



RESEARCH ARTICLE

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Key Points:

- Net surface energy flux is reconstructed from satellite data and reanalyses
- High correlation between multiannual mean reconstruction and simulations
- Reduced heating of East Pacific Ocean 1986–2008 at odds with simulations

Supporting Information:

- Tables S1 and S2 and Figures S1–S4

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Combining satellite observations and reanalysis energy transports to estimate global net surface energy fluxes 1985–2012

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Abstract Two methods are developed to estimate net surface energy fluxes based upon satellite-derived reconstructions of radiative fluxes at the top of atmosphere and the atmospheric energy tendencies and transports from the ERA-Interim reanalysis. Method 1 applies the mass-adjusted energy divergence from ERA-Interim, while method 2 estimates energy divergence based upon the net energy difference at the top of atmosphere and the surface from ERA-Interim. To optimize the surface flux and its variability over ocean, the divergences over land are constrained to match the monthly area mean surface net energy flux variability derived from a simple relationship between the surface net energy flux and the surface temperature change. The energy divergences over the oceans are then adjusted to remove an unphysical residual global mean atmospheric energy divergence. The estimated net surface energy fluxes are compared with other data sets from reanalysis and atmospheric model simulations. The spatial correlation coefficients of multiannual means between the estimations made here and other data sets are all around 0.9. There are good agreements in area mean anomaly variability over the global ocean, but discrepancies in the trend over the eastern Pacific are apparent.

1. Introduction

The absolute mean value of net radiation imbalance at the top of atmosphere (TOA) is a key climate variable, providing an estimate of total energy gain of the Earth system and a link between radiative forcing, ocean heat uptake, and surface temperature response. It has been estimated to be 0.5 to 1 W/m² for the global mean in recent studies [Hansen et al., 2011; Loeb et al., 2012; Trenberth et al., 2014; Wild et al., 2015] using changes in total ocean heat content [Lyman and Johnson, 2014; Trenberth et al., 2014; Smith et al., 2015; Roemmich et al., 2015] and making assumptions about minor energy sinks including the land, the atmosphere, and the cryosphere. Although satellite data provide regional coverage of top of atmosphere radiative fluxes, the net surface fluxes display much larger uncertainty due to the lack of constraints from global observations [Trenberth et al., 2009; Wild et al., 2013, 2015].

The net energy fluxes at the Earth's surface, including short- and long-wave radiation and the sensible and latent heat fluxes, are very important for the study of surface temperature change and the atmospheric and oceanic circulations. The surface fluxes also control the water cycle since the incoming short-wave radiation provides much of the energy required for surface water evaporation. Net downward surface energy can accumulate within the ocean, leading to long time scale effects on the climate. Therefore, accurate estimation of the surface energy fluxes is essential for understanding both the short-term temperature hiatus [Easterling and Wehner, 2009; Knight et al., 2009; Trenberth and Fasullo, 2013a; Huber and Knutti, 2014; Watanabe et al., 2014] and long-term climate change [Otto et al., 2013]. It is difficult to obtain accurate absolute surface fluxes from satellites due to complicated atmospheric conditions affecting the retrieval processes in particular relating to the numerous surface variables required by turbulent flux bulk formulae [Schmetz, 1991].

The net input of radiation fluxes at TOA are modulated by the atmosphere and redistributed by lateral energy transports [Keith, 1995; Chiodo and Haimberger, 2010; Lucarini and Ragone, 2011; Trenberth and Fasullo, 2013a; Mayer and Haimberger, 2012; Mayer et al., 2014; England et al., 2014; Loeb et al., 2015]. Meehl et al. [2011] and

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Trenberth and Fasullo [2013b] also demonstrated that the vertical energy redistribution in the oceans is likely to have contributed substantially to the slowing in the rate of global average surface temperature increase in the last 15 years. Assessment of where the net accumulation of energy in the climate system is being stored within ocean basins [*Balmaseda et al.*, 2013; *Drijfhout et al.*, 2014; *Llovel et al.*, 2014; *Desbruyères et al.*, 2014; *Roemmich et al.*, 2015] is required for understanding the mechanisms of energy redistribution associated with internal variability and therefore the surface temperature variations.

The currently available surface flux data sets have some limitations. Observed data from in situ measurements are sparsely distributed in space, while satellite-derived retrievals contain substantial uncertainties and require further validation. Observationally based data, reanalysis estimates, and climate model simulations show a large spread in the data and large unrealistic global imbalances when turbulent and radiative flux products are combined [*Trenberth et al.*, 2009; *Stephens et al.*, 2012; *Wild et al.*, 2013]. In this study, we apply an atmospheric energy divergence approach [*Chiodo and Haimberger*, 2010; *Mayer and Haimberger*, 2012] using two different methods to estimate the net downward surface energy fluxes by combining reconstructed net radiation fluxes at TOA [*Allan et al.*, 2014] with the energy tendencies and lateral divergence simulated by the ERA-Interim reanalysis [*Dee et al.*, 2011; *Berrisford et al.*, 2011].

2. Data and Methods

2.1. Data Sets

The key data set is the ECMWF (European Centre for Medium-Range Weather Forecasts) ERA-Interim reanalysis (ERA-Interim) [*Dee et al.*, 2011]. Various observational data are assimilated to a weather forecast model to provide representations of atmospheric states. Although it has some known problems, such as the lack of volcanic aerosols and the omission of the 11 year solar cycle [*Dee et al.*, 2011], it provides a comprehensive representation of atmospheric variables and estimates of energy divergences and fluxes required for this study. The net radiation flux at TOA is based on the recent reconstruction by *Allan et al.* [2014] using satellite observations from the Clouds and the Earth's Radiant Energy System (CERES) [*Loeb et al.*, 2012] and Earth Radiation Budget Satellite wide field of view (WFOV; 72 day mean) nonscanning instrument [*Wong et al.*, 2006], ERA-Interim reanalysis, and climate model simulations applying the Atmospheric Model Intercomparison Project 5 (AMIP5) experimental setup with prescribed observed sea surface temperature (SST) and sea ice and realistic radiation forcings [*Taylor et al.*, 2012]. The net TOA flux is adjusted to ensure agreement with an observational estimate over the period 2005–2010, primarily determined by observed 0–2000 m ocean heating rate [*Loeb et al.*, 2012; *Allan et al.*, 2014]. The TOA reconstructions are updated using the latest version (version 2.8) of CERES data. Another important update from *Allan et al.* [2014] is that prior to March 2000, reconstructed radiative fluxes are adjusted separately for each hemisphere rather than applying a global adjustment. This adjustment ensures that deseasonalized anomalies in radiative fluxes match the WFOV variability for 0–60°S and 0–60°N regions. Further details of the additional adjustment procedures are described in *Allan et al.* [2014]. The updated net downward TOA radiation flux will be referenced as F_T .

Sixteen AMIP5 models are used in this study, and one member from each model is chosen. Data from a five-member ensemble of the UPSCALE (UK on Partnership for Advanced Computing in Europe (PRACE): weather-resolving Simulations of Climate for global Environmental risk) [*Mizielinski et al.*, 2014] simulations are also used here. UPSCALE is from a global atmospheric model (Hadley Centre Global Environmental Model version 3 global atmosphere-only simulations (HadGEM3-A-GA3) [*Walters et al.*, 2011]) at 25 km resolution, which is employed to produce an extended AMIP simulation up to 2011 using the Operational Sea Surface Temperature and Sea Ice daily high-resolution Analysis [*Donlon et al.*, 2012]. The only differences between these five-ensemble-member runs are their initial conditions: each member was perturbed by randomly altering the lowest-order bit in the 3-D potential temperature field.

The recently available ECMWF twentieth century atmospheric reanalysis from ERA-CLIM (European Reanalysis of Global Climate Observations) project (hereafter ERA20C) is used here for comparison purpose; it is a single-member reanalysis, and it assimilates observations of surface pressure and surface marine winds; SST, sea ice, and realistic radiative forcings are prescribed [*Poli et al.*, 2013]. The atmospheric energy divergence from the MERRA (Modern-Era Retrospective Analysis for Research and Applications) reanalysis is also used for the comparison of net surface energy fluxes. A large quantity of observational data is assimilated in the MERRA system using a three-dimensional variational data assimilation analysis algorithm [*Rienecker et al.*, 2011].

