



# *Advancing research for seamless Earth system prediction*

Article

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<b>Abstract:</b>	<p>Whether on an urban or planetary scale, covering timescales of a few minutes or a few decades, the societal need for more accurate weather, climate, water and environmental information has led to a more seamless thinking across disciplines and communities. This challenge, at the intersection of scientific research and society's need, is amongst the most important scientific and technological challenges of our time. The "Science Summit on Seamless Research for Weather, Climate, Water, and Environment" organized by the World Meteorological Organization (WMO) in 2017, has brought together researchers from a variety of institutions for a cross-disciplinary exchange of knowledge and ideas relating to seamless Earth system science. The outcomes of the Science Summit, and the interactions it sparked, highlight the benefit of a seamless Earth system science approach. Such an approach has the potential to break down artificial barriers that may exist due to different observing systems, models, time and space scales, and compartments of the Earth system. In this context, the main future challenges for research infrastructures have been identified. A value cycle approach has been proposed to guide innovation in seamless Earth system prediction. The engagement of researchers, users and stakeholders will be crucial for the successful development of a seamless Earth system science that meets the needs of society.</p>
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## **Advancing Research for Seamless Earth System Prediction**

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95 **Capsule Summary:** The WMO convened the “Science Summit on Seamless Research for  
96 Weather, Climate, Water, and Environment” to guide the Commission for Atmospheric Sciences  
97 (CAS-17) on future scientific research needs and requirements.

98 **Abstract**

99           Whether on an urban or planetary scale, covering timescales of a few minutes or a few  
100 decades, the societal need for more accurate weather, climate, water and environmental  
101 information has led to a more seamless thinking across disciplines and communities. This  
102 challenge, at the intersection of scientific research and society’s need, is amongst the most  
103 important scientific and technological challenges of our time. The “Science Summit on Seamless  
104 Research for Weather, Climate, Water, and Environment” organized by the World  
105 Meteorological Organization (WMO) in 2017, has brought together researchers from a variety of  
106 institutions for a cross-disciplinary exchange of knowledge and ideas relating to seamless Earth  
107 system science. The outcomes of the Science Summit, and the interactions it sparked, highlight  
108 the benefit of a seamless Earth system science approach. Such an approach has the potential to  
109 break down artificial barriers that may exist due to different observing systems, models, time and  
110 space scales, and compartments of the Earth system. In this context, the main future challenges  
111 for research infrastructures have been identified. A value cycle approach has been proposed to  
112 guide innovation in seamless Earth system prediction. The engagement of researchers, users and  
113 stakeholders will be crucial for the successful development of a seamless Earth system science  
114 that meets the needs of society.

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121 **Body Text**

122           Fundamental changes in the environment, an ever-growing global population, especially  
123 in vulnerable regions like coastal zones, and rapid changes in technologies create new challenges  
124 and opportunities. At the same time, natural events with high impact (e.g., resulting from hydro-  
125 meteorological hazards or air pollution) continue to reveal the vulnerability of people and the  
126 infrastructures they rely on. Making society more resilient to the impacts of such events, whose  
127 characteristics may be amplified under a changing climate, requires a coordinated research effort  
128 and new investments to build the observing and prediction systems of the future. To enable all  
129 nations of the world to benefit the scientific and technical knowledge and advancements need to  
130 be made more accessible and usable through international efforts, such as undertaken by the  
131 World Meteorological Organization (WMO).

132           With a focus on establishing the organization's future research agenda, the Commission  
133 for Atmospheric Sciences (CAS) of WMO convened in October 2017 for a Science Summit.  
134 More than 120 scientists (Fig. 1) from 47 countries participated in this conference, which aimed  
135 to garner the scientific community's views, and share knowledge and strategic thinking (see  
136 online supplement for further information about the Science Summit). The presentations, panel  
137 discussions and breakout groups in World Cafes (Fig. 2, and online supplement), focused on  
138 seamless prediction of the Earth system and on how science can serve society. Identifying key  
139 challenges and requirements for future infrastructure, innovation and resources and the  
140 sustainable development of science were on the agenda (Hov et al. 2017). Here we highlight the  
141 key outcomes of the Science Summit and the discussions it sparked, together with the  
142 requirements that are needed to implement successfully the future seamless Earth system science  
143 agenda.

144 **Seamless Prediction and Science for Society**

145           The Earth system is characterized by complex non-linear physical, chemical, and  
146 dynamical processes acting on a vast range of spatial and temporal scales (e.g., Lucarini et al.  
147 2014). The memory of the Earth system components and the associated coupled processes (e.g.,  
148 ocean-atmosphere, land-atmosphere, ocean-ice-atmosphere, atmospheric composition, air  
149 quality) act as seamless sources of predictability. Mitigating and adapting to the impacts of  
150 weather extremes and changing environmental conditions requires detailed information on all  
151 relevant scales, and tailored predictions for a broad variety of user needs. These demands can  
152 only be addressed through a seamless approach to Earth system science that encompasses the  
153 processes acting on the various scales and in all compartments of the Earth system---including  
154 human-induced changes---and their interactions (Sidebar; Shapiro et al. 2010; Nobre et al. 2010).  
155 Advancing Earth system observation, analysis, and prediction capabilities as an international  
156 community, and providing valuable information to the benefit of society was postulated by  
157 Shapiro et al. (2010) as our grand challenge for the future.

