

1 **A White Paper**
2 **Resulting from the US CLIVAR Workshop**
3 **“Translating Process Understanding to Improve Climate Models”**

4 NOAA GFDL, Princeton, New Jersey

5 October 15-16, 2015

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12 **1. Introduction**

13 This white paper synthesizes the oceanic, atmospheric and climate, modeling and
14 observational, communities’ input on the need for a coordinated effort to translate process
15 understanding into climate model improvements. It aims to assess the need for launching a new
16 effort and addresses the questions of what form such an effort ought to take, which areas need to
17 be tackled, and how such an effort might be implemented.

18 During the past 12 years, NSF and NOAA have supported Climate Process Teams (CPTs), a
19 concept that was initiated by US CLIVAR, to translate process understanding into climate model
20 improvements with the aim of reducing model biases. With the currently funded CPTs coming to
21 an end, there is a need to review their benefits and devise a plan for future efforts.

22 A steering committee formed by the US CLIVAR Process Study and Model Improvement
23 Panel (PSMIP) conducted a survey of modeling centers, process study groups, enhanced
24 observing projects, recent satellite missions, recent CPTs, and US CLIVAR Working Groups to
25 collect feedback on the utility of CPTs and the continued need for such efforts. The survey
26 specifically targeted the large modeling centers and a wide range of process studies/
27 observational efforts (see full list in Appendix). The results of the [modeling center survey](#) and
28 the [process studies and observational projects survey](#), available through the embedded links,
29 confirmed broad community interest for a scoping workshop.

30 A workshop was therefore convened by US CLIVAR (PSMIP) at the encouragement of the
31 US Inter-Agency Group, to seek input from the observational, modeling, and theoretical
32 communities on how to achieve a translation of process understanding into climate model
33 improvements. The workshop was funded by NSF, NOAA, and DOE, with local support from
34 NOAA GFDL. The steering committee organized the workshop, reported to the Inter-Agency
35 Group on its outcomes by means of a short report and teleconference, and has now compiled this
36 white paper to provide feedback to the community.

37 The community-wide workshop with open participation was held at GFDL, on October 15-16,
38 2015 to discuss possible mechanisms to translate process understanding to model developments
39 and to identify processes for which newly available observational data and understanding could
40 inform future model improvements. Workshop attendance was limited to 90 participants and was
41 quickly filled to capacity, but web streaming made the workshop talks available to anyone who
42 could not participate in person (a further ~80 remote participants).

43 Scheduled over two full days, the workshop included invited oral presentations, posters,
44 breakout sessions, and participant discussions. Oral presentations were solicited from
45 representatives of modeling centers, who were requested to highlight model biases and
46 weaknesses. Process study group representatives were invited to describe newly developed
47 process understanding from observational and theoretical studies, which could inform model
48 improvement. These presentations were solicited to be as inclusive and as broad as possible, but
49 they could not represent all the interests expressed in the survey due to time constraints.

50 An agenda for the workshop is available [online](#). It targeted the ocean, atmosphere, cryosphere,
51 and land, and the interactions between these climate components. After some discussion early
52 on, and from the feedback of the surveys, it was thought best to maintain the focus of these
53 efforts on physical climate-related processes and to encourage discussion on the interaction
54 amongst components of the climate system, which had been identified as important areas for
55 model improvement in the surveys.

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58 **2. Need for process translation into model development**

59 In the past decade, CPTs brought together observationalists, theoreticians, process modelers,
60 and model developers to work closely on improving parameterizations of a particular process in
61 one or more global models. The US CLIVAR-sponsored CPTs were initiated in 2003, followed
62 by a second round in 2010, and included funding from both NOAA and NSF, with some
63 involvement from NASA. They focused on improvements in Intergovernmental Panel on
64 Climate Change (IPCC)-class models (particularly at GFDL and NCAR) used for climate change
65 simulations. Other NOAA-sponsored CPTs were designed to specifically improve NOAA
66 models, including NCEP and GFDL models. Previous CPTs have largely focused on low-latitude
67 cloud processes in the atmosphere and ocean eddy and mixing processes. In addition, two CPTs
68 had a cryospheric focus, as summarized in Table 1.

69 Following the past experience with CPTs, there is strong recognition from the community that
70 bringing process experts together with climate modelers is a useful means of improving
71 representation of physical processes in large-scale models. The past CPTs have led to important
72 improvements in IPCC-class models; examples include: new cloud parameterizations (e.g.,
73 CLUBB as implemented in the Community Atmosphere Model; Bogenschutz et al. 2013), new

74 subgrid-scale effects of photosynthetically available radiation in ice-covered waters (Long et al.
75 2015), new ocean model representations of shear-driven mixing (Jackson et al. 2008),
76 hydraulically controlled flow and mixing in straits (Wu et al. 2007), bottom boundary mixing
77 (Legg et al. 2006), and mixed layer submesoscale restratification (Fox-Kemper et al. 2008).
78 These improvements are included in one or more Coupled Model Intercomparison Project phase
79 5 (CMIP5) models. Recently, NOAA-sponsored CPTs also led to operational implementations
80 into the NCEP model (e.g., dry EDMF boundary layer parameterization; Han et al. 2016). By
81 focusing in depth on a single problem for a five year period, CPTs have accelerated scientific
82 understanding of particular processes, for example by providing a more complete picture of the
83 ocean internal wave energy distribution ([CPT website](#)) and stimulating research into ocean
84 submesoscale processes (Boccaletti et al. 2007). Through involvement in the CPTs, strong and
85 enduring links have developed between specific scientific communities in academia and model
86 developers, for example through the continuing presence at the modeling center of process
87 experts originally hired as CPT liaisons.