Table 1. Data Sets and Their Properties

Data Set	Period	Resolution (Latitude × Longitude)	References
CERES EBAF v2.8	2000–2012	1.0° × 1.0°	Loeb <i>et al.</i> [2012]
Reconstruction	1985–2012	1.0° × 1.0°	Allan <i>et al.</i> [2014]
ERA-Interim	1985–2012	0.7° × 0.7°	Dee <i>et al.</i> [2011]
ERA20C	1985–2010	0.7° × 0.7°	Poli <i>et al.</i> [2013]
MERRA	1985–2012	0.5° × 0.7°	Rienecker <i>et al.</i> [2011]
HadCRUT4 v4.2.0.0	1985–2012	5° × 5°	Morice <i>et al.</i> [2012]
AMIP5 models	1985–2008		
ACCESS1-0		1.25° × 1.875°	Bi <i>et al.</i> [2013]
CanAM4		2.79° × 2.81°	Arora <i>et al.</i> [2011]
CCSM4		0.94° × 1.25°	Gent <i>et al.</i> [2011]
CMCC-CM		0.75° × 0.75°	Scoccimarro <i>et al.</i> [2011]
CNRM-CM5		1.40° × 1.41°	Voltaire <i>et al.</i> [2012]
CSIRO-Mk3-6-0		1.87° × 1.875°	Rotstayn <i>et al.</i> [2010]
FGOALS-s2		1.66° × 2.81°	Li <i>et al.</i> [2013]
GFDL-CM3		2.0° × 2.5°	Delworth <i>et al.</i> [2006]
GISS-E2-R		2.0° × 2.5°	Schmidt <i>et al.</i> [2014]
HadGEM2-A		1.25° × 1.875°	Collins <i>et al.</i> [2011]
INM-CM4		1.5° × 2.0°	Volodin <i>et al.</i> [2010]
IPSL-CM5A-LR		1.89° × 3.75°	Dufresne <i>et al.</i> [2013]
MIROC5		1.39° × 1.41°	Watanabe <i>et al.</i> [2011]
MPI-ESM-LR		1.85° × 1.875°	Raddatz <i>et al.</i> [2007]
MRI-CGCM3		1.11° × 1.13°	Yukimoto <i>et al.</i> [2012]
NorESM1-M		1.89° × 2.5°	Zhang <i>et al.</i> [2012]
UPSCALE	1985–2011	0.35° × 0.23°	Mizielinski <i>et al.</i> [2014]

Observed surface temperature data are from Hadley Centre/Climatic Research Unit gridded surface temperature data set 4 (HadCRUT4) [Morice *et al.*, 2012]. All data used in this study are monthly mean diagnostics accumulated from higher time resolution data and are listed in Table 1.

2.2. Methods

2.2.1. Surface Energy Flux From Mass-Adjusted Divergence

Following Berrisford *et al.* [2011], the total energy (E) in an atmospheric column can be written as

$$E = \frac{1}{g} \int_0^1 (Lq + C_p T + \varphi_s + k) \frac{\partial p}{\partial \eta} d\eta, \quad (1)$$

where L , q , C_p , T , φ_s , and k are the latent heat of condensation of water, specific humidity, the specific heat capacity of air at constant pressure, temperature, surface geopotential, and kinetic energy ($(\mathbf{V} \cdot \mathbf{V})/2$; \mathbf{V} is the horizontal wind velocity vector), respectively. p is the pressure, g is the gravitational acceleration, and η is the hybrid vertical coordinate which is a function of pressure and surface pressure [Simmons and Burridge, 1981]. The total energy tendency $\frac{\partial E}{\partial t}$ in each atmospheric column can be expressed as

$$\frac{\partial E}{\partial t} = -\nabla \cdot \frac{1}{g} \int_0^1 \mathbf{V}(h + k) \frac{\partial p}{\partial \eta} d\eta + F_A. \quad (2)$$

The total energy input to the atmosphere $F_A = F_T - F_S$, where F_T is the net downward radiation flux (difference between the absorbed solar radiation and the outgoing long-wave radiation) at TOA and F_S is the net downward energy flux including contributions from both radiation flux and turbulent heat fluxes at surface. The moist static energy $h = Lq + C_p T + \varphi$ (φ is the geopotential). Note that a further term could be added to the right-hand side of equation (2), to represent a budget residual, which in reanalysis data would be due to analysis increments and numerical effects. Rearranging equation (2) allows F_S to be obtained from

$$F_S = F_T - \frac{\partial E}{\partial t} - \nabla \cdot \frac{1}{g} \int_0^1 \mathbf{V}(h + k) \frac{\partial p}{\partial \eta} d\eta. \quad (3)$$

The total energy tendency, $\frac{\partial E}{\partial t}$, is small compared with other terms and can be calculated from time series of E computed from ERA-Interim analyses, while $\nabla \cdot \frac{1}{g} \int_0^1 \mathbf{V}(h + k) \frac{\partial p}{\partial \eta} d\eta$ is the energy divergence (E_D). The horizontal

flux in E_D is not simply the flux of total energy from equation (1) but incorporates the flux of enthalpy [Boer, 1982; Trenberth and Solomon, 1994].

For mass consistency, the output E_D from ERA-Interim should be mass adjusted, because during the assimilation procedure, observations reset the surface pressure field, whereas the mass fluxes are not adjusted accordingly [Graversen et al., 2007; Berrisford et al., 2011]. Based on Mayer and Haimberger [2012],

$$E_{D\text{mass}} = E_D - (\bar{h} + \bar{k})(M_{\text{DIV}} + M_{\text{TEND}} + P - E_{\text{vap}}), \quad (4)$$

where M_{DIV} and M_{TEND} are the vertically integrated total mass divergence and tendency obtained from the ERA-Interim reanalyses. The difference between evaporation (E_{vap}) and precipitation (P) is calculated from total column water vapor (w) content based on the method of Trenberth et al. [2001]:

$$E_{\text{vap}} - P = \frac{\partial w}{\partial t} + \nabla \cdot \frac{1}{g} \int_0^1 q \mathbf{V} \frac{\partial p}{\partial \eta} d\eta = w_{\text{TEND}} + w_{\text{DIV}}, \quad (5)$$

where w_{DIV} is the vertically integrated water vapor divergence and w_{TEND} is the total column water vapor tendency which can be calculated from the time series of total column water vapor content. Both are obtained from ERA-Interim; this method is considered more accurate than using $E_{\text{vap}} - P$ directly from the reanalysis, since water vapor is assimilated, but precipitation is a simulated variable that is highly dependent upon model parameterizations. It includes water mass transfer due to phase change between water vapor and liquid water. The phase change between liquid water and ice in the atmosphere has been ignored, and the horizontal water transport due to cloud advection is also neglected since these terms are small, \bar{h} and \bar{k} are the vertical average of moist static energy and kinetic energy, respectively, which can be computed from analyzed ERA-Interim fields.

From equation (3), we can have the net surface energy flux from the mass-adjusted energy divergence (F_{mass}):

$$F_{\text{mass}} = F_T - \frac{\partial E}{\partial t} - E_{D\text{mass}}. \quad (6)$$

Similar procedures are applied to MERRA data [Mayer et al., 2013] to obtain mass-adjusted total energy divergence $E_{D\text{mass}} - \text{MERRA}$ which is substituted into equation (6) to obtain the net downward surface flux $F_{\text{mass}} - \text{MERRA}$.

