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# Components of near-surface energy balance derived from satellite soundings – Part 1: Net available energy

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## Abstract

This paper introduces a relatively simple method for recovering global fields of nearsurface net available energy (the sum of the sensible and latent heat flux or the difference between the net radiation and surface heat accumulation) using satellite visi-

- <sup>5</sup> ble and infra-red products derived from the AIRS (Atmospheric Infrared Sounder) and MODIS (MOderate Resolution Imaging Spectroradiometer) platforms. The method focuses on first specifying net surface radiation by considering its various shortwave and longwave components. This was then used in a surface energy balance equation in conjunction with satellite day-night surface temperature difference to derive 12 h dis-
- <sup>10</sup> crete time estimates of surface, system heat capacity and heat accumulation, leading directly to retrieval for surface net available energy. Both net radiation and net available energy estimates were evaluated against ground truth data taken from 30 terrestrial tower sites affiliated to the FLUXNET network covering 7 different biome classes. This revealed a relatively good agreement between the satellite and tower data, with
- <sup>15</sup> a pooled root mean square deviation of 98 and 72 W m<sup>-2</sup> for net radiation and net available energy, respectively, although both quantities were underestimated by approximately 25 and 10%, respectively relative to the tower observations. Analysis of the individual shortwave and longwave components of the net radiation revealed the downwelling shortwave radiation to be the main source of this systematic underestima-<sup>20</sup> tion.

### 1 Introduction

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An important manifestation of climate change is widespread alteration of the composition of the energy balance at the Earth's surface (Trenberth et al., 2009; Wild et al., 2013). Given the importance of being able to predict the consequences of climate change, both measurement and modelling of the components of surface energy balance attract significant attention from a broad range of related scientific disciplines



(Stephens et al., 2012). Two such disciplines are hydrology and meteorology, which share a common interest in resolving the balance between sensible, *H*, and latent,  $\lambda E$ , heat fluxes over a broad range of spatial and temporal scales (Anderson et al., 2012). Net available energy,  $\Phi$ , is a core variable used to predict the magnitude of *H* and  $\lambda E$  given it is defined as the sum of these two fluxes (Wright et al., 1992; Migletta et al., 2009; Anderson et al., 2012),

 $\lambda \Phi = E + H$ 

(1)

(2)

The utility of this definition arises from being able to also specify  $\Phi$  as the difference between the net broadband radiation,  $R_N$ , and the rate of heat accumulation, G, below the plain across which  $R_N$  is specified,

 $\Phi = R_{\rm N} - G$ 

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Given  $R_N$  is routinely measured using net radiometers this affords an opportunity to specify  $\Phi$  and hence either H or  $\lambda E$ . For example, in modelling studies  $\lambda E$  is invariably specified as a function of  $\Phi$  using the ubiquitous equations such as those of Penman (1948) for open water or Monteith (1965) for land surfaces (Mu et al., 2011; Mallick et al., 2014a). Despite being the rate of change of heat stock in terrestrial environments, G is often interpreted as the "ground heat flux" and attempts to measure this using heat flux plates are commonplace (Mayocchi and Bristow, 1995; Sauer and Horton, 2005; Heitman et al., 2010). These measurements prove somewhat less reliable than  $R_N$  due

- to greater spatial heterogeneity in ground heat uptake (Gao et al., 1998; Tittebrand and Berger, 2009; Verhoef et al., 2012) allied to the fact that significant heat capacity resides in other elements of the land surface (Ochsner et al., 2007). As a result, *G* proves problematic in surface energy balance studies and is either ignored (Foken et al., 2006; Foken, 2008) or treated somewhat superficially (Choudhury, 1987), despite being sig-
- <sup>25</sup> nificant under a broad range of conditions (Santanello and Friedl, 2003; Ochsner et al., 2007). Large scale estimates of *G* are useful in the context of regional and global evapotranspiration modelling and for verification of regional and Global Circulation Models (Kergoat et al., 2011).



The arrival of satellite retrievals for many of the components of  $R_N$  has opened up opportunities to develop large scale estimates of this variable and hence  $\lambda E$  (Batra et al., 2006; Mu et al., 2007; Anderson et al., 2012). For example, retrievals for the components of  $R_N$  have been available through the International Satellite Cloud Climatology Project (ISCCP) (Pinker and Laszlo, 1992; Stephens et al., 2012); the Earth Radiation Budget Experiment (ERBE) (Priestley et al., 2011); Clouds and Earth's Radiant Energy System (CERES) (Mlynczak et al., 2011; Chen et al., 2013) on board of the NASA Earth Observing System (EOS) and Tropical Rainfall Measuring Mission (TRMM) satellites (Wielicki et al., 1998). Several studies have reported the estimation of  $R_N$  using a combination of MODIS (MOderate Resolution Imaging Spectroradiometer) atmospheric and land products over USA, China, and India (Cai et al., 2007; Mallick et al., 2009; Bisht and Bras, 2010, 2011) or NOAA-14 (National Oceanic and

Atmospheric Administration) data over the Tibetan Plateau (Ma et al., 2002).

