2e**). The oriented  
369 thin-section was analysed under an optical microscope and spacing, aperture and the lengths of  
370 microfractures recorded. Photo-micrographs were merged and the scanline was measured  
371 orthogonally to the bounding NE-SW meso-fracture (**Fig. 1f**).

372

### 373 5 1D fracture population results

#### 374 5.1 Fracture length, aperture and intensity/spacing results

375 MLE distribution fitting and KS tests were performed for all datasets and different types of  
376 distribution (exponential, log-normal, power-law). The recorded range values of trace length and  
377 aperture (or vein width) for each of datasets are shown in **Table 2**. In Table 3 and Table 4 in the  
378 Supplementary Data File, we report the MLE distribution fitting results for both non-truncated  
379 (exponential, log-normal and power-law distributions) and for truncated (power-law distribution)  
380 populations for trace length and aperture, respectively.

381 Length population datasets yielded values, rounded to the nearest order of magnitude, centred  
382 at ca. 10 m for the sub-regional scale,  $10^{-1}$  m at mesoscale and  $10^{-4}$  m at microscale (**Fig. 6a**). Aperture

populations are centred between ca.  $10^{-3}$  m for the mesoscale dataset and ca.  $10^{-5}$  m for the microscale dataset (**Fig. 6**).

The plots in **Fig. 6** give an insight into the relationship between cumulative frequency/intensity (inverse spacing), and length or aperture. For example, at the mesoscale (**Fig. 6b** right), the intensity of fractures with  $> 25$  mm aperture is about  $0.03 \text{ m}^{-1}$  corresponding to a 34 m spacing. Similarly, the intensity of fractures with  $> 0.4$  mm aperture is between  $0.45 \text{ m}^{-1}$  and  $11.2 \text{ m}^{-1}$  corresponding to 8.9 cm to 2.2 m spacing, respectively. At microscales (**Fig. 6b** left), the intensity of fractures with  $> 2.9 \cdot 10^{-5}$  m aperture is about  $155.51 \text{ m}^{-1}$  corresponding to 6 mm spacing whilst the intensity of fractures with  $> 3.9 \cdot 10^{-6}$  m aperture is about  $1555 \text{ m}^{-1}$  corresponding to a 0.64 mm spacing. These spacing values do not take into account any systematic spatial arrangement. Using the Marrett et al. (2018) spatial correlation analysis on our 1D fracture samples, we see fractal clustering with 0.02 - 0.2 m spaces twice as likely compared to a random distribution at Castletown, Thurso and Brims Ness. These are arranged in 1 - 3 m wide clusters spaced at 1 - 20 m. At larger scale, Dounreay shows cluster widths of 650 m spaced with 5 km intervals. In contrast, St. John's Point shows anti-clustered (regularly spaced) fractures at ca. 2 m (see plots in the Supplementary Data File).

To examine the possible influence of mechanical stratigraphy on fracture scaling across the Orcadian Basin in Caithness, we indicate on the fracture size plots, selected sedimentary unit thickness values reported in previous studies (**Fig. 6**). These include sedimentary laminae thickness (0.3 mm) at microscale, bedding-range thicknesses of the Lower Stromness Formation (20 cm to 5 m) at mesoscale, and thicknesses of the Ham-Skarfskerry and Latheron Subgroups at sub-regional scales (data from Andrews et al., 2016). Also, the approximate boundary between faults that can be imagined in seismic reflection images and smaller-scale structures is shown in **Fig. 6** (yellow arrows) based on well-known empirical displacement-length relationships (a 10 m displacement corresponding to a length of ca. 100 m, following Kim and Sanderson, 2005).

#### *5.1.1 Analysis of uncertainties: validity of data populations and reliability of best-fit distributions*

In any statistical analysis, the sampled population should be large enough to give a statistically acceptable representation of the population and to properly determine the distribution type and its parameters (Bonnet et al., 2001). The sample sets are statistically valid for most samples after the first 20 measurements (grey area in **Fig. 7**) because the cumulative fracture intensity of the population data and its standard deviation (black and green curves, respectively) become reasonably stable. The uncertainty in the cumulative fracture intensity reduces progressively towards the end of the scanline.

## 416    **6    The scalability of fracture attributes**

### 417    *6.1    Slope determination – MLE approach*

418    The complete (non-truncated) populations show that a log-normal distribution best describes the data  
419    as they show consistently high percentage fitting values. However, the choice of the best-fit  
420    distribution should not be based on the complete population because the distribution tails  
421    (corresponding to the largest and smallest size fractures) are biased (see also Supplementary Data  
422    File). We therefore also investigated progressively truncated populations in order to validate the  
423    hypothesis. The fitting results for complete log-normal and truncated power-law datasets are  
424    generally similar (see Supplementary Data Files), suggesting that either type of distribution can  
425    successfully describe the size attribute data.

426

### 427    *6.2    Multiscale analysis*

428    Trace length distribution data from all transects have been normalised using sample line length (c.f.  
429    Marrett et al., 1999) and are displayed together on a single population plot (**Fig. 8a**) which enables us  
430    to assess scaling over 8 orders of magnitude ( $10^{-4}$  to  $10^4$ ). The grey region in **Fig. 8a** shows that the  
431    multiscale data can be described by a power-law distribution with overall scaling coefficient close to  
432    a slope of -1 centred on a 1 m length fracture with a 1 m spacing. This power-law distribution implies  
433    fractal or self-similar behaviour of the length parameter over 8 orders of magnitude which effectively  
434    means that the fracture array maintains the same statistical properties of intensity and length at all  
435    scales assessed here.

436

437        The aperture datasets collected in the meso- and micro-scale transects are also shown on a  
438    single population plot (**Fig. 8b**) and show evidence for an overall power-law scaling over 4 orders of  
439    magnitude ( $10^{-6}$  to  $10^{-2}$ ) also with a slope of -1. However, the best-fit line is centred on a 1 mm wide  
440    fracture with a 1 m spacing. This overall slope is indicative of a fractal distribution or self-similar  
441    behaviour of the aperture parameter over 4 orders of magnitude which means that the fracture array  
442    maintains the same relationship between intensity and aperture at all scales assessed here.

443

444        In order to assess whether stratigraphic units influenced the fracture scaling, the estimated  
445    overall thickness of the Devonian rocks in Caithness by Donovan (1975) and the smallest scale  
446    bedding planes of ca.  $10^{-4}$  m that were observed in thin-section are shown on **Fig. 8a** (dashed red

lines). These limits approximately span the range of fracture lengths recorded and the absence of obvious slope-changes between the limits suggest that stratigraphic element have not played a role in determining the fracture scaling.

### 6.3 Length-Aperture correlations

Trace length and aperture or vein width data are plotted side by side to illustrate the positive correlation between these attributes over 4 orders of magnitude (**Fig. 8c**). A linear scale length vs. aperture scatterplot in **Fig. 9a** shows that the data are clustered towards the origin, reflecting the greater frequency of smaller fractures expected for a power-law distribution (Vermilye and Scholz, 1995, Olson, 2003, Schultz et al., 2008). The plot of logarithmic length vs. logarithmic aperture in **Fig. 9b** shows two clusters of data which correspond to the mesoscale population (larger datasets in the centre of the figure) and the microscale population (bottom left dataset). Small aperture mesoscale data are poorly resolved, plotting at either 0.01 or 0.05 mm due to the effect of using the thickness comparator in the field. In the distribution plots, this artefact is removed conventionally by only plotting the highest cumulative frequency for each aperture value. In contrast, however, in the aperture vs. length plot each individual data point of the cloud is statistically equally important, although this results in increased uncertainty at lower aperture values. The logarithmic plot for veins only (triangles in **Fig. 9b**), showing a clear positive power-law relationship between aperture and length, has less pronounced artefacts and permits an appraisal of the relationship between these two parameters. Line fitting methods suggest a slope of 0.65 or larger with a  $R^2$  of 0.75 (red line in **Fig. 9b**) for all fracture data in this study. A comparison of veins (triangles) with other fractures including joints (grey dots in **Fig. 9b**) might further suggest that veins tend to be shorter for any given aperture.

## 7 2D population analysis

### 7.1 2D sampling locations

The 2D analysis was conducted at sub-regional scale on a bathymetric map from the near offshore (**Fig. 10a**) and on the mesoscale using a photograph of a large rock pavement outcrop **Fig. 10b**) to provide quantitative assessments of fracture connectivity and self-similarity. The offshore data provides access to a much larger area compared to onshore, however, the nature of the fractures



themselves can only be constrained by extrapolation from adjacent onshore exposures. We chose to perform 2D analysis on these areas for two main reasons. First, both contain large numbers of fractures spread over a large plan view area and therefore were most likely to provide a statistically meaningful analysis using different 2D methods (e.g. circular scanline windows and box counting). Second, the difference in size between the two areas gives an insight into fracture scaling properties. The fracture interpretation of the bathymetric image enabled analysis of the fracture length distribution for comparison with the 1-D results, and a topological fracture network analysis of the fracture nodes.

The bathymetry map used for this study is a high-resolution multibeam dataset provided by MeyGen Ltd (IXsurvey Ltd, 2009) in the area between St. John's Point and Stroma Island where the Devonian rocks are exposed on the sea floor which has been washed clean by the action of strong water currents (**Fig. 10a**, raw image in Supplementary Data File). The largest structures in the bathymetry were interpreted by the data providers as fault and fractures. Moreover, the ENE-oriented structures have the same strike as minor faults and fractures observed on the coastal platform at St. John's Point (and Stroma Island). The northernmost and longest lineament aligns well with a small bay where we observed intense faulting (and folding) related to mineralized, sinistral ENE-striking faults (classified as Group 3 structures). The N-S oriented structures are most likely to be reactivated faults (Dichiarante, 2017).

A similar 2D analysis was carried out using a mesoscale photograph taken at Brims Ness (location in Fig. 1b and raw image in Supplementary Data File). Distortion effects were minimized by analysing a single photo taken orthogonally to the outcrop pavement and by conducting the analysis in a circular area to avoid edge distortions. These structures are thought to be associated with dextral reactivation of the Bridge of Forss Fault (BFF) and, based on their similar style, associated mineralization and kinematics are inferred to be the same age as the Group 3 structures dated as Permian by Dichiarante et al. (2016).

#### *7.1.1 2D Fracture patterns*

Interpreted faults from the bathymetric data show ENE-WSW and NNW-SSE orientations. ENE-WSW trending faults dominate in this region (see SJ rose diagram in **Fig. 5a**) and show corridor arrangements in the sense of Questiaux et al. (2010). The orientations of these faults are comparable

to the two main fault sets seen onshore in locations such as St. John's Point (**Fig. 5c**). NNW-SSE trending faults are regularly spaced (100 to 200 m) in the central part of the area, while the ENE-WSW trending faults are present across the entire survey. The latter set show two different spacing values: less than 100 m for the shorter structures and about 1000 m for larger structures.

The Brims Ness photo shows three different sets of fractures: N-S, NE-SW and WNW-ESE trending (**Fig. 10b**). The N-S and NE-SW trending structures form the majority of the fractures. Most fractures have straight traces and crosscut each other. Three larger WNW-ESE and NNE- to NE-trending faults were detected. A single curved WNW-ESE trending fault was also identified (**Fig. 10b**).

## 7.2 Fracture topology results and fracture connectivity

The bathymetric topology is comprised of 698 I-, 123 Y- and 117 X-nodes, respectively (yellow, cyan and red squares in **Fig. 10a**) whilst the outcrop topology is composed of 916 I-, 240 Y- and 202 X-nodes, respectively (yellow, cyan and red squares in **Fig. 10b**).

I-type nodes are regularly distributed in the area while Y- and X-type nodes mainly occur in the central part of the bathymetry map, where longer ENE-WSW trending faults occur (**Fig. 10a**). X- and Y-type nodes, which contribute most to connectivity of the 2D system, are mainly localized where the ENE-WSW trending faults crosscut NNW-SSE trending structures.

