



1 **Uncertainty in regional estimates of capacity for carbon capture and storage**

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10 **Abstract.** Carbon capture and storage (CCS) is a potentially important technology for
11 the mitigation of industrial CO₂ emissions, however the majority of the subsurface
12 storage capacity is in geological strata for which there is relatively little information,
13 the so-called saline aquifers. Published estimates of the potential storage capacity of
14 such formations, based on limited data, often give no indication of the uncertainty,
15 despite there being substantial uncertainty associated with the data used to
16 calculate such estimates. Using only publicly available data, a group of experts
17 independently estimated the storage capacity of 7 regional saline aquifers. The
18 experts produced a wide range of estimates for each aquifer due a combination of
19 using different published values for some variables and differences in their
20 judgements of the aquifer properties such as area and thickness. The range of
21 storage estimates produced by the experts shows that there is significant
22 uncertainty in such estimates, in particular the experts' range does not capture the
23 highest possible capacity estimates, meaning that by not accounting for uncertainty,
24 such regional estimates may underestimate the true storage capacity. The result is
25 applicable to single values of storage capacity of regional potential, but not to
26 detailed studies of a single storage site.

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28

29 **1. Introduction**

30 Geological storage of carbon dioxide (CO₂) has been proposed as a potential
31 technological solution to help reduce emissions of greenhouse gases, given the
32 continued use of fossil fuels to meet much of the world's energy requirements. In
33 carbon capture and storage (CCS), the CO₂ produced from industrial sources is
34 captured and transported to a geological storage site and injected deep into the
35 subsurface where it is stored indefinitely in the pore space of the rocks. So-called



36 saline aquifers, rock formations where the pore space is filled with brines too saline
37 for useful extraction, offer the largest storage capacity (Holloway, 1997). However,
38 unlike hydrocarbon reservoirs, such formations often have limited legacy data. In
39 order to identify potential storage sites that are worth the investment required for
40 detailed assessment, attempts have been made to characterise regional saline
41 aquifers using this legacy data (e.g. Wilkinson et al., 2013). However care must to
42 taken to account for the substantial uncertainty associated with such regional
43 assessments. The capacity of a geological formation to store CO₂ securely is a first-
44 order concern in any storage assessment. Lack of capacity is one of the highest risks
45 to carbon capture and storage projects (Polson et al., 2012) and uncertainty impacts
46 the design of transport and injection networks (Sanchez Fernandez et al., 2016).
47 Previous work on the subject is limited, though Calvo et al. (2019) studied the
48 influence of the uncertainty in storage capacity due to uncertainty in thermophysical
49 properties (pressure and temperature of the reservoir).
50
51 Many published regional studies of CO₂ storage capacity quote single values for the
52 capacity of individual formations, sometimes with ranges allowing for uncertainty in
53 a single parameter such as the proportion of porespace that can be utilised for
54 storage ('storage efficiency') e.g. Medina et al. (2011). The reporting of individual
55 studies varies, but some provide storage estimates to 6 significant figures, implying a
56 precision of greater than 0.001 %. However, this precision is clearly unachievable,
57 since the commonly used methodologies for capacity calculation of so-called saline
58 aquifers (e.g. Goodman et al. 2011) requires inputs which are inherently variable
59 over the area of assessment, such as the thickness of the formation, net:gross ratio
60 (the proportion of usable reservoir within the overall unit thickness), and porosity.
61 When offshore locations are considered, data are usually available from only a small
62 number of borehole penetrations, often with a spacing between boreholes of several
63 kilometres. While there are published methods for dealing with such uncertainty
64 (Burruss et al., 2009; Smith et al., 2011), estimates of the variability of each input
65 parameter must be made, and suitable software employed for the calculation.
66 Consequently the use of single-value storage estimates is both quicker and cheaper
67 than full probabilistic assessments.



68

69 Furthermore, capacity assessments will largely depend on expert interpretation of
70 geological data, and are therefore dependent on the prior knowledge and
71 experience of individual experts (see Curtis, 2012, for summary). Studies have shown
72 that geological experts are subject to a range of cognitive biases, as are all
73 individuals (Kahneman et al., 1982), that combined with differences in prior
74 experience can influence their interpretation of data leading to subjective results
75 (e.g. Phillips, 1999; Polson and Curtis, 2010; Bond et al., 2012). As a result, an
76 estimate of the accuracy of single-value storage capacities is of practical use, not
77 least with assessments already published but lacking an assessment of uncertainty.
78 This is of particular practical importance where a storage estimate falls close to a
79 cut-off value, below which, for example, a potential storage unit may be rejected as
80 being too small to be economically viable. For example, a regional screening study
81 (Wilkinson et al., 2010) rejected all units below an arbitrary 50 Mt of estimated CO₂
82 storage capacity. For an individual storage project the minimum acceptable storage
83 capacity value is likely to be determined by the volume of CO₂ to be stored over the
84 project lifetime.

