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1 **Communicating Uncertainty in Climate**
2 **Information for China: Recommendations and**
3 **Lessons Learned for Climate Services**

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ABSTRACT

24
25 Uncertainty is an inherent characteristic of climate forecasts and projections. While there is
26 an expanding body of international research focussed on identifying what climate information
27 users need to know about uncertainty, and how this should be communicated, very little of
28 this has been conducted in a Chinese cultural context. In this paper we report on the findings
29 of interviews with climate experts (n=28) and (potential) users of climate information in
30 China (n=18) at seasonal and multidecadal timescales, with the objective of addressing the
31 following research questions: 1) What information about uncertainty in climate forecasts and
32 projections is currently provided to users in China?; 2) What do climate experts believe that
33 users need to know about uncertainty?; 3) What information about uncertainty would
34 (potential) users like to receive?; 4) What challenges do providers and users perceive with
35 respect to the communication of uncertainty? We find that while seasonal forecasts are
36 predominantly presented deterministically current and potential users are aware that there is
37 uncertainty associated with them. Climate experts highlight the probabilistic nature of
38 forecasts and the conditional nature of forecast quality, as areas for communication
39 development. Interviews with (potential) users indicate a) that preferences for deterministic
40 information are not unanimous; b) probabilities associated with conditions being above/below
41 normal may only be considered useful for decision making if they are >60%; and c)
42 preference for forecasts that provide probability of user-relevant thresholds being crossed. At
43 multidecadal timescales we observe lower engagement with projections, and less evidence of
44 interaction between providers and recipients, suggesting that development of climate services
45 at multidecadal timescales will need to first highlight the added value of these. We present
46 key recommendations for communicating uncertainty in seasonal forecasts, and exploring the
47 potential value of multidecadal projections.

48 Key words: climate services, communication, uncertainty, user needs, China.

49 **1. Introduction**

50 Uncertainty in climate information arises from a range of sources including deficits in
51 understanding, modelling limitations and the inherent unpredictability of the climate system
52 (Risbey and O’Kane, 2011; Slingo and Palmer, 2011). At longer timescales, uncertainties
53 about future greenhouse gas emissions scenarios become increasingly important (Hulme and
54 Dessai, 2004). Effective mobilisation of scientific information requires recipients to have
55 some awareness of the uncertainty surrounding it and the quality of the underlying scientific
56 process that produced the information (Fischhoff and Davis, 2014). If these are not
57 adequately communicated to those using this information to inform decision making, it may
58 result in a false sense of certainty, maladaptive decision making and a loss of trust in
59 providers (Macintosh, 2013, LeClerc and Joslyn, 2015). It is therefore important to identify
60 appropriate ways to characterise and communicate uncertainty in climate information. While
61 the question of how to communicate uncertainty in climate information has received an
62 increasing attention in international research (e.g. Taylor et al., 2015), there is a dearth of
63 research on how to convey this in a specifically Chinese cultural context. This paper brings
64 together qualitative evidence from a series of interviews with climate scientists, along with
65 current and potential Chinese climate information users to explore: current provision of
66 uncertainty information in climate forecasts and projections, users’ preferences for receiving
67 this information, and the communication challenges faced by providers and users.

68 Many fields have explored how to best characterise and communicate scientific
69 uncertainty (e.g. Spiegelhalter et al., 2017). In the context of climate services, Otto et al.,
70 (2016) highlight the challenge of adequately characterising uncertainty in climate
71 information, while tailoring it to meet the needs of users. The World Meteorological
72 Organisation has promoted more responsible forecast provision by moving away from
73 potentially misleading deterministic products (typically using mean model outputs) towards

74 expressing climate anomalies in terms of probabilities (e.g. tercile categories based on the
75 distribution of model outputs) (Goddard et al., 2010). Studies suggest that when tailored to a
76 specific task and the recipients' cognitive processes, non-experts can effectively use
77 probabilistic forecasts in decision-making (Joslyn and Le Clerc, 2012; Joslyn and Le Clerc,
78 2013; Savelli and Joslyn, 2013). Indeed, providing information about uncertainty may help
79 to sustain trust in cases where forecasts do not match subsequent events (e.g. false alarms)
80 (Joslyn and Le Clerc, 2013). Nonetheless, organisational users of climate and weather
81 information services may vary in their tolerance for uncertainty in climate information (e.g.
82 Taylor et al., 2015).