The number of connections per line ( $n_{C/L}$ ) and number of connections per branches ( $n_{C/B}$ ) are respectively 1.18 and 1.1 for the bathymetry image, and 1.53 and 1.22 for the outcrop analysis (on a scale value between 0 and  $\infty$  for  $n_{C/L}$  and between 0 and 2 for  $n_{C/B}$ ). This indicates low overall connectivity for the fracture systems exposed in 2D. The  $n_{C/L}$  is also shown in a ternary I-Y-X plot (inset in the bottom left of **Fig. 10 a and b**).

For the bathymetry dataset, the nodal density map shows that the large majority of nodes are aligned along a series of ENE-WSW trending faults (**Fig. 11a-b top**). The density map shows that Y and X-nodes are mainly associated with NNW-SSE trending faults and are responsible for producing most of the connectivity of the system (**Fig. 11a-b bottom**).

### 7.3 Assessing self-similarity on 2D maps

Circular scanlines were performed to investigate the connectivity of specific smaller areas of the fracture network on the bathymetry map and mesoscale outcrop photograph (44 and 22 circular scanlines carried out, respectively – see **Fig. 12**). Circular scanline windows of three different diameters were used. The numbers of X-, Y- and I-nodes for each scanline are plotted in the ternary diagrams: blue for small, orange for intermediate and green larger scanlines. The data generally spread out from the centre of the ternary plot (**Fig. 12a** right and **Fig. 12b** right) and the overall data-spread is clearly unrelated to the size of the performed scanlines.

Box counting methods (Bonnet et al., 2001) were performed in the red-boxed areas shown in **Fig. 12a** at the mesoscale and regional scale to assess whether there is self-similarity in the 2D fracture patterns (**Fig. 12a-b**). Box counting assesses the presence of fractures in 2D squares of increasing size and the box dimension should be more than 1.0 but less than 2.0 (Hirata, 1989). The normalized population plot in **Fig. 12c** shows a self-similarity over 1 order of magnitude for both the bathymetry (**Fig. 12c**, red) and the mesoscale datasets (**Fig. 12c**, blue). The box dimension obtained at the two different scales of analysis were -1.77 for the outcrop photograph and -1.81 for the bathymetry map (**Fig. 12c**). Both best-fit curves yielded  $R^2$  values of 0.99. The almost identical slopes of ca -1.8 show that the 2D spatial distribution of fractures sampled at the two different ranges of scale, almost three orders of magnitude apart, is the same within the resolution of the box-counting method.

## 8 Discussion

### 8.1 Self-similar fault and fracture scaling

Fracture attribute analyses are often conducted on field outcrop analogues because they can provide useful information to bridge the gap between faults imaged in geophysical datasets (e.g. seismic reflection profiles) and fractures observed in borehole data. Our findings show that the aperture distributions for individual datasets – particularly at the mesoscale – the whole sample is often best described by a log-normal distribution. While it is difficult to unequivocally fit a power-law due to sampling bias (truncation and censoring as discussed in Section 3.1.1) our new MLE approach, which progressively truncates and censors samples until the best fit emerges, shows that a power-law distribution can provide an at least equally valid, and oftentimes better, description of the data.

571 When our data are combined from microscale to regional scales, a power-law distribution of  
572 fracture aperture and trace length attributes emerges over 4 and 8 orders of magnitude, respectively  
573 (**Fig. 8c**). Variability in the fracture intensity level (y-axis intercept) and in the slope is particularly  
574 apparent for the aperture and length datasets at the mesoscale (**Fig. 8c**). This could be attributed to  
575 more natural variability at this scale resulting from local factors such as the proximity to major  
576 structures, lithology control or sampling. It might also be because this is the scale we have sampled  
577 the most (highest number of transect datasets). When viewed on the multiscale plot, the effect of this  
578 variability at a given scale is reduced as the plots all sit close to a power-law slope of just less than  
579  $-1.0$  (**Fig. 8c**) for both aperture and length. We suggest this approach of assessing the scaling of  
580 attributes over a large-scale range to help reducing uncertainty due to variability in individual  
581 datasets. If we are correct, then it implies that, at different magnifications (or scales), the dataset  
582 structure can be interpolated to other scales within that range. Our findings are in general agreement  
583 with Hooker et al. (2014) who found for a large number of sandstone-hosted opening-mode fractures  
584 that the aperture scaling exponent (slope) was  $0.8 \pm 0.1$ .

585

586 Mechanical stratigraphy at different scales is known to affect the aspect ratio of faults, limiting  
587 their vertical size and increasing layer-parallel growth; strata-bound opening mode fracture aperture  
588 are more likely to be log-normal (e.g. Gillespie et al., 1999). Known mechanical stratigraphic  
589 boundaries for Devonian rocks in Caithness relative to individual datasets are included in **Fig. 8a** (e.g.  
590 cm scale beds at mesoscale), but they do not seem to affect the distribution plots, suggesting that it is  
591 not unreasonable to use power-law distributions to describe these data. The absence of a stratabound  
592 signature is consistent with the host rock being well cemented during the deformation (which is also  
593 consistent with Clair – see section 2.2). Previous studies (Odling et al., 1999) of fracture length over  
594 many orders of magnitude (1 cm to 1 km) from the comparable Devonian sandstones in the Hornelen  
595 Basin (Norway) showed that, while individual datasets show log-normal distributions, the collective  
596 datasets are reasonably well described by a power-law distribution. Their 2D (when normalizing the  
597 data by area) exponent overall for joint lengths was  $-2.0$  which would be equivalent to a value of  
598  $-1.0$  if normalised by length therefore in agreement with our study.

599

600 A number of previous studies has used data collected at one scale to extrapolate to another (for  
601 example Odling, 1999; Odling et al., 1999; Marrett et al., 1999; Hooker et al., 2009). Clearly, caution  
602 should be applied when using datasets acquired at a given scale to estimate a fracture attribute on  
603 other scales. Censored data might bias the choice of distribution function that best-fit the data  
604 suggesting that log-normal may seem more appropriate even when this is not the case in reality.

605 However, by extending the scale observation (i.e. by applying a multiscale approach), we reduce the  
606 potential effects of censoring, truncation and variability due to individual datasets on the overall result  
607 and also extend the estimation range for the size parameters such as length, aperture and intensity.  
608 The multiscale approach, together with the analysis of truncated individual samples, has enabled us  
609 to be more confident in concluding that both single- and multi-scale populations follow a power-law  
610 distribution.

611  
612 Although, our result remains to be tested with more datasets, the correlation we observe  
613 between aperture and length (**Fig. 9**) can provide a basis for a good estimation of frequency and  
614 fracture attributes for large-scale (regional) fractures (see next section). The scaling exponent (0.65)  
615 between aperture and length falls between the sub-linear ( $\text{exp} = 0.5$ ) and linear scaling ( $\text{exp} = 1.0$ )  
616 expected for opening mode and shear fractures (faults) respectively (Olson, 2003; Schultz et al.,  
617 2008). It is however consistent with suggestions that variations from theoretical values for the fracture  
618 length/aperture relationship are caused by interactions between segments (see Olson, 2003;  
619 Mayrhofer et al., 2019).

620

## 621 8.2 *Applications to offshore fractured reservoirs*

### 622 8.2.1 *1D prediction for reservoir volumetrics*

623 In this section we use our analogue scaling data to predict intensity, kinematic aperture, and length in  
624 the Clair subsurface reservoir making use of a fracture model published by Coney et al. (1993). This  
625 enables us to illustrate how the analogue fracture scaling relationships established onshore can be  
626 applied as to estimate sizes and intensities of fractures in subsurface reservoirs (**Fig. 13**). The early  
627 fracture model was based on well and aeromagnetic data collected for Clair by Coney et al. (1993).  
628 They identified three hierarchical orders of fractures arranged in corridors (defined as closely spaced  
629 sub-parallel fractures sets) in the Clair Group spaced at 30 - 35 m, 100 - 200 m and 1 - 1.5 km. We  
630 regard these corridors as being equivalent to the fracture clusters we observe in our data. We assume  
631 the Clair fracture aperture and length attributes scale in a comparable way to the Caithness analogue.  
632 Aperture data collected by Franklin (2013) and observed on drillcore 206-13z on Clair cores (see  
633 example in Supplementary Data File) broadly supports this assumption. Our 1D analysis indicated  
634 fracture clusters at 3 m spacing and 600 m and our 2D maps show corridors (clusters) spaced at sub  
635 100 m, between 100 and 200 m and 1000 m. This shows that fracturing in the analogue system is  
636 behaving in a similar manner to Clair with hierarchical clustering at metre, decametre and kilometre

scales. We plot the Coney et al. (1993) fracture corridor spacing on the analogue Caithness 1D scaling curves shown schematically on **Fig. 13** taking into account the sub-linear length/aperture relationship established above. The reservoir spacing values (inverse of fracture intensity) provides a predictive constraint on the fracture size attributes (light grey regions) that might be expected. Predicted fracture lengths of 30 – 60 m length (for fractures with 30 – 35 m spacing), 100 – 150 m length (for fractures of 100 – 200 m spacing) and 1 – 2 km length (for fractures with 1 - 1.5 km spacing) are directly constrained by the Caithness data (dark grey are in the plot). Values of kinematic aperture can similarly be estimated. Fractures with 30 – 35 m spacing measured in our field analogue have estimated apertures of about ca. 8 cm. For more widely spaced Clair structures (more than 100 m), values of aperture can be extrapolated by extending the Caithness aperture slope to larger scales (light grey area in **Fig. 13**). For example, for fractures spaced 100 - 200 m and 1 - 1.5 km, average aperture or fault width is estimated to be 0.8 to 1 m and 10 - 20 m, respectively (light yellow lines in **Fig. 13**). The uncertainty on these estimates is large (see width of shaded area on plot) and more data is needed to provide estimate errors, nonetheless the result provides an insight into the possible fracture intensity and apertures in the subsurface reservoir.

#### 8.2.2 2D prediction of permeability distribution

The analysis of 2D datasets using the nodal counting method has shown low connectivity for the overall systems due to the dominance of I-type nodes compared to Y- and X-type. It has been suggested that areas of poor -or no-exposure, greatly increase the level of subjective bias during the collection of fracture data (Andrews et al., 2019). These exposures issues tend to introduce more I-nodes and decrease the estimate of connectivity. Regions of relatively higher connectivity are localized at the intersection between larger and smaller structures. The high connectivity is specific to certain areas of the 2D fracture network where fracture corridors intersect at sub-regional scale or fractures cluster at mesoscale. If fluid transport is correlated with fracture trace connectivity, we should expect the permeability and fluid transport properties within the 2D network also to be heterogeneous.

In the analogue bathymetric map dataset, we observed 1D spacing ranges similar to those observed by Coney et al. (1993) for the Clair Field. We recognize 100 – 200 m spacing for NNW-SSE trending faults and less than 100 m and 1 km for ENE-WSW trending faults. Connectivity results from the bathymetry data have shown that these fractures are locally well-connected in plan-view and

scanline analysis results have shown that these fractures are potentially permeable with kinematic apertures of about  $10^{-1}$  m to 10 m producing, in the latter case, corridors of partially open fractures where these are clustered. These localized regions are believed to provide most of the connectivity of the 2D system and fluid-flow, which is consistent with the distribution of mineralization observed in the field along corridor-type structures (e.g. the White Geos Fault locality described by Dichiarante et al., 2016).

The combination of the connectivity information in plan-view derived from the bathymetry map and the fracture dip information (see **Fig. 5**) derived from fieldwork shows that fracture corridor structures and fracture intersections will be useful in constraining the main fluid-flow anisotropy that should be considered when developing an effective drilling strategy. In general, the calculated steeply plunging fault/fracture intersections would seem to favour horizontal drilling as opposed to vertical drill orientations (**Fig. 14**).

