Comparison of ERA40 and NCEP/DOE near-surface datasets with other ISLSCP-II datasets

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Abstract

The fields of 2-m temperature, relative humidity, precipitation, downward, short-wave and long-wave radiation, net radiation, sensible and latent heat flux from the ERA40 and NCEP/DOE reanalyses are compared with each other and with other independent ISLSCP-II data sets, where available. There are differences in the climatologies of the different data sets, but generally they show consistent patterns of the major seasonal anomaly fields. The anomaly patterns are coherent, showing warm, dry seasonal biases associated with reduced precipitation and cloudiness, and the converse. This confirms that major changes in the atmospheric circulation patterns, with the associated differences in surface temperature, humidity, precipitation, cloud fields and incoming surface SW and LW radiation fluxes are captured by both reanalyses and the comparison ISLSCP-II datasets.
1. Introduction

Near-surface meteorology data sets were extracted for the years 1986-95 for the second International Land-Surface Climatology Project (ISLSCP-II) from the European Centre for Medium-range Weather Forecasts (ECMWF) re-analysis [ERA40, Simmons and Gibson 2000; Källberg et al. 2004; Uppala et al. 2005; http://www.ecmwf.int/research/era/], and by the Center for Ocean-Land Atmosphere Studies (COLA) from the National Centers for Environmental Predictions (NCEP)/Department of Energy (DOE) Atmospheric Model Intercomparison Project (AMIP)-II Reanalysis. [http://wesley.wwb.noaa.gov/reanalysis2/, Kanamitsu et al. 2002]. For brevity we will refer to the two reanalyses as ERA40 and NCEP2. In this paper we compare five basic climate parameters; near-surface temperature, relative humidity, precipitation, incoming short-wave and long-wave radiation fluxes from the two reanalyses with other ISLSCP-II data sets [Hall et al. 2006]. We also intercompare the surface sensible and latent heat fluxes from the two reanalyses, for which there is no corresponding global observational time-series over land (although we have an ocean climatology). The purpose of this analysis, which is purely qualitative, is to give users of the ISLSCP-II datasets a broad visual overview of the differences and similarities between some key datasets.

1.1 The concept of reanalysis and data assimilation

Routine analyses are produced by operational meteorological centers in real time several times each day using the current version of the centers’ global forecast and analysis system. These models are constantly being updated and improved, so that over time fundamental climatological properties of the model are also changed, sometimes drastically. This makes a long time series of operational analyses useless for examining long-term trends or variations in climate. A reanalysis
is a way to produce a dynamically consistent global analysis of the state of the atmosphere over an extended period of time (many years or decades) with no gaps in space or time. This is done by using a “frozen” version of the analysis model, and performing a retrospective analysis using the historic archive of observations. This allows for the use of more observational data, as many high-quality observations are not available to the operational centers in real-time.

Observational networks do not cover the entire globe uniformly, but have gaps in space and vary in their coverage over time. However, observational measurements provide the best estimate of the state of the atmosphere where they are taken. On the other hand, a geophysical fluid-dynamical model of the atmosphere containing parameterizations of important physical processes like radiative transfer, convection, turbulent transfer and diffusion of heat, moisture and momentum, can provide a complete global simulation of the atmosphere and is also used for forecasting purposes as it can be integrated forward in time. However, the models are imperfect representations of the atmosphere (as many small-scale processes are parameterized), and are prone to systematic errors, drift, and limitations owing to their finite spatial and temporal resolutions. A reanalysis is a combination of model and measurement, using observations to constrain the dynamical model to optimize between the properties of complete coverage and accuracy. The insertion of observational information into the model integration is called data assimilation. Operational meteorological centers developed data assimilation as a means to generate initial conditions for dynamical numerical models that are consistent both with the model and the observed state of the atmosphere, thereby improving forecasts. However, the data assimilation practiced by operational centers is an adjustment of state variables – essentially an additional term in the predictive equations called an analysis increment. Because of the addition of these increments, reanalyses frequently do not locally conserve mass, energy, water or
momentum. This can make operational reanalyses challenging to use in certain science applications.

Although the reanalysis model is frozen, there are significant changes with time in the observational datasets, especially during the last few decades in satellite data. The ISLSCP-II decade of 1986-1995 does include significant changes in the satellite observations; and especially the introduction of microwave sensor data in 1987. Note that the changes with time in the observational systems means that these reanalyses still cannot determine trends on decadal timescales, although, as we shall show, the seasonal anomalies are coherent and useful.

1.2 Description of the two reanalyses

1.2.1 ERA40 reanalysis system

The European Centre for Medium-range Weather Forecasts (ECMWF) re-analysis [ERA40, Källberg et al. 2004; Uppala et al. 2005] covers the period from September 1957 to August 2002. Links can be found at http://www.ecmwf.int/research/era/ for many aspects of ERA40, including documentation of the cycle 23r4 Integrated Forecast System that was used for this reanalysis; and a summary and discussion of the observations available at different times during the 40-year reanalysis. The ERA40 model has 60 atmospheric levels in the vertical, from the top of the model at 0.1 hPa to the lowest model level at about 10 m above the surface. The spectral resolution is T159 (triangular truncation at wave number 159) with a corresponding resolution of about 125 km in grid point space. A so-called reduced Gaussian grid (N80, described in http://www.ecmwf.int/products/data/technical/ gaussian/) is used for many physical processes and the land surface parameters. The ISLSCP products have been interpolated from this grid to the uniform 1x1 deg ISLSCP Earth grid, as much as possible consistent with the land-sea mask
definitions. The ERA40 land sea mask (LSM) and the ISLSCP LSM are used to ensure that when possible only land points are transformed into land points, and only sea points are used for sea points. The ERA40 analysis uses the so-called 3-dimensional variational method, where a cost function in relation to observations and background model field is minimized. The weighting of the different parts of the cost function is controlled by estimates of observation errors and model background errors. The spreading of observations in the horizontal and the vertical is controlled by horizontal and vertical correlation of the background errors. Most satellite observations (e.g. from infrared sensors) are used by computing radiances from the model fields (using a forward model) and by comparing them with the satellite radiances. The analysis system uses a wide range of other observations, from conventional radiosonde and synoptic (SYNOP) observations, to ocean winds from satellite scatterometry. The ISLSCP period of 1986-1995 spans changes in the satellite observations used in the analysis. The satellite microwave data from the Special Sensor Microwave/Imager (SSM/I) was introduced in 1987 and the European Remote-Sensing Satellite (ERS) data in 1991.

