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1103 **DELIVERABLE ONE: REPORT ON THE DST(S) FOR YOUR NATIONAL APPLICATIONS AREA**

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**Climate Change Science Program**

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1108 **Uses and Limitations of Observations, Data, Forecasts, and Other Projections in**

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**Decision Support for Selected Sectors and Regions**

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1111 **Chapter 3. “Decision Support System for Assessing Hybrid Renewable Energy Systems”**

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**David S. Renné**

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**National Renewable Energy Laboratory**

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**1617 Cole Boulevard**

1117

**Golden, CO 80401-3393**

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1119 **1. Introduction**

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The national application area addressed in this chapter is the deployment of renewable energy technologies.

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Renewable energy technologies are being used around the world to meet local energy loads, to supplement grid-wind

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electricity supply, to perform mechanical work such as water pumping, to provide fuels for transportation, to provide

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hot water for buildings, and to support heating and cooling requirements for building energy design. Numerous

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organizations and research institutions around the world have developed a variety of decision-support tools to address

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how these technologies might perform in a most cost-effective manner to address specific applications. This chapter

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will focus on one specific tool, known as the Micropower Optimization Model, Hybrid Optimization Model for Electric

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Renewables (HOMER), that has been under consistent development and improvement at the US Department of

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Energy’s National Renewable Energy Laboratory and is used extensively around the world.

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1130 HOMER relies heavily on knowledge of the renewable energy resource available to the technologies being analyzed.  
1131 Renewable energy resources, particularly for solar and wind technologies, are highly dependent on weather and climate  
1132 phenomena, and are also driven by local microclimatic processes. Given the absence of a sufficiently dense ground  
1133 network of reliable solar and wind observations, we must rely on validated numerical models, empirical knowledge of  
1134 microscale weather characteristics, and collateral (indirect) observations derived from Earth observations, such as  
1135 reanalysis data and satellite-borne remote sensors, to develop reliable knowledge of the geospatial characteristics and  
1136 extent of these resources. Thus, the Decision Support System (DSS) described in this chapter includes HOMER as an  
1137 end-use application and is described in the context of the renewable energy resource information required as input, as  
1138 well as some intermediate steps that can be taken to organize these data, using Geographic Information Systems (GIS)  
1139 software, to facilitate the application of HOMER.

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1143 2a. Description of the HOMER DSS

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1145 The HOMER DSS described in this chapter consists of three main components: (1) the renewable energy resource  
1146 information required to estimate technology performance and operational characteristics, (2) (optional) organization of  
1147 the resource data into a GIS framework so that the data can be easily imported into the decision support tool, and (3)  
1148 NREL's Micropower Optimization Model known as HOMER, which ingests the renewable resource data for  
1149 determining the optimal mix of power technologies for meeting specified load conditions at specified locations. This  
1150 section describes each of these components separately. Although climate-based Earth observational data are primarily  
1151 relevant only to the first component, some related Earth observation information could also be associated with the  
1152 second and even the third component. Furthermore, it will be apparent that the first component is of major importance  
1153 in the successful use of the HOMER DSS.

1154

1155 Although HOMER handles a number of power technologies, we will focus our attention in this chapter on solar and  
1156 wind technologies and the resources required to run these technologies.

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1159 Solar and Wind Resource Assessments

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1161 The first component of the HOMER DSS is properly formatted, reliable renewable energy resource data. The significant  
1162 data requirements for this component are time-dependent measurements of wind and solar resources, as well as Earth  
1163 observational data and data from numerical models, to provide the necessary spatial information for these resources,  
1164 which can vary significantly over relatively small distances due to local microclimatic effects. Because of this natural  
1165 variability, it is necessary to examine these energy resources geospatially in order to determine optimal siting of  
1166 renewable energy technologies; alternatively, if a renewable energy technology is sited at a specific site in order to meet  
1167 a nearby load requirement (such as a solar home system), it is necessary to know what the resource availability is at that  
1168 location, since microclimatic variability may make even nearby data sources irrelevant.

1169

1170 Examples of the products derived from the methodologies described below can be found for many areas around the  
1171 world. One significant project that has recently been completed is the Solar and Wind Energy Resource Assessment  
1172 (SWERA) Project, which provided high-resolution wind and solar resource maps for 13 countries around the world.  
1173 SWERA was a project funded by the Global Environment Facility and was cost-shared by several technical  
1174 organizations around the world: NREL; the State University of New York at Albany, the NASA's Langley Research  
1175 Center, and the USGS/EROS Data Center in the U.S.; Riso National Laboratory in Denmark; the German Aerospace  
1176 Institute (DLR); the Energy Resources Institute (New Delhi, India); and the Brazilian Spatial Institute (INPE) in Sao  
1177 Jose dos Campos, Brazil. The United Nations Environment Programme (UNEP) managed the project. Besides the  
1178 solar and wind resource maps and underlying data sets, a variety of other relevant data products came out of this  
1179 program. All of the final products and data can be found on the SWERA archive, hosted at the UNEP/GRID site,  
1180 collocated with the USGS/EROS data center in Sioux Falls, South Dakota (<http://swera.unep.net>).

1181

1182 For wind resource assessments, NREL's approach, known as Wind Resource Assessment Mapping System (WRAMS)  
1183 relies on mesoscale numerical models such as MM5 or Weather Research and Forecasting (WRF), which can provide  
1184 simulations of near-surface wind flow characteristics in complex terrain or where sharp temperature gradients might  
1185 exist (such as land-sea contrasts). Typically, these numerical models use available weather data, such as the National

1186 Climatic Data Center’s Integrated Surface Hourly (ISH) data network and National Center for Atmospheric Research-  
1187 National Centers for Environmental Protection (NCAR-NCEP) reanalysis data as inputs. In coastal areas or island  
1188 situations NREL’s wind resource mapping also relies heavily on SeaWinds data from the Quikscat satellite to obtain  
1189 near-shore and near-island wind resources. WRAMS also relies on Global Land Cover Characterization (GLCC) 1-km  
1190 and Regional Gap Analysis Program (ReGAP) 200-m land cover data, as well as Moderate Resolution Imaging  
1191 Spectroradiometer (MODIS) data from the Aqua and Terra Earth Observation System satellites, to obtain information  
1192 such as percent of tree cover and other land use information. This information is used not only to determine roughness  
1193 lengths in the numerical mesoscale models but also to screen sites suitable for both wind and solar development in the  
1194 second component of the HOMER DSS.

1195  
1196 The numerical models are typically run at a 2.5-km resolution. However, wind resource information is often reported at  
1197 the highest resolution at which a digital elevation model (DEM) can provide. Globally this has traditionally been 1-km  
1198 resolution; however, in some cases in the US 400-m DEM data are available. Furthermore, the Shuttle Radar Topology  
1199 Mission has now been able to provide users with a  
1200 90-m DEM for much of the world. Thus, additional steps are needed beyond the 2.5-km resolution model output to  
1201 depict wind resources at the higher resolutions offered by these DEM’s. This can be accomplished by using a  
1202 secondary high-resolution mesoscale model, empirical methods, or both. For example, with NREL’s WRAMS  
1203 methodology, GISD-based empirical modeling tools have been developed to modify results from the numerical models  
1204 that appear to have provided unreliable results in complex-terrain areas.