### 8.2.3. *Application of multiscale analysis to equivalent subsurface reservoir*

Our study shows that a multiscale 1D and 2D data analysis of the Orcadian Basin analogue provides a useful insight to aid understanding of the fracture patterns in a sub-surface reservoir (in this case the Clair Field). The size and scaling of aperture and length play an important control on reservoir permeability (e.g. Odling et al., 1999; Olson, 2003; Mäkel, 2007). Our mesoscale description of the aperture scaling and multiscale description of fracture length, together with the aperture/length relationship, provides a useful constraint on the 1D fracture size distributions and enables us to estimate the kinematic aperture of the largest fractures in the analogue system even though we have not sampled them directly. In doing this, we have made assumptions about the nature of the large-scale fracture corridors in the Orcadian Basin which also applies to the Clair Field. The uncertainty arises in the degree to which the largest-scale structures, most likely to be faults, have a kinematic aperture in the sense of the mesoscale structures. This assumption would need to be rigorously tested using available core and image log data. The extent to which these larger fractures are likely to only be open or partially open in the sub-surface (e.g. Laubach, 2003) needs careful consideration in any application of the analogue to the subsurface fluid flow predictions. If the decametre to kilometre length fractures are faults rather than opening mode fractures, then their contribution to fluid flow may not be significant. We note that an open fracture with 14 cm kinematic aperture was recognized in drillcore from well 208-8 in the Clair Field (Franklin, 2013). A brecciated and mineralised fault with a width of 20 cm was recorded by Dichiarante (2017) in core 206-13z from Clair (see Supplementary Data File). Fracture fills of the kind seen in Caithness (and Clair) are not

703 always bad for the hydrocarbon potential of a fractured reservoir. Wall rock fragments, early fracture-  
704 hosted hydrothermal minerals, and fills of younger porous sediment all have the ability to act as  
705 natural proppants that hold fractures open in the long term and counteract the tendency for the present  
706 day stress field to close open fracture networks in sub-surface reservoirs (Holdsworth et al., 2019,  
707 2020). These fracture fills will however reduce permeability dramatically from the ‘cubic law’  
708 relationships of ideal parallel-sided open fractures (Nelson, 1985; Laubach, 2003). It seems likely  
709 that both the opening mode fractures and faults are capable of transporting fluids, therefore justifying  
710 the application of the analogue scaling relationships.

711

712 When diagenetic or other fracture fill is present, the spatial and connectivity properties have  
713 a more important impact on rock permeability (Philip et al., 2005). Our 1D fracture size analysis is  
714 extended by the 2D approach that captures fracture interaction, clustering and connectivity to describe  
715 map-scale spatial variability of the system. These relationships can be directly applied to the Clair  
716 Field and other equivalent sub-surface reservoirs by calibrating the fracture size populations from  
717 drill core and image log data, the spatial properties from seismic attribute data, and the fracture fills  
718 from core description.

719

720 Finally, the methodology we employed in this study may be applied in a range of geological  
721 contexts ranging from hydrocarbon exploration, geothermal reservoir analyses, carbon capture and  
722 deep radioactive waste disposal facilities (e.g. see Primaleon et al., 2020). The straightforward  
723 multiscale approach allows direct comparison between analogues and sub-surface targets and is easy  
724 to apply to different areas, dataset-types and scales to provide important constraints for reservoir  
725 modelling and prediction at regional scales.

726

## 727 **9 Conclusions**

728 The Devonian rocks of the Orcadian Basin in Caithness provide a plausible analogue for the main  
729 reservoir in the Clair Field and other equivalent offshore fractured reservoirs hosted in similar tight  
730 sandstone strata. We applied a multiscale fracture analysis methodology as an alternative to the use  
731 of single-scale datasets to characterise the fracture attributes. We advocate an extended approach that  
732 integrates datasets collected at different scales and combines 1D and 2D analysis. In our example,  
733 statistical analysis provides a useful insight into the nature and scalability of the natural fracture  
734 networks. Specifically:



- Our 1D analysis has shown that the population distribution of length and aperture of the onshore datasets are best described using truncated power-law distributions.
- The multiscale approach shows scale-invariance. The scalability of single dataset can be extended from 1 - 2 orders of magnitude (single plots) to up to 4 and 8 orders of magnitude (side by side plots) for aperture and trace length, respectively. This illustrates that the multiscale approach improves the confidence using power-law scaling distributions to describe natural fracture systems.
- The correlation between fracture aperture and length is well represented by sub-linear power-law scaling (exponent = 0.65) over 4 orders of magnitude. Although, this remains to be tested with more microscale datasets and understanding of the large-scale fault properties, we suggest that this methodology provides a good estimation of fracture attributes and their scaling properties.
- Using the normalised spatial correlation approach, we detected fracture clusters at 3 m and 600 m. In 2D, we observed fracture corridors spaced at < 100 m, 150 m and 100 m.

An associated topological 2D analysis has provided the following additional insights:

- Box counting methods have shown the self-similarity of fracture counts over about 1 order of magnitude for at bathymetry- and outcrop-scales. The datasets have almost identical slopes showing that the fracture arrays over different scale ranges have the same 2D spatial distribution which confirms the hierarchical scaling of fracture clusters from the spatial analysis.
- The overall connectivity of the 2D system is low and very similar on the two scales of observation studied. However, connectivity is highly variable in the system and appears to be mainly focused along fracture corridors or clusters at a largescale and on the longer structures at the mesoscopic outcrop scale.

Our study demonstrates how a multiscale 1D size distribution and 2D spatial analysis of an onshore analogue may provide a better understanding of fracture scaling in a geologically equivalent subsurface reservoir, in this example the Clair reservoir. The method allows a prediction of fracture or fault size and their clustering properties. The spatial information, in particular, together with fracture geometry (i.e. dip data) provides important constraints on the possible permeability anisotropy. The nature of fracture fills and insitu stress should also be considered when planning exploratory drilling or modelling the reservoir.

769 **Data availability**

770 Fracture data and results of topological analysis are available at [doi:10.15128/r1cv43nw819](https://doi.org/10.15128/r1cv43nw819)

771

772 **Author Contributions**

773 AD designed and conducted the research, interpreted the data and prepared the manuscript. KM  
774 assisted with data analysis and manuscript preparation. RH designed the study and assisted with  
775 manuscript preparation. TB assisted with data analysis, ED assisted with data collection and analysis.

776

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781

782 **Supplement**

783 A supplementary data file containing the statistical method and raw images used in the analysis is  
784 available at <http://xxxxxxxxxxxxxx>.

785

786 **Competing Interests**

787 The authors declare that they have no conflicts of interest.

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1022 **FIGURES**

1023 Fig. 1: (a) Location map of the North Sea with the outline of the Orcadian Basin (light blue area). (b)  
1024 Schematic geological map of the North Scotland showing the interpreted fault lineaments by Wilson  
1025 et al. (2010) and the trace of the regional scale transects (WTR1 and WTR2) and the location of the  
1026 sub-regional transects (DO and SJ). (c) Example of Landsat aerial image showing the trace of the  
1027 sub-regional scale transect at Dounreay (DO). (d) Oblique view of the platform at Castletown. The  
1028 meter ruler shows the trace of the transect CTr1 (mesoscale). (e) Outcrop photograph of the NE-SW  
1029 fault zone where the sample for the thin-section SK04 was collected (yellow star), (f) thin-section  
1030 photograph (top) with an example of one of the microphotographs showing one fracture. The trace of  
1031 the scanline is shown by a continuous red line. CW = Cape Wrath, GGF = Great Glen Fault, fr =  
1032 fracture, SK = Skarfskerry.

1033 Fig. 2: (a) Diagram summarizing how the aperture of a fracture is related to its morphology, aperture  
1034 and fill and the general influence of an imposed stress. (b-e) Different fracture aperture and fill types  
1035 associated with oil in the Orcadian Basin. (b) Photomicrograph of open fissure with oil fill and wall  
1036 rock fragments, Thurso Bay foreshore; (c) photomicrograph of partial calcite fill with vuggy oil fill,  
1037 Dounreay; (d) photomicrograph of oil-filled brecciated calcite in dilational jog, Dounreay; (e)  
1038 Outcrop photo of calcite and red sandstone fill of inferred Permian age, Skarfskerry foreshore (see  
1039 **Fig. 1**). All thin sections are taken in plane polarized light, with scale bar = 1mm.

1040 Fig. 3: (a) Synthesis of 1D and 2D methodologies to estimate fracture attributes: (i) scanline sampling  
1041 (or transect), (ii) window sampling, (iii) circular scanline window and (iv) box counting method  
1042 (modified after Zeeb et al., 2013). L = box counting size, l = box size grid.

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1044 Fig. 4: Population distribution plots for (a) exponential (linear-logarithmic axes), (b) log-normal  
1045 (logarithmic-linear axes) and (c) power-law (logarithmic-logarithmic axes) distributions with relative  
1046 best-fit equations (top) and sketch of physical meaning (bottom). Note Hooker et al (2013) proposed

1047 an alternative classification for stratabound that they termed ‘height bounded’. In the distribution  
1048 plots, datasets are shown as black diamonds and typical best-fits are shown as red dashed lines. (d)  
1049 Examples of density maps showing higher connectivity where Y- and X-nodes occur, (e) ternary plots  
1050 showing that the overall system shown in Fig. d is isolated and (f) self-similarity plot method from a  
1051 (iv).

1052 Fig. 5: Rose diagrams of fracture orientation data for the transects at (a) regional scale, (b,c) sub-  
1053 regional scale and (d,i) mesoscale. Locations are given in Fig. 1. Note that the mesoscale transect  
1054 trend is corrected to be the same as the transects at sub-regional scale (dashed blue lines in rose  
1055 diagrams). (j) Ternary plot providing an estimation of the different type of fracture branches  
1056 intersecting each transect. N = number of fractures, MAX = maximum, CI = 95% confidence interval.  
1057 Lower hemisphere equal area projections of measured offshore data at (k) St. John’s Point and (l)  
1058 Caithness. Note that the best-fit of faults and fractures data collected onshore at St. John’s Point  
1059 (yellow diamond in the top stereonet) is consistent with the best-fit of fault and fractures data collected  
1060 in Caithness (yellow diamond in the bottom stereonet, Dichiarante, 2017). MAX = maximum density,  
1061 MEAN = mean density.

1062 Fig. 6: Cumulative distribution plots of (a) fracture and fault trace length for transects at (left)  
1063 microscale, (centre) mesoscale and (right) sub-regional scale and (b) kinematic aperture for transects  
1064 at (left) microscale, (right) mesoscale. On the plots reported stratigraphic layer thicknesses are shown  
1065 as grey boxes. The Ham-Skarfsskerri Subgroup (177m) and the Latheron Subgroup (114m) from  
1066 Anders et al. (2016) are shown on the sub-regional scale plot, The Lower Stromness Flagstone (5m)  
1067 on the sub-regional scale and mesoscale plots. At mesoscale plot the average thickness of beds (ca.

1068 20cm) is also plotted. On the microscale plot, the thickness of individual laminae (ca. 0.3mm) is  
1069 shown. Dashed lines and number refer to values discussed in text.

1070 Fig. 7: Fracture intensity and standard deviation as function of fracture number for (a) sub-regional  
1071 scale, (b) mesoscale and (c) microscale transects. Fracture intensity is unstable for a relatively small  
1072 number ( $< 20$ ) of detected fractures (grey area).

1073 Fig. 8: Cumulative frequency plots of (a) fracture length and (b) fracture aperture. (c) Side by side  
1074 population distribution plots of length (right side of the plot) and aperture (left side of the plot). Note  
1075 that the distance between the datasets in different localities (down to mesoscale) represents the  
1076 relationship in terms of order magnitude between aperture and length. Add information about  
1077 thickness of Devonian rock on Fig. 8a.

1078 Fig. 9: (a) Length vs. aperture scatter plot and (b) Log of length vs. log of aperture for veins (triangles)  
1079 and other structures (circle). Linear regression for veins on logarithmic plot is shown (dashed red  
1080 line). Fig. 10: (a) 2D analysis of bathymetric data from the area between St. John's Point and Stroma  
1081 Island with lineament interpretation and I-, Y- and X-nodes, rose diagrams of lineaments and ternary  
1082 plot of node-types proportions. b) 2D analysis of outcrop pavement photograph with lineament  
1083 interpretation and I, Y and X nodes and ternary plot of node-types proportions. MAX = maximum  
1084 density.