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Table 1: Summary of previous CPT efforts, including lead-PI, agency, dates and modeling centers involved.

Climate process team topic	Lead PI	Funding agency	Dates	Modeling centers involved
Ocean eddy mixed layer interactions	Raf Ferrari (MIT)	NOAA/NSF	2003-2008	GFDL, NCAR
Gravity current entrainment	Sonya Legg (Princeton)	NOAA/NSF	2003-2008	GFDL, NCAR
Low latitude cloud feedbacks on climate sensitivity	Chris Bretherton (UW)	NOAA/NSF	2003-2006	GFDL, NCAR, GSFC
Improving the subtropical Sc-Cu transition	Joao Teixeira (JPL/Caltech)	NOAA	2010-2013	NCEP, NCAR
Cloud parameterization and aerosol indirect effects	Vince Larson (UWisc)	NOAA/NSF	2010-2015	NCAR, GFDL
Ocean mixing processes associated with high-spatial heterogeneity in sea ice	Meibing Jin (U Alaska)	NOAA/NSF	2010-2013	NCAR, GFDL
Internal wave driven mixing	Jen MacKinnon (SIO)	NOAA/NSF	2010-2015	NCAR, GFDL

Improving turbulence and cloud processes in the NCEP global models	Steve Krueger (U Utah)	NOAA	2014-2017	NCEP, NCAR
Representing calving and iceberg dynamics in global climate models	Olga Sergienko (U Princeton)	NOAA	2013-2016	GFDL
Cloud and boundary layer processes	Chris Bretherton (UW)	NOAA	2014-2017	NCEP, GFDL

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These past CPTs have brought together people from modeling and observational communities who otherwise would have had little opportunity for interaction. The US CLIVAR CPTs have typically included 7-12 PIs, as well as several postdoctoral researchers, some of whom were placed at the modeling centers. The modeling centers benefited from the exposure to new ideas, the physical insight obtained from observational data, and the involvement of groups looking at the specifics of the process from different angles. Modeling centers funded by different agencies with different missions have been able to pool their resources to tackle a particular scientific problem (e.g., low cloud parameterization problem). The academic community has gained access to modeling center expertise, models and computer resources, and knowledge about the requirements and limitations of climate models. A long-term outcome of such interaction is the synthesis of results from numerous process experiments into forms suitable for reference by model developers. Examples include the “Table of observations” condensing observations of oceanic overflows into a convenient reference (Legg et al. 2009) and the synthesis of ocean mixing data (Waterhouse et al. 2014).

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In summary, the past CPTs have been an effective mechanism to facilitate interaction between process experts and model developers focused around improvements in the representation of particular processes. Nonetheless, the success of the past CPTs does not diminish the need for future activities designed to bring together climate modelers and process experts. Such activities should improve upon the structure of past CPTs, incorporating those elements that have proven successful, while making modifications to enhance their effectiveness and relevance. Only a limited number of processes have so far been targeted with CPTs, as seen in Table 1. Numerous processes, and their interactions, remain poorly represented in large-scale models, and such models still have many biases that may be improved by better process representation, as detailed in later sections. In many cases, it is the interactions between climate components (e.g., ocean-atmosphere, land-ocean, ice-ocean), processes, and the ways in which parameterizations might interact, that remain poorly represented and uncertain in weather and climate models. The experts involved in processes not included in past CPTs may still remain relatively unconnected

119 to climate model developers. Modelers, theorists, and observationalists are generally not
120 collocated, and modeling centers and process study scientists may receive funds from different
121 sources. The ability for different modeling centers to work together to advance science depends
122 on coordination between their different funding agencies. Often, there remains a mismatch
123 between the disparate scientific results obtained from process studies and the information that a
124 model developer can use. For this, synthesis is needed. For example, the numerous process
125 studies of estuaries and river outflows could be synthesized to provide a reference against which
126 to compare climate model representations. Hence, a need for specific mechanisms for
127 coordinated funding to bring together scientists from academia and different modeling centers to
128 focus on particular model improvements still remains.

129 The climate modeling environment has evolved since the first CPTs in 2003. One of the
130 objectives of the 2015 workshop was therefore to explore new ways for making these
131 coordinated activities more efficient and relevant. Breakout sessions were planned to brainstorm
132 alternatives to the past CPT approaches and examine ways in which hurdles to translating
133 understanding to climate model improvement could be overcome.

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136 **3. Format of teams**

137 The CPT approach sought to bring together observationalists, theoreticians, and modelers to
138 improve model representation of targeted processes. Aspects of the CPT approach originally
139 supported by NSF and NOAA have now been espoused by other projects and agencies, including
140 DOE, NASA, and ONR. A number of programs, within the aforementioned agencies, seek to
141 engage the university/research community in model improvement. Between 2011 and 2015,
142 ONR funded a multi-institution five-year Departmental Research Initiative (DRI) on “Unified
143 Physics for Extended-Range Prediction.” DOE has just announced a new funding opportunity for
144 “Climate model development and validation” in the context of its “Accelerated Climate
145 Modeling for Energy” (ACME) project. NASA’s “Modeling, Analysis and Prediction,” and
146 NOAA’s “Climate Test Bed” are regular components of the “Research Opportunities in Space
147 and Earth Sciences” (ROSES), as well as the Climate Program Office’s regular request for
148 proposals.