The analysis of T and q at the 2-m level and the snow depth analysis are part of a separate surface analysis, which uses a successive correction method. Because large areas over land do not have snow depth observations, a weak relaxation is applied to a snow depth climatology that is specified. ERA40 uses an optimal interpolation of soil water [Douville et al. 2000], which adds soil water increments based on analysis increments in 2-m T and q. The land-surface scheme for ERA40 and its parameters are discussed in Van den Hurk et al. [2000] and its performance over the boreal forest in Betts et al. [2001]. ERA40 has a four layer soil model with a representation of frozen and unfrozen soil. In each grid box, there are tiles for bare soil and two classes of vegetation, short and tall, each with a specified vegetation type with a fixed set of parameters: fractional area, rooting distribution, leaf area index, roughness and canopy resistance. None of the
vegetation parameters vary with time. Snow is represented by a single layer, which lies on top of bare ground and short vegetation, so that it is directly coupled to the atmosphere; but under the canopy for tall vegetation, with a separate energy balance that is less strongly coupled to the atmosphere. A report series evaluating ERA40 is available at http://www.ecmwf.int/research/era/Products/

This paper extends the work of Betts and Beljaars [2003], which compared intercompared ERA40 with other ISLSCP-II data sets. Evaluations of ERA40 have been made using river basin budgets for the Mississippi, Mackenzie, and Amazon basins [Betts et al. 2003a, b, 2005; Betts and Viterbo 2005]. A useful time-series analysis of the trends and variability in the CRU, ERA40 and the earlier NCEP/NCAR analyses of surface air temperature is given in Simmons et al. [2004]. A global evaluation of the hydrological cycle in ERA40 is given in Hagemann et al. [2005]. The ERA40 system has a spin-up of the precipitation field at high latitudes for the first 24-36 hours, associated with a problem in the moisture analysis [Betts et al. 2003b]. The ERA40 reanalysis has a high bias in precipitation over the tropical oceans, stemming from a problem in the use of satellite radiances in the analysis of humidity [Troccoli and Kållberg, 2004]. ERA40 has a known error in the diurnal cycle of precipitation over land (a bias towards precipitation too early in the day) which is larger in the tropics [Betts and Jakob 2002] than the mid-latitudes. Betts [2006] analyzes the coupling of the diurnal cycle of temperature in ERA40 to the net long-wave radiation field. Here we average over the diurnal cycle. There is also a cold bias over ice-covered oceans in both the Arctic and the Antarctic [Betts and Beljaars 2003], relating to the assimilation of infrared data satellite data, which affects part of the ISLSCP-II period, 1989-1996 (it was identified and largely corrected as the reanalysis progressed). The eruption of Pinatubo in 1991, which put volcanic aerosol into the stratosphere, impacts the infrared radiances, which in turn impact the analysed tropical circulation and rainfall, mainly over the tropical oceans.
The output from ERA40 is an analysis every 6h at the standard synoptic times: 00, 06, 12 and 18 UTC. From each analysis 6-h forecasts are run; and twice a day from the 00 and 12 UTC analyses, these forecasts are extended to 36h, with fields archived every three hours. The ISLSCP-II data includes both analyses every six hours, and model forecast data every three hours. In this paper, the ERA40 data comes primarily from monthly averages of the four daily 0-6h short-range forecasts, except for precipitation, where we shall show the monthly averages from 24-36h forecasts. All flux variables are averages accumulated during a model forecast, while some variables, such as temperature, humidity and wind are instantaneous fields. The ERA40 ISLSCP-II dataset comprises some fixed fields, soil and snow variables (monthly time-step), surface fluxes (monthly and 3-hrly), near surface variables (monthly, 3-hrly) at the lowest model level (roughly 10m above the surface); and a level about 100m above the surface (intended for driving off-line land-surface models in a loosely coupled mode). There are also the conventional 2-m temperature and dewpoint, and the 10-m wind that are computed using the boundary layer model from the predicted variables at the surface and the lowest model level. In addition, there is a separate 2-m analysis of temperature and dewpoint (6-hrly).

