

PRE-PRINT: Suitability of legacy oil and gas subsurface data for nascent geoenergy activities onshore United Kingdom

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Suitability of legacy oil and gas subsurface data for nascent geoenergy activities onshore United Kingdom

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35 **Abstract**

36 The decarbonisation of energy systems to achieve net zero carbon emissions will likely see the
37 rapid development of carbon capture and storage, energy storage in the subsurface and
38 geothermal energy projects. Subsurface data such as seismic reflection surveys and borehole
39 data are vital for geoscientists and engineers to carry out comprehensive assessments of both
40 the opportunities and risks for these developments. Here, legacy subsurface data from onshore
41 hydrocarbon and coal exploration in the UK are collated and their suitability for net zero
42 geoenergy activities, specifically geothermal. We provide a description of the spatial coverage
43 and a chronology of the acquisition of key seismic reflection and borehole data, as well as
44 examine data resolution and limitations. We discuss the implications of spatial variability in
45 subsurface datasets and the associated subsurface uncertainty. This variability is vitally
46 important to understanding the suitability of data for decision making. We examine societal
47 aspects of data uncertainty and discuss that when the same data are used to communicate
48 subsurface uncertainty and risk, the source of the data should also be considered, especially
49 where data are not easily publicly accessible. Understanding the provenance and quality criteria
50 of data are vitally important for future geoenergy activities and public confidence in subsurface
51 activities. Finally, we ask should there be minimum data collection criterion, such as resolution
52 requirements, ahead of subsurface activities with potentially significant impacts to the
53 environment, economy and society?

54 **Keywords:** data, geoenergy, onshore geology, subsurface, uncertainty, geostatistics, public
55 perception

56 .

57 1. Introduction

58 Achieving a transition to net zero carbon emissions from energy systems is one of the
59 most pressing challenges facing society globally (Rogelj et al., 2015). The UK government has
60 set a legally binding target to reduce greenhouse gas emissions to net zero by 2050 (see Climate
61 Change Act government, 2008), which, to achieve, will require decarbonising both industrial
62 and residential energy systems (e.g. Cooper and Hammond, 2018;Broad et al., 2020). There
63 will likely be a need for subsurface activities, whether as part of industrial clusters and the
64 development of carbon capture and storage (CCS) (e.g. Alcalde et al., 2019), decentralised
65 energy systems and the use of geothermal energy (Lloyd, 2018). The exploration and
66 production of shale gas has highlighted concerns not only about the compatibility with low
67 carbon energy systems and mitigating climate change (Partridge et al., 2017), but also the
68 ability to predict the behaviour of the subsurface for hazards such as induced seismicity
69 (Bommer et al., 2015). A fundamental question regarding the use of the subsurface for future
70 decarbonisation pathways is whether existing data are both extensive enough and of
71 sufficiently high-quality to adequately assess the potential contribution and impact of
72 subsurface activities and their role in a net zero future.

73 Data are fundamental to understanding the risk and uncertainties associated with
74 subsurface activities (e.g. Baker et al., 1999;Ross, 2004;Bles et al., 2019) and are used by
75 geoscientists, environmental scientists and engineers to carry out comprehensive assessments
76 of the resource or storage potential as well as risk assessment. Informed decisions, for example,
77 policy change, come from the analysis, interpretations and modelling of such data, and
78 ultimately the scientific communication of the results. Characterising subsurface uncertainties
79 is a vital part of risk management, covering operational safety, environmental and economic
80 risks, as well as being key to characterising any resource potential. Society is increasingly
81 concerned with the environmental risks and impacts of subsurface activities (van Os et al.,
82 2016b;a) which can have a direct impact on communities, for example, as the result of induced
83 seismicity (van der Voort and Vanclay, 2015), subsidence (Franks et al., 2010), environmental
84 pollution (O'Rourke and Connolly, 2003), health (Holdren et al., 2000) and rapid changes to
85 community life (Schafft et al., 2013). These impacts also have broader consequences for both
86 the public and the industries involved, for example, protests or project delays (Bradshaw and
87 Waite, 2017;Short and Szolucha, 2019) leading to very serious issues with gaining social
88 licence. As a result of the increased scrutiny with which subsurface activities have come under,
89 the need for effective communication is becoming increasingly vital to ensuring that
90 geoscientific know-how reaches all those involved and impacted (Stewart and Gill, 2017;van
91 der Bles et al., 2019). This comes at a time where the UK's (and the world's) ambition to
92 decarbonise energy systems could, despite the predicted shift from fossil fuels to new, lower
93 carbon energy sources, require subsurface activities at significant industrial scale (e.g.
94 Stephenson et al., 2019). This study synthesises the legacy oil and gas subsurface datasets from
95 the UK landmass, with the purpose of providing an unbiased view of the implications for future
96 geoenergy activities using examples for geothermal and unconventional hydrocarbons.

97 Geological data derived from sparse borehole coverage measuring geological attributes
98 indirectly are always innately uncertain. In communicating subsurface risks, experts often
99 discuss the degrees of uncertainty inherent in subsurface characterisation, however, this is often
100 without considering the target audience. Importantly, it has been shown that when experts avoid
101 (or deny) discussing the uncertainties as part of public communication that it can drive distrust
102 in science and organisations (e.g. Sjöberg, 1998;Frewer et al., 2003). One suggested
103 mechanism to improve risk communication is to focus on 'what is being done to reduce the

104 uncertainty' (Frewer et al., 2002). Nascent activities, such as the recent introduction of
105 hydraulic fracturing for shale gas in the UK, may as a result of their relative immature
106 deployment, be associated with greater uncertainty, particularly regarding the extent of, for
107 example, resources or potentially negative environmental impacts. What may have been an
108 acceptable level of uncertainty and risk in the past, or in other jurisdictions, is no longer socially
109 perceived as acceptable and, as argued by Beck et al. (1993), that disasters (or the highest
110 impact events) shape perceptions of risk. The introduction of new subsurface activities, such
111 as hydraulic fracturing or the development of CCS, may, due to their immature development
112 be initially associated with greater uncertainty, particularly regarding how far their potentially
113 negative effects extend within the subsurface (Krause et al., 2014).

114 To describe uncertainty requires a recognition that the knowledge is limited, that
115 "known unknowns" are identified, and acknowledging that there may also be "unknown
116 unknowns" (Pérez-Díaz et al., 2020). Quantifying uncertainty makes it possible to analyse how
117 interpretations might differ from reality (Pérez-Díaz et al., 2020). Ackoff (1989) defines data
118 as symbols that represent properties of objects, events and their environment, and are the
119 products of observation. The Data, Information, Knowledge, Wisdom (DIKW) model which
120 Ackoff (1989) described can be applied to subsurface data and information (**Table 1**); this
121 differentiation between data and information is somewhat subjective in many areas of
122 geosciences, specifically with respect to geophysical or remote sensing data, where processing
123 of the data are required to enable a geological interpretation or analysis. **Table 1** provides a
124 summary of typical subsurface data, information and knowledge sources. The accuracy of any
125 subsurface interpretation or analysis is dependent on the quantity, spatial distribution and
126 quality of data available. Important considerations are the data requirements for both business
127 and regulators to effectively make decisions on resources and safety criteria, but also to
128 reassure the public and assess the likelihood of an activity impacting local communities and
129 society.

130 This study describes the characteristics of legacy oil and gas subsurface datasets that
131 describe the deep subsurface of the UK landmass, which in UK legislation is defined as any
132 land at a depth of at least 300 metres below surface level (The Infrastructure Act, 2015). This
133 study synthesises these datasets and includes examples of why data resolution and quality are
134 also an important consideration for future geenergy activities and public confidence in
135 subsurface activities.

136

137 **2. Subsurface UK data**

138 The geology of the UK landmass contains a geological record all the way back to the
139 Archean, and includes a history of subduction zones, volcanic arcs, continental rifts and
140 mountain belts (Woodcock and Strachan, 2012). While extensive geological mapping of the
141 UK dates back to the 19th century and is summarised in the now famous map by William Smith
142 (Smith, 1815), it was not until 1918 that the first deliberate deep oil and gas well was drilled,
143 Hardstoft-1 in Derbyshire, to a depth of ~950m (Morton, 2014). In the period preceding the
144 Second World War (1939), there were several early seismic reflection experiments by Anglo-
145 Iranian Oil Company (Jones, 1937). From the 1950s systematic geophysical data acquisition
146 began for oil, gas and coal exploration. Seismic reflection surveys have subsequently become
147 the primary subsurface geophysical method employed for oil and gas exploration. Acquisition
148 of seismic reflection data, and the drilling of deep boreholes continued onshore in the UK, with
149 the late 1980s being the peak of onshore seismic data acquisition (see section Seismic

150 Reflection for details). Much of the current understanding of the deep subsurface onshore in
151 the UK results from data acquired for the exploration and production of oil coal and gas.
152 However, whilst these data have advanced our understanding of the geology in the UK, the
153 implications of using these data on public trust and perceptions of risk has been little considered
154 in the literature. Demonstrably independent, impartial research based on data is essential for
155 ensuring social licence.

156 This study focuses on the coverage of both seismic reflection surveys and boreholes,
157 and the spatial distribution and acquisition history of these data. Data included in this study are
158 derived from UK Onshore Geophysics Library (UKOGL), the British Geological Survey
159 (BGS), and the Oil and Gas Authority (OGA); sources are listed in **Table 2**. Detailed
160 description of these data can be found in subsequent sections, but as an overview, there are ~76
161 136km of 2D seismic reflection data, ~2400km² of 3D seismic reflection data and 2242
162 released oil and gas exploration boreholes (not including those completed in 2018, 2019 and
163 2020 as these data have not as yet been released). The BGS Single Onshore Borehole Index
164 database is the most accurate consolidated record of boreholes in the UK landmass, therefore,
165 the present analysis has used records derived from the BGS, OGA and UKOGL to provide a
166 comprehensive view. Throughout this study the term “borehole” is used regardless of depth of
167 penetration or verticality. The now outdated BGS Geothermal Catalogue (Rollin, 1995),
168 included temperature data which had to be digitised into a tabulated format from analogue
169 records (now available in scanned PDF file format available from the National Geoscience
170 Data Centre). Precision of location data was then only specified either to 10 m or 100 m.
171 Comparisons made to the UK Continental Shelf relate to data included as “Surveys as
172 Consented 2D”, available from the OGA National Data Repository. In this study, comparisons
173 have been made with data from the Netherlands (NLOG), where alongside onshore oil and gas
174 exploration and production, there is significant exploitation of geothermal energy. This data
175 which is managed by the Geological Survey of the Netherlands (TNO) on behalf of the Ministry
176 of Economic Affairs and Climate.

177

178 **3. Methodology**

179 This study describes the spatial distribution and characteristics of data available from
180 the BGS, OGA and UKOGL (**Table 2**), comprising primarily geological and geophysical data
181 collected for oil and gas exploration and production. Geological parameters and concepts are
182 often not directly observed or measured, but interpreted from these data (Pérez-Díaz et al.,
183 2020), therefore, this study considered data to be measurements in wells, and post-stack seismic
184 reflection data (**see Table 1**).

