

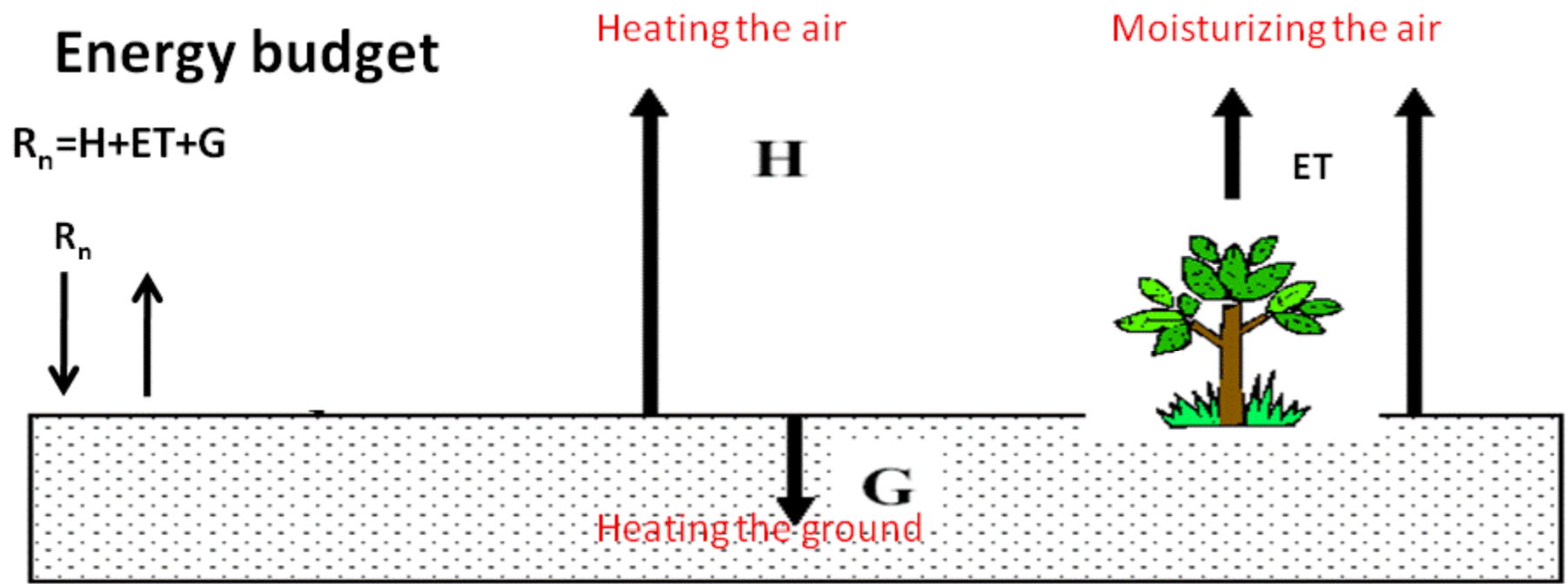


Yale University

Feb. 17, 2011

Satellite Estimation of Land Surface Radiation and Energy Budgets

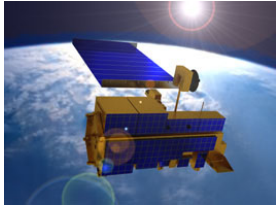
Dr. Shunlin Liang, Professor
Department of Geography
301-405-4556, sliang@umd.edu
<http://www.glue.umd.edu/~sliang>



$$R_n = R_n^s + R_n^l = (1 - \alpha)F_d^s + \epsilon F_d^l - \sigma \epsilon T^4$$

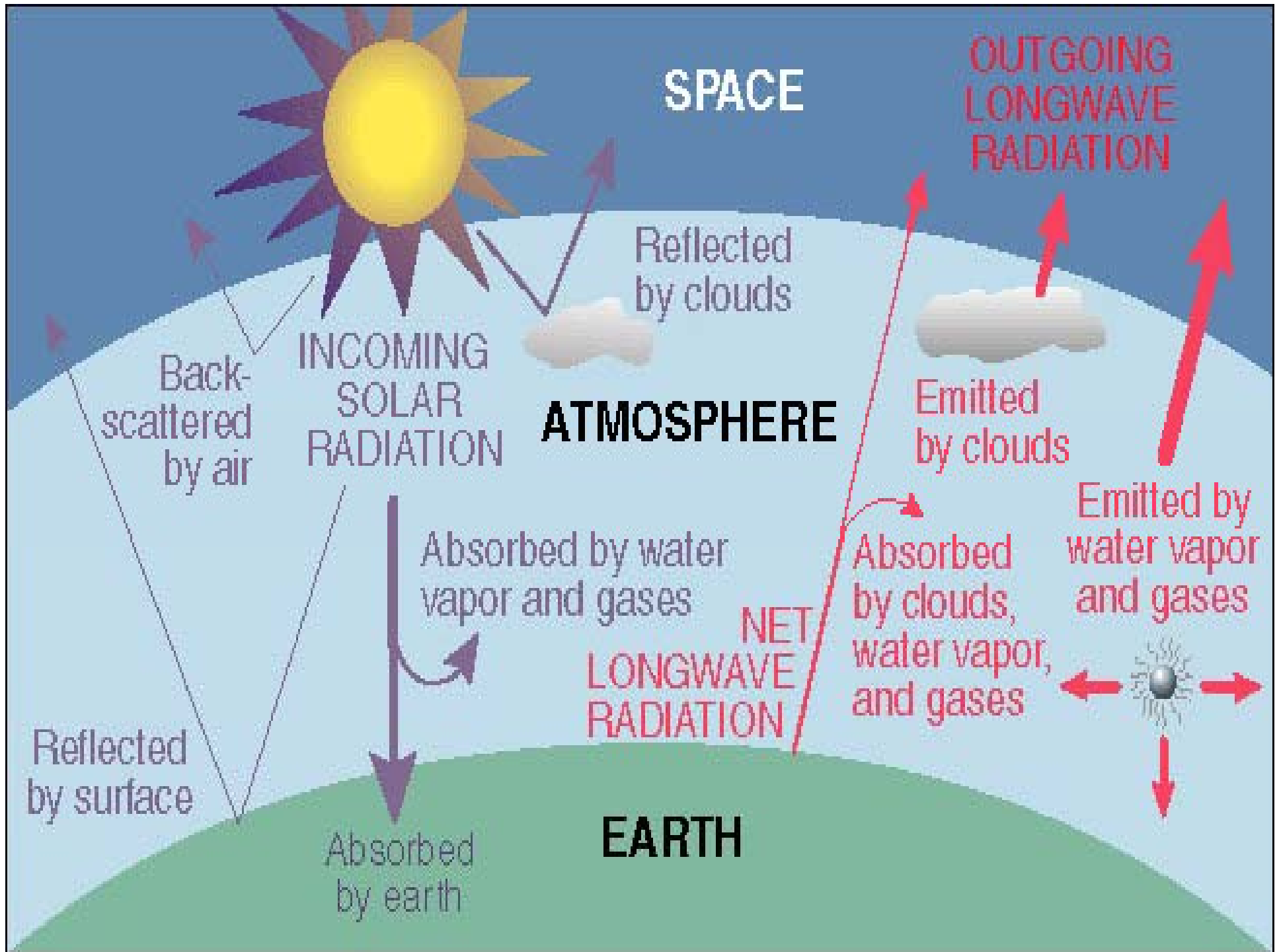
Net radiation
 albedo
 Insolation
 Longwave downward radiation
 Emissivity
 Skin temperature