158           *A definition of seamless prediction*

159           The original usage of “seamless” (Palmer et al., 2008) referred to predictions across the  
160 range of weather and climate time scales. Since then, the definition has evolved towards the idea  
161 of predicting “the spatial-temporal continuum of the interactions among weather, climate and  
162 Earth system” (Brunet et al., 2010).

163           In 2015, WMO and the World Bank compiled an economic assessment of meteorological  
164 and hydrological services, conceptualizing the connections between the production and delivery  
165 of those services into a value chain (WMO et al. 2015). This value chain links the production and  
166 delivery of these services to user decisions and to the outcomes and values resulting from those

167 decisions. The main components are observation, modeling, forecasting, and services delivery.  
168 This approach strengthens the role of user needs in the development of weather and climate  
169 products. At the same time, however, it does not include feedback and co-design mechanisms  
170 that would put user needs at the heart of the research and development phase. The value cycle  
171 approach (Day 1999) extends the idea of a value chain, originally developed in an economic  
172 context (Porter 1985), by adding interactions with users to the process. Such a value cycle  
173 approach provides a useful means to guide Earth system science and ensure its societal benefit.  
174 The generation and delivery of weather and climate services can be depicted in such a value  
175 cycle (Fig. 3). This encompasses the production (observing, modelling, forecasting) of  
176 information, the dissemination to users (ways of provision, communication and tailor-made  
177 products), perception and decision making, and the outcomes and values. The interaction with the  
178 users is essential for the exploration of “what works” in terms of relevance, quality and impact. The  
179 processes connecting those steps along the cycle and the feedback between them are essential for  
180 its functioning. For instance, it allows to explore how new technologies may help to enhance forecast  
181 products or methods like climate downscaling.

182         Extending the concept of seamless prediction to draw on expertise from social sciences  
183 together with users’ knowledge and experience will help to improve the development of  
184 knowledge and services. Nowadays, we thus expand the initial definition of seamless prediction  
185 to consider also the need of users, stakeholders and decisions makers for information that is  
186 continuous and consistent despite the different sources from which the information is generated.  
187 This seamless prediction approach thus encompasses all compartments of the Earth system,  
188 including human-induced modifications and their consequences, but also all elements of the  
189 value cycle.

190 Seamless Earth system science, guided by the value cycle approach (Fig. 3), will allow us  
191 to understand better and simulate more completely the inherent feedbacks and to generate and  
192 deliver user-specific information on changes in the Earth system, over minutes to centuries in  
193 time, and local to global scales in space. Further, it will enable an assessment of the resulting  
194 benefits to society.

195 The need for such a seamless prediction approach that considers inherent feedbacks is  
196 underpinned by the fact that human activities like water management or various other climate  
197 policies can directly modify the very system that we want to predict. Two examples of why such  
198 interactions need to be considered to allow for the best possible predictions across a wide range  
199 of applications are given below.

- 200 1) Depending on the availability of water resources and their management on sub-  
201 seasonal time scales, stakeholders might decide to mitigate the impact of a heat  
202 wave by modifying urban microclimates through water buffers and green spaces  
203 or irrigating surrounding fields. This in turn may feed back through surface fluxes  
204 on to the local and mesoscale weather patterns (e.g., Grimmond et al. 2010;  
205 Steenbergen et al. 2011; Shepherd 2013; Oke et al. 2017; Chen and Jeong 2018)
- 206 2) On longer time scales, we also have to consider changes in land use, such as  
207 urbanization, deforestation, expansion or reduction of agricultural land, as well as  
208 construction of infrastructure, including photovoltaic- and wind power plants. The  
209 associated change in surface albedo and roughness will locally influence water  
210 and energy surface fluxes of the Earth system and may lead to regional influences  
211 on weather patterns (e.g., Erickson, 1992; Baidya Roy et al. 2004; Pielke 2005).

212 In this framework, accelerating improvements in prediction and services requires  
213 comprehension of the complexity of the technological and human dimensions of the value cycle  
214 together with the interactions, synergies and feedbacks between the various components of the  
215 Earth system. This integrated approach broadens the Earth system science's traditional approach  
216 to include socio-economic themes.

