

1 **“Uses and Limitations of Observations, Data, Forecasts, and Other Projections**
2 **in Decision Support for Selected Sectors and Regions”**

3
4 **(Climate Change Science Program, Synthesis and Assessment Product [SAP] 5.1)**

5
6 **Executive Summary**

7
8 Earth information—the diagnostics of Earth’s climate, water, air, land, and other dynamic processes—is
9 essential for our understanding of humankind’s relationship to our natural resources and our environment.

10 Earth information can inform our scientific knowledge, our approach to resource and environmental
11 management and regulation, and our stewardship of the planet for future generations. New data sources,
12 new ancillary and complementary technologies in hardware and software, and ever-increasing modeling
13 and analysis capabilities characterize the current and prospective states of Earth science and are a harbinger
14 of its promise. A host of Earth science data products is enabling a revolution in our ability to understand
15 climate and its anthropogenic and natural variations. Crucial to this relationship, however, is understanding
16 and improving the integration of Earth science information in the activities that support decisions
17 underlying national priorities: ranging from homeland security and public health to air quality and natural
18 resource management.

19
20 Also crucial is the role of Earth information in improving our understanding of the processes and effects of
21 climate as it influences or is influenced by actions taken in response to national priorities. Global change
22 observations, data, forecasts, and projections are integral to informing climate science.

23
24 The Synthesis and Assessment Product (SAP), “Uses and Limitations of Observations, Data, Forecasts, and
25 Other Projections in Decision Support for Selected Sectors and Regions” (SAP 5.1), examines the current
26 and prospective contributions of Earth science information in decision support activities and their
27 relationship to climate change science. The SAP contains a characterization and catalog of observational
28 capabilities in an illustrative set of decision support activities. It also contains a description of the

29 challenges and promise of these capabilities and discusses the interaction between users and producers of
30 information (including the role, measurement, and communication of uncertainty and confidence levels
31 associated with decision support outcomes and their related climate implications).

32

33 **Decision Support Tools and Systems**

34 In 2002, the National Aeronautics and Space Administration (NASA) formulated a conceptual framework
35 in the form of a flow chart (Figure 1) to characterize the link between Earth science data and their potential
36 contribution to resource management and public policy. The framework begins with Earth observations,
37 including measurements made *in situ* and from airborne and space-based instruments. These data are input
38 into Earth system models that simulate the dynamic processes of land, the atmosphere, and the oceans.
39 These models lead in turn to predictions and forecasts to inform decision support tools (DST).

40

41 In this framework, DSTs are typically computer-based models assessing such phenomena as resource
42 supply, the status of real-time events (e.g., forest fires and flooding), or relationships among environmental
43 conditions and other scientific metrics (i.e., water-borne disease vectors and epidemiological data). These
44 tools use data, concepts of relations among data, and analysis functions to allow analysts to build
45 relationships—including spatial, temporal, and process-based—among different types of data, merge layers
46 of data, generate model outcomes, and make predictions or forecasts. Decision support tools are an element
47 of the broader decision making context or Decision Support System (DSS). DSSs include not just computer
48 tools but the institutional, managerial, financial, and other constraints involved in the decision-making
49 process.

50

51 The outcomes in these decision frameworks are intended to enhance our ability to manage resources
52 (management of public lands and measurements for air quality and other environmental regulatory
53 compliance) and evaluate policy alternatives (as promulgated in legislation or regulatory directives)
54 affecting local, state, regional, national, or even international actions. To be exact, for a variety of reasons,
55 many decisions are not based on data or models. In some cases, formal modeling is not appropriate, timely,

56 or feasible for all decisions. But among decisions that are influenced by this information, the flow chart
57 (Figure 1) characterizes a systematic approach for science to be connected to decision processes.