85

86 For this study, an assessment of the accuracy of storage capacity estimates was
87 conducted as part of a study of an area of the UK territorial waters, in the Inner
88 Moray Firth area of the North Sea (Fig. 1). Subsurface geological data were available
89 from boreholes drilled by the petroleum industry, both as individual well records
90 released by the UK Government, and summarised as scientific publications. The
91 subsea strata are largely siliciclastics, of Devonian to Jurassic age. They rest
92 unconformably on strata that were affected by the lower Palaeozoic Caledonian
93 Orogeny (Andrews et al., 1990), which are here considered to be basement (i.e. to
94 have no storage potential). To the east of the area, there is a variable-thickness
95 cover of Cretaceous Chalk, a fine-grained pelagic limestone, here not considered as a
96 potential store as it lacks an obvious seal. Questions concerning the presence of a
97 suitable seal, trapping structures and potential leakage pathways were addressed in
98 the wider study but are not reported here.

99

100 **2. Materials and Methods**

101

102 A group of graduate students who had been trained in the methodology of storage
103 capacity estimation and in at least basic geology relevant to CO₂ storage, assessed
104 the capacity of the potential saline aquifers in the area. All the students were
105 studying for a Masters of Science degree in Carbon Capture and Storage, and can be
106 considered to be 'expert' in the subject, though their prior backgrounds are variable
107 ranging from geosciences to engineering. The experts had to identify the potential
108 reservoir ¹formations (saline aquifers) within the area using the scientific literature,
109 then collect the input information required to perform the basic storage capacity
110 estimates (surface area, thickness, porosity, net:gross ratio), using The product of
111 these parameters is an estimate of the volume of porewater within the aquifer,
112 which may be compressed or partly displaced allowing for the storage of CO₂.

113

$$114 \quad M = AhNG\Phi\rho E \quad (1)$$

115

116 where M is the mass of CO₂ that can be stored, A is the area that defines the region
117 being assessed, H is the thickness of the saline aquifer, NG is the net:gross ratio, Φ is
118 the porosity, ρ is the density of CO₂ and E is a storage efficiency factor.

119

120 For surface area the experts were directed to maps within (14) and (15); each expert
121 independently estimated the area. Uncertainty in this parameter is therefore due to
122 the variable interpretation of the same data from expert to expert. For the other
123 parameters, the experts were expected to locate suitable data, primarily using web-
124 based search tools. The uncertainty in these parameters is therefore determined by
125 the total number and range of published values; the ease with which experts could
126 find relevant information; and the interpretation by the experts of the applicability
127 and reliability of the data that they located.

128



129 For the purposes of this paper, the values for each variable provided by the experts
130 were combined with constant values of CO₂ density (650 kg/m³) and storage
131 efficiency (the proportion of porespace that can be utilised for storage, here taken to
132 be 0.02), and the total storage capacities were re-calculated for each expert using
133 Equation 1. This approach was undertaken to remove non-geological effects from
134 the results, such as variation in estimated CO₂ density due to the use of different
135 equations of state or pressure / temperature conditions of burial, and also any
136 calculation errors. These individual estimates are hereafter referred to as experts'
137 estimates however they are not the estimates calculated by the individual experts,
138 but the estimates re-calculated by the authors using the data collected by each
139 expert. For each geological unit, the standard deviation of the storage estimates was
140 calculated across the set of individual storage volume estimates. All experts gave
141 express permission for their data to be used for this purpose.

142
143 In order to determine the full range of possible estimates from the expert derived
144 values, storage estimates were calculated for all possible combinations of the
145 variables. The resulting distribution of the storage estimates, $P(M)$, gives an
146 indication of their uncertainty. However as this method does not take into account
147 the real uncertainty in each variable (which is unknown), $P(M)$ is not the probability
148 distribution of the storage capacity.