217

### 218 *Meeting the needs of society*

219 Tackling and reducing risks of natural hazards and disasters depends increasingly upon  
220 interdependencies between people, their environment and hazards (Paton and Johnston 2006;  
221 Eiser et al. 2012). For example, Barros et al. (2014) analyzed nearly 4,000 stream gage records in  
222 the eastern and southeastern United States. They reported increases of one order of magnitude in  
223 the specific flood discharge for high-frequency events (e.g. the 2- and 10-year return period) in  
224 counties with large increases in population density between 1990-2010 according to the US  
225 census, and in particular in the Houston area. Using population density as an indirect metric of  
226 urbanization (lifeline infrastructure and increase in paved areas in new developments), and thus  
227 landscape hardening, this implies reduced conveyance and storage capacity in the downstream  
228 network for the same weather event or risk level. Given that a one order of magnitude increase in  
229 the specific flood discharge was found for such high-frequency events in Houston already, then  
230 much worse conditions should be expected for extreme low-frequency events such as Hurricane  
231 Harvey in 2017.

232 Take the general case of a tropical cyclone forecast to make landfall in an urban area.  
233 Based on a probable landfall forecast, authorities have to monitor water storage of dams  
234 surrounding the area and the drainage system status across the city, and reconcile the timeliness

235 of all information sources. Their operational decisions then feed back into the system behavior.  
236 For example, releasing water from a reservoir to prevent dam failure may result in magnifying  
237 the flood threat. To improve the prediction of such events, and thus increase resilience, the  
238 coupled natural- and infrastructure drainage systems and contributing areas need to be  
239 represented in models with a high level of granularity. A continuous monitoring of system  
240 changes in land-use, population density and drainage systems, especially in upstream  
241 contributing areas will allow the representation in models to be updated on a regular basis.

242 Introducing land-use and other anthropogenic effects, allows us to predict the impacts of  
243 extreme weather events more effectively (impact based forecasting). The step forward is to  
244 ensure the timeliness, granularity and flexibility of the information that is required for successful  
245 decision making processes. For instance, traffic management (road, airports, railways, etc.) in an  
246 urban area during and after landfall of a tropical cyclone needs high granularity (i.e. resolution,  
247 level of details) of information, but also flexibility in providing details at required time intervals.

#### 248 *A co-design approach*

249 It is important to ensure flexibility in the development of products and services while also  
250 maintaining standards for quality. Only a co-designed approach that involves all relevant parties  
251 will allow this novel service provision based on seamless Earth system information to work.  
252 Expanded services require more collaboration among disciplines, sectors, and organizations.

253 The energy sector provides examples of where scientific progress improves functionality  
254 and service delivery through a co-design approach. At present, the world is undergoing a global  
255 energy transition with increasing shares of energy derived from renewable energy systems that  
256 are intrinsically weather and climate dependent (REN21 2017; IEA 2017; Siefert and Hagedorn  
257 2017). Ramps in wind- or photovoltaic power production occur due to their weather-dependent



258 capacity. They threaten the security of energy supply if not predicted with the required accuracy.  
259 Power plant- and grid operators must incorporate these energy sources into existing fossil-fuel  
260 dominated power grids and manage their variable weather-dependent outputs based on tailored  
261 predictions. These challenges result in new definitions of high impact weather---such as the  
262 occurrence or non-occurrence of low stratus clouds that strongly affect solar power production---  
263 that must be considered by scientists and forecast providers.

264 A secure and economic integration of renewable energy sources thus relies on accurate  
265 forecasts of the potential power production, and these in turn on improved weather forecasts,  
266 including an estimate of forecast uncertainty. The energy sector requires data for multiple  
267 timescales to respond to current user needs. Further, it uses data for infrastructure planning and  
268 for responding to future energy demands. The value cycle approach could help facilitate the  
269 integration of user's needs into the science planning, thus becoming a concrete tool for co-  
270 design.

271

## 272 **Future Infrastructure**

273 Earth system sciences are extremely data and compute intensive. They are increasingly a  
274 big data problem, involving a huge number of different kinds of observations and diverse  
275 modeling and data processing outputs. A new machine learning frontier is bridging between  
276 outputs and sector-specific services. Turning these opportunities and challenges into a benefit for  
277 society requires a paradigm shift in scientific methodologies and a strengthening of collaboration  
278 across different sectors. Science that serves society requires planning to ensure that resources---  
279 financial, technical, physical, organizational and human---can meet future requirements.

280 *Earth system computing and machine learning*

281 Advances in numerical weather prediction since the 1950s and in climate predictions and  
282 projections more recently have gone hand in hand with progress in scientific computing and  
283 observational capabilities. Meeting societal needs requires simulating finer scales with more  
284 complex physical processes, assimilating more data, coupling models for the different  
285 compartments of the Earth system and running large ensembles to produce more accurate and  
286 reliable forecasts, while also providing information about their uncertainty. This has resulted in  
287 research and operational centers using some of the largest high performance computing (HPC)  
288 systems worldwide. The steady increase of skill obtained with more complex forecasting systems  
289 run on increasingly larger HPC facilities and the availability of new diverse and extended  
290 observational datasets for data assimilation (e.g., from modern satellite systems), has led to what  
291 is known as a ‘quiet revolution’ in numerical weather prediction (Bauer et al. 2015).

