

# The sub-seasonal to seasonal prediction (S2S) project database

Article

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# 1 The Sub-seasonal to Seasonal Prediction (S2S) Project Database

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29	
30	Summary: A database containing sub-seasonal to seasonal forecasts from 11 operational
31	centres is available to the research community and will help advance our understanding of
32	the sub-seasonal to seasonal time range.
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- 51 Abstract
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53 Demands are growing rapidly in the operational prediction and applications communities for 54 forecasts that fill the gap between medium-range weather and long-range or seasonal 55 forecasts. Based on the potential for improved forecast skill at the sub-seasonal to seasonal 56 time range, a sub-seasonal prediction (S2S) research project has been established by the 57 World Weather Research Program/World Climate Research Program. A main deliverable of 58 this project is the establishment of an extensive database, containing sub-seasonal (up to 60 59 days) forecasts, 3-weeks behind real-time, and reforecasts from 11 operational centers, 60 modelled in part on the THORPEX Interactive Grand Global Ensemble (TIGGE) database for 61 medium range forecasts (up to 15 days).

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63 The S2S database, available to the research community since May 2015, represents an 64 important tool to advance our understanding of the sub-seasonal to seasonal time range that 65 has been considered for a long time as a "desert of predictability". In particular, this database 66 will help identify common successes and shortcomings in the model simulation and 67 prediction of sources of sub-seasonal to seasonal predictability. For instance, a preliminary 68 study suggests that the S2S models underestimate significantly the amplitude of the Madden 69 Julian Oscillation (MJO) teleconnections over the Euro-Atlantic sector. The S2S database 70 represents also an important tool for case studies of extreme events. For instance, a multi-71 model combination of S2S models displays higher probability of a landfall over Vanuatu 72 islands 2 to 3 weeks before tropical cyclone Pam devastated the islands in March 2015.

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#### 1) Sub-seasonal to seasonal prediction

75 Demands are growing rapidly in the operational prediction and applications communities for 76 forecasts that fill the gap between medium-range weather (up to 15 days) and long-range or 77 seasonal (3-6 months) forecasts. Skillful sub-seasonal to seasonal prediction (forecast range 78 more than 2 weeks but less than a season) provides an important opportunity to inform 79 decision makers of, for example, changes in risks of extreme events or opportunities for 80 optimizing resource management decisions. Although many challenges remain to make subseasonal forecasts sufficiently reliable, skillful and tailored for users, a great return on 81 82 investment in weather and climate science and model development is to be expected if the 83 science and forecast products of sub-seasonal to seasonal prediction can be successfully 84 connected to societal applications.

85 Weather-related hazards, including slow onset of long-lasting events such as drought and 86 extended periods of extreme cold or heat, trigger and account for a large proportion of 87 disaster losses, even during years with other very large geophysical events (e.g., Haitian and 88 Chilean earthquakes) Munich Re: (source 89 http://www.iii.org/sites/default/files/docs/pdf/munichre-010715.pdf). While many end-users 90 have benefited by applying weather and climate forecasts in their decision-making, there 91 remains ample evidence to suggest that such information is underutilized across a wide 92 range of economic sectors (e.g., Morss et al., 2008; Rayner et al., 2005; O'Connor et al., 2005; 93 Pielke and Carbone, 2002; Hansen, 2002). This may be explained in part by the presence of 94 'gaps' in our forecasting capabilities at the sub-seasonal time scale and in part by the 95 complexity of processes and the numerous facets involved in decision making. Developing

96 countries are most affected by major gaps in access to forecasts and knowledge. The goal of
97 the Sub-seasonal to Seasonal Prediction (S2S) Project and its associated database is to help
98 fill these gaps.

99

#### 100 **2)** The S2S Project

101 Sub-seasonal forecasting, bridging a gap between the more mature weather and climate 102 prediction communities, is at a relatively early stage of development. Forecasting the day-to-103 day weather is often considered as an atmospheric initial condition problem. Most of the 104 current operational medium-range forecasting systems (forecasts up to day 15) are not 105 coupled to an ocean model, although there can be an influence from ocean (e.g. Bender and 106 Ginnis 2000) and land conditions (e.g. Koster et al, 2010). Forecasting at the multi-season to 107 multi-annual range depends strongly on the slowly-evolving components of the earth system 108 such as the sea surface temperature. In between these two time scales is sub-seasonal to 109 seasonal variability (defined here as the time range between 2 weeks and 2 months). 110 Forecasting for this time range has so far received much less attention than medium-range 111 and multi-season prediction despite the considerable socio-economic value that could be 112 derived from such forecasts. This timescale is critical for proactive disaster mitigation efforts. 113 It is considered a difficult time range since the lead time is sufficiently long that much of the 114 memory of the atmospheric initial conditions is lost and it is too short for the variability of the 115 ocean to have a strong influence. However, recent research has indicated important 116 potential sources of predictability for this time range such as the MJO, the state of ENSO, soil 117 moisture, snow cover and sea ice, stratosphere-troposphere interactions, ocean conditions 118 and tropical-extratropical teleconnections (see for example review in Vitart et al., 2015).