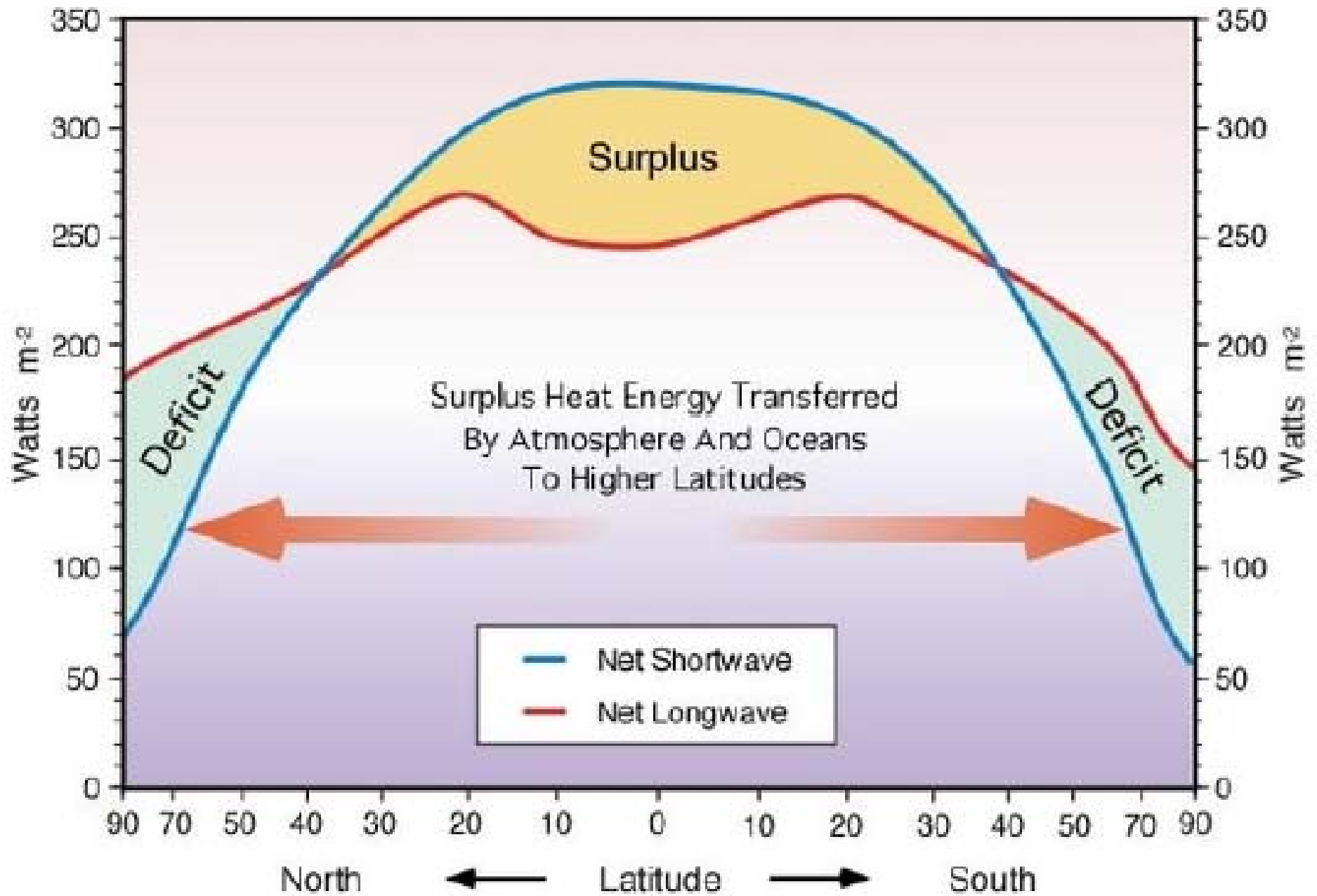
Radiation budget

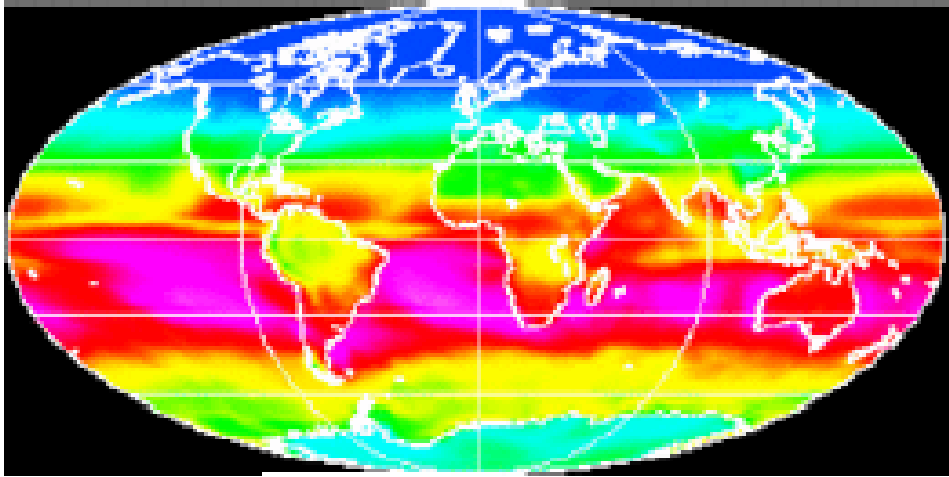


Outlines

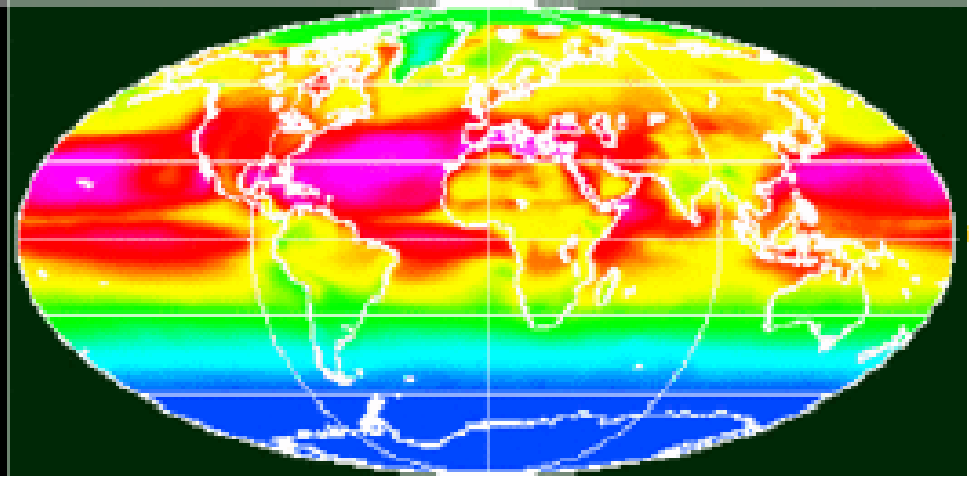
- ♣ Background
- ♣ Insolation/PAR
- ♣ Albedo
- ♣ Longwave net radiation
- ♣ Evapotranspiration
- ♣ Summary



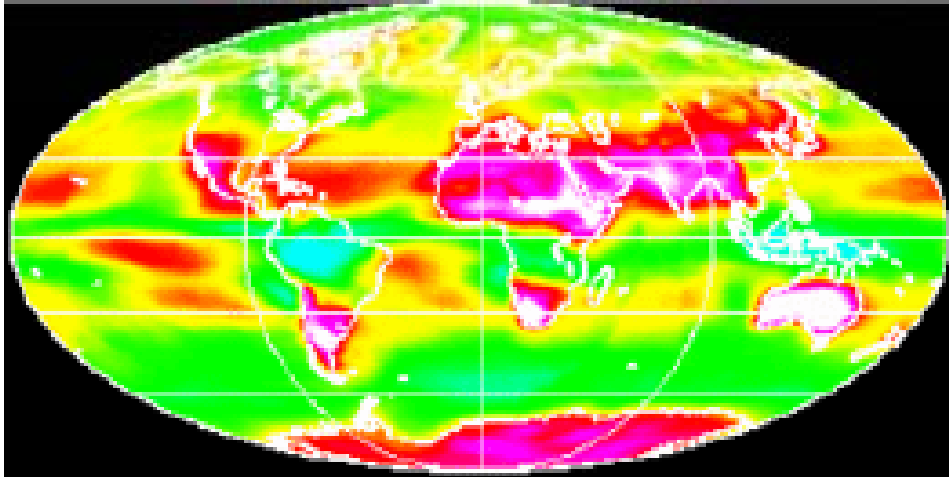




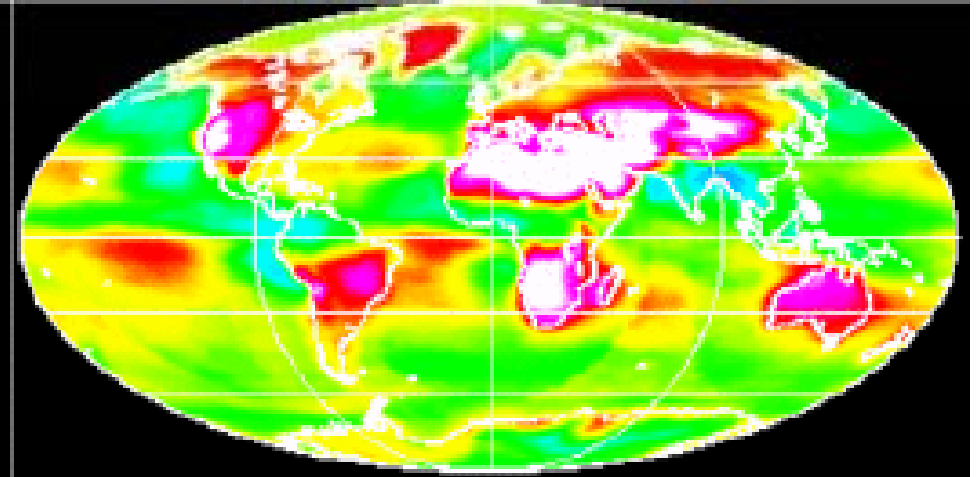
Average Net Shortwave Radiation at the Earth's Surface:
January 1984-1991 (K)



Average Net Shortwave Radiation at the Earth's Surface:
July 1984-1991 (K)



Average Net longwave Radiation at the Earth's Surface:
January 1984-1991 (K)



Average Net longwave Radiation at the Earth's Surface:
July 1983-1990 (K)



radiation and energy budgets in the climate models

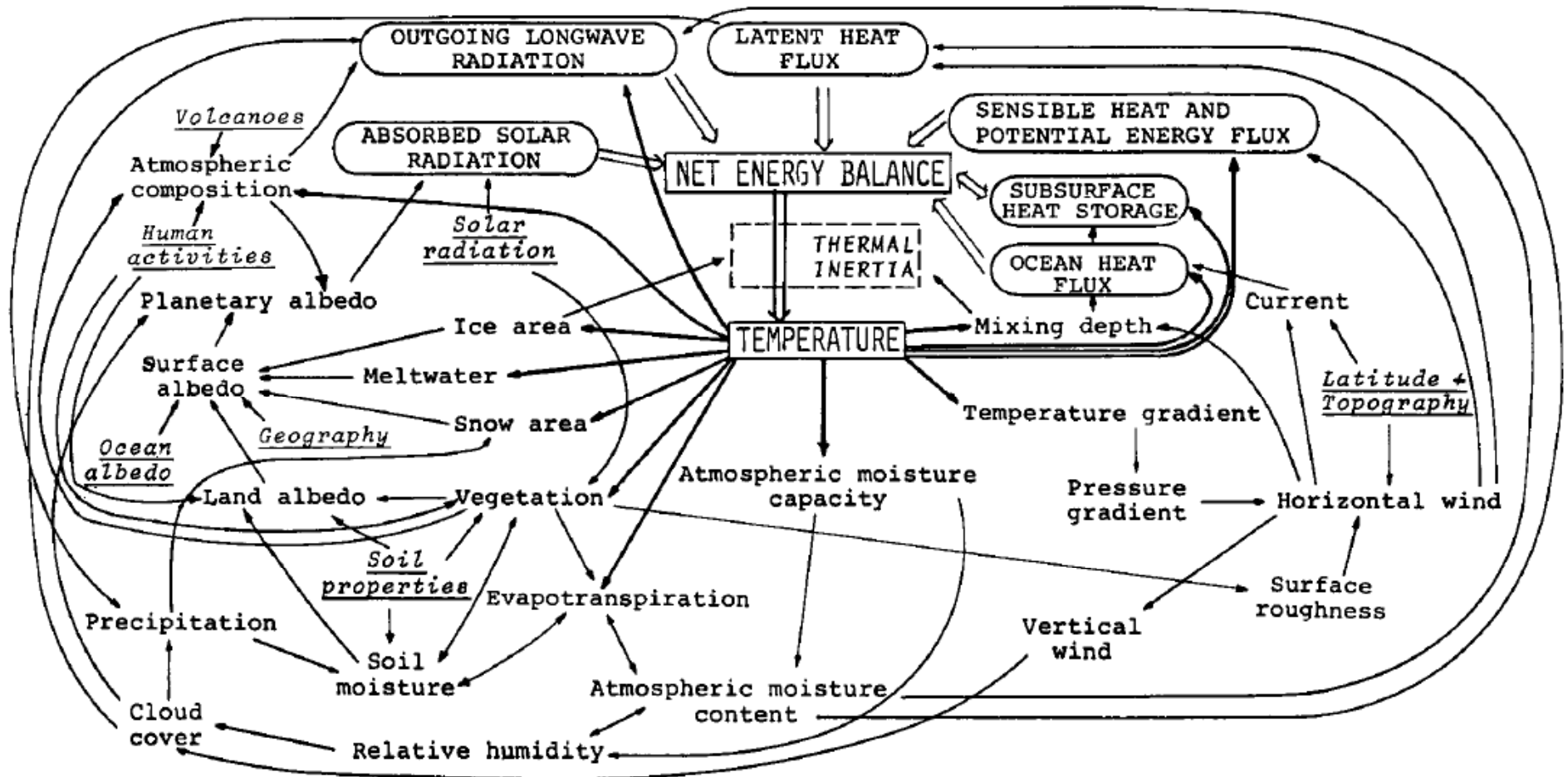
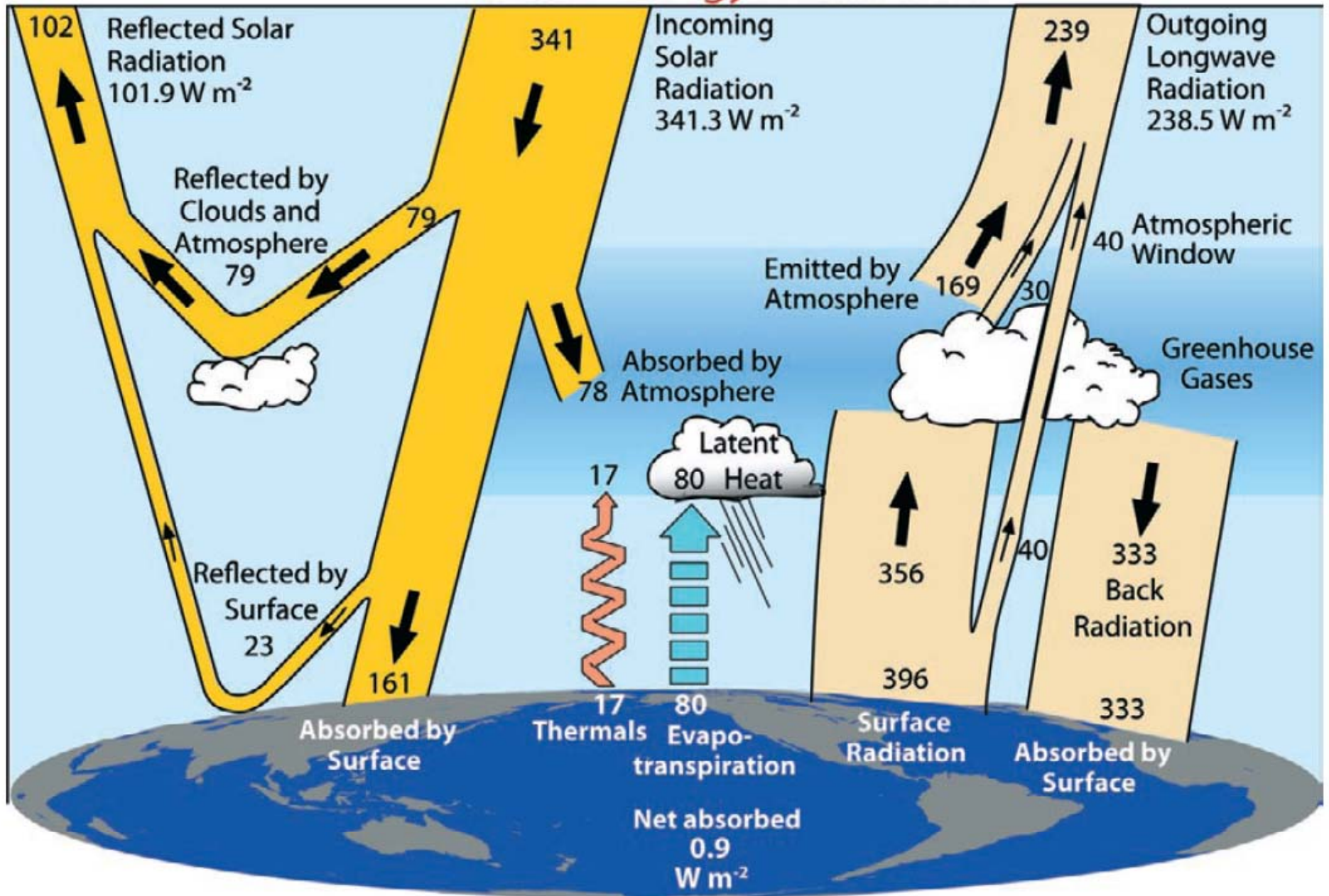