2.2.2. Surface Energy Flux From Model Residual Divergence

Another way to estimate the atmosphere energy divergence is to calculate it directly from ERA-Interim as a residual of energy fluxes [Chiodo and Haimberger, 2010; Mayer and Haimberger, 2012]:

$$E_{D\text{res}} = F_{T-\text{ERA}} - F_{S-\text{ERA}} - \left(\frac{\partial E}{\partial t} \right)_{\text{fc}}, \quad (7)$$

where $F_{T-\text{ERA}}$ and $F_{S-\text{ERA}}$ are the energy fluxes at the TOA and surface computed directly from the ERA-Interim 12 hourly forecasts, where their radiation components (short-wave and long-wave) are calculated from the radiation transfer model based on the atmospheric states. $F_{S-\text{ERA}}$ also includes turbulent fluxes simulated by the reanalysis. The term $\left(\frac{\partial E}{\partial t} \right)_{\text{fc}}$ is mass-corrected forecasted total energy tendency [Mayer and Haimberger, 2012] and is preferred over analyzed tendencies to be consistent with forecasted TOA and surface fluxes. The calculated $E_{D\text{res}}$ can be used to estimate the surface flux (F_{res}) using the reconstructed TOA flux and total energy tendency from ERA-Interim analyses.

$$\begin{aligned} F_{\text{res}} &= F_T - \frac{\partial E}{\partial t} - E_{D\text{res}} \\ &= F_{S-\text{ERA}} + (F_T - F_{T-\text{ERA}}) + \left(\frac{\partial E}{\partial t} \right)_{\text{fc}} - \frac{\partial E}{\partial t} \end{aligned} \quad (8)$$

The accuracy of this divergence relies on the accuracy of the atmospheric properties, the radiative transfer through the atmosphere, and the turbulent energy calculations at the surface. It is known that ERA-Interim does not represent aerosol forcing due to volcanic eruptions, most notably following the Mount Pinatubo eruption [Allan et al., 2014], which might affect the divergence ($E_{D\text{res}}$) accuracy since the radiation fluxes are affected by aerosols. Although the constraint on divergence is poor, hence the need for mass adjustment, data assimilation constrains parameters toward an observed atmospheric state; with the inclusion of analysis increment, $\left(\frac{\partial E}{\partial t} \right)_{\text{fc}} - \frac{\partial E}{\partial t}$, the effect of aerosol-related biases on F_{res} will be reduced.

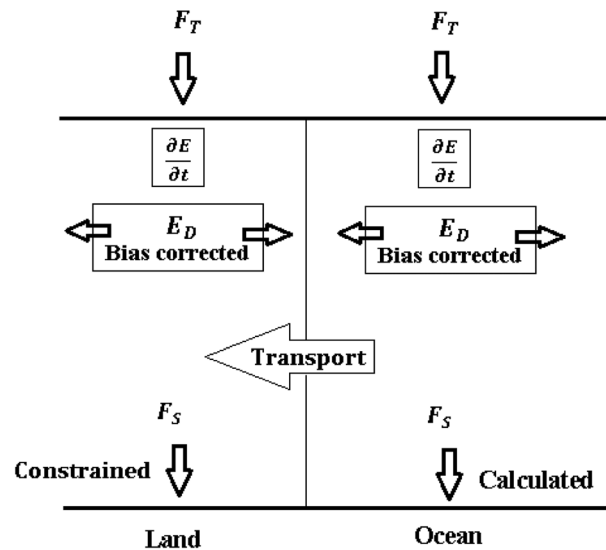


Figure 1. Schematic illustrating the energy flow terms used in the estimation of surface energy fluxes over land and ocean.

local changes (2001–2008 mean minus 1986–2000 mean; not shown here) and the global mean RMS (root-mean-square) of the multiannual mean differences (2001–2008 mean minus 1986–2000 mean) is about 8.5 Wm^{-2} . The area mean $E_{D_{\text{mass}}}$ over land is also large (about 2 Wm^{-2} over 2001–2005). A strategy was required to address these problems. The schematic diagram shown in Figure 1 illustrates the energy flow terms used in the estimation of net surface energy fluxes. The left and right columns depict the energy flow over land and ocean, respectively, and there is a net energy transport from the ocean column to land column [Wild et al., 2015]. The steps for estimating the monthly net surface energy fluxes are as follows:

1. Remove the global mean divergence from $E_{D_{\text{mass}}}$.
2. We already have F_T and $\frac{\partial E}{\partial t}$; assuming that we have the correct monthly net surface energy flux data over land, the monthly vertically integrated energy divergence can be calculated over land using energy balance equation.
3. The globe is divided into 15° latitude band (30° over Antarctic). The mean discrepancy between mass-corrected divergence and the one derived from step (2) over land is redistributed evenly over ocean grid points to keep the total divergence unchanged across each band.
4. The monthly net surface energy flux over the ocean can then be calculated using bias-corrected divergence.

In step (2), it will be ideal to use net surface energy flux calculated from $E_{D_{\text{mass}}}$ as the initial estimation over land, but as mentioned above the derived fluxes have unrealistically large regional changes (2001–2008 mean minus 1986–2000 mean) over land, so the surface energy flux from ERA-Interim ($F_{S-\text{ERA}}$) over land is used as the initial estimation. In order to correct the unrealistic trend and large anomaly variability of $F_{S-\text{ERA}}$ as discussed in section 3.3 (which would imply large unrealistic temperature variations or land heat capacity), a simple method described in the next section is applied to estimate the monthly net energy flux variability based on the observationally constrained surface temperature changes over land.

2.4. Net Energy Flux Over Land

The mean global land flux is estimated using the simple relationship of $F_S = C \frac{\Delta T}{\Delta t} + \varepsilon$, where C is the effective mean surface land heat capacity, $\frac{\Delta T}{\Delta t}$ is the global land mean surface temperature change rate, and ε is a constant indicating the energy flux penetrating beneath the surface layer. Data from five UPSCALE ensemble members are used for this estimation. The land surface model in UPSCALE simulations is JULES (Joint UK Land Environment Simulator), which has an explicit representation of the surface energy balance for vegetation, capturing the weaker coupling that exists between the canopy and underlying soil [Best et al., 2011]. The effective land heat capacity depends on the soil and canopy properties and the soil water content. After

2.3. Adjustment Constraints

Since a large quantity of observational data are assimilated into ERA-Interim, it is expected that both $E_{D_{\text{mass}}}$ and $E_{D_{\text{res}}}$ will provide reasonable spatial structures, but the $E_{D_{\text{res}}}$ has a multiannual (2001–2005) global mean value of -0.9 Wm^{-2} which is not physically reasonable since it is expected that the global averaged E_D should be zero to guarantee energy conservation. This is because atmospheric models do not, in general, have a closed budget for the atmospheric energy, as a result of inconsistent treatment of turbulent cascades of kinetic energy and water mass [Lucarini and Ragone, 2011; Liepert and Previdi, 2012; Lucarini et al., 2014]. Even though the global mean $E_{D_{\text{mass}}}$ is close to zero ($\sim 10^{-4}$), the net surface flux derived from it has unrealistically large