292 Moving to high-resolution, complex and probabilistic Earth system analysis and  
293 forecasting systems will, however, require substantially more computing and data handling  
294 resources. Contrary to the reliance on the steady micro-processor performance development in  
295 the past, these need to be provided by a concerted effort between mathematical, algorithmic and  
296 programming environment developments, taking also into account affordable electric power  
297 levels. Further, the developments should focus on more heterogeneous, specialized hardware  
298 options (Lawrence et al. 2018), like different kinds of processors, and explore artificial  
299 intelligence methods where applicable (Dueben and Bauer 2018). These challenges receive  
300 worldwide attention currently and spawn significant funding programs, for example through the  
301 Future and Emerging Technology High-Performance Computing program of the European  
302 Commission, the Department of Energy Exascale Earth System Model effort in the US, and  
303 comparable large-scale science-technology programs in Japan and China.

304           Substantial advances have been made in the assimilation of traditional and new types of  
305 data into models, using them for the development of verification methodologies, as well as for  
306 the generation of nowcasting- and other prediction products. New developments in data  
307 assimilation, like ultra-rapid data assimilation algorithms, allow the gap between forecasts from  
308 nowcasting and numerical weather prediction to be closed and can form the base for seamless  
309 prediction from minutes to hours.

310           Machine learning and big data techniques provide new possibilities to complement and  
311 expand on our seamless prediction system, in particular for very short-time decision making  
312 problems (time scale of minutes). Information (e.g., about road conditions) can be shared  
313 instantaneously and processed by smart networks (e.g., interconnected cars), issuing an  
314 automatic warning to the full network.

315           The emerging wealth of data further provides the chance to add inductive, data-driven  
316 science to theory-driven, deductive science (Hey et al. 2009). Additional opportunities arise for  
317 multidisciplinary research that can enrich service provision using seamless Earth system  
318 information by providing visual analytics and appropriate storylines. Storylines can help to make  
319 information about possible developments of the Earth system and their impact more  
320 comprehensible to users (Hazeleger et al. 2015). In such a storyline approach, numerical models  
321 can e.g. be used to create a set of physically plausible realizations of an extreme weather event in  
322 an altered climate and the possible impacts. Instead of probability information, which often  
323 suffers from uncertainties in model simulations, this event-oriented approach provides a set of  
324 possible development scenarios, which might be more accessible to some users (Shepherd et al.  
325 2018). Tapping into the potential of these technological opportunities to further enhance our  
326 seamless Earth system prediction capabilities needs supporting scientific virtual or physical

327 infrastructures that facilitate their exploitation (e.g., regional research center network, monitoring  
328 capacity in least developed countries and small island developing states, computing facilities and  
329 high-speed connectivity). Capitalizing the full benefit of these new types of data, methods and  
330 systems for our prediction approaches remains an ongoing challenge, however, requiring  
331 continuous investments in research, infrastructure and human resources.

332 *The data management issue*

333 The increasing volume of data, both from observations and models, may make data  
334 handling and transfer computationally unaffordable or even infeasible and thus result in  
335 immobility of data. New data-management models are required, such as moving from existing  
336 centralized data storage, processing- and analysis systems to more distributed systems or cloud-  
337 based solutions. Big data approaches must be applied to large spatio-temporal data such as  
338 gridded forecasts, satellite imagery, and large volumes of non-conventional observations of  
339 weather or the environment (Lu et al. 2016). Future distributed infrastructure should build on  
340 modular components of formats, methods, and systems, including the full chain from observation  
341 operators, retrieval and nowcasting algorithms, data assimilation components, numerical models,  
342 monitoring and alert systems, exchange formats, verification, diagnostics, quality control, to  
343 intercomparison tools. In this case, maintenance of observational systems and data management  
344 are critical elements to ensure the sustainable development of knowledge and science for society.

345 *Accessibility of observations*

346 The availability and accessibility of observations is key to skillful predictions and  
347 indispensable for developing, maintaining and further enhancing the skills of our prediction  
348 systems. That is why a long tradition of standardizing and sharing data, starting from the late  
349 19th century has developed in the meteorological community (Pudykiewicz and Brunet 2008). It

350 is recognized that weather forecasting is a shared, global challenge that must be addressed  
351 collectively. Under the auspices of WMO, the worldwide access and exchange of observational  
352 data from national networks and the fleet of national and international satellites has therefore  
353 been organized in an efficient way.

354 In this context, the highest priority should be given to ensuring data availability and the  
355 best possible exchange of information. All relevant observations must be available to improve  
356 nowcasting, for assimilation into multi-scale models, and to ensure long-term monitoring of  
357 essential climate variables. Earth system data---because of their essential role for the security of  
358 society and environmental disaster prevention---thus need to be “Findable, Accessible,  
359 Interoperable, and Re-usable (FAIR)” (Wilkinson et al. 2016).