149 The buy-in of elements of the CPT approach by applied/mission-driven agencies prompts the
150 question of the potential benefit of simultaneously testing novel, process-derived
151 parameterizations in the diversity of models supported by the different agencies, against the cost
152 of negotiating different priorities. Realistically, the modeling centers each have their particular
153 strengths, so with more institutions involved, an approach in which individual institutions divide
154 up the workload according to their strengths may be more effective. Differences in agency
155 mission, e.g., NOAA’s interest in seamless prediction from weather to seasonal climate

156 timescales, as opposed to NASA’s interest in guiding the design and value of new satellite
157 missions, could thus lead each agency to prioritize the improvement of the representation of
158 different processes.

159 The survey respondents and workshop participants, however, were clear in their support for
160 future efforts to involve multiple modeling centers. The resulting diversity of expertise and
161 approach leads to better science, and academic experts prefer to enhance the overall state of
162 knowledge, rather than tie their efforts to a single agency’s model—especially now that multi-
163 model ensembles have become available from dozens of modeling centers worldwide through
164 the many phases of CMIP.

165 A few management issues were brought up as potential concerns when coordinating multiple
166 centers, but the feeling was that the benefits greatly outweigh the disadvantages, given that some
167 of the more macroscopic biases (e.g., in the climatologies of the eastern tropical basins including
168 the double Intertropical Convergence Zone (ITCZ) in the tropical Pacific) are long-standing
169 problems shared across many models. Logistical challenges for multi-agency, multi-modeling
170 center efforts include a lack of human resources in modeling centers to appropriately engage
171 with a multitude of university investigators, especially face-to-face in meetings; coordination
172 amongst different agencies with different priorities; identification of individuals to lead such an
173 effort and effectively communicate across the diverse group of experts and agencies; and dealing
174 with the different procedures inherent to each agency. A key hurdle for a successful translation
175 of process understanding into models is communication amongst the project participants. This
176 was overcome in the past through regular face-to-face meetings. Another key point that was
177 emphasized is that some of the most useful work that results from the CPTs occurs towards the
178 tail end of the project (e.g., the examination of the impact of Nordic overflow parameterizations
179 on AMOC variability, Danabasoglu et al. 2010), or even after the project has officially ended (in
180 terms of funding). Recognition and support for this “analysis tail” would help to extract the most
181 value from the project.

182 A well-conceived scientific focus around specific processes and biases (for some earlier
183 CPTs) was a key factor in the success of past CPTs. However, experience from past CPTs shows
184 that new or improved parameterizations of a process in a climate model do not always lead to a
185 reduction in model biases or improved representation of climate phenomena. This is due to often
186 unpredictable and complex interactions between different physical processes.

187 In addition to retaining the process-specific focus, some workshop participants and survey
188 respondents suggested that CPTs should also be focused on specific model biases. Candidate
189 problems would be those that required holistic consideration of the coupled ocean-atmosphere-
190 land-ice system, and for which knowledge/understanding of the processes (e.g., from
191 observations or theoretical process studies) was at a sufficiently advanced stage, but had not been
192 translated into climate model applications yet. New approaches to building teams for model

193 improvement, as described below, might allow for more efficient translation of the scientific
194 successes seen in past CPTs to model bias improvements.

195 Though the challenges associated with the multi-model/agency approach are considerable, and
196 while single modeling center/agency efforts can be an effective mechanism for improving a
197 single model, the higher payoff to the community as a whole of a multi-model, multi-agency
198 effort encourages the additional efforts and resources required for overcoming these hurdles with
199 planned action.

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202 **4. Opportunities for translating process understanding to model** 203 **improvements**

204 The workshop included presentations and discussion on biases within climate models,
205 relevant process understanding, and areas where that understanding might be in a suitable state
206 of readiness for translation into climate model developments. Only a subset of the climate
207 science community could attend the workshop, and the topics discussed were naturally
208 dependent on the individuals involved and topics highlighted in the survey results. As such, the
209 opportunities identified at the workshop should be considered illustrative examples of the kinds
210 of topics that could be tackled in future activities. Such an illustrative list is presented in Table 2.

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212 **4.1 Modeling biases/areas requiring improvement**

213 Information on key biases within climate model simulations was provided through a response
214 to a survey questionnaire and talks and breakout group discussions at the workshop. Much of this
215 information was provided by several modeling centers themselves with breakout group
216 discussions allowing a wider community perspective. The biases encompass all climate system
217 components, including the ocean, atmosphere, land, and sea ice, and the coupling among them.
218 In many cases, these biases can influence simulated biogeochemistry, the carbon cycle, and the
219 transient climate system response. For example, adequately representing the upper ocean and the
220 mixed-layer is important for not only representing short time scale atmosphere-ocean interaction,
221 but also because of its important role in primary productivity.

222 Many of the highlighted biases have existed through many generations of climate model
223 development. For example, the presence of a double ITCZ, the warm sea surface temperature
224 (SST) biases in eastern ocean boundary upwelling regions, and biases in the position and
225 strength of the Gulf Stream and associated biases in sea surface temperature, sea surface salinity
226 and surface fluxes of heat, are persistent deficiencies in climate models. A number of biases
227 discussed were specific to the atmosphere, for example, the generally deficient diurnal cycle of
228 convection and precipitation. Biases are found in aspects of variability, for example the eastward
229 propagation of the Madden-Julian Oscillation (MJO) is in general poorly simulated. Biases in the

230 relative proportion of liquid versus ice in mixed phase clouds with large impacts on radiative
231 fluxes, particularly over the higher latitudes, were also discussed. Ocean-specific biases include
232 those associated with water mass transformation in the Southern Ocean, and those related to
233 processes driving shelf-open ocean exchange. Cryosphere biases related to poor simulation of
234 snow on sea ice and ice sheet-ocean interactions, particularly in fjords, were also noted. Coastal
235 interactions more generally were raised as a concern, including biases associated with estuarine
236 processes and the influence of river runoff on coastal oceans. Finally, there was some discussion
237 of terrestrial biases, including vegetation biases over the continental United States.