1085 Fig. 11: Lineament and density maps of nodes for (a) the bathymetry fault network and (b) the fault  
1086 network in pavement. All nodes density map (top) Y, X- type nodes density map allowing a  
1087 qualitative assessment of connectivity (bottom).

1088 Fig. 12: (a left) 2D topological map of bathymetric data and (b left) 2D topological map of outcrop  
1089 pavement photograph (Brims Ness) showing box counting area and example of performed circular  
1090 scanlines. Ternary plots of circular scanlines performed on (a right) bathymetric data and (b right)

1091 outcrop scale photograph. Note in the ternary plot from the bathymetry data the 22 circular scanlines  
 1092 resulted in 16 distinct proportions of I-, Y- and X- nodes. Box counting method applied to (c)  
 1093 bathymetric data and (d) outcrop scale photograph. (e) Logarithmic-logarithmic scale plot showing  
 1094 the result obtained from the maps in d and e. Data are normalized by box size and number of fractures.

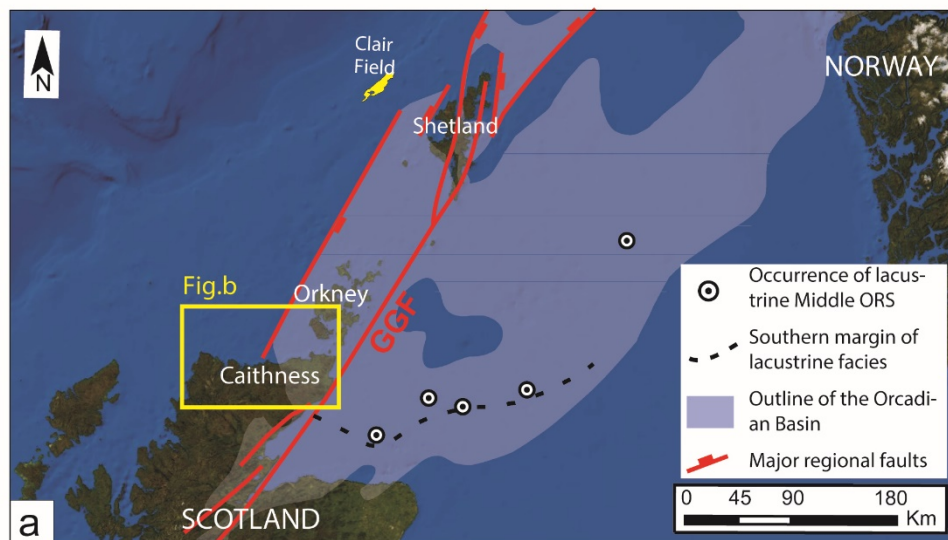
1095 Fig. 13: Sketch of the side by side population distribution plots of fracture lengths and apertures from  
 1096 **Fig. 8c**. The dark grey areas represent the region where all the aperture (left) and length (right) plots  
 1097 are localized. Coloured lines represent the distributions at each scale. Orange horizontal lines  
 1098 represent the reported spacing values for Clair (Coney et al., 1993) and yellow vertical lines represent  
 1099 the relative estimated aperture values using trends from this study. Note that we extrapolate the  
 1100 aperture (light grey area) using the slope derived from the microscale and mesoscale datasets.

1101 Fig. 14: Schematic block diagram created by combining offshore 2D density map of connectivity and  
 1102 onshore dip values

1103 **Table 1:** Basic parameters, definitions and equations provided by 1D and 2D methods (Zeeb et al.,  
 1104 2013 *modified*)

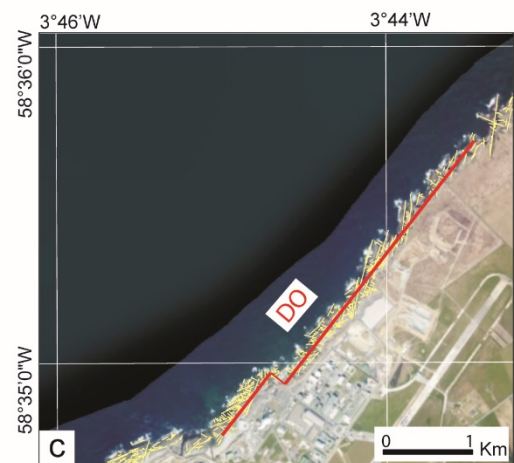
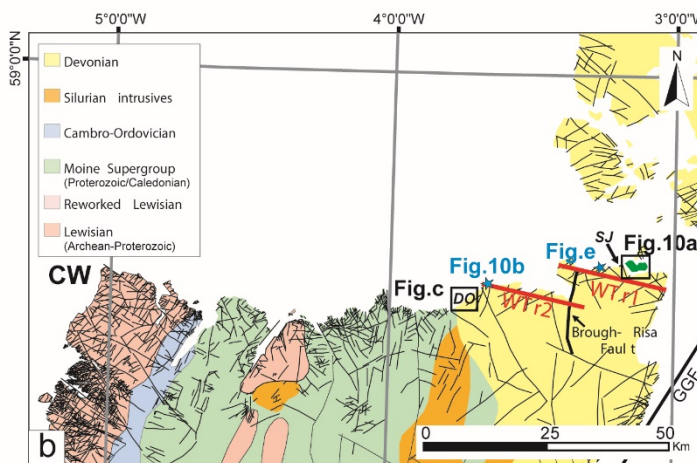
1105 **Table 2:** Transect data. GPSs = GPS position of the starting point,  $N$ = total number of sampled  
 1106 fractures,  $J$ = joint,  $V$ = vein,  $FnI$ = Fracture without infill,  $T$ = tensile,  $Dx$ = dextral,  $Sn$ = sinistral,  $I-I$  =  
 1107 "isolated branch", delimited by two I-nodes.  $I-Y$  and  $I-X$  = "singly connected" branches, delimited by  
 1108 one I-node and one Y or X-node.  $YY$ ,  $YX$  and  $XX$  = "multiply connected" branches, delimited by two  
 1109 Y or X-nodes or one Y and one X-node.

1110



Regional Scale

Sub-regional Scale



Mesoscale

Microscale

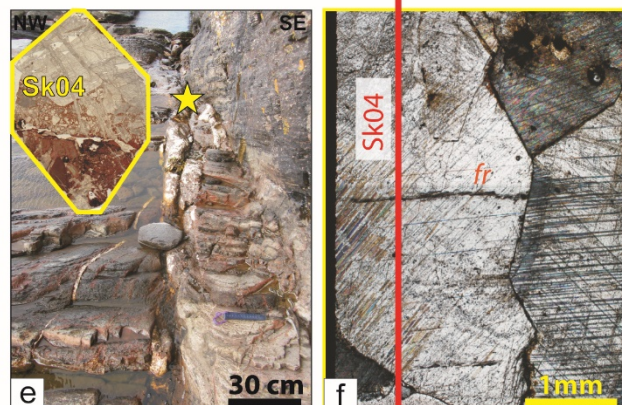


Fig. 1



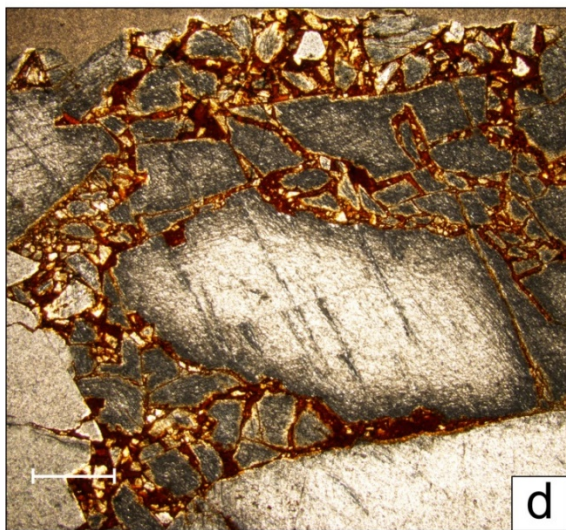
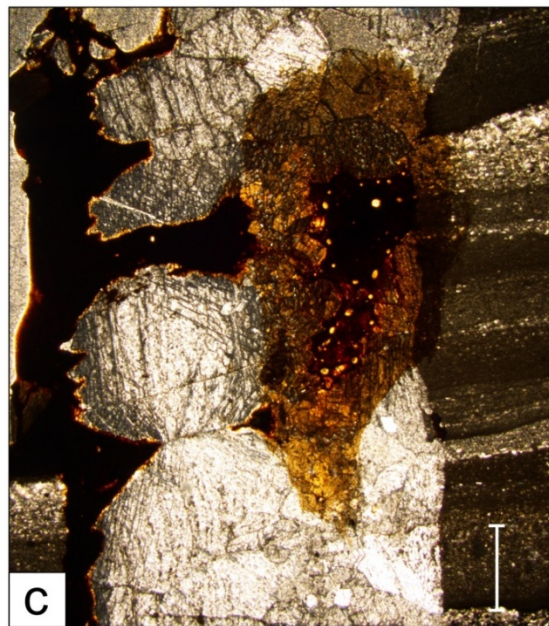
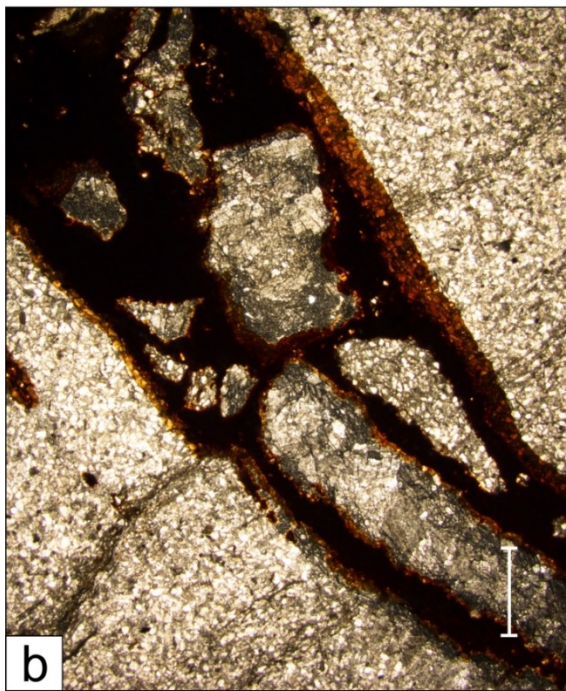
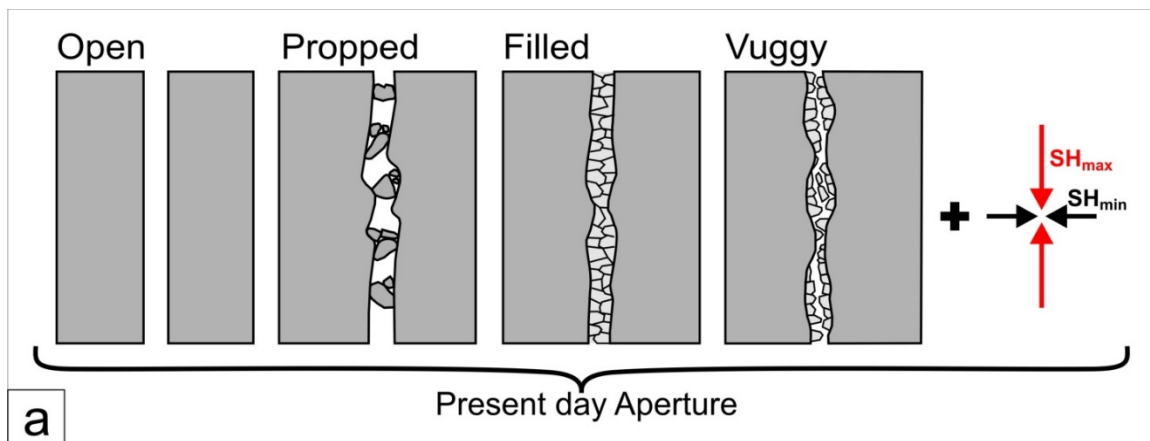