185

186 *3.1. Spatial Statistics*

187 Using geostatistical techniques (quadrant analysis, point density, average nearest
188 neighbour, and global Moran’s I), the extent and distribution of geospatial subsurface data have
189 been quantified. The results of this analysis have then been assessed in terms of the possible
190 impact on subsequent interpretation and analysis. All spatial statistics were computed in
191 ArcMap 10.6.1. The quadrant analysis point density and total line length was computed for a
192 given area of 10km by 10km area (area of 100km²). While for individual well locations point
193 density was used, in contrast for the BGS Geothermal Catalogue where individual wells have

194 more than one measurement, a quadrant analysis was used. For spatial statistics, the P-value is
195 used to assess if there is spatial pattern among the features and, therefore, the probability that
196 the observed spatial pattern was random. A small P-value is indicative of a low probability that
197 the observed spatial pattern is the result of random processes. The Z-values are standard
198 deviations, for example, for a z value of 2.5 the result is 2.5 standard deviations. The study has
199 differentiated between shallow and deep wells based on true vertical depth (TVD); a deep well
200 being defined as one completed to a depth >300m.

201 These statistics have then been used to assess the distribution of the legacy data and
202 discuss their suitability for assessing the risks and opportunities of the future use of the
203 subsurface, specifically geothermal and unconventional hydrocarbon extraction. The coverage
204 of both well and seismic reflection data have been analysed with respect to the domestic and
205 non-domestic heat demand to assess the data available for geothermal resource characterisation
206 in demand hot spots. The study has used heat demand data for the year 2009 from Taylor et al.
207 (2014). The original data are annual heat demand provided at a 1km by 1km resolution and in
208 units of kWh/km². In this study, the data were reduced, using an aggregated mean, to a 5km by
209 5km resolution to simplify the boundaries of heat demand, and then converted to MWh/km².
210 These data have been used to compare areas of heat demand to the distribution of subsurface
211 data.

212

213 *3.2. Quantitative Analysis of Seismic Reflection Data*

214 Given the recent controversy surrounding the extraction of unconventional
215 hydrocarbon resources (Williams et al., 2017), and the differing interpretations reported in both
216 peer-reviewed literature (Anderson and Underhill, 2020) and other reports published
217 (Cuadrilla, 2019), this study has carried out an analysis of the post-stack characteristics of 3D
218 seismic reflection data from across the Craven Basin in Lancashire (Figure 1), the area where
219 shale gas hydraulic fracturing activities were carried out between 2011 and 2019. The study
220 describes the characteristics of the seismic data that relate to the ability to interpret geological
221 features, for example, faults and analysed the frequency content and how this relates to the
222 vertical and horizontal resolution of data. This comprises analysis of the bandwidth and
223 acquisition parameters, which have been evaluated and described for the first time. A
224 trapezoidal Ormsby filter was used to filter frequencies content of the data to give an indication
225 of how reflector continuity is related to the dominant frequency of the data. The frequency
226 content of both the original and filtered datasets was analysed using SeisLab 3.0 (Rietsch,
227 2020).

228

229 **4. Data Analysis**

230 *4.1. Well Data*

231 There are 1 335 511 boreholes in the UK landmass recorded BGS Single Onshore
232 Borehole Index (SOBI), with the depth of investigation varying from just a few meters to >3000
233 m (**Figure 2b**). SOBI includes 334 757 boreholes have no details on depth; most boreholes
234 included in SOBI (851 963) are to a depth <30m. A further 136 650 boreholes investigate a
235 depth range between 30 and 500m. A histogram of the oil and gas exploration wells by depth
236 from the UKOGL database shows that >70% penetrate less than 1000m TVD (**Figure 2b**).

237 There are only 2885 wells deeper than 500m, comprising oil and gas exploration and
238 production wells and other deep boreholes. The OGA dataset includes only the 2242 released
239 oil and gas wells; the UKOGL database also includes an additional 643 coal boreholes. Oil and
240 gas exploration and production boreholes therefore account for less than 1% of all the boreholes
241 drilled in the UK. The spatial density of the shallow boreholes between 30m and 500m onshore
242 the UK can be seen in **Figure 3**. Despite there being in excess of 1 million boreholes, many
243 which are related to site investigations in populated areas, there are areas of the UK where there
244 are hardly any boreholes, notably west Wales and Scotland. **Figure 4** shows a series of spatial
245 point density maps for deep boreholes using data provided by UKOGL. The deepest onshore
246 well penetration in the UK is the Seal Sands No. 1 well, drilled to a total depth of 4169 m TVD
247 (Johnson et al. 2011). The mean depth of a drilled oil and gas borehole is 1152m. Of these
248 boreholes only ~151 wells, extend deeper than 2000m TVD, and just 13 are deeper than 3000m.
249 Spatially, these are not distributed equally across the onshore of the UK. Nearly all the deep
250 wells, because of being drilled for hydrocarbon exploration and production are in either the
251 Carboniferous Basins of northern England and the Midlands, or the Mesozoic Basins of
252 southern England. For the wells in the UKOGL database, nearest neighbour analysis estimates
253 a Z-score of -82.77, indicating that the data are clustered and there is a less than 1% likelihood
254 that this clustered pattern is random. Global Moran's I analysis, indicates that wells are
255 clustered with respect to depth, with a Z-score of 53.57, and less than 1% likelihood that the
256 clustering is random. A histogram of wells drilled onshore the UK by year shows that over
257 ~70% of the onshore wells in the UK were drilled prior to 1990. Since 3D seismic reflection
258 data acquisition onshore UK did not start until the 1990s, that means that all these wells were
259 drilled based on interpreted 2D seismic reflection data. As would be expected there is a spatial
260 coincidence of both boreholes and seismic reflection data. A total of 644 boreholes are co-
261 located with 3D seismic reflection data, and 1578 wells located within 100m of a 2D seismic
262 reflection line.

263 *4.1.1. Core and downhole log data*

264 The BGS maintain a database of over 10 000 onshore borehole samples, comprising a
265 range of materials including core, core samples, individual hand specimens, bulk samples,
266 unwashed cuttings, washed and dried cuttings, plugs, powders and bulk samples. These include
267 those collected as part of onshore oil and gas exploration and production borehole drilling. The
268 relative spatial density of these data can be seen in **Figure 5a** (database can be searched online).
269 The BGS hold an archive of digital geophysical downhole log data from boreholes distributed
270 across the UK which is not available to other researchers. Basic well metadata, such as location
271 spud and completion date, is also held by UKOGL, but digital log data are only available
272 through formal release agents. There is no single publicly available record of all downhole logs
273 onshore UK. The BGS hold a record of ~4541 wells with digital geophysical logs, which
274 includes both oil and gas exploration wells and other boreholes including mine gas and coal
275 bed methane wells. The spatial density of these data is shown in **Figure 5b**. In addition to the
276 BGS records of geophysical logs, well data are available through the OGA's appointed data
277 release agents, who hold an inventory of digital log data for onshore wells. There is both no
278 agreed standard log dataset agreed, and companies make choices based on specific
279 requirements, so the data set is not consistent.

280

281 *4.1.2. Temperature data*

282 The BGS Geothermal Catalogue is a published compilation of temperature and heat
283 flow measurements from across the onshore UK. **Figure 5c** shows the location of individual
284 wells with temperature measurements and **Figure 5d** shows the number of temperature
285 measurements in a 10km by 10km quadrant. Average nearest neighbour analysis returns an
286 observed mean distance of 1668m compared with an expected mean distance of 9538m. This
287 returns a nearest neighbour ratio of 0.1879, with Z-score of -60.31 and less than 1% likelihood
288 that this is random indicating that the data are strongly clustered. Global Moran's I analysis,
289 indicates that location of temperature measurements is clustered with respect to depth, with a
290 Z-score of 35.303, and less than 1% likelihood that the clustering is random. As well as spatial
291 clustering, the measurements of temperature in the boreholes are also over a limited depth
292 range. As described by Rollin (1995), there are ~2600 temperatures at over 1150 sites. Of these,
293 geothermal gradients are estimated in the dataset for ~1700 measurements. Over 90% of the
294 temperature data are from depths less than 2000m and ~27% are from a depth shallower than
295 500m (**Figure 6a**). While the dominant trend is one of increasing temperature with depth, there
296 is no simple relationship. These temperatures in the catalogue are used to estimate geothermal
297 gradients using a modified air surface temperature. These estimates of geothermal gradient
298 were not used in this study, as the method of determining land surface temperature is an
299 oversimplification and not accurate without correction. There are only 116 temperature
300 measurements from depths greater than 2000m, and as **Figure 6a** shows there is a very
301 significant vertical sampling bias.

302 An analysis of the distribution of the temperature data with respect to the domestic and
303 non-domestic heat demand in the UK (Taylor et al. 2014) finds 141 of the measurements (~8%)
304 are within high heat demand areas. **Table 3** lists the four largest areas with a heat demand
305 >10 000 MWh/km² and the associated deep data associated with each area. **Figures 9a-d** are
306 maps of London, Birmingham, Manchester and Glasgow with the location of temperature
307 measurements plotted, as well as the location of deep well and 2D and 3D seismic reflection
308 surveys over the same geographical areas. In some heat demand hot spots there are multiple
309 temperature measurements, and in some cases, these are across multiple wells. However, there
310 are areas of high heat demand with no temperature measurements in the database, for example,
311 Leeds and Greater Manchester. Although temperatures from wells drilled since the 1990 are
312 not currently captured in the BGS Geothermal Catalogue, across the areas of highest heat
313 demand, i.e. Manchester to Liverpool area (**Table 3**) there are 79 deep wells.

314

315 5. Seismic Reflection Data

316 5.1. Seismic reflection data coverage

317 The location, line length (in the case of 2D) and area (in the case of 3D) of seismic
318 reflection data onshore UK have been analysed to determine the spatial distribution of the data.
319 **Figure 8a** shows the location of all 2D seismic reflection lines. Onshore UK there are ~75
320 871km of 2D seismic reflection data which cover an area of ~100 000km². As with the deep
321 wells, it is almost exclusively in either the Carboniferous Basins of Northern England and the
322 Midlands, or the Mesozoic Basins of Southern England. The density of data varies
323 dramatically, with the maximum coverage being 700km in a single 10km² quadrant and the
324 minimum being 7km. Across the onshore sedimentary basins the greatest coverage of 2D data
325 are located across the Wessex and East Midlands Basins (**Figure 8b**). As shown in **Figure 8a**,
326 over 90% of the 2D seismic reflection data onshore UK was acquired prior to 1990. The mean
327 length of a 2D seismic line is 8.2km and the longest individual 2D seismic line is 67.4km. As

328 a comparison, in the 10 000km² offshore area of the UK East Irish Sea Basin there are 72 454
329 km of 2D seismic reflection lines; approximately 10 times the data density in one offshore
330 basin compared with the whole of the UK onshore.

331 Three-dimensional seismic reflection data onshore UK is limited to just 32 surveys
332 (**Figure 8c**) covering an area of ~2400km² (covering ~1% of the UK). As a comparison, the
333 Netherlands has a land area of ~ 41 543km² across which there is ~14 000km² of onshore 3D
334 seismic reflection data (equivalent to ~34% of the Netherlands). Onshore the UK the largest
335 onshore 3D survey is 363km², which is the Lincswold02 3D survey. Using the current (as of
336 April 2020) Petroleum and Exploration Development Licences (PEDL) outlines from the
337 OGA, there are 12 PEDL which have complete 3D seismic coverage. Presently, 114 out of 181
338 of the current PEDL have no 3D seismic coverage and 19 have less than 10% coverage. **Figure**
339 **9b** is a histogram of 3D seismic reflection area acquired by year onshore UK, and with only
340 638km² acquired since 2010. Of these surveys 5 are within the prospective shale gas
341 exploration areas identified by the BGS (Andrews, 2013). These prospective areas total ~20
342 000km, however, there has only been 452km² of new 3D seismic acquisition in these areas,
343 which amounts to ~2% of the total prospective areas.