58 For purposes of providing an organizational framework, the CCSP provides additional description of
59 decision support:

60

61 In the context of activities within the CCSP framework, decision-support resources,
62 systems, and activities are climate-related products or processes that directly inform or
63 advise stakeholders in order to help them make decisions. These products or processes
64 include analyses and assessments, interdisciplinary research, analytical methods
65 (including scenarios and alternative analysis methodologies), model and data product
66 development, communication, and operational services that provide timely and useful
67 information to decision makers, including policymakers, resource managers, planners,
68 government officials, and other stakeholders. (*“Our Changing Planet,” CCSP FY2007,*
69 *Chapter 7, p. 155).*

70

71 **Our Approach**

72 Our approach to this SAP has involved two overall tasks. The first task defines and describes an illustrative
73 set of DSTs in areas selected from a number of areas deemed nationally important by NASA and also
74 included in societal benefit areas identified by the intergovernmental Group on Earth Observations (GEO)
75 in leading an international effort to build a Global Earth Observation System of Systems (GOESS) (see
76 Tables 1 and 2).

77

78

79 The areas we have chosen as our case studies are air quality, agricultural efficiency, energy management,
80 water management, and public health. As required by the *SAP 5.1 Prospectus*, in the case studies we:

81

- 82 • explain the observational capabilities that are currently or potentially used in these tools;

- 83 • identify the agencies and organizations responsible for their development, operation, and
84 maintenance;
- 85 • characterize the nature of interaction between users and producers of information in delivering
86 accessing and assimilating information;
- 87 • discuss sources of uncertainty associated with observational capabilities and the decision tools and
88 how they are conveyed in decision support context and to decision makers; and
- 89 • describe relationships between the decision systems and global change information, such as
90 whether the tools at present contribute or in the future could contribute to climate-related
91 predictions or forecasts.

92

93 Because our purpose in this first task is to offer case studies by way of illustration rather than a
94 comprehensive treatment of all DSTs in all national applications, in our second task we have taken steps to
95 catalog other DSTs which use or may use, or which could contribute to, forecasts and projections of climate
96 and global change. The catalog is an exciting first step toward an ever-expanding inventory of existing and
97 emerging DSTs. The catalog can be maintained on-line for community input, expansion, and updating to
98 provide a focal point for information about the status of DSTs and how to access them.

99

100 The information in this report is largely from published literature and interviews with the sponsors and
101 stakeholders of the decision processes, as well as publications by and interviews with the producers of the
102 scientific information used in the tools.

103

104 **Our Case Studies**

105 We illustrate the following DSTs:

- 106 1. The Production Estimate and Crop Assessment Division and its Crop Condition Data Retrieval
107 and Evaluation (PECAD/CADRE) system of the US Department of Agriculture, Foreign
108 Agricultural Service (FAS). PECAD/CADRE is the world's most extensive and longest running

109 (over two decades) operational user of remote sensing for evaluation of worldwide agricultural
110 productivity.

111 2. The Community Multiscale Air Quality (CMAQ) modeling system of the US Environmental
112 Protection Agency (EPA). CMAQ is a widely used, US continental/regional/urban-scale air
113 quality decision support tool.

114 3. The Hybrid Optimization Model for Electric Renewables (HOMER), a micropower optimization
115 model of the US Department of Energy's National Renewable Energy Laboratory (NREL).
116 HOMER is used around the world to optimize deployment of renewable energy technologies.

117 4. Decision Support System to Prevent Lyme Disease (DDSPL) of the US Centers for Disease
118 Control and Prevention (CDC) and Yale University. DDSPL seeks to prevent the spread of the
119 most common vector-borne disease, Lyme disease, of which there are tens of thousands of cases
120 annually in the US

121 5. RiverWare, developed by the University of Colorado-Boulder's Center for Advanced Decision
122 Support for Water and Environmental Systems (CADSWES) in collaboration with the Bureau of
123 Reclamation, Tennessee Valley Authority, and the Army Corps of Engineers. RiverWare is a
124 hydrologic or river basin modeling system that integrates features of reservoir systems, such as
125 recreation, navigation, flood control, water quality, and water supply, in a basin management tool
126 with power system economics to provide basin managers and electric utilities a method of
127 planning, forecasting, and scheduling reservoir operations.