Fig. 2

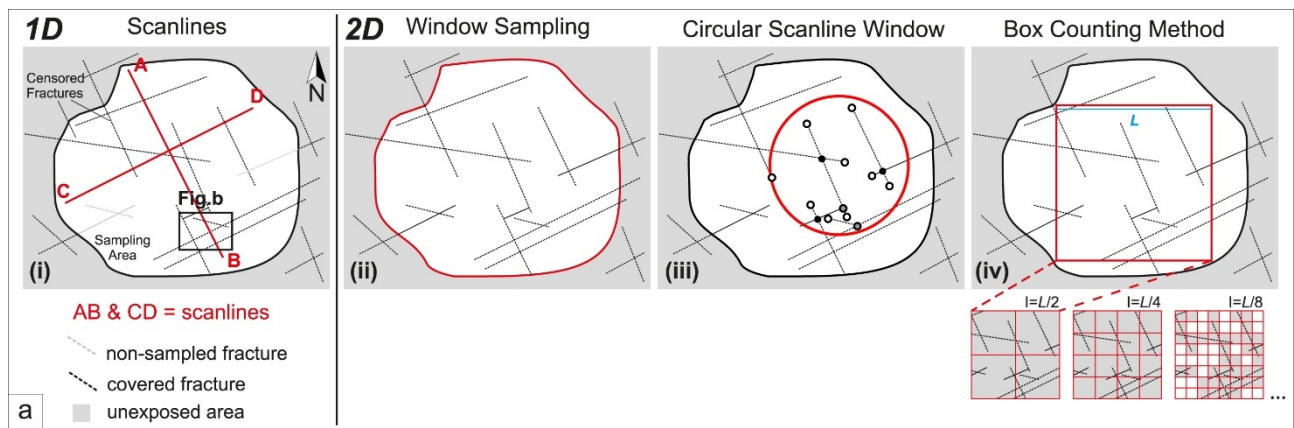


Fig. 3



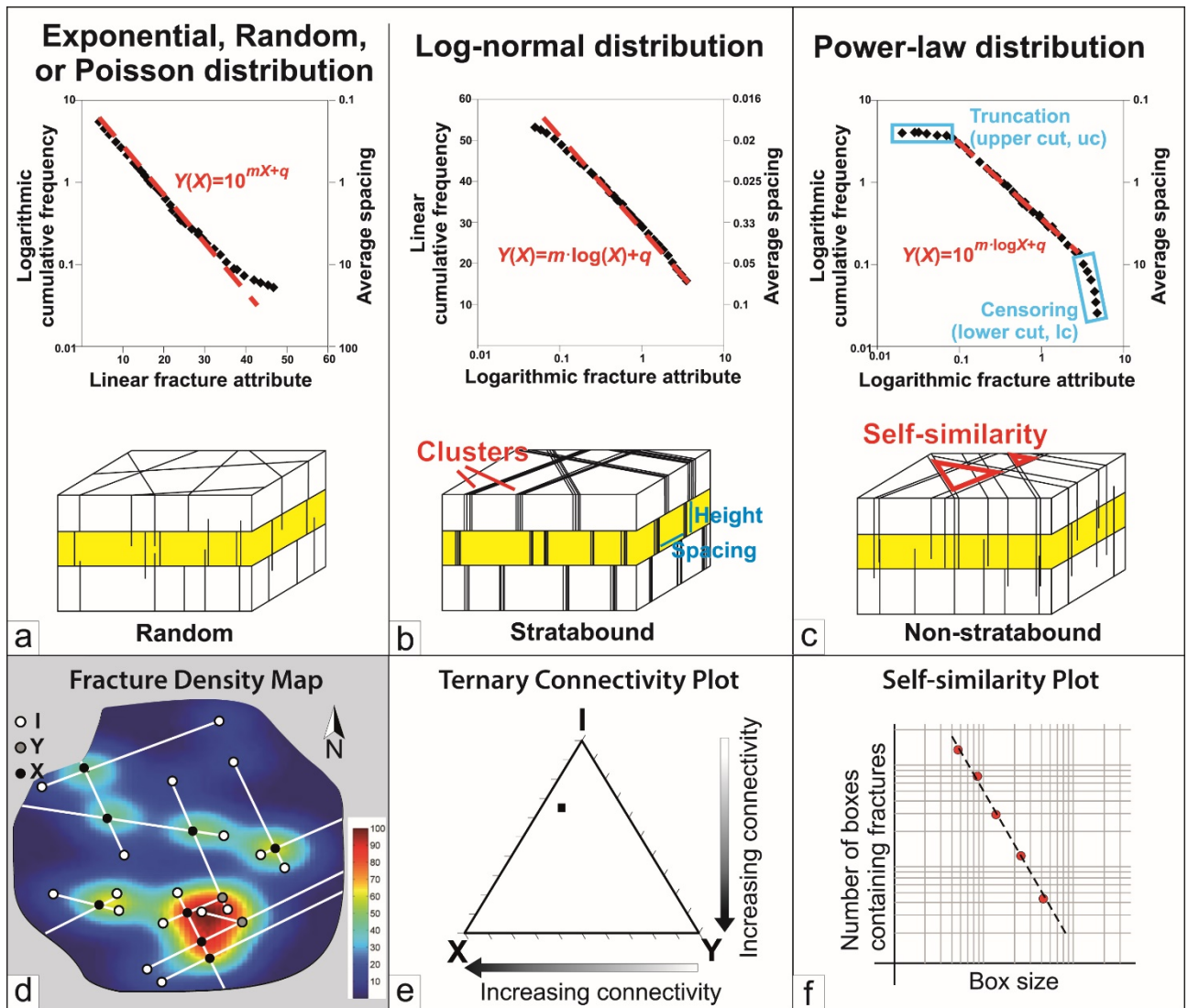


Fig. 4

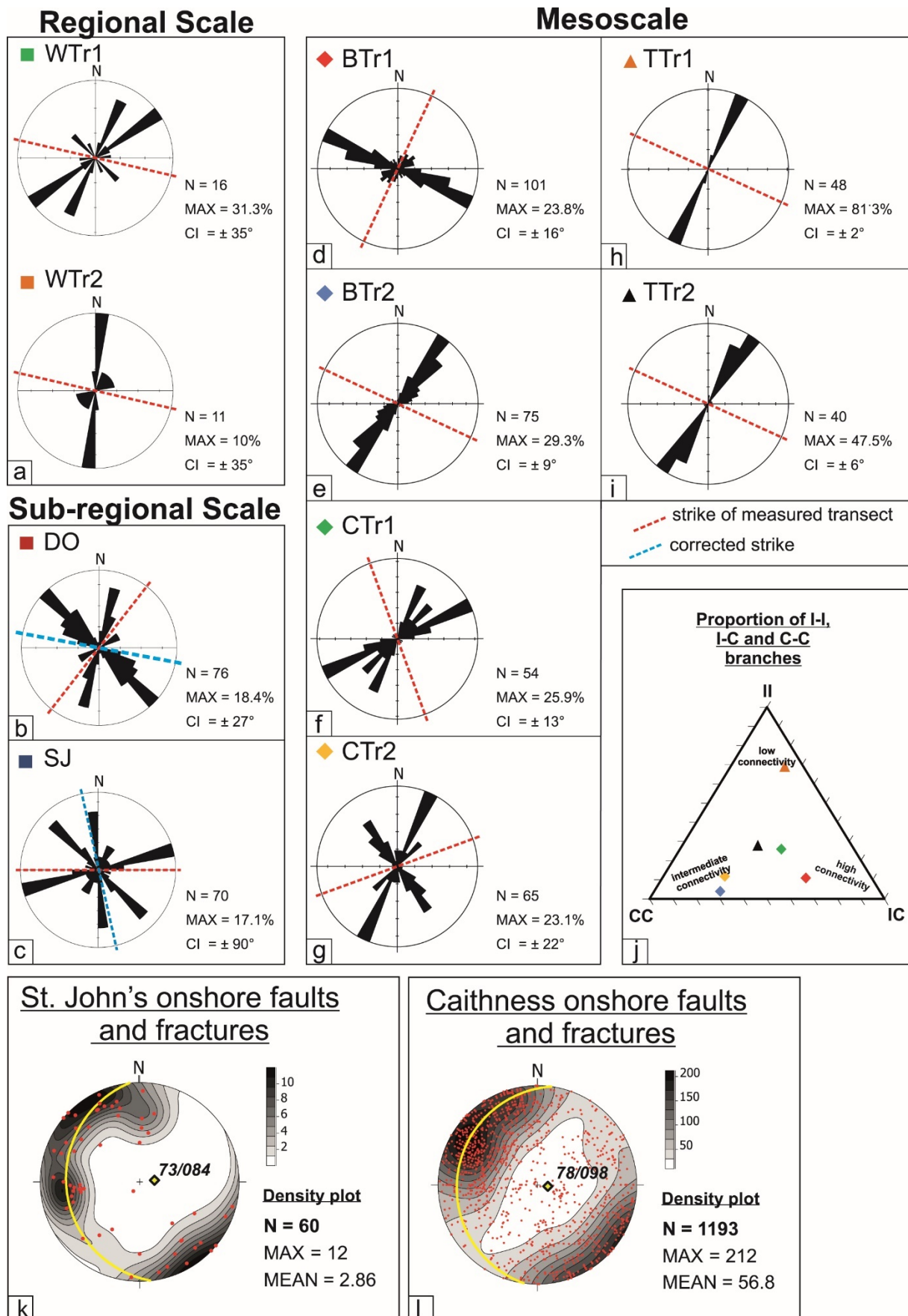