1205  
1206 The numerical models generally provide outputs at multiple levels above the ground. The WRAMS methodology  
1207 provides values at a single specified height above the ground, nominally 50 m, or near the hub-height of modern-day  
1208 large wind turbines (although with the recent advent of larger and larger wind turbines, hub heights are approaching 100  
1209 m, so this standard height designation is changing). Where measured data are used to assess wind resources, a simple  
1210 “power law” relationship is used to extrapolate the measured data to the desired height (Elliott *et al.*, 1987), i.e.

1211  
1212 
$$V_R/V_a = (Z_R/Z_a)^\alpha \quad (1)$$

1213

1214 where  $\alpha$ , the power law coefficient, is normally assumed to be  $1/7$ ,  $V_R$  is the wind speed at reference height  $Z_R$   
1215 (nominally, 50 m), and  $V_a$  is the wind speed at the measurement height  $Z_a$ .

1216

1217 The output of the WRAMS methodology is typically a value of wind power density at every grid-cell representative of  
1218 an annual average (in order to produce monthly values, the procedure outlined above would have to be repeated for each  
1219 month of the year). For mapping purposes, a classification scheme has been set up that relates a “wind power class” to  
1220 a range of wind power densities. The classification scheme ranges from 1 to  $>7$ , and applies to a specific height above  
1221 ground. Normally, for grid-connected applications, a wind power class of 4 or above is best, while for small wind  
1222 turbine applications where machines can operate in lower wind speeds, wind power class of 3 or above is suitable. Of  
1223 course, the wind maps are not intended to identify sites at which large wind turbines can be installed, but rather are  
1224 intended to provide information to developers on where they might most effectively install wind measurement systems  
1225 for further site assessment. The maps also provide a useful tool for policy makers to obtain reliable estimates on the  
1226 total wind energy potential for a region.

1227

1228 Other well-known approaches besides NREL’s WRAMS methodology are also used to produce large-area wind  
1229 resource mapping. For example, Riso National Laboratory calculates wind speeds within 200 m above the Earth’s  
1230 surface using the Karlsruhe Atmospheric Mesoscale Model (KAMM). Although KAMM also uses NCEP/NCAR  
1231 reanalysis data, the model is based on large-scale geostrophic winds, and simulations are performed for classes of  
1232 different geostrophic wind. The classes are weighted with their frequency to obtain statistics for the simulated winds.  
1233 The results can then be treated as similar to real observations to make wind atlas files for the Wind Atlas Analysis and  
1234 Application Program (WASP), which are employed to predict local winds at a much higher resolution than KAMM can  
1235 provide. WASP calculations are based on wind data measured or simulated at specific locations and includes a complex  
1236 terrain flow model, a roughness change model, and a model for sheltering obstacles. More on WASP can be found at  
1237 <http://www.wasp.dk/>.

1238

1239 Due to the scarcity of high-quality, ground-based solar resource measurements, large-area solar resource assessments in  
1240 the US have historically relied on the analysis of surface National Weather Service cloud cover observations. These  
1241 observations are far more ubiquitous than solar measurements, and allowed NREL to develop a 1961 to 1990 National

1242 Solar Radiation Database for 239 surface sites. However, more recently in the US more and more reliance has been  
1243 placed on GOES visible channel data to obtain surface reflectance information that can be used to derive high-resolution  
1244 (~10-km) site-time specific solar resource data (see for example Perez, *et al.*, 2002). In fact, this approach has become  
1245 commonplace in Europe, using Meteosat data. And the NASA Langley Research Center has recently completed a 20-  
1246 year worldwide 100-km resolution Surface Solar Energy Data set derived from International Satellite Cloud  
1247 Climatology Project data, which is derived from data collected by all of the Earth's geostationary and polar orbiting  
1248 satellites (<http://eosweb.larc.nasa.gov/sse>).

1249  
1250 The use of satellite imagery for estimating surface solar resource characteristics over large areas has been studied for  
1251 some years, and Renné *et al.* (1999) published a summary of approaches developed around the world. These satellite-  
1252 derived assessments require good knowledge of the aerosol optical depth over time and space, which can be obtained in  
1253 part from MODIS and Advanced Very High Resolution Radiometer (AVHRR) data from polar orbiting environmental  
1254 satellites. The assessments provide information both on Global Horizontal Irradiance (GHI), which is useful for  
1255 estimating resources available to flat plate collectors such as photovoltaic panels or solar water heating systems, and  
1256 Direct Normal Irradiance (DNI), which is needed for determining the resources available to solar concentrators that  
1257 track the sun.

1258  
1259 Besides NREL and NASA, other organizations perform similar types of high-resolution solar resource data sets. For  
1260 example, the German Space Agency (DLR) has been applying similar methods to Meteosat data for developing solar  
1261 resource maps and data for Europe and northern Africa. DLR was also involved in the SWERA project and applied  
1262 their methodologies to several SWERA countries.

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#### 1265 Geospatial Toolkit

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1267 Recently, NREL has begun to format the solar and wind resource information into GIS software-compatible formats,  
1268 and has incorporated this information, along with other geospatial data relevant to renewable energy development, into  
1269 a Geospatial Toolkit (GsT). The GsT is a stand-alone, downloadable, and executable software package that allows the

1270 user to overlay the wind and solar data with other geospatial data sets available for the region, such as transmission  
1271 lines, transportation corridors, population (load) centers, locations of power plant facilities and substations, land use and  
1272 land form data, terrain data, etc. Not only can the user overlay various data sets of their choosing, there are also simple  
1273 queries built into the toolkit, such as the amount of “windy” land (e.g. Class 3 and above) available within a distance of  
1274 10-km of all transmission lines (minus specified exclusion areas, such as protected lands). The GsT developed at NREL  
1275 makes use of the Environmental Science and Research Institute’s (ESRI) MapObjects software, although other  
1276 platforms, including on-line, Web-based platforms, could also be used.

1277  
1278 In a sense, the GsT is a DSS, since it allows the user to manipulate resource information with other critical data relevant  
1279 to the deployment of renewable energy technologies to assist decision makers in identifying and conducting preliminary  
1280 assessments of possible sites for installing these systems and supporting renewable energy policy decisions. However,  
1281 up to now NREL has only prepared GsT’s for a few locations: the countries of Sri Lanka, Afghanistan, and Pakistan;  
1282 Hebei Province in China; the state of Oaxaca in Mexico; and the state of Nevada in the US. By the time of publication  
1283 of this chapter, additional toolkits may also be available. As with the resource data, all toolkits developed by NREL are  
1284 available for download from NREL’s Web site. Those toolkits developed under the SWERA project are also available  
1285 from the SWERA Web site.