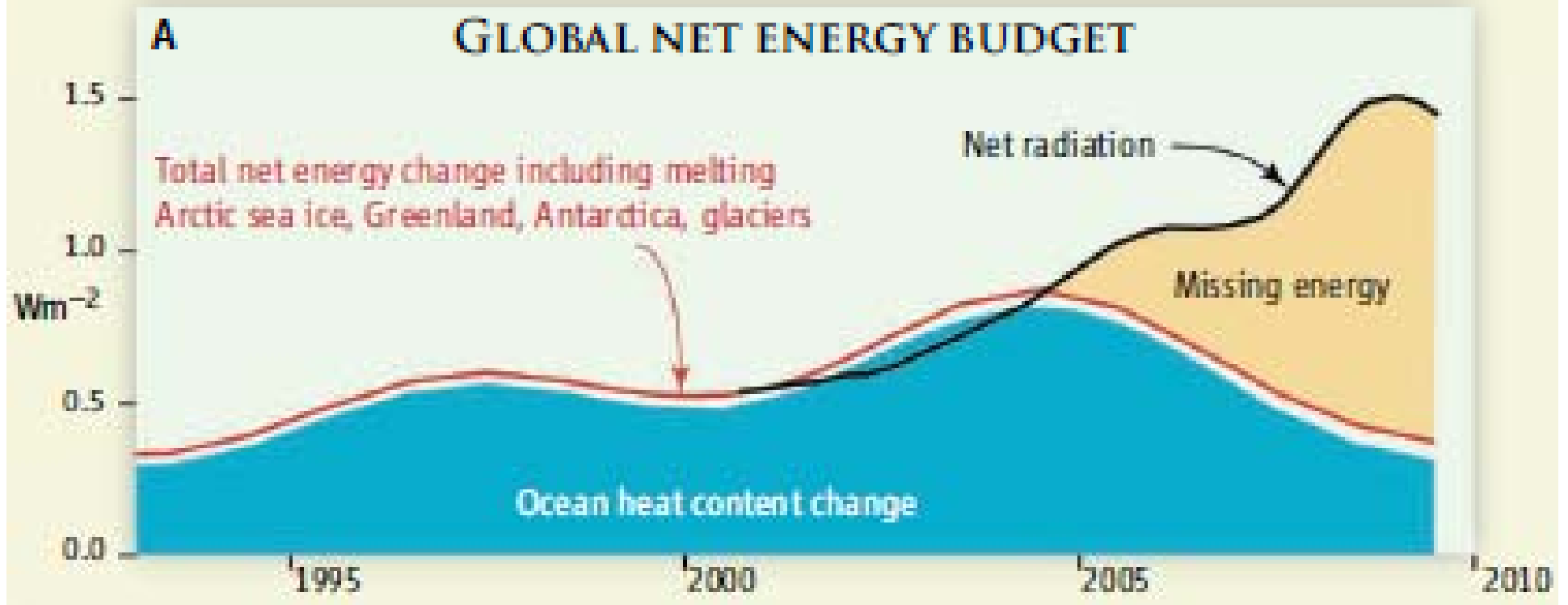
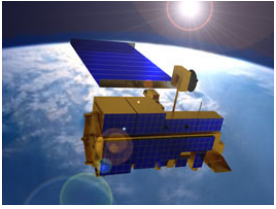
FIG. 1. Climate feedback diagram.

(Rohock, 1985)

Global Energy Flows $W m^{-2}$



Trenberth et al. (2009)



Trenberth, Science, 2010

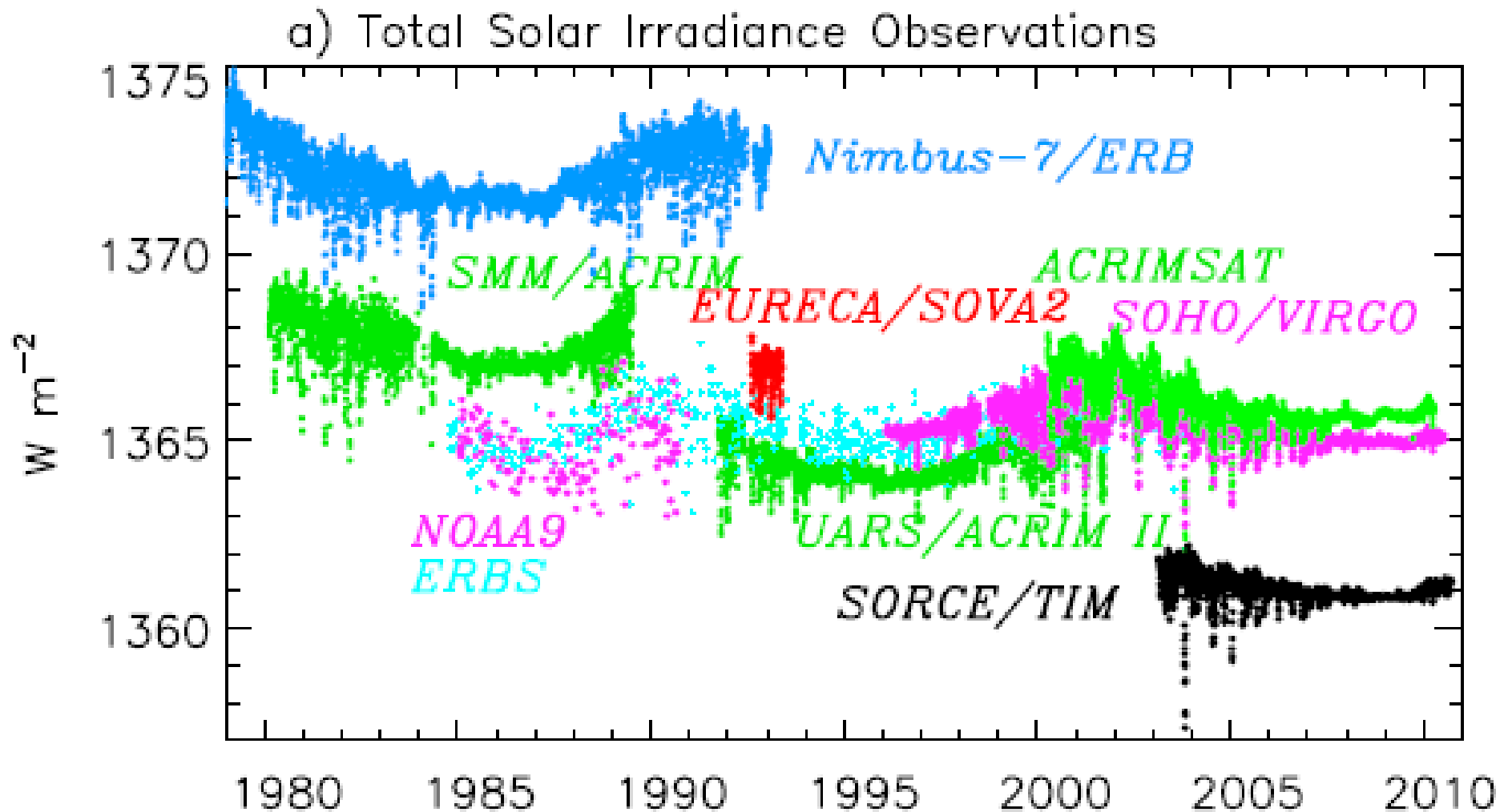


Figure Space-borne total solar irradiance (TSI) measurements are shown on “native” scales with offsets attributable to calibration errors. Instrument overlap allows corrections for offsets and the creation of a composite TSI record.

TABLE 1. Global mean clear- and all-sky SW, LW, and net TOA radiative fluxes, solar irradiance, and CRE for satellite-based data products (units in $W m^{-2}$).

Product name	CERES					ISCCP FD
	ERBE S-4	ES-4 Ed2_rev1	SRBAVG- nonGEO Ed2D_rev1	SRBAVG- GEO Ed2D_rev1	GEWEX SRB Version 2.86	
Time period	02/85 – 01/89				03/00 – 02/2006	
Solar irradiance	341.3	341.3	341.3	341.3	341.8	341.5
LW (All sky)	235.2	239.0	237.7	237.1	240.4	235.8
SW (All Sky)	101.2	98.3	96.6	97.7	101.7	105.2
Net (All Sky)	4.9	4.0	7.0	6.5	-0.3	0.5
LW (Clear Sky)	264.9	266.6	266.4	264.1	268.1	262.3
SW (Clear Sky)	53.6	49.3	51.2	51.1	54.5	54.2
Net (Clear Sky)	22.8	25.4	23.7	26.2	19.2	25.0
LW CRE	29.7	27.6	28.7	27.0	27.7	26.5
SW CRE	-47.6	-49.0	-45.4	-46.6	-47.2	-51.0
NET CRE	-17.9	-21.4	-16.7	-19.7	-19.5	-24.5

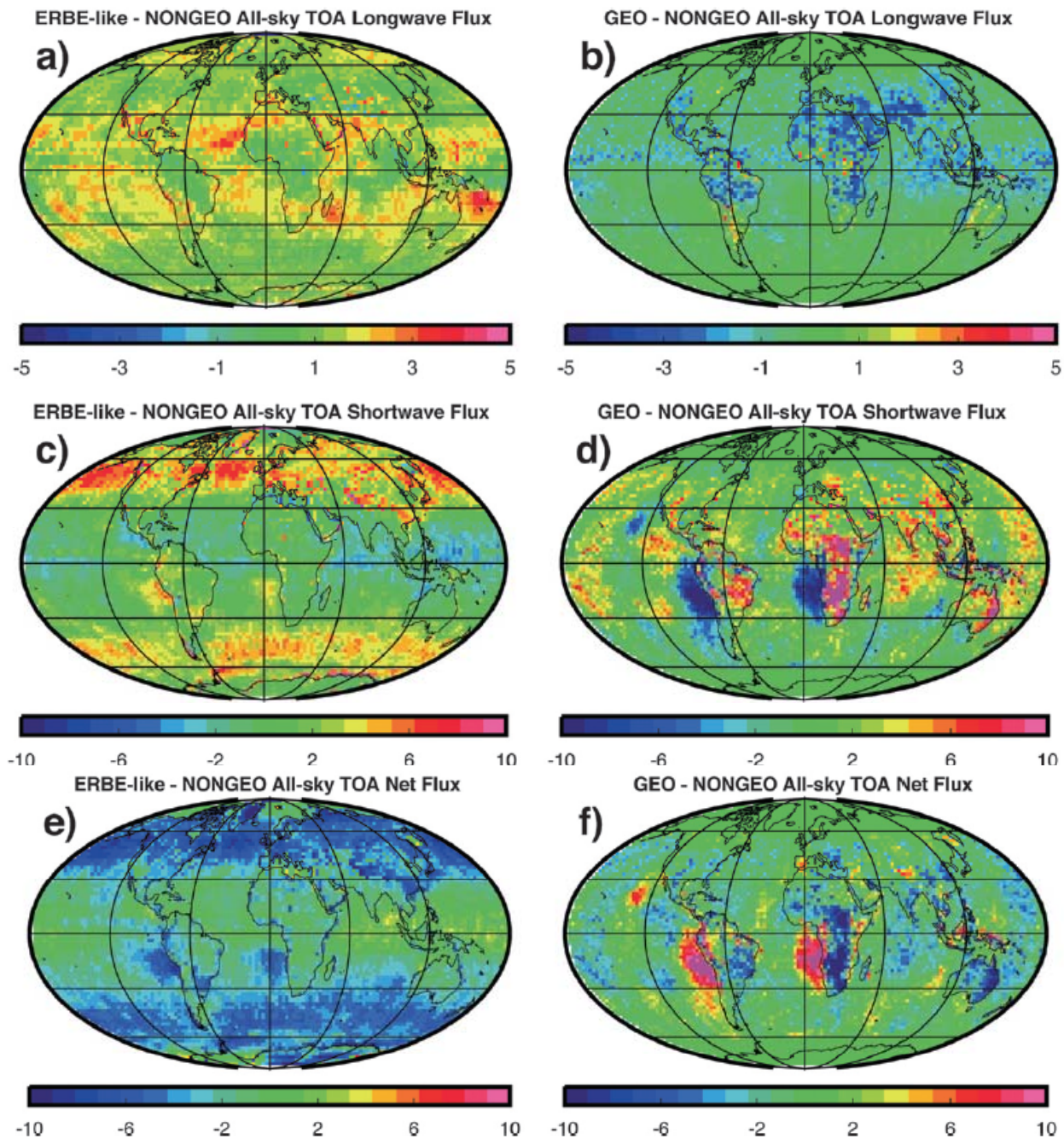
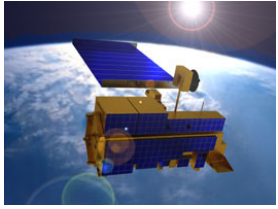


FIG. 1. Annual mean TOA flux difference between (left) CERES ERBE-like and CERES SRBAVG-nonGEO and (right) CERES SRBAVG-nonGEO and SRBAVG-GEO for (a), (b) LW; (c), (d) SW; and (e), (f) net for the year 2002.