149

150 **3. Results**

151

152 There are 7 geological units (which are either Formations or Members in formal
153 nomenclature; Cameron et al., 1990; Richards et al, 1993) that are potential storage
154 reservoirs in the area, henceforth called storage units. Figure 2 shows $P(M)$ as a
155 cumulative density function for each formation and Table 1 shows the median and
156 range of the individual expert estimates and the 5th, 95th and median of $P(M)$. Both
157 show a wide range of possible estimates for the storage capacity. The range of $P(M)$
158 is typically between 2 and 6 times the median value, though in the case of the Orrin
159 formation, the range is 13 times the median.

160



161 The median values of the expert estimates tend to be similar to the median of the
162 distribution (within 10 %, except the Hopeman Sandstone which is within 20 %). The
163 individual expert estimates tend to cover the range from the 5th to 95th percentiles of
164 $P(M)$, though in 3 formations the minimum expert estimate exceeds the 5th
165 percentile of $P(M)$ and in the case of the Hopeman Sandstone Formation, the lowest
166 expert estimate as at around the 15th percentile. For 2 formations, the maximum
167 expert estimate is less than the 95th percentile of $P(M)$ and for all formations, the
168 highest value of $P(M)$ exceeds the maximum expert estimate by between 40 % and
169 120%.

170

171

172 The 5th to 95th percentiles expressed as a percentage of the median value of $P(M)$
173 can range from 8-62% for the 5th percentile and 170-307% for the 95th percentile
174 (the expert estimates show a similar range; Table 1). Figure 3 shows the range of
175 $P(M)$ against the number of unique values for the surface area, thickness, net:gross
176 and porosity. Surface area and thickness coincide because there are the same
177 number of unique values for all formations.

178

179 **4. Discussion**

180

181 The storage capacity estimates of 7 saline aquifers by a group of experts shows that
182 any single estimate by 1 expert might be a gross under or overestimation of the
183 median storage capacity. Even using a cohort of experts to provide independent
184 estimates of the storage capacity does not cover the full range of possible values
185 using just the data that those same experts collected. In particular, the range of
186 expert estimates underestimated the highest values of the storage capacity by at
187 least 40% (and up to 120%). As there is no reasons to assume that any one
188 combination of variables is more or less likely than any other, all possible
189 combinations must be assumed to have the same probability. Hence the storage
190 capacity calculated using all minimum or maximum values for all variables are
191 equally likely as any other individual combination, though there are more



192 combinations of variables that will produce storage capacities around the median
193 value than the extremes, making an estimate around the median more likely overall.

194

195 It is therefore evident that the uncertainty associated with a single estimate of CO₂
196 storage capacity for a saline aquifer is large compared to the precision with which at
197 least some published values are presented. Given both the small database upon
198 which estimates are typically based, and the inherent variability of the geological
199 parameters involved, the result is perhaps not surprising. The exercise upon which
200 this paper is based was conducted using only publicly available data. The experts had
201 access to a science library, and to the internet. It is apparent that the vast majority of
202 the data were derived by web-searching, including in most cases the data from the
203 library which must obviously be located before it can be consulted. A source of
204 uncertainty within the estimates is therefore the choice of search terms entered into
205 internet search tools, which could be crucial in either locating or missing key data
206 sources. In this study, porosity tends to have fewer independent sources in the
207 literature than the other parameters, leading to potential underestimation of the
208 uncertainty in comparison to other parameters and hence a smaller range of
209 calculated storage capacity values for this parameter. The ability to calculate the
210 uncertainty in a storage capacity estimate is therefore limited by data availability and
211 uncertainty is likely to be underestimated if this is not taken into account. In the case
212 of the Mains Formation, the range of calculated capacities is comparable to the
213 median value (Fig. 3), as all the experts located a single published porosity value. In
214 other words, the range of storage estimates is partly controlled by the number of
215 published values, and their accessibility or ease of location. In an extreme case as
216 with the Mains Formation, the range of $P(M)$ is likely to be underestimated.

217

218 A further potential source of variability in the storage estimates is the influence of
219 the individual assessors. Both personal judgement and previous experience have
220 been shown to influence geological interpretation (Polson and Curtis, 2010). In this
221 case, personal judgement is exercised when faced with parameters for which several
222 data values are available, with no indication of which are more representative of the
223 regional mean, and with no objective method of ranking the accuracy or importance



224 of the values. One approach under these circumstances is simply to average the
225 available values; the resulting mean clearly depends on which data have been
226 located by the individual expert.