119 The fundamental goals of the sub-seasonal to seasonal prediction (S2S) research project are 120 to improve forecast skill and understanding on the sub-seasonal to seasonal timescales, and 121 to promote its uptake by operational centers and by the applications community (Vitart et al, 122 An extensive database containing sub-seasonal (up to 60 days) forecasts and 2012). 123 reforecasts (sometimes known as hindcasts) has been created to enable research to 124 operational pathways to accomplish these goals. It is modelled in part on the THORPEX 125 Interactive Grand Global Ensemble (TIGGE) database for medium range forecasts (up to 15 126 days) (Bougeault et al, 2010) and the Climate-System Historical Forecast project (CHFP) 127 (http://wcrp-climate.org/index.php/wgsip-chfp/chfp-overview) for seasonal forecasts. The 128 research is organized around a set of six topics (Madden-Julian Oscillation, Monsoons, Africa, 129 Extremes, Teleconnections and Verification), each intersected by the cross-cutting research 130 and modeling issues, and applications and user needs. The latest science plans of each sub-131 project are available online (<u>http://www.s2sprediction.net/documents/reports</u>). Some of the 132 main research questions include:

- What is the benefit of a multi-model forecast for sub-seasonal to seasonal
   prediction and how can it be constructed and implemented?
- What is the predictability of extreme events and how can we identify windows of
   opportunity for sub-seasonal to seasonal prediction?
- 137
- What is the best initialization strategy for a forecasting system that includes
   ocean, land and cryosphere? What is the optimal way to generate an ensemble of
   sub-seasonal to seasonal forecasts?

141	•	What is the impact of horizontal and vertical resolution of atmosphere and ocean				
142		models on sub-seasonal to seasonal forecasts?				
143	•	What are the origins of the systematic errors affecting sub-seasonal to seasonal				
144		forecasts?				
145	•	How well do state-of-the-art models represent tropical-extratropical				
146		teleconnections?				
147	•	What forecast quality attributes are important when verifying S2S forecasts and				
148		how should they be assessed?				
149	•	What are current S2S forecasting capabilities for daily weather characteristics				
150		relevant to agriculture, water resource management and public health, such as				
151		heavy rainfall events, dry spells and monsoon onset/cessation dates?				
152	•	How well do we understand the fundamentals of predictability and dynamical				
153		processes of the sub-seasonal variability?				
154						

155 **3)** Description of the S2S database

156 The S2S database builds on the experience of creating the TIGGE database and can be seen 157 as its extension to the longer forecasts ranges. The S2S database includes near real-time ensemble forecasts and reforecasts up to 60 days from 11 centers: Australian Bureau of 158 159 Meteorology (BoM), China Meteorological Administration (CMA), European Centre for 160 Medium-Range Weather Forecasts (ECMWF), Environment and Climate Change Canada 161 (ECCC), the Institute of Atmospheric Sciences and Climate (CNR-ISAC), Hydrometeorological Centre of Russia (HMCR), Japan Meteorological Agency (JMA), Korea Meteorological 162 163 Administration (KMA), Météo-France/Centre National de Recherche Meteorologiques 164 (CNRM), National Centers for Environmental Prediction (NCEP) and the United Kingdom's 165 Met Office (UKMO). A key difference with the TIGGE database, is that the S2S database 166 includes reforecasts, whereas none are included in the TIGGE database. For short-range 167 weather forecasts, model error is not usually so dominant that a reforecast set is needed, but 168 for the sub-seasonal to seasonal range model error is too large to be ignored. Therefore an 169 extensive reforecast set spanning several years is needed to calculate model bias. Such 170 reforecasts in some cases can also be used to evaluate skill. The models are also generally 171 different from the TIGGE models. For instance, S2S models can have the atmospheric 172 component coupled to an ocean model and an active sea ice model (Table 1).