Outlines

- ♣ Background
- ♣ Insolation/PAR
- ♣ Albedo
- ♣ Longwave net radiation
- ♣ Evapotranspiration
- ♣ Summary

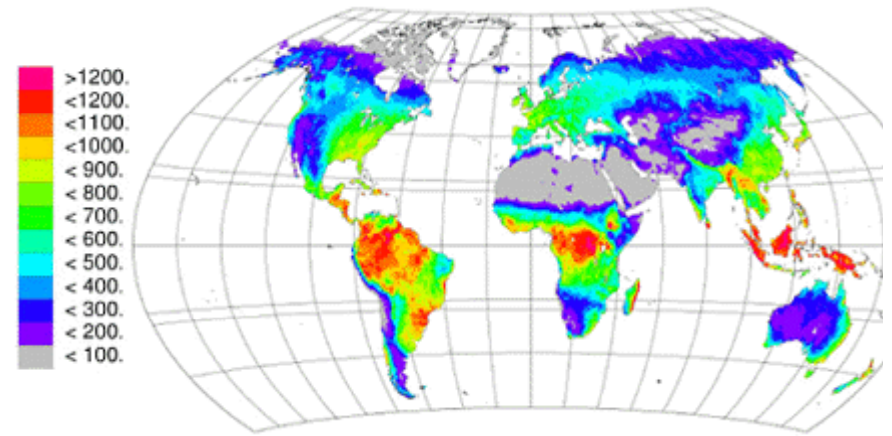
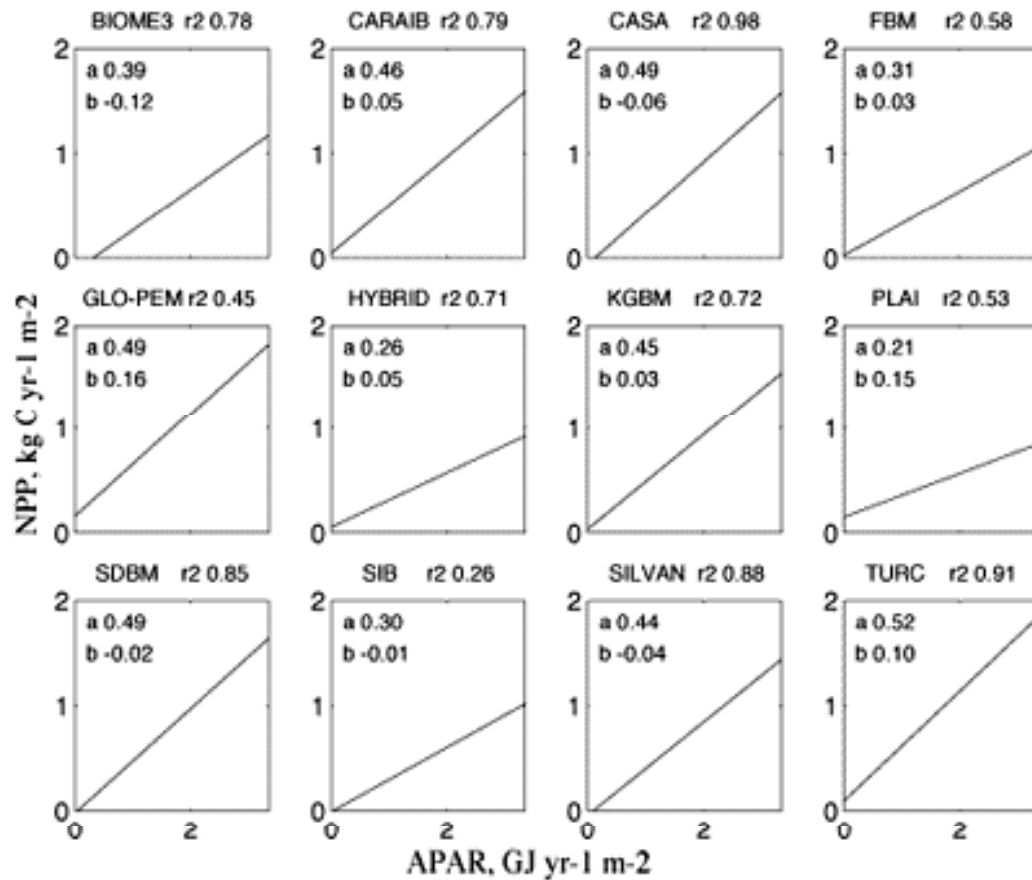


Production Efficiency Principles:

$$GPP = \varepsilon_g \bullet fPAR \bullet PAR$$

$$NPP = \varepsilon_n \bullet fPAR \bullet PAR$$

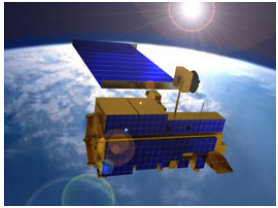
where PAR (MJ m⁻²) is in a time period (day, month), FAPAR is the fraction of PAR absorbed by vegetation canopy, and ε_g is the light use efficiency (LUE, g C MJ⁻¹ PAR) in GPP calculation, and ε_n is the light use efficiency in NPP calculation.



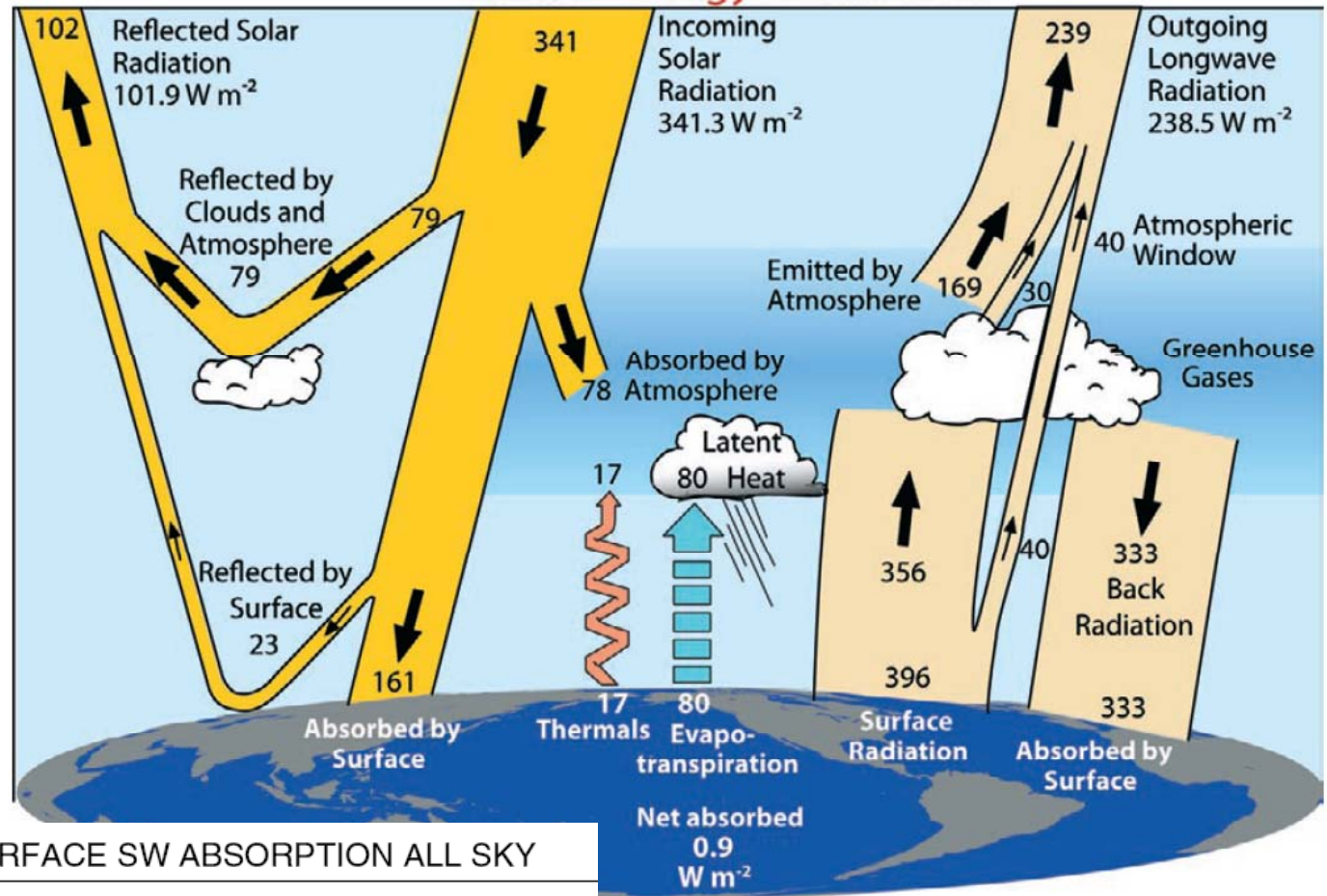
Annual net primary production (g C m⁻² yr⁻¹) estimated as the average of all model NPP estimates.

Grid cell level regression of net primary production (NPP) (kg C yr⁻¹ m⁻²) against absorbed photosynthetically active radiation (APAR) (GJ yr⁻¹ m⁻²)

Cramer, et al., 1995, Net primary productivity model inter-comparison activity, IGBP/GAIM report series 5