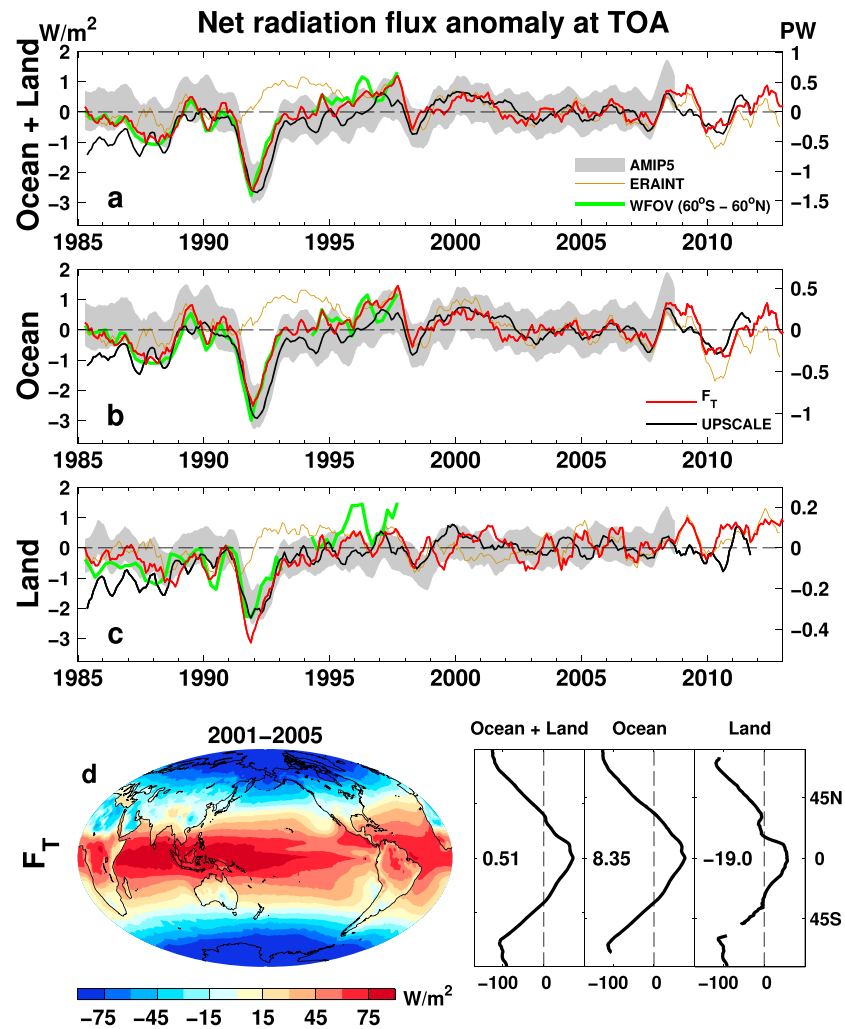


Figure 2. Deseasonalized anomaly (relative to the 2001–2005 period) time series of mean net downward radiation fluxes at TOA over (a) the globe, (b) the global ocean, and (c) the global land, for data sets of AMIP5, ERAINT, WFOV, F_T , and UPSCALE. The shaded areas of AMIP5 are 16-member ensemble mean ± 1 standard deviation. All lines are 6 month running mean. The WFOV anomaly (60°S–60°N) is relative to 1985–1999 period; its line is three data point (three 72 day means) running mean and is adjusted to match F_T . The y axis unit is W/m^2 on the left and PW on the right. (d) The multiannual (2001–2005) mean from F_T . The area mean (W/m^2) is displayed in the zonal mean plot.

testing we found high correlations between energy flux and the rate of surface temperature change if $\frac{\Delta T}{\Delta t}$ is calculated from consecutive months; e.g., the climatology of F_5 in April will correlate well with $\frac{\Delta T}{\Delta t}$ calculated from the climatology difference between April and March, so the effective land heat capacity C and the constant ε are calculated by regression using the climatology of F_5 and climatological $\frac{\Delta T}{\Delta t}$. The anomaly time series from modeled and reconstructed (from C , $\frac{\Delta T}{\Delta t}$, and ε) land surface mean fluxes are plotted in Figure S1 in the supporting information. The correlation coefficients (r) between monthly anomalies (reference period 2001–2005) are all above 0.6. The plotted lines are 6 month running means, and the inflated reconstructed lines (red) are multiplied by the ratio of the standard deviation between modeled and reconstructed monthly flux anomalies (values in red in the plot). The variability in F_5 is generally well captured, although there are exceptions, notably over the Mount Pinatubo eruption period since the constant seasonal C is used while in reality it should vary under anomalous situations; as discovered by *Iles and Hegerl* [2014], the models underestimate the precipitation over Pinatubo eruption period which affects the soil moisture content, therefore affecting the relations between temperature change and energy fluxes. Another factor affecting the net surface energy flux variability is the snow and ice melting. While there are considerable limitations, this

method was applied to ensure that large biases in the variability in F_S over land did not diminish the realism of diagnosed F_S over ocean which is the goal of the present study.

Five sets of the regression coefficients from five UPSCALE members using the above method are applied to the global land mean surface temperature (skin temperature) rate of change $\frac{\Delta T}{\Delta t}$ from ERA-Interim to get five proxies of mean surface flux; the ensemble mean is used as our estimated global land mean surface net energy flux. Based on *Beltrami et al.* [2002], the mean net energy flux over the continental lithosphere is 0.0391 W/m^2 over 1950–2000, where the mean land surface temperature change from HadCRUT4 [*Morice et al.*, 2012] is about 0.0138 K/yr (from regression). Based upon the 1985–2012 mean surface temperature change of 0.0298 K/yr from HadCRUT4 we estimate the mean of the reconstructed net surface flux as 0.08 W/m^2 over this period. Setting this flux to zero is also reasonable [*Trenberth et al.*, 2009]. Combining algorithms in sections 2.3 and 2.4, the estimated 2-D net surface energy fluxes over land maintain the spatial structure of $F_S - \text{ERA}$, but the monthly area-weighted mean values match those from the simple model ($F_{\text{SFC}} = C \frac{\Delta T}{\Delta t} + \varepsilon$) and the long-term mean (1985–2012) is anchored to 0.08 W/m^2 .

3. Net Downward Energy Fluxes

3.1. Net Radiation Flux at TOA

The reconstructed net downward radiation flux anomalies at TOA are updated from *Allan et al.* [2014] using the latest version (version 2.8) of CERES data and adjusting pre-CERES variability to match the interannual anomalies from the WFOV instrument for each hemisphere separately rather than using the 60°S – 60°N near-global mean. The TOA flux anomaly time series are plotted in Figure 2 for the global mean, the global ocean, and the global land, respectively. The reference period is from 2001 to 2005, but WFOV has a reference period of 1985–1999 and is adjusted, for clarity, to match the mean F_T (reconstruction) anomaly over this period. There is good agreement between variability depicted by F_T and the other data sets over the global ocean and the globe. The correlation coefficients (r) between F_T and ERAINT, UPSCALE, or AMIP5 monthly anomaly time series are 0.63, 0.60, and 0.58 over the global ocean and 0.64, 0.44, and 0.46 over the land, respectively. All these correlations are significant based on the two-tailed test using Pearson's critical values at the level of 5%. The degree of freedom of the time series is calculated by first determining the time interval between effectively independent samples [*Yang and Tung*, 1998] but additionally assuming that periods separated by 12 or more months are independent. Although ERAINT does not represent changes in aerosol emissions, most notably following the Mount Pinatubo eruption in 1991, the correlation coefficient between F_T and ERAINT is still the highest. This reflects the realistic monthly variability of atmospheric circulation patterns through the extensive assimilation of conventional and satellite data by ERA-Interim.

The area-weighted multiannual mean net downward energy fluxes from F_T (Figure 2d) over 2001–2005 are 0.51, 8.35, and -19.0 W/m^2 for the globe, the global ocean, and the global land, respectively. The difference is mainly due to the albedo difference between the land and the ocean. The large-energy deficit over land should be compensated by the horizontal energy transport from ocean to land [*Mayer and Haimberger*, 2012; *Trenberth and Fasullo*, 2013b].

3.2. Net Energy Flux at the Surface

The multiannual mean (2001–2005) net surface energy fluxes from F_{mass} are plotted in Figure 3a, and zonal mean variations from F_{mass} , F_{res} , $F_{\text{mass}} - \text{MERRA}$, ERAINT, ERA20C, UPSCALE, and AMIP5 data sets are plotted in Figures 3b–3d. The area-weighted means are displayed in the zonal mean plot. The multiannual means for other data sets are in Figure S2. F_{mass} and $F_{\text{mass}} - \text{MERRA}$ are calculated from the spatially filtered $E_{D_{\text{mass}}}$ and $E_{D_{\text{mass}} - \text{MERRA}}$, respectively, using a Hoskins spectral filter [*Sardeshmukh and Hoskins*, 1984] with an attenuation of 0.1 at wave number 106 [*Berrisford et al.*, 2011]. A filter is necessary due to the noise generated by data assimilation, highlighting that spatial patterns must be interpreted with caution.