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174 Because S2S is a research project, the real-time forecasts are only available with a 3-week 175 delay. Table 1 displays the main characteristics of the S2S models. Tables 2, 3, 4, 5 and 6 176 show the list of variables which have been requested for the S2S archive, which include 177 standard variables at many pressure levels, together with a large number of single-level 178 variables including thermodynamic, hydrological, and surface flux fields. However, some 179 models are providing just a subset of the requested variables. The list of variables provided 180 model be found by each here: can 181 https://software.ecmwf.int/wiki/display/S2S/Provided+parameters. Pressure level fields are 182 available in the stratosphere at 50 and 10 hPa to facilitate the diagnostic of sudden 183 stratospheric warming events and their downward propagation. The frequency of archiving is 184 once a day except for maximum and minimum near surface temperature and total 185 precipitation which are available 4 times a day (computed over 6-hour periods). The data is archived in GRIB2 format, and a conversion to NetCDF will be made available. There are plans 186 187 to add some oceanic variables in the near future, from the coupled ocean-atmosphere

188	models: sea surface salinity, depth of the 20 degree isotherm, heat content in the top 300 m,
189	salinity in top 30 meters, U and V surface current and sea surface height. It is also planned to
190	include sea-ice thickness for the models which have a dynamical sea-ice model.
191	
192	The S2S database is a database of "opportunity", which means that the forecasts have not
193	been produced specifically for the S2S project following an agreed protocol. Table 1
194	highlights differences in model setup between the operational centers. The main differences
195	between real-time forecasts from different centers include:
196	
197	• The forecast time range varies from 32 to 60 days
198	• The horizontal resolution of the atmospheric model varies from a few hundreds
199	kilometers resolution to about 30 kilometers.
200	• The ensemble size varies from 4 to 51 members. This reflects a different of
201	strategy between operational centers. The centers producing a low number of
202	ensemble members typically produce forecasts in lag mode (combining ensemble
203	members from different start dates to produce an ensemble forecast).
204	• The frequency of initializing forecasts varies. Some models are run in burst mode
205	on a sub-weekly basis with a large ensemble size (e.g. ECMWF, BoM, ECCC), whereas
206	other models are run in continuous mode on a daily basis with a smaller ensemble
207	size (e.g. NCEP, UKMO, CMA, KMA). Other models (e.g. CNRM) are run on a monthly
208	basis.
209	• Some models have an atmosphere component coupled to an ocean and a sea ice

model (e.g. UKMO, NCEP, CNRM, CMA) while other use a combination of persistence

211	of initial conditions and climatology to define the oceanic and sea ice boundary
212	conditions (e.g. JMA, ECCC).
213	
214	The configuration of the reforecasts also varies greatly between the models:
215	
216	• Some models have a re-forecast set covering a period exceeding 30 years (e.g.
217	JMA, BoM), while other re-forecast sets span a much shorter number of years
218	(e.g. NCEP, UKMO)
219	• Some reforecasts are produced progressively "on the fly" (as at ECMWF), while
220	others are computed all at once prior to operational implementation (e.g., BoM,
221	NCEP).
222	• The ensemble size can vary from just 1 member (e.g. CNR-ISAC) to 33 members
223	(BoM).
224	• Some models have reforecasts produced on a daily basis (e.g. NCEP) while others
225	have reforecasts on a sub-weekly basis (e.g., BoM, ECMWF) and others have
226	reforecasts on a monthly basis (e.g CNRM).
227	
228	There is much greater diversity between the various S2S forecast systems than in other
229	databases for medium and seasonal time ranges (e.g. TIGGE, EUROSIP, CHFP). Very different
230	strategies are currently in use. For example, some centers take advantage of their seasonal

and climate systems, while other centers employ systems used for weather forecasting. This

highlights the current lack of consensus on the best practice for sub-seasonal prediction

unlike for medium-range and seasonal forecasting and diversity of priorities of operational

centers. One of the goals of the S2S project is to make recommendations on the optimal

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configuration of sub-seasonal systems. The S2S database will enable these issues to be addressed by clustering the models sharing similar characteristics (e.g. coupled oceanatmosphere models vs atmosphere-only models; lag vs burst initialization...) and comparing their forecast skill scores.

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Despite the differences in system set-up, there are enough commonalities between them to make inter-comparisons or multi-model combinations possible, as will be shown in Section 3. For instance, almost all of the S2S systems produce real-time ensemble forecasts every Thursday, and have reforecasts covering the period 1999-2010. Therefore, it is possible to create a multi-model combination of the S2S models every Thursday, calibrated using the common period 1999-2010.