83 While the last decade has seen a proliferation of research on the communication of
84 climate information, relatively few peer reviewed studies have taken place in China, with a
85 majority focussing in North America, Europe and Australia. This represents a gap in
86 understanding, as the comparatively limited amount of work focussed on user needs and
87 climate and weather communication in China suggests that cross-cultural differences exist.
88 For instance, while the ordinal nature of the yellow, amber, red 'traffic light' weather
89 warning system is well understood in the UK and US (e.g. Lesch et al., 2009; Taylor et al.,
90 2019), this is not the case in China, where red is not intuitively interpreted as representing a
91 higher threat level than amber (e.g. Wong and Yan, 2002; Lesch et al., 2009). Moreover,
92 scoping work has suggested that institutional preferences for deterministic information may
93 be particularly strong in China (Golding et al., 2017a; Norbert et al., 2015). Given China's
94 exposure to climate hazards such as heavy rainfall, flooding, tropical cyclones, heatwaves
95 and drought, addressing this is important for the development of effective climate services
96 (Hewitt et al., 2018, Wang et al., 2020).

97 As part of the Newton Fund's Climate Science for Services Partnership (CSSP) China
98 programme, a project on improving the treatment of uncertainty was undertaken, with a core

99 focus on addressing the following research questions: 1) What information about uncertainty
100 in climate forecasts and projections is currently provided to users? 2) What do climate
101 information providers believe that users need to know about uncertainty? 3) What
102 information about uncertainty would users (and potential users) like to receive? 4) What
103 challenges in the communication of uncertainty exist for the providers and users of climate
104 information?

105 In this paper, we report findings from a series of interviews with climate experts and
106 (potential) climate information users to address these questions, and outline the key
107 recommendations emerging from this work.

108 **2. Methodology**

109 ***2.1 Expert interviews***

110 Between November 2017 and April 2018, we conducted 28 expert interviews with
111 climate scientists from China and the UK. Eighteen had expertise in seasonal precipitation
112 forecasting for China (China=13, UK=5). Ten (all Chinese scientists) had expertise in
113 multidecadal temperature and precipitation projections relevant to China. As Part 1 of the
114 interviews required a constrained geographic focus, we concentrated on regions where other
115 CSSP China projects were focussing on climate services for seasonal precipitation
116 forecasting (Middle Yangtze) and climate adaptation (Lower Yangtze) (e.g. Bett et al., 2017;
117 Golding et al., 2017a; 2017b; Sun et al., 2019). Experts were identified through the CSSP
118 China programme and a review of the literature. Participants were approached through the
119 UK Met Office, China Meteorological Administration (CMA), Institute of Atmospheric
120 Physics (IAP) and the 2018 Forum on Regional Climate Monitoring, Assessment and
121 Prediction for Asia (FOCRAII). Interviews comprised two stages, and were conducted in
122 English or Mandarin. Part 1 elicited expert judgements about the importance of different
123 sources of predictability and uncertainty in seasonal forecasts or multidecadal projections.

124 Part 2 focussed on experts' perceptions of users' needs for receiving information about
125 uncertainty including: what they believed users needed to know about uncertainty, their
126 experience of providing this information, and any challenges they had encountered or
127 anticipated with respect to communicating uncertainty. In this paper, we focus on responses
128 to Part 2, with in-depth analyses of Part 1 reported in Grainger et al. (2018).

129 **2.2 User interviews**

130 Between March 2018 and July 2018, we interviewed 18 current and potential users of
131 climate information. Participants were initially recruited through contacts in CMA and other
132 CSSP China projects, and asked to recommend other potential contacts who may be willing
133 to take part. As two other projects within the CSSP China programme were conducting
134 interviews with some of the same target participants, a joint interview protocol combining
135 questions from each of the projects was developed to limit the risk of stakeholder fatigue
136 (Verdon-Kidd et al., 2012). Participant characteristics are summarised in Table 1. Five of the
137 18 participants identified as decision makers, while 11 had intermediary roles as either in-
138 house meteorologists ($n=3$) or researchers/analysts ($n=8$) who provide information to advise
139 decision makers. The remaining two were academic researchers. Six participants currently
140 received seasonal climate forecasts, while two received multidecadal projections. The
141 remainder ($n=10$) did not currently receive climate information, but were interested in doing
142 so (seasonal=5, multidecadal=2, general=4). All organisations operated at either a city,
143 province or river basin level.