128

129 Taken together, these DSTs demonstrate a rich variety of applications of observations, data, forecasts, and
130 other predictions. In four of our studies, agricultural efficiency, air quality, water management, and energy
131 management, the DSTs have become well established as a basis for public policy decision making. In the
132 case of public health, our lead author points out reasons why direct applications of Earth observations to
133 public health have tended to lag behind these other applications and thus is a relatively new application
134 area. He also reminds us that management of air quality, agriculture, water, and energy—in and of

135 themselves—have implications for the quality of public health. The DST he selects is a new, emerging tool
136 intended to assist in prevention of the spread of infectious disease.

137

138 Our selection also varies in the geographic breadth of application, illustrating how users of these tools tailor
139 them to relevant regions of analysis and how, in some cases, the geographic coverage of the tools carries
140 over to their requirements for observations. For instance, PECAD/CADRE is used for worldwide study of
141 agricultural productivity and has data requirements of wide geographic scope, HOMER can be used for
142 renewable energy optimization throughout the world, and DDSPL focuses on the eastern, upper Midwest,
143 and West Coast portions of the US. CMAQ is used to predict air quality for the contiguous US as well as
144 regions and urban locales. RiverWare provides basin managers and electric utilities a method of planning,
145 forecasting, and scheduling reservoir operations.

146

147 With the exception of DDSPL, none of the DSTs we considered for potential selection, nor those we
148 discuss in this report, have to date made extensive use of climate change information or been used to study
149 the effect of a changing climate. However, in all cases, the developers and users of these DSTs fully
150 recognize their applicability to climate change science. In the discussion of the five DSTs presented in this
151 SAP, the authors describe how climate data and/or predictions might be used in these DSTs so that long-
152 range decisions and planning might be accomplished.

153

154 **Overview of the Chapters**

155 We next summarize the case studies. For each case study, we describe the DST and its data sources,
156 highlight potential uses as well as limits of the DSTs, note sources of uncertainty in using the tools, and
157 finally, discuss the link between the DST and climate change and variability. After our summary, we offer
158 general observations about similarities and differences among the studies.

159

160 *Agricultural Efficiency*: The Production Estimate and Crop Assessment Division (PECAD) of the US
161 Department of Agriculture, FAS is the world's most extensive and longest running operational user of
162 remote sensing data for evaluation of worldwide agricultural productivity. PECAD supports the FAS

163 mission to collect and analyze global crop intelligence information and provide periodic estimates used to
164 inform official USDA forecasts for the agricultural market, including farmers; agribusiness; commodity
165 traders and researchers; and federal, state, and local agencies. PECAD is often referred to as
166 PECAD/CADRE with one of its major automated components known as the Crop Condition Data Retrieval
167 and Evaluation (CADRE) geospatial database management system. Of all the DSTs we consider in this
168 report, CADRE has the oldest pedigree as the operational outcome of two early, experimental earth
169 observation projects during the 1970s and 1980s: the Large Area Crop Inventory Experiment (LACIE) and
170 the Agriculture and Resources Inventory Surveys through Aerospace Remote Sensing (AgRISTARS).

171

172 Sources of data for CADRE include a large number of weather and other earth observations from US,
173 European, Japanese, and commercial systems. PECAD combines these data with crop models, a variety of
174 GIS tools, and a large amount of contextual information, including official government reports, trade and
175 news sources, and on-the-ground reports from a global network of embassy attaches and regional analysts.

176

177 Potential future developments in PECAD/CADRE could include space-based observations of atmospheric
178 carbon dioxide (CO₂) measurements and measurement of global sea surface salinity to improve
179 understanding of the links between the water cycle, climate, and oceans. Other opportunities for enhancing
180 PECAD/CADRE include improvements in predictive modeling capabilities in weather and climate.