238 Some of the noted biases are associated with coarse horizontal resolution. The horizontal
239 resolution in workhorse ocean climate models is typically on the order of 1° with increased
240 meridional resolution in the tropics. This results in resolution that is not adequate to represent
241 baroclinic instability and mesoscale eddies. Ocean currents tend to be too weak in mid-latitude
242 boundary currents and often do not have the correct vertical structure. The coarse model
243 resolution results in a poorly simulated Gulf Stream path and associated heat transport and as a
244 consequence SST biases in the Gulf Stream are amongst the largest in the world's oceans. The
245 Gulf Stream also plays an important role in the Atlantic Meridional Overturning Circulation
246 (AMOC) and associated decadal modes of variability. In the atmosphere, coarse resolution
247 adversely impacts the flow over topography, with many consequent impacts, and some
248 horizontal transport effects, such as the simulation of atmospheric rivers. Increases in model
249 resolution will likely lead to improvement in some of these simulated aspects. However, given
250 the need to run climate simulations for long timescales and multiple ensemble members, many
251 model simulations will continue to be run at resolutions in which these biases are problematic.
252 The transition to models with the capacity for regional refinement was noted and may alleviate
253 some resolution-dependent issues. However, this raises the need for parameterizations that are
254 scale-aware and valid for use across a large range of resolutions; a need that was noted by
255 climate modeling centers and meeting participants.

256 In many cases, studies have provided insights on the causes for model biases. New diagnostic
257 capabilities have become available, such as satellite simulators within climate models that ensure
258 consistent comparisons to satellite observations. These are aiding our understanding of the
259 underlying causes and consequences of certain model biases. Process knowledge and
260 observational data have advanced in many areas, providing the potential for translation into
261 model improvements. Below we outline some example processes that impact specific model
262 biases and are in a state of readiness for translation into climate model developments. The list is
263 by no means exhaustive. Many other candidate topics are possible. However, working through
264 these topics does give some indication of common requirements that are needed to make
265 advancements.

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267 **4.2 Process-understanding in a state-of-readiness for implementation in climate models**

268 A series of talks and breakout sessions targeted areas where advances in process knowledge
269 and observational information have been made that could be used to address some climate model
270 biases. These processes spanned the climate system, including aspects of different climate
271 components and the coupling between them. In some cases, the processes mapped directly on the
272 climate model biases presented. However, even when this was not the case, it was generally felt
273 that improved process representation in many areas would lead to improved and more reliable
274 models.

275 Table 2 contains a list of some of the processes that were discussed, including information on
276 the motivation for addressing them. In addition to improving specific processes within the
277 model, there was a stated need to incorporate new model capabilities, such as the inclusion of
278 estuary and fjord modules to better represent riverine discharge and coupling to the ocean. There
279 was also recognition that some key issues in atmospheric model physics are presently related to
280 the connections between the different parameterizations, highlighting the need for unified
281 parameterizations (e.g., unified boundary layer and moist convection parameterizations). The
282 importance of focusing on processes occurring at the interfaces (e.g., air-sea interaction) was also
283 discussed in some detail.

284 The breakout groups focused at length on the state of readiness of specific processes for
285 translation into climate model improvements. In general, this included processes that were felt to
286 significantly impact important climate model biases, have significant process knowledge, and
287 have observational constraints that would facilitate parameterization/process model
288 developments. Below we address in more detail some example topics including the biases or
289 phenomena that might be impacted, the data and understanding that exist, and how the
290 translation of that information could be used to improve models. The example topics span the
291 atmosphere, ocean, cryosphere, and coupled system but by no means represent an exhaustive list.
292 Indeed the workshop highlighted numerous candidate topics of this type and undoubtedly even
293 these are just a subset of the possible processes in a state of readiness to be improved in models
294 through coordinated activities. The examples discussed below were not chosen based on their
295 relative importance or readiness relative to other candidate topics but instead because adequate
296 information existed from the workshop materials to more fully flesh them out. We provide these
297 examples primarily because they allow insights on the factors that determine the readiness of
298 processes for incorporation into models.

299 300 *4.2.1 Atmosphere example - Moist convection*

301 Moist convection (in the boundary layer, shallow and deep) plays a crucial role in the climate
302 system and a realistic representation of moist convection in weather and climate models is
303 essential for the accurate prediction of a variety of phenomena, from weather to seasonal and

304 climate change prediction, e.g., the diurnal cycle of convection and precipitation over land and
305 ocean; severe storms; MJO; the Monsoon; El Niño-Southern Oscillation (ENSO); and cloud-
306 climate feedbacks.