Despite the contrasting methods and data sets, the multiannual means for the period 2001–2005 from all data sets show similar spatial structures and zonal means except for the MERRA data which show much stronger fluxes over the central Indian Ocean and central western Pacific. The spatial correlation coefficients of multiannual means between estimations and other data sets are all around 0.9. Over the oceans, despite ~ 10 – 20 W/m^2 differences present in the zonal means (Figure 3c), all data sets capture the positive

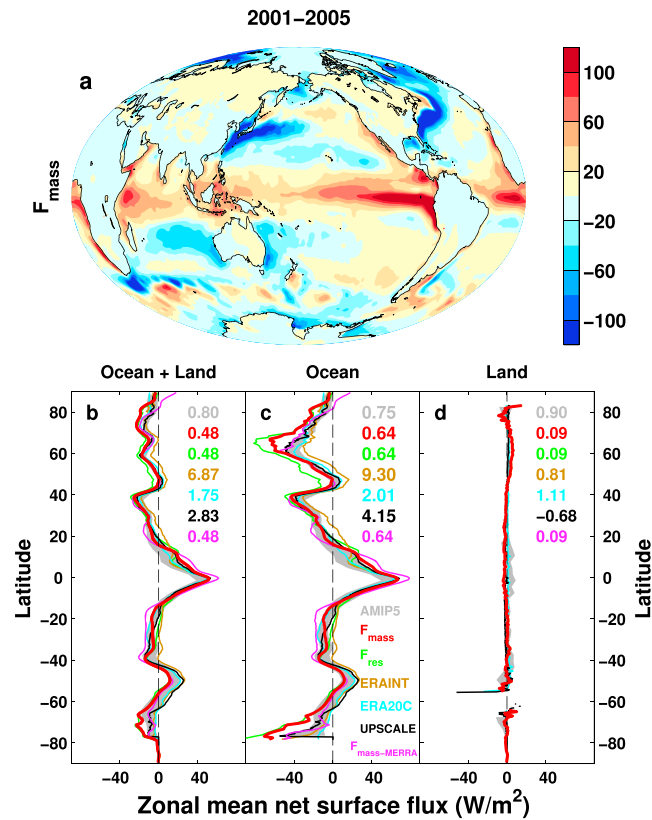


Figure 3. (a) Multiannual (2001–2005) mean net downward energy fluxes (in W/m^2) at surface from F_{mass} . Zonal mean variations from AMIP5, F_{mass} , F_{res} , ERAINT, ERA20C, UPSCALE, and $F_{mass} - MERRA$ are in the bottom plot for (b) the globe, (c) the global ocean, and (d) the global land, respectively. The shaded areas of AMIP5 are 16-member ensemble mean ± 1 standard deviation. The area mean is displayed in the zonal mean plot.

by single grid points at the southern tip of South America and northern tip of the Antarctic Peninsula require further investigation.

The mean northward total meridional atmospheric energy transports calculated from E_{Dmass} , E_{Dres} , and $E_{Dmass} - MERRA$ are also plotted in Figure 4a. Peak magnitudes of around 5 PW (1 PW = 10^{15} W) close to 40°S and 40°N are broadly consistent with Mayer and Haimberger [2012] and Lucarini and Ragone [2011] and coincide with the maximum in baroclinic activity [Lucarini and Ragone, 2011]. The transport from $E_{Dmass} - MERRA$ has stronger magnitude at 40°S/40°N compared with the other estimates. The transport from E_{Dmass} is of larger magnitude than that from E_{Dres} in the northern and southern hemisphere subtropics, consistent with Mayer and Haimberger [2012].

Due to flux constraints over land, the area mean fluxes from both F_{mass} and F_{res} are identical. Their spatial structures and zonal mean variations are also very similar (Figure 3 and Figure S2a), but the magnitudes differ in places as shown in Figure 5a. F_{res} is larger in magnitude than F_{mass} in the South Indian Ocean but smaller in the North Indian Ocean. F_{res} is smaller over the central, western, and northwest Pacific but has larger values over the subtropical gyre of North Pacific, as well as over southeast Pacific.

Although the mean surface flux spatial structure of ERAINT (Figure S2b) is similar to the derived ones, its area mean fluxes are unrealistically large over the global ocean ($9.30 Wm^{-2}$ in Figure 3c) compared with ocean observations [Llovel et al., 2014; Roemmich et al., 2015] which are on the order of 0–1 W/m^2 . ERA-Interim surface fluxes are substantially larger than F_{mass} over the oceans as shown in Figure 5b, except for the area near the equator, and this can be seen clearly from the zonal mean variations (Figure 3c). ERA20C simulates large fluxes into the Southern Ocean, more flux from ocean to atmosphere over the whole Indian Ocean and

downward energy flux over the equatorial central and East Pacific areas due to the interaction between the tropical instability waves [Willett et al., 2006] and the equatorial Pacific cold tongue [Martínez-García et al., 2010] controlled by ocean mixing [Moum et al., 2013]. The evaporation is less, and there is lower outgoing long-wave radiation over this cold region compared with surrounding regions. The negative downward fluxes over the Gulf Stream in the North Atlantic and Kuroshio Currents in the North Pacific are due to heat and moisture transports from the warm ocean surface to the cold atmosphere above [Kwon et al., 2010]. Over the global land, the UPSCALE simulation has a similar large-magnitude residual flux ($-0.68 W/m^2$) to the ERAINT flux ($0.71 W/m^2$) because it does not have a closed energy budget [Lucarini and Ragone, 2011]. This is in part because the high-resolution version of the UPSCALE simulations used was not recalibrated using observations since a key aim of this project was to understand the influence of resolution upon mean climate. The unrealistically large-magnitude values at around 55 and 65°S (Figure 3d) that are caused

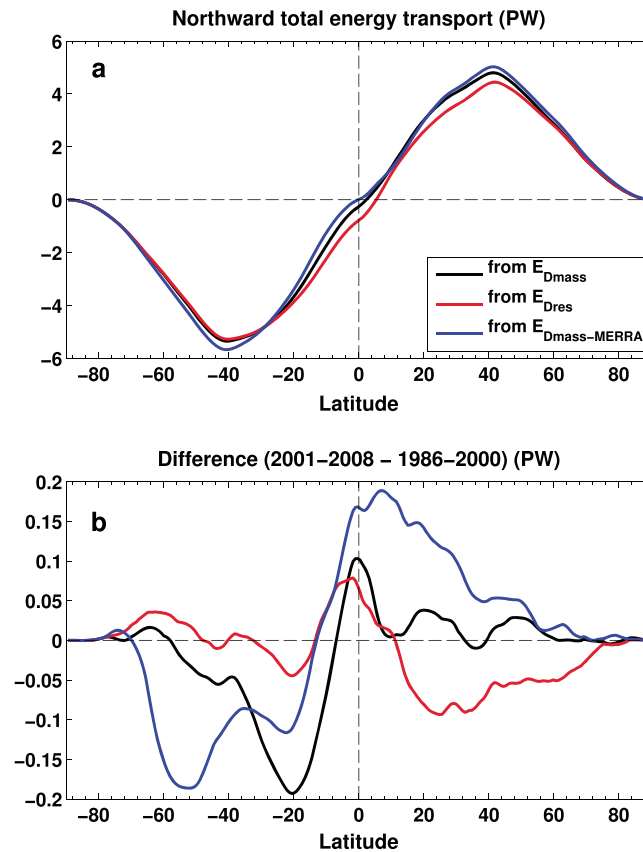


Figure 4. (a) Multiannual mean (2001–2005) northward total meridional energy transport (unit is PW) calculated from E_{Dmass} , E_{Dres} , and $E_{Dmass-MERRA}$ and (b) multiannual mean difference (2001–2008 minus 1986–2000).

[Lucarini and Ragone, 2011; Mayer and Haimberger, 2012; Guemas *et al.*, 2013; Allan *et al.*, 2014; Mayer *et al.*, 2014; Drijfhout *et al.*, 2014], considering the changes of multiannual means in the net downward energy fluxes at both TOA and surface is informative. A preliminary assessment of the multiannual mean changes (2001–2008 mean minus 1986–2000 mean) from reconstruction (F_T , F_{mass} , F_{res} , and $F_{mass-MERRA}$), UPSCALE, and AMIP5 data sets are presented in Figure 6. As discussed by Allan *et al.* [2014], all three data sets show decreased TOA net fluxes over the tropical East Pacific (Figure 6, left column). The magnitudes of the TOA flux changes over oceans are much smaller than those at the surface.