144 Interviews were conducted in Mandarin or English. Participants were first asked about
145 their organisation's approach to uncertainty. This was followed by questions about
146 information about uncertainty in climate products currently received (current users only),
147 preferences for receiving information about uncertainty in climate products, and any
148 challenges in using or interpreting this information. Those interested in seasonal forecasts

149 were asked to provide feedback on the format and layout of a Chinese translation of a
 150 seasonal forecast produced by the Met Office for the Three Gorges Dam (Bett et al., 2017).

151

152 Table 1 Characteristics of participants in interviews with current and potential users of
 153 climate information

	<i>Sector</i>	<i>Role</i>	<i>Status</i>
1	Energy/Water	Intermediary	Current user: Seasonal
2	Energy/Water	Intermediary	Current user: Seasonal
3	Urban	Intermediary	Potential user
4	Urban	Intermediary	Potential user
5	Urban	Intermediary	Potential user
6	Water/Urban	Intermediary	Potential user
7	Energy/Urban	Decision maker	Potential user
8	Urban	Intermediary	Current user: Seasonal
9	Commercial	Decision maker	Potential user
10	Energy/Urban	Decision maker	Potential user
11	Urban	Intermediary	Potential user
12	Water/Urban	Intermediary	Potential user
13	Energy	Decision maker	Current user: Seasonal
14	Water	Intermediary	Current user: Seasonal
15	Water	Decision maker	Current user: Seasonal
16	Commercial	Intermediary	Potential user
17	Academia	Researcher	Current user: Multidecadal
18	Academia	Researcher	Current user: Multidecadal

154

155 **2.3 Analysis**

156 Thematic analysis, a procedure for identifying and coding key themes in qualitative
 157 data (Guest et al., 2011), was used to analyse the interviews. We applied a mixture of
 158 deductive coding, where we examined whether themes suggested by prior research were
 159 present in the interviews (e.g. preference for deterministic information), and inductive
 160 coding, where themes emerge from interviews.

161 **3. Results and discussion**

162 *3.1 What information about uncertainty in climate forecasts and projections is currently*
163 *provided to users?*

164 *3.1.1 Deterministic forecasts*

165 Around half of the Chinese experts in seasonal forecasting (6 out of 13) reported
166 direct experience of providing forecasts to sectoral decision makers. They indicated that
167 numeric probabilities were rarely delivered to end-users in formal forecast communications.
168 However, where established relationships between providers and users existed, informal
169 discussions relating to uncertainty (e.g. forecast reliability) do take place. This was consistent
170 with findings from user interviews, where only 1 of 6 seasonal forecast recipients reported
171 receiving probabilistic information. This took the form of qualitative probabilistic statements
172 (e.g. likely, unlikely, etc). The remaining 5 reported receiving deterministic forecasts,
173 although one noted that being provided with ranges (i.e. min, max) did capture forecast
174 uncertainty to some extent.

175 *“It’s deterministic. It gives a max/min range and mean in mm but not probabilities.”*

176 [Interviewee 2, Energy and Water Sector]

177 *“The range itself is already a symbol of uncertainty.”* [Interviewee 15, Water Sector]

178 *3.1.2 Awareness of uncertainty*

179 While most seasonal forecast users received deterministic forecasts, all were aware
180 that forecasts are inherently uncertain. Despite the potential for “false-alarms” to undermine
181 trust (White and Eiser, 2006; Trainor et al., 2015; Ripberger et al.,2015), seasonal forecast
182 users indicated a generally high level of trust in CMA, despite recognition that forecasts are
183 not always accurate.

184 *“Uncertainty comes with the forecast. The certain forecast itself brings uncertainty.”*

185 [Interviewee 1, Energy and Water Sector]

186 *3.1.3. Limited uptake of multidecadal projections*

187 At multidecadal timescales, we found less evidence of demand for longer term
188 projections amongst decision makers. Interviews with experts also indicated lower levels of
189 interaction between providers and non-academic users of projections. Scientists reporting
190 experience of providing climate projections to the central government indicated that they had
191 little direct interaction or feedback from policy makers. Both academic users of climate
192 projections reported attending training workshops run by climate scientists. However, they
193 received only raw or processed data, without additional summaries of uncertainty.