Fig. 5

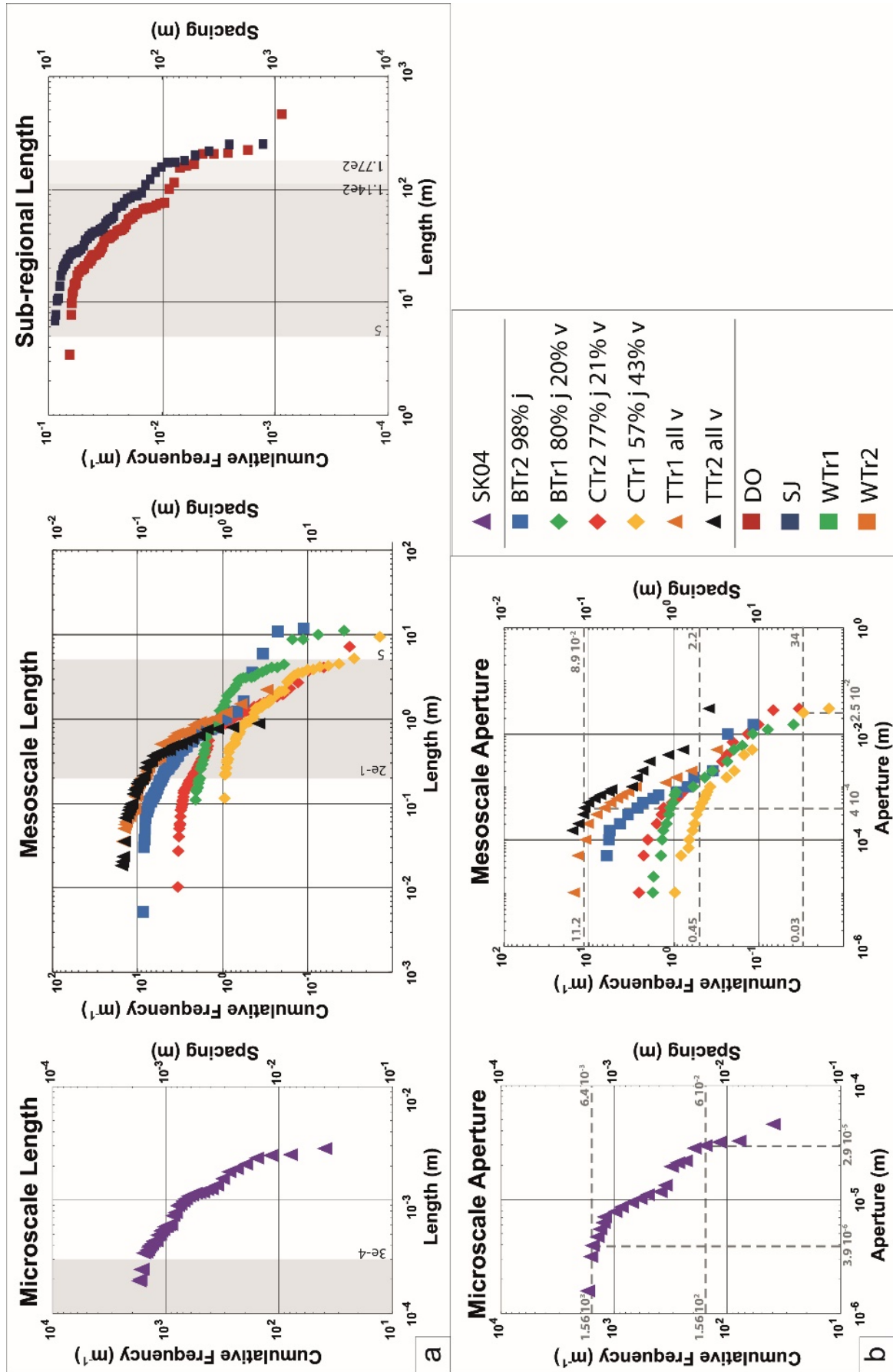


Fig. 6

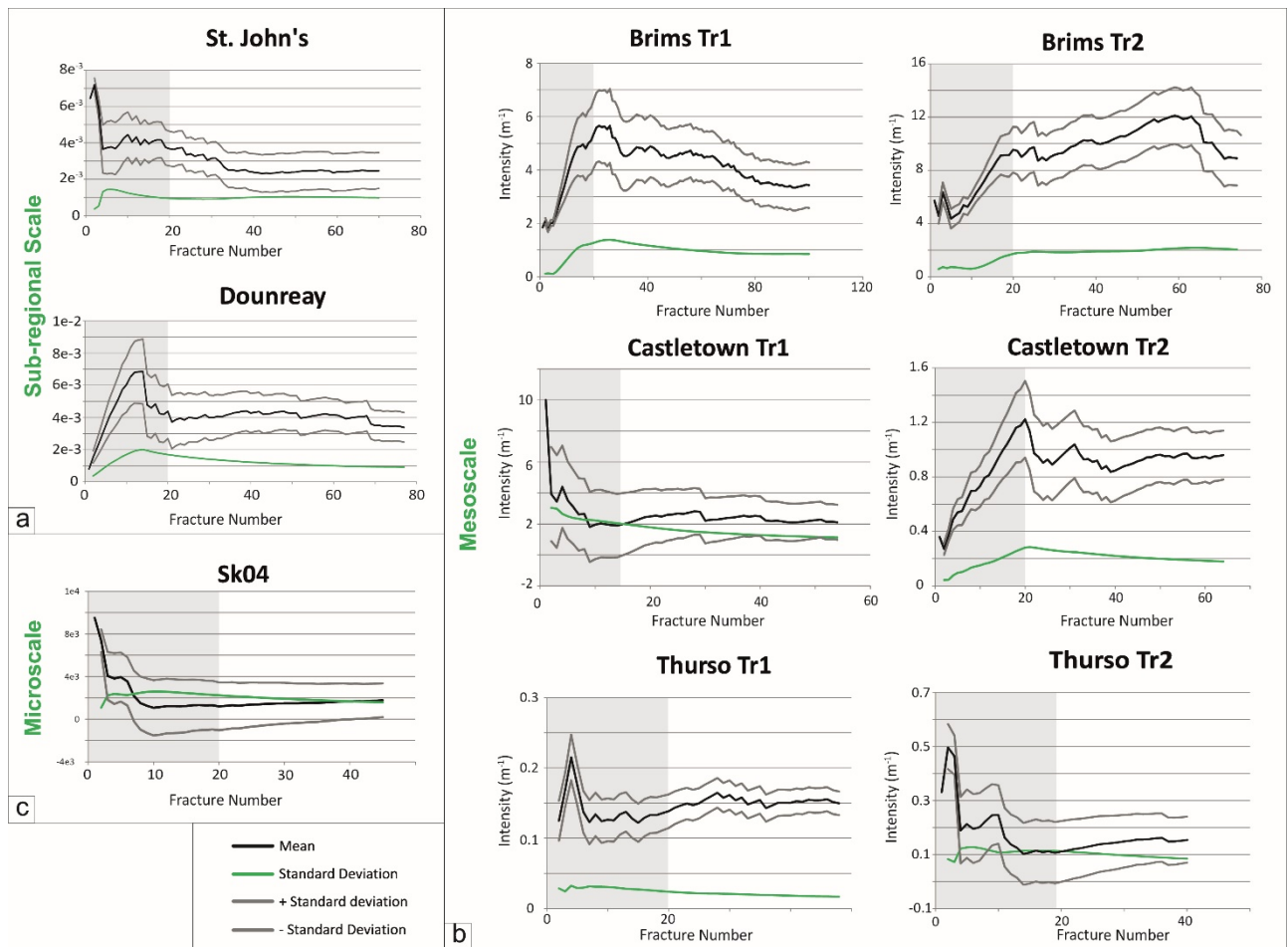


Fig. 7

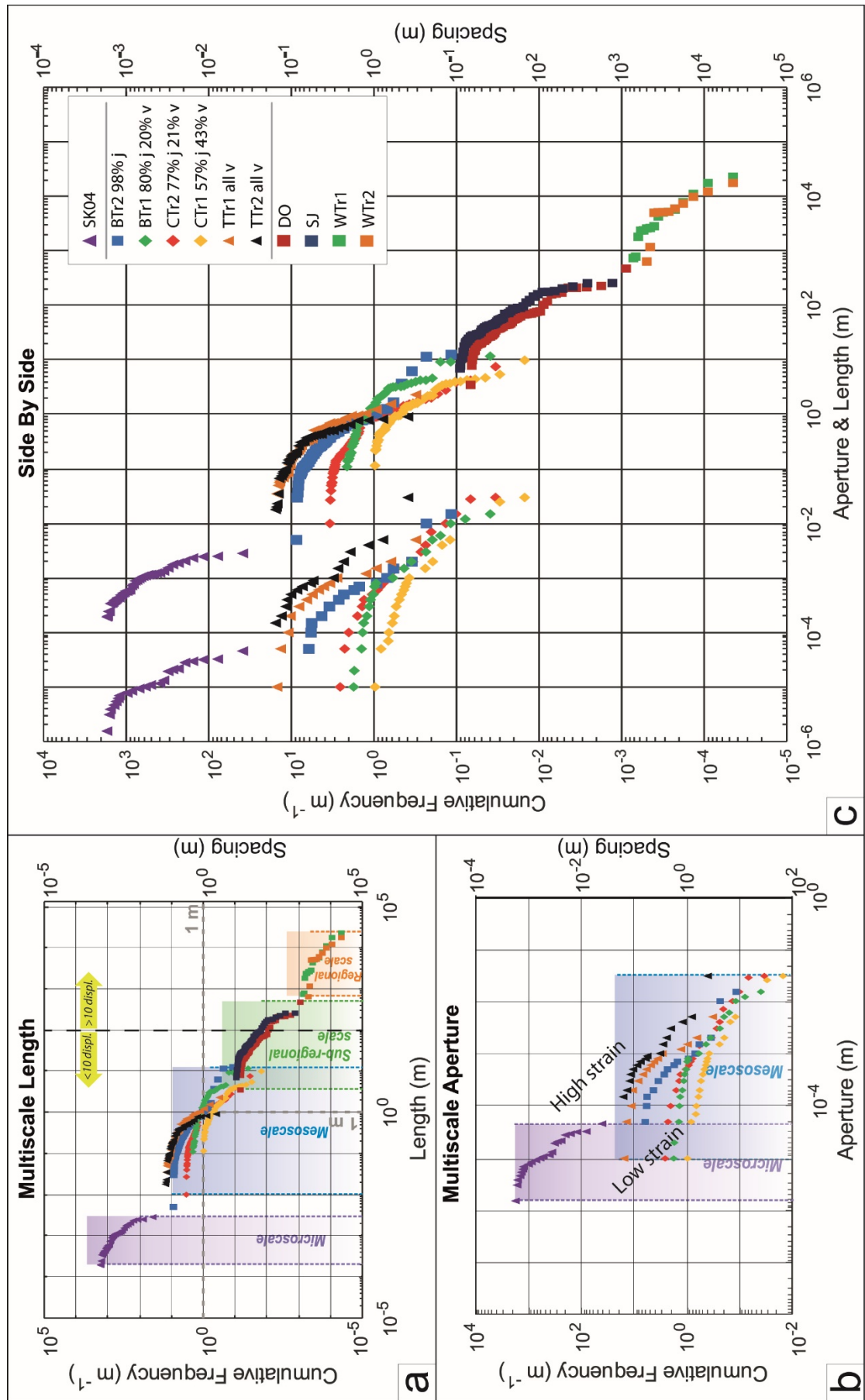