344 When the coverage of 2D and 3D seismic reflection data are compared with the
345 domestic and non-domestic heat demand across the UK, only ~500km of the existing 2D
346 seismic reflection data intersect areas of domestic heat demand above 10 000 MWh/km²
347 annually. This is <1% of the 2D seismic reflection data. There is no 3D seismic reflection data
348 in these areas. **Table 3** summarises the coverage of data and the total length of 2D seismic data
349 and the number of wells within the ten largest areas where heat demand is >10 000 MWh/km².
350 As well as the limited availability of 2D seismic reflection data, there are also only a handful
351 of deep wells and wells with temperature measurements in these areas. **Figure 7** shows the four
352 largest areas, London, Birmingham, Manchester and Liverpool, and the coverage of deep data.
353 These data indicate that there is notable paucity of well and seismic data for geothermal
354 exploration in these areas. Whilst some PEDL licences include urban areas, active exploration
355 and acquisition of seismic data in such built up condition is practically impossible due to the
356 level of disruption and is therefore, considered uneconomic and societally unacceptable.

357

358 *5.2. Seismic reflection data quality*

359 The study has looked at the quality of 3D seismic reflection data specifically within the
360 PEDL licence where hydraulic fracturing took place at two wells between 2018 and 2019.
361 There are 43km of 2D lines across the PEDL and a single 3D seismic reflection survey.
362 Interpretations of this 3D seismic survey have been described previously with implications for
363 both exploitation of resources (Clarke et al., 2018) and for the evaluation of induced seismicity
364 (Anderson and Underhill, 2020). Anderson and Underhill (2020) recently described the
365 structural setting of the area and the implications for induced seismicity, for example,
366 geological faults below seismic resolution. Here the geophysical characteristics of the 3D
367 survey are described, focusing on the frequency content and the implications for the resolution
368 and quality of the data. **Figure 10** shows how the frequency spectrum for the 3D data varies
369 by depth (in two-way-time [TWT]) of investigation. To examine the impact of frequency
370 content on the quality of the seismic reflection data, **Figure 11** shows example seismic sections
371 of the original post-stack seismic volume (**Figure 11a**) with different high frequency cut offs
372 applied at 90 Hz (**Figure 11b**), 60 Hz (**Figure 11c**) and 40 Hz (**Figure 11d**). The difference
373 between the original and filtered seismic reflection data are shown in **Figure 12**. Filtering out

374 the high frequency component (>90 Hz) of the 3D survey (**Figure 11b**) makes almost no
375 difference to the seismic image (**Figure 12a**), aside from some high frequency noise in the near
376 surface section (upper most 500m TWT) section. Filtering out the component >60Hz removes
377 some coherent energy above 1500m, but below this there is very little difference (**Figure 12b**).
378 Filtering out >40Hz component results in removing coherent energy in the interval shallower
379 than 1500m as well as some deeper coherent energy (**Figure 12c**). In this area, the exploration
380 targets were at ~1000m. While there is overall a higher frequency content at shallower depths,
381 this does not contribute to improving the overall interpretability of the data and suggests that
382 much of the higher frequency content could be noise rather than coherent energy. Frequency is
383 a key parameter controlling the resolution of faults in seismic images. The maximum vertical
384 resolution is directly related to the ability to distinguish individual reflecting surfaces (Yilmaz,
385 2001) and in the case of the Bowland-12 survey is approximately 60 m at the target intervals.
386 For the horizontal resolution, assuming that the Fresnel zone is reduced to a small circle by 3D
387 migration (Brown, 2011), then in the case of the Bowland-12 survey the horizontal resolution
388 can be estimated to be ~40m. The frequency content of the data and resulting estimated
389 resolution means it is difficult to distinguish layers below this limit. The implications of the
390 vertical and horizontal resolution of both 2D and 3D seismic data for shale gas exploration and
391 other geenergy activities is explored in the discussion.

392

393 6. Discussion

394 Like in many countries, the acquisition of subsurface data onshore UK has been driven
395 by the exploration and exploitation of natural resources. This means that the data that exist to
396 investigate the subsurface are biased and often display clustering, as is evidenced by this study.
397 Pérez-Díaz et al. (2020) break down the process of transforming geoscientific data to
398 geological knowledge into acquisition, processing, analysis, interpretation and modelling. The
399 findings presented here show that quantification of sampling bias, data clustering and
400 underlying limitations are vital to understand prior to analysis, interpretation and modelling of
401 the data.

402

403 6.1. Subsurface Mapping and Geoenergy

404 The ability to create accurate models of the subsurface relies on data being
405 representative of the area of interest. Data acquisition in oil and gas exploration is location
406 biased, and often clustered, because it is acquired to test a geological scenario that may have
407 multiple objectives. This clustering has been demonstrated using spatial statistics. Onshore oil
408 and gas exploration wells exhibit significant clustering, as do the temperature data that are
409 frequently acquired in these wells. Of the total onshore area of the UK, i.e. ~243 000km², the
410 76 136km of 2D seismic data covers an area of ~109 900km². This means that less than half
411 of the total onshore area of the UK is covered by any subsurface image. As noted previously,
412 when compared with the offshore of the UK, where seismic acquisition is easier, the onshore
413 UK has a relative paucity of both 2D and 3D seismic reflection data and a significant deficit
414 in the relative quality of the information that can be derived from these data.

415 3D seismic reflection data cover a total of just 2400km² of the onshore UK. The limited
416 extent of any single 3D seismic survey onshore the UK limits the ability to map or extend our
417 geological knowledge and understanding. The largest onshore survey is 363km² (Lincswold-

418 02) and is approximately 30km by 12km. Similarly, the limited extent to which surveys are
419 adjacent to one another and form a patchwork from which larger areas can be mapped is in the
420 same location where the Lincwold-02 is adjacent to and overlaps with the Saltfleetby-99 survey
421 and together cover ~380km². Despite the UK Government encouraging and overseeing shale
422 gas exploration and a number of companies having embarked on shale gas exploration
423 programmes (see Selley, 2012) only 638km² of 3D seismic reflection data has actually been
424 acquired across ~20 000km the prospective areas since 2010 (i.e. ~3% of the prospective area).
425 Overall, the paucity of 3D seismic data onshore the UK limits the ability to interpret geological
426 structure and trends beyond a handful of areas. Despite the critical role that 3D seismic
427 reflection data have in exploration and exploitation, and their importance in future geoenergy
428 activities such as CCS, there is a limit to their resolution and, therefore, the features that can
429 be resolved to characterise the full complexity and heterogeneity of the subsurface. For future
430 geoenergy projects, operators should report the parameters and resolution of their seismic
431 reflection surveys ahead of consents being given, for example, to hydraulically fracture.

432 As is now well documented, induced seismicity felt by the local population has been
433 associated with hydraulic fracturing at two prospective shale gas sites in the UK (Clarke et al.,
434 2014;Clarke et al., 2019). At both Preese Hall (Clarke et al., 2014) and Preston New Road
435 (Clarke et al. 2019), the focus of studies has largely been the monitoring and prediction of
436 seismicity using passive seismic techniques (e.g. Clarke et al., 2019). However, the
437 observations and interpretations of the geology prior to the hydraulic fracturing and the
438 suitability of 2D and 3D seismic reflection data to make confident interpretations has received
439 limited consideration. The analysis presented on frequency content and resolution of the
440 Bowland-12 3D survey indicate that the ability to interpret structural discontinuities, such as
441 faults, which could be reactivated during hydraulic fracturing, is fundamentally limited by the
442 extent and quality of the data. In the case of the Preese Hall-1 well, the geological and
443 geophysical interpretations for the hydraulic fracture plan were based on 2D seismic data
444 (Green et al., 2012). If there is even moderate structural complexity then the migration process
445 in a vertical plane may be inadequate to capture this (Brown, 2011). The limitations for
446 geological interpretation are compounded by the sparsity and spacing of the 2D seismic
447 reflection data. The use of 3D seismic reflection data reduces the uncertainty in pre-drill
448 characterisations and predictions (Brown, 2011), including the presence and geometry of faults.
449 By acquiring 3D seismic reflection data it may have been possible to improve the structural
450 interpretation of faulting within the basin, as also suggested by Green et al. (2012). At both
451 Preston New Road wells (PNR-1 and PNR-2) the hydraulic fracture planning did utilise 3D
452 seismic reflection data. It has been described previously (Clarke et al., 2019) that the
453 reactivated fault which resulted in the induced seismicity was not imaged using the Bowland-
454 12 3D seismic reflection survey. The analysis of the post-stack seismic data here suggests that
455 ahead of any planned drilling or hydraulic fracturing it would have been possible to report that
456 the data would not be suitable for interpreting faults with either vertical (throw) or horizontal
457 (heave) displacements below the 40m and 60m estimated resolutions respectively. In addition,
458 it is possible that the resolution of the data are lower than estimated from the seismic frequency
459 because the higher frequencies in the Bowland-12 3D data do not contribute to the overall
460 interpretability of the data (**Figure 12a-c**). Given these constraints, the interpretation of a fault
461 with a vertical offset of less than 50m would be highly uncertain. The overall accuracy and
462 precision of the 3D seismic reflection data for structural interpretations is limited by the vertical
463 and horizontal resolution of the data. It should also be noted that the in-situ stress data for west
464 England (Kingdon et al., 2016;Fellgett et al., 2018) highlights that most faults are likely to be
465 optimised for strike-slip failure. Faults which are dominantly dip slip displacement are
466 frequently simpler to identify in seismic reflection data as they juxtapose intervals with

467 different seismic properties against each other. However, dominantly strike-slip faults do not
468 juxtapose differing intervals against one another, and therefore are frequently more difficult to
469 interpret in seismic reflection data. In strike slip stress regimes, there may be an increased risk
470 of induced seismicity, where faults are more difficult to identify with equivalent data in
471 dominantly dip slip settings.

472 For geothermal energy this study highlights that in areas of high heat demand there is
473 limited existing subsurface data (see **Table 3**). Both well and seismic reflection data show
474 significant clustering, with the well data also have a sampling bias with respect to depth. The
475 ability to predict subsurface properties, such as temperature, relies on calibrating models
476 against existing data. If the existing data are clustered, and there is a significant sampling bias
477 then making predictions, based on models, away from data rich areas inevitably comes with an
478 increased uncertainty. As discussed by Bond (2015), the way in which these uncertainties are
479 communicated in geosciences is important from a social and economic perspective because the
480 public are increasingly concerned with the decision-making processes and the associated risks
481 and uncertainties.