Unfortunately, in the absence of direct observations of *G* at spatial scales and coverage of satellite  $R_N$ , retrievals for  $\Phi$  have had to rely on the parameterisation of *G* using surface temperature, albedo and vegetation index information (Bastiaanssen et al., 1998; Batra et al., 2006) or by assigning some fixed proportion of  $R_N$  (Choudhury, 1987; Humes et al., 1994) in satellite-based surface energy balance models (Mecikalski et al., 1999; Anderson et al., 2012). But studies by Murray and Verhoef (2007), Hsieh

- et al. (2009) and recently Verhoef et al. (2012) also demonstrated that *G* is, by definition, a highly dynamic quantity, and that the ratio  $G/R_N$  can range anywhere from 0.05 to 0.50 depending on the time of day, soil moisture and thermal properties, and vegetation density. Therefore, methods that are able to provide defensible estimates of *G* in conjunction with  $R_N$  would clearly be of great benefit to this area for determin-
- <sup>25</sup> ing  $\Phi$  directly from satellite data and without relying unduly on any offline calibration. In this paper we present a method for retrieving  $R_N$  and  $\Phi$  based on exploring both satellite radiance data and day–night surface temperature difference. The approach is necessarily simple in order to avoid over reliance on models in the pre-processing and to reflect the fact that the focus of this work is the production of satellite estimates of  $\Phi$



for use in a simple Bowen ratio  $\lambda E$  specification framework as detailed in a companion paper by Mallick et al. (2014b). Taking advantage of the extensive network of terrestrial eddy covariance tower sites (Baldocchi et al., 2001) which record direct measurements of  $R_N$ , H and  $\lambda E$ , we use these measurements to derive independent non-radiative estimates of  $\Phi$  in order to critically evaluate our satellite estimates of this quantity.

## 2 Methodology

# 2.1 Net radiation

The approach for estimating  $R_N$  uses the Atmospheric Infrared Sounder (AIRS) radiation products, although we have also made use of the MODIS surface reflectance and

solar zenith angle products where necessary.  $R_N$  is generated by considering the following balance between net shortwave ( $R_{NS}$ ) and longwave ( $R_{NL}$ ) radiation at or near the Earth's surface,

$$R_{\rm N} = R_{\rm NS} + R_{\rm NL} = (1 - \alpha)R_{\rm S\downarrow} + R_{\rm L\downarrow} - R_{\rm L\uparrow}$$

where  $\alpha$  is the surface albedo and  $R_{L\downarrow}$  and  $R_{L\uparrow}$  are the downwelling and upwelling thermal radiative fluxes and  $R_{S\downarrow}$  is the downwelling shortwave radiative flux (all fluxes specified in W m<sup>-2</sup>). Our chosen reference level for  $R_N$  is the near surface given this corresponds to the flux-based tower estimates we used in the evaluation. Therefore, surface  $R_{S\downarrow}$  was estimated from its top-of-atmosphere clear sky counterpart  $R_{S0\downarrow}$  and AIRS cloud cover fraction (*f*) following Hildebrandt et al. (2007),

<sup>20</sup> 
$$R_{S\downarrow} = (1 - f)\tau_A R_{S0\downarrow}$$

where  $\tau_A$  is the clear sky transmissivity of the atmosphere which we assume is 0.75 (Cano et al., 1986; Thornton and Running, 1999; Hildebrandt et al., 2007; Gubler et al., 2012). Although clearly a simplification, a constant clear sky transmissivity is widely



(3)

(4)

used (e.g. Massaquoi, 1988; Bindi et al., 1992; Choudhury, 2001; Hildebrandt et al., 2007; Mallick et al., 2009) in recognition of the absence of robust alternatives. In addition, exploiting the AIRS cloud cover fraction data in Eq. (4) should help accommodate the effects of variations in both the aerosol optical depth (Kaufman and Koran, 2006; Quass et al., 2010) and atmospheric water vapor (Adhikari et al., 2006).

The terrestrial surface albedo was generated using the MODIS Aqua–Terra surface reflectances  $r_i$  following Liang et al. (1999),

$$\alpha = \sum_{i=1}^{N} p_i r_i + 0.0036$$

where  $r_i$  are the mid-point reflectances within the 0.62–0.67; 0.841–0.876; 0.459– 0.479; 1.230–1.250; 1.628–1.653; 1.628–1.653; and 2.105–2.155µm wavelength bands and  $p_i$  are the weightings for each wavelength bands taken as  $p_i = [0.3973;$ 0.2382; 0.3489; -0.2655; 0.1604; -0.0138; 0.0682] (Liang et al., 1999, 2002). The albedo of the ocean varies according to the cosine of solar zenith angle (Jin et al., 2004). In the present case, a constant albedo of 0.04 was assumed for oceans given the satellite radiances are nadir.

Many of the longwave components of the radiative balance are very closely related to the raw IR radiances being measured by AIRS. Given these are not in the public domain we have attempted to recover them as follows, although in future we would anticipate using the raw IR radiances more directly if possible.  $R_{\rm NL}$  was calculated as,

<sup>20</sup> 
$$R_{\rm NL} = R_{\rm L\downarrow} - R_{\rm L\uparrow} = \varepsilon_{\rm C} \varepsilon_{\rm S} \sigma T_{\rm C}^4 - \varepsilon_{\rm S} \sigma T_{\rm S}^4$$

where  $\sigma$  is the Stefan–Boltzmann constant (5.67 × 10<sup>-8</sup> W m<sup>-2</sup> K<sup>-4</sup>),  $T_{\rm C}$  is the columnar air temperature, and  $\varepsilon_{\rm C}$  and  $\varepsilon_{\rm S}$  are the column and surface emissivities. Among the different schemes for calculating  $\varepsilon_{\rm C}$  we have used the formulation proposed by Prata (1996) given this appears to be the most reliable (Niemela et al., 2001; Bisht and Bras,

(5)

(6)

2010, 2011). This scheme uses AIRS total precipitable water ( $\xi$ ) information to estimate  $\varepsilon_{\rm C}$  as,

 $\varepsilon_{\rm C} = 1 - (1 + \xi) e^{-(1.2 + 3\xi)^{0.5}}$ 

The columnar air temperature  $T_{\rm C}$  in Eq. (6) is taken as the average of the 2 m and 1000 hPa pressure level AIRS temperatures in an attempt to reflect a weighting toward the lower troposphere when specifying  $R_{\rm L\downarrow}$ .  $T_{\rm S}$  and  $\varepsilon_{\rm S}$  are taken directly from the AIRS skin temperature and surface emissivity products.