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#### 1287 HOMER: NREL’s Micropower Optimization Model

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1289 The primary decision support tool that makes up the DSS being described here is HOMER, NREL’s Micropower  
1290 Optimization Model. HOMER is a computer model that simplifies the task of evaluating design options for both off-  
1291 grid and grid-connected power systems for remote, stand-alone, and distributed generation applications. HOMER’s  
1292 optimization and sensitivity analysis algorithms allow the user to evaluate the economic and technical feasibility of a  
1293 large number of technology options and to account for variation in technology costs and energy resource availability.  
1294 HOMER can also address system component sizing and the adequacy of the available renewable energy resource.  
1295 HOMER models both conventional and renewable energy technologies:

#### 1296 **Power sources:**

- 1297 • solar photovoltaic

- 1298 • wind turbine
- 1299 • run-of-river hydropower
- 1300 • Generator: diesel, gasoline, biogas, alternative and custom fuels, co-fired
- 1301 • electric utility grid
- 1302 • microturbine
- 1303 • fuel cell
- 1304 **Storage:**
- 1305 • battery bank
- 1306 • hydrogen
- 1307 **Loads:**
- 1308 • daily profiles with seasonal variation
- 1309 • deferrable (e.g., water pumping and refrigeration)
- 1310 • thermal (e.g., space heating and crop drying)
- 1311 • efficiency measures

1312

1313 In order to find the least cost combination of components that meet electrical and thermal loads, HOMER simulates  
1314 thousands of system configurations, optimizes for lifecycle costs, and generates results of sensitivity analyses on most  
1315 inputs. HOMER simulates the operation of each technology being examined by making energy balance calculations for  
1316 each of the 8,760 hours in a year. For each hour, HOMER compares the electric and thermal load in the hour to the  
1317 energy that the system can supply in that hour. For systems that include batteries or fuel-powered generators, HOMER  
1318 also decides for each hour how to operate the generators and whether to charge or discharge the batteries. If the system  
1319 meets the loads for the entire year, HOMER estimates the lifecycle cost of the system, accounting for the capital,  
1320 replacement, operation and maintenance, and fuel and interest costs. The user can obtain screen views of hourly energy  
1321 flows for each component as well as annual costs and performance summaries.

1322

1323 This and other information about HOMER are available on NREL's Web site: <http://www.nrel.gov/homer/>. The Web  
1324 site also provides extensive examples of how HOMER is used around the world to evaluate optimized hybrid renewable



1325 power systems to meet load requirements in remote villages. Figure 1 shows a typical example of an output graphic  
1326 available from HOMER.

1327

1328 In order to accomplish these tasks, HOMER requires information on the hourly renewable energy resources available to  
1329 the technologies being studied. However, typically hour-by-hour wind and solar data are not available for most sites.  
1330 Thus, the user is requested to provide monthly or average information on solar and wind resources; HOMER then uses  
1331 an internal weather generator to provide the best estimate of a simulated hour-by-hour data set, taking into consideration  
1332 diurnal variability if the user can provide an indication of what this should be. However, these approximations  
1333 represent a source of uncertainty in the model. For those locations where a GsT is available, the GsT offers a  
1334 mechanism for the user to easily ingest data from the toolkit into HOMER for the specific location of interest. However,  
1335 since the toolkit contains only monthly solar and wind data, the limitations described above still apply. More  
1336 information on the weather generator can be found in the HOMER Help files.

1337

1338 The HOMER developers have implemented various methods to facilitate access to reliable resource data that provide  
1339 some of the input for simulations. For example, a direct link with the NASA SSE data site enables the user to download  
1340 monthly and annual solar data from any location on Earth. The 100-km resolution NASA data have become a  
1341 benchmark of solar resource information, due to the high quality of the modeling capability used to generate the data,  
1342 the fact that the SSE is validated against numerous ground stations, and the fact that it is global in scope and now covers  
1343 a 20-year period. However, the data set is still limited by a somewhat coarse resolution and no validation in areas where  
1344 ground data do not exist. The procedures used to generate the SSE also have problems where land-ocean interfaces  
1345 occur, and in snow-covered areas.

1346

1347 Linking HOMER to higher-resolution regional solar data sets would likely improve these uncertainties somewhat, but in  
1348 general these data sets are also limited to monthly and seasonal values. However, since these methods rely on  
1349 geostationary satellite data that provide frequent imagery of the Earth's surface, an opportunity exists to produce hourly  
1350 time series data for up to several years at a 10-km resolution. This option will require significant data storage and  
1351 retrieval capabilities on a server, but such a possibility now exists for future assessments.

1352

1353 Wind data available to HOMER is also generally limited to annual and at best monthly values. The standard HOMER  
1354 interface allows the user to also designate a Weibull “k” value if this information is available. The Weibull k is a  
1355 statistical means of defining the frequency distribution of the long-term hourly wind speeds at a location; this value can  
1356 vary substantially depending on local terrain and microclimatic conditions. HOMER also has a provision for the user to  
1357 designate the diurnal range of wind speeds and the timing when maximum and minimum winds occur. This  
1358 information then provides improved simulation of the hour-by-hour wind values. The difficulty is that there may be  
1359 applications where even these statistical values are not known to the user and are not available from the standard wind  
1360 resource maps produced for a region, but this limitation may not be critical and requires further study to determine the  
1361 impact on model output uncertainties.

1362

#### 1363 2b. Access to the HOMER DSS

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1365 HOMER was originally developed and has always been maintained by the National Renewable Energy Laboratory.  
1366 The model can be downloaded free of charge from NREL’s Web site at <http://www.nrel.gov/homer/default.asp>. The  
1367 user is required to register, and registration must be updated every six months. The Web site also contains a variety of  
1368 guides for getting started and using the software.

1369

1370 Resource information required as input to HOMER is generally freely available at the Web sites of the institutions  
1371 developing the data. These institutions also generally maintain and continuously update the data. For example,  
1372 renewable energy resource information can be found in several places on NREL’s Web site, such as <http://rredc.nrel.gov>  
1373 or [www.nrel.gov/GIS](http://www.nrel.gov/GIS). NASA solar energy data, which can be easily input to HOMER, is available at  
1374 <http://eosweb.larc.nasa.gov/sse>. In fact, there is a specific feature built into HOMER that automatically accesses and  
1375 inputs the SSE data for the specific location that the model is analyzing. Wind and solar resource data for the 13  
1376 SWERA countries can be found at <http://unep.swera.net>. This Web site is currently undergoing expansion and  
1377 upgrading by the USGS/EROS Data Center in Sioux Falls, SD, and will eventually become a major clearing house for  
1378 resource data from around the world in formats that can be readily ingested into tools such as HOMER.