482 The subsurface will likely be required to deliver a low carbon energy transition in the
483 UK, for example the deployment of CCS, energy storage (methane and hydrogen), for the
484 continued, but sustainable extraction of natural resources (Stephenson et al., 2019) and likely
485 vital for long term disposal of radioactive waste. However, our ability to sustainably exploit
486 the subsurface relies on our ability to predict and model it accurately. Given the vintage of
487 much of the existing seismic reflection data, a consideration of future geoenergy projects
488 should be whether existing data are suitable or whether a step change in onshore seismic data
489 quality (and coverage) will be required to both fully understand the opportunity and to
490 demonstrate that activities will have a low impact on communities and the environment. The
491 variability in the extent and quality of existing data across the UK means that decision makers
492 should include an assessment on the suitability of data from the project inception phase.

493

494 *6.2. Governance and Regulatory Challenges*

495 In the UK, the governance and regulation of deep subsurface activities involves
496 different decision makers and regulatory bodies, including the Oil and Gas Authority, the
497 Environment Agency and The Health and Safety Executive. Hawkins (2015) highlighted that
498 in the case of hydraulic fracturing the existing conventional oil and gas regulation failed to
499 translate into adequate controls for the shale gas industry. The transition from the dominant use
500 of the deep subsurface in the UK being for fossil fuel production in the offshore areas, to a
501 more complex and multi-faceted system onshore, potentially raises questions on the suitability
502 of existing governance and regulation structures in managing activities. An example could be
503 the move to localised energy systems for the use of geothermal energy (Lloyd, 2018). As
504 highlighted by this study, both the coverage and quality of existing subsurface data vary
505 considerably across UK regions and communities. Consideration to governance, regulation and
506 guidelines should be addressed ahead of expansion of these nascent subsurface activities and
507 should consider best practice guidance on the minimum data requirements ahead of activities
508 which perturb the subsurface to design and implement more effective mitigation measures for
509 the potential impacts on communities and the environment. Given that subsurface data have
510 inherent resolution limitations, and that hydraulic fracturing by its very nature perturbs the
511 subsurface, it could be argued that there should be a minimum requirement for data resolution

512 ahead of such activities. At present there are no minimum standards or expectations for the data
513 which decisions must be based on.

514 The exploration and production of unconventional hydrocarbons which use hydraulic
515 fracturing methods have brought into sharp focus the challenges in confidently predicting the
516 subsurface. There is typically a larger uncertainty in subsurface interpretations using 2D
517 seismic reflection data compared with 3D seismic reflection data, with reduced uncertainty a
518 function of both improved areal coverage and the benefits of 3D migration (Bacon et al., 2007).
519 The Consolidated Onshore Guidance (Oil and Gas Authority, 2018) specifies that “a map and
520 seismic lines showing faults near the well and along the well path” should be included but
521 makes no specific reference to demonstrating the suitability of the underlying data on which
522 those interpretations are made. There is no requirement for the operators to demonstrate that
523 the seismic reflection data are specifically suitable for the activity that is being planned. The
524 required information relates to primarily to interpretations (or knowledge).

525 How industry and society utilise the deep subsurface is likely to change as a result of
526 the need to decarbonise energy systems. This change undoubtedly will bring about new
527 regulations and guidance. The status quo of adopting previous practice from either onshore or
528 offshore oil and gas exploration and production, for nascent geenergy activities is unlikely to
529 be a justifiable position and new frameworks should consider the inherent uncertainty and
530 possible impacts of deep subsurface activities.

531

532 6.3. *Communities and Science Communication*

533 Risks associated with subsurface development are a major public issue for UK citizens,
534 especially since 2011 when hydraulic fracturing led to seismic activity at Preese Hall (Clarke
535 et al., 2014). Moreover, strong public opposition to hydraulic fracturing and subsurface
536 development appear to be linked to the uncertainty associated with seismic activity, even
537 though few UK residents have actual first-hand experiences with high hazard seismic events
538 (Cotton, 2015; Szolucha, 2018). Nevertheless, not all UK regions and communities are equally
539 exposed to subsurface development. That is, there are significant regional and community
540 variations in the subsurface resources, and as shown here the quantity and quality of this data
541 relates directly to the uncertainty in characterising these resources, and the assessment of the
542 risks. This unequal distribution of subsurface risk is also compounded by various
543 interpretations of risk. Social science research suggests that variations in perceptions of risk
544 are explained by geography, culture, socioeconomic status, ethnicity, race and gender (Flynn
545 et al., 1994). As just one example of the importance of context, consider the case of hydraulic
546 fracturing in Oklahoma (USA), a state highly dependent on oil and gas development. The
547 perceived risks associated human induced seismicity among Oklahoma residents are less of a
548 concern than perceived risks associated with pollution, especially to water and poisoning of
549 livestock (Campbell et al., 2020). Thus, when subsurface data are mapped out across the UK it
550 demonstrates the potential for enormous variation in interpretation of risk according to the
551 spatial location of wells as well as the constellation of community and demographic
552 combinations that may together shape risk perceptions (e.g. Kropp, 2018). This distribution of
553 perception of risk has yet to be explored in the UK using subsurface data, though ecosystem
554 services suggests there are good reasons to undertake such an analysis in the future.

555 There is an increasing public demand for high quality information that is accurate,
556 consistent, complete, timely and representative (e.g. Wang and Strong, 1996). This analysis

557 suggests that seismic reflection and borehole data represent an information source that can be
558 used to contribute to information quality and aid in the communication of subsurface risk.
559 However, simply reporting information, even high-quality information, is probably not enough.
560 Social science research suggests that credible information sources are highly important in
561 conveying actual risk (Renn and Levine, 1991). Thus, where data are uncertain or complex the
562 public is likely to rely on experts to help them make sense of subsurface risks that may be
563 reflected in those data. As a result, trust in the experts and institutions is likely to have an
564 important impact on general perceptions about risks associated with subsurface development.

565 The interpretation of these subsurface data open up an important opportunity for
566 geoscientists to help engage UK citizens about the levels of uncertainty and subsurface risks
567 associated with energy development (e.g. Buchanan et al., 2014). However, with opportunities
568 also come challenges. That is, while this study is one of the first to map the onshore UK
569 subsurface, much of the underlying data are produced by industry. Thus, information presented
570 by geoscientists will be constantly evaluated within the context of industry trust (Wray et al.,
571 2008; Wachinger et al., 2013; Seeger et al., 2018). The challenge, then, is to convey meaningful
572 information about uncertainty and risk when data generated may be viewed as suspicious,
573 especially when it is not publicly accessible. Therefore, one of the biggest obstacles in
574 conveying accurate perceptions of risk to UK residents may rest in the fact that frequently
575 subsurface data are generated by industry (Wachinger et al., 2013), although in geosciences,
576 these data may subsequently be available for regional synthesis. Such challenges, however, are
577 not usual in risk analysis as researchers find that stakeholders are often perceived to
578 communicate risk through the selective use of data that advances their own interests (Leiss,
579 1995). Future social science research might test public perceptions about trust in different types
580 of subsurface data. That is, are some types of subsurface data likely to be trusted more by the
581 public? If so, why? Which types of data could be best used to communicate the nature of
582 subsurface risks? What organisations are best placed to communicate data about subsurface
583 risks? Why? These are just a few of the issues that geoscientists may confront when attempting
584 to map the landscape of subsurface risk.

585

586 **7. Conclusions**

587 After over a century of subsurface data collection onshore UK, there remains significant
588 subsurface uncertainties, which in part are due to the quality and accessibility of existing key
589 subsurface datasets. This study highlights a paucity of both well and seismic data across the
590 onshore UK. All subsurface interpretations, be it for well-established activities such as
591 conventional oil and gas exploration and production, or new activities as part of the energy
592 transition, rely on these geophysical or geological data. These interpretations and models are
593 fundamentally limited by the inhomogeneous datasets and the resolution of them. Onshore oil
594 and gas production in the UK currently accounts for <1% of the total production from the UK
595 (OGA, 2020) and the limited scale of resources, when compared to the offshore, that has
596 restricted further data collection, with companies prioritising the offshore areas of the UK
597 Continental Shelf. The lack of extensive and high-quality data could be a fundamental
598 limitation on the expansion of nascent low carbon subsurface activities and technologies. The
599 attention with which the public are now putting on all new energy activities will require
600 geoscientists to clearly articulate the limitations of currently available datasets, and these
601 limitations should highlight areas where new data collection is needed, both to improve
602 coverage, and to improve resolution. The ability to understand and quantify uncertainties in a
603 subsurface description is key to effectively reducing safety, environmental, health and

604 economic risks. Gaining new knowledge through data acquisition cannot be guaranteed to de-
605 risk a subsurface outcome, however, the new knowledge can be vital in the decision-making
606 processes.

607 The analysis and statistical measures shown here for the onshore UK subsurface
608 datasets can be used to determine priority areas for future data collection. But the analysis does
609 not address what is enough data for a given activity. There needs to be a concerted effort across
610 geosciences and social sciences to understand what defines an acceptable level of uncertainty,
611 financial risk, and environmental risk. This study raises the question is there a need for
612 regulators to demand minimum data standards as part of the planning process prior to
613 subsurface activities taking place? There is more than ever a social dimension to subsurface
614 uncertainty. Explaining the information contained within the data are as important as the data
615 itself. Never has the spotlight been so focused on the ability of geoscientists to predict the
616 subsurface.

617

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634

635 **Data Availability**

636 **3D Seismic Data.** The 3D seismic reflection presented in this study are available from the
637 UKOGL but restrictions apply to the availability of these data, which were used under licence
638 for the current study, and so are not publicly. Data are however available from the authors upon
639 reasonable request and with permission of the UKOGL. See www.ukogl.com.

640 **OGA Onshore 2D and 3D Seismic Data.** The location of 2D and 3D seismic data onshore the
641 UK analysed during this study is available from the Oil and Gas Authority (OGA) at
642 https://maps.ukogl.org.uk/arcgis/rest/services/public/public_seismic_BritNatGrid/Mapserver

643 **BGS Borehole Locations.** The BGS borehole location dataset analysed during this study is
644 available from the www.bgs.ac.uk.

645 **UKOGL Borehole Locations.** The location of the UKOGL borehole locations used in this
646 study are available from UKOGL but restrictions apply to the availability of these data, which
647 were used under licence for the current study, and so are not publicly available. Data are
648 however available from the authors upon reasonable request and with permission of UKOGL.
649 See www.ukogl.com.

650 **BGS Geothermal Catalogue.** The BGS Geothermal Catalogue data analysed during this study
651 is available from <http://nora.nerc.ac.uk/id/eprint/512272/>

652 **BGS Geophysical Logs and Borehole Samples.** The location of geophysical logs and
653 borehole samples used in this study are available from the BGS but restrictions apply to the
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655 publicly available. Data are however available from the authors upon reasonable request and
656 with permission of BGS. See www.bgs.ac.uk.