#### 2.2 Surface heat capacity, ground heat flux and net available energy

The definition of G stems from consideration of the non-steady state surface energy balance,

$$c\frac{\mathrm{d}T_{\mathrm{S}}(t)}{\mathrm{d}t} = R_{\mathrm{N}}(t) - \lambda E(h) - H(t) = G(t)$$

where *c* is the aggregate surface system heat capacity. The AIRS sounder platform samples twice daily at 01:30 and 13:30 hours local time (LT). Despite being somewhat coarse, taking a discrete time, backward difference approximation of Eq. (8) with a sample interval of  $\Delta t = 12$  h equivalent to that of the AIRS pass gives,

 $\Delta T_{\rm S}(t) = b_1 R_{\rm N}(t) + b_2$ 

20

(9)

(10a) (10b)

(8)

(7)

where  $\Delta T_{\rm S}$  is the day-night surface temperature change,  $b_1 = \Delta t/c$  and  $b_2 = -\Phi(t)\Delta t/c$ . If we assume that the system is approximately in equilibrium over a 24 h cycle, and that at 01:30 LT  $\Phi \approx 0$  (for all 30 sites analysed in this study  $\Phi(01:30) < 0.05\Phi(13:30)$ ; see also Tamai et al., 1998; Mamadou et al., 2014), then this gives the following simultaneous equations:

$\Delta T_{\rm S}(13:30) = b_1 R_{\rm N}(13:30) + b_2$	
$-\Delta T_{\rm S}(13:30) = b_1 R_{\rm N}(01:30)$	



which can be solved analytically to derive  $b_1$  and  $b_2$  and hence  $\Phi$  and c for each grid cell in the AIRS global array.

Equation (10) is a coarse approximation of Eq. (8) and hence potentially suffers from a number of deficiencies. Firstly, diurnal symmetry in  $\Delta T_S$  is only appropriate when one considers weekly or monthly average behaviour, and that there are no additional heat losses or gains to/from stores beyond the domain defined by the single heat capacity *c*. In this study we examined the monthly average behaviour because AIRS only gives partial global coverage on the daily timescale due to both cloud effects and the non-overlapping swath width of the sensor. Interaction with additional long term heat stores is an issue in systems such as the oceans where there can be a persistent heat

- <sup>10</sup> stores is an issue in systems such as the oceans where there can be a persistent heat loss/gains to/from deeper water over timescales of weeks to months, although relative to the diurnal fluctuation of stored surface heat this tends to be small (Stramma et al., 1986). Secondly,  $\Phi$  can be either positive or negative at 01:30 LT, although it tends to be only a fraction (never exceed 5% of afternoon  $\Phi$ ) of its 13:30 LT value due to the
- <sup>15</sup> supply of relatively small energy at night compared to the day. This may be less true for areas of land in the height of winter with cloud full days and over the sea where significant daytime heat accumulation could in part be re-released as night time latent and sensible heat. Thirdly, Eqs. (9) and (10) attribute the magnitude of the daytime *G* to the nighttime net long wave radiative balance, which is obviously rather uncertain. Finally,
- <sup>20</sup> all the terms in Eq. (8) are highly dynamic and yet are treated as constant or varying linearly over the 12 h sample interval. It is difficult to predict what the consequences of this are, given it depends on the pattern of radiative forcing through the day which can vary significantly in both time and space.

### 2.3 Satellite datasets

<sup>25</sup> In the present study two different data sources were used for the estimation of  $R_N$ and  $\Phi$ , AIRS and MODIS. The AIRS sounder is carried by the NASA Aqua satellite, which was launched into a sun-synchronous low Earth orbit on 4 May 2002 as part of the NASA Earth Observing System. It gives near global coverage twice daily at



1.30 a.m.-p.m. from an altitude of 705 km. Level 3 standard monthly day-night data products of air temperature and relative humidity profiles, cloud cover fraction, surface emissivity, near-surface air temperature and surface-skin temperature and columnar total precipitable water at 1° × 1° spatial resolution were obtained for 2003 from
the online data archive of AIRS, distributed through NASA Mirador data holdings (http://mirador.gsfc.nasa.gov/). The monthly products are simply the arithmetic mean, weighted by counts, of the daily data of each grid box. The multi-day merged products have been used here because the IR retrievals are not cloud proof and the multi-day product gave decent spatial cover in light of the missing cloudy sky data. The data products were obtained in hierarchical data format (HDF4) along with their latitude-longitude projection.

We have used the MODIS Aqua atmospheric product datasets (MYD08\_D3) (http: //modis-atmos.gsfc.nasa.gov/index.html) at  $1^{\circ} \times 1^{\circ}$  spatial resolution for extracting the solar zenith angle field. For generating the surface albedo fields we used narrowband

<sup>15</sup> surface reflectances from combined MODIS Terra–Aqua 16 day data (MCD43C4) products acquired from the MODIS data archive (http://ladsweb.nascom.nasa.gov/data/ search.html). The native spatial resolution of the MODIS datasets is 0.05°. Therefore, all the narrowband surface reflectances were first resized into 1° by 1° to make them compatible with the AIRS spatial resolution and then the broadband surface albedo was generated from these data.