307 Unfortunately moist convection is notoriously difficult to parameterize in weather and climate
308 models. A variety of moist convection parameterizations have been developed over the last few
309 decades (since the start of numerical weather and climate prediction in the 1960s) but many
310 problems in its representation still need to be resolved. Even with the advent of global cloud
311 resolving models (with horizontal grid resolutions from 1 to 10 km—capable of at least explicitly
312 representing cluster/mesoscale dynamics), the community will need to develop improved moist
313 convection parameterizations—not only for shallow convection and the transition to deep
314 convection, but for deep convection as a whole as well—since at these resolutions the resolved
315 dynamics part is not able to explicitly represent key processes such as turbulent lateral
316 entrainment.

317 An important modern topic of parameterization research is the development of unified
318 parameterizations of all turbulent and convective processes in the Earth’s atmosphere (including
319 shallow and deep moist convection). In fact, the last few years have seen the advent of different
320 approaches to solve this unification problem, which are being implemented in operational
321 weather and climate prediction models. In particular, much work has been performed on
322 approaches based on assumed probability density functions (PDFs; e.g., Bogenschutz et al. 2013)
323 or on optimal blends of eddy-diffusivity (ED), typically used to parameterize more local mixing,
324 and mass-flux (MF), typically used for moist convection (EDMF; e.g., Siebesma et al. 2007).
325 Although versions of PDF and EDMF parameterizations have been tested and implemented with
326 some success recently, it is fair to say that no parameterization that fully unifies the
327 representation of all turbulent and convective processes has yet been implemented in
328 atmospheric models.

329 Extending these new approaches to deep convection, which would allow realistic
330 representation of the transition from shallow to deep convection for example, is perceived as
331 crucial for the development of more accurate weather and climate models. A particularly
332 important topic in this respect is the representation (parameterization) of the more complex
333 convective structures that exist when moist convection gets deeper than the boundary layer and
334 cloud microphysics starts to play a key role in the dynamics and thermodynamics. Over the last
335 several years much work has been done using multi-scale modeling framework (MMF)
336 approaches, where 2D cloud resolving models (CRMs) are embedded in a climate model grid-
337 box (e.g., Randall et al, 2003). MMF approaches can be particularly useful to improve
338 understanding of the interactions between deep convection and the large-scale dynamics.
339 However, MMF approaches are still too (computationally) expensive and often suffer from

340 similar parameterization issues as regular weather and climate models (e.g., clouds, boundary
341 layer, shallow convection – all still need to be parameterized in these CRMs).

342 Multiple-plume convection parameterizations (e.g., Suselj et al. 2013) have grown in
343 popularity in recent years, to try to represent the complexity of moist convection and its interplay
344 with the surrounding environment. But fully unified convection parameterizations, extending
345 from boundary layer and shallow convection to deep convection, still need to tackle significant
346 challenges. These include: the coupling of convection parameterizations to cloud micro and
347 macrophysics parameterizations; downdraft parameterizations; and the role and representation of
348 cold pools.

349 A clear advantage of unified parameterizations of boundary layer mixing and moist
350 convection (as opposed to the more traditional parameterization modularity) is that the
351 interaction of moist convection with the sub-cloud layer occurs in a much more natural
352 (continuous) manner without the need for ad hoc cloud base closures, for example. In addition, it
353 should also improve the representation of the interaction of moist convection with the land and
354 ocean surface. In this context, particular attention should be paid to air-sea flux
355 parameterizations and the interaction with ocean mixing, and the interaction of moist convection
356 with sub-grid orography and gravity waves.

357 Important modern topics of research in numerical weather prediction such as data-
358 assimilation, ensemble prediction, and high-resolution modeling (with horizontal resolutions of
359 the order of 1 - 10 km) have major implications for the development of future convection
360 parameterizations. In particular, the development and successful implementation of stochastic
361 and scale-aware convection parameterizations will be crucial to improve the accuracy and
362 reliability of weather, seasonal, and climate prediction.

363 364 *4.2.2 Ocean example – mesoscale eddy life cycles*

365 Ocean mesoscale eddies, generated through baroclinic instability in regions of strong
366 horizontal density gradients, are smaller than the grid-scale of most global models that are
367 routinely used for climate-scale simulations. The effect of such eddies on buoyancy transports
368 and restratification has typically been represented by variants of the Gent and McWilliams
369 (1990) parameterization. New global models are increasingly becoming eddy permitting, at least
370 at low latitudes. Yet, they are unable to represent the full range of mesoscale eddy activities,
371 including their feedback on surface mixed layers, surface fluxes, large-scale current structure,
372 and the processes by which eddies dissipate or transfer energy to larger scales.

373 Improved representation of the full life cycle of mesoscale eddy energy would impact two
374 major biases in climate models: Western boundary currents (which in turn impact SST biases,
375 AMOC, and decadal variability), and Southern Ocean and subpolar North Atlantic mixed layers
376 (which in turn impact primary productivity and carbon uptake). Progress in improving the

377 representation of these processes would be facilitated by the analysis of new high-resolution (up
378 to 1/50°) simulations, and by making use of new potential vorticity budget diagnostics (e.g.,
379 recently implemented in CESM).

380 Several new parameterization ideas have shown promise in tests, including resolution-
381 dependent mesoscale eddy parameterization (Hallberg 2013), the addition of stochastic
382 backscatter (Jansen et al. 2015a) and use of negative viscosity or non-Newtonian visco-elastic
383 dynamics (Mana and Zanna 2014). The loss of energy from mesoscale eddies to lee waves and
384 hence to diapycnal mixing has been explored by Nikurashin and Ferrari (2011). Jansen et al.
385 (2015b) and Eden and Greatbatch (2008) have developed frameworks for accounting for the full
386 mesoscale eddy energy budget.