194 “... *the group who is working on these GCM models, they will give us like a one-day*
195 *or two-day training and also we have regular meetings. And they will use examples to*
196 *tell us, what are the uncertainties... or why they cannot change this.*” [Interviewee 17,
197 Academic researcher]

198 *3.1.4 Current provision: Summary*

199 There is currently limited formal provision of information about uncertainty in
200 climate products. Seasonal forecasts are usually presented deterministically, although issues
201 related to forecast quality and expert confidence may be conveyed informally. This is
202 consistent with earlier research indicating a predominance of deterministic information in
203 climate provision in China (Norbert et al., 2015). However, as has been observed in other
204 contexts (Morss et al., 2008), users understand that uncertainty surround deterministic
205 forecasts

206 While we observed strong connections between seasonal forecast users and CMA
207 providers, there appears to be less interaction between providers and recipients of climate
208 projections. This resonates with findings from work with Chinese water managers, showing
209 that while there is frequent interaction between CMA and users, when it comes to forecasts
210 and observations, this is not yet the case for projections (Khosravi et al., 2020). Indeed, when

211 it comes to long-term planning historical observations may be preferred to projections
212 (Khosravi et al., 2020). Lack of engagement with projections, may also reflect a stronger
213 focus on mitigation than adaptation in China's climate policy (Engels, 2018; He et al., 2013).

214 ***3.2 What do experts think that users need to know about uncertainty?***

215 Experts' beliefs about users' needs do not always correspond with actual needs
216 (Bruine de Bruin & Bostrom, 2013). However, it is important to identify what experts believe
217 users need to know to avoid misleading interpretations, and where expert and user
218 perceptions differ..

219 *3.2.1 Trade-off between completeness and comprehension*

220 All of the experts interviewed agreed that uncertainty should be communicated to users.
221 However, most perceived a tension between providing a detailed account of probability and
222 reliability and information that is readily understandable. Trade-offs between detail and
223 understandability are recognised in the wider climate communication literature (Stephens et
224 al., 2012), but may be of particular importance in China, where there is currently limited
225 exposure to probabilistic forecasts. Most experts felt that information needed to be tailored to
226 specific users, with Chinese providers noting variability in the level of complexity that
227 different users wanted and had the capacity to understand.

228 *3.2.2 Perceived preference for deterministic information*

229 While current provision of seasonal forecasts is largely deterministic, several Chinese
230 experts felt that probabilistic information should be provided. However, some raised
231 concerns about this may not be accepted or understood. Echoing previous findings (Norbert
232 et al., 2015), they perceived a preference for deterministic information amongst users. When
233 asked how they thought that probabilities should be presented, one participant suggested
234 verbal descriptions may be more acceptable to users than numeric probabilities. Others
235 highlighted a need for education about the probabilistic nature of forecasts.

236 *3.2.3 Explaining the conditional nature of forecast quality*

237 Few scientists believed that users required a full account of sources of predictability in
238 seasonal forecasts. However, some felt that credibility may be bolstered by users believing
239 that experts know these things. Echoing this, an intermediary commenting on the seasonal
240 briefing indicated that while decision makers within their organisation were unlikely to
241 consult such a document - relying instead on advice provided ‘in house’ - its ‘scientific’
242 appearance may instil confidence in its quality. While experts agreed that detailed
243 descriptions of sources of predictability and uncertainty were not needed, many felt that some
244 explanation of why these things affected forecast quality should be provided. The El Nino-
245 Southern Oscillation (ENSO) was identified as the most important source of predictability for
246 seasonal precipitation in the Middle Yangtze, with subjective judgements of forecast quality
247 tending to be conditional on whether it was an El Nino year or not. It was therefore suggested
248 that brief explanations for forecast quality varying from year-to-year could be beneficial.
249 Indeed, one provider expressed concern that year-to-year variability could harm credibility.