1379

#### 1380 2c. Definition of HOMER information requirements

1381

1382 The ideal input data format to HOMER is an hourly time series of wind and solar resource data covering a complete  
1383 year (8,760 values). In addition, the wind data should be representative of the wind turbine hub height that is being  
1384 analyzed within HOMER. Unfortunately data sets such as these are seldom available at the specific locations for which  
1385 HOMER is being applied. More typically, the HOMER user will have to identify input data sets from resource maps  
1386 (even within the GsT, the resource data are based on what is incorporated into the map, which, in the case of wind, may  
1387 represent only a single annual value). Because monthly and annual mean data are more typically available, HOMER  
1388 has been designed to take monthly mean wind speeds (in m/s) and monthly mean solar resource values (in kw-h/m<sup>2</sup>-  
1389 day). In the case of wind, HOMER also allows for the specification of other statistical parameters related to wind speed  
1390 distributions and diurnal characteristics. Furthermore, if the wind data available for input to HOMER do not represent  
1391 the same height above the ground as the wind turbine's hub height being analyzed, HOMER has internal algorithms to  
1392 adjust for this. The user must specify the height above the ground for which the data represent, and a power law  
1393 conversion adjusts the wind speed value to the hub height of the specific wind turbine being analyzed. HOMER then  
1394 utilizes an internal weather generator that takes the input information and creates an hour-by-hour data profile  
1395 representing a one-year data file. Then, HOMER calculates turbine energy output by converting each hourly value to  
1396 the energy production of the machine using the manufacturer's turbine power curve.

1397

1398 Besides the mean monthly wind speeds, the statistical parameters required by HOMER to generate the hourly data sets  
1399 include the following:

1400

- 1401 • The altitude above sea level (to adjust for air density, since turbine performance is typically rated at sea level);
- 1402 • The Weibull k value, which typically ranges from 1.5 to 2.5, depending on terrain type;
- 1403 • An auto-correlation factor, which is a measure of how strongly the wind speed in 1 hour depends (on average)  
1404 on the wind speed in the previous hour (these values typically range from 0.85 to 0.90);
- 1405 • A diurnal pattern strength, which is a measure of how strongly the wind speed depends on the time of day  
1406 (values are typically 0.0 to 0.4); and
- 1407 • The hour of the peak wind speed (over land areas this is typically 1400 to 1600 local time)

1408

1409 In the US as elsewhere, wind resource maps often depict the resource in terms of wind power density, in units of watts-  
 1410  $m^{-2}$  rather than in wind speeds. In this case, the wind power density must be converted back to a mean wind speed. The  
 1411 relationship between wind power density (P) and wind speed (v) is given as follows:

1412

$$1413 \quad P = \frac{1}{2}\rho \sum_i v_i^3, \quad (2)$$

1414

1415 where  $\rho$  is the density of the air and  $i$  is the individual hourly wind observation. Since the frequency distribution of  
 1416 wind speed over the period of a year or so follows a Weibull distribution shape, the wind power density can be  
 1417 converted back to a wind speed if the “k” factor in the Weibull distribution is known, as well as the height above sea  
 1418 level of the site (to determine the air density).

1419

1420 2d. Access to and use of the HOMER DSS among the federal, state, and local levels

1421

1422 Because of the easy access to HOMER and to the related resource assessment data products, the HOMER DSS is freely  
 1423 available to all government and private entities in the US and worldwide. Thousands of users from all economic sectors  
 1424 are using HOMER to evaluate renewable energy technology applications, particularly for off-grid use.

1425

1426 2e. Variation of the HOMER DSS by geographic region or characteristic

1427

1428 A key feature of HOMER is the evaluation of specific renewable energy technologies and related energy systems for  
 1429 different regions and for different applications. The HOMER model contains information on renewable energy  
 1430 technology characteristics; however, these characteristics, such as power curves of difference wind turbine models,  
 1431 generator fuel curves, and other factors are not affected by location. Because of the location-specific dependency of  
 1432 resource data, use of data that is not representative of the specific region of analysis will introduce additional  
 1433 uncertainties in the model results. Thus, the user should evaluate the accuracy and relevancy of any default information  
 1434 that is built into HOMER, or any resource data chosen as input to HOMER before completing the final analyses.

1435

1436 **3. Observations used by the HOMER DSS now and of potential use in the future**

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1438 This section focuses on the Earth observations (of all types, from remote sensing and *in situ*) used or of potential use in  
1439 the HOMER DSS.

1440

1441 3a. Kinds of observations being used

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1443 In the previous section we provided a description of the renewable energy resource assessment related to solar and wind  
1444 technologies that are required as input to HOMER when these technologies are being modeled. As noted in that section,  
1445 developing this resource information requires the use of a variety of Earth observations. In this section we list out these  
1446 observations for each resource category, as well as other types of observations relevant to the HOMER DSS.

1447

#### 1448 Wind Resources

1449

1450 The ideal observational platform for obtaining reliable wind resource data to be input into HOMER would be calibrated  
1451 wind speed measurements from a meteorological tower installed at the location of interest. These measurements should  
1452 be obtained at the hub height of the wind turbine being modeled, should be of sufficient sampling frequency to provide  
1453 hourly measurements, and should be of sufficient quality and duration to result in at least one full year of continuous  
1454 measurements. Although measurements of this quality are typically necessary at project sites where significant  
1455 investments in large grid-connected wind turbines are anticipate, and where a decision has already been made to  
1456 implement a large-scale project, it is extremely rare that this level of observation is available for most HOMER  
1457 applications, where the user is examining potential applications for proposed projects. Thus, some indirect means to  
1458 establish wind characteristics at a proposed site, such as extrapolating wind resource measurements available from a  
1459 nearby location or developing a wind resource map such as described in Section 2, is required. The major global data  
1460 sets typically used by NREL for wind resource assessment are summarized in table 1.

1461

1462 More discussion on some of these data sets is provided below.

1463

#### 1464 Surface Station Data

1465

1466 In the US, as well as in most other countries, the main source of routine surface wind observations would be  
1467 observations from nearby national weather stations, such as those routinely maintained to support aircraft operations at  
1468 airports. These data can be made available to the user from the National Climatic Data Center (NCDC) in the form of  
1469 the Integrated Surface Hourly (ISH) data set. This database is composed of worldwide surface weather observations  
1470 from about 20,000 stations, collected and stored from sources such as the Automated Weather Network (AWN), the  
1471 Global Telecommunications System (GTS), the Automated Surface Observing System (ASOS), and data keyed from  
1472 paper forms (see, [http://gcmd.nasa.gov/records/GCMD\\_C00532.html](http://gcmd.nasa.gov/records/GCMD_C00532.html)).