Fig. 8

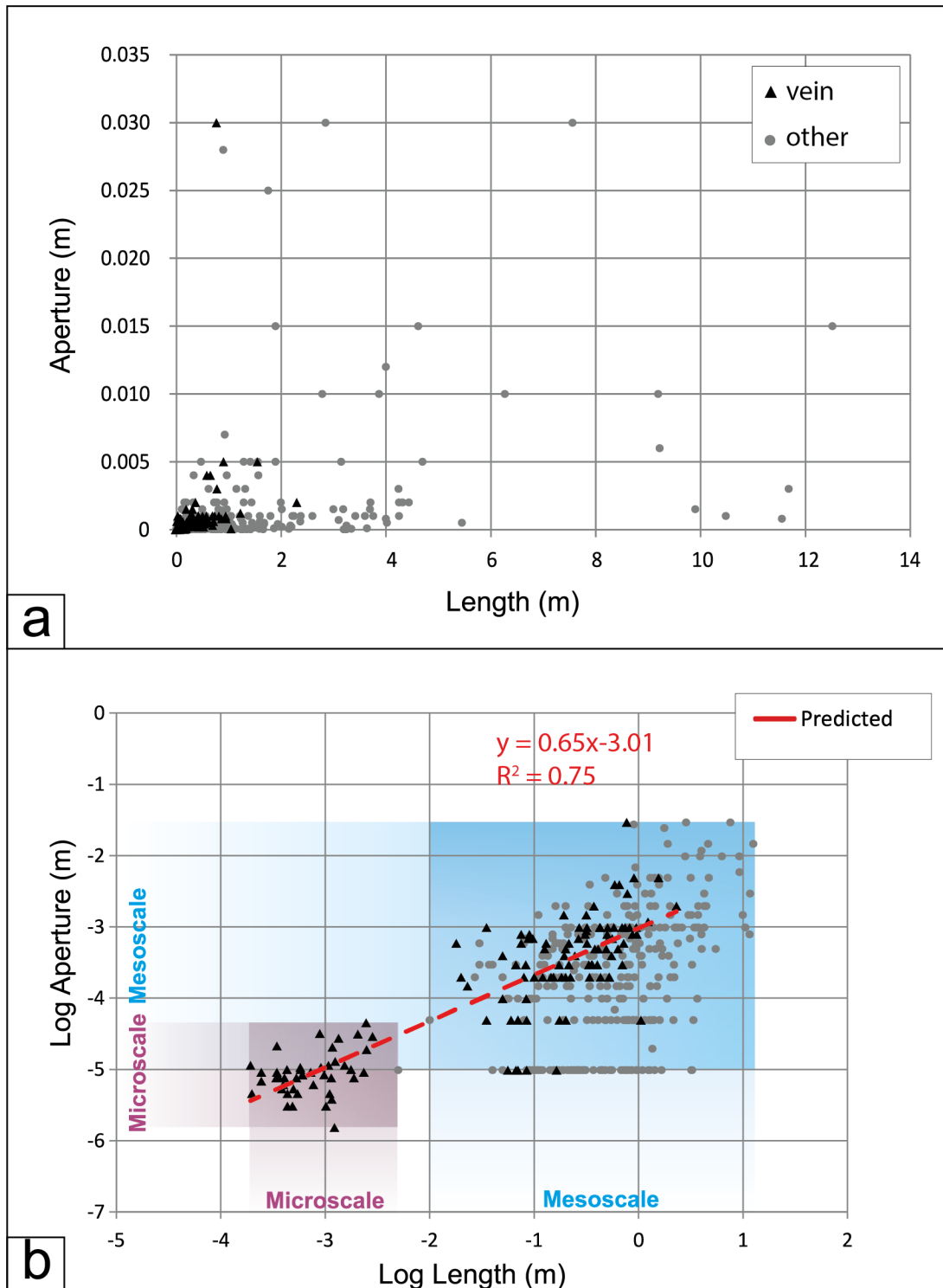


Fig. 9



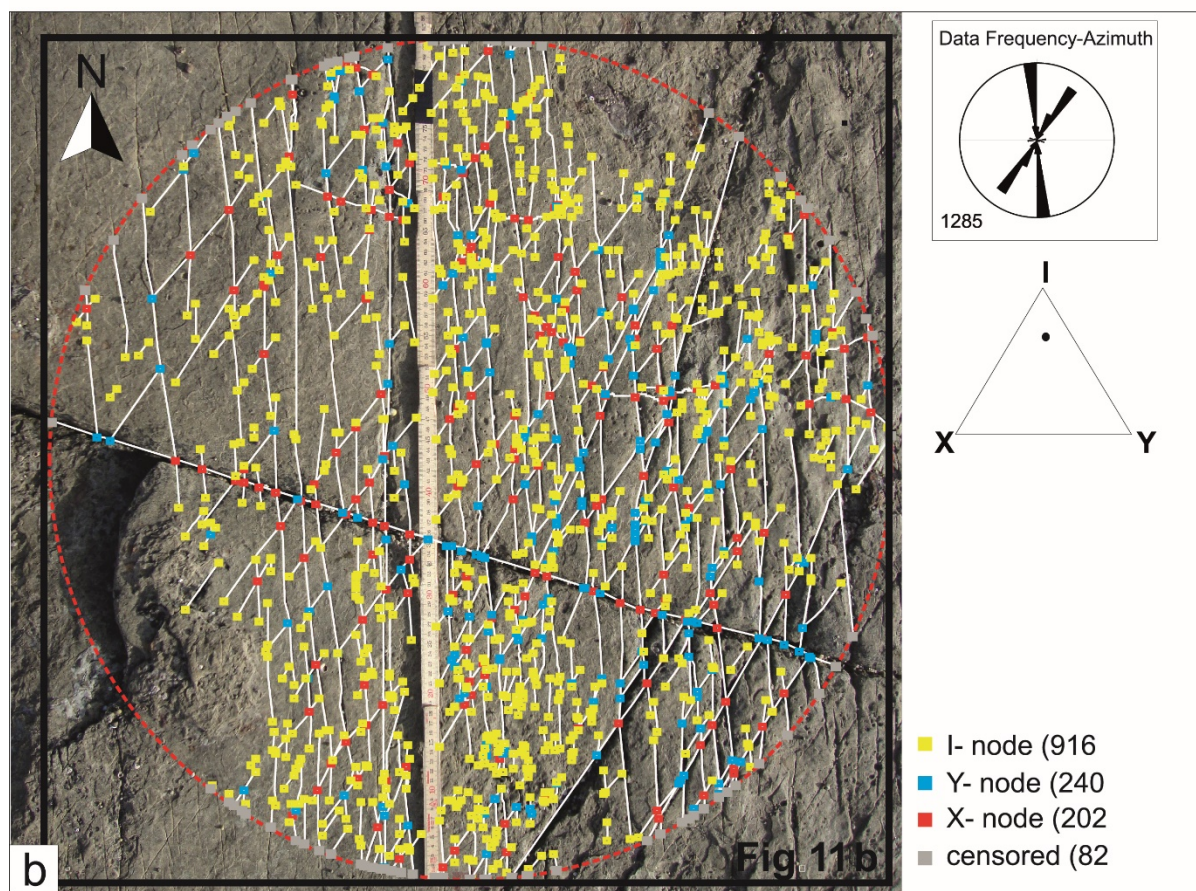
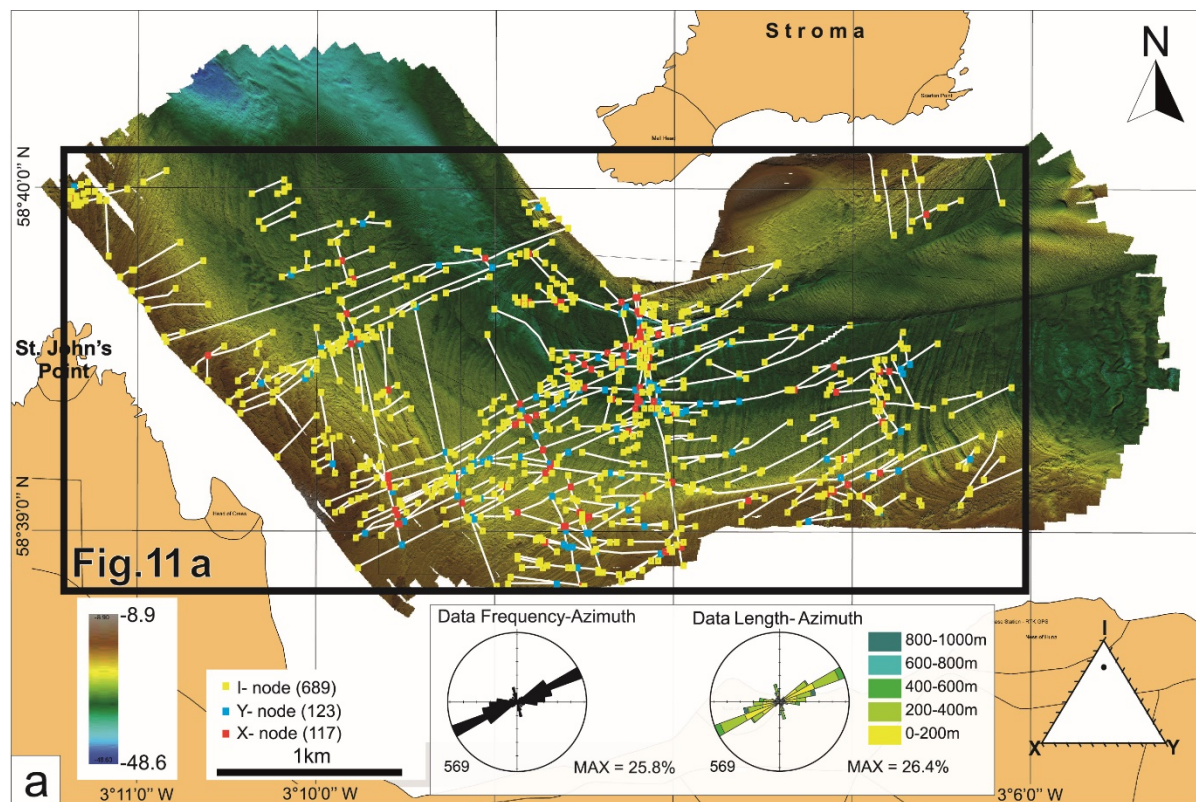