657 **UKERC Heat Demand.** The domestic and industrial heat demand data from Taylor et al.,
658 (2014) is available from
659 <https://data.ukedc.rl.ac.uk/browse/edc/efficiency/residential/Buildings/DS4DS>

660 **OGA Offshore 2D Seismic Data.** The data used to compare the offshore coverage to the
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666

667 **Author Contributions**

668 **MI:** conceptualization (lead), data curation (lead), formal analysis (lead), investigation (lead),
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671 analysis (supporting), writing – review and editing (supporting). **PS:** validation (supporting),
672 writing – review and editing (supporting); **RD:** conceptualization (supporting), writing –
673 original draft (supporting), writing – review and editing (supporting); **AK:** data curation and
674 evaluation (supporting), writing – review and editing (supporting)

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824

825 **Table 1.** Typical data types and their classification based on the Ackoff (1989) hierarchical
 826 model

Data	Information	Knowledge
Well depth		
Well locations		
Samples (from wells)		
Geophysical logs (wells)		
Temperature (from wells)		
Fluid sample (from wells)		
	Geothermal gradient	
Raw seismic field data		
Processed seismic reflection data		
	Seismic horizons	
		Fault geometry

827

828

829 **Table 2.** Data sources used in the quantitative analysis of available data onshore the UK

Data type	Collection	Source	Format	N=
2D seismic	Onshore 2D seismic locations	OGA	Lines	9283
3D seismic	Onshore 3D seismic locations	OGA	Polygons	32
Wells	UKOGL: deep wells locations	UKOGL	Points	4156
Wells	Onshore well locations	OGA	Points	2242
Wells	All boreholes locations	BGS	Points	1 335 511
Wells	Temperature measurements	BGS	Points	1712
Borehole Samples	Samples from boreholes	BGS	Database	10 427
Geophysical logs	Geophysical logs by well	BGS	Database	4541 (digital) 6454 (paper)

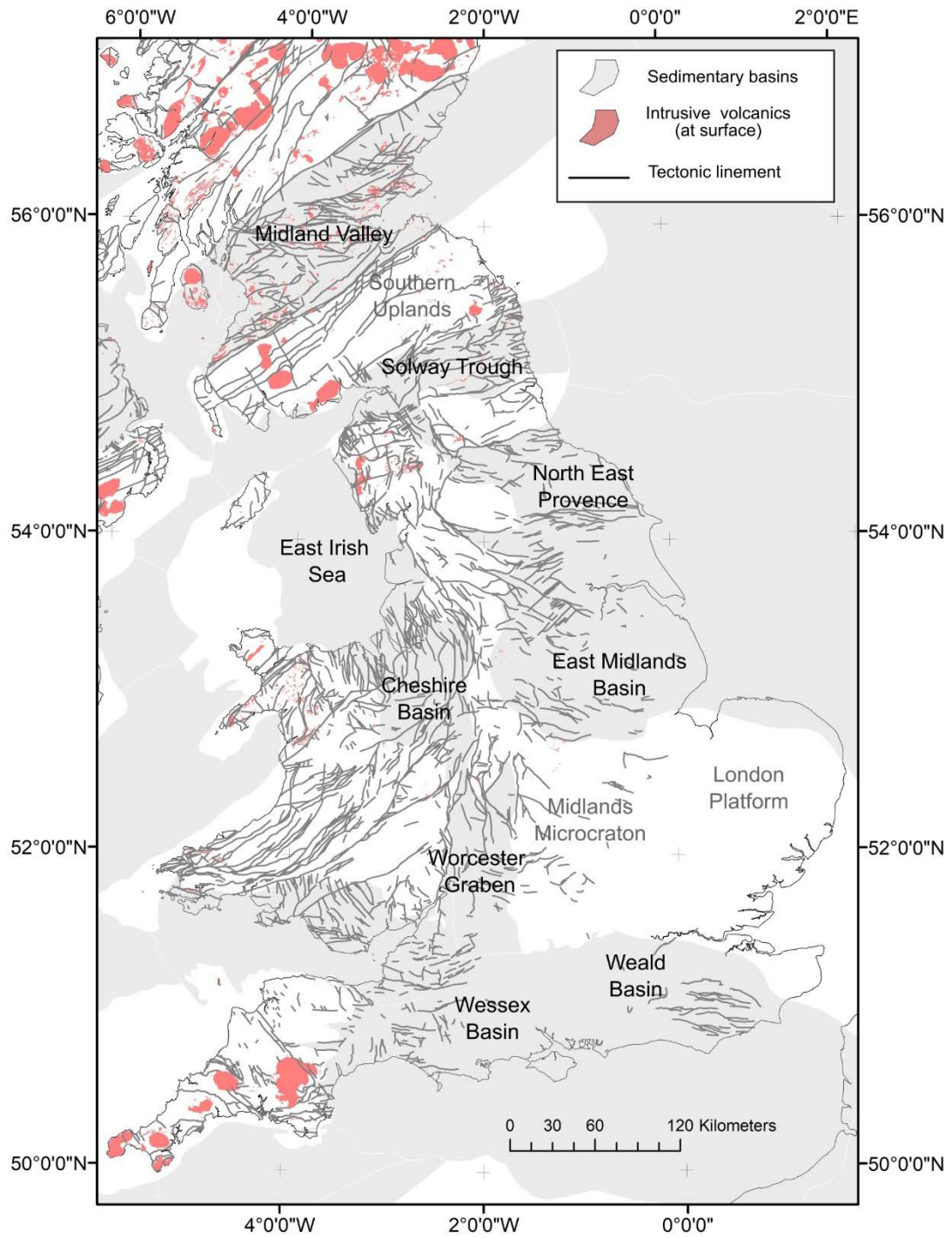
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831

832 **Table 3.** Data coverage for 10 largest areas of annual domestic heat demand above 10 000
 833 MWh/km². Both extent of 2D seismic reflection data and number of deep wells are quantified
 834 within these areas. See Figure 7 for map view of London, Birmingham, Manchester and
 835 Liverpool.

City/Town	Area of City/Town (km²)	2D seismic reflection data (km)	Deep wells (>300m TVD)
London	1295	0	14
Birmingham	492	0	7
Manchester	370	139	5
Liverpool	182	27	5
Glasgow	133	0	0
Newcastle upon Tyne	121	0	1
Leeds	81	0	0
Nottingham	81	20	3
Bristol	80	0	1

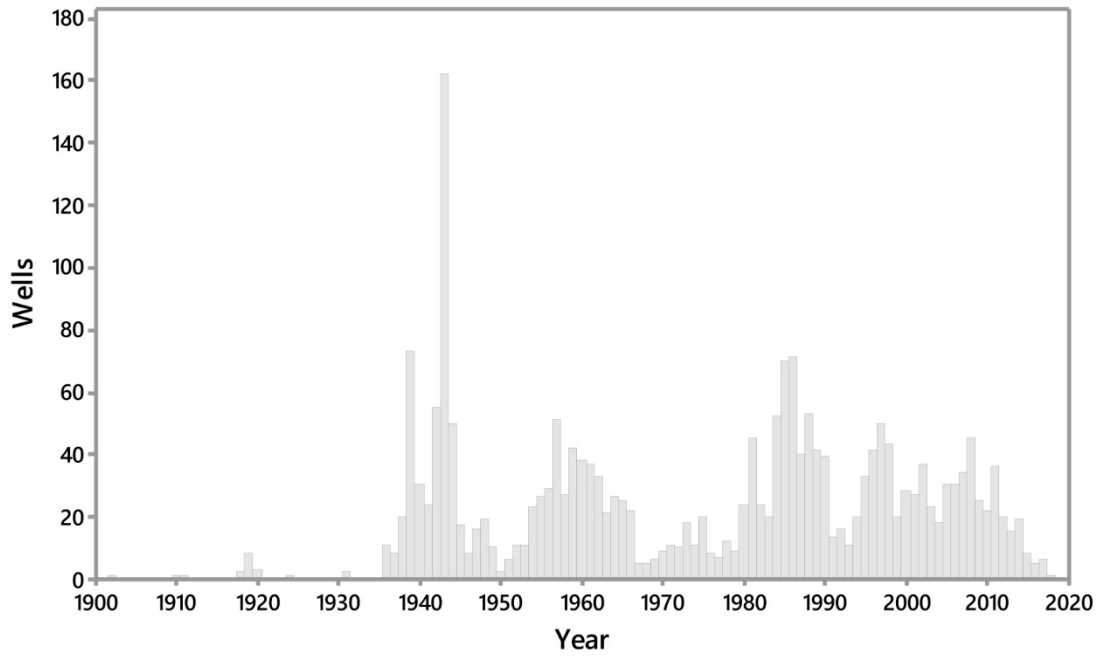
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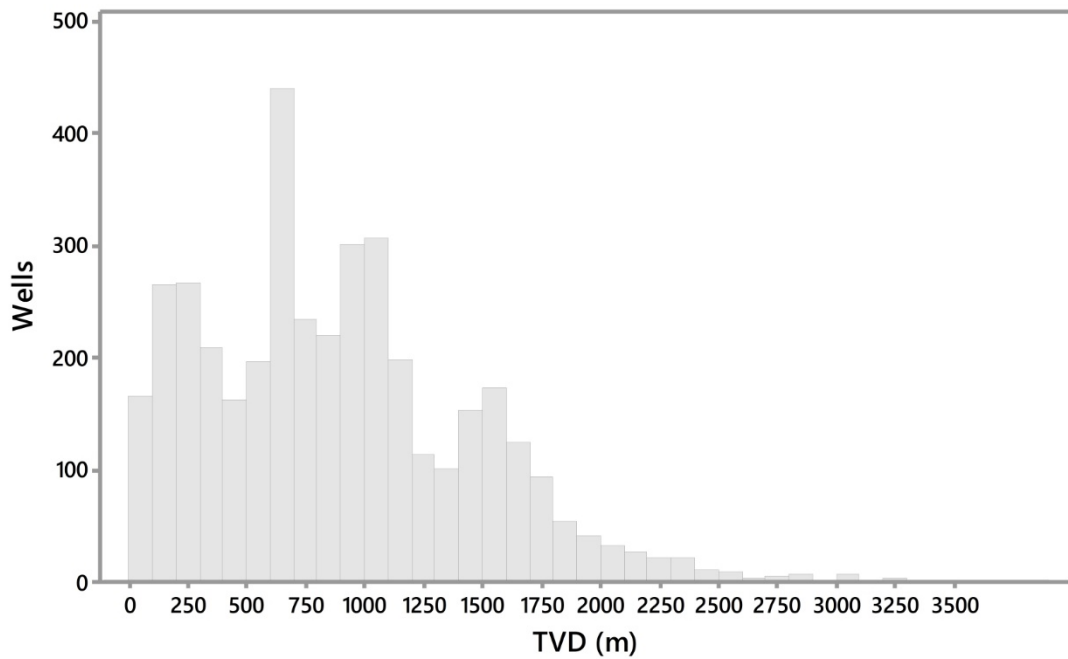
837

838 **Fig. 1.** Simplified geological map of the UK, showing the outlines of sedimentary basins,
 839 position of volcanics at surface, and major tectonic liniments.