1.2.2 NCEP2 reanalysis system

The Center for Ocean-Land Atmosphere Studies (COLA) near-surface meteorology data set for ISLSCP-II has been derived from the NCEP/DOE AMIP-II reanalysis (NCEP2) [Kanamitsu et al. 2002] that covers the years from 1979-2003. The purpose of this NCEP/DOE Reanalysis is to provide an improved version of the original NCEP/National Center for Atmospheric Research (NCAR) reanalysis [Kalnay et al. 1996; Kistler et al. 2001] for use by the AMIP-II project [Gleckler, 1996; http://www-pcmdi.llnl.gov/projects/amip/NEWS/amipnl8.pdf] for general circulation model validation. The NCEP2 reanalysis uses a very similar analysis
system to the NCEP/NCAR reanalysis and an upgraded version of the same general circulation model, with known errors fixed and assimilation of a more complete stream of observational data after 1993. To co-register the NCEP2 reanalysis to the ISLSCP 1° Earth grid, the reanalysis data set was re-gridded using bi-linear interpolation from its native T62 Gaussian grid resolution (192 x 94 grid boxes globally) to the 1° spatial resolution required by ISLSCP-II. When possible, NCEP2 land grid points are mapped to ISLSCP land grid points, and NCEP2 water grid points are mapped to ISLSCP water grid points. On occasions where there is no overlap of like surfaces between the two grids, a straight interpolation is performed. The NCEP2 reanalysis scheme is also a three-dimensional variational scheme called spectral statistical interpolation [Parrish and Derber 1992]. The NCEP reanalysis system is thoroughly reviewed in Kalnay et al. [1996]: here we summarize the error fixes, model changes, new components and boundary conditions implemented for NCEP2 [Kanamitsu et al. 2002]. Errors found and corrected include misregistration of PAOBS (subjective sea-level pressure analyses) over the Southern Hemisphere for 1979-1992, mishandling of observed snow-cover data between 1974-1994, a problem in humidity diffusion that led to extremely noisy snow analysis maps, spatial discontinuities in the relationships between relative humidity and cloudiness, and problems in the representation of ocean albedo and snowmelt. Major changes to the model physics for NCEP2 were the implementation of a new planetary boundary layer scheme [Hong and Pan 1996] and shortwave radiation parameterization [Chou and Lee 1996]. Minor changes include an increase in the frequency of full radiation calculations from 8 to 24 times daily, retuning of the stratus cloud and convective parameterizations, improved cloud-top cooling and calculation of radiation on the full Gaussian grid. Two new system components may have a particularly strong impact on the surface fields in ISLSCP-II. An assimilation of precipitation was incorporated to improve the simulation of soil wetness. The difference of pentad-averaged reanalysis rainfall from the Xie and Arkin [1997]
precipitation estimate is subtracted from infiltration into the top soil layer in an effort to improve the land surface hydrology. Similarly, adjustments to observed snow cover analyses based on model predicted snow cover replaced a simple empirical estimate for more realistic winter hydrology. Finally, several fixed fields were changed, including new data for desert albedos, ozone, Northern Hemisphere snow cover; and the sea surface temperatures and sea-ice data used are consistent with AMIP-II. The land surface scheme used in NCEP2 is the same as for the original NCEP/NCAR reanalysis. It is a relatively simple two-layer soil with simple parameterizations of surface flux exchanges, including restrictions on evapotranspiration caused by plant vascular systems [Pan and Mahrt 1987]. This scheme includes no interannual or seasonal variations in vegetation cover, and minimal spatial variability of surface properties.

Some data comparisons of the COLA NCEP2 dataset are contained in a report by Zhao and Dirmeyer [2003], who discuss the production of a hybrid data set for the second Global Soil Wetness Project [Dirmeyer et al. 2006]. There are several studies of the hydrologic cycle of the NCEP2 reanalysis. Roads et al. [2002] evaluated the NCEP2 water and energy budgets on a global scale. Roads et al. [2003] is a comprehensive analysis of the water and energy budgets for the Mississippi river basin comparing the two NCEP reanalyses with operational model products and an off-line hydrologic model [Maurer et al., 2001]. Roads [2003] evaluates the tropical precipitation of both NCEP reanalyses against TRMM data. Lu et al. [2005] compared the simulations of soil wetness to observations and the earlier NCEP/NCAR reanalysis and found some improvements, although Dirmeyer et al. [2004] describe shortcomings in the ability of NCEP2 to simulate interannual variations of soil wetness where long-term observations are available.

The analysis increment for the NCEP2 reanalysis is six hours, and output data are routinely reported every hour or six hours. In order to satisfy the ISLSCP-II requirement for 3-hourly data,
twice daily 36 hour forecasts were made and hourly output were obtained from the 24-36 hour forecasts. The 24-36 hour forecasts were chosen instead of the 0-12 hour forecasts to minimize initial "spin-up/spin-down" problems. The one-hour data were later combined to produce 3-hourly data. Time averaging was performed on the native reanalysis grid before regridding. The fields from the NCEP2 reanalysis that are provided for ISLSCP-II are near surface meteorological fields, fluxes of heat, moisture and momentum at the surface, and land surface state variables, all with a spatial resolution of 1° in both latitude and longitude. There are five temporal categories of data; time invariant and monthly mean annual cycle fields (together referred to as “fixed” fields); monthly mean fields; monthly-3-hourly (mean diurnal cycle) fields, and 3-hourly fields. Two types of variables exist in this data; instantaneous fields (primarily state variables), and average fields (primarily flux fields expressed as a rate).

1.3 Differences between NCEP2 and ERA40 Reanalyses

Note that there are fundamental differences between the NCEP2 and ERA40 reanalysis products (Table 1), which are not evident to the user of the ISLSCP-II versions of the products, since they have been co-registered on the same 1° grid, and presented with the same time structure. These differences may contribute beyond what might be expected simply from the disparities in model physics and assimilated data streams. For example, the differences in the forecast intervals used are a response to very different spin-up behaviours in the two products, particularly for precipitation and surface fluxes.

1.4 Comparison with ISLSCP-II data
In this paper, we show comparisons of the two reanalyses with three other ISLSCP-II datasets [Hall et al. 2006]. For temperature and relative humidity, we compare with the ISLSCP-II surface meteorology data set re-gridded from the Climatic Research Unit, University of East Anglia data set [New et al. 1999, 2000]. This we shall refer to as the CRU dataset. Temperature is a primary variable with extensive coverage, which is directly analyzed. Where observations are sparse and over mountainous terrain, this gridded temperature analysis may not be representative. However New et al. [2000] treat vapor pressure (from which we compute relative humidity, RH) as a secondary variable, and the gridded analysis uses not only station observations where available (converting monthly mean RH to vapor pressure for some stations) but also synthetic data (found by estimating monthly dewpoint from monthly mean minimum temperature), which is added to the analysis in regions of sparse data.

For precipitation we compare with the Global Precipitation Climatology Project (GPCP) Version 2 Satellite-Gauge (SG) combination global gridded monthly precipitation dataset [Adler et al. 2003], which is considered the best all-round monthly precipitation in the ISLSCP-II set, that is derived from observations as opposed to reanalysis [Hall et al., 2006]. Over land, the GPCP SG consists of a standard gauge analysis with climatological bias correction, in combination with a community-based satellite-only product to improve estimates where gauges are sparse. Furthermore, the GPCP SG provides a seamless transition to that satellite-only product alone over the oceans and other un-gauged regions. Thus, the GPCP SG is globally complete, albeit with reduced confidence over the oceans and at high latitudes. For brevity, this will be referred to in this paper as the GPCP precipitation. Further discussion of this GPCP dataset, and the other precipitation data sets in the ISLSCP-II collection is given in Hall et al. [2006].