250 *3.2.4 Limited interaction with recipients of multidecadal projections*

251 For multidecadal projections, providers had far less interaction with recipients and thus
252 fewer expectations about what information about uncertainty users required. Some felt that
253 provision information about different areas of uncertainty (e.g. natural variability, model
254 uncertainty, scenario uncertainty) could be useful. Indeed, other recent studies suggest that
255 some potential user are unaware of the scenario-based nature of projections, conflating them
256 with forecasts (Khosravi et al., 2020). One participant suggested providing confidence
257 categories and ‘worst case scenario’ statements. However, these were acknowledged as
258 speculative suggestions.

259 *3.2.5. Maintaining credibility*

260 Credibility and legitimacy are core components of usable climate knowledge (Lemos et
261 al., 2012). The seasonal forecast providers interviewed emphasised that this is especially true
262 in a Chinese cultural context, where adherence to procedures and hierarchy within the
263 delivery processes may have an importance beyond the forecast information itself. Indeed,
264 one expert expressed that this could be more important than the accuracy of the forecast.

265 *3.3.6. Summary: Experts' perceptions of user needs*

266 Experts perceived trade-offs between providing detailed explanations of uncertainty and
267 overloading users with information. They agreed that probabilistic forecast information
268 should be communicated, though opinions on how to do this varied. Likewise, explaining
269 why the performance of forecast models varies from year-to-year was felt to be potentially
270 useful. At multidecadal timescales, the experts interviewed had less interaction with
271 recipients of this information, and thus fewer expectations regarding user needs.

272 ***3.4 What information about uncertainty would (potential) users like to receive?***

273 *3.4.1 Going beyond deterministic forecasts*

274 Consistent with prior research (Norbert et al., 2015), experts perceived a user
275 preference for deterministic forecasts. However, user interviews revealed a more nuanced
276 picture. While most potential users preferred deterministic forecasts (4 in 5), all six
277 experienced users wished to receive probabilistic information and some detail about the
278 forecast process.

279 *“Yes, the probability is actually very essential... I would love to receive a relatively*
280 *clear statement like there's an 80% of probability to have such a weather.”*

281 [Interviewee 13, Energy Sector]

282 Nonetheless, some noted that while they welcomed probabilistic information, higher
283 level decision makers within their organisation required deterministic input.

284 [On probabilities] *“They just ask directly for an accurate number. This is how it*
285 *works.”* [Interviewee 14, Water sector]

286 This highlights that in some organisations decision makers do not consult forecast
287 information directly, relying on interpretations from technical staff who have scientific
288 expertise in areas other than climatology (e.g. hydrology, engineering). Climate service
289 development should therefore take into account that the end-users of climate products may
290 not always be decision makers, but those who advise them. Nonetheless, engaged decision
291 makers do exist, as illustrated by one energy sector decision maker who actively sought
292 explanations for why particular climatic conditions are expected.

293 *“....at the end of January in 2015 and 2016, during that very cold period, I learned*
294 *from the (online) forum that there was an abnormal weather in the Arctic pushed a*
295 *cold air toward here. With such an explanation, I began to understand how does this*
296 *cold air happen.”* [Interviewee 13, Energy sector]

297 3.4.2 High probabilities and seasonal extremes

298 While current forecast users wanted to receive probabilistic information, a caveat was
299 that only very high probabilities (>80%) were felt to be useful for decision making. Users
300 were less inclined to engage with lower probabilities (<60%) as the forecast may be
301 perceived as lacking credibility.

302 *“If the forecast is highly probable, for example, more than ninety, it's useful. But if it's*
303 *less than ninety or less than eighty or whatever, it's not.”* [Interviewee 15, Water
304 Sector]

305 *“40-50% probabilities are confusing and seem not so credible.”* [Interviewee 1,
306 Energy Sector]

307 This resonates with international research on seasonal forecast uptake, where lower
308 probabilities may be perceived as too uncertain to support decision making (e.g. Bruno

309 Soares and Dessai, 2016). However, the quotes above were made in response to statements
310 about seasonal conditions being above or below historical averages, where probabilities close
311 to 50% may be interpreted as reflecting a lack of knowledge. Hence, the usefulness of
312 forecasts showing lower probabilities of seasonal extremes may be perceived differently.
313 Indeed, a desire for forecasts for seasonal extremes or user-defined thresholds was expressed
314 throughout the user interviews. This is consistent with research in other countries, where
315 tercile-based forecasts have been found to have limited value for decision making (Haines,
316 2019). Additionally, some participants indicated a preference for communications linking
317 forecasts to specific decisions and actions; with three expressing a preference for reports
318 where explicit recommendations are provided. Again, this resonates with findings in the
319 broader literature, emphasising demand for forecasts linked to specific actions (dePerez et al.,
320 2015; Nkiaka et al., 2019; Weyrich et al., 2019).