At the surface, the estimated changes over land areas are small from estimation (F_{mass} , F_{res} , and $F_{mass-MERRA}$), but the flux changes over Russia are slightly larger than in the UPSCALE and AMIP5 simulations. Figures 6b and 6d show the increasing downward energy flux over the North Pacific and Southern Ocean (increased ocean heat uptake), but negative flux changes over the central Pacific, North Indian Ocean, and North Atlantic. Although the individual surface flux components are not reconstructed, considering those simulated by ERAINT, the changes appear to be dominated by latent heat fluxes. Comparing with atmospheric model simulations, although both ensemble means from UPSCALE and AMIP5 simulations show decreased fluxes into the central Indian Ocean and North Atlantic (Figures 6i and 6l), the big differences are over the eastern Pacific, where simulated increases in downward flux are opposite to the estimations in Figures 6b, 6d, and 6f. The estimated surface flux from MERRA ($F_{mass-MERRA}$ in Figure 6f) is even noisier than those from F_{mass} and F_{res} , but it also displays decreased net downward energy flux over the eastern Pacific. This has been identified as an important region in determining aspects of the recent slowing rate of global surface temperature rise [Kosaka and Xie, 2013; Trenberth and Fasullo, 2013a; Meehl *et al.*, 2014]. On one hand, the cooling eastern Pacific will suppress turbulent energy transport from ocean to the atmosphere, so the net downward flux

the North and South Atlantic subtropical gyres (Figure 5c). As stated earlier, the $F_{mass-MERRA}$ (Figure 5d) has larger values over the central Indian Ocean and central western Pacific but smaller values over much of the eastern Pacific. UPSCALE shows the common feature of smaller flux over the North Indian Ocean and larger-energy flux over the Southern Ocean, but the strong flux over the western Pacific and smaller-energy flux over the eastern Pacific are not apparent in other data sets (Figure 5e). The ensemble means from AMIP5 simulations show much lower fluxes into the western Pacific (Figure 5f), and this is mainly contributed from Centro Euro-Mediterraneo sui Cambiamenti Climatici (CMCC), Centre National de Recherches Météorologiques (CNRM), Flexible Global Ocean-Atmosphere-Land System (FGOALS), Goddard Institute for Space Studies (GISS), Meteorological Research Institute (MRI), and Institute of Numerical Mathematics Climate Model version 4 (INM-CM4) model simulations as shown in Figure S3.

3.3. Changes in Downward Energy Flux

In order to investigate where the energy is moving through the climate system

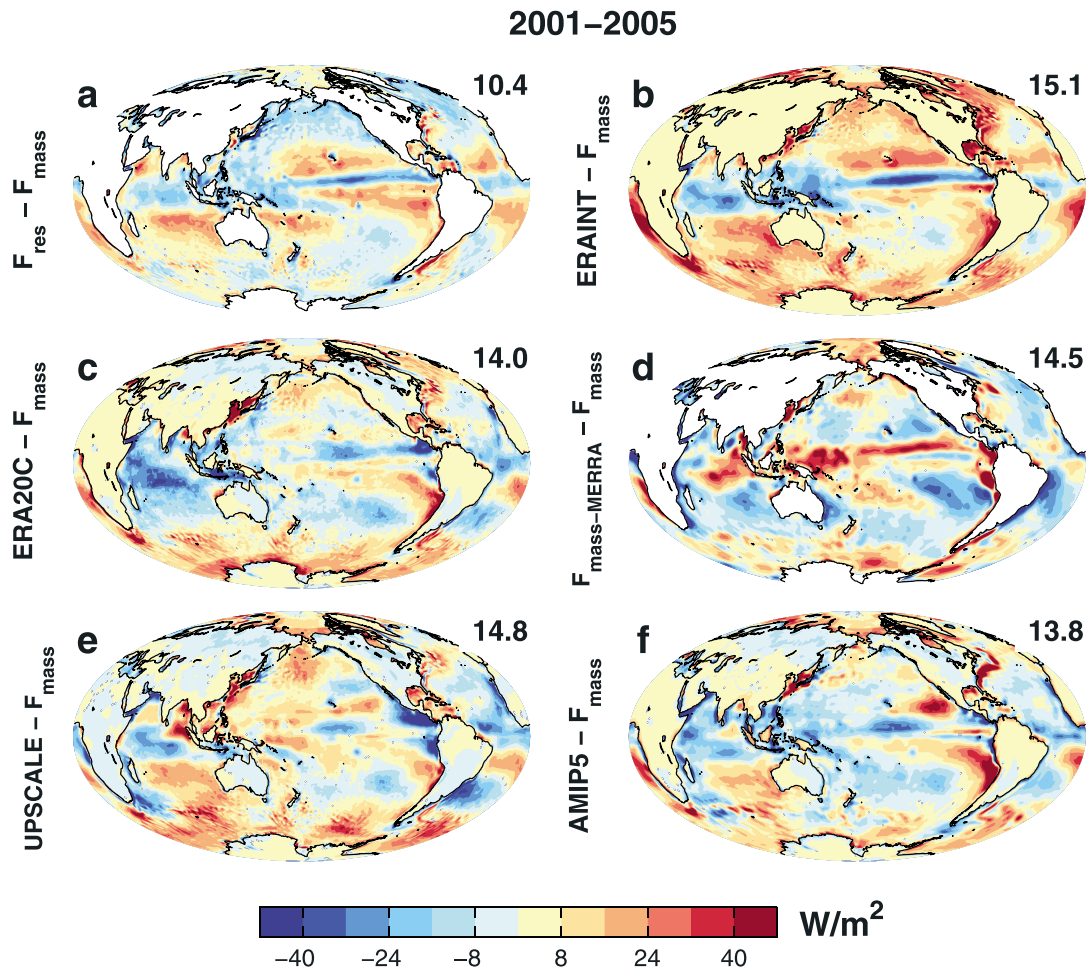


Figure 5. Multiannual (2001–2005) mean net downward surface energy flux (in W/m^2) differences between F_{mass} and (a) F_{res} , (b) ERAINT, (c) ERA20C, (d) $F_{\text{mass}} - \text{MERRA}$, (e) UPSACLE, and (f) AMIP5. The grid points of zero values are marked white, and the RMS differences are given at the top right corner.

would be increased over this region; on the other hand, as demonstrated by England *et al.* [2014], the cooling is due to the observed pronounced strengthening in Pacific trade winds which are not represented fully by AMIP simulations. The increased winds will cause more evaporation, so more latent heat transports to the atmosphere. Brown *et al.* [2014] also showed that the surface cooling over eastern Pacific will enhance the reflected short-wave radiation, therefore reduce the net downward energy flux.

The eastern tropical Pacific region marked in Figures 6b, 6d, and 6f covers 20°N – 20°S and 210°E to the west coast of the Central America. The mean TOA flux change (2001–2008 mean minus 1986–2000 mean) over this area (Figure 6a) is -2.1 W/m^2 , while the surface flux changes from F_{mass} (Figure 6b) and $F_{\text{mass}} - \text{MERRA}$ (Figure 6f) are -3.9 W/m^2 and -4.6 W/m^2 , respectively. Since the total energy tendency is almost zero over this area, the corresponding changes in vertical flux divergence (equal to net surface flux minus net TOA flux; Figure 6c) over this area are -1.8 W/m^2 and -2.5 W/m^2 , respectively. The negative signs indicate that vertical flux divergence decreased and consequently, divergence of horizontal energy transports increased in the 2001–2008 period compared to the 1986–2000 mean (compare equation (6)), so both changes in TOA fluxes and atmospheric energy transport contribute roughly equally to the reduced downward surface fluxes over the eastern tropical Pacific from these two mass-adjusted data sets. For F_{res} (Figure 6d) the mean change in surface flux over this area is about -0.5 W/m^2 and the corresponding mean change in vertical flux divergence (Figure 6e) is about 1.6 W/m^2 , which is opposite to the mean changes in vertical flux divergence of F_{mass} and $F_{\text{mass}} - \text{MERRA}$, implying that increased horizontal energy transport into the East Pacific region offsets much of the reduction in TOA downward fluxes leading to a smaller change in surface fluxes in this case. The net surface flux change obtained