For the surface radiation, we compare the downwelling shortwave (SWdown) and longwave (SWdown) radiation fluxes with the corresponding radiation fluxes from the surface
radiation budget dataset (for which we use the acronym SRB) developed by Stackhouse et al. [2004] and Cox et al. [2004] from the ISCCP cloud data [Rossow and Schiffer 1999]. The algorithms used for the SW fluxes are documented in Pinker and Laszlo [1992] and Gupta et al. [1992] for the LW radiation fluxes. Meteorological profile information is developed at the processing grid resolution from the NASA Data Assimilation Office Goddard Earth Observing System version 1 (GEOS-1) reanalysis [Schubert et al. 1995]. Note that this is a different and earlier reanalysis, which has a different set of temperature and humidity biases from NCEP2 and ERA40.

For the surface sensible and latent heat fluxes, we have no comparison data set over land, but we will compare the reanalyses over the oceans with a monthly climatology, derived from marine observations [DaSilva et al. 1994]. This climatology does end however in 1993, so we use the 1986-1993 climatology for comparison.

2. Comparison of reanalyses with selected ISLSCP-II products

The advantage of surface fields from model reanalyses is that they have complete coverage at 3-hourly time resolution (in this ISLSCP-II data). In contrast, surface observations are not uniformly distributed globally, and are sparse over many regions in the tropics, where only monthly mean data may be available, or even just climatology. Surface observations can be interpolated to a common grid, as is the case for our CRU dataset, but where observations are sparse; the gridded analysis may not be representative. Models not only assimilate data, but in regions of missing data, the global model will compute a complete set of fields from continuity and the dynamic and thermodynamic equations. This means that model products have biases, related to the specific set of model equations and the parameterizations for unresolved physical
processes. The NCEP and ECMWF analysis-forecast systems differ in their model structure, physical parameterizations and horizontal and vertical resolution, and in their methods of processing the input observations. Consequently there are differences between the model surface fields. In this short paper, we cannot analyze in detail all the differences between the reanalyses. Our intent is to help the users of the ISLSCP-II data assess which products might be useful for different purposes. In many cases the differences between reanalyses and other datasets will give an estimate in our uncertainty in a given variable. We will give a broad overview on seasonal timescales of differences at the surface by comparing the reanalysis fields with independent data from the ISLSCP-II data collection. The ISLSCP data set contains several other near-surface products, some derived directly from surface observations, and some from satellite observations. We will present the 10-year mean climatology for DJF (December, January, February) and JJA (June, July, August) for the two reanalyses and a corresponding dataset; and the difference fields of the reanalyses from this dataset, as well as the difference of the two reanalyses. We will also compare anomaly fields for two seasons (DJF, 1991-1992 and JJA, 1988), where the anomalies are computed from the separate climatologies for each reanalysis or other data set. The first, DJF, 1991-1992, has a warm anomaly in the eastern Pacific, the onset of an El Nino event; and the second, JJA 1988 has a cold anomaly in the eastern Pacific, and also corresponded to a drought over the continental United States. A full set of these anomaly fields for all four seasons is available for ERA40 at

http://www.ecmwf.int/research/demeter/d/inspect/catalog/research/era/diagnostics/ISLSCP-II/

2.1 Surface temperature and relative humidity fields

2.1.1 The 2-m temperature
Figures 1 and 2 show the 2-m temperature distribution from ERA40, NCEP2 and the CRU data (over land) for the 10-year (1986-1995) DJF and JJA climatologies and selected difference fields. The three climatologies on the left (panels (a), (b) and (c)) show that the two reanalyses are similar to each other and to the CRU analysis over land. The very low JJA temperatures (below –55 °C) over the Antarctic icecap do not appear with the contouring shown. Three difference fields are shown on the right in panels (d), (e) and (f); for ERA40-CRU, NCEP2-CRU, and NCEP2-ERA40 respectively. ERA40 tends to be a little cooler in the tropics than the CRU analysis, and a little warmer in mid- and high latitudes, especially over the boreal forest in the northern winter. These ERA40 temperature biases are consistent with those shown in Betts et al. [2003b] for the Mackenzie River basin and Betts et al. [2005] for the Amazon basin. The pattern of the NCEP2 temperature bias differs from ERA40. The cool bias over the Amazon and Sahara is a little larger for NCEP2 than ERA40, but over the Sahel in JJA, the two reanalyses have opposite biases. The Eurasian warm bias during DJF is larger in NCEP2 further to the east than in ERA40. Although sea surface temperature is specified, NCEP2 has a slightly warmer 2-m temperature over much of the oceans. The comparisons over regions of sparse data and over high terrain should be viewed with some caution. There are differences in the orography used in the two analyses, and in mountainous regions data coverage is generally and stations are often limited to the valleys. In polar regions where we have no CRU data, NCEP2 is warmer than ERA40 in the winter hemisphere and cooler in the summer hemisphere. ERA40 has a known cold bias over ice-covered oceans in both the Arctic and the Antarctic, relating to the assimilation of infrared satellite data (which was identified as the reanalysis progressed, so that only the years, 1989-1996, are affected). The difference pattern in Figures 1(f) and 2(f) resemble 1(e) and 2(e) more closely than they do 1(d) and 2(d); which means that, away from polar regions, ERA40 has generally smaller biases over land than NCEP2.
Figure 3 shows the anomaly fields (each from their own climatologies) for two seasons, DJF, 1991-1992 (left) and JJA, 1988 (right) for the two reanalyses and the CRU analysis over land. Despite the differences in the climatologies, the anomaly patterns are quite similar, showing that both reanalyses capture the broad character of differences in seasonal weather regimes.