360 *Improved collaboration in the Global Weather Enterprise*

361 As new players are entering the field, the infrastructure and management culture of this  
362 data exchange need to be modernized. These new players, including commercial data service  
363 providers, are generating and providing observations of our environment, or creating their own  
364 prediction systems. One example can be found in the development and management of sensor  
365 systems, where a transition is underway from solely sparse public sector data sources using high  
366 cost, but well-characterized and standardized equipment and mobile monitoring, to the use of  
367 blended data that includes lower cost sensors deployed by public and private actors. More non-  
368 conventional data will become available, for example, from mobile phones, cars, and other  
369 internet-connected devices, most of which will be owned by private companies or individuals.  
370 This results in increasing volumes of data in the public domain of varying quality, provenance  
371 and reliability, supplied by a much wider range of sources. On the one hand, this opens up the  
372 opportunity for advancing prediction systems and for the production of improved user-tailored

373 products. On the other hand, this requires a policy on data usage and sharing, e.g. following the  
374 FAIR concept mentioned above, and the means to ensure interoperability of systems and  
375 methods with data from other science disciplines or sectors.

376           Collaboration among the private and public sectors and partnerships in the context of the  
377 Global Weather Enterprise (Thorpe and Rogers 2018) are vital to ensure that as much of this data  
378 as possible are available to as many people as possible, including full accessibility for research  
379 purposes. At the same time, these data and technologies must be used in ways that ensure  
380 decisions are made based on information that is of the right quality for the task at hand (Lewis  
381 and Edwards 2016). An open question is how the growth of private sector capabilities can  
382 strengthen and not weaken the overall investment on the value cycle, and the continuous  
383 improvement and availability of Earth system information. Companies with a weather-oriented  
384 business recognize that this capability has to be built on the public investment in the global  
385 observing system, in models and tools that form the bedrock of their operations, and in long-term  
386 atmospheric research (Thorpe and Rogers 2018). The development of public-private partnerships  
387 further necessitates a clear definition of the roles of the different players in providing  
388 information. This applies in particular to warnings and other information that are highly critical  
389 for the society. Such a policy and mutual agreement among all players involved could be crucial  
390 to prevent unplanned breakdowns in the provision of essential Earth system data for the benefit  
391 of society. Thus, WMO is promoting a dialogue among different players to ensure a coordinated  
392 growth of the Global Weather Enterprise.

393

394

395 **Nurturing Scientific Talents**

396           An innovative and diverse workforce is needed to advance seamless Earth system  
397 science. Developing science guided by the value cycle requires an interdisciplinary approach and  
398 mind-set alongside the capability to work in depth in individual disciplines. The technical side of  
399 the value cycle requires expertise in aspects of data handling and understanding emerging  
400 technologies, in computational sciences, in managing and improving infrastructure, developing  
401 and running coupled prediction models and various other components. Developing products for  
402 end-users, improving the information provided for decision making, and considering aspects of  
403 vulnerability and risk in predictions requires a consideration of risk communications, behavioral  
404 sciences, and economic aspects. Early exposure to and training in an interdisciplinary scientific  
405 approach is essential in building the links between natural and social sciences.

406           There are various challenges that the new generation of scientists encounter during their  
407 career, which may vary for different regions in the world. Access to data, tools and infrastructure  
408 and scientific publications, as well as the possibility to attend conferences and workshops and  
409 thus to interact with the international research community are examples. These may apply in  
410 particular to scientists in developing countries (Dike et al. 2018), but also to scientists at under-  
411 resourced universities and research institutions. The development of cloud-based solutions to  
412 provide access to data and tools could be a means to foster research worldwide. Together with  
413 improvements in information technology and research infrastructure within developing countries  
414 and an investment plan for highly-qualified human resources, these new solutions might help to  
415 prevent researchers from moving to other countries because they expect a better support for their  
416 research (Polcher et al. 2011).

417           The development and retention of scientific talents could benefit from the development of  
418 scientific educational hubs, both virtual and real. There is a need to connect people in academia,

419 government, and the private sector, facilitating the improvement of both local and global  
420 research collaborations and providing an open forum for broader participation. Three examples  
421 from Africa for such an approach are given here. The “Science for Weather Information and  
422 Forecasting Technology (SWIFT)<sup>1</sup>” project is jointly funded by research and development funds,  
423 through the UK’s Global Challenges Research Fund Africa. SWIFT aims to enhance the weather  
424 prediction capability from hourly to seasonal timescales in four African countries. The project  
425 connects universities and forecasters from the UK with those in Senegal, Ghana, Nigeria and  
426 Kenya to maintain and further increase local research capacities. It works with forecast users  
427 from various sectors toward tailored provision of weather forecasts and improved response to  
428 high-impact weather events. The recently established East African Institute for Fundamental  
429 Research (EAIFR)<sup>2</sup> in Rwanda, a partner of the International Centre for Theoretical Physics,  
430 addresses the need in Rwanda and the region for MScs and PhDs in various areas of physics,  
431 both fundamental and applied. The African Institute for Mathematical Sciences, a pan-African  
432 network of centers of excellence, offers a structured Master’s in mathematical sciences and  
433 focuses on scientific training, cutting-edge research and public engagement. One important  
434 component of any of these actions is ensuring fluent and sustained connections amongst  
435 scientists from developed and less developed countries and from different sectors. This would  
436 benefit science as a whole.