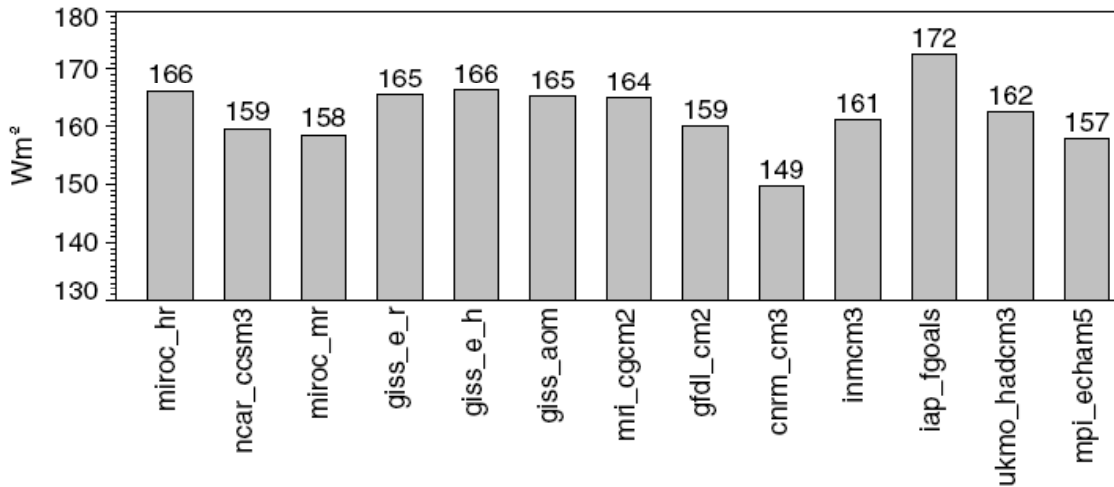


Global Energy Flows $W m^{-2}$

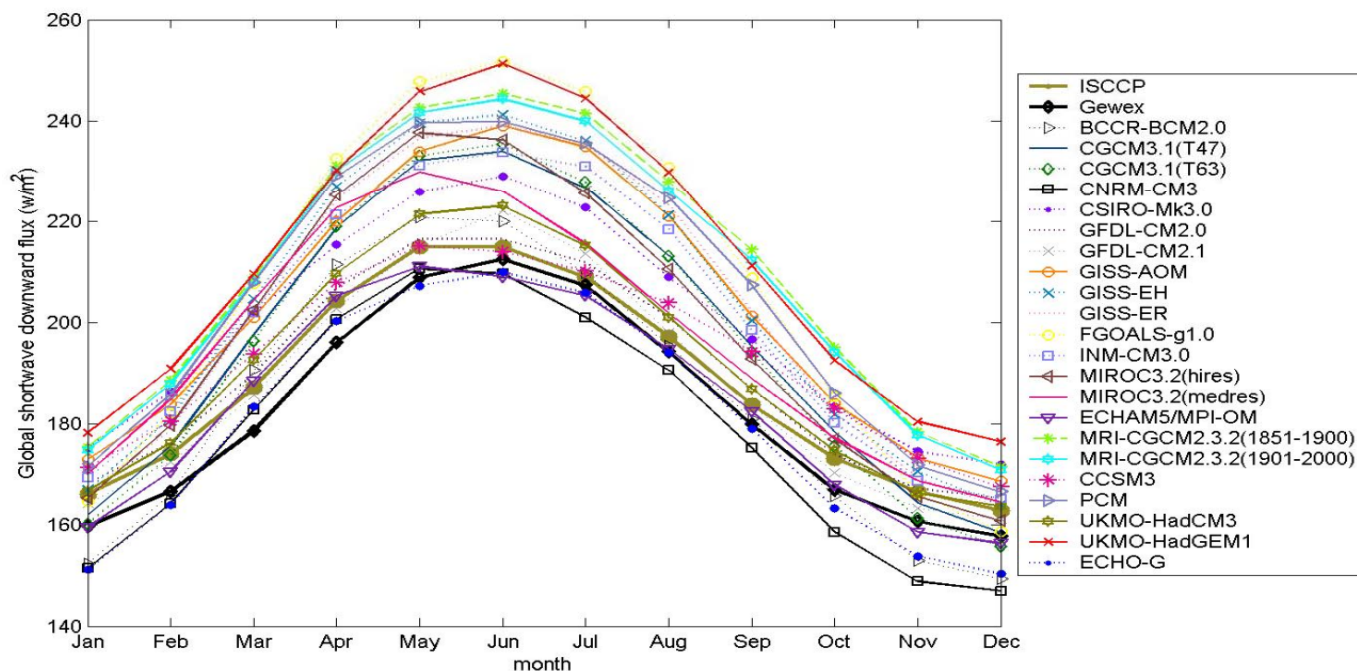


Model range: **24**
 Standard dev.: **5.5**

IPCC AR4 : SURFACE SW ABSORPTION ALL SKY



Radiation budgets in IPCC AR4 GCMs

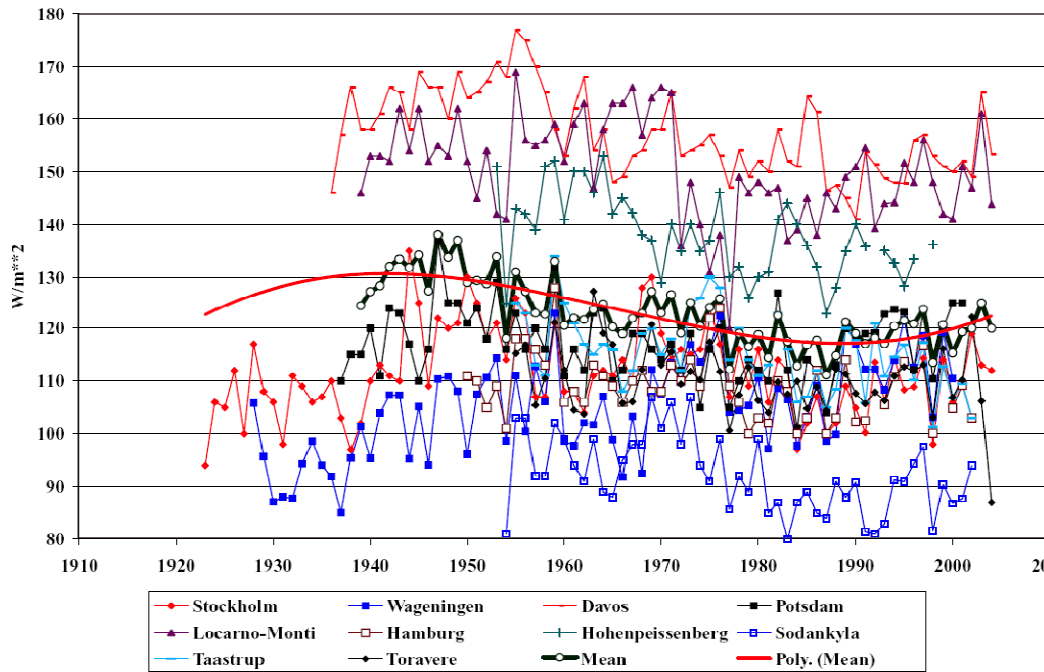


IPCC AR4 GCM model simulations and two satellite products

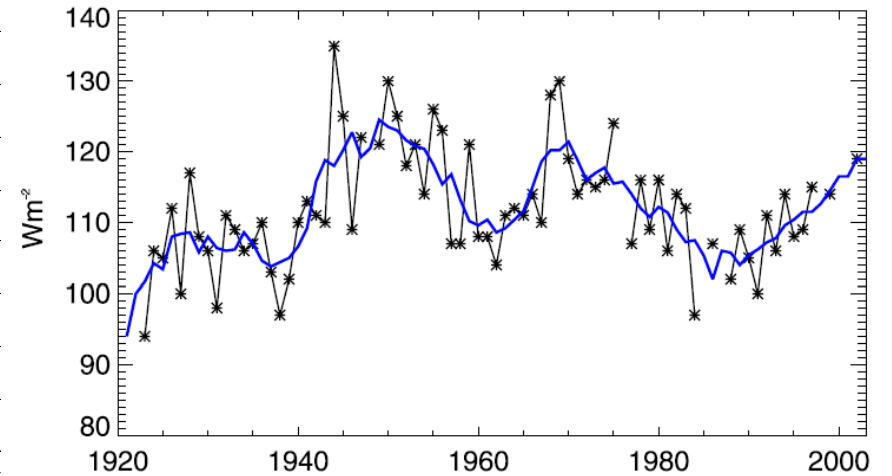
Table 1 Estimated averaged Insolation over land surfaces (Wm^{-2})

dataset	ISCCP-FD [125]	NRA [125]	ERA-40 [125]	JRA [125]	Trenberth et al. [125]	AMIP II GCMs mean [140]	GEBA mean [140]	IPCC AR4 GCM mean	GEWEX
Feb. 1985- April 1989	190.1	224.1	177.2	206.4		178	169	175	
Mar 2000 – May 2004	188.8	225.4	-	207.4	184.7				182.4