1473

#### 1474 Satellite-Derived Ocean Wind Data

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1476 Ocean wind data can be obtained from the SeaWinds Scatterometer (see <http://manati.orbit.nesdis.noaa.gov/quikscat/>)  
1477 mounted aboard NASA's Quick Scatterometer (QuickSCAT) satellite. QuickSCAT was launched on June 19, 1999 in a  
1478 sun-synchronous polar orbit. A longer-term ocean winds data set is available from the Special Sensor  
1479 Microwave/Imager (SSM/I) data products as part of NASA's Pathfinder Program. The SSM/I geophysical dataset  
1480 consists of data derived from observations collected by SSM/I sensors carried onboard the series of Defense  
1481 Meteorological Satellite Program (DMSP) polar orbiting satellites (see  
1482 [http://www.ssmi.com/ssmi/ssmi\\_description.html#ssmi](http://www.ssmi.com/ssmi/ssmi_description.html#ssmi)). An example of how Scatterometer data were used in support  
1483 of a wind resource assessment in Pakistan is provided in figure 2 (see also  
1484 [http://www.nrel.gov/international/rr\\_assess\\_pakistan.html](http://www.nrel.gov/international/rr_assess_pakistan.html)). Airborne or space borne Synthetic Aperture Radar systems  
1485 can also provide information on ocean wind data, although these data are not commonly used for this purpose in the US,  
1486 since Scatterometer data products are more readily and freely available.

1487

#### 1488 Reanalysis Upper Air Data

1489

1490 The US reanalysis data set was first made available in 1996 to provide gridded global upper air and vertical profiles of  
1491 wind data derived from 1,800 radiosonde and pilot balloon observations stations (Kalnay, *et al.* 1997). The reanalysis  
1492 data were prepared by NCAR-NCEP and can be found at <http://www.cdc.noaa.gov/cdc/reanalysis/>. An early analysis of

1493 the data set (Schwartz, George, and Elliott, 1999) showed that for wind resource assessments the dataset was a  
1494 promising tool for gaining a more complete understanding of vertical wind profiles around the world but that  
1495 discrepancies with actual radiosonde observations still existed. Since that time, continuous improvements have been  
1496 made to the NCAR-NCEP dataset, and it is has become an ever-increasingly important data source for contributing to  
1497 reliable wind resource mapping activities.

1498

#### 1499 Digital Terrain Data

1500

1501 Digital Elevation Models (DEM) have been accessed from the USGS/EROS data center. These models consist of a  
1502 raster grid of regularly spaced elevation values that have been derived primarily from the USGS topographic map series.  
1503 The USGS no longer offers DEMs, and for the US these can now be accessed from the National Elevation Dataset  
1504 (<http://ned.usgs.gov/>). The Shuttle Radar Topographic Mission (SRTM) offers much higher resolution terrain data sets,  
1505 which are now beginning to be used in some wind mapping exercises. These are also being distributed by USGS/EROS  
1506 under agreement with NASA (<http://srtm.usgs.gov/>).

1507

#### 1508 Digital Land Cover Data

1509

1510 Land cover data are used to estimate roughness length parameters required for the mesoscale meteorological models  
1511 used in the wind mapping process. Data from the Global Land Cover Characterization dataset provide this information  
1512 at a 1-km resolution (see <http://edcsns17.cr.usgs.gov/glcc/background.html>). The Moderate Imaging Spectroradiometer  
1513 (MODIS) is used to obtain global percent tree cover values at a spatial resolution of 0.5 km (Hansen, *et al.*, 2003).  
1514 Existing natural vegetation is also being mapped at a 200-m resolution as part of the USGS Regional Gap Analysis  
1515 program. Gap analysis is a scientific method for identifying the degree to which native animal species and natural  
1516 communities are represented in our present-day mix of conservation lands (Jennings and Scott, 1997).

1517

#### 1518 Solar Resources

1519

1520 As with wind, the ideal solar resource data set for incorporation into HOMER would be data derived from a quality,  
1521 calibrated surface solar measurement system consisting of a pyranometer and a pyrliometer that can provide a  
1522 continuous stream of hourly data for at least one year. Such data are seldom available at the site for which HOMER is  
1523 being applied. Although interpolation to nearby surface radiometer data sets can be accomplished with reasonable  
1524 reliability, we usually resort to an estimation scheme to derive an *in-situ* data set. The solar resource assessments that  
1525 NREL and others undertake make use of several different observational datasets, such as ground-based cloud cover  
1526 measurements, satellite-derived cloud cover measurements, or the use of the visible channel from satellite imagery data.  
1527 The major global data sets used for solar resource assessments are summarized in table 2.

1528

1529 More discussion on some of these data products is described below.

1530

#### 1531 World Radiation Data Center

1532

1533 Since the early 1960s the World Radiation Data Center, located at the Main Geophysical Institute in St. Petersburg,  
1534 Russia, has served as a clearinghouse for worldwide solar radiation measurements collected at national weather stations.  
1535 The WRDC is under the auspices of the World Meteorological Organization. A Web-based data set was developed by  
1536 NREL in collaboration with the WRDC and can be accessed at <http://wrdc-mgo.nrel.gov/>. This data archive covers the  
1537 period 1964 to 1993. For more recent data, the user should go directly to the WRDC home page at  
1538 <http://wrdc.mgo.rssi.ru/>.

1539

#### 1540 Aerosol Optical Depths (AOD)

1541

1542 After clouds, atmospheric aerosols have the greatest impact on the distribution and characteristics of solar resources at  
1543 the Earth's surface. However, routine *in-situ* observations of this parameter have only recently begun. Consequently, a  
1544 variety of surface-based and satellite-based observations are used to derive the best information possible of the temporal  
1545 and spatial characteristics of the atmospheric AOD. The most prominent of the surface data sets is the AERONET  
1546 (<http://aeronet.gsfc.nasa.gov/>), a network of automated multiwavelength sun photometers located around the world.  
1547 This network also has links to other networks, where the data may be less reliable. AERONET data can be used to



1548 provide ground-truth data for different satellite sensors that have been launched on a variety of sun-synchronous  
1549 orbiting platforms since the 1980s, such as the Total Ozone Mapping Spectrometer (TOMS), the Advanced Very High  
1550 Resolution Radiometer (AVHRR), the Moderate Resolution Imaging Spectroradiometer (MODIS), and the Multi-Angle  
1551 Imaging Spectroradiometer (MISR), the latter two mounted on NASA's Terra satellite. As noted by Gueymard (2003)  
1552 determination of AOD from satellite observations is still subject to inaccuracies, particularly over land areas, due to a  
1553 variety of problems such as insufficient cloud screening or interference with highly reflective surfaces. The Global  
1554 Aerosol Climatology Project (GACP), established in 1998 as part of the NASA Radiation Sciences Program and the  
1555 Global Energy and Water Experiment (GEWEX), has as its main objectives to analyze satellite radiance measurements  
1556 and field observations in order to infer the global distribution of aerosols, their properties, and their seasonal and  
1557 interannual variations and to perform advanced global and regional modeling studies of the aerosol formation,  
1558 processing, and transport (<http://gacp.giss.nasa.gov/>).