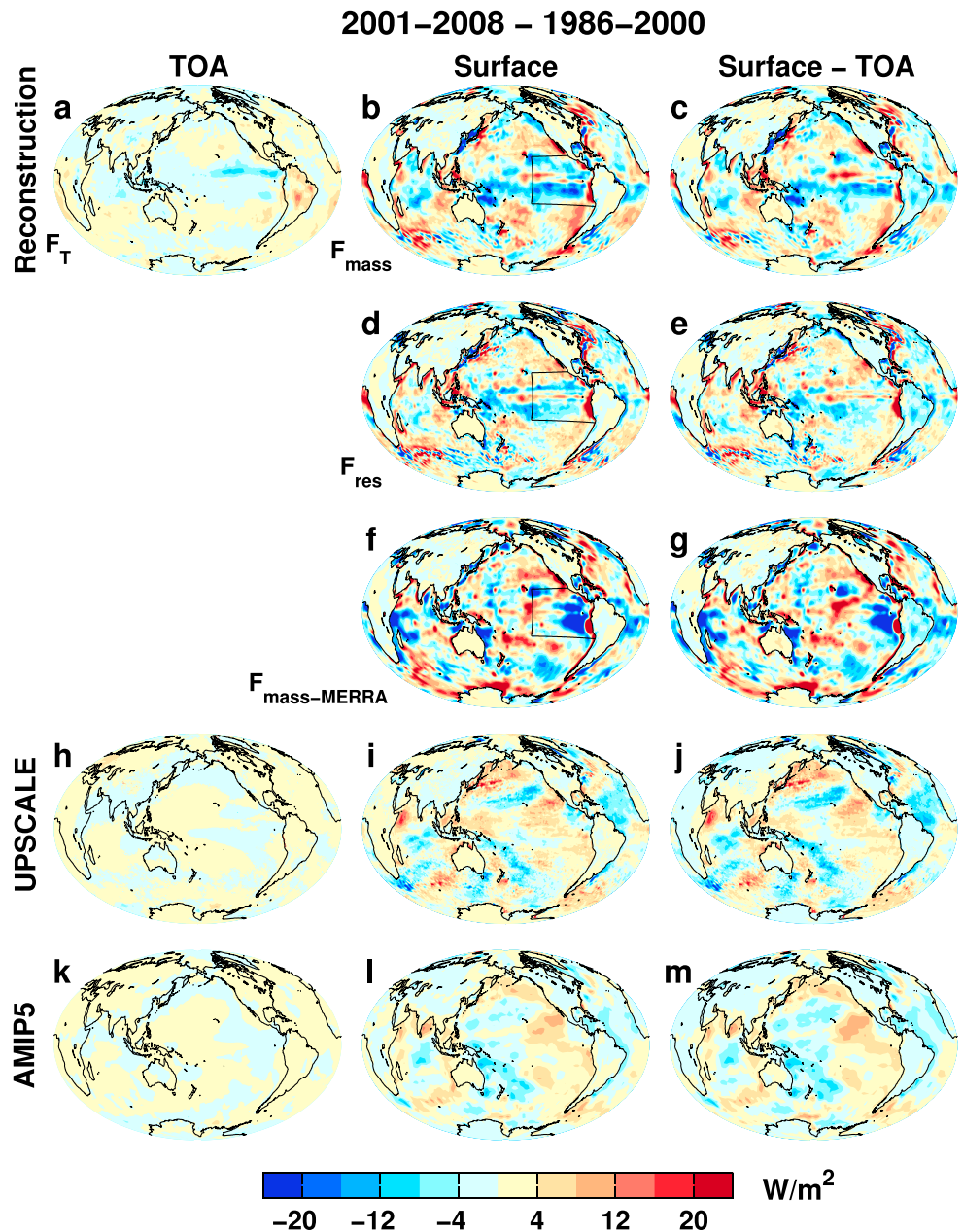


Figure 6. Change in net energy fluxes (W/m^2 ; 2001–2008 minus 1986–2000) at (left column) TOA, at (middle column) surface, and (right column) the difference between fluxes at surface and TOA from reconstructions (F_{mass} , F_{res} , and $F_{\text{mass}} - \text{MERRA}$), UPSCALE, and AMIP5 data sets. (a–c) The reconstructions based on Allan *et al.* [2014] at the TOA and the mass correction method using ERA-Interim data, (d and e) based on the residual method using ERA-Interim data, (f and g) the estimates from the mass correction method using MERRA reanalysis data, (h–j) the five-ensemble-member mean of the UPSCALE simulations, and (k–m) the 16-ensemble-member mean from the AMIP simulations. The marked area in Figures 6b, 6d, and 6f is from 20°N–20°S and 210°E to the west coast of Central America.

from E_{Dres} is weaker than those obtained from E_{Dmass} and $E_{\text{Dmass}} - \text{MERRA}$; since E_{Dmass} and $E_{\text{Dmass}} - \text{MERRA}$ are computed from analyzed state quantities, they are considered more realistic than E_{Dres} which is computed from model forecasts. Changes in TOA fluxes are about -0.5 W/m^2 for UPSACLE and AMIP5 data (Figures 6h and 6k). The changes at the surface (Figures 6i and 6l) are 2.2 W/m^2 and 3.3 W/m^2 , and the corresponding mean divergence changes of horizontal energy transport (Figures 6j and 6m) are 2.7 W/m^2 and 3.8 W/m^2 , respectively, implying that increased horizontal energy transport by the atmosphere into the region dominates the

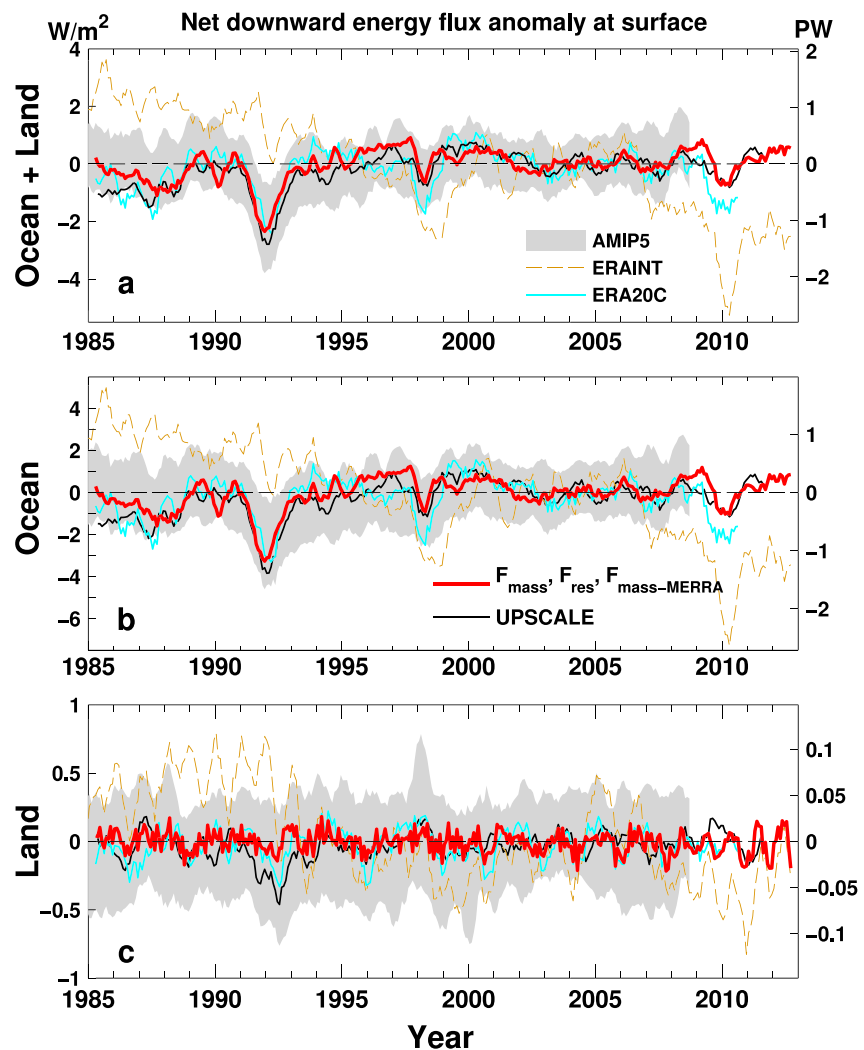


Figure 7. Deseasonalized anomaly (relative to the 2001–2005 period) time series of mean net downward energy fluxes at surface over (a) the globe, (b) the global ocean, and (c) the global land, from data sets of AMIP5, ERAINT, ERA20C, derived (F_{mass} , F_{res} , and $F_{\text{mass}} - \text{MERRA}$), and UPSCALE. The light grey shadings denote the \pm standard deviations of the 16 AMIP5 simulations. All lines are 6 month running mean. The y axis unit is W/m^2 on the left and PW on the right.

simulated changes in the surface fluxes. The divergence difference over the eastern tropical Pacific between the mass-adjusted data and those from model simulations requires further study.