(a)

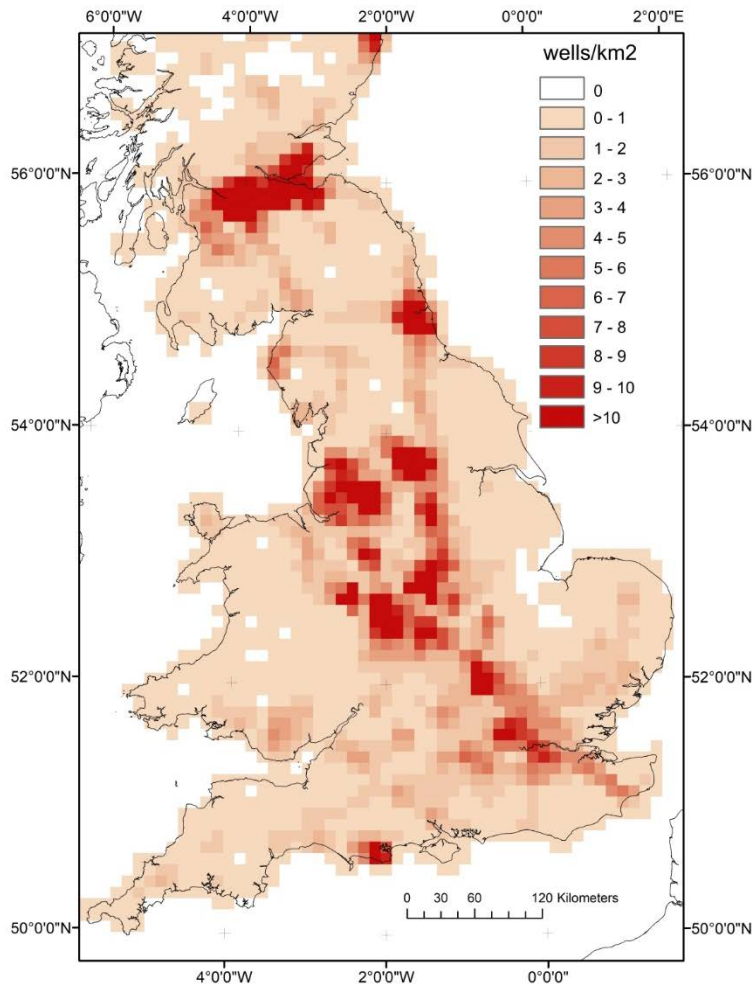


(b)



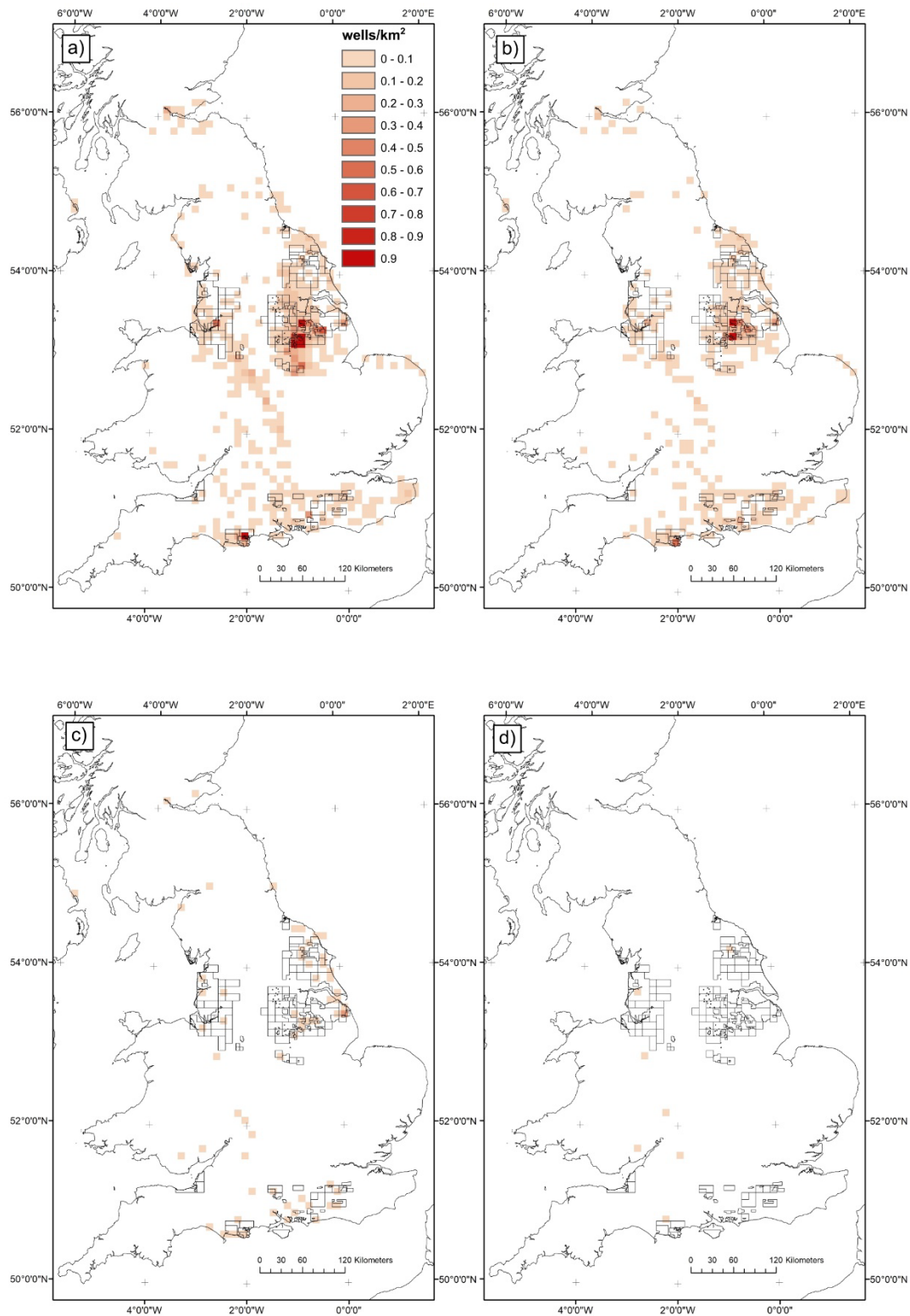
840

841 **Fig. 2.** Histograms of wells drilled onshore the UK from the UKOGL database a) by year and
842 b) TD TVD



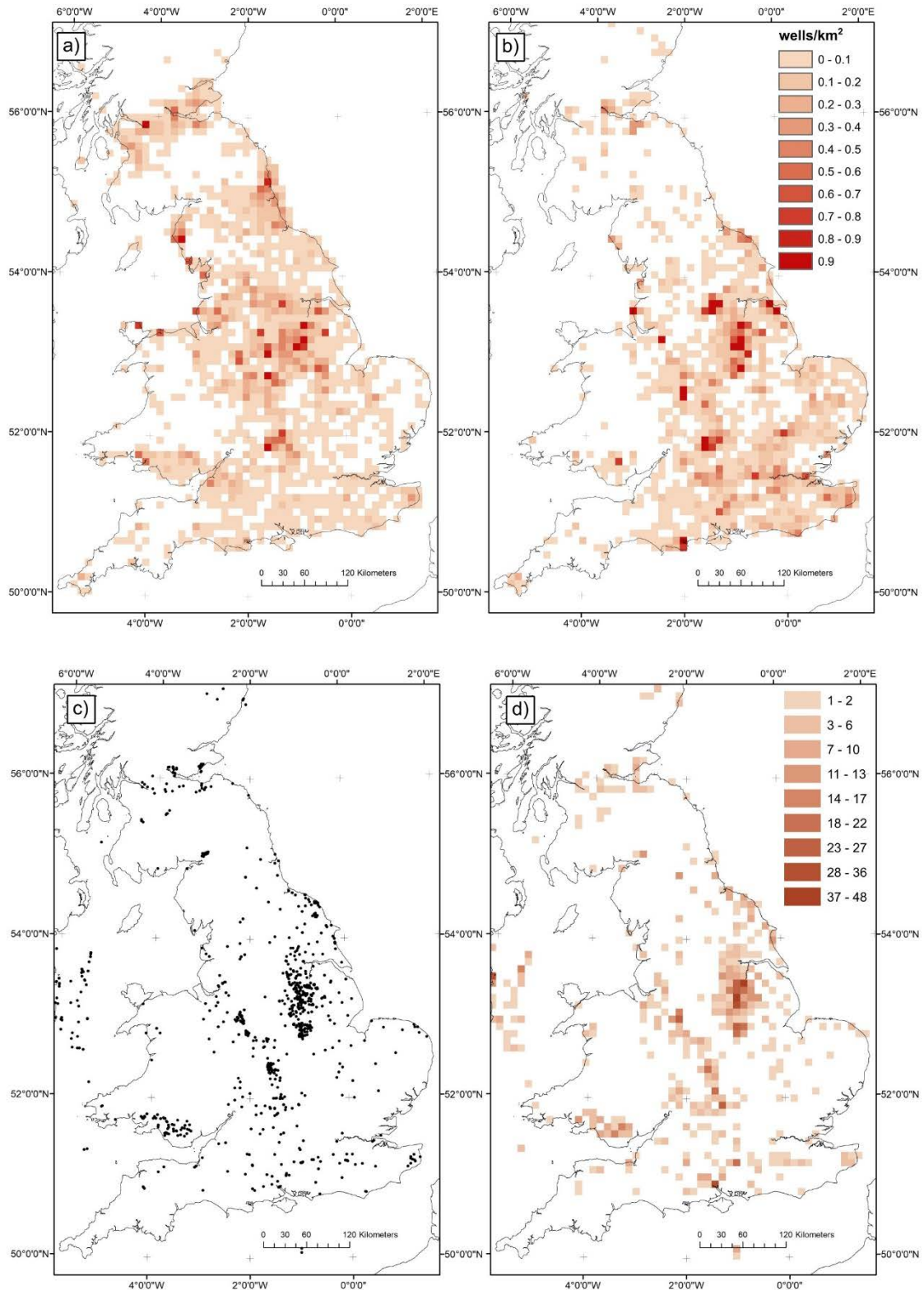
843

844 **Fig. 3.** Well density of all boreholes held by the BGS with a TD between 30 and 300 m.



845

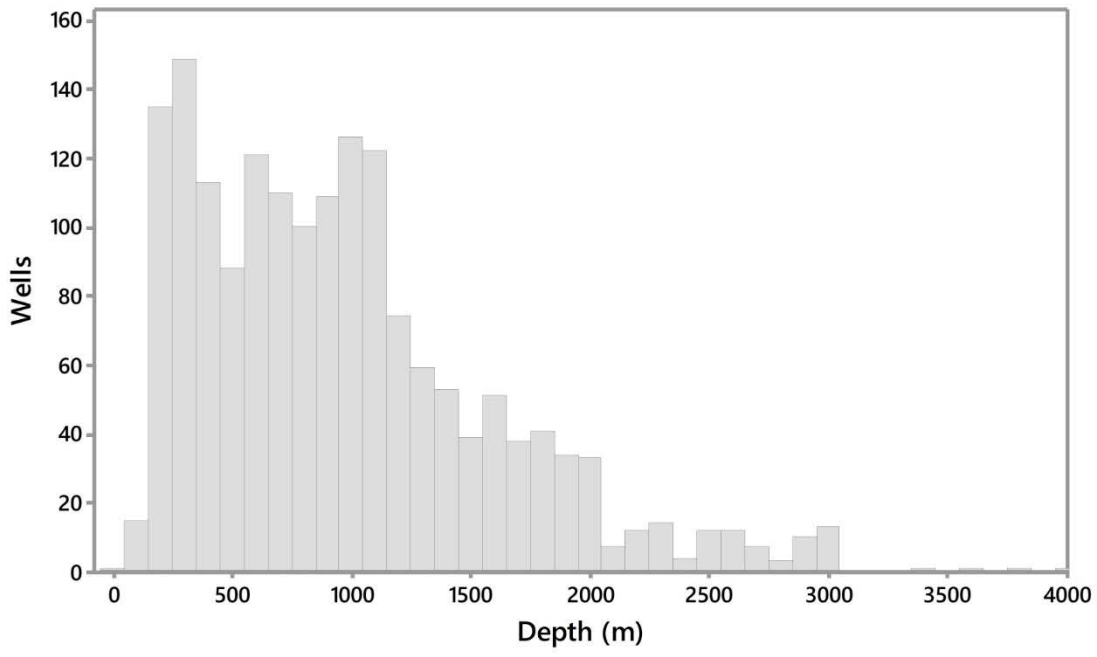
846 **Fig. 4.** Map showing the density of wells from UKOGL database with total depths (TVD) of:
 847 a) >500m; b) >1km; c) >2km; d) >3km. Includes location of current PEDL