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The database is currently updated routinely with near real-time forecasts and reforecasts from nine data providers, namely, JMA, NCEP, BoM, ECMWF, UKMO, CMA, CNRM, CNR-ISAC and HMCR. Data from ECCC and KMA will be available soon. The S2S database is hosted by two archiving centers, ECMWF and CMA, and was opened to the public on 6 May 2015 at ECMWF via the Data Portal and ECMWF Web API (Application Programming Interface) and in November 2015 at CMA. Users can register, visit the data portal and browse the contents of the database, and are encouraged to use the ECMWF Web API to download data in batch.

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255 By the end of 2015, about 300 users from 42 countries had registered and had already 256 executed over 200,000 requests to extract about 30 Terabytes of data from ECMWF. ECMWF and CMA are working together closely to ensure the timely synchronization of the two 257 258 databases. The S2S database at ECMWF be accessed at can

<u>http://apps.ecmwf.int/datasets/data/s2s\_and\_http://apps.ecmwf.int/datasets/data/s2s-</u>
 <u>reforecasts\_for\_the\_reforecasts.</u> The S2S\_database at CMA can be accessed at
 <u>http://s2s.cma.cn/.</u>

At CMA, about 22 Terabytes of forecast and re-forecast data have been collected from ECMWF. S2S data is archived on tapes into the MARS system (same archiving system as at ECMWF) and also stored into a large online storage system with a preprocessed unified form. The CMA data portal, as the ECMWF data portal, provides descriptions of the models from the different centers and S2S data parameters, in addition to the data download service. Two ways of searching and accessing the data are supported: free text search and faceted search. The method of downloading data is similar to the e-commerce "shopping-cart" through a "Data cart". All the S2S data can be accessed by HTTP currently and OPeNDAP in the near future. The S2S data in GRIB2 format can be directly downloaded at CMA, and data in NetCDF format obtained through online conversion.

**4)** Examples of use of the S2S database

#### 4.1 Multi-model prediction

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283 In order to monitor the S2S forecasts, a basic set of products has been developed, including 284 ensemble mean anomalies for few meteorological parameters and some atmospheric 285 indices. These products are generated routinely at ECMWF from each individual forecast 286 system and for a multi-model combination. Figure 1 shows an example of multi-model 287 prediction of 2-meter temperature anomalies from three S2S models, along with the 288 verification. This figure shows that a cold event in the northeast of US and Canada in 289 February 2015 was well predicted for the day 12-18 time range. These S2S products will be 290 made available on the ECMWF public website to support the S2S community with a 3-week 291 delay by the end of 2016.

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#### 4.1 The strong March 2015 MJO event

294 The S2S dataset can be used to assess the performance of current state-of-the-art sub-295 seasonal to seasonal forecasting systems to predict recent extreme events. For instance 2015 296 witnessed an exceptional MJO event in March; it exhibited record amplification resulting in 297 the largest amplitude ever recorded (above 4 standard deviation; Marshall et al. 2016) and 298 triggered the formation of twin tropical cyclones, one on each side of the Equator. The 299 amplification was promoted by the unusually warm waters near the dateline (Marshall et al. 300 2016), which preceded development of strong El Nino conditions in the eastern Pacific later 301 in the year. The surface westerly winds that developed in the western Pacific as a result of 302 this March MJO event with twin cyclones likely enhanced the development of the strong El Niño later in the year. It is encouraging to see that all the models and the multi-model combination (black line in Fig. 2a) forecasted a strong MJO event more than 2 weeks in advance (Figure 2a). Most models also predicted the occurrence of an MJO event 3 weeks in advance (black line in Fig. 2b), although the amplitude is generally underestimated, and no ensemble member predicted such a strong amplitude event.

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This record-strength MJO event also contributed to the formation of Tropical Cyclone Pam, which intensified to Category 5 strength and hit the islands of Vanuatu in the south Pacific on 13 March with devastating effects. Around 15 people were killed and many buildings were destroyed. The cyclone was the second strongest on record in the southern Pacific, second only to Zoe (2002). It is regarded as the worst natural disaster in Vanuatu's history. The cyclone formed on 6 March east of the Solomon Islands and was classified as a tropical storm on 9 March.