387 New observational and process modeling efforts that are underway (ONR DRI including
388 FLEAT, LATMIX and ASIRI) will likely provide new insights for understanding the evolution
389 of eddies and fronts, and the important role that they play in controlling near-surface
390 stratification. Other recent observations (DIMES, Gille et al. 2012) have focused on the
391 interaction between eddies, topography, and diapycnal mixing. In both the upper and deeper
392 ocean, these observations allow for a better understanding of the interaction between internal
393 waves and eddies.

394 In summary, better representation of the eddy energy lifecycle is possible by synthesizing
395 several new parameterization ideas, incorporating the loss of eddy energy to upscale transfer, and
396 dissipation at bottom topography with new understanding from recent field programs, thereby
397 improving representation of the impact of mesoscale eddies on mixed layers, surface fluxes, and
398 large-scale energetic currents. Such a synthesis would involve theory (e.g., geostrophic
399 turbulence inverse energy cascade), observations, high-resolution modeling, and involve
400 interactions between different parameterization components (e.g., interactions between
401 mesoscale eddy parameterizations and abyssal mixing via the generation of lee waves,
402 interactions between mesoscale eddies and mixed layer parameterizations via re-stratification).

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404 *4.2.3 Cryosphere example – Snow on Sea Ice*

405 For much of the year, sea ice has an extensive and highly variable snow cover. This snow
406 exhibits high-spatial heterogeneity and is greatly impacted by factors such as wind redistribution.
407 Snow on sea ice is a highly insulative material (Sturm et al. 1998, 2002) with one of the highest
408 albedos of all natural materials (Perovich et al. 2002). These aspects of the snow play a primary
409 control on sea ice mass budgets and coupled interactions. The snow cover greatly attenuates light
410 transmission to the ice and ocean with consequent impacts on ice and ocean biota. The relative
411 importance of different snow impacts varies by season and is likely to change with changing
412 climate conditions. For example, in winter snow insulates and slows ice growth, while in
413 summer, the highly reflective snow reduces surface melt. As such, the state and variability of

414 snow conditions on sea ice has implications for coupled climate feedbacks and the transient
415 response of sea ice to changing forcing.

416 Climate models simulate large discrepancies in the snow conditions on sea ice (Hezel et al.
417 2012; Light et al. 2015). This influences the climate response to perturbations in forcing and the
418 strength of the surface albedo feedback in future climate projections (Holland and Landrum
419 2015). The physical treatment of snow processes in climate models is quite simple and has
420 remained largely unchanged over multiple model generations. For example, climate models
421 typically assume a constant density snow pack with no liquid water content (e.g., Hunke et al.
422 2014). They also generally exclude factors, such as blowing snow and snow metamorphosis,
423 which impacts the snow mass budgets, its spatial heterogeneity, thermal properties, and surface
424 reflectivity.

425 While the climate model representation of snow on sea ice has remained quite simple,
426 considerable advances have been made in understanding the processes driving variations in snow
427 conditions. Observational data indicate important changes in the thickness of the snow cover
428 (e.g., Webster et al. 2014) that are coupled to and likely feedback on the changing sea ice state.
429 Observations have also provided insights on what influences blowing snow and its redistribution
430 (e.g., Dery and Tremblay 2004; Leonard et al. 2008), factors that influence snow metamorphosis
431 (e.g., Sturm and Massom 2010), and how snow modifies the coverage and location of melt ponds
432 (Perovich and Polashenski 2012). In some cases, aspects of this knowledge have been
433 encapsulated in process models (e.g., Dery and Tremblay 2004; Lecomte et al. 2011) that can
434 provide avenues for parameterization developments for large-scale climate models. New
435 observations, for example from Operation IceBridge surveys (Kurtz and Farrell 2011), are also
436 providing a larger scale perspective on varying snow conditions. The wealth of observations and
437 process knowledge available should allow for significant advances in the process representation
438 for snow on sea ice within climate models. This would improve simulated feedbacks with
439 important implications for the projected climate response.

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441 *4.2.4 Coupled system example – Eastern Boundary upwelling systems*

442 Eastern boundary upwelling systems are regions of high biological productivity and play an
443 important role in the carbon cycle. The dynamics of the processes controlling these regions are
444 highly coupled and dependent on wind forcing, cloud processes, and ocean dynamics. SST biases
445 at eastern boundaries are a longstanding problem with climate models. Various hypotheses have
446 been proposed for their existence, including inadequate stratocumulus cloud representation, weak
447 upwelling and coastal currents, and teleconnection of errors from remote (equatorial) regions.
448 The SST biases are important to climate variability and predictability in these regions, and
449 upwelling biases are important to projections of how coastal ecosystems respond to changing

450 climate, including fishery and other impacts. Consequently, upwelling is one of the research foci
451 of CLIVAR: <http://www.clivar.org/research-foci/upwelling>.

452 The interactions between clouds and SST (i.e., the coupling between atmospheric and ocean
453 physics) play a key role in modulating both the cloud properties and the SST. It is well known
454 that the SST biases extend far beyond the upwelling regions and much of these biases can only
455 be accounted for by biases in cloud cover and liquid water. There is a potential positive feedback
456 in this tight coupling problem: cloud biases lead to SST biases, which in turn lead to even more
457 marked cloud biases, and so on. Biases in cloud cover and liquid water are, to first degree, a
458 manifestation of issues in the vertical thermodynamic structure of the atmospheric boundary
459 layer. New and improved parameterizations of the atmospheric boundary layer turbulence and
460 convection (e.g., PDF-based, EDMF) are essential for the improvement of these cloud and SST
461 biases.