181

182 One of the largest technology gaps in meeting PECAD requirements is the practice of designing earth
183 observation systems for research rather than operational use, limiting the ability of PECAD/CADRE to rely
184 on data sources from non-operational systems. PECAD analysts require dependable inputs, implying use
185 of operational systems that ensure continuous data streams and that minimize vulnerability to component
186 failure through redundancy.

187

188 Sources of uncertainty can arise at each stage of analysis, from the accuracy of data inputs to the
189 assumptions in modeling. PECAD operators have been able to benchmark, validate, verify, and then
190 selectively incorporate additional data sources and automated decision tools by way of detailed engineering

191 reviews. Another aspect of resolving uncertainty in PECAD is the extensive use of a convergence
192 methodology to assimilate information from regional field analysts and other experts. This convergence of
193 evidence analysis seeks to reconcile various independent data sources to achieve a level of agreement to
194 minimize estimate error.

195

196 The relationship between climate and agriculture is complex, as agriculture is influenced not only by a
197 changing climate, but agricultural practices themselves are a contributory factor (e.g., in affecting land use
198 and influencing carbon fluxes. At present, PECAD is not directly used to address these dimensions of the
199 climate-agriculture interaction. However, many of the data inputs for PECAD are climate-related, thereby
200 enabling PECAD to inform understanding of agriculture as a “recipient” of climate-induced changes. For
201 instance, observing spatial and geographic trends in the output measures from PECAD can contribute to
202 understanding how the agricultural sector is responding to a changing climate. Likewise, trends in
203 PECAD’s measures of the composition and production of crops could shed light on the agricultural sector
204 as a “contributor” to climate change (for instance, in terms of greenhouse gas emissions or changes in soil
205 that may affect the potential for agricultural soil carbon sequestration). PECAD may also be influenced by,
206 as well as a barometer of climate-induced changes in land use, such as conversion from food production to
207 biomass fuel production.

208

209 ***Air Quality:*** The EPA CMAQ modeling system has been designed to approach air quality by including
210 state-of-the-science capabilities for modeling tropospheric ozone, fine particles, toxics, acid deposition, and
211 visibility degradation. CMAQ is used to guide the development of air quality regulations and standards and
212 to create state implementation plans for managing air emissions. CMAQ also can be used to evaluate
213 longer-term as well as short-term transport from localized sources and to perform simulations using
214 downscaled regional climate from global climate change scenarios.

215

216 The CMAQ modeling system contains three types of modeling components: a meteorological modeling
217 system for the description of atmospheric states and motions, emission models for man-made and natural
218 emissions that are injected into the atmosphere, and a chemistry-transport modeling system for simulation

219 of the chemical transformation and fate. Inputs for CMAQ, and their associated regional meteorological
220 model, mesoscale model version 5 (MM5), can include, but are not limited to, the comprehensive output
221 from a general circulation model, anthropogenic and biogenic emissions, description of wildland fires, land
222 use and demographic changes, and meteorological and atmospheric chemical species measurements by *in-*
223 *situ* and remote sensing platforms, including satellites and aircraft.

224

225 CMAQ can be used to study questions such as: How will present and future emission changes affect
226 attainment of air quality standards? Will present and future emissions and/or climate/meteorological
227 changes affect the frequency and magnitudes of high pollution events? How will land use changes due to
228 urbanization and global warming affect air quality? How does long-range air pollution from other regions
229 affect US air quality? How will changes in the long-range transport due to the climate change affect air
230 quality? How does wildland fire affect air quality and will climate change affect wildland fire and
231 subsequently air quality? How sensitive are the air quality predictions to changes in both anthropogenic and
232 biogenic emissions?

233

234 **Energy Management:** HOMER is a micropower optimization model of the US Department of Energy's
235 NREL. HOMER is able to calculate emission reductions enabled by replacing diesel-generating systems
236 with renewable energy systems in a micro-grid or grid-connected configuration. HOMER helps the user
237 design grid-connected and off-grid renewable energy systems by performing a wide range of design
238 scenarios. HOMER can be used to address questions such as: Which technologies are most cost-effective?
239 What happens to the economics if the project's costs or loads change? Is the renewable energy resource
240 adequate for the different technologies being considered to meet the load? HOMER does this by finding the
241 least-cost combination of components that meet electrical and thermal loads.