1559 Other sources of aerosol optical depth data include the Global Ozone Chemistry Aerosol Transport (GOCART) model  
1560 (<http://code916.gsfc.nasa.gov/People/Chin/gocartinfo.html>) which is derived from a chemical transport model. An  
1561 older dataset, the Global Aerosol Dataset (GADS), which can be found at [http://www.lrz-](http://www.lrz-muenchen.de/~uh234an/www/radaer/gads.html)  
1562 [muenchen.de/~uh234an/www/radaer/gads.html](http://www.lrz-muenchen.de/~uh234an/www/radaer/gads.html), is a theoretical data set providing aerosol properties averaged in space  
1563 and time on a  $5^{\circ} \times 5^{\circ}$  grid. (Koepke, *et al.*, 1997).

#### 1564 Other Renewable Energy Resources

1565 Although the scope of this chapter focuses on wind and solar energy resources, it is evident that many of the Earth  
1566 observation data sets listed above can apply to other renewable energy resources as well. For example, hydropower  
1567 resources can be determined by analysis of high resolution DEM data, along with knowledge of the rainfall amounts  
1568 over specific watersheds and the land use characteristics of these watersheds. Biomass resource assessments can be  
1569 enhanced through use of MODIS data as well as other weather-related data, and through evaluation of MODIS and  
1570 AVHRR data to determine the Normalized Vegetation Index (NVI).

1571 3b. Limitations on the usefulness of observations

1572

1573 In the absence of direct solar and wind resource measurements at the location for which HOMER is being applied, the  
1574 observations described in Section 3a, when used in the wind and solar resource mapping techniques described in  
1575 Section 2, will together provide useful approximations of the data required as input to HOMER. However, the  
1576 observations all have limitations in that they do not explicitly provide direct observation of the data value required for  
1577 the mapping techniques but only approximations based on the use of algorithms to convert a signal into the parameter of  
1578 interest. These limitations for some of these data sets can be summarized here:

1579

1580 Surface Station Data: These are generally not available at the specific locations at which HOMER would be applied, so  
1581 interpolation is required. Furthermore, they generally do not have actual solar measurements, but rather proxies for  
1582 these measurements (i.e., cloud cover). The wind data are generally collected at 10 m above the ground or less, and the  
1583 anemometer may not be in a well-exposed condition. When the station observations are derived from human  
1584 observations, they represent samples of a few minutes duration every 1 or 3 hours; therefore, many of the observations  
1585 are missing. For those stations that have switched from human observations to Automated Surface Observation Stations  
1586 (ASOS), the means of observation have changed significantly from the human observations, representing a  
1587 discontinuity in long-term records. Occasionally, the location of the station is changed without changing the station ID  
1588 number, which can also cause a discontinuity in observations. Similarly, equipment changes can cause a discontinuity in  
1589 observations

1590

1591 Satellite-Derived Ocean Wind Data: These data are not based on direct observation of the wind speed at  
1592 10 m above the ocean surface, but rather from an algorithm that infers wind speeds based on the wave height  
1593 observations provided by the scatterometers or Synthetic Aperture Radar

1594

1595 Satellite-Derived Cloud Cover and Solar Radiation Data: These data sets are derived from observations of the  
1596 reflectance of the solar radiation from the Earth-atmosphere system. Although it could be argued that this method does  
1597 provide a direct observation of clouds, the solar radiation values are determined from an algorithm that converts  
1598 knowledge of the reflectance observation, the incoming solar radiation at the top of the atmosphere, and the  
1599 transmissivity characteristics of the atmosphere to develop estimates of solar radiation.

1600

1601 Aerosol Optical Depth: Considerable research is underway to improve the algorithms used to convert multi-spectral  
1602 imagery of the Earth's surface to aerosol optical depth. The satellite-derived methods have additional shortcomings  
1603 over land surfaces, where irregular land-surface features make application of the algorithms complicated and uncertain.

1604

1605

1606 3c. Reliability of the observations

1607

1608 For those observations that provide inputs to the solar and wind resource data, their reliability can vary from parameter  
1609 to parameter. Generally all of the observations used to produce data values required for solar and wind assessments  
1610 have undergone rigorous testing, evaluation, and validation. This research has been undertaken by a variety of  
1611 institutions, including the institutions gathering the observations (e.g., NASA and NOAA) as well as the institutions  
1612 incorporating the observations into resource mapping techniques (e.g., NREL). Many of the satellite-derived  
1613 observations of critical parameters will be less reliable than *in-situ* observations; however, satellite-derived observations  
1614 must still be used due to the scarcity of *in-situ* measurement stations.

1615

1616 3d. What kinds of observations could be useful in the near future

1617

1618 All of the observations currently available will continue to be of critical value in the near future. For renewable energy  
1619 resource mapping, improved observations of key weather parameters (wind speed and direction at various heights above  
1620 the ground and over the open oceans at higher and higher spatial resolutions, improved ways of differentiating snow  
1621 cover and bright reflecting surfaces from clouds, etc.) will always be of value to the renewable energy community.  
1622 New, more accurate methods of related parameters such as aerosol optical depth would result in improvements in the  
1623 resource data. All of these steps will lead to improvements in the quality of outputs from renewable energy decision  
1624 support tools such as HOMER.

1625

1626 **4. Uncertainty**

1627

1628 Application of the HOMER DSS involves a variety of input data types, all of which can have a level of uncertainty  
1629 attached to them. HOMER addresses uncertainties by allowing the user to perform sensitivity analyses for any  
1630 particular input variable or combination of variables. HOMER repeats its optimization process for each value of that  
1631 variable and provides displays to allow the user to see how results are affected. An input variable for which the user has  
1632 specified multiple values is called a sensitivity variable, and users can define as many of these variables as they wish.  
1633 In HOMER, a “one-dimensional” sensitivity analysis is done if there is a single sensitivity variable, such as the mean  
1634 monthly wind speed. If there are two or more sensitivity variables, the sensitivity analysis is “two” or “multi-  
1635 dimensional.” HOMER has powerful graphical capabilities to allow the user to examine the results of sensitivity  
1636 analyses of two or more dimensions. This is important for the decision maker, who must factor in the uncertainties of  
1637 input variables in order to make a final judgment on the outputs of the model.

1638  
1639 The amount of uncertainty associated with resource data is largely dependent on how the data are obtained and on the  
1640 nature of the analysis being undertaken. For some types of analyses, very rough estimates of the wind resource would  
1641 be sufficient; for others, detailed hourly average data based on surface measurements would be necessary. Quality *in-*  
1642 *situ* measurements of wind and solar data in formats suitable for renewable energy applications over a sufficient period  
1643 of time (one year or more) can have uncertainties of less than  $\pm 3\%$  of the true value. However, when estimation  
1644 methods are required, such as the use of Earth observations and modeling and empirical techniques, uncertainties can be  
1645 as much as  $\pm 10\%$  or more. These uncertainties are highest for shorter-term data sets, and are lower when annual  
1646 average values are being used, since throughout the year errors in the estimation methods have a tendency to  
1647 compensate among the individual values.