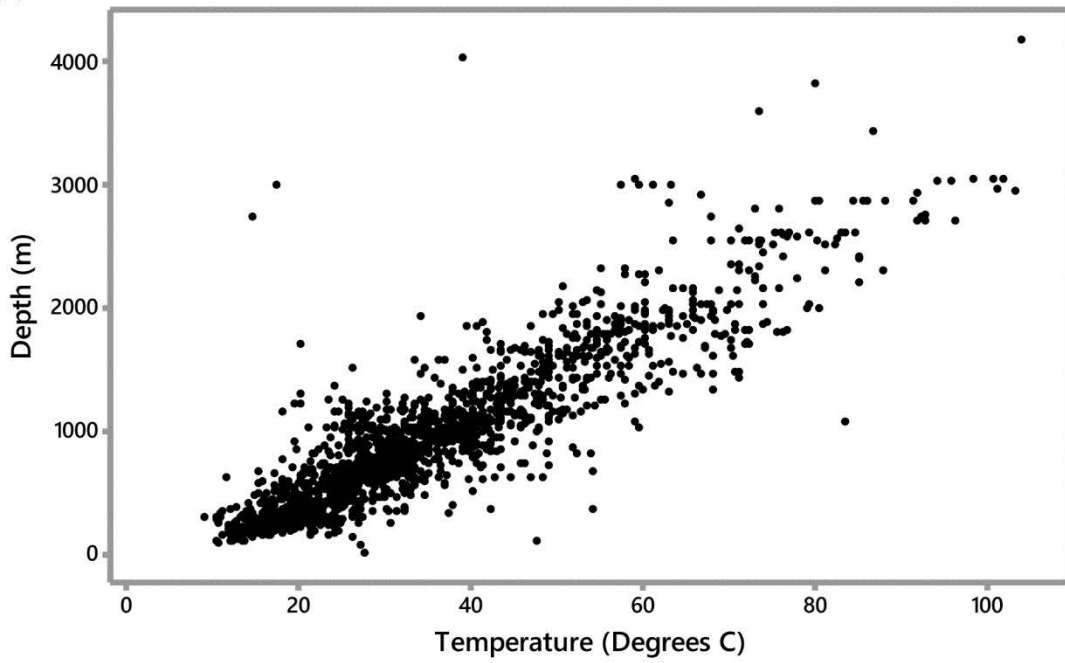
848

849 **Fig. 5.** Spatial density of wells with a) geophysical logs and b) rock samples and maps showing
 850 c) the location of wells with temperature measurements and d) the number of temperature
 851 measurements per 10km².

(a)

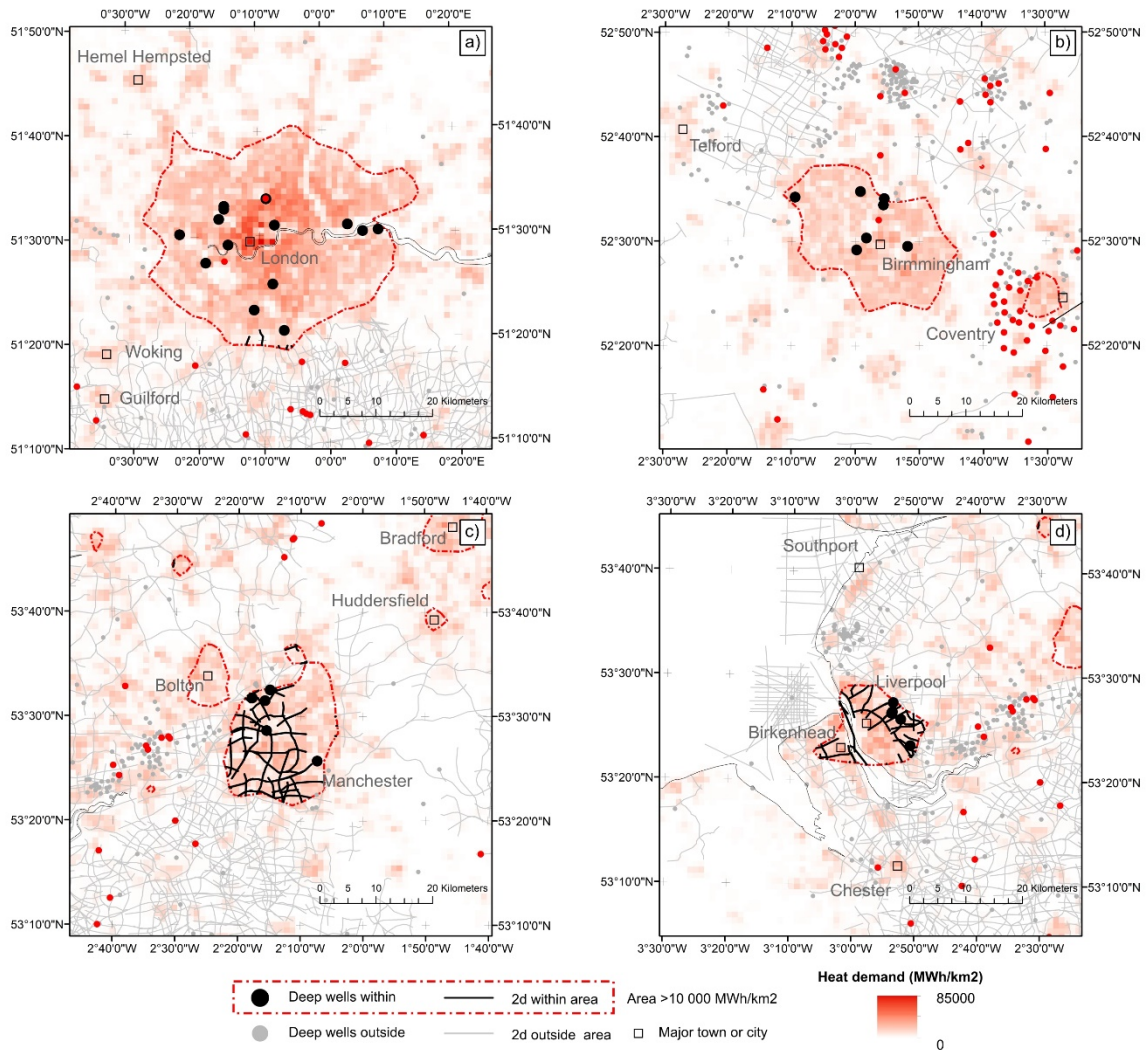


(b)



852

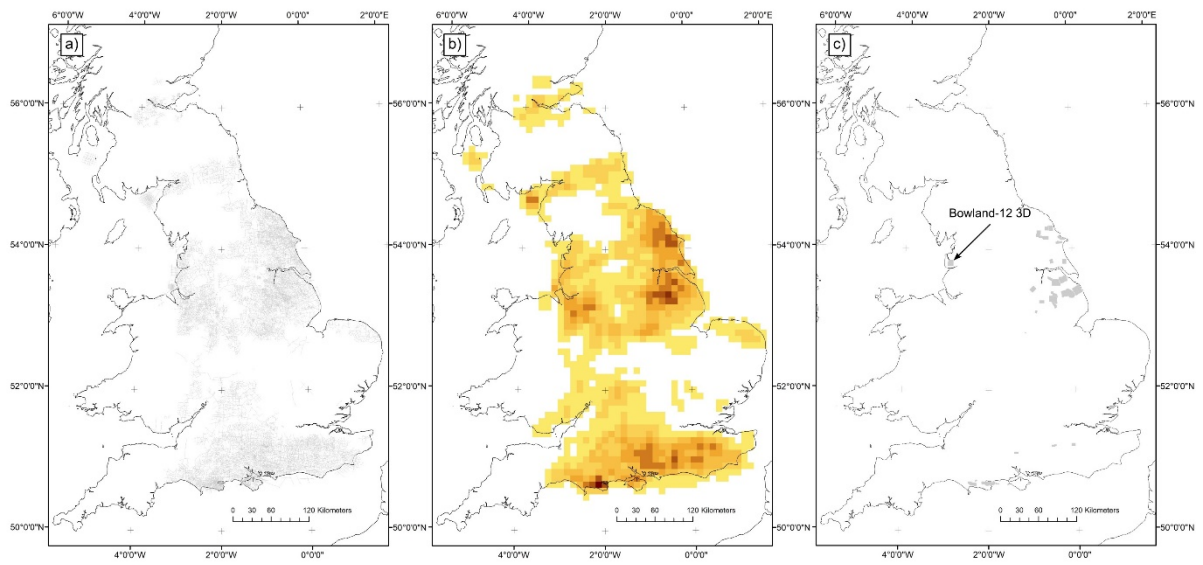
853 **Fig. 6.** Histogram of a) temperature measurements by depth and b) temperature vs depth plot.



854

855 **Fig. 7.** Subsurface data coverage, showing the location of both 2D seismic data and the deep
 856 wells from UKOGL in relation to heat demand across a) London, b) Birmingham, c)
 857 Manchester and d) Liverpool.