### 2.4 Sensitivity analysis

A general sensitivity analysis was carried out in order to assess the effects of the propagation of uncertainty onto the estimates of G,  $R_N$  and  $\Phi$ . For this analysis the input terms were assigned uniform prior distributions of  $\pm 10\%$  for all parameters other than

temperatures for which ±1 K uniform prior distributions were assumed. These assumed ranges resemble the stated uncertainties as given in the AIRS support literature (Aumann et al., 2003; Hearty et al., 2014). The sensitivity of each output to each input was calculated assuming an average, locally linear sensitivity. These were expressed as



the change in output per unit change in input, normalised by the median value of each. Only absolute sensitivities > 0.1 were considered significant. The standard deviation of the estimated distributions of *G*,  $R_N$  and  $\Phi$  were used as the summary statistic for the measurement uncertainty of the proposed methodology.

## 5 2.5 Evaluation of $R_N$ and $\Phi$

To evaluate the satellite values of  $R_N$  and  $\Phi$  we have made use of the extensive FLUXNET terrestrial tower network (Baldocchi et al., 2001). Clearly, there is a scale conflict here with the satellite retrievals being 1° whilst the individual tower observations are for scales of the order of 1 km or less. The tower  $R_N$  are from the broadband net radiometer sensors located on each tower. In the absence of reliable measures of G at the tower scale and in order to derive genuinely independent measures of  $\Phi$  against which to evaluate the satellite data, we have taken the tower net available energy as the sum of the measured sensible and latent heat flux i.e. Eq. (1). Thereby we have assumed that the eddy covariance flux measurements are able to close the energy balance (i.e.  $R_N - G = \lambda E + H$ ), the implications of which will be discussed below. We have chosen 30 sites covering a broad range of geographical locations selected from 7 land cover types including; evergreen broadleaf forest (DBF), savanna (SAV), grass-

land (GRA) and cropland (CRO). A comprehensive list of the site characteristics are provided in Table 1. Each tower evaluation dataset is comprised of the 13:30 LT samples of  $R_N$ , H and  $\lambda E$  which correspond with the satellite overpass. Again, the evaluation is based on pooling these data into weighted monthly average values. For the evaluation we have elected to compare all 12 months of data for 2003 given this had the best overlap between the FLUXNET and AIRS databases.



## 3 Results

Table 2 shows the results from the sensitivity analysis. For *G* we see the importance of the long wave specification and in particular  $\varepsilon_{\rm C}$ . The standard deviation of the estimate of *G* from the ensemble is  $18 \,{\rm W} \,{\rm m}^{-2}$  giving approximate 95% confidence detection limits of  $\pm 36 \,{\rm W} \,{\rm m}^{-2}$  on the estimates.  $R_{\rm N}$  is sensitive to all components of the radiation balance calculation as expected (Table 2). The standard deviation of the estimate of  $R_{\rm N}$ 

- from the ensemble is 40 W m<sup>-2</sup> giving approximate 95 % detection limits of  $\pm$ 80 W m<sup>-2</sup> on the estimates (Table 2). Not surprisingly the sensitivity results for  $\Phi$  mirror those of  $R_{\rm N}$  albeit with a marginally higher ensemble standard deviation of 44 W m<sup>-2</sup> (Table 2).
- <sup>10</sup> The locations of the 30 terrestrial evaluation sites are marked in Fig. 1. Figure 2 shows annual average, global satellite scenes for 13:30 LT  $R_N$ , c, G and  $\Phi$  for 2003. Missing data in the images are mainly due to missing data in the AIRS soundings at high latitudes or over the mountain belts where it is difficult to profile air temperature and relative humidity reliably. In addition, persistent cloudy conditions also prevent reli-<sup>15</sup> able retrieval and hence are rejected although these will be less evident in the monthly or annual average data.

Figure 2a shows the global distribution of  $R_{\rm N}$  which generally decreases with latitude as expected.  $R_{\rm N}$  also decreases over land due to the generally higher albedo resulting in reduced absorption of the net shortwave radiation (Giambelluca et al., 1997; Gao and Wu, 2014) or relatively higher surface temperature increasing the net longwave component, especially over the drier regions (Liang et al., 1998; Trenberth, 2011). As a result the magnitude of  $R_{\rm N}$  was around 200–300 W m<sup>-2</sup> over the dry desert regions whereas the oceanic values of  $R_{\rm N}$  were 450–700 W m<sup>-2</sup>.

Figure 2b shows the global distribution of *c* (surface heat capacity). The oceanic values of 4 to 8 MJ m<sup>-2</sup> K<sup>-1</sup> are equivalent to 1 to 2 m of sea water, which appears reasonable on the daily time step to which they relate (Stramma et al., 1986; Schwartz, 2007). These oceanic values are somewhat noisy due to the small day–night temperature differences observed for the oceans giving a relatively poor signal to noise ratio.