242

243 The earth observation information serving as input to HOMER is centered on wind and solar resource
244 assessments derived from a variety of sources. Wind data include surface and upper air station data,
245 satellite-derived ocean and ship wind data, and digital terrain and land cover data. Solar resource data

246 include surface cloud, radiation, aerosol optical depth, and digital terrain and land cover data from both *in-*
247 *situ* and remote sensing sources.

248

249 All of the input data for HOMER can have a level of uncertainty attached to them. HOMER allows the user
250 to perform sensitivity tests on one or more variables and has graphical capabilities to display these results
251 to inform decision makers. As a general rule, the error in estimating the performance of a renewable energy
252 system over a year is roughly linear to the error in the input resource data.

253

254 One of the largest challenges in HOMER is the absence of direct or *in-situ* solar and wind resource
255 measurements at specific locations to which HOMER is applied. In addition, in many cases, values are not
256 based on direct measurement at all but are approximations based on the use of algorithms to convert a
257 signal into the parameter of interest as is the case with most satellite-derived data products. For example,
258 satellite-derived ocean wind data are not based on direct observation of the wind speed above the ocean
259 surface but from an algorithm that infers wind speed based on wave height observations. Observations of
260 aerosol optical depths (for which considerable research is underway) can be complicated by irregular land-
261 surface features that complicate the application of algorithms for satellite-derived measures.

262

263 For renewable energy resource mapping, improved observations of key weather parameters (for instance,
264 wind speed and direction at various heights above the ground, particularly at the hub height of wind energy
265 turbine systems, and over the open oceans at higher and higher spatial resolutions, and improved ways of
266 differentiating snow cover and bright reflecting surfaces from clouds) will be of value to the renewable
267 energy community. New, more accurate methods of related parameters, such as aerosol optical depth,
268 would also improve the resource data.

269

270 The relationship between HOMER and global change information is largely by way of the dependence of
271 renewable energy resource input measurements on weather and local climate conditions. Although
272 HOMER was not designed to be a climate-related management decision-making tool, by optimizing the
273 mix of hybrid renewable energy technologies for meeting load conditions, HOMER also enables users to

274 respond to climate change and variability in their energy management decisions. HOMER could be used to
275 evaluate how renewable energy systems can be used cost-effectively to displace fossil-fuel-based systems.
276

277 **Public Health:** The DDSPL is operated by the US CDC and Yale University to address questions related to
278 the likely distribution of Lyme disease east of the 100th meridian, where most cases occur. Lyme disease is
279 the most common vector-borne disease in the US, with tens of thousands of cases annually. Most human
280 cases occur in the Eastern and upper Midwest portions of the US, although there is a secondary focus along
281 the West Coast. Vector-borne diseases are those in which parasites are transmitted among people or from
282 wildlife to people by insects or arthropods (as vectors, they do not themselves cause disease). The black-
283 legged tick is typically the carrier of the bacteria causing Lyme disease.

284

285 Early demonstrations during the 1980s showed the utility of earth observations for identifying locations and
286 times that vector-borne diseases were likely to occur, but growth of applications has been comparatively
287 slow. Earth observing instruments have not been designed to monitor disease risk; rather, data gathered
288 from these platforms are “scavenged” for public health risk assessment. DDSPL uses satellite data and
289 derived products, such as land cover together with meteorological data and census data, to characterize
290 statistical predictors of the presence of black-legged ticks. The model is validated by field surveys. The
291 DDSPL is thus a means of setting priorities for the likely geographic extent of the vector; the tool does not
292 at present characterize the risk of disease in the human population.