Global radiation for Europe for sites with more than 50 years observation



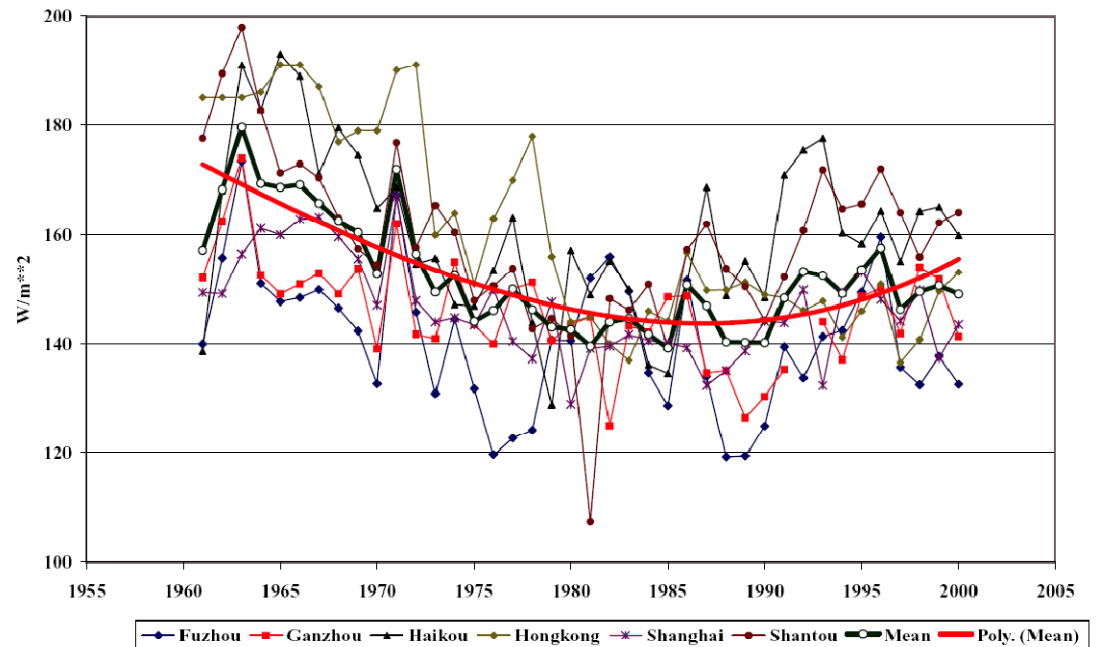
Stockholm



“global dimming”: Substantial decline in solar radiation the surface

“global brightening”: Substantial increase in solar radiation the surface

Best 6 stations in southeast China



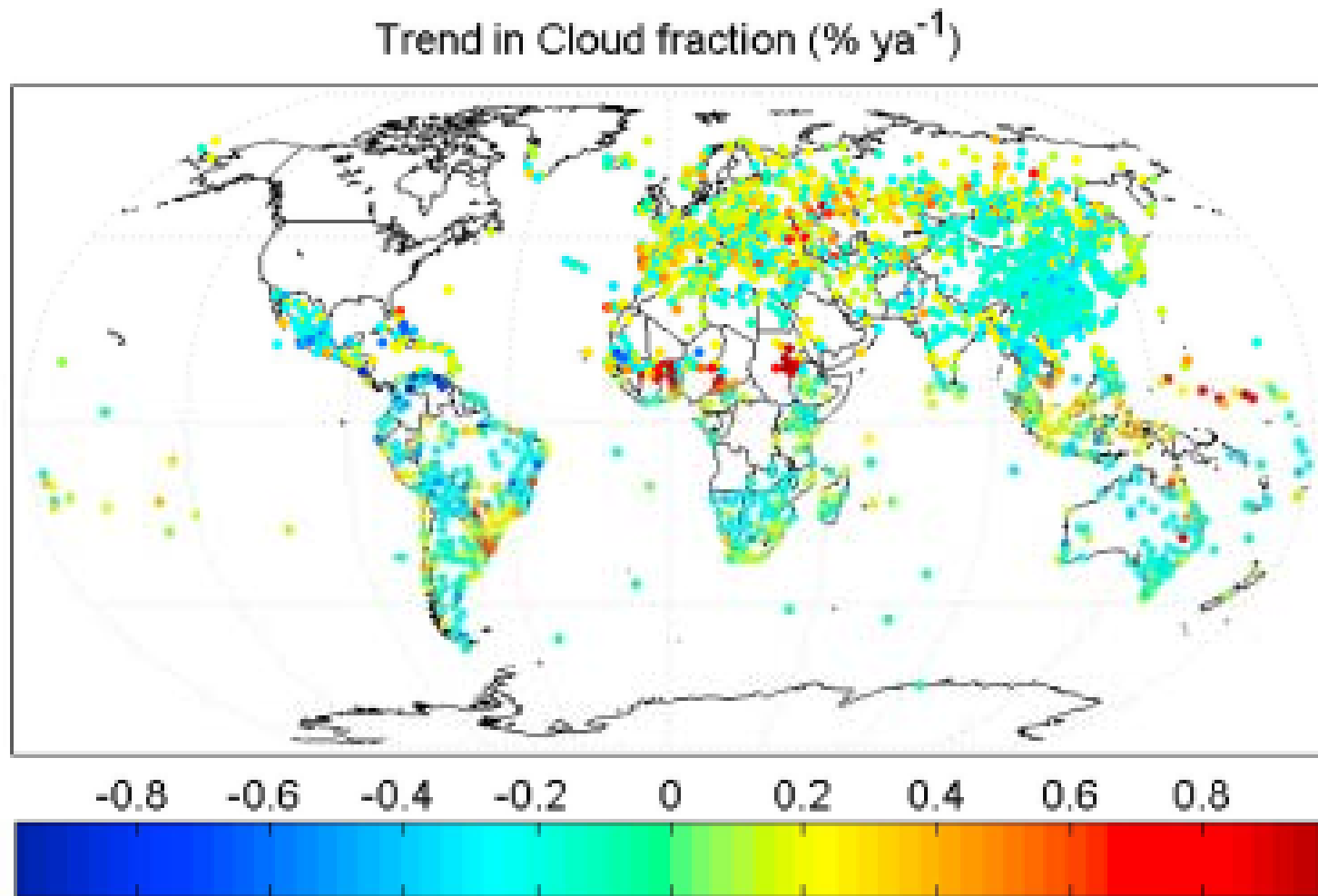
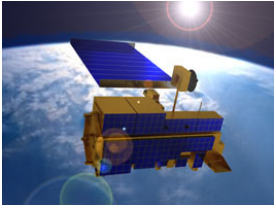
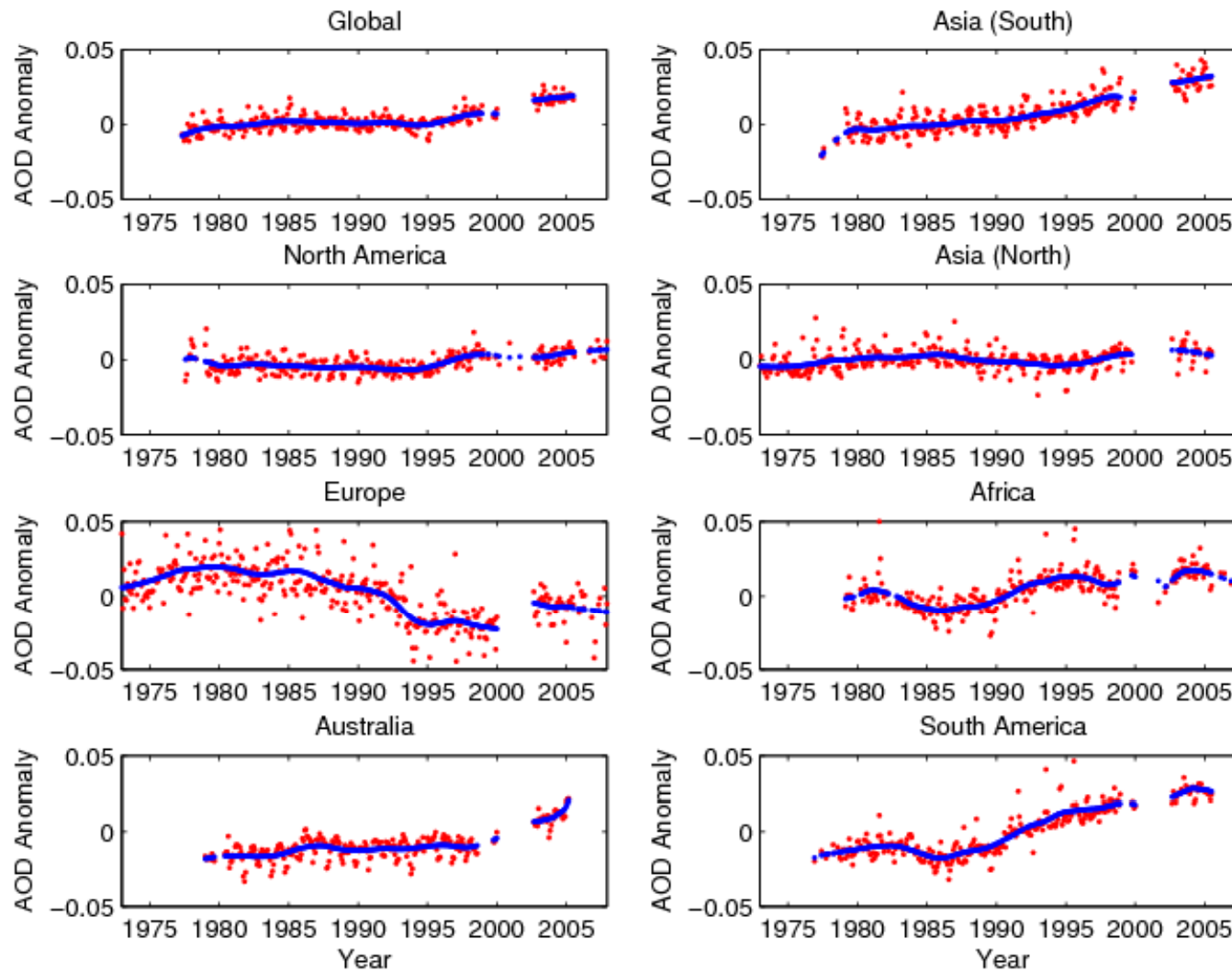


Figure 6. Linear trend of daily total cloud coverage from 1973 to 2008. The monthly total cloud cover fraction anomaly is derived and used to calculate the linear trend using the Mann-Kendall trend test method, and only stations that pass the 95% significance level in the Mann-Kendall trend test are shown. Some sites over North America and some European countries changed the observational method from human visual observations to instrument observations during the 1990s and are excluded because they show obvious discontinuities in total cloud coverage.



Wang, K., R. Dickinson and **S. Liang**, (2009), Clear sky visibility has decreased over land globally from 1973 to 2007, *Science*, 323, 1468-1470



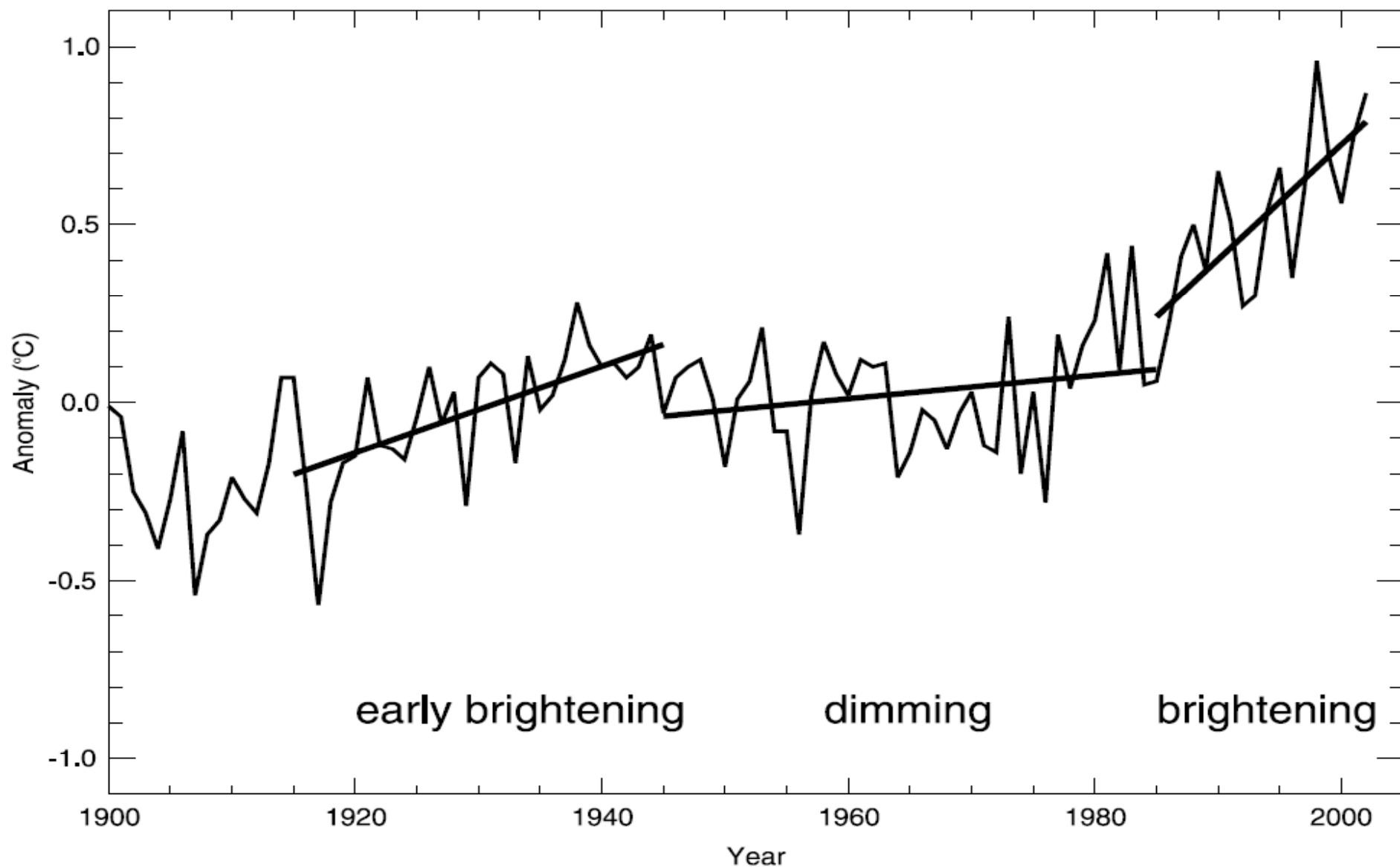


Figure 9. Observed 2-m temperature anomalies over global land surfaces during the 20th century. There is indication for a suppression of greenhouse-induced warming through “global dimming” between the 1950s and 1980s, and an enhancement through “brightening” between the 1920s and 1940s as well as from the 1980s onward. Anomalies with respect to the 20th century average. Units are °C. Adapted from *Wild et al.* [2007].

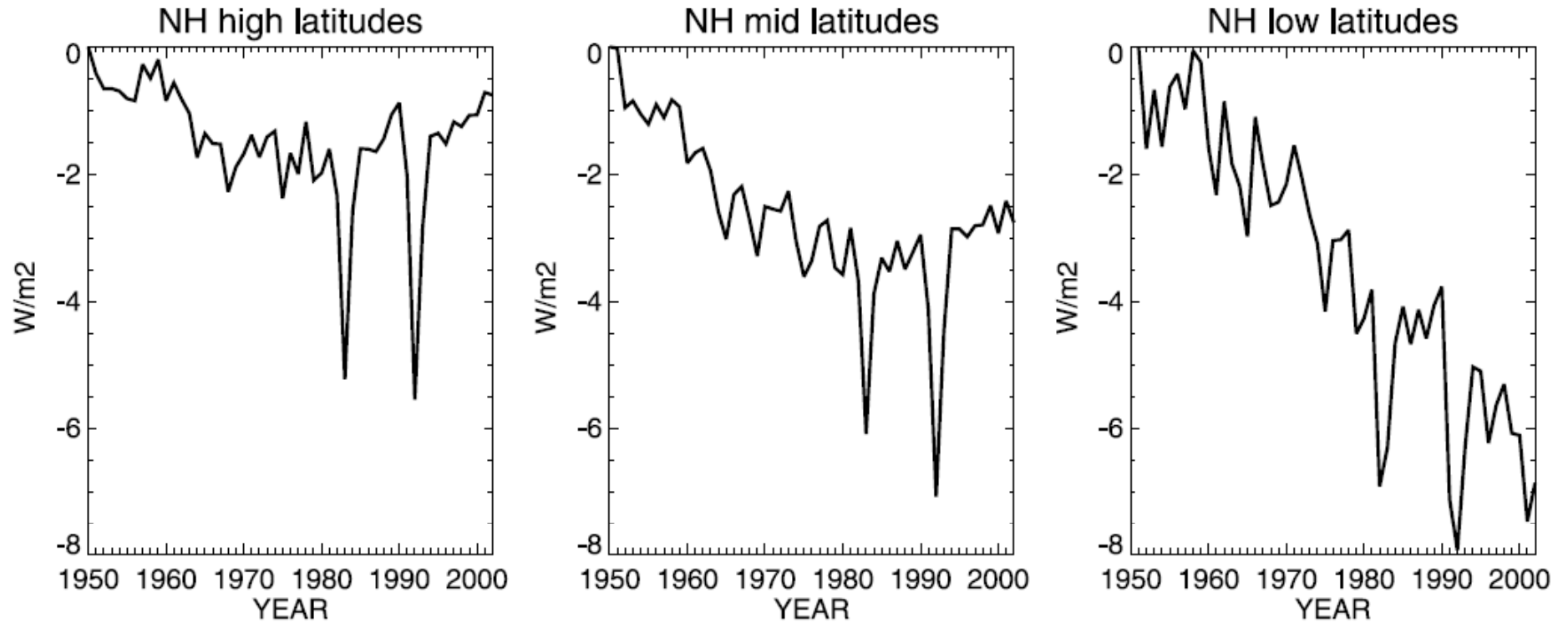
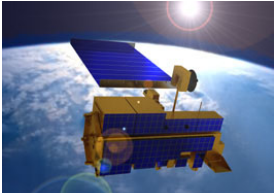


Figure 5. Simulated annual clear-sky surface solar radiation anomalies over the period 1950–2000 in different latitude belts of the Northern Hemisphere: High latitudes (60°N – 90°N), middle latitudes (30°N – 60°N), and low latitudes (0° – 30°N). Simulations done with the aerosol-climate modeling system ECHAM5 HAM [Stier *et al.*, 2005, 2006]. Reference value is 1950. Units are W m^{-2} .