462 Recent work has confirmed that the representation of physical processes in the eastern
463 boundary of ocean basins in models is resolution-dependent in a significant manner. Higher
464 atmosphere horizontal resolution is key to simulating atmospheric jets and obtaining a coastal
465 SST bias reduction (Gent et al. 2010). High-vertical resolution, particularly within the planetary
466 boundary layer, can improve coupled model biases (Harlaß et al. 2015). Higher ocean resolution
467 (to at least 0.1°) is needed to resolve coastal currents/fronts and upwelling (Small et al. 2015).

468 The theory controlling features relevant to the SST biases have been explored in a number of
469 studies. The dynamics of the atmospheric coastal jet have been examined in the framework of
470 hydraulic theory (Samelson 1992), which has been applied to the California Jet. Validity of the
471 theory is being tested for the Benguela Jet (Small et al. 2015), but what key parameters control
472 the jet structure remains to be explored. The linear dynamics of coastal upwelling have been
473 extensively explored in McCreary et al. (1987), Fennel et al. (2012), and Junker et al. (2015).
474 These dynamics were found by Small et al. (2015) to explain much of the errors in one particular
475 climate model (NCAR's CCSM4)—due to the biased off-coast structure of wind forcing. The
476 UCLA ROMS group and others have extensively explored non-linear ocean dynamics, including
477 eddies, frontal filaments, and submesoscale vortices (Capet et al. 2008, references therein).

478 Extensive observations also exist for the three most biased regions in the world, namely the
479 southeast Atlantic/Benguela system, the southeast Pacific/Humboldt current and the northeast
480 Pacific/California Current. For example the Coastal Ocean Dynamics Experiment (CODE)
481 collected some very useful data for the California Jet, the VOCALS campaign (Mechoso et al.
482 2014) took extensive observations for the southeast Pacific and Humboldt current system, and
483 the ongoing PREFACE (Prediction of Tropical Atlantic climate and its impacts,
484 <http://preface.b.uib.no/>) campaign is gathering extensive data of the oceanographic dynamics of
485 the southeast Atlantic/Benguela current system. Additional existing data can be brought to bear
486 on understanding the relative controls on SST conditions in these regions.

487 Given the importance of the SST biases in these regions, advances in theoretical
488 understanding, and extensive observations that exist, model developments are both needed and
489 possible. A number of possible approaches to address this issue include:

- 490 • Enhancements in atmospheric horizontal resolution to at least 0.5° , which are currently
491 being run in some models, and appear sufficient to reduce the SST bias in some regions
492 (Small et al. 2015).
- 493 • Mesh-refined atmosphere and/or ocean components, which are becoming available in
494 some models, and can provide regional refinement in coastal zones.
- 495 • Atmospheric parameterizations to better represent the boundary layer turbulence,
496 convection, clouds (e.g., PDF-based, EDMF), and coastal jets.
- 497 • Ocean parameterizations to mimic the effect of narrow coastal upwelling and coastal
498 oceanic jets in coarser resolution models.

500 **4.3 Implementation-ready processes likely to benefit model improvement**

501 The workshop identified many different opportunities for model improvement by
502 incorporating new process understanding. These opportunities are summarized in Table 2, but
503 any future activities should not be limited to the topics discussed in this document or at the
504 workshop given that the interests of only a subset of the community have been considered thus
505 far. As highlighted in discussions at the workshop and illustrated by examples above, there are
506 some common elements for relevant implementation-ready processes. These include:

- 507 1. Processes that have an important influence on the simulation characteristics. This can
508 include processes for which improved representation ameliorates biases in the simulated
509 climatological state or, just as importantly, increases the realism of climate variability and
510 feedbacks or influences the simulated biogeochemistry and carbon cycle.
- 511 2. Processes that have an adequate level of understanding. This understanding should be
512 informed by theoretical considerations and observational analysis.
- 513 3. Adequate human capital (e.g., theoreticians, observers, modelers) that can synthesize and
514 enhance the relevant process knowledge to enable translation into model improvements.

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Table 2. Processes/phenomena identified at the workshop that participants felt were in a reasonable state of readiness for translation into climate model improvements.

Process / Phenomenon	Potential Bias Improvement	Motivation	References
Estuarine/fjord – ocean interactions	Salinity near estuaries and rivers; coastal ocean stratification	Allows for riverine nutrient & heat transport; impacts coastal biogeochemistry	Geyer and MacCready (2014)
Atmospheric boundary layer and land surface interaction	Forecast biases on sub-seasonal timescales	Improvements in soil moisture coupling to atmospheric boundary layer	Kumar et al. (2014)
Equatorial mixing	Cold tongue bias	Can influence simulated variability (ENSO), surface coupling	Sasaki et al. (2013)
Eddy life cycle and energetics	Mixed layer depth; primary production	Controls on vertical ocean exchange, upper ocean stratification	Jansen et al. (2015b)
Eastern boundary upwelling	Warm regional SST bias	Improve coupled interactions and feedbacks; impacts on BGC	Small et al. (2015)
Western boundary currents	SST, surface heat fluxes, and oceanic heat transport	Potential links to AMOC, decadal variability	Carton et al. (2014), Hu et al. (2015)
Swell	Southern ocean mixed layer bias	Potentially influence ocean transient response	Fan and Griffies (2014)
Shelf-open ocean exchange	Ocean water mass and density structure	Potential influence on shelf biogeochemical processes including upwelling driven primary production, hypoxia, and low pH events	Bryan et al. (2015)
Gravity wave drag	Large-scale circulation	Improved wind stress and coupling	Geller et al. (2013)
Atmospheric moist convection	Diurnal cycle of precipitation	Improvements to tropical climate and variability	Pearson et al. (2014)
Mixed-phase clouds	Radiation biases, precipitation biases	Potential influence on cloud feedbacks	Pithan et al. (2014)
Snow on sea ice	Snow and albedo biases	Influences polar feedbacks	Hezel et al. (2012)