1648  
1649 Based on wind turbine and solar technology operating characteristics, it is possible that the error in estimating a  
1650 renewable energy system performance over a year is roughly linear to the error in the input resource data. For example,  
1651 for wind energy systems, even though the power of the wind available to a wind turbine is a function of the cube of the  
1652 wind speed, it turns out that the turbine operating characteristics, where turbines typically do not provide any power at  
1653 all until a certain threshold speed is reached, and then the power output increases linearly with wind speed until the  
1654 winds are so high that the turbine must shut down. This results in an annual turbine power output that is roughly linear  
1655 to the mean annual wind speed for certain mean wind speed rangers. This would mean that, in some cases, an

1656 uncertainty in the annual wind or solar resource of  $\pm 10\%$  results in an uncertainty of expected renewable energy  
1657 technology output of approximately  $\pm 10\%$ .

1658

1659

## 1660 **5. Global change information and the HOMER DSS**

1661

1662 This section expands the discussion of the HOMER DSS to include the relationship of HOMER and its input data  
1663 requirements with global change information

1664

### 1665 5a. Reliance of HOMER DSS global change information

1666

1667 As shown in the previous section, a number of observations that provide information on global change are also used in  
1668 either direct or indirect ways as input to HOMER. These observations related primarily to the renewable energy  
1669 resource information that is required for HOMER applications. Renewable energy system performance is highly  
1670 dependent on the local energy resources available to the technologies. The extent and characteristics of these resources  
1671 is driven by weather and local climate conditions, which happens to be the primary area in which Earth observational  
1672 systems monitoring climate change are addressing. Thus, as users seek access to observations to support renewable  
1673 energy resource assessments, they will invariably be seeking certain global change observational data.

1674

1675 Specifically, users will be seeking global change data related to atmospheric properties that support the assessment of  
1676 solar and wind energy resources, such as wind and solar data, and atmospheric parameters important for estimating  
1677 these data. For example, major data sets used in solar and wind energy assessments include long-term reanalysis data,  
1678 climatological surface weather observations, and a variety of satellite observations from both active and passive  
1679 onboard remote sensors.

1680

1681 Key factors in affecting the choice of these observational data are their relevance to conducting reliable solar and wind  
1682 energy resource assessment, their ease of access, and low or no cost to the user. The extensive list of observational data

1683 being used in the assessment of renewable energy resources represents strong leveraging of major, taxpayer-supported  
1684 observational programs that are geared primarily for global change assessment.

1685

1686 There is also an important consideration regarding the potential influence of long-term climate change on the renewable  
1687 energy resources that are used as input into HOMER. Through the Intergovernmental Panel on Climate Change there  
1688 has been a significant improvement in the reliability and spatial resolution of General Circulation Models (GCM) used  
1689 to estimate the impacts of greenhouse gas emissions on climate change. As weather patterns change under changing  
1690 climate conditions, wind and solar energy resources at a specific location can also change over time. The GCM results  
1691 indicate that these renewable energy resources can be measurably different 50 to 100 years from now than today in  
1692 specific locations and regions. These changes may have a noticeable impact on the results of HOMER simulations in  
1693 the future; however, significant uncertainties exist in GCM results. Until these uncertainties are reduced sufficiently,  
1694 implementation of GCM results will produce unreliable HOMER simulations.

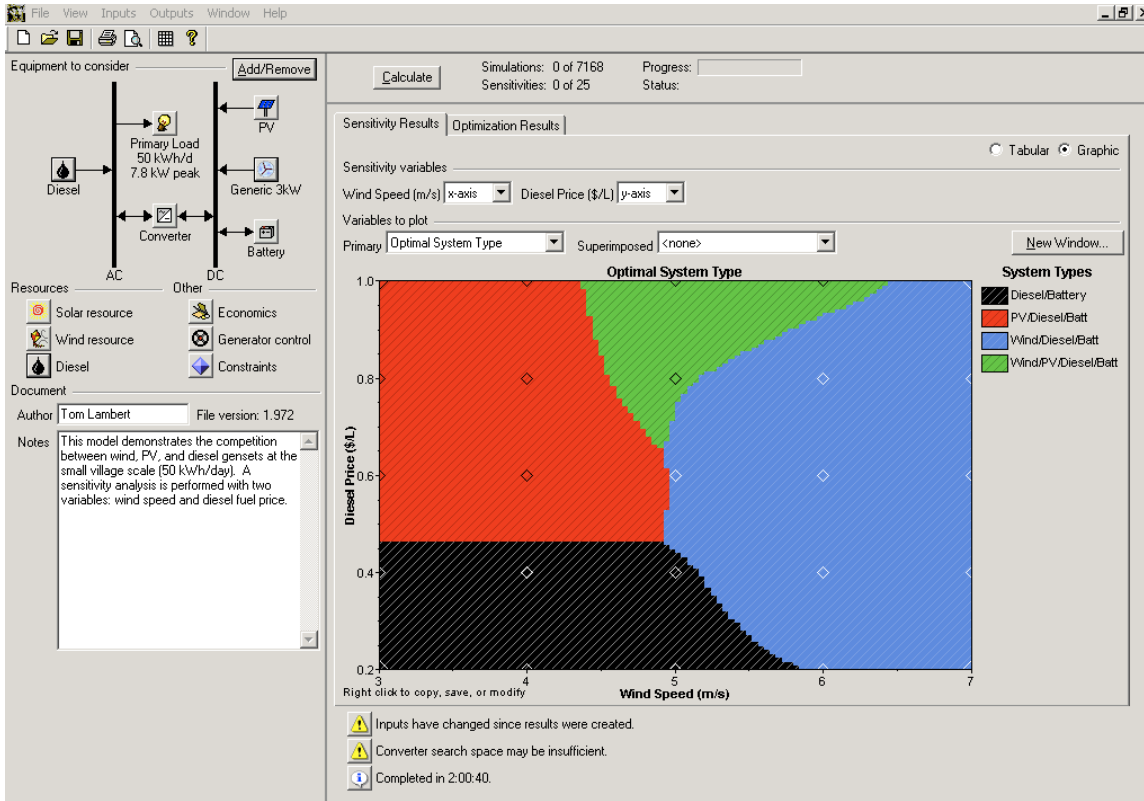
1695

1696 5b. How the HOMER DSS can support climate-related management decision-making among US government agencies

1697

1698 Although HOMER was not intentionally designed to be a climate-related management decision-making tool, the  
1699 HOMER DSS has attributes that can support these decisions. For example, as we explore mechanisms for mitigating  
1700 the growth of carbon emissions in the atmosphere, the HOMER DSS can be deployed to evaluate how renewable energy  
1701 systems can be used cost-effectively to displace energy systems dependent on fossil fuels. Clearly, the science results  
1702 and global change data and information products coming out of our reanalysis and satellite-borne programs are of  
1703 critical importance to HOMER for supporting this decision-making process. Given that the pertinent observational data  
1704 sets have been developed primarily by federal agencies, these data sets tend to be freely available or available at a  
1705 relatively small cost, given the costs involved in making the observations in the first place. However, as we have noted  
1706 in previous sections, the use of global change observations as input to the resource assessment data required by  
1707 HOMER is not the optimal choice of data; ideally, *in-situ* (site-specific) measurements of wind and solar data relevant  
1708 to the technologies being analyzed would be the most useful and accurate data to have for HOMER, if they were  
1709 available.