227

228 Personal judgement is required when estimating net:gross ratio, as the most
229 common source of data are borehole logs with a summary lithology column showing
230 whether the sediments within the reservoir interval are interpreted as sandstone,
231 silty sandstone, siltstone or mudstone (there are no significant limestones in the
232 study area). Clearly mudstone is non-reservoir, and sandstone is potentially
233 reservoir, but a more-or-less arbitrary boundary between the two must be drawn. A
234 more experienced wireline log interpreter might choose to ignore the summary
235 lithology column of the composite log, and choose a value of, for example, the
236 gamma ray log as an arbitrary cut-off between reservoir and non-reservoir, or
237 estimate porosity (see below) and use an arbitrary minimum value of c. 10 % for
238 reservoir.

239

240 The most important control on the quality of the estimate of reservoir thickness is
241 probably the number of borehole logs used to estimate the mean value. The most
242 commonly used sources of data in this study (Andrews et al., 1990 and Cameron,
243 1993), typically present 3 summary borehole logs of each storage unit. However the
244 experts had access to 28 other composite (summary) borehole logs from the region,
245 released by the UK Government. Some experts choose to use the entire suite of logs
246 provided, others used only a subset. Even if all logs are used, it is possible to use a
247 range of methods to calculate mean regional thickness. For example, one can simply
248 calculate the mean of the storage unit thickness data; or one could to construct a
249 map and interpolate contours, then estimate mean thickness by some simple
250 graphical method involving dividing the storage unit into zones of constant thickness
251 interval and calculating an average thickness weighted to the areas of the zones. It is
252 also possible to use commercial software to perform both the contouring and the
253 reservoir volume calculation, in which case calculating the mean thickness is
254 unnecessary. Each of these approaches will result in different estimates of the
255 thickness of the reservoir (or final gross reservoir volume).



256

257 For porosity, literature values can be utilised if they exist, though if a range is given
258 then the mean must be estimated. Sometimes porosity data are only provided
259 graphically (as a cross-plot of porosity versus log permeability) and the mean value
260 can only be estimated visually as the points are frequently too dense to be read
261 individually from the graphs. Alternatively porosity can be calculated from borehole
262 logs using standard methods - using Formation Density Compensated (FDC) and
263 Compensated Neutron (CNL) logs for example - either manually or by using
264 petrophysical computer software if the wireline logs are available in digital form.
265 Again, the choice of method will influence the result. Measured porosity data are
266 most commonly from within hydrocarbon fields, where the spatial density of
267 boreholes is greatest. Whether the porosity of oilfield reservoirs is representative of
268 the associated aquifer, or is systematically higher and thus introduces a systematic
269 error in the estimate of aquifer porosity, is a controversial issue (e.g. Wilkinson &
270 Haszeldine, 2011) for which a judgement is necessary. In a commercial study, it is
271 possible to purchase porosity data measured from borehole core; unsurprisingly
272 none of the experts choose this option in this study.

273

274 The study reported here could be considered to be typical of regional studies
275 conducted with the aim of ascertaining which geological units in a region are worthy
276 of further study, i.e. a scoping study. The data available to the experts will be only a
277 fraction of the total data collected from the area, and the data must obviously be
278 located before being utilised. In any hydrocarbon province, it is unlikely that all
279 possible data can be used in a regional scoping study, due to the large (often very
280 large) volumes of data that have been collected, and due to the non-availability of
281 some (or much) of the data due to commercial confidentiality. Unless there are
282 previously published syntheses of data with calculated averages of parameters such
283 as the thickness of storage units, then some proportion of the total data will be
284 selected and utilised, inherently introducing uncertainty into the result.
285 Furthermore, the experts in this study could not spend unlimited periods of time
286 searching for data, or in processing it once obtained. Again, this restriction is likely to
287 be encountered in a regional scoping study, where many potential stores must be



288 assessed within a fixed budget. The North Sea is also typical of hydrocarbon
289 provinces in that there are a large number of boreholes drilled into relatively small
290 areas (i.e. producing hydrocarbon fields) and relatively small numbers of boreholes
291 in the much larger intervening areas. The spacing of the boreholes (data density) is
292 probably not atypical of other offshore hydrocarbon provinces, though onshore
293 hydrocarbon provinces may have much higher borehole densities (i.e. boreholes per
294 square kilometre). Borehole records in the UK are released by the Government, so
295 that the density of available data may be comparable to other areas of the world
296 where borehole density is greater but where drilling results are not so readily
297 available due to commercial confidentiality.