437         The gender disparity in the scientific, technological, engineering and mathematics  
438 (STEM) disciplines is another factor that may limit access to the full potential of an emerging  
439 generation of scientific talents. Although not fully understood yet, the gender disparity may be  
440 attributed to conscious and unconscious gender biases, education systems and society, challenges

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<sup>1</sup> More information on SWIFT is available at <https://africanswift.org/>

<sup>2</sup> More information on EAIFR is provided at <https://eaifr.ictp.it/>



441 in work-life balance, lack of long-term career opportunities in academia, attitudes about career  
442 choice, and a lack of role models. Aspects of harassment, marginalization and isolation might  
443 further add to the gender disparity. An in-depth analysis of the factors resulting in gender  
444 disparity is beyond the scope of this paper. UNESCO (2017) provides a more detailed  
445 assessment, together with examples from research and practice on how encourage women and  
446 girls to pursue a career in STEM.

447 Working to break down the barriers described above and to create opportunities that  
448 foster future scientific talents with an interdisciplinary education are key to supporting an  
449 emerging generation of scientists. As future leaders in the field it is in their hands to progress the  
450 work toward a seamless Earth system science that will benefit governments, institutions and  
451 society.

452

### 453 **Innovation and Resources**

454 Our environment has a widespread and important impact on many industries,  
455 including energy, transportation, public health, and agriculture. Organizations in all these  
456 industries are using Earth system data to inform their operations, planning and decisions. In a  
457 science driven inquiry framed by real-world problems, there is a growing interaction between  
458 science and applications. The proposed seamless Earth system science will advance scientific  
459 knowledge of the system itself, improve predictive capabilities, and foster policy oriented  
460 research. It will also enable the provision of products and services at all timescales and to all  
461 sectors and applications, and will hence facilitate the transition to a seamless provision of Earth  
462 system information.

463           The way we organize science and its connection with stakeholders need to change if we  
464 are to develop a more flexible system tailored for answering emerging and urgent societal  
465 requirements, expressed in the Paris Agreement of 2015 (UNFCCC 2015), the 2030 Agenda for  
466 Sustainable Development (UN 2016), or the Sendai Framework for Disaster Risk Reduction  
467 (UNISDR 2016). Research requires a balanced approach, combining long-term activities that  
468 will support continuous improvement alongside short-term innovation for targeted challenges.  
469 Both are needed to progress towards the longer-term goal of seamless Earth system prediction.  
470 The implementation of a feedback loop along the value cycle (sidebar) and across the interfaces  
471 will help to ensure a continuous interaction between users, operations and science. As an  
472 example, in the satellite sector, scientists who are designing the satellite observing system, those  
473 who are developing products, and the user community work together to determine how satellite  
474 data can better inform decision making (Brown and Escobar 2019). In this value cycle  
475 framework innovation can be promoted by focusing research activities, improving access to  
476 interdisciplinary datasets and tools for application development, and mobilizing resources around  
477 key societal needs.

478

## 479 **Recommendations**

480           From the discussions during the Science Summit and beyond, a number of  
481 recommendations emerged as cornerstones to shape the WMO research agenda for the years to  
482 come.

- 483           • A better integration between the needs of stakeholders, decision-makers and other  
484           users, and the implementation of seamless Earth system prediction must be facilitated.

485           Science has to work together with users to explore ways of integrating data from

486 different observing systems, models and other prediction products, as well as from the  
487 different compartments of the Earth system to enable the provision of information that  
488 is accurate, smooth and consistent across time and space scales.

489 • A mechanism must be developed for a rolling review of user requirements that will  
490 help shape priorities in Earth system science and involve user groups through  
491 effective feedback mechanisms, inter-dependencies, and mutual trust. To ensure that  
492 our developments meet the increasing demands of users and society for more  
493 sophisticated, integrated services, this mechanism must be based on a continuous  
494 exchange of information between the science and user communities. This is the  
495 prerequisite for co-designing the development of new and user-oriented services. The  
496 implementation of a value cycle, with well-defined connections at the interfaces along  
497 the cycle, is seen as a promising approach to realize the concept of co-design.

498 • The focus on emerging technologies and methodologies, like new observing  
499 platforms, lower-cost sensors, artificial intelligence, “extreme” (Exabyte and further)  
500 data management and supercomputing must be strengthened. The increasing  
501 availability of a vast amount of data opens up new opportunities for improving  
502 predictions and services. At the same time, new challenges emerge when it comes to  
503 the diversity of data sources, to aspects of data handling or to recent developments in  
504 supercomputing. Fruitful collaborations between computing experts and industry  
505 could thus help explore new ways of creating seamless Earth system predictions. As a  
506 pioneering endeavor, the international ExtremeEarth initiative  
507 (<http://www.extremearth.eu/>) aims at bringing together academia, private companies  
508 and operational centers to drive future developments in large-scale computing and

509 data intensive methodologies. Together with these emerging technological  
510 opportunities comes the need to implement strategies to ensure that the information  
511 provided is of the right quality and content to allow well-informed decision making.