321 *3.4.3 Preferred forecast formats*

322 While detailed scientific information about forecast processes was not desirable for
323 most users, having some explanation was felt to be helpful. Seven participants expressed a
324 preference for concise sentences qualifying the forecast in terms of: types of climate
325 variability considered; justification of high probability for particular conditions; model
326 reliability; and overall forecaster confidence. For probabilistic information, five participants
327 indicated a preference for numerical/graphical information linked to weather variables and
328 expected ranges. A further three also expressed a preference for tailored statements (e.g.
329 spatially appropriate and related to specific decisions or actions). One potential user from the
330 urban transport sector, elaborated on the type of statement that they would like to receive.

331 *“...this March or April, there's (a high) probability to have high temperature, heavy*
332 *rain ... along our [transport] Line 1 or Line 2 and the intensity of it. If we can just*

333 *know it in advance, that would be really helpful for us.*” [Interviewee 4, Urban
334 Transport Sector]

335 3.4.5 *Support for academic users of multidecadal projections*

336 For multidecadal projections, our sample size made it impossible to identify common
337 themes and preferences with respect to how users would like to receive this information.
338 However, the academic users interviewed indicated a wish for support in integrating
339 projection uncertainties into their research.

340 3.4.6 *Summary: User preferences*

341 We find that while those with lower experience of seasonal forecasts may prefer
342 deterministic forecasts, experienced users wish to receive probabilistic information. However,
343 where forecast probabilities are near to climatology they may not be perceived as credible.
344 We also observe demand for forecasts linked to user-defined thresholds rather than historical
345 averages, and for decision-relevant advisory statements. While lengthy technical descriptions
346 are unlikely to be directly consulted, short explanations justifying the forecast are welcomed.

347 **3.5 *What challenges in the communication of uncertainty exist for the providers and users*** 348 ***of climate information?***

349 3.5.1 *Tolerance for uncertainty*

350 While climate experts perceived preferences for deterministic information to be a
351 barrier to the provision of probabilistic seasonal forecasts, we found that this was not
352 unanimously the case, with more experienced users wishing to receive probabilistic
353 information. Recognising that users vary in their tolerance for uncertainty is important.

354 3.5.2 *Reconciling scientific feasibility with user preferences*

355 Experts identified the management of user expectations as a key challenge. Tensions
356 between what users want and what can reasonably be provided by the available science was
357 underscored in user interviews. For instance, seasonal forecast users often expressed

358 preferences for spatial resolutions that are currently impossible to realise. Similarly, 80% was
359 cited as a threshold for decision making for forecasts for above/below average conditions, a
360 value that may rarely be reached. While we recommend that providers work toward
361 providing forecasts for user-relevant thresholds and extremes, this is contingent on its
362 scientific feasibility.

363 *3.5.3 Understanding*

364 Both expert and user interviews highlighted challenges related to understanding and
365 interpreting complex scientific information. When asked to provide feedback on the sample
366 seasonal precipitation forecast for the Middle Yangtze, most participants focussed their
367 attention on the summary box at the top of the page instead of the more detailed information
368 in the main body. This demonstrates the importance of ensuring that the most salient
369 characteristics of communication contain the most important information, and that this is
370 easily understood (Klopprogge et al., 2007; Spiegelhalter et al., 2011).

371 *3.5.4 Limited provider-user interaction for multidecadal projections*

372 At multidecadal timescales a key challenge for providers was a lack of feedback from
373 non-academic users. The two academic users interviewed did however indicate that while
374 they received training on global climate models they were currently receiving raw and
375 processed data without means to integrate uncertainties into their own research.