1710 Figure 1: Example of HOMER output graphic. The column on the left provides a diagram showing the load  
1711 characteristics and the types of equipment considered to meet the load. The optimal system design graphic shows the  
1712 range within specified diesel fuel prices and wind energy resources for which various system types are most economical  
1713 (e.g., a wind/diesel/battery system becomes the most optimal configuration to meet the load requirement for wind  
1714 speeds greater than 5 m/s and fuel costs at 0.45 to 0.75\$/l.  
1715

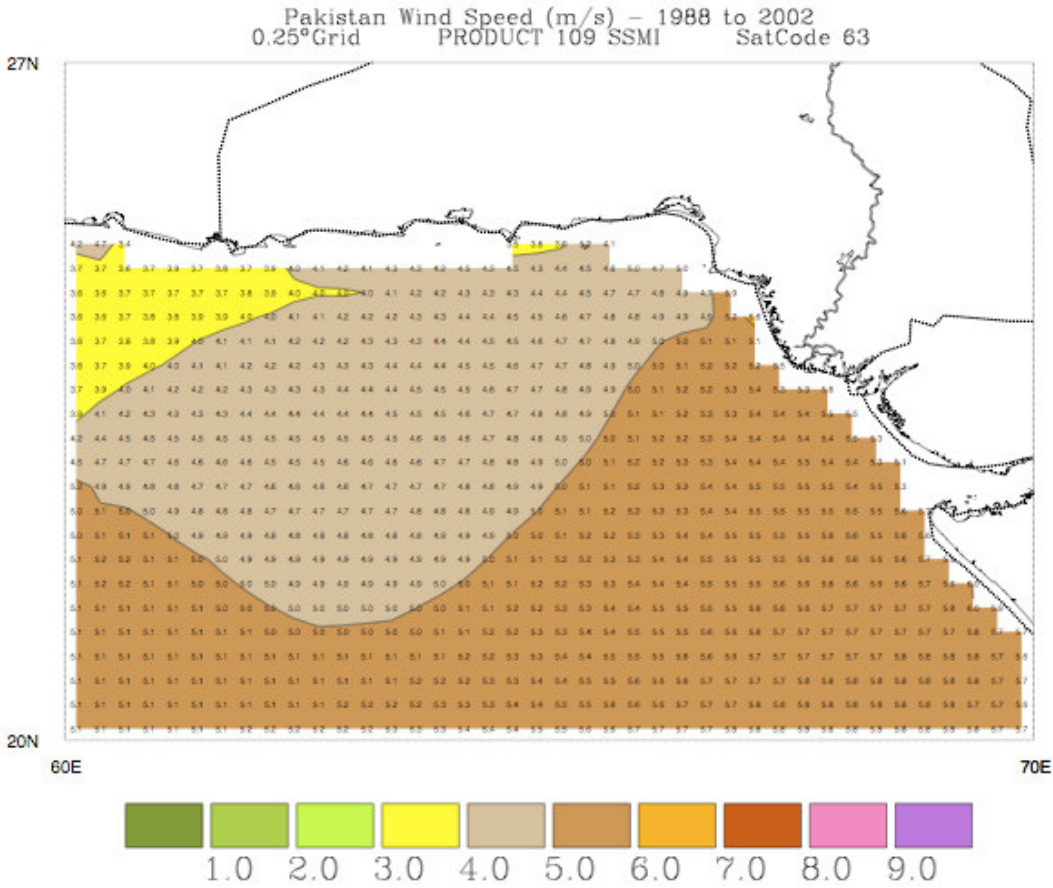


1716  
1717

1718 Figure 2. Example of ocean wind resource assessment output for the offshore regions of Pakistan. These data were  
1719 derived from the SeaWinds scatterometer aboard NASA's QuickSCAT satellite. The assessment provides estimated  
1720 mean annual wind speeds at 10-m above the ocean surface, averaged over the period 1988 to 2002.

1721

1722



1723



1724 Table 1: Major Global Data Sets Used by NREL for Wind Resource Assessment

1725

<b>Data Set</b>	<b>Type of Information</b>	<b>Source</b>	<b>Period of Record</b>
Surface station data	Surface observations from more than 20,000 stations worldwide	NOAA/NCDC	Variable up to 2006
Upper air station data	Rawinsonde and pibal observations at 1,800 stations	NCAR	1973–2005
Satellite-derived ocean wind data	Wind speeds at 10 m above the ocean surface gridded to 0.25 <sup>o</sup>	NASA/JPL	1988–2006
Marine climatic atlas of the world	Gridded (1.0 <sup>o</sup> ) statistics of historical ship wind observations	NOAA/NCDC	1854–1969
Reanalysis upper air data	Model-derived gridded (~200-km) upper air data	NCAR-NCEP	1958–2005
Global upper air climatic atlas	Model-derived gridded (2.5 <sup>o</sup> ) upper air statistics	NOAA/NCDC	1980–1991
Digital geographic data	Political, hydrograph, etc.	ESRI	N/A
Digital terrain data	Elevation at 1-km spatial resolution	USGS/EROS	N/A
Digital land cover data	Land use/cover and tree cover density at 0.5-km resolution	NASA/USGS	N/A

1726

1727 Table 2: Major global data sets used for solar resource assessments

1728

1729

<b>Data Set</b>	<b>Type of Information</b>	<b>Source</b>	<b>Period of Record</b>
Surface station data	Surface cloud observations from more than 20,000 stations worldwide	NOAA/NCDC	Variable up to 2006
World Radiation Data Center	Surface radiation observations from over 1,000 stations worldwide	WRDC, St. Petersburg	1964–1993
Satellite imagers	Imagery from the visible channel of geostationary weather satellites, 1-km resolution	NASA/NOAA	1997– present
International Satellite Cloud Climatology Project	Used in the 1 <sup>0</sup> global surface solar energy meteorological data set	NASA/SSE	1983–2003
AERONET	Observations of aerosol optical depth from around the world	NASA/Goddard	Variable depending on station
GACP	Aerosol optical depths (generally over oceans) at 1 <sup>0</sup> x 1 <sup>0</sup> from AVHRR data	NASA	1981–2005
MODIS, MISR, TOMS	Aerosol optical depth	NASA	Variable since

			1980s
GOCART	Aerosol optical depth for turbid areas	NASA	March 30-May 3, 2001
GADS	Aerosol optical depth derived from theoretical calculations and proxies		Compilation of Measurements & Models
Digital geographic data	Political, hydrography, etc.	ESRI	N/A
Digital terrain data	Elevation at 1-km spatial resolution	USGS/EROS	N/A
Digital land cover data	Land use/cover and tree cover density at 0.5-km resolution	NASA/USGS	N/A

1730