2.1.2 The 2-m relative humidity

Figures 4 and 5 show the relative humidity (RH) distribution (with respect to water saturation) from ERA40, NCEP2 and the CRU data (over land) for the 10-year DJF and JJA climatologies. The differences between the reanalyses are small over the oceans. Over land, NCEP2 generally has a higher RH than ERA40. With respect to the CRU analysis, ERA40 is drier at most high northern latitudes in DJF, while elsewhere the ERA40 bias is small. The NCEP2 RH bias is generally positive over land. As a result, RH in NCEP2 is generally higher than in ERA40, which would correspond to a lower mean lifting condensation level. Figure 6 shows the anomaly fields for DJF 1991-1992, and JJA, 1988 (the same seasons as Figure 3). There is some correspondence in the anomaly patterns of RH: generally the ERA40 anomaly extremes are larger than those of the CRU dataset, and for NCEP2 they are larger still. Comparing Figures 3 and 6, we see the correspondence in the summer hemisphere of warm-dry and cool-wet anomalies, as expected: for example, the warm, dry conditions associated with the summer drought over the United States in JJA 1988.

However New et al. [2000] treat vapor pressure (from which we compute relative humidity, RH) as a secondary variable, and the gridded analysis uses not only station observations where available (converting monthly mean RH to vapour pressure for some stations) but also synthetic data (found by estimating monthly dewpoint from monthly mean minimum temperature)
in regions of limited observations. As a result there is considerably more uncertainty in their humidity analysis that in their temperature analysis, especially over regions of sparse data.

2.2 Precipitation fields

Precipitation in the reanalyses is most affected by the spin-up of the dynamic fields in mid-latitudes [Betts et al. 2003b]. For ERA-40, the ISLSCP-II dataset includes monthly precipitation computed from the 0-6, 0-12, 12-24 and 24-36h forecasts so user can assess this spinup. Betts and Beljaars [2003] includes some examples. For NCEP2, the ISLSCP-II dataset is derived from the 24-36h forecasts as discussed in section 1.2.2. In this section we compare the 24-36h forecast precipitation for both ERA40 and NCEP2 with the observationally-based GPCP data set, which combines a standard gauge analysis over land (with climatological bias correction) with a community-based satellite-only product to improve estimates where gauges are sparse. Thus, this GPCP analysis is globally complete, albeit with reduced confidence over regions such as open oceans and at high latitudes, where it is a satellite-only product. Figures 7 and 8 show the precipitation distribution from ERA40, NCEP2 and the GPCP data for the 10-year DJF and JJA climatologies. They have the same structure as earlier figures with the three climatologies on the left and the difference fields on the right. Although the climatologies are generally similar, the reanalyses have significant biases. Over the tropical oceans both reanalyses have more rainfall than GPCP, with ERA40 greater than NCEP2. Roads [2003] discusses the high bias of the NCEP2 reanalysis with respect to the Tropical Rainfall Measuring Mission (TRMM) satellite precipitation. The high bias of tropical precipitation in the ERA40 reanalysis, stems from a problem in the use of satellite radiances in the analysis of humidity [Troccoli and Källberg 2004]. ERA40 also has a negative DJF bias over the Amazon. For NCEP2, the biases over the tropics are
smaller than in ERA40. In mid-latitudes, the NCEP2 biases are generally positive over the oceans in the winter hemisphere, negative over the oceans in the summer hemisphere, and positive over the summer continents. The corresponding mid-latitude biases of ERA40 from GPCP are generally smaller. The difference fields between NCEP2 and ERA40 show that NCEP2 has generally more precipitation over the summer continents, and less over the tropical oceans; where there are also differences in the location and width of the convergence zones in the two reanalyses. Over Africa in JJA, the ITCZ precipitation in both reanalyses does not extend as far north as in the GPCP analysis. The biases for ERA40 are a little smaller with the 24-36h forecast precipitation than for the 0-6h forecast precipitation (not shown, although the ISLSCP-II dataset includes both for ERA40), so for most users 24-36h precipitation is the preferred choice for the reanalyses. ERA40 has a known error in the diurnal cycle of precipitation over land (a bias towards precipitation too early in the day) which is larger in the tropics [Betts and Jakob, 2002] than the mid-latitudes. Similar biases in the diurnal cycle exist for NCEP2. However, the seasonal means that we show average over the diurnal cycle in the reanalyses.

Figure 9 compares the ERA40, NCEP2 and GPCP precipitation anomalies for DJF, 1991-1992 and JJA, 1988. Despite the differences in their means, the anomaly patterns are remarkably similar, especially considering that the three analyses differ in their native horizontal resolution. Generally the anomalies for the higher resolution ERA40 are a little closer to the GPCP analysis than for NCEP2, which has generally slightly larger anomalies. Precipitation in the reanalyses is entirely a computed field, while the GPCP analysis is derived from a blend of satellite data for the cloud field and rain-gage data, primarily over land. Again if Figures 3, 6 and 9 are compared, we see coherent anomaly patterns in the summer hemispheres with high precipitation associated with cool-wet anomalies and the converse. This suggests that reanalyses have a good representation of the major circulation differences for the two seasons shown.
2.3 Surface radiation budget

The SRB dataset [Stackhouse et al. 2004; Cox et al. 2004] computes the SW and LW radiation fluxes from the ISCCP cloud data set [Rossow and Schiffer 1999], using the algorithms of Pinker and Laszlo [1992] and Gupta et al. [1992], and meteorological profile information from the GEOS-1 reanalysis [Schubert et al., 1995]. This is an earlier reanalysis than NCEP2 and ERA40, with a different set of biases in temperature and humidity. The ERA40 and NCEP2 reanalyses use their own radiation codes, meteorological profiles and model cloud fields to compute their radiation fluxes. The satellite-based datasets for shortwave are generally similar to each other, because they typically use cloud information from the International Satellite Cloud Climatology Project (ISCCP). One problem with the SRB SW dataset is that it under-estimates the downward shortwave flux over the central part of the Tibetan Plateau [Masuda, 2004], and the bias extends to a wide area of elevated terrain in western China, according to comparison between the SRB SW and an empirical estimation based on sunshine duration (calibrated with ground-based observations) by Xu et al [2005]. It is likely that the SRB retrieval model assumed too low values of clear-air transmittance for this region. It is not known whether this low bias extends to other regions of elevated terrain.