316 Previous studies (e.g. Vitart, 2009) have demonstrated that state-of-the-art extended-range 317 forecasting systems can simulate the modulation of tropical cyclone activity by the MJO, with 318 an increase risk of tropical cyclone activity over the South-West Pacific when the MJO is in 319 Phase 6 and 7. In order to assess the skill of the S2S models to predict the probability of a 320 tropical cyclone hitting Vanuatu, tropical cyclones have been tracked in each ensemble 321 forecast member from CMA, JMA, NCEP, ECMWF and BoM using the algorithm described in 322 Vitart et al. (1997). Figure 3 shows the probability of a tropical cyclone strike within a 300 km 323 radius for the multi-model combination of the 5 real-time forecasts starting on 19 and 26 324 February 2015 and verifying on the weekly period 9-15 March 2015 when Pam hit the islands 325 of Vanuatu. Figure 3 suggests that this event had some extended-range predictability, the

326 multi-model combination indicating an increased risk of tropical cyclone strike probability in 327 the vicinity of Vanuatu (indicated by a black dot in Figure 3) 2 to 3 weeks in advance. The 328 multi-model also predicted the possibility of a tropical cyclone strike in the western Pacific, 329 which is consistent with the twin tropical cyclone genesis associated to the strong MJO event 330 of March 2015. The multi-model forecast from 26 February also predicted an increased risk 331 of tropical cyclone strike east of Madagascar and over the northwest coast of Australia which 332 could correspond respectively to tropical Storm Haliba (7-10 March 2015) and tropical 333 cyclone Olwyn (8-14 March 2015).

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#### 336 4.3 MJO Teleconnections in the Northern Extratropics

337 Accurate predictions of MJO events are not sufficient for successful sub-seasonal forecasts. 338 The ability to predict the impact of MJO events on the global circulation is crucial. By acting 339 to excite the NAO, the MJO affects European weather (Cassou 2008; Lin et al. 2009) and 340 North Atlantic significant ocean wave heights (Marshall et al. 2015). Cassou (2008) and Lin et 341 al. (2009) showed that the probability of a positive phase of the NAO is significantly increased 342 about 10 days after the MJO is in Phase 3 (Phase 3 + 10 days), and significantly decreased 343 about 10 days after the MJO is in Phase 6 (Phase 6 + 10 days). The probability of a negative 344 phase of the NAO is decreased (increased) about 10 days after the MJO is in Phase 3 (Phase 345 6). The impact of the MJO on two other Euro-Atlantic weather regimes, the Atlantic Ridge 346 and Scandinavian blocking, is much weaker.

347 Vitart and Molteni (2010) showed that a set of ECMWF reforecasts using cycle 32R3 348 displayed realistic MJO teleconnections over the Northern Extratropics, consistent with the 349 observed impacts (Cassou 2008; Lin et al. 2009). Lin et al. (2010) further found that the MJO 350 has a significant impact on the intra-seasonal NAO skill scores using the ECCC model. This 351 section evaluates whether the MJO teleconnections in the Northern Extratropics are 352 adequately simulated in the reforecasts from the S2S database. We do this by forming 500 353 hPa geopotential height composites 10 days after an MJO is in Phase 3 for all cases when the 354 predicted MJO has amplitude larger than one standard deviation. Only the reforecasts 355 covering the period from January to April have been considered.

356 Figure 4 shows that the models generally capture the spatial pattern of the teleconnection 357 but tend to overestimate the intensity of the MJO teleconnections in the North Pacific and 358 underestimate its projection onto the positive phase of the NAO over the North Atlantic 359 basin. This underestimation could be explained by the analysis being based on a single 360 observed realization whereas the model composites are averaged over several ensemble 361 members. Since not a single ensemble member reproduced the intensity of the teleconnection in the North Atlantic sector as strongly as in the analysis, it follows that 362 363 underestimation of the MJO impact over the Atlantic (Vitart and Molteni (2010) is a real 364 deficiency, common to several models. The under-representation of the MJO impact over 365 the Euro-Atlantic sector is likely to limit the predictability and predictive skill over the North 366 Atlantic and Europe in the sub-seasonal time range and therefore is an important aspect to 367 be analyzed.

368

#### **5)** Other activities

370 The above examples give a flavour of the potential scope for research that the database 371 offers. This database will also help to assess the potential of current operational S2S systems 372 to forecast the extreme events around the globe, which are discussed in the BAMS special 373 annual supplement on extremes, and other events which have led to major humanitarian aid 374 responses. Three important aspects of the S2S database---namely that it contains (a) an 375 archive of real-time forecasts (3 weeks delayed), (b) accompanying re-forecast sets, and (c) 376 that these outputs are from WMO-recognized systems used currently for operational 377 forecasts---make it a uniquely powerful tool for improving operational forecasts and 378 exploring and prototyping decision support elements based on S2S forecast information. The 379 WMO Lead-Centre for Long-Range Forecast Multi Model Ensembles (LC-LRFMME) will have access to the S2S database and will obtain the real-time forecasts without the 3-week 380 381 embargo, enabling National Meterorological and Hydrological Services (NMHSs) to utilize 382 real-time forecast information in a few years time once the necessary research has been 383 done to estimate and document skill and approval has been obtained by WMO. The S2S 384 database will augment the resources available to developing countries to enable the research 385 in early warning system products. The S2S project is using the database to train young 386 developing-country scientists to access the data, perform the necessary research, and 387 collaborate with international experts.