512 • New policies on data management and use must be developed, taking into account the  
513 growing field of Earth system information providers. The different contributors to the  
514 Global Weather Enterprise from the public, private and academic sectors need to co-  
515 operate even more fully than in the past if the seamless approach to Earth system  
516 prediction is to become reality.

517 • The education of scientists, particularly in developing regions, must be fostered, in  
518 order to exploit the full potential of the seamless Earth system prediction worldwide.  
519 Building academic training around the concept of the value cycle presented in this  
520 paper would be a first priority for better linking the academic community to WMO  
521 operational activities. The emerging opportunities of online communication tools to  
522 broaden access to training and information sharing and the establishment of  
523 educational hubs should improve accessibility to scientific resources and bring the  
524 global research community closer together.

525 • An international coordinating mechanism must be established that ensures the  
526 development of basic and applied themes of the seamless Earth system prediction,  
527 combined with a strong link to the different regional needs. Regional dedicated  
528 networks for interconnecting academic and operational institutions are needed for the  
529 most vulnerable regions. These networks should elaborate the relevant scientific  
530 questions that need to be addressed to make regions more resilient to environmental  
531 extremes, and international bodies and organizations (e.g., WMO, the International

532 Science Council, Intergovernmental Oceanographic Commission of UNESCO, the  
533 Belmont Forum, and FutureEarth) should facilitate this together.

534 These recommendations will guide the research and operation dialogue at 2019 WMO Congress,  
535 ensuring effective connections of Earth system science with societal needs, paving the way to the  
536 development of seamless Earth system prediction capabilities. Engaging researchers, users and  
537 stakeholders in advancing seamless Earth system science will be crucial to ensure that the  
538 delivery of information about the changing environment addresses the needs of society.

539

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670 **Sidebar1: The value cycle approach to seamless Earth system science and the Global**  
671 **Weather Enterprise**

672

673 Earth system:

674 Following Shapiro et al. (2010), the Earth system encompasses the atmosphere and its chemical  
675 composition, the oceans, land/sea ice and other cryosphere components, as well as the land  
676 surface, including surface hydrology and wetlands, lakes, and human activities. On short time  
677 scales, it includes phenomena that result from the interaction between one or more components,  
678 such as ocean waves and storm surges. On longer time scales (e.g., climate), the terrestrial and  
679 ocean ecosystems, including the carbon and nitrogen cycles and slowly varying cryosphere  
680 components (e.g., the large continental ice sheets and permafrost), are also part of the Earth  
681 system.

682

683

684 Global Weather Enterprise:

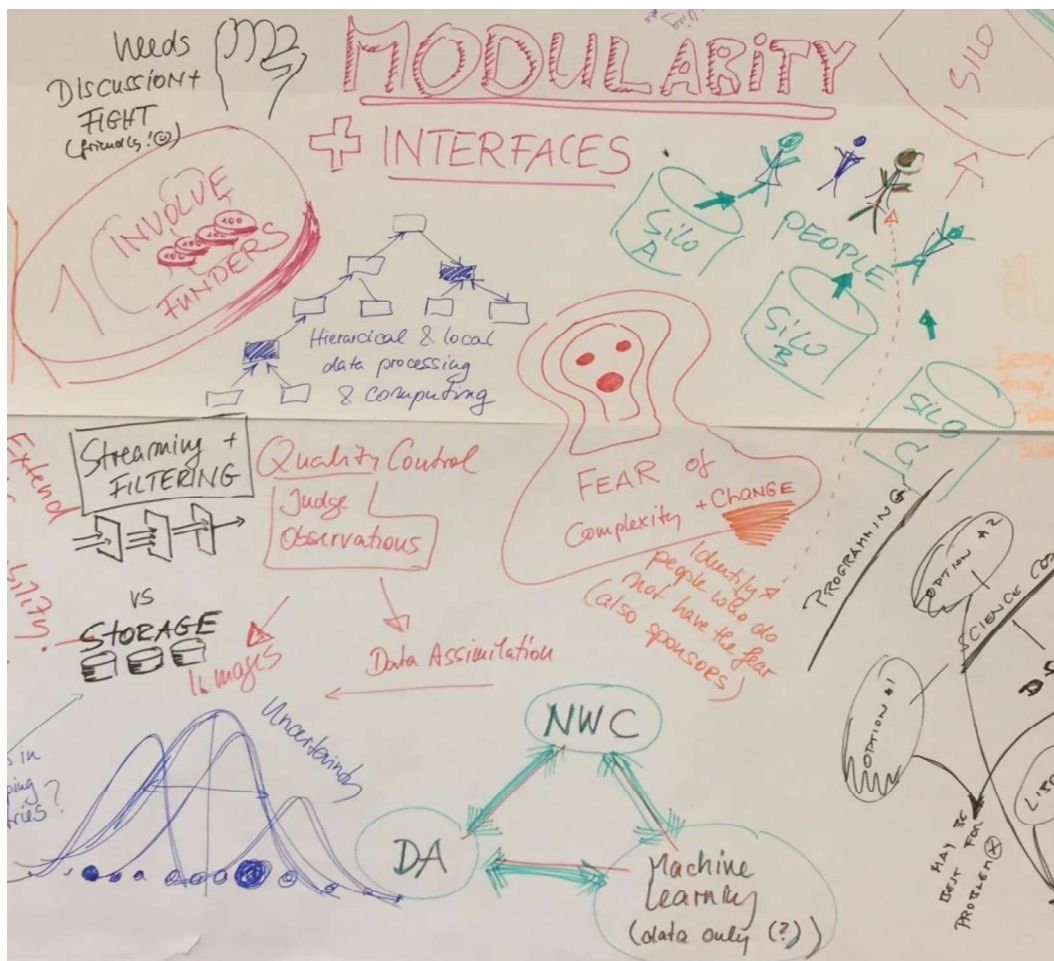
685 Following Thorpe and Rogers (2018): “The term Global Weather Enterprise (GWE) has been  
686 coined to describe the totality of activities by individuals and organizations to enable weather  
687 information to be created and provided to society... The enterprise includes the full value chain  
688 of scientific research, observations of the Earth system, numerical models encoding the laws of  
689 physics applied to the system, supercomputing to integrate the models and observations, weather  
690 and hydrological forecasts from hours to weeks and potentially months ahead, and business-  
691 specific products and services enabling economic benefit and jobs to be created. The health of  
692 the whole enterprise strongly depends on the strength of each component.”