The downward surface SW flux, SWdown, can be written as the difference between the clear-sky flux and the surface SW cloud forcing (SSWCF):

\[ \text{SWdown} = \text{SWdown(clear)} + \text{SSWCF} \]

For both reanalyses and SRB dataset, SWdown(clear) is a model calculation from the top-of-the-atmosphere (TOA) incoming flux, modified by the atmospheric absorption. Since all three calculations use different radiation models, together with slightly different atmospheric structure
and composition, some small differences (not shown) can be expected in the their estimates of SWdown(clear). However, the primary differences between the three estimates of SWdown come from the computation of the impact of the cloud field, SSWCF. Whereas, the SRB estimate is based on the TOA flux measurements of the observed cloud field, ERA40 and NCEP2 calculate their cloud fields using their model parameterizations. We presume therefore that the SRB SWdown estimate is closer to the truth (since it is based on an observed cloud field); so that the model differences from SRB represent primarily model bias, associated with errors in the model cloud fields.

**2.3.1 Incoming short-wave radiation flux, SWdown**

Figure 10 shows (left) the 10-year climatology of surface SWdown for southern summer DJF from ERA40, NCEP2 and SRB and (right) the difference of ERA40 and NCEP2 from SRB climatology (top and middle right), and (bottom right) the difference, NCEP2-ERA40. The three difference patterns have some similarities. For DJF, SWdown for ERA40 is systematically low in the tropics (-10 to -50 Wm$^{-2}$), suggesting too much reflective cloud cover. SWdown is too large (suggesting too little cloud), for the southern ocean, for the stratocumulus areas off the western edge of continents, and for some continental regions. For NCEP2, the bias pattern is similar and the biases are generally larger (e.g. -20 to -90 Wm$^{-2}$ in the tropics), except over Antarctica (where the uncertainty of the SRB estimate becomes large because of the difficulty of distinguishing clouds over background ice). Figure 11 is the corresponding plot of SWdown for northern summer, JJA. Again we see that the reanalyses have low biases over the tropical oceans, and high biases over the stratocumulus regimes and parts of the northern continents. Again the NCEP2 biases are larger, especially over the northern continents, where clearly NCEP2 has too little cloud cover. The ERA40 biases over the northern continents are mixed and generally smaller. The low
bias of the SRB SWdown data over the Tibetan plateau [Masuda 2004 is consistent with the positive difference we see for ERA40-SRB. In the tropics, differences in the location and width of the ITCZ is responsible for the banded structure near the equator in the difference NCEP2-ERA40. Comparing Figures 10 and 11 with the corresponding Figures 7 and 8 for the precipitation bias shows that some (but not all) of the model errors in cloud cover over the tropical oceans are not surprisingly associated with errors in the model precipitation field. Figure 12 compares anomaly fields for ERA40, NCEP2 and SRB (from their own climatologies) for DJF, 1991-1992 (left) and (right) JJA 1988. Encouragingly, despite the biases in their climatologies, the major anomaly signals are similar in the two reanalyses and the SRB dataset except at very high latitudes. Comparing Figures 3, 6, 9 and 12, the coherence of the anomaly patterns can be again seen in the summer hemispheres, with the association of high SWdown with higher temperature but lower precipitation and RH. For this pair of seasons, the SRB anomalies are generally closer to the ERA40 anomalies over the tropical oceans.

Differences in SWup are not shown here: they reflect differences in background albedo, as well as the albedo with snow cover. The two reanalyses have different specified background albedos, and the SRB dataset albedo differs substantially from ERA40 [Betts and Beljaars 2003]. The ISLSCP-II data collection contains six other albedo products [Hall et al. 2006]; and this important surface parameter is still not known to sufficient accuracy for climate studies.

2.3.2 Incoming long-wave radiation flux, LWdown

LWdown depends on the atmospheric structure and composition, as well as on the cloud-base temperature, and retrievals from satellite data often differ significantly. Uncertainty in cloud base temperature is a source of uncertainty in the SRB dataset, and it uses atmospheric temperature and humidity profiles from GEOS-1, a different reanalysis. For the ERA40 and
NCEP2 reanalyses, differences in the LW flux are primarily related to differences in model cloud cover and atmospheric temperature and moisture structure. Generally LWdown will have a low bias if low- and medium-level cloud cover is underestimated, because these clouds are nearly black in the infrared.

Figure 13 shows the DJF climatology and difference fields, and Figure 14 is the corresponding JJA comparison. For ERA40 the differences from SRB are generally small over the oceans. The stratocumulus regions, which have too little low cloud, have correspondingly reduced LWdown in ERA40 and NCEP2. There are some regions in the sub-tropics where LWdown is higher in ERA40 and NCEP2, which correspond to regions with a high cloud bias. At higher latitudes, especially in the summer hemisphere, NCEP2 has a lower LWdown over the oceans than ERA40 and SRB (by -10 to -20 Wm\(^{-2}\)). For DJF over the northern continents, ERA40 has a higher LWdown than SRB, especially over western Eurasia; for NCEP2, this region of higher LWdown is further east. When we compare the difference field (NCEP2-ERA40 in Figure 13(f) with Figure 1(f) and Figure 10(f), we see the influence of the temperature difference in eastern Russia and the added impact of reduced cloud cover over Europe in NCEP2. For JJA over the northern continents, ERA40 and SRB agree quite closely, but NCEP2 has a lower LWdown (-10 Wm\(^{-2}\)) over land as well as over the ocean. The high values of SWdown in Figure 11(e) suggest that the bias of LWdown in Figure 14(e) is also caused by too little cloud cover. There are surprisingly large differences between the reanalyses over the Arctic in JJA: with NCEP2 values a little less than SRB and ERA40 much greater. At high latitudes in the winter hemisphere, the SRB estimates of the cloud field have larger uncertainties. In addition, large cold biases in winter in the skin temperatures in the GEOS-1 reanalysis adversely affect the SRB LWdown; so we cannot say with any confidence that the SRB LWdown is more accurate over the continents. Differences in the upward long wave fluxes reflect primarily differences in radiometric skin temperature, and we do
not show this comparison. The GEOS-1 reanalysis has cold biases in wintertime skin temperature, which give a large negative bias in upward LW flux [see Betts and Beljaars 2003]. In addition, there is still considerable uncertainty in model skin temperatures, because these depend strongly on poorly known (as well as conceptually questionable) roughness lengths for heat [e.g. Betts and Beljaars 1993].