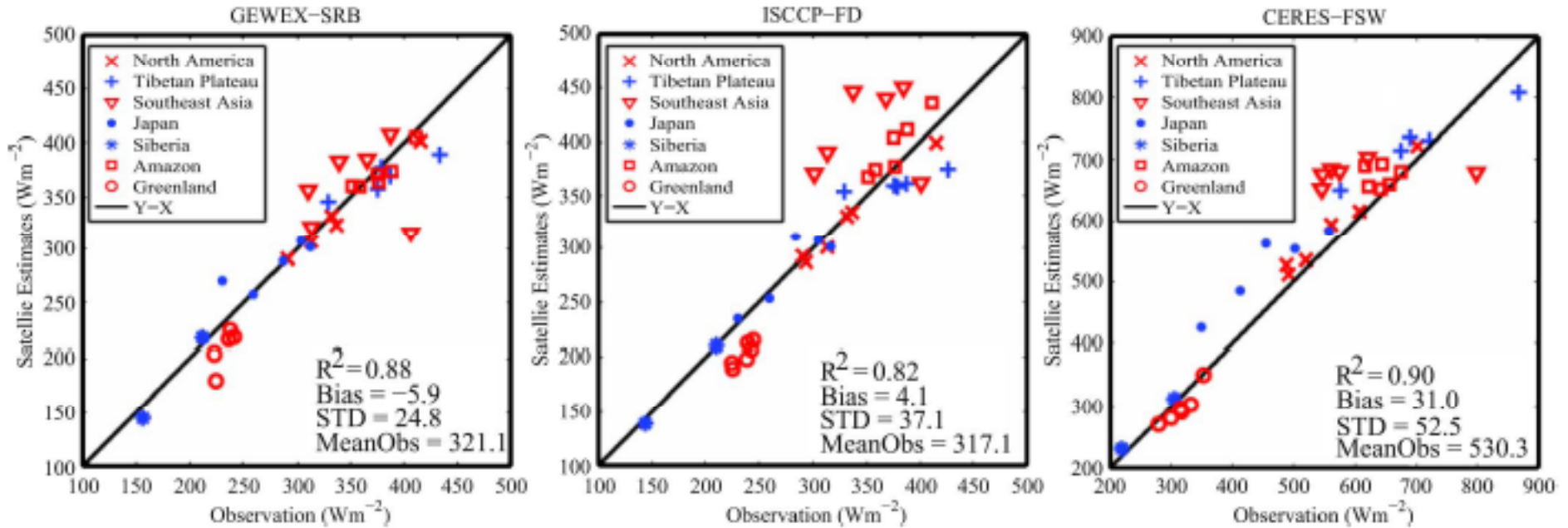


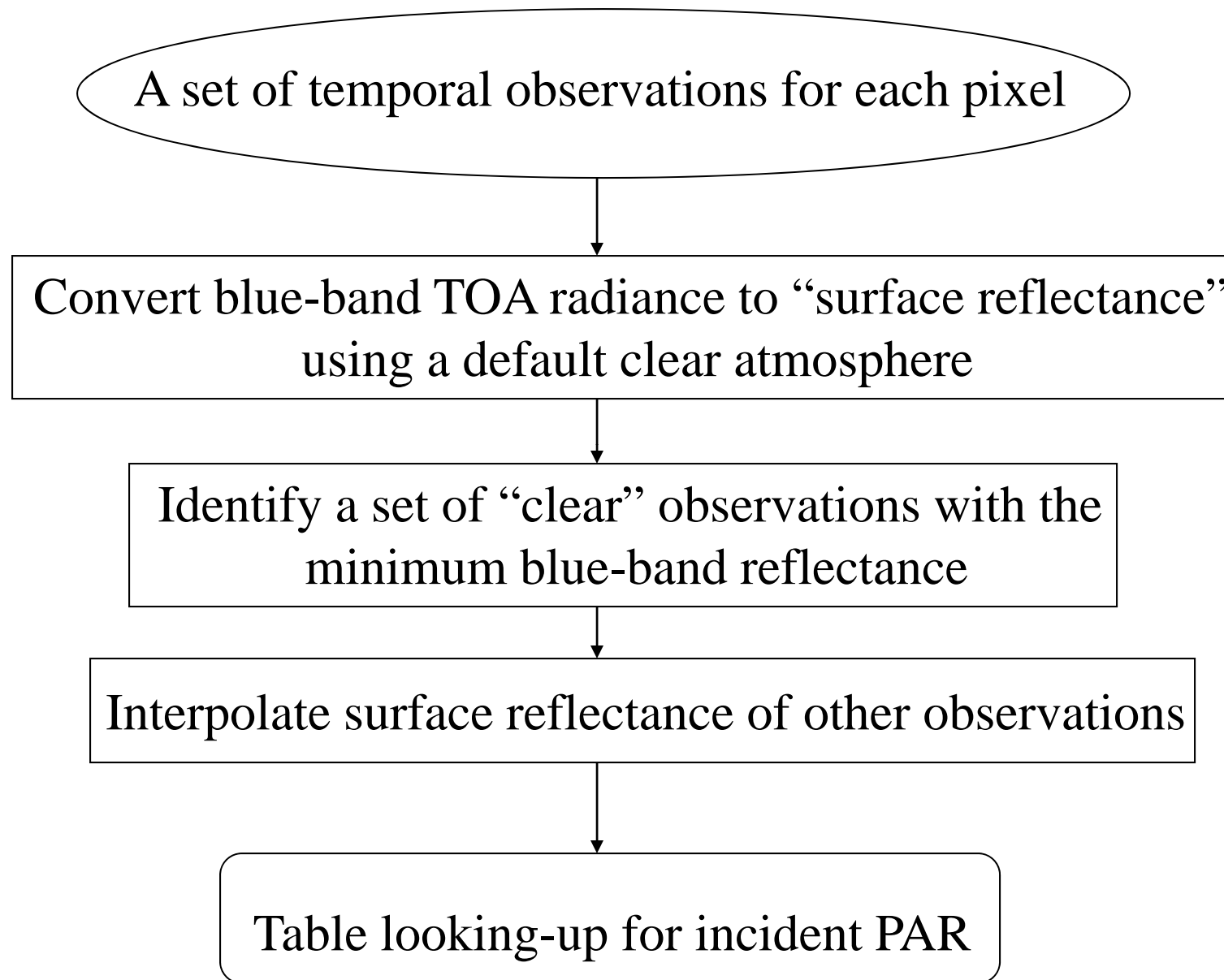
Fig. 2. Comparison of three-year mean DSSR between satellite estimates and ground measurements at 36 sites in seven regions, averaged from three hourly or hourly values over 2000–2002.

Gui, et al., *IEEE GRSL*, 2010



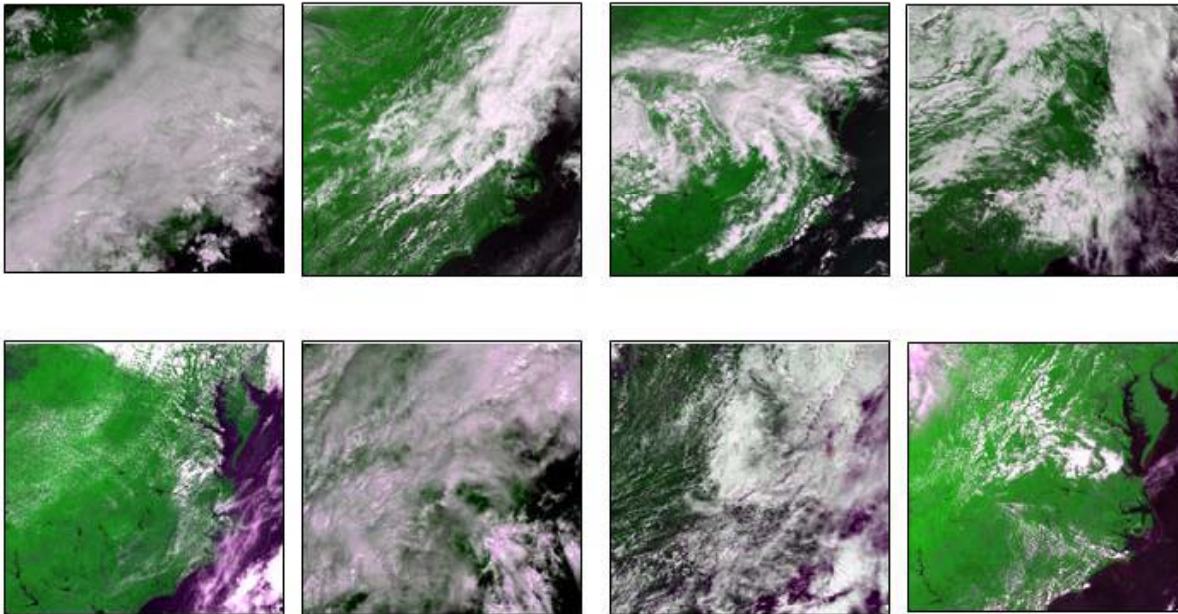
Estimating incident solar radiation

- **Liang, S.**, Zheng, T., Liu, R., Fang, H., Tsay, S.C., & Running, S. (2006). Mapping incident Photosynthetically Active Radiation (PAR) from MODIS Data. *Journal of Geophysical Research-Atmospheres*, 111, Art. No. D15208, doi:15210.11029/12005JD006730.
- **Liang, S.L.**, Zheng, T., Wang, D.D., Wang, K.C., Liu, R.G., Tsay, S.C., Running, S., & Townshend, J. (2007). Mapping high-resolution incident photosynthetically active radiation over land from polar-orbiting and geostationary satellite data. *Photogrammetric Engineering and Remote Sensing*, 73, 1085-1089
- Liu, R., **Liang, S.**, He, H., Liu, J., & Zheng, T. (2008). Mapping photosynthetically active radiation from MODIS data in China. *Remote Sensing of Environment*, 112, 998-1009
- Wang, D., **Liang, S.**, & Zheng, T. (2010). Estimation of daily-integrated PAR from MODIS data. *International Journal of Remote Sensing*, 31(6), 1661, doi:10.1080/01431160903475407
- Zheng, T., **Liang, S.**, & Wang, K.C. (2008). Estimation of incident PAR from GOES imagery. *Journal of Applied Meteorology and Climatology*, 47, 853-868