388

**6)** Conclusions

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The S2S database, a key component of the WWRP-WCRP Sub-seasonal to Seasonal Prediction
Project science plan, is currently open to the public. It contains reforecasts and also near real-

393	time sub-seasonal to seasonal forecasts from all the major operational centers. This database
394	represents an important tool to advance our understanding of the sub-seasonal to seasonal
395	time range that has been considered for a long time as a "desert of predictability". Use of
396	this database by the research community can include:
397	- Assess the average forecast skill of sub-seasonal to seasonal predictions in a statistical
398	way through the large number of reforecasts and near-real time forecasts;
399	- Assess the potential predictability of the S2S models and identify forecast windows of
400	opportunity;
401	- Perform case studies to assess the skill of the model during a specific period or event;
402	- Identify sources of predictability, dynamical processes and their impact on the
403	forecast skill scores (e.g. sudden stratospheric warmings, MJO and its
404	teleconnections, sea-ice, soil initial conditions);
405	- Assess the models capability to represent these key dynamical processes that are
406	sources of sub-seasonal predictability so as to guide ongoing model development
407	- Assess the benefit of a multi-model approach on sub-seasonal time scale and
408	estimate the effective ensemble size of the multi-model ensemble as in Pennell and
409	Reichler (2011) for climate models.
410	- Assess the representation of model uncertainty in the current operational systems;
411	- Assess the potential benefit of sub-seasonal to seasonal forecasts in applications;
412	- Compare the strategies for model initialization (e.g. burst vs lag ensemble
413	initialization).

415 Work is ongoing to extend the list of oceanic and sea-ice variables and improve the 416 conversion of the data into NetCDF. There are also plans to automatically compute some 417 products from the database (e.g MJO, North Atlantic Oscillation, El-Niño Southern Oscillation, 418 Sudden Stratrospheric Warming indices, weather regimes, tropical cyclone tracks...) and 419 make them available to the community to avoid multiple computations of the same indices. 420 For example the International research Institute for Climate and Society (IRI) at Columbia 421 University also plans to make available a user-oriented subset of products from the S2S 422 database hosted at ECMWF and CMA. 423 424 425 Acknowledgements 426 Duane Waliser's contribution was carried out on behalf of the Jet Propulsion Laboratory, 427 California Institute of Technology, under a contract with NASA. The part of Mikhail Tolstykh's 428 contribution (HMCR model diagnostics) was funded by Russian Science Foundation (grant No. 429 14-37-00053). The authors would like to thank Gilbert Brunet and two anonymous reviewers 430 for their suggestions and comments which helped improve this manuscript. 431 References: 432 Bender, Morris A., Isaac Ginis, 2000: Real-Case Simulations of Hurricane-Ocean Interaction 433 Using A High-Resolution Coupled Model: Effects on Hurricane Intensity. Monthly Weather 434 Review: Vol. 128, No. 4, pp. 917-946.

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#### 497 Figure captions:

Figure 1: Multi-model comparisons. A possible use of the database is to make comparisons between the outputs of different forecasting centers. The image shows forecasts of 2-meter temperature anomalies from three S2S ensemble mean forecasts and a verification panel based on ECMWF reanalysis (ERA-INTERIM, Dee et al. 2011). The forecast start date is 22 January 2015 and the forecast range is days 12–18. The areas where the ensemble forecast is not significantly different from the ensemble climatology, according to a Wilcoxon-Mann–Whitney (WMW) test (see for example Wonacott and Wonacott 1977), are blanked.

505 Figure 2: Phase diagram showing MJO index forecasts from five S2S systems. Forecasts are

506 initiated on a) 5 March 2015 and b) 26 February 2015 and are represented in colored lines.