Figure 15 shows that the anomaly fields for the two reanalyses and SRB are rather similar and rather small (< ±20W m⁻²), suggesting again that both reanalyses represent the major changes in the atmospheric circulation. Comparing Figure 3 and 15 for DJF, shows the association of the LWdown anomalies with temperature anomalies in winter.

2.3.3 Net radiation flux, Rnet

Figure 16 compares the JJA climatology and difference fields for JJA for Rnet. Comparing with the patterns in Figure 11 for SWdown and Figure 14 for LWdown, we see that Rnet differences are dominated by SWdown differences in the tropics, while LW flux differences are dominant at high latitudes. There are of course differences in the surface albedo and the upward LW fluxes, which we have not shown. Note the large differences in Rnet over the Sahara, with NCEP2 having the lowest values, which we shall see in the next section lead to low values of sensible heat flux.

3. Surface sensible and latent heat fluxes (SH and LH)

The partition of the surface Rnet into sensible and latent heat fluxes is of fundamental importance over land in driving the diurnal cycle of the atmospheric boundary layer. However the two reanalyses differ substantially in this energy partition, and we have no comparison ISLSCP-II
data product for evaluation. We therefore compared the sensible and latent heat fluxes from ERA40 and NCEP2 with each other, and over the oceans with the DaSilva climatology [DaSilva et al. 1994]. This covers the years 1945-1993, and has both the long-term mean and the monthly anomaly fields. For our comparison we use the 8-yr DaSilva climatology for 1986-1993. Six additional Figures for the climatologies, difference and anomaly fields for SH and LH are available in a supplementary file via Web browser or via Anonymous FTP from ftp://ftp.agu.org/apend/jd/2006JD007174" (Username = "anonymous", Password = "guest").

3.1 Sensible heat flux

ERA40 has a higher SH over most of the ocean in both seasons than the DaSilva climatology, except for a band in the southern ocean and over parts of the Gulf Stream and Kuroshio currents. In DJF in the Arctic Circle, ERA40 has a negative heat flux over the ice, whereas the DaSilva climatology has an unrealistically large upward heat flux. The NCEP2 SH fluxes are less than the DaSilva climatology over much of the oceans especially at mid-latitudes in winter; so that over most of the oceans NCEP2 has a smaller SH flux than ERA40. Over land as well the NCEP2 SH flux is generally less than ERA40, except for a few regions in summer. Over the Sahara NCEP2 has rather low values of SH heat flux in JJA because of the low values of Rnet (see Figure 16). The reanalyses have a similar structure in their anomalies however, although the NCEP2 anomalies are generally larger, consistent with Figure 3. The DaSilva anomaly fields have considerable uncertainty as observations are sparse in many regions, and we have averaged only three months. (See Figures S1, S2 and S3 in Supplementary file.)

3.2 Latent heat flux
Over the oceans, evaporation is generally higher in NCEP2 than ERA40 except in the summer hemisphere, where ERA40 is higher. There is a tendency for evaporation in ERA40 over the oceans to be higher in southern hemisphere mid-latitudes and lower in some regions of the tropics than the DaSilva climatology (which is again unrealistic over the Arctic in winter). Over land, NCEP2 has generally a higher latent heat flux than ERA40, which is consistent with its lower SH flux. The anomaly fields show some similarities between the reanalyses, but again the NCEP2 anomalies are larger than those of ERA40. There are considerable differences over the tropical oceans between the reanalyses for JJA 1988. Again, the DaSilva anomalies may be less reliable on a seasonal basis. The summer drought over the USA in 1988 is picked up in both reanalyses; with a stronger signal in the NCEP2 reanalysis (consistent with Figure 6). (See Figures S4, S5 and S6 in Supplementary file.)

4. Discussion

We have given a broad overview on seasonal timescales of differences at the surface by comparing the reanalysis fields with independent data from the ISLSCP-II data collection. Our purpose is to help the users of these data to assess which products might be useful for different purposes. In many cases the differences between reanalyses and other datasets give an estimate in our uncertainty in a given variable. Models not only assimilate data, but in regions of missing data a global model analysis will compute a complete set of fields using short-term forecasts of the model. Consequently, the ERA40 and NCEP2 reanalyses have biases that relate to their specific analysis-forecast systems and their choice of physical parameterizations. The comparison ISLSCP-II datasets each have their own issues of bias and representivity. In-situ surface
observations can be interpolated to a common grid, as is the case for our CRU dataset, but where observations are sparse; the gridded analysis may not be representative. The GPCP precipitation analysis is a blend of in-situ precipitation measurements and satellite data, and biases over the open oceans and the arctic regions are unknown. The SRB dataset is derived from satellite visible and infrared observations using parameterized radiation codes and atmospheric profiles coming from another earlier reanalysis with its own biases.

It is clear from the figures that there are systematic differences in the climatologies of the two reanalyses and the other ISLSCP-II data sets that we have shown for comparison. For temperature and RH, we compared the reanalyses with the CRU data set [New et al. 1999]. ERA40 has generally smaller biases of temperature over land than NCEP2. The CRU humidity analysis has more uncertainty in some regions because synthetic observations are added where there are few observations [New et al. 2000]. ERA40 has small biases in RH, often slightly negative, and NCEP2 has in many regions a small positive RH bias. The NCEP2 RH anomalies are larger than those for ERA40, which in turn are larger than those in the CRU analysis.

For precipitation we compared with the GPCP Satellite-Gauge combination monthly precipitation dataset [Adler et al. 2003]. Over the tropical oceans, both reanalyses have more rainfall than GPCP, with ERA40 greater than NCEP2. This is a known error in the ERA40 reanalysis [Troccoli and Källberg 2004]. ERA40 also has a negative bias over the Amazon in DJF. In mid-latitudes, the NCEP2 biases (from GPCP) are generally positive over the oceans in the winter hemisphere, negative over the oceans in the summer hemisphere, and positive over the summer continents. The corresponding mid-latitude biases from GPCP are generally smaller for ERA40.