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527 **5. Pathways to future teams**

528 While there was consensus on the success and effectiveness of the current CPT format, a long
529 discussion focused on possible improvements to the structure of future teams. While many of the
530 current and past CPTs have successfully focused on specific processes, there is the realization
531 that new efforts could be focused on questions related to the interactions between different
532 components of the climate system. An example includes the physics of the coupled ocean-
533 atmosphere system in upwelling regions giving rise to SST biases. The highly focused approach
534 of CPTs could be extended to include specific climate phenomena which emerge from multiple
535 interacting processes. These efforts have merit from a “pure” scientific angle and should not be
536 exclusively tied to a specific improvement in model fidelity.

537 It would be useful to explore process translation themes that would attract interest from
538 multiple modeling centers and agencies, including both weather and climate prediction centers.
539 In this context, data-assimilation is a tool that could help bridge the gaps between these different
540 communities. Ensemble prediction has not been studied in detail by previous CPTs, and brings
541 slightly different challenges in terms of parameterization (e.g., stochastic physics). Sensitivity
542 experiments (e.g., model resolution or parameter uncertainty in parameterizations) could also be
543 used to better identify parameters and processes responsible for coupled biases.

544 New computational capabilities now allow for experimental global simulations with ultra high
545 resolution (order of a few km, even if only for a few days) for both the oceans and atmosphere.
546 These new revolutionary global simulations are unique tools to understand atmospheric and
547 oceanic processes at scales between 1 km and the more commonly used grid resolutions of 50-
548 100 km in climate model simulations. In particular, in the atmosphere, these models will provide
549 unique insight into the role of deep convection and mesoscale dynamics and novel ideas on how
550 to move forward with parameterizations of moist convection in climate and weather models.

551 In terms of new observational capabilities, large datasets (e.g., from new Autonomous
552 Underwater Vehicle capabilities, satellite data) and data mining capabilities (making use of big
553 data) could potentially lead to new developments. In particular satellite observations, which often
554 provide a global view of certain variables and processes, have not been at the core of previous
555 CPTs. However, satellite data has been used for data assimilation and validation of weather and
556 climate models. While *in situ* observations or high-resolution models, such as Large-Eddy
557 Simulation (LES) can adequately resolve processes missing in climate models, satellite data
558 rarely possesses the spatial and temporal resolutions to completely represent these physical
559 processes. However, satellite observations offer comprehensive nearly global datasets that have
560 yet to be completely exploited in parameterization development endeavors. Recent examples
561 (e.g., Suzuki et al. 2013) on how to successfully use satellite observations to improve specific
562 physical processes in climate models offer significant promise in this respect. Focused efforts for

563 parameterization development should be encouraged that take advantage of satellite observations
564 and of optimal combinations of high-resolution modeling, *in situ*, and satellite observations.

565 In summary, there are a variety of exciting challenges and opportunities, from focusing on
566 process-interaction to exploiting new computational capabilities and satellite datasets, which will
567 help shape new CPT projects.

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570 **6. Summary and Conclusions**

571 The workshop highlighted key current biases across the plethora of climate and weather
572 models developed and maintained by seven different modeling centers in the US. Past CPTs
573 have led to significant model improvements and helped develop strong and enduring links
574 between specific scientific communities in academia and model developers. The success of past
575 CPTs does not diminish the need for future activities in this arena, as numerous processes remain
576 poorly represented. Better representations of specific processes—as well as the complex
577 interactions between processes—and between ocean, atmosphere, land, and cryosphere
578 components, are likely to reduce still pervasive model biases. Hence, there is consensus on the
579 need for future efforts to harness expertise from the observational, theory, and modeling
580 communities and form dedicated teams that achieve synergy in improvement of climate models.

581 The workshop participants strongly recommend such activities continue in the future. There is
582 consensus that new activities should retain many aspects of the past CPTs. These include the
583 formation of teams involving modelers, observationalists, and theoreticians, based in both
584 modeling centers and academia, and the funding of postdocs dedicated to the task. Workshop
585 participants also gave strong support to multi-modeling center, multi-agency approaches, well-
586 suited to deliver sustainable and comprehensive improvements to climate models.
587 Recommendations for new developments include enlarging the scope of such activities to
588 consider not only teams built around the theme of improving the representation of a specific
589 process but also new teams focused on coupled processes and model component interactions to
590 address specific biases or climate phenomena. New activities must consider the emerging
591 computational and expanded observational capabilities, as well as the challenges associated with
592 the growth in observational and model data. Future mechanisms to facilitate the translation of
593 process understanding to improvements in climate models will be broadly welcomed by the
594 climate science community.

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812 *Appendices to be included in final report:*

813 1. *Modeling Center Survey Responses*
814 2. *Process Study Survey Responses*
815 3. *List of workshop attendees*
816 4. *Workshop agenda*
817 5. *List of acronyms*