However, behind this noise the pattern of oceanic c appears relatively uniform as one might expect. Over land c varies between 0.05–0.5 MJ m<sup>-2</sup> K<sup>-1</sup> with wetter tropical and high latitude areas showing significantly higher values than the drier, less vegetated areas as expected. The soil equivalent depth of this heat capacity is approximately 5 0.01 m, which again appears reasonable for a daily time step (Li and Islam, 1999), although in heavily vegetated areas c is obviously comprised of a more complex ag-

gregation. Figure 2c shows the global distribution of G. These are the 13:30 LT values, hence being net positive as an annual average. Between 20° N-S G is approximately 10 to

- 20 % of  $R_{\rm N}$ , and this rises to more than 40 % above 50° N–S (Hsieh et al., 2009). Given 10 this opposes the pattern of  $R_{\rm N}$  one would conclude either some deficiencies in the way G is specified here or that  $R_{\rm N}$  partitions into latent heat far more effectively than surface heating in these warm wet environments (Liu et al., 2005). Again, terrestrial values are lower than their oceanic equivalents mainly due to the lower heat capacity as
- well as reduced  $R_{\rm N}$  as discussed above. This also highlights the role of the vegetation 15 layer in preventing ground heating (Baker and Baker, 2002; Bounoua et al., 2010). The Sahara appears particularly prominent in this scene with high rates of midday heat accumulation which appears to be associated with a combination of moderate net radiation and relatively high heat capacity. The heterogeneity in this region appears to

be related to the pattern of bare darker rock. 20

Figure 2d shows the global distribution of  $\Phi$  which follows a similar pattern to  $R_{\rm N}$ as expected, although the pattern of G shown in Fig. 2c dictates that the N-S gradients in  $\Phi$  are somewhat stronger than those of  $R_{\rm N}$ . Before discussing these results we consider their evaluation. Figure 3a shows the pooled evaluation of  $R_{\rm N}$  which pro-

duced an overall correlation of  $r = 0.88 (\pm 0.03)^{1}$ . Assuming both tower and satellite 25 observations are linearly related through some "true" value, then the pooled values are co-related by  $R_{\rm N}$ (satellite) = 0.75 (±0.02)  $R_{\rm N}$ (tower) + 23.37 (±8.20) i.e. a small but significant underestimation in  $R_N$  (satellite) relative to  $R_N$  (tower). The root mean square

<sup>&</sup>lt;sup>1</sup>All uncertainties are expressed as  $\pm$  one standard deviation unless otherwise stated.



Introduction

References

**Figures** 

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Interactive Discussion

deviation (RMSD) between the two was  $98 \text{ W m}^{-2}$ . The biome specific statistics for  $R_N$  are given in Table 3 which reveals correlations ranging between 0.65 (EBF) to 0.96 (ENF), RMSD ranging between 74 (GRA) to 127 W m<sup>-2</sup> (EBF) and regression statistics ranging between 0.58 (±0.08) to 0.87 (±0.04) for the gain and -32.40 (±23.73) to 107.45 (±39.93) for the offset.

Figure 3b shows the evaluation for  $\Phi$  which produced pooled statistics of r = 0.87 (±0.03),  $\Phi$ (satellite) = 0.90 (±0.03)  $\Phi$ (tower) – 2.43 (±8.19) and RMSD of 72 W m<sup>-2</sup>. The biome specific statistics for  $\Phi$  are also given in Table 3 showing correlations ranging from 0.70 (EBF) to 0.95 (ENF), RMSD ranging between 62 (GRA & SAV) to 88 (EBF) W m<sup>-2</sup> and regression coefficients ranging between 0.66 (±0.08) to 1.01 (±0.05) and –65.25 (±27.07) to 108.71 (±32.10) for the gain and offset, respectively.

Figure 4 shows a sample of monthly time series for  $\Phi$  for both the satellite and the towers. The sites were selected to represent the biome classes considered here and also ones for which complete annual data sets for 2003 were available. These results

show the satellite estimates generally track the trends in the tower data and hence the pooled statistics are not masking the within site variability. Again, the site-wise comparative statistics for these data are given in Table 2.

#### 4 Discussion

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For R<sub>N</sub> the statistics relating the satellite and tower data are comparable with the re<sup>20</sup> sults of: Bisht et al. (2005) who obtained 74 W m<sup>-2</sup> RMSD when evaluating MODIS Terra geophysical land products over the Southern Great Plains of the US; Jacobs et al. (2004) who obtained a 12.2 % error when evaluating GOES (Geostationary Operational Environmental Satellite) geostationary data over Southern Florida; Cai et al. (2007) who obtained 13.7 % error when evaluating MODIS Terra–Aqua data over China; Bisht and Bras (2010) who obtained 39–51 W m<sup>-2</sup> RMSD over the central US using MODIS terra atmospheric data at 5–10 km spatial resolution; Hwang et al. (2013) and Hou et al. (2014) who reported RMSD in instantaneous and daily R<sub>N</sub>



to be 58–142 W m<sup>-2</sup> and 37–40 W m<sup>-2</sup> over South east Asia and China, respectively, using MODIS terra data products. Stackhouse et al. (2000) evaluated the International Satellite Cloud Climatology Project (ISCCP) data to have errors in the range 10 to  $15 \text{ W m}^{-2}$  in monthly average shortwave and longwave radiative fluxes. When these errors are compounded in the derivation of  $R_{\text{N}}$  and compared with tower data, an RMSD of the order of 98 W m<sup>-2</sup> appears reasonable considering the spatial resolution of the AIRS data (1° × 1°).

There have been very few attempts to retrieve satellite estimates of  $\Phi$  and compare these with ground truth data, although the statistics from our attempt appear to parallel results of Stisen et al. (2008) who studied a single site in the Senegal River basin using 5 km spatial resolution MSG (Meteosat Second Generation) geostationary satellite data and obtained a correlation of r = 0.51 and an RMSD of 14 to 17% of the mean in comparison to the surface measurements.