The incoming radiation fluxes at the surface are a critical component of the surface radiation budget, which are heavily modified by clouds. So we compared the downward SW and
LW fluxes from the reanalyses with those in the SRB dataset [Stackhouse et al. 2004], which computes them from the ISCCP cloud data set [Rossow and Schiffer 1999] using atmospheric profiles from the GEOS-1 reanalysis [Schubert et al. 1995]. The SRB downward SW flux, which is derived from the observed cloud field, probably has smaller biases than the reanalysis fluxes, which calculate their cloud fields using their model parameterizations. The biases in the climatology of the reanalyses are significant: they suggest too much reflective cloud over the tropical oceans (more for NCEP2 than ERA40), except for the stratocumulus regimes, where reanalysis cloud cover is too low. Over the northern continents in summer, NCEP2 has too little cloud cover, so that downward SW is too large. The corresponding ERA40 biases are mixed and generally smaller.

The downward LW depends on the atmospheric structure and composition, as well as cloud-base temperature. Uncertainty in cloud base temperature is a source of uncertainty in the SRB dataset, as are atmospheric boundary layer temperatures, which come from the GEOS-1 reanalysis. For the reanalyses, differences in the LW flux are primarily related to differences in model cloud cover and atmospheric temperature and moisture structure. Generally, over the tropical oceans, the differences in downward LW for the reanalyses come from differences in cloud cover; so that there is a positive bias where model cloud cover is too high (trade-wind regimes) and a negative bias where model cloud cover is too low (stratocumulus regimes, and northern summer mid-latitudes for NCEP2). Over the winter continents, the SRB LW has significant biases coming from a cold bias in the near-surface temperatures in the GEOS-1 reanalysis. We showed a single comparison of the surface net radiation flux for the northern summer. In the tropics, differences in cloud cover lead to differences in SWdown and Rnet, while at high latitudes the substantial LW flux differences dominate. Rnet of course drives the important surface energy budget over land.
The two reanalyses differ considerably in their SH and LH fluxes over land, which are of importan
to ISLSCP; yet we have no comparison data set over land for evaluation. For many regions over land NCEP2 has more evaporation and reduced SH in comparison to ERA40, but larger seasonal anomalies. The larger seasonal anomalies for the NCEP2 reanalysis are consistent with Dirmeyer et al. [2004], who noted that the annual range of soil wetness was smaller for ERA40 than NCEP2. Ferranti and Viterbo [2006] concluded from a study of the European summer of 2003 that variability in the European Centre model climate is dampened by the soil moisture increments. It is likely that the different methods used in the reanalyses’ land surface models for controlling drifts of soil wetness (see sections 1.2.1 and 1.2.2) are responsible, so uncertainty remains in the energy flux partition over land.

These biases in the climatologies of the reanalyses, as well as the uncertainties in the datasets which we have compared them with, must be considered by users of the datasets in the ISLSCP-II collection. In contrast, a striking and encouraging feature of all our comparisons is that the anomaly fields derived from each climatology show similar features in both datasets and reanalyses. The anomalies are also coherent in the sense that we see warm, dry seasonal biases associated with reduced precipitation and cloudiness, and the converse. This confirms that major changes in the atmospheric circulation patterns, with the associated differences in surface temperature, humidity, precipitation, cloud fields and incoming surface SW and LW radiation fluxes are captured by both reanalyses and our other comparison ISLSCP-II datasets. Thus, the anomaly fields are useful in identifying seasonal changes in the climatological patterns. For precipitation and SWdown, the large-scale anomalies for the higher resolution ERA40 appear to be a little closer to the comparison ISLSCP-II data set than are the NCEP2 anomalies, but there are regions over land where this is not so. Our advice to the user of these ISLSCP-II data can only be very general, since these data will be used for such a wide variety of analyses.
Acknowledgments

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Table 1. Some differences between the NCEP2 and ERA40 reanalysis products

<table>
<thead>
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<th>Property</th>
<th>NCEP/DOE</th>
<th>ERA40</th>
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<td>T159 (approx. 125 km on a reduced</td>
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<td>resolution</td>
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<td>humidity errors</td>
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<td>Analyses &amp; 0-6h forecasts</td>
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<td>(plus 0-12, 12-24 and 24-36h, for</td>
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<td>precipitation)</td>
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List of Figures.

Figure 1(a), (b) and (c) 2-m temperature distribution from ERA40, NCEP2 and the CRU data (over land) for the 10-year (1986-1995) DJF climatologies and (d), (e) and (f) difference fields for ERA40-CRU, NCEP2-CRU, and NCEP2-ERA40.

Figure 2 As Figure 1 for JJA climatologies and difference fields.

Figure 3. 2-m temperature anomaly fields from ERA40, NCEP2 and the CRU data (over land) for (a), (b) and (c) DJF 1991-1992 and (d), (e) and (f) for JJA 1988.

Figure 4. As Figure 1 for 2-m RH

Figure 5. As Figure 2 for 2-m RH

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Figure 7. As Figure 1 for 24-36h forecast precipitation and GPCP precipitation.

Figure 8. As Figure 7 for JJA climatology.

Figure 9. As Figure 3 for precipitation anomalies.

Figure 10. As Figure 1 for surface SWdown from reanalyses and SRB data.

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Figure 16. As Figure 11 for surface Rnet.
Six additional Figures are included in this file 2006JD007174supplement.pdf. These go with the discussion in section 3.

Figure S1(a), (b) and (c). Sensible Heat (SH) from ERA40, NCEP2 and the DaSilva climatology over oceans for the 10-year (1986-1995) DJF climatologies and (d), (e) and (f) difference fields for ERA40-DaSilva, NCEP2-DaSilva, and NCEP2-ERA40.

Figure S2. As Figure S1 for JJA climatologies and difference fields.

Figure S3. SH anomaly fields from ERA40, NCEP2 and DaSilva over oceans for (a), (b) and (c) DJF 1991-1992 and (d), (e) and (f) for JJA 1988.

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Figure S5. As Figure S2 for LH

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