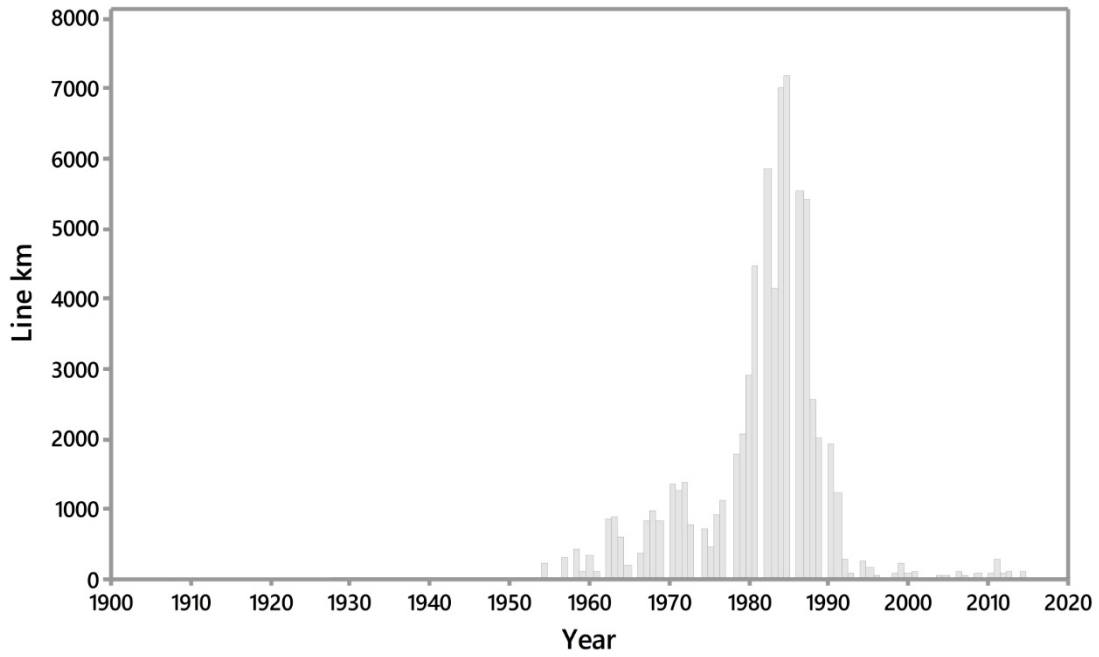
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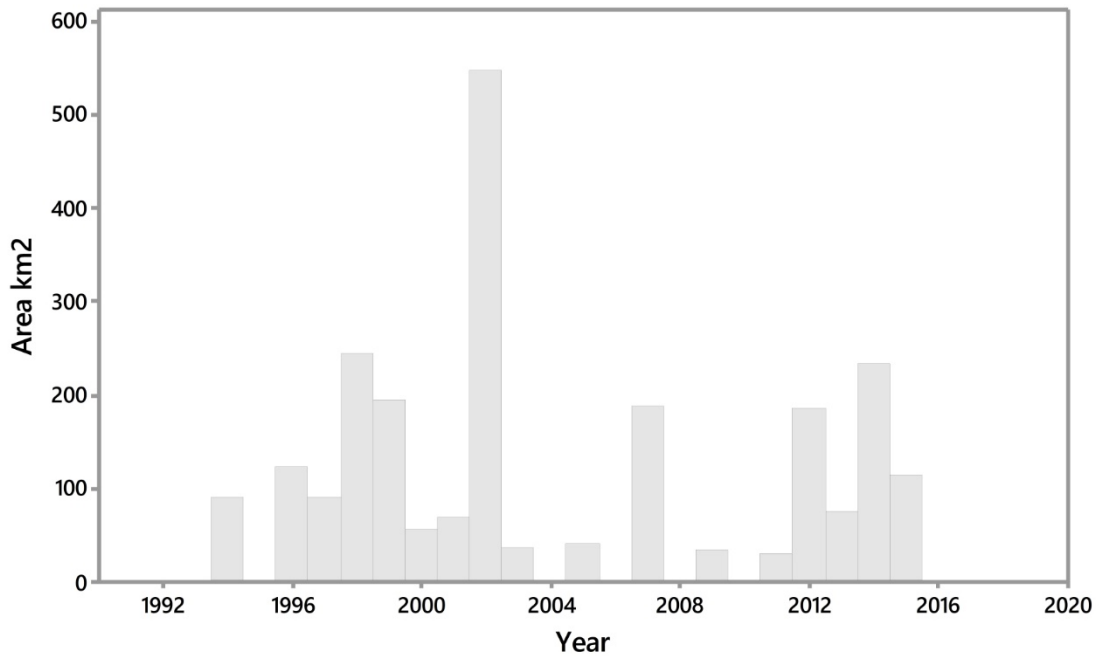
859

860 **Fig. 8.** a) 2D seismic data across the UK and b) the number of line km of 2D seismic per 10km^2
 861 and c) location of all 3D seismic reflection surveys onshore the UK (grey polygons) with the
 862 location of Bowland-12 3D survey highlighted.

(a)

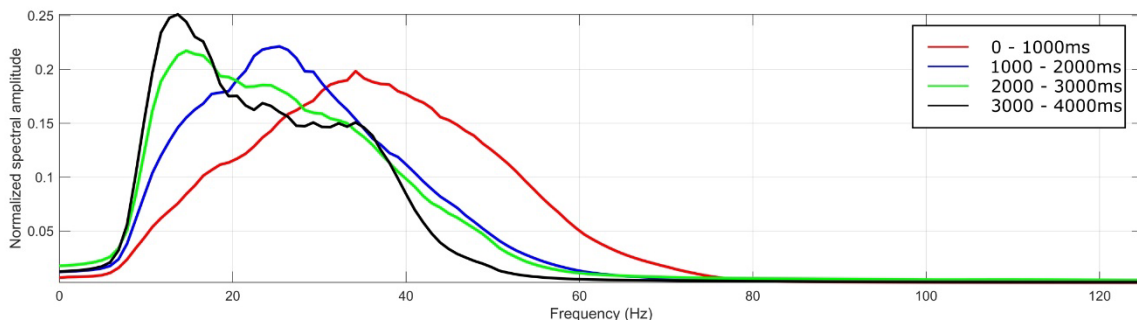


(b)



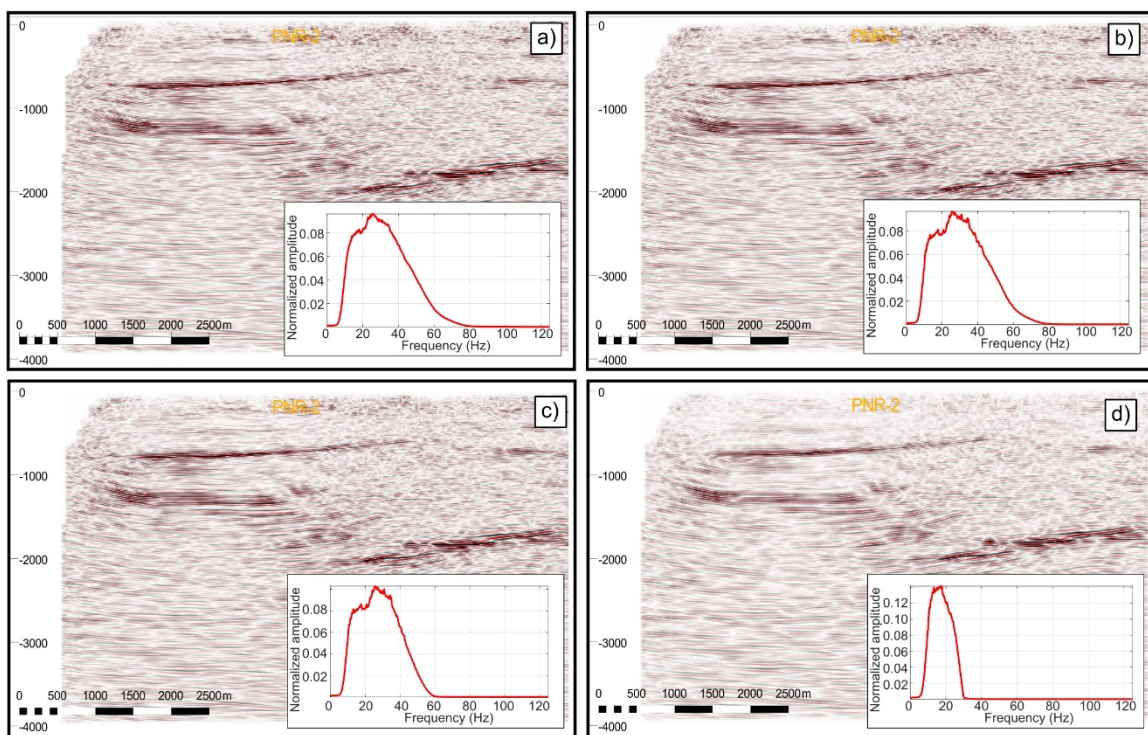
863

864 **Fig. 9.** Histograms of a) length of 2D seismic lines acquired onshore the UK by year, and b)
865 area of 3D seismic surveys acquired onshore the UK by year



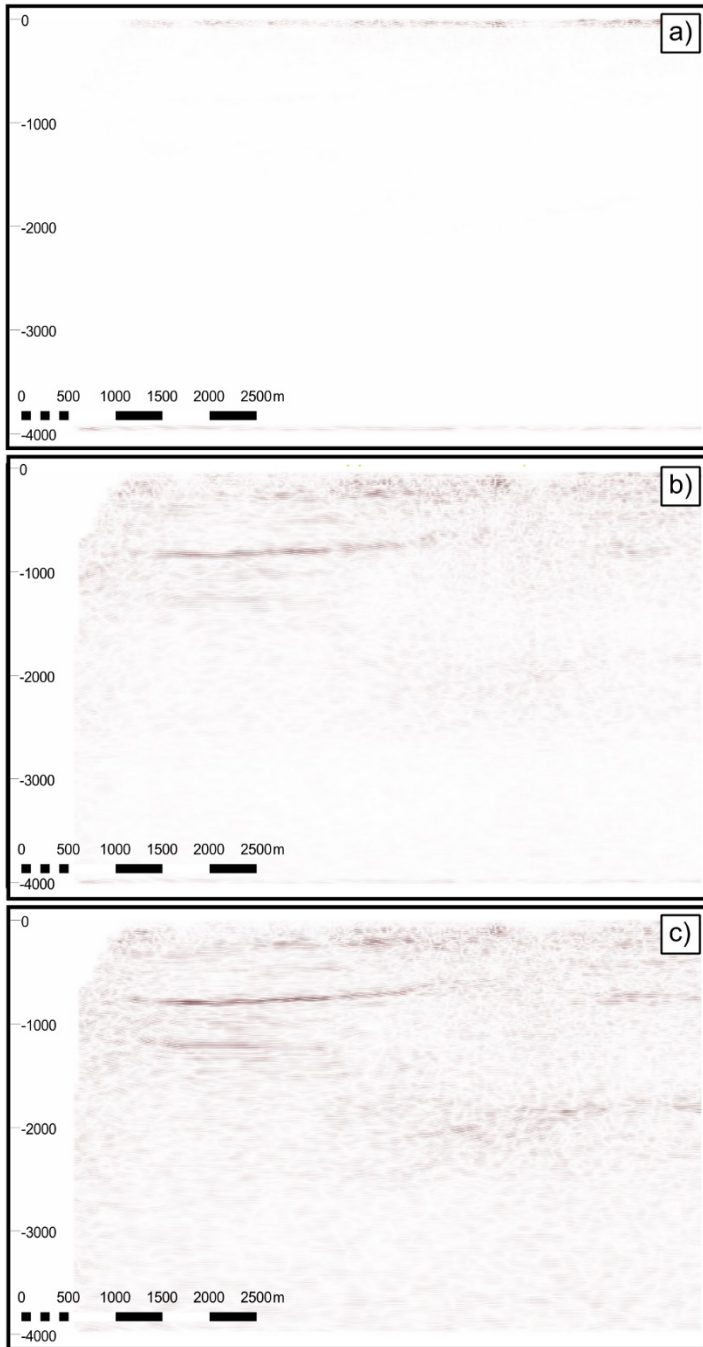
866

867 **Fig. 10.** a) A simple comparison of the frequency content of the Bowland-12 3D seismic
 868 reflection survey for different time intervals. The frequency content decreases with depth.



869

870 **Fig. 11.** Comparison of seismic sections adjacent to the Preston New Road 2 well. a) unfiltered;
 871 b) low pass filter cut at 90Hz; c) low pass filter cut at 60 Hz and d) a low pass filter cut at 40
 872 Hz. The section is orientated E-W (XL 1234).



873

874 **Fig. 12.** Difference between the original seismic data and a) low pass filter cut at 90Hz; b) low
 875 pass filter cut at 60 Hz and c) low pass filter cut at 40 Hz. The section is orientated E-W (XL
 876 1234).

877