376 *“Here are the raw data... many tiers of data. We copy from them (information*
377 *providers), and we use the exact same way they presented data and we use it in our*
378 *research. So, for me, this is a big issue.” [Interviewee 17, Academic]*

379 *3.5.5 Summary: Challenges*

380 While preferences for deterministic information can be a challenge for the
381 communication of uncertainty, experts may overestimate this. However, managing user
382 expectations of what science can feasibly provide is of critical importance, especially if

383 tailoring climate information products to decision relevant thresholds or extremes, where high
384 probabilities of exceedance may be rare. While it is important for climate information
385 products to address decision making needs, it should be recognised that in some cases end-
386 users may be intermediaries advising decision makers rather than decision makers
387 themselves. While we have less evidence regarding user needs for multidecadal projections,
388 our findings do highlight a current dearth of products for academic users that allow them to
389 integrate uncertainties into their own work.

390 **4. Recommendations and lessons learned**

391 Based on the findings of this this work, we produced a set of recommendations for
392 providers on the treatment of uncertainty in climate information for climate services in China
393 (Grainger et al., 2019). In this section we outline recommendations for the communication of
394 uncertainty in seasonal forecasts, developing the provision of multidecadal projections, and
395 reflect on the challenges and lessons learned in undertaking this work.

396 ***4.1 Communicating uncertainty in seasonal forecasts***

397 *4.1.1. Work to provide seasonal forecasts that are based on user-relevant thresholds*

398 Forecasts for the exceedance of user-defined thresholds may be more useful than
399 forecasts for conditions being above/below average. Indeed, likelihoods between 40%-60%
400 for above/below average conditions were felt to offer little useful information. We
401 recommend that, where scientifically appropriate, providers work with users to identify
402 decision-relevant thresholds and explore whether providing information about likelihood of
403 exceedance is possible.

404 *4.1.2 Explain conditionality*

405 Forecast quality is conditional on sources of predictability, such as ENSO, meaning
406 that forecast models may perform better in some years than others. To avoid confusion and
407 loss of trust that may result from this variability in forecast quality, we recommend that

408 forecasters explain that forecast performance (i.e. skill) is conditional on these sources of
409 predictability (e.g. precipitation forecasts for the Middle Yangtze being more reliable in El
410 Nino years).

411 *4.1.3 Provide an indication of forecasters' confidence in the forecast quality*

412 Non-specialist users may not want to receive detailed technical information. However,
413 having an indication of forecasters' confidence regarding forecast quality is valued by users.
414 Some users already receive this through discussions with CMA providers.

415 *4.1.4 Provide forecasts based on climatology when skill is low*

416 Where seasonal forecasts do not perform better than historical data (climatology) they
417 cannot provide added value, and may be misleading. However, Chinese forecast providers
418 indicate that they cannot fail to provide a forecast when requested by users. Where
419 forecasting models lack skill, we recommend that forecasts be provided based on
420 climatology, with it being explained that that the decision to base the forecast on models or
421 observations is made on the best available science.

422 *4.1.5 Ensure that the most important decision-relevant information is in the summary box*

423 Our user interviews did not reveal a unanimous consensus as to the precise format in
424 which information regarding probability and forecast quality presented (e.g. graphs, numeric
425 probabilities). However, feedback on the precipitation forecast for the Middle Yangtze,
426 highlighted the importance of ensuring that the most critical information is the most salient.
427 When presented with the forecast, most participants focussed predominantly on the summary
428 box at the top of the page, with limited attention to the more detailed text and diagrams
429 below. Where a briefing style is used, the summary at the start should contains the most
430 decision relevant elements of the forecast. This could include: likelihood of decision relevant
431 threshold exceedance, spatial resolution of forecast, and any user-specific advisory statements
432 provided as part of the forecast.

433 ***4.2. Developing the provision of multidecadal projections***

434 Throughout this project, we observed low engagement with climate projections, and
435 limited interaction between providers and users. To develop the provision of multidecadal
436 climate services it may be necessary to actively explore the potential for climate projections
437 to inform long term planning decisions with users. For instance, through examining the
438 benefits of using climate change projections alongside historical observations.