As seen in Fig. 3a there is a systematic underestimation of  $R_{\rm N}$  relative to the tower values which exceeds the typical accuracy of net radiometer measurements of 20 W m<sup>-2</sup> quoted by Foken (2008). We examined this underestimation in more detail wherever possible, by evaluating three of the individual radiation components of  $R_{\rm N}$  ( $R_{\rm S\downarrow}$ ,  $R_{\rm L\downarrow}$  and  $R_{\rm L\uparrow}$ ). All tower sites provided measurements of  $R_{\rm S\downarrow}$  (but not  $R_{\rm S\uparrow}$ ). Figure 3c shows  $R_{\rm S\downarrow}$  is systematically underestimated at the satellite scale with  $R_{\rm S\downarrow}$ (satellite) = 0.70 (±0.02)  $R_{\rm S\downarrow}$ (tower) + 68 (±12.24) which accounts for the mismatch of  $R_{\rm N}$ (satellite)  $\approx 0.75R_{\rm N}$ (tower). Before attempting to account for the various reasons for this underestimation it is important to realise that, unlike the IR components, the shortwave components are all-sky retrievals i.e. like the tower data they do not omit cloudy sky conditions. As a result, any bias in the shortwave is not as a result of biased sampling when comprising the monthly average. Besides, the omission of non-clear sky data would tend to lead to  $R_{\rm SI}$ (satellite) >  $R_{\rm SI}$ (tower).

If we assume  $\tau_A$  to be the principal reason for  $R_{S\downarrow}$  (satellite) <  $R_{S\downarrow}$  (tower) then a global value of 0.75 would be, on average, too low. Given the relatively well defined relationship between  $R_{S\downarrow}$  (satellite) and  $R_{S\downarrow}$  (tower) seen in Fig. 3c one would imagine



that a more sophisticated dynamic representation of  $\tau_A$  would offer only marginal improvements. Recalibration of  $\tau_A$  using the tower data is a possibility although we have avoided this given the AIRS cloud cover fraction and scale mismatch between the satellite and tower could also be implicated in the observed bias. For example, the diffuse

fraction of R<sub>Sl</sub>(tower) can become enriched by surface reflected solar radiation, particularly in undulating terrain (Dubayah and Loechel, 1997; Sultan et al., 2014). Nonlinear scaling effects of surface albedo (Oliphant et al., 2003; Salomon et al., 2006) can also be implicated in this because surface albedo interacts nonlinearly with surface characteristics such as surface wetness and land surface temperature (Ryu et al., 2008) or
 the leaf area index (Hammerle et al., 2008).

To probe the specification of  $R_N$  further we investigated the individual longwave radiation components in relation to measures of these fluxes available for a limited subset (14) of tower sites. From Fig. 3d and e it appears that there is quite good agreement between the satellite and tower data for both  $R_{L\downarrow}$  and  $R_{L\uparrow}$  and that any mismatch is <sup>15</sup> insufficient to explain the discrepancy in  $R_N$ . This is somewhat surprising for two reasons. Firstly, unlike the shortwave component,  $R_L$ (tower) is all sky whilst  $R_L$ (satellite) is only from clear sky conditions where IR retrieval is possible. As a result, one would anticipate very significant differences in the monthly average values of the longwave components. However, it is difficult to predict the effect of this biased sampling on

<sup>20</sup>  $R_{\rm NL}$ (satellite) given that cloud interacts with both  $R_{\rm L\downarrow}$  and  $R_{\rm L\uparrow}$  in complex ways. Secondly, one would anticipate significant scaling effects from the  $T^4$  nonlinearity in Eq. (6) which can result in a disproportionate contribution of warmer elements within the system to both  $R_{\rm L\downarrow}$  and  $R_{\rm L\uparrow}$  (Kustas and Norman, 2000; Lakshmi and Zehrfuhs, 2002; Corbari et al., 2010). The fact that these affects are not seen to any significant de-<sup>25</sup> gree could point to compensating errors in the analysis but does not distract from the central message of the importance of the bias in the shortwave when accounting for  $R_{\rm N}$ (satellite) <  $R_{\rm N}$ (tower).

Figure 3b and Table 3 show that  $\Phi$ (satellite)  $\approx 0.90\Phi$ (tower) suggesting a slight compensation for the underspecification of  $R_{S1}$  through the underspecification of *G* from the



satellite data. However, this evaluation assumes the energy balance to be closed in the tower data (i.e.  $R_N - G = \lambda E + H$ ), which typically is not the case,  $\lambda E + H$  often falling short of  $R_N - G$  by 20% (Wilson et al., 2002). Because the causes of this energy imbalance remain controversial (see Foken, 2008 for review), it is difficult to estimate how much the tower values of  $\lambda E + H$  are actually biased low and hence the extent to which this bias affects our evaluation. The errors in the tower data are believed to be associated with different footprint characteristics for the different instruments used (Lin et al., 2008). For example,  $R_N$  observations typically have a footprint size of ~ 10 m<sup>2</sup> whilst