293

294 Future use of DDSPL partly depends on whether the goal of disease prevention or the goal of treatment
295 drives public health policy decisions. In addition, studies have shown that communication to the public
296 about the risk in regions with Lyme disease often fails to reduce the likelihood of infection. Use of the
297 DDSPL is also limited by restrictions on the dissemination of detailed information on the distribution of
298 human disease. The role of improved Earth science data is unclear in terms of improving the performance
299 of DDSPL because at present the system has a level of accuracy deemed “highly satisfactory.” Future use
300 may instead require a model of sociological/behavioral influences among the population.

301

302 Standard statistical models and in-field validation are used to assess the uncertainty in decision making
303 with DDSPL. The accuracy of clinical diagnoses also influences the ultimate usefulness of DDSPL as an
304 indicator tool to characterize the geographic extent of the vectors.

305

306 The DDSPL is one of the few public health DSTs that has explicitly evaluated the effects of climate
307 variability. Using outputs of a Canadian climate change model, study has shown that with warming global
308 mean temperatures by the year 2050 to 2080 the geographic range of the tick vector will decrease at first,
309 with reduced presence in the southern boundary, and then expand into Canada and the central region of
310 North America where it now absent. The range also moves away from population concentrations.

311

312 **Water Management:** RiverWare was developed and is maintained by CADSWES in collaboration with the
313 Bureau of Reclamation, Tennessee Valley Authority, and the Army Corps of Engineers. It is a river basin
314 modeling system that integrates features of reservoir systems, such as recreation, navigation, flood control,
315 water quality, and water supply in a basin management tool, with power system economics to provide basin
316 managers and electric utilities a method of planning, forecasting, and scheduling reservoir operations.
317 RiverWare uses an object-oriented software engineering approach in model development. The object
318 oriented software-modeling strategy allows computational methods for new processes, additional
319 controllers for providing new solution algorithms, and additional objects for modeling new features to be
320 added easily to the modeling system. RiverWare is data intensive in that a specific river/reservoir system
321 and its operating policies must be characterized by the data supplied to the model. This allows the models
322 to be modified as new features are added to the river/reservoir system and/or new operating policies are
323 introduced. The data-intensive feature allows the model to be used for water management in most river
324 basins.

325

326 Riverware is menu driven through a graphical user interface (GUI). The basin topology is developed
327 through the selection of a reservoir, reach, confluence, and other necessary objects and by entering the data
328 associated with each object manually or through importing files. Utilities within RiverWare provide a
329 means to automatically execute many simulations, to access data from external sources, and to export

330 model results. Users also define operating policies through the GUI as system constraints or rules for
331 achieving system management goals (e.g., related to flood control, water supplies, water quality,
332 navigation, recreation, and power generation). The direct use of earth observations in RiverWare is limited.
333 Unlike traditional hydrologic models that track the transformation of precipitation (e.g., rain and snow) into
334 soil moisture and streamflow, RiverWare uses supplies of water to the system as input data. These data are
335 derived from a hydrologic model where direct use of earth observations can be and have been made.
336 Application of RiverWare is limited by the specific implementation defined by the user and by the quality
337 of the input data. It has tremendous flexibility in the kinds of data it can use, but long records of data are
338 required to overcome the issue of data non-stationarity.

339

340 The specific application of RiverWare in the context of mid- or long-range planning for a specific river
341 basin will reflect whether decisions may rely on global change information. For mid-range planning of
342 reservoir operations, characterization and projections of interannual and decadal-scale climate variability
343 (e.g., monitoring, understanding, and predicting interannual climate phenomena such as the El Nino-
344 Southern Oscillation) are important. For long-term planning, global warming has moved from the realm of
345 speculation to general acceptance. The impacts of global warming on water resources, and their
346 implications for management, have been a major focus in the assessments of climate change. The estimates
347 of potential impacts of climate change on precipitation have been mixed, leading to increasing uncertainty
348 about the reliability of future water supplies.