298

299 While the uncertainty of estimated storage capacities will vary from study to study,
300 and can be reduced by costly data collection (or possibly purchase) for any given
301 geological unit, the results here suggests that there is significant uncertainty in any
302 storage capacity estimate that does not include a site-specific estimate of
303 uncertainty. Note that this analysis does not take account of uncertainty in CO₂
304 density or storage efficiency. Storage efficiency, unless constrained on a unit-by-unit
305 basis, can introduce an order-of-magnitude uncertainty to a storage estimate (e.g.
306 Scottish Centre for Carbon Storage, 2009). The geological variability of a storage unit
307 hence appears to impart less uncertainty into the storage estimate than the storage
308 efficiency.

309

310 It is not possible to estimate the likely uncertainty of any single storage capacity
311 estimate as there is no way to know whether it is at lower, middle of upper range of
312 $P(M)$. However, these results show that the storage capacity could range from less
313 than 10% to over 300% of any single value. This supports the recommendation of
314 Chadwick et al. (2008) that a (single) calculated storage capacity that is similar to the
315 quantity of CO₂ to be stored should be regarded as a cautionary indicator for the
316 suitability of a storage unit for a particular project.

317

318 Data for this study was limited to that in the public domain which is probably
319 realistic for a regional study, where a potentially large number of candidate aquifers



320 are assessed for first-order suitability for storage (e.g. 1). It is probably not applicable
321 to a detailed study of a single aquifer, where every effort is made to reduce key
322 uncertainties and where confidential data may be available. For example, in the
323 estimation of aquifer thickness, every borehole log that penetrates the storage unit
324 could be utilised, removing the subjective element of choice associated with taking a
325 subset of the available data. It is also likely that a more rigorous approach to
326 uncertainty would be used in a single aquifer study, generating a reliable estimate of
327 the likely range of capacity. For this reason, the range of uncertainty for a detailed,
328 single aquifer study should be substantially less than that derived here.

329

330 5. Conclusions

331

332 The average standard deviation in CO₂ capacity for the storage units studied here is ±
333 64 %. This is substantially greater than the implied precision of many published
334 storage estimates. The geological uncertainty of a single storage capacity estimate
335 for a storage unit with no other assessment of uncertainty might be in the range of
336 30 – 245 % of the estimated value, or 6 to 520 % more conservatively . For storage
337 units where capacity is on the borderline of being economic or otherwise useable,
338 this uncertainty may materially influence the decision of acceptance or rejection of
339 the candidate unit. It should also be recognised that the analysis here does not
340 exclude the possibility of the useable, real-world, storage capacity of a candidate
341 storage unit being zero, due to for example, an unfixable leakage pathway or
342 regulatory issues.

343

344 Uncertainty documented in this study is due to a mixture of spatial variability in the
345 parameters combined with only limited data availability; the number of independent
346 (prior) estimates that are located for each parameter; and the variation in
347 interpretation of the same data by different experts. The range and standard
348 deviation values in this study should be considered to be minimum values. The
349 overall uncertainty is likely to be significantly larger as several sources of uncertainty
350 are not accounted for in this study, in particular uncertainty due to storage efficiency
351 could be larger than the geological uncertainty assessed here. Therefore a single



352 assessment of a storage capacity of a geological unit, with no associated assessment
353 of uncertainty, should be considered to have at least this degree of uncertainty in
354 the absence of other information.

355

356 **Author contribution**

357

358 MW designed the initial concept and supervised the storage assessment exercise. DP
359 performed the majority of the data analysis and interpretation.

360

361 **Competing interests**

362

363 The authors declare that they have no conflict of interest.

364

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366

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Table 1 – Range of individual expert and distribution ($P(M)$) of storage capacity estimates. Numbers in brackets are values expressed as a percentage of the median.

Storage unit	Expert Median (Mt CO ₂)	Expert Min (Mt CO ₂)	Expert Max (Mt CO ₂)	P(M) Median (Mt CO ₂)	P(M) 5 th percentile (Mt CO ₂)	P(M) 95 th percentile (Mt CO ₂)
Burns Sandstone Member	1905	119 (6%)	5381 (282%)	1755	144 (8%)	5035 (287%)
Beatrice Formation	120	37 (31%)	192 (160%)	110	25 (23%)	202 (185%)