- air properties (e.g. air temperature, humidity) have footprint sizes of > 1 km<sup>2</sup>. By way of illustration, if, in the worst case, the entire energy imbalance was to be attributed exclusively to  $\lambda E + H$  (i.e.  $R_N - G$  are quantified correctly), then the true midday  $\lambda E + H$ could be some 20 % greater (Wilson et al., 2002). As a result, the present bias seen in Table 3 would change to  $\Phi$ (satellite)  $\approx 0.72\Phi$ (tower) again implicating  $R_{S\downarrow}$  as the main source of bias in the satellite retrievals for both  $R_N$  and  $\Phi$ .
- <sup>15</sup> The method we present here for estimating  $R_N$  contrasts with more sophisticated model-based approaches which attempt to accommodate the complexity of atmospheric radiative transfer explicitly (e.g. Fouquart and Bonnel, 1980; Mlawer et al., 1997; Bisht and Bras, 2010; Hou et al., 2014). There are several reasons for adopting this stance. Firstly, the estimates of  $R_N$  need to be sympathetic with the simple <sup>20</sup> dynamic energy balance used to accommodate *G* when specifying  $\Phi$ . Secondly, the
- estimates of  $\Phi$  are intended for use in a simple Bowen ratio framework for specifying satellite latent heat (Mallick et al., 2014b) and we believe it to be important that the complexity of the methods used here are commensurate with those used in the simple Bowen ratio approach. Related to this, we have tried to restrict the approach to
- <sup>25</sup> largely using only AIRS data which provides the satellite soundings required for the Bowen ratio estimates. This single platform approach is to ensure the estimates do not suffer unduly from blending different data sources. Finally, complex radiative transfer approaches are also prone to the effects of significant uncertainties (Betts et al., 1993; Morcrette, 2002; Seidel et al., 2010).



### 5 Conclusions

We have only evaluated the satellite retrievals using data from terrestrial sites, and clearly it would be worthwhile repeating this for the ocean retrievals if possible. We have held back on this evaluation here because of the lack of an extensive network of instan-

- taneous latent and sensible heat flux or radiative flux data over the oceans, although we note that the SEAFLUX project within the Global Energy and Water Experiment (GEWEX) initiative should give rise to such a database in the near future. From the terrestrial evaluation we would argue that the methodology employed here shows some promise for specifying both *R*<sub>N</sub> and Φ, although the results suggest the need for improvements particularly in the specification of *R*<sub>S↓</sub>. More detailed studies evaluating the representativeness of each tower site footprint in relation to the 1° scale within which
- it is situated could prove useful in this regard as would methods for cloud-proofing the satellite retrievals under persistent cloudy sky conditions. Similarly, evaluation under extreme conditions (e.g. high altitudes and latitudes) is also required.
- Having derived and provisionally evaluated Φ for terrestrial systems this is available to schemes for estimating latent and sensible heat using satellite data. Given we have resorted to the minimal amount of calibration in deriving Φ it would appear sensible if a similar philosophy were adopted in developing satellite-based schemes for these important fluxes. With the availability of high spatial and temporal resolution geostation-
- ary imager-sounder data from GOES and, in the near future, GIFTS (Geosynchronous Interferometric Fourier Transform Spectrometer) and INSAT (Indian National Satellite)-3D our present approach could be extended to derive Φ at the half hour time scale. In addition to opportunities in specifying large scale surface heat and water vapour fluxes, the heat capacity estimates made here clearly carry information on variations in
- terrestrial properties such as surface moisture storage and we envisage that studies to develop this concept further could prove fruitful, particularly because of the emergence of satellite microwave data against which the results could be compared.



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Biome type	Site name, Country	Latitude	Longitude	Reference
Evergreen broadleaf forest (EBF)	Palagkaraya, Indonasia	-2.35	114.04	Hirano et al. (2007)
	Puechabon, France	43.74	3.6	Reichstein et al. (2003)
	Caxiuana Forest-Almeirim, Brazil	-1.72	-51.46	Carswell et al. (2002)
	Manaus – ZF2 K34, Brazil	-2.61	-60.21	de Araújo et al. (2004)
	Santarem-Km67, Brazil	-2.86	-54.96	Hutyra et al. (2007)
	Santarem-Km83, Brazil	-3.02	-54.58	Goulden et al. (2004)
Mixed forest (MF)	Vielsalm, Belgium	50.31	5.99	Aubinet et al. (2001)
	Tomakomai National forest, Japan	42.73	141.52	Hirano et al. (2003)
	Changbaishan, China	42.4	128.09	Zhang et al. (2006)
Grassland (GRA)	Oensingen1 grass, Switzerland	47.29	7.73	Ammann et al. (2007)
	Neustift/Stubai Valley, Austria	47.12	11.32	Hammerle et al. (2008)
	Goodwin Creek, USA	34.25	-89.87	unpublished data
	Bugacpuszta, Hungary	46.69	19.61	Gilmanov et al. (2007)
	Dripsey, Ireland	51.99	-8.75	Jaksic et al. (2006)
Cropland (CRO)	ARM Southern Great Plains, USA	36.61	-97.49	Fischer et al. (2007)
	Bondville, USA	40.01	88.29	Meyers et al. (2004)
	Tsukuba, Japan	36.05	140.03	Saito et al. (2005)
Evergreen needleleaf forest (ENF)	Le Bray, France	44.72	-0.77	Granier et al. (2000a)
	Duke Forest – loblolly pine, USA	35.98	-79.09	Katul et al. (2003)
	Blodgett forest, USA	38.89	-120.63	Goldstein et al. (2000)
	Howland forest, USA	45.2	-68.74	Hollinger et al. (1999)
Deciduous broadleaf forest (DBF)	Harvard Forest EMS Tower (HFR1), USA	42.54	-72.17	Urbanski et al. (2007)
	Univ. of Michigan Biological Station, USA	45.56	-84.71	Gough et al. (2009)
	Willow Creek, USA	45.81	-90.08	Cook et al. (2004)
	Hesse Forest – Sarrebourg, France	48.67	7.06	Granier et al. (2000b)
	Hainich, Germany	51.08	10.45	Anthoni et al. (2004)
	Morgan Monroe State forest, USA	39.32	-86.41	Baldocchi et al. (2001)
	Takayama, Japan	36.15	137.42	Saigusa et al. (2002)
Savanna (SAV)	Tonzi Ranch, USA	38.43	-120.97	Baldocchi et al. (2004)
	Skukuza, South Africa	-25.02	31.49	Scholes et al. (2001)