For the reconstructed surface fluxes (F_{mass} and F_{res}), the global changes from the 1990s to the 2000s (see Table S1 in the supporting information) are consistent with *Allan et al.* [2014], who considered the TOA net imbalance; there is an increase in net downward flux at the surface due to the recovery from Pinatubo [*Smith et al.*, 2015]. Consistency with global mean TOA fluxes is expected since the surface flux estimates are based upon these TOA reconstructions and atmospheric heat capacity is small so cannot uptake a significant fraction of the top of atmosphere imbalance [*Palmer and McNeill*, 2014]. The ocean heat uptake is also increasing since over 90% of the excess energy into the Earth system is stored in the ocean [*Trenberth and Fasullo*, 2013a]. Consistency between global mean surface and TOA flux changes also applies to ERA20C reanalysis, UPSCALE, and AMIP5 simulations (see Table S1). *Smith et al.* [2015] highlighted the decline of TOA net downward radiation flux from 1999 to 2005, which potentially contributed to the recent warming slowdown. Consistent with *Smith et al.* [2015], similar calculations of two 5 year means centered at 1999 and 2005 from net downward surface energy fluxes show declines of 0.31 Wm^{-2} (reconstruction), 0.51 Wm^{-2} (UPSCALE), 0.07 Wm^{-2} (AMIP5), and 0.26 Wm^{-2} (ERA20C). The differences between flux changes

at TOA and surface (Figures 6h–6k) include the total energy tendency and divergence. The patterns are very similar to those surface changes and imply that the atmospheric energy divergence is the dominant factor affecting the surface flux changes, since both changes of TOA flux and atmospheric energy tendency are relatively small.

The changes of northward total meridional energy transports calculated from $E_{D_{\text{mass}}}$, $E_{D_{\text{res}}}$, and $E_{D_{\text{mass}} - \text{MERRA}}$ are also plotted in Figure 4b. Energy transports from mass-corrected divergences show the increase of northward transport in the northern hemisphere, but the energy transport from $E_{D_{\text{res}}}$ shows a decrease. It is mixed in the southern hemisphere where transport derived from $E_{D_{\text{res}}}$ displays a small-energy transport while both calculations from $E_{D_{\text{mass}}}$ and $E_{D_{\text{mass}} - \text{MERRA}}$ indicate an increase of poleward energy transport between 10–55°S and 15–70°S. The effect of the temporal discontinuities on these changes [Mayer *et al.*, 2013] in the reanalysis, due to artifacts of the observing system, merits further investigation, although the effect is most significant for the partition of the latent and dry static energy and less prominent when considering the total transport [Trenberth and Fasullo, 2013b].

The deseasonalized anomaly (calculated relative to the 2001–2005 period) time series of the area-weighted mean net downward energy fluxes at the surface from different data sets are plotted in Figure 7 for the globe, the global ocean, and the global land. The time series from all derivations (F_{mass} , F_{res} , and $F_{\text{mass}} - \text{MERRA}$) are identical by design. The light grey shadings are ± 1 standard deviation of the 16 AMIP5 simulations. All lines are 6 month running means. The ERAINT data are also plotted for reference purpose; spurious trends are explained by latent heat flux changes over the ocean [Chiodo and Haimberger, 2010] and from long-wave radiation over the land. There is good agreement between derived fluxes and those from AMIP5, ERA20C, and UPSCALE data sets over the globe. The correlation coefficients between derived and AMIP5, ERA20C, and UPSCALE are 0.38, 0.52, and 0.47, which are significant based on the two-tailed test using Pearson's critical values at the level of 5%. Over the global ocean, the coefficients are 0.33, 0.52, and 0.45, which are also statistically significant. Over land the correlation coefficient between derived and ERA20C is 0.60. The correlation coefficients between other data sets can be found in Table S2, and the correlation coefficient maps are in Figure S4. Future work will consider in more detail the variability across individual ocean basins and comparisons with independent data sets [Drijfhout *et al.*, 2014; Mayer *et al.*, 2014; Desbruyères *et al.*, 2014; Roemmich *et al.*, 2015] contributing toward understanding of variation in energy flows into the ocean.

4. Summary

Surface fluxes are a crucial component of the climate system, yet global-scale observational estimates are highly uncertain [Wild *et al.*, 2015]. To complement the existing set of surface flux data sets, an alternative method is developed. The net downward energy fluxes at the Earth's surface are estimated through the combination of the reconstructed TOA radiation fluxes [Allan *et al.*, 2014] and the atmospheric energy divergences (Figure 1) which are calculated using two distinct methods: (1) mass-adjusted energy divergence computed from ERA-Interim reanalysis [Trenberth *et al.*, 2001; Mayer and Haimberger, 2012; Berrisford *et al.*, 2011] and (2) the residual from the difference between the energy fluxes at the TOA and the surface from ERA-Interim.

To correct for unrealistic variability in energy fluxes over the land an adjustment was applied using a simple mean relation between surface flux and surface temperature change in UPSCALE climate model simulations which are strongly dependent upon the model's land surface component, JULES. By setting the global energy divergence to zero, applying the corrected surface fluxes over land and adjusting atmospheric energy divergence from the ocean to the land accordingly the net surface energy flux over ocean could be derived. Although this method relies upon the gross relationship between surface temperature change rate and energy fluxes from a simulation and other assumptions it was found that the sensitivity of the ocean surface flux changes to the methods applied over land are relatively small compared to the differences among data sets.

The accuracy of the resultant surface fluxes relies heavily on the quality of the reanalysis. The current version of ERA-Interim has some known problems including drifts in energy fluxes and deficient radiative forcing changes relating to anthropogenic and natural aerosols and problems with mass divergence and

conservation [Berrisford *et al.*, 2011]. All these will affect the quality of our product. The assimilation of various observed fields into the model draws toward an observed atmospheric state, so the aerosol effect on the mass-adjusted energy divergence ($E_{D_{\text{mass}}}$) should be less than the effect on $E_{D_{\text{res}}}$, but the accuracy of the divergence relies on other factors too, such as model spin-up and large time sampling errors, as discussed by Chiodo and Haimberger [2010].

Different data sets capture the general global patterns of the multiannual mean net downward surface fluxes despite the contrasting methods involved. The spatial correlation coefficients of multiannual means (2001–2005) between the reconstruction and other data sets are all around 0.9. The area mean surface flux anomaly time series shows reasonable agreement with AMIP5 ($r=0.33$), ERA20C ($r=0.52$), and UPSCALE ($r=0.45$) simulated monthly anomalies over the global ocean.

The change between the 1990s and 2000s over the eastern Pacific differs between data sets: while climate model-simulated surface fluxes increase over the period [Katsman and van Oldenborgh, 2011], the reconstruction indicates a reduced net downward surface flux. The cooling surface suppresses the air-sea turbulent energy exchange, but the strengthening of the observed trade winds [England *et al.*, 2014] over this area will reduce the net downward energy flux. Feedback involving low-altitude cloud and reflected short-wave radiation may also amplify this response [Brown *et al.*, 2014]. Since the estimated surface fluxes are strongly dependent upon the ERA-Interim as well as the MERRA reanalysis which both have temporal homogeneity issues [Mayer *et al.*, 2013], further verification of these products with other data sets from observations, reanalysis, and model simulations is required in order to further understand the strengths and weaknesses of the current methodology.

Assessing the degree to which SST patterns are driving or being driven by surface flux changes in this region merits investigation [Mayer *et al.*, 2014; Drijfhout *et al.*, 2014; Desbruyères *et al.*, 2014]. More detailed assessments of recent changes in surface energy fluxes entering distinct ocean basins [Mayer *et al.*, 2014; Desbruyères *et al.*, 2014] will contribute toward improved understanding of energy flows and internal variability in the climate system.

Acknowledgments

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