693

694 Figure 1: Participants of the Science Summit, held from 20 – 22 October 2017 at the World  
695 Meteorological Organization’s headquarters in Geneva, Switzerland. A full list of participants is  
696 provided in the online supplement.

697



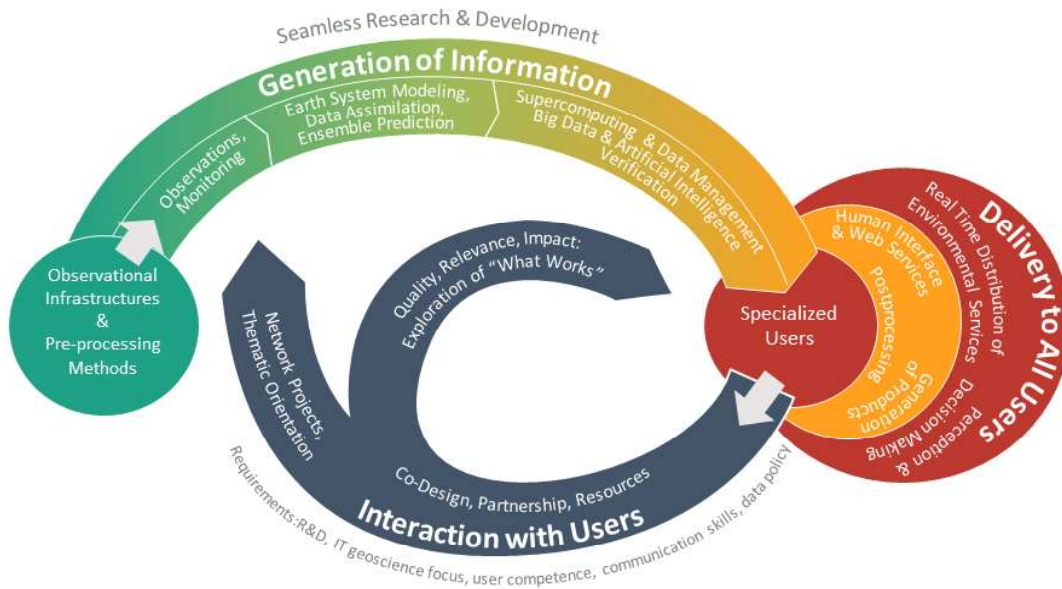
698

699 Figure 2: An impression of the discussion in the World Cafe. This setup allowed all members of

700 the Science Summit to participate and express their views and vision, both verbally and by

701 drawing on the table cloths.

## SCIENCE FOR SERVICES JOURNEY



702 Figure 3: Technical developments on seamless Earth system science need to go hand in hand  
 703 with informed advancement of observations, monitoring capabilities and advanced assimilation  
 704 and Earth system modelling and other prediction methods, which are the backbone of existing  
 705 meteorological services. This sketch of the value cycle identifies the fundamental bricks of our  
 706 system and details the interfaces along the value cycle. It encompasses the generation of  
 707 information (observations and their infrastructure, modelling, forecasting; green to yellow),  
 708 postprocessing, the generation of products and suitable interfaces (yellow), as well as the  
 709 dissemination to users (red) and the perception and decision making. The interaction with users  
 710 (gray), e.g. through co-design of projects, is essential for the exploration of user-oriented  
 711 services.