#### **Table 1.** Eddy covariance sites used for the evaluation of the satellite derived $R_N$ and $\Phi$ .

**BGD** 11, 11825-11861, 2014 **Components of** near-surface energy balance derived from satellite soundings K. Mallick et al. Title Page Abstract Introduction Conclusions References Tables Figures 14 Þ١ ► ◄ Close Back Full Screen / Esc Printer-friendly Version Interactive Discussion  $(\mathbf{\hat{H}})$ 

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**Table 2.** Sensitivity analysis results. The forcing data are taken for mid-summer on the Southern Great Plains, US. Sensitivities are locally linear, averaged across the ensemble response and expressed as dimensionless relative changes. Only absolute sensitivities > 0.1 are shown.  $N = 10^5$  realisations.





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<b>Table 3.</b> Comparative statistics for the satellite and tower derived $R_N$ and $\Phi$ for a range of
biomes. Values in parenthesis are $\pm$ one standard deviation.

				Φ						
Biome	RMSD (Wm <sup>-2</sup> )	Gain	Offset	r	N	RMSD (W m <sup>-2</sup> )	Gain	Offset	r	Ν
EBF	126.67	0.58 (±0.08)	107.45 (±39.93)	0.65 (±0.09)	69	87.67	0.66 (±0.08)	108.71 (±32.10)	0.70 (±0.09)	65
MF	104.21	0.82 (±0.07)	-32.40 (±23.73)	0.89 (±0.08)	36	87.29	0.97 (±0.10)	-65.25 (±27.07)	0.86 (±0.09)	32
GRA	74.29	0.73 (±0.05)	51.37 (±15.88)	0.88 (±0.06)	59	61.51	0.83 (±0.07)	15.71 (±16.21)	0.86 (±0.07)	53
CRO	89.13	0.73 (±0.08)	35.62 (±28.57)	0.84 (±0.09)	36	53.31	0.99 (±0.10)	-0.23 (±23.98)	0.87 (±0.09)	36
ENF	85.45	0.87 (±0.04)	-26.83 (±14.78)	0.96 (±0.04)	48	66.7	1.01 (±0.05)	-56.56 (±15.27)	0.95 (±0.05)	46
DBF	92.77	0.71 (±0.05)	21.74 (±15.15)	0.85 (±0.06)	84	71.57	0.88 (±0.06)	-16.70 (±14.23)	0.85 (±0.06)	80
SAV	103.98	0.69 (±0.08)	56.28 (±36.08)	0.87 (±0.11)	23	61.98	0.97 (±0.11)	-14.42 (±37.68)	0.88 (±0.25)	23
Pooled	98.21	0.75 (±0.02)	23.37 (±8.20)	0.88 (±0.03)	355	72.26	0.90 (±0.03)	-2.43 (±8.19)	0.87 (±0.03)	335

EBF = Evergreen broadleaf forest, MF = Mixed forest, GRA = Grassland, CRO = Cropland, ENF = Evergreen needleleaf forest, DBF = Deciduous broadleaf forest, SAV = Savanna.







**Figure 2.** Global fields for annual average 13:30 LT: (a) net radiation  $R_N$  (W m<sup>-2</sup>); (b) surface heat capacity, c (MJ m<sup>-2</sup> K<sup>-1</sup>); (c) surface heat accumulation rate, G (W m<sup>-2</sup>); (d) net available energy,  $\Phi$  (W m<sup>-2</sup>), for 2003.





**Figure 3.** Comparison of satellite and tower monthly average 13:30 LT (a)  $R_N$  and (b)  $\Phi$ . For details of the site characteristics see Table 1. For the comparative statistics see Table 2. The solid line is the pooled linear regression given in Table 2. Comparison of satellite and tower monthly average 13:30 LT (c)  $R_S$ , (d)  $R_{L\downarrow}$  and (e)  $R_{L\uparrow}$  for a selection of sites for which tower data for  $R_S$  (360 data points),  $R_{L\downarrow}$  (159 data points) and  $R_{L\uparrow}$  (159 data points) were available. The linear fit (solid line) between the two sources of  $R_S$  is,  $R_S$ (AIRS) = 0.70 (±0.02)  $R_S$ (tower) + 67.68 (±12.24); r = 0.84 (±0.03). The linear fit (solid line) between the two sources of  $R_S$  is,  $R_S$ (AIRS) = 1.03 (±0.03)  $R_{L\downarrow}$ (tower) – 36.91 (±10.05); r = 0.95 (±0.03). The linear fit (solid line) between the two sources of  $R_{L\downarrow}$  is,  $R_{L\downarrow}$ (AIRS) = 1.03 (±0.03)  $R_{L\downarrow}$ (tower) – 36.91 (±10.05); r = 0.95 (±0.03). The linear fit (solid line) between the two sources of  $R_{L\downarrow}$  is,  $R_{L\downarrow}$ (AIRS) = 0.91 (±0.02)  $R_{L\uparrow}$ (tower)+20.43 (±8.77); r = 0.96 (±0.02). The dashed lines are 1 : 1 in all cases.

 $(+ EBF; \times MF; \circ GRA; * CRO; \lor ENF; \diamondsuit DBF; \Box SAV)$ 







**Figure 4.** Satellite (grey) and tower (black) time series of monthly average 13:30 LT net available energy  $\Phi$  for a selection of sites for 2003.