507 The grey and the black thick solid lines represent the verification and the multi-model

508 ensemble respectively. The MJO index is based on a combined Empirical Orthogonal Function

509 (EOF) analysis using fields of near-equatorially-averaged 850-hPa and 200-hPa zonal wind

510 and outgoing longwave radiation (OLR) (Wheeler and Hendon 2004). The RMM1 and RMM2

511 give an information on the location of the MJO: Indian Ocean (quadrant 2 and 3), Maritime

512 Continent (quadrant 4 and 5), western pacific (quadrant 6 and 7) and western hemisphere

513 (quadrant 8 and 1). The amplitude of the MJO is represented by the distance to the center,

514 and the inner circle represents one standard deviation.

Figure 3: Probability anomalies of a tropical storm strike within 300 km radius from the multimodel ensemble (combination of ECMWF, NCEP, CMA, JMA and BoM forecasts). The forecasts were initialized on 26 February 2015 (top panel), 19 February 2015 (bottom panel) and cover the weekly period 9-15 March 2015, which corresponds to a forecast range of day 12-18 (top panel) and day 19-26 (bottom panel). The black dot in each panel represents the location of landfall of tropical cyclone Pam over Vanuatu islands.

521	Figure 4: MJO Phase 3 10-day lagged composites of 500 hPa geopotential height anomaly
522	from ECMWF, NCEP, JMA and BoM over the Northern Extratropics for the period January to
523	April 1999 to 2010 (common re-forecast period) and ERA-Interim (left panel). Red colors
524	indicate positive anomalies. Blue colors indicate negative anomalies. The contours are plotted
525	every 10 meters.
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	Time-		Ens						Ocean	Sea-ice
Model	range	Resolution	size	Freq	Rfc	Rfc period	Rfc freq	Rfc size	Coupling	coupling
				twice					YES	NO
BoM	d 0-62	~2x2 L17	33	weekly	fixed	1981-2013	6/month	33		
СМА	d 0-60	~1x1 L40	4	daily	fixed	1994-2014	daily	4	YES	YES
					on the				NO	NO
ECCC	d 0-32	0.45x0.45 L40	21	weekly	fly	1995-2012	weekly	4		
ECMWF	d 0-46	0.25/0.25 day 0-10	51	twice	on the	past 20y	2/week	11	YES	NO
		0.5x0.5 after day 10		weekly	fly					
		L91								
					On the				NO	NO
HMCR	d 0-61	1.1x1.4 L28	20	weekly	fly	1985-2010	weekly	10		
CNR-							Every 5		NO	NO
ISAC	d 0-31	0.8x0.56 L54	41	weekly	fixed	1981-2010	days	1		
				twice					NO	NO
JMA	d 0-33	~0.5x0.5 L60	25	weekly	fixed	1981-2010	3/month	5		
					on the				YES	YES
KMA	d 0-60	~0.5x0.5 L85	4	daily	fly	1996-2009	4/month	3		
CNRM	d 0-61	~0.7x0.7 L91	51	monthly	fix	1993-2014	2/month	15	YES	YES
NCEP	d 0-44	~1x1 L64	16	daily	fixed	1999-2010	day	4	YES	YES
					on the				YES	YES
UKMO	d 0-60	~0.5x0.8 L85	4	daily	fly	1996-2009	4/month	3		

**Table 1:** Main characteristics of the 11 contributions to the S2S database where:

**Time range:** Forecast lead time in day

**Resolution:** Longitude and latitude resolution in degrees. The number after the letter L

546 represents the number of vertical levels.

548 **Ens size:** Number of members in the real-time forecast ensemble.

549 **Freq**: How often (Frequency) the forecasts are run.

**Rfc:** Re-forecast (hindcast) are run using the actual forecast model but for past several years
on the same (or nearby) calendar day as the forecast. The re-forecast is used to calibrate the
actual forecast. There are two types of reforecasts:

fixed: Some operational centers (e.g. NCEP) use the same version of their model
("frozen" version) to produce real-time S2S forecasts over a period of several years

555 (typically 4-5 years). Therefore, the reforecasts are produced once, often before the

- 556 first real-time forecast is produced, and used for several years to calibrate the real-
- 557 time forecasts.

558 **on-the-fly:** Other operational centers (e.g. ECWMF) update their model version

559 several times per year. In order to ensure model consistency between real-time

560 forecasts and re-forecasts, the re-forecasts are produced continuously just before

561 the real-time forecast they will be used to calibrate. For example, at ECMWF, every

week, a set of reforecast is produced starting the same day and same month as the

563 next real-time forecast (e.g. 1<sup>st</sup> January 2015) but for the past 20 years (1<sup>st</sup> January

564 1995 to 2014).