439 ***4.3 Challenges, limitations and lessons learned***

440 While our sample sizes falls within recommended ranges for qualitative interviews
441 (Sim et al, 2018), sampling was constrained by the need for introductions to be made through
442 established contacts. While this work enabled us to identify a coherent set of themes,
443 challenges and recommendations, we must acknowledge that the user sample was
444 comparatively small and concentrated in state-run water and energy organisations. Other
445 sectors and private organisations were underrepresented. Despite an initial intention to focus
446 on decision makers as end-users of climate information, we found that within some
447 organisations decision makers do not directly consult climate information, relying on others
448 to interpret this information and provide recommendations. Nonetheless, in tailoring climate
449 products, it is important to understand the choices they will inform, even if they will not be
450 directly consulted by decision makers.

451 A key challenge in this research was the limited uptake and interest in multidecadal
452 projections amongst (potential) users. However, a current lack of engagement with
453 multidecadal projections does not indicate that these timescales are irrelevant to decision
454 making. Historical observations may be used in preference to projections for long term
455 planning (Khosravi et al., 2020). Bringing together providers and potential users to raise
456 awareness of the added value that projections may provide may be crucial to promote uptake;

457 an approach taken by Sun et al. (2019), who brought together providers, decision makers and
 458 researchers to explore climate services for urban sector adaptation.

459 Another notable challenge related the fact that there were not always direct Chinese
 460 analogues for English-language terminology related to uncertainty in climate information
 461 (e.g. probability, accuracy, reliability and skill having distinct meanings) (American
 462 Meteorological Society, 2020). The lack of a common vocabulary for discussing different
 463 aspects of uncertainty with experts did pose a challenge, suggesting the need for appropriate
 464 terminologies to be identified for cross-cultural collaborations.

465 **5. Summary**

466 This work was undertaken to examine the current provision of information about
 467 uncertainty in climate forecasts and projections for China, assess users’ preferences and
 468 experts’ perceptions of user needs, and explore the challenges associated with communicating
 469 uncertainty. Key recommendations are summarised in Table 2.

470

471 Table 2: Summary of Recommendations

Recommendation	Timescale
Where the underlying science permits, work to provide seasonal forecasts that are based on user-relevant thresholds.	Seasonal forecasts
Explain conditionality (i.e. why forecasts may perform better in some years than others).	
Provide an indication of forecasters’ confidence in the forecast quality.	
When the skill of forecast models is low, provide forecasts based on climatology, explaining to users that in some years historical data provides the best guide to seasonal conditions.	
Ensure that the most important decision-relevant information is placed in the part of the document most likely to be noticed first (e.g. the summary box on seasonal briefings)	
Where historical observations alone are used for long-term planning decisions, explore the potential added value that climate projections could provide by bringing together providers, decision makers and intermediaries.	Multidecadal projections

Identify who within the user organisation will receive and use climate products (e.g. decision makers, intermediaries, both)	General
Identify the type of choices that climate products will be expected to inform, even if they will not be directly consulted by decision makers.	
As there are not always direct Chinese translations for English words describing different aspects of uncertainty (e.g. probability, reliability, accuracy, skill), identify terminology that can be effectively used to refer to these in cross-cultural collaborations.	

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At seasonal timescales we find that current provision is mainly deterministic.

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However, while experts perceive a preference for deterministic information amongst users,

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this is not universally the case, with experienced users wishing to receive probabilistic

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forecasts. Nonetheless, when it comes to forecasts presented as likelihood of above/below

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average, only high probabilities (>60%-80%) are perceived as useful, with probabilities

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~50% perceived as not conveying useful information. As anomaly-based forecasts for

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above/below average conditions can be challenging to integrate into decision making

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processes, we recommend that developers of seasonal climate services for China explore the

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feasibility of providing probabilistic forecasts based on user-defined thresholds. Our

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exploration of preferences for receiving information about uncertainty did not identify a

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“most preferred” format. However, it did highlight the importance of having ‘summary

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boxes’ that contain all decision critical information. While detailed technical information may

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be of limited interest, many users did welcome having some explanation and justification for

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the forecast. Indeed, as forecast performance depends on sources of predictability, we suggest

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that short statements regarding forecast quality be provided. For instance, when model skill is

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low, providing forecasts based on historical averages (climatology) and explaining that this

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represents the best available science, may offer a credible way to address the conditional

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nature of forecast quality. At multidecadal timescales limited user engagement made it

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impossible to provide evidence-based recommendations for communication. However our

492 findings suggest that the development of climate services at multidecadal timescales will
493 require exploration of the added value that projections may provide for long-term planning.

494

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500

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