349

350 **General Observations**

351 Application of all of the DSTs involves a variety of input data types, all of which have some degree of
352 uncertainty in terms of their accuracy. The amount of uncertainty associated with resource data can depend
353 heavily on how the data are obtained. Quality *in-situ* measurements of wind and solar data suitable for
354 application in HOMER are can have uncertainties of less than $\pm 3\%$ of true value; however, when
355 estimation methods are required, such as the use of earth observations, modeling, and empirical techniques,
356 uncertainties can be as much as $\pm 10\%$ or more. The DSTs address uncertainty by allowing users to
357 perform sensitivity tests on variables. With the exception of HOMER, a significant amount of additional

358 traditional on-the-ground reports are a critical component. In the case of PECAD/CADRE, uncertainty is
359 resolved in part by extensive use of a convergence methodology to assimilate information from regional
360 field analysts and other experts. This brings a large amount of additional information to PECAD/CADRE
361 forecasts, well beyond the automated outputs of DSTs. In RiverWare, streamflow and other hydrologic
362 variables respond to atmospheric factors such as precipitation, and obtaining quality precipitation estimates
363 is a formidable challenge, especially in the western US where orographic effects produce large spatial
364 variability and where there is a scarcity of real-time precipitation observations and poor radar coverage.

365

366 In terms of their current or prospective use of climate change predictions or forecasts as *DST inputs*, or the
367 contributions of *DST outputs* to understanding, monitoring, and responding to a changing climate, the
368 status is mixed. DDSPL is one of the few public health decision support tools that has explicitly evaluated
369 the potential impact of climate change scenarios on an infectious disease system. None of the other DSTs at
370 present is directly integrated with climate change measurements, but all of them can and may in the future
371 take this step. PECAD/CADRE's assessment of global agricultural production will certainly be influenced
372 by observations and forecasts of climate change and variability as model inputs, just as the response of the
373 agricultural sector to a changing climate will feedback into PECAD/CADRE production estimates.

374 HOMER's renewable energy optimization calculations will be directly affected by climate-related changes
375 in renewable energy resource supplies and will enhance our ability to adapt to climate-induced changes in
376 energy management and forecasting. Air quality will definitely be affected by global climate change. The
377 ability of CMAQ to predict those affects is conditional on acquiring accurate predictions of the
378 meteorology under the climate change conditions that will take place in the US and accurate emission
379 scenarios for the future. Given these inputs to CMAQ, reliable predictions of the air quality and their
380 subsequent health affects can be ascertained. It was noted that there is great difficulty in integrating climate
381 change information into RiverWare and other such water management models. The multiplicity of
382 scenarios and vague attribution of their probability for occurrence, which depends on feedback among
383 social, economic, political, technological, and physical processes, complicates conceptual integration of
384 climate change impacts assessment results in a practical water management context. Furthermore, the

385 century timescales of climate change exceed typical planning and infrastructure design horizons in water
386 management.

387

388 **Audience and Intended Use**

389 The *CCSP SAP 5.1 Prospectus* describes the audience and intended use of this report:

390

391 This synthesis and assessment report is designed to serve decision makers and
392 stakeholder communities interested in using global change information resources in
393 policy, planning, and other practical uses. The goal is to provide useful information on
394 climate change research products that have the capacity to inform decision processes. The
395 report will also be valuable to the climate change science community because it will
396 indicate types of information generated through the processes of observation and research
397 that are particularly valuable for decision support. In addition, the report will be useful
398 for shaping the future development and evaluation of decision-support activities,
399 particularly with regard to improving the interactions with users and potential users.

400

401 There are a number of national and international programs focusing on the use of Earth
402 observations and related prediction capacity to inform decision support tools (see
403 Table 3, “Related National and International Activities”). These programs both inform
404 and are informed by the CCSP and are recognized in the development of this product.

405 (*CCSP Synthesis and Assessment Product 5.1, Prospectus for “Uses and Limitations of*
406 *Observations, Data, Forecasts, and Other Projections in Decision Support for Selected*
407 *Sectors and Regions,” 28 February 2006)*

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