Fig. 10



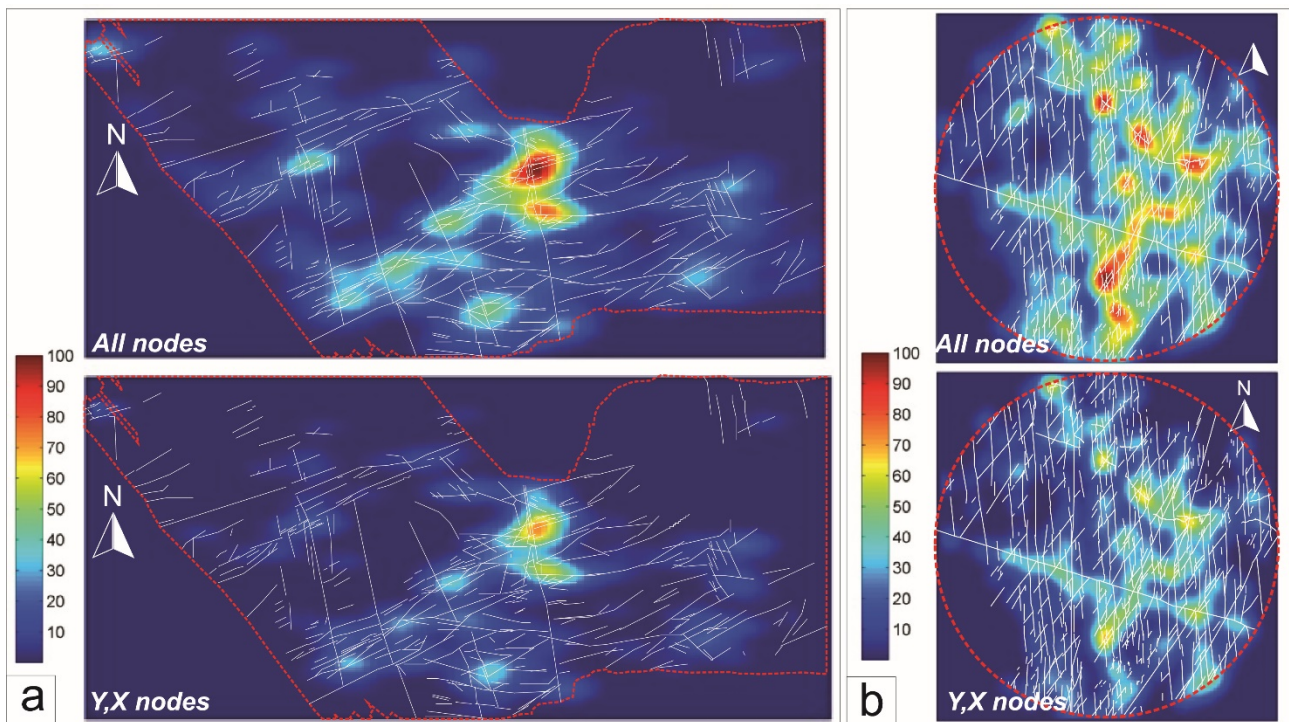


Fig. 11

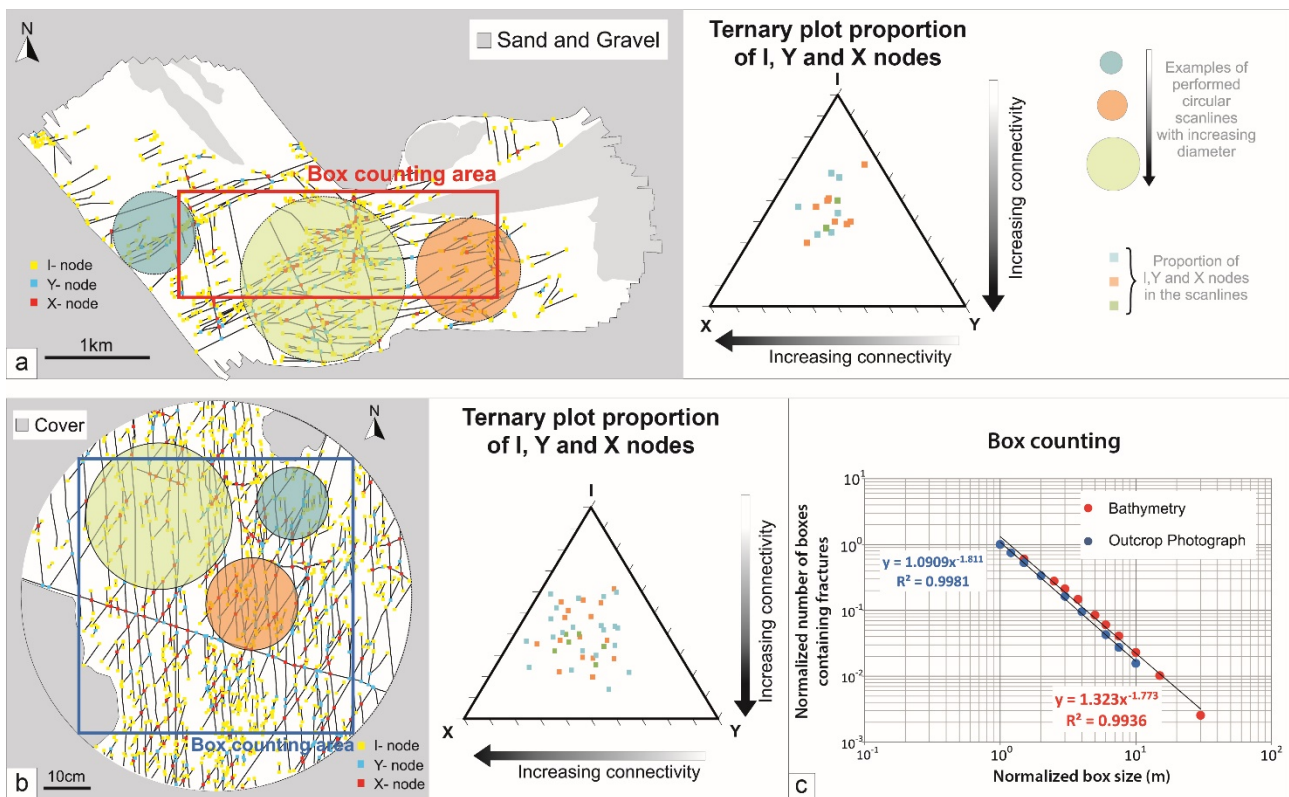


Fig. 12



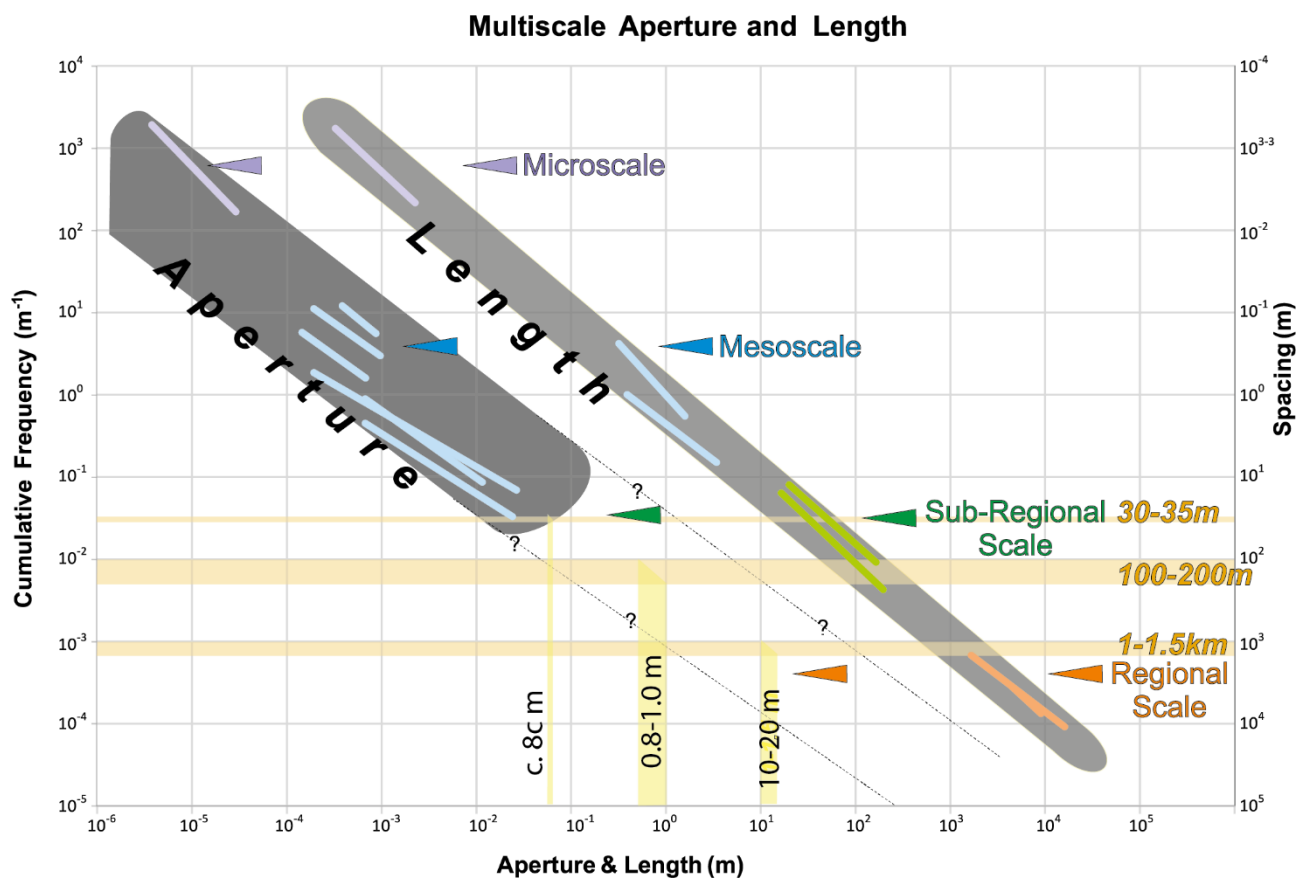


Fig. 13

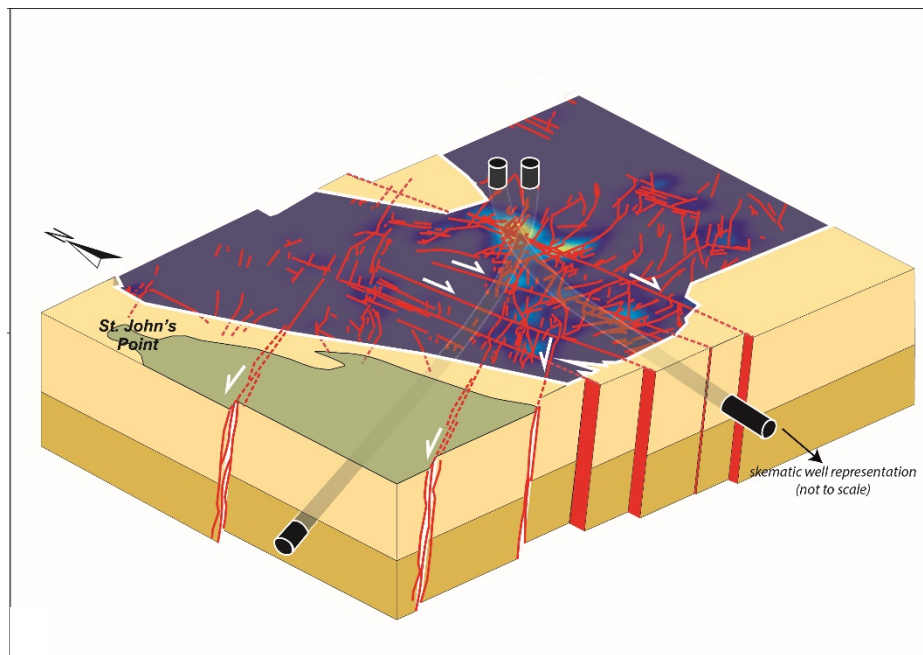


Fig. 14

Parameter	Definition	Scanline Sampling	Window Sampling	Circular Scanline	Box Counting Method
Orientation	Orientation of a fracture on a sampling plane (1D) or sampling volume (3D)	YES	YES	-	-
Spacing ( $s$ )	Spacing between consecutive fractures [m] (1D)	$S = l/l$	-	-	YES
Length ( $l$ )	Length of fracture intersecting the scanline (1D) or sampling area (2D)	YES	YES	-	YES
Aperture ( $a$ )	Aperture of fracture intersecting the scanline (1D)	YES	YES	-	-
Intensity or Frequency ( $I$ )	Number of fractures ( $N$ ) per unit length ( $L$ ) [ $m^{-1}$ ] (1D)	$I = N/L$	-	-	-
Density ( $D$ )	Number of fractures ( $N$ ) per unit area ( $A$ ) [ $m^{-2}$ ] (2D)	-	$D = N/A$	$D = m/2\pi r^2$	YES

Name	GPSs	$N$	Type			Kinematic			Termination					Spacing Range [m]		Length Range [m]		Aperture Range [m]	
			$J$	$V$	$Fnl$	$T$	$Dx$	$Sn$	$IY$	$IX$	$YY$	$YX$	$XX$	From	to	From	to	From	to
WTr1	ND1835175022	16	-	-	-	-	-	-	-	-	-	-	-	$3.7 \cdot 10^3$	$3.4 \cdot 10^2$	$2.3 \cdot 10^4$	$7.4 \cdot 10^2$	-	-
WTr2	ND0305471126	11	-	-	-	-	-	-	-	-	-	-	-	$3.8 \cdot 10^3$	$1.78 \cdot 10^1$	$1.8 \cdot 10^4$	$6.4 \cdot 10^2$	-	-
DO	NC9834067080	76	-	-	-	-	-	-	-	-	-	-	-	$2.6 \cdot 10^2$	0.8	$4.8 \cdot 10^2$	3.5	-	-
SJ	ND2931274823	70	-	-	-	-	-	-	-	-	-	-	-	$1.5 \cdot 10^2$	1.2	$2.6 \cdot 10^2$	7	-	-
BTr1	ND0432271142	99	80	20	1	94	5	2	21	7	27	19	16	1.3	$4 \cdot 10^{-3}$	7.6	$10^{-2}$	$3 \cdot 10^{-2}$	$1 \cdot 10^{-5}$
BTr2	ND0436071157	75	73	-	2	75	-	-	10	11	8	22	21	$8 \cdot 10^{-1}$	$2 \cdot 10^{-3}$	12	$5 \cdot 10^{-3}$	$1.5 \cdot 10^{-2}$	$5 \cdot 10^{-5}$
CTr1	ND1888569104	54	31	23	-	14	-	-	10	4	9	4	0	3.2	$5 \cdot 10^{-3}$	12	0.1	$1.5 \cdot 10^{-2}$	$1 \cdot 10^{-5}$
CTr2	ND1892269088	65	50	14	1	8	-	-	7	11	17	12	0	4.6	$2 \cdot 10^{-2}$	9	0.11	$3 \cdot 10^{-2}$	$1 \cdot 10^{-5}$
TTr1	ND1089969071	48	-	48	-	48	-	-	11	0	3	1	0	$2 \cdot 10^{-1}$	$3 \cdot 10^{-3}$	2.3	$3.5 \cdot 10^{-2}$	$5 \cdot 10^{-3}$	$1 \cdot 10^{-5}$
TTr2	ND1091469036	39	-	39	-	39	-	-	13	0	6	8	2	0.33	$5 \cdot 10^{-3}$	0.9	$1.8 \cdot 10^{-2}$	$3 \cdot 10^{-2}$	$1.5 \cdot 10^{-4}$
SK04	ND2613574584	45	-	-	-	-	-	-	-	-	-	-	-	$2.2 \cdot 10^{-4}$	$1.2 \cdot 10^{-6}$	$2.8 \cdot 10^{-3}$	$1.9 \cdot 10^{-4}$	$4.6 \cdot 10^{-5}$	$1.5 \cdot 10^{-6}$