Orrin Formation	96	18 (18%)	785* (819%)	102	16 (16%)	179 (176%)
Mains Formation	197	95 (48%)	245 (124%)	186	116 (62%)	316 (170%)
Hopeman Sandstone Formation	263	114 (43%)	457 (174%)	220	66 (30%)	490 (223%)
Findhorn Formation	1381	565 (40%)	3632 (263%)	1471	626 (43%)	3431 (233%)
Strath Rory Formation	763	75 (10%)	2300 (302%)	724	75 (10%)	2219 (307%)

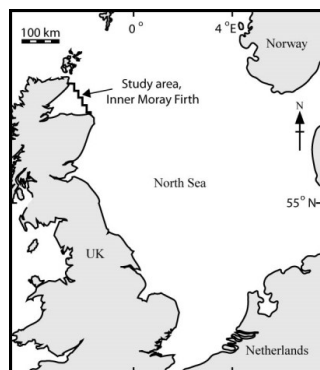
457 * This is significantly higher than the 95th percentile due to 1 expert estimating the
 458 volume of the formation to be significantly higher than the other experts.
 459

460

461

462 **Figure Legends**

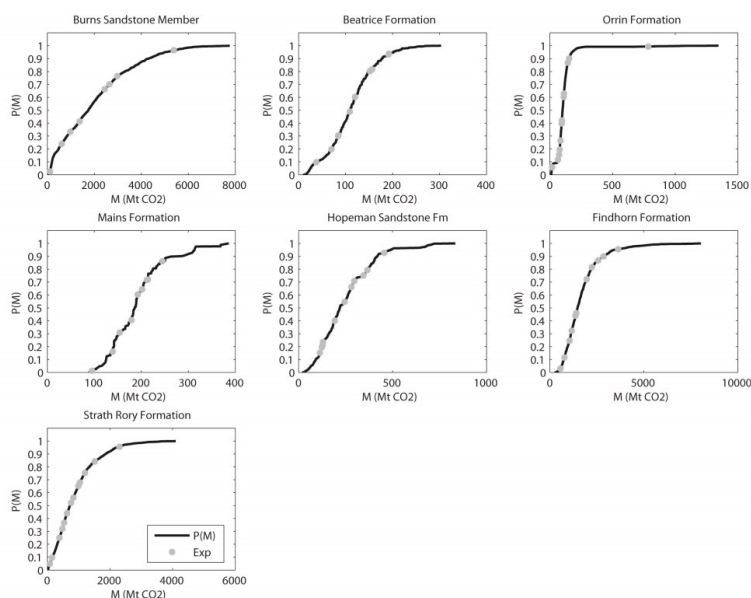
463



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465 Figure 1 – location map of study area.

466



467

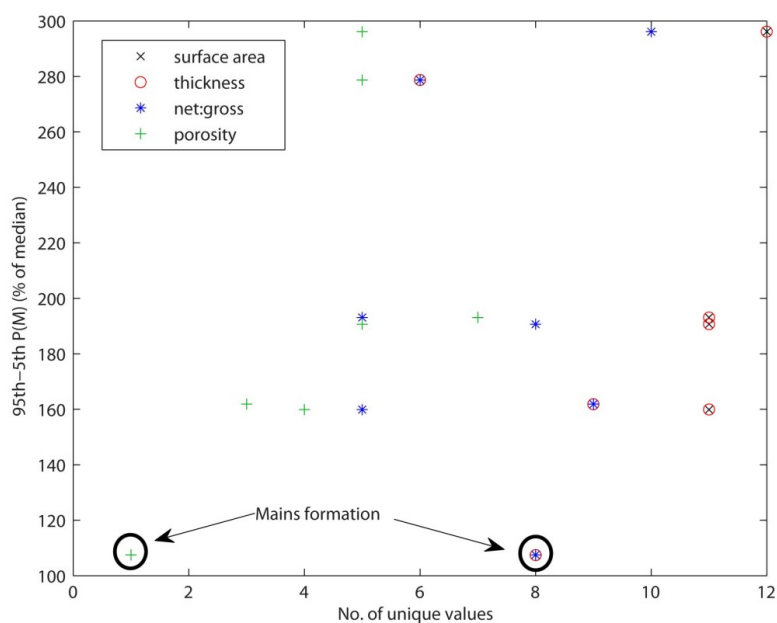
468 Figure 2. Range of storage capacity estimates using the different values for variables

469 found by group of experts for 7 saline aquifers. Range is shown as a cumulative

470 density function but does not represent the true probability density function for



471 each aquifer.



472

473 Figure 3. The Range of $P(M)$ (5th -95th percentile) against number of unique values for
474 the area, thickness, net:gross and porosity.