562

567

565 **Rfc period:** The number of years the reforecasts are run. In some centers, the number of re-566 forecast years is fixed, but the list of years varies from year to year. For instance the re-

568 **Rfc freq:** How often the reforecasts are run.

569 **Rfc size**: The number of ensemble members for reforecasts.

forecast years at ECMWF cover the past 20 years.

570	Ocean coupling: Indicates if the atmospheric component is coupled to a dynamics ocean
571	model
572	Sea-ice coupling: Indicates if an active dynamical sea ice model is included or not.
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Name	Abbreviation	Unit	Frequency
Geopotential height	gh	gpm	Instantaneous once a day (00Z)
Temperature	t	К	Instantaneous once a day (00Z)
U-velocity	u	m s-1	Instantaneous once a day (00Z)
V-velocity	v	m s-1	Instantaneous once a day (00Z)

590	Table 2: 3-D parameters available	e on 10 pressure levels (	(1000, 925,	850, 700, 5	500, 300, 200,
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591 100, 50 and 10 hPa) from all models.

Name	Abbreviation	Unit	Frequency
Specific	q	kg kg-1	Instantaneous
humidity			

606	Table 3: 3-D parameter available on 7 pressure levels (1000, 925, 850, 700, 500, 300, 200)
607	from all models.
608	
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	Name	Abbreviation	Unit	Frequency
	Vertical pressure velocity	w	pa s-1	once a day
623				
624	Table 4: The following parameter	er is available at	500 hPa	
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Name	Abbreviation	Unit	Frequency
Potential	pv	K m2 kg-1 s-1	once a day
vorticity			

<sup>641</sup> Table 5: The following parameter is available only at 320K.

Name	Abbreviation	Unit	Frequency
10 meter u	10u	m s-1	Instantaneous once a day (00Z)
10 meter v	10v	m s-1	Instantaneous once a day (00Z)
САРЕ	саре	kg-1	Daily average
Skin temperature	skt	К	Daily average
Snow depth water	sd	kg m-2	Daily average
equivalent			
Snow density	rsn	kg m-3	Daily average
Snow fall water	sf	Kg m-2	Accumulated once a day
equivalent			
Snow albedo	asn	%	Daily average
Soil moisture top	sm20	kg m-3	Daily average
20cm			
Soil moisture top	sm100	kg m-3	Daily average
100cm			
Soil temperature to	st20	К	Daily average
20cm			
Soil temperature top	st100	К	Daily average
100cm			
Surface air max	mx2t6	К	Instantaneous 4 time a day
temperature			
Surface air min	mn2t6	К	Instantaneous 4 times a day
temperature			

Surface air	2t	К	Daily average
temperature			
Surface air dewpoint	2d	К	Daily average
temperature			
Sea surface	wtmp	К	Daily average
temperature			
Sea ice cover	сі	proportion	Daily average
Surface pressure	sp	Ра	Instantaneous once a day (00Z)
Mean sea level	msl	Ра	Instantaneous once a day (00Z)
pressure			
Total cloud cover	tcc	%	Daily average
Total column water	tcw	Kg m-2	Daily average
Total precipitation	tp	Kg m-2	Accumulated 4 times a day
Convective	ср	Kg m-2	Accumulated once a day
precipitation			
Northward turbulent	nsss	N m-2 s	Accumulated once a day
surface stress			
Eastward turbulent	ewss	N m-2 s	Accumulated once a day
surface stress			
Water runoff and	ro	kg m-2	Accumulated once a day
drainage			
Surface water runoff	sro	kg m-2	Accumulated once a day
Land sea mask	lsm	Proportion	Instantaneous once a day (00Z)

		of land	
Orography	orog	gpm	Instantaneous once a day (00Z)
Soil type	slt	Categorical	Instantaneous once a day (00Z)
Top net thermal	ttr	W m-2 s	Accumulated once a day
radiation			
Surface latent heat	slhf	W m-2 s	Accumulated once a day
flux			
Surface net solar	ssr	W m-2 s	Accumulated once a day
radiation			
Surface net thermal	str	W m-2 s	Accumulated once a day
radiation			
Surface sensible heat	sshf	W m-2 s	Accumulated once a day
flux			
Solar radiation	ssrd	W m-2 s	Accumulated once a day
downwards			
Surface thermal	strd	W m-2 s	Accumulated once a day
radiation downwards			

**Table 6:** *List of single level parameters* 





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668	not significantly different from the ensemble climatology, according to a Wilcoxon-Mann-
669	Whitney (WMW) test (see for example Wonacott and Wonacott 1977), are blanked.

676





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