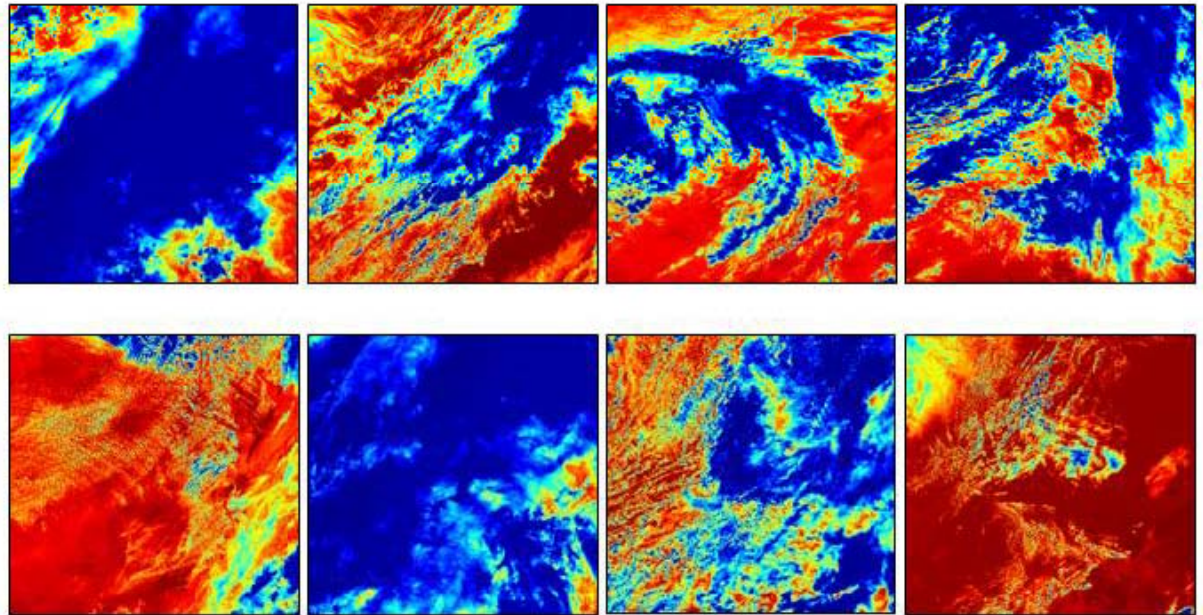


Flowchart of the algorithm using the temporal signatures

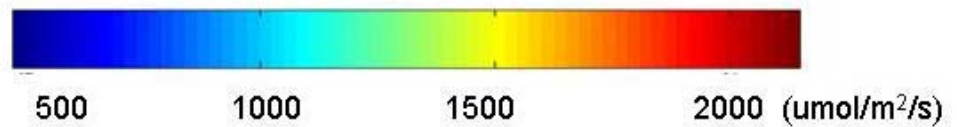
(Liang, et al., 2006)

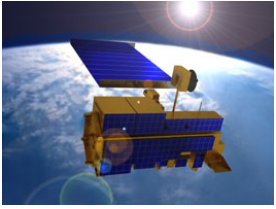


MODIS TOA radiance on 8 days in 2003: May 22, May 25, May 29, May 31, June 5, June 7, June 8, and June 10

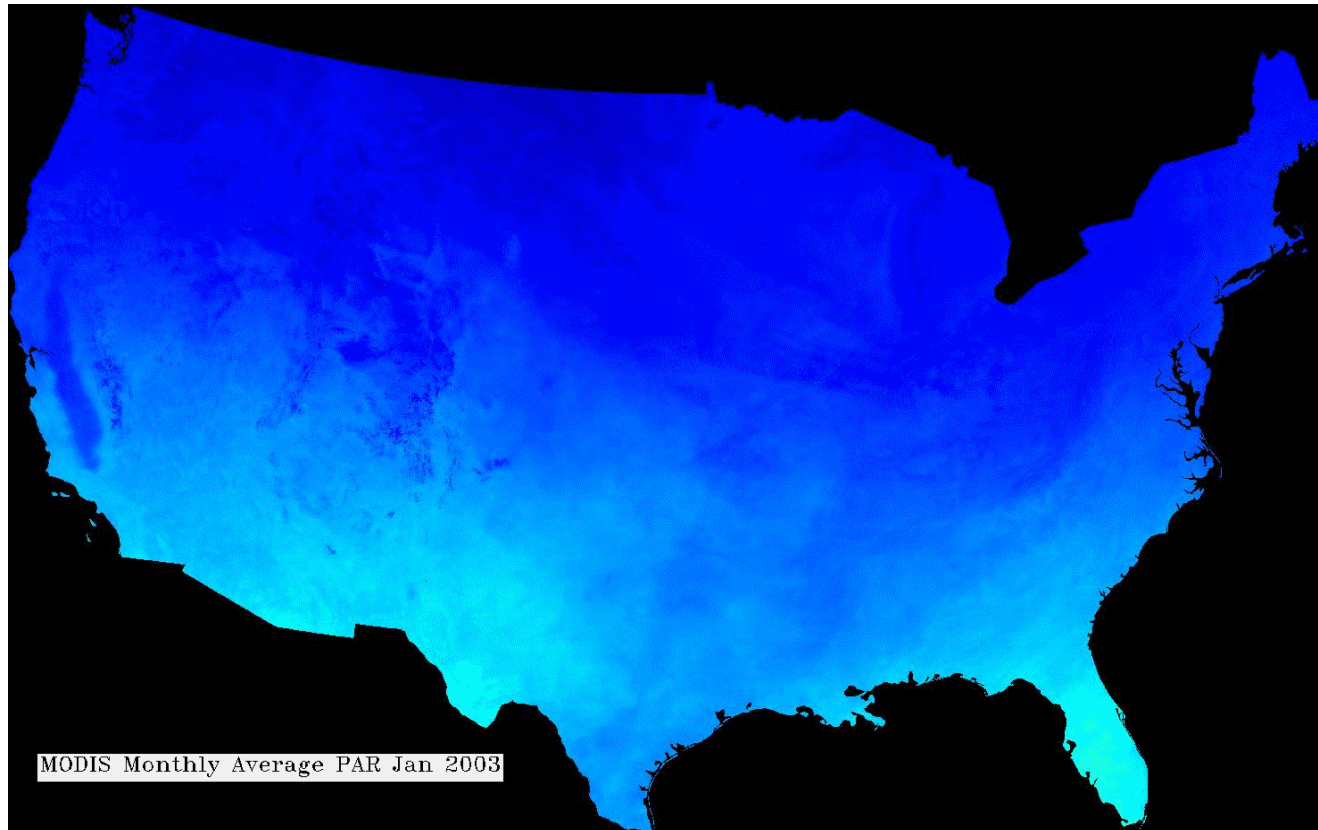


Estimated incident PAR

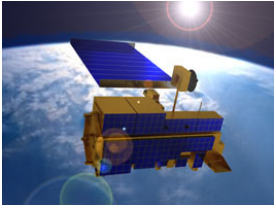




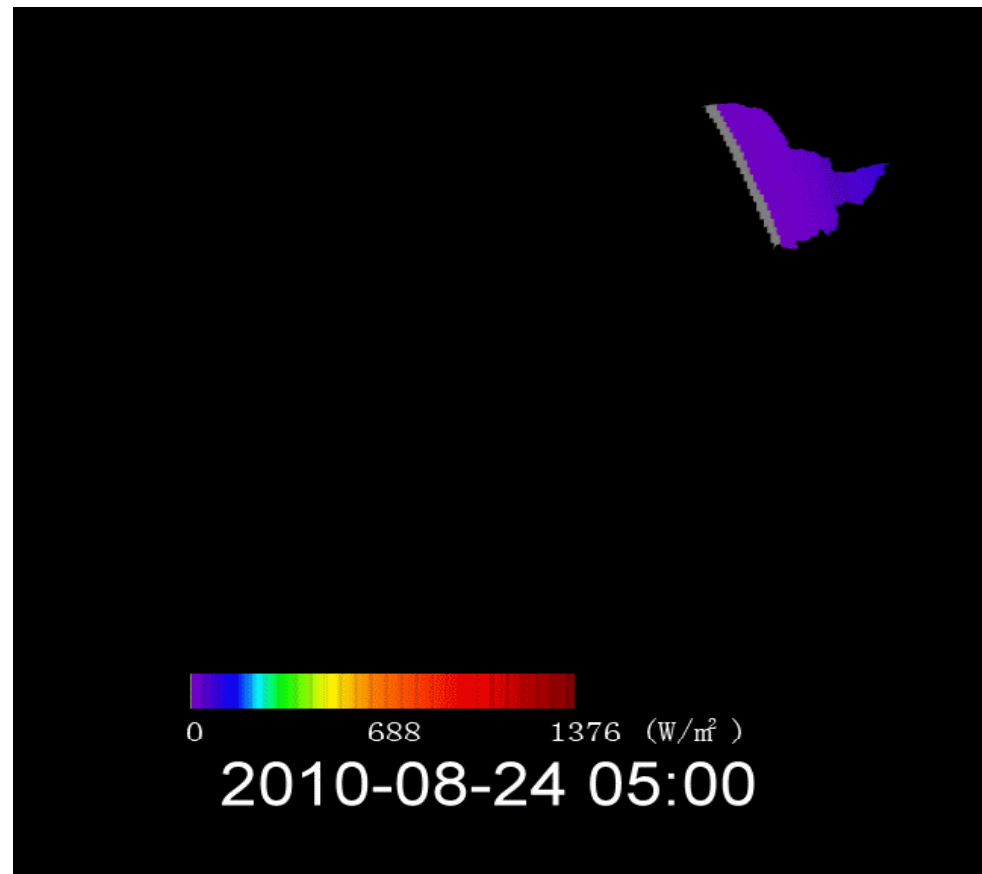
Monthly PAR from MODIS



Three years (2003-2005) daily incident PAR at 4km over North America available now; Bob Cook has kindly agreed to distribute them at ORNL DAAC.

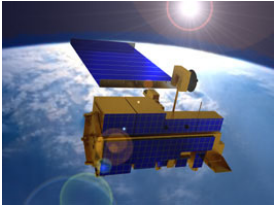


Downwelling Shortwave Radiation

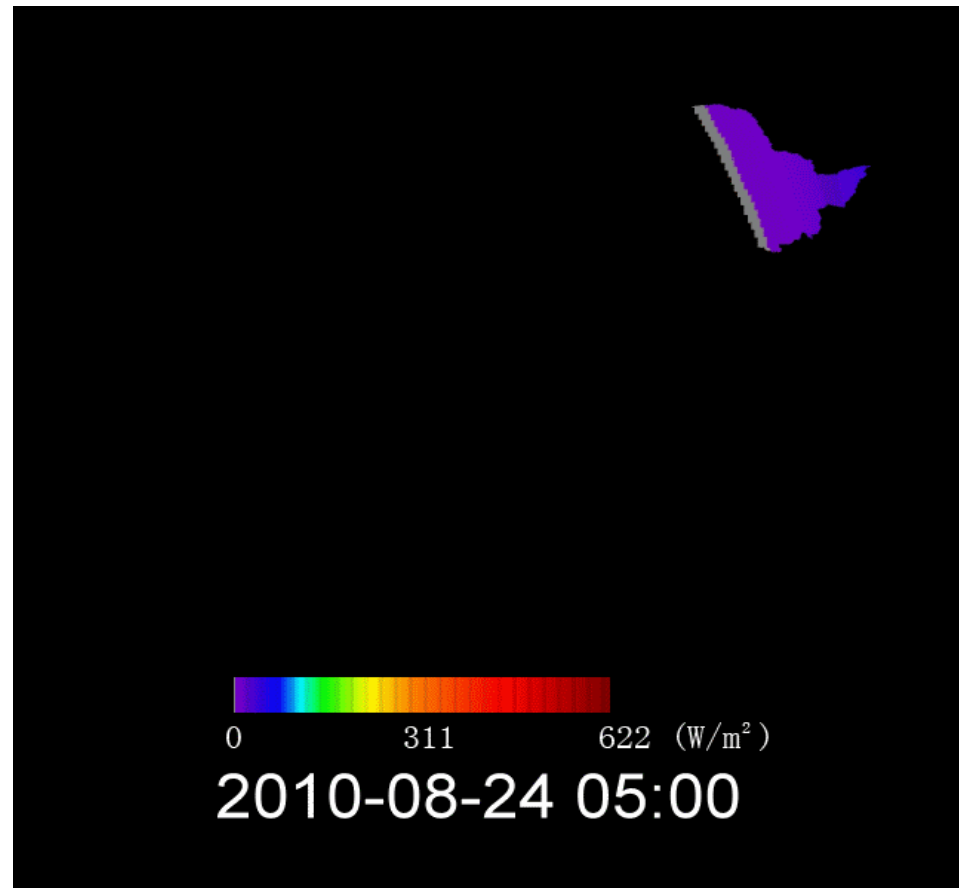


Downward Surface Shortwave Radiation of China from MTSAT images.

3 hours ,5km, (2008-2010, global)

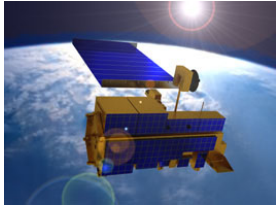


Photosynthetically Active Radiation



Photosynthetically Active Radiation of China from MTSAT images.

3hours, 5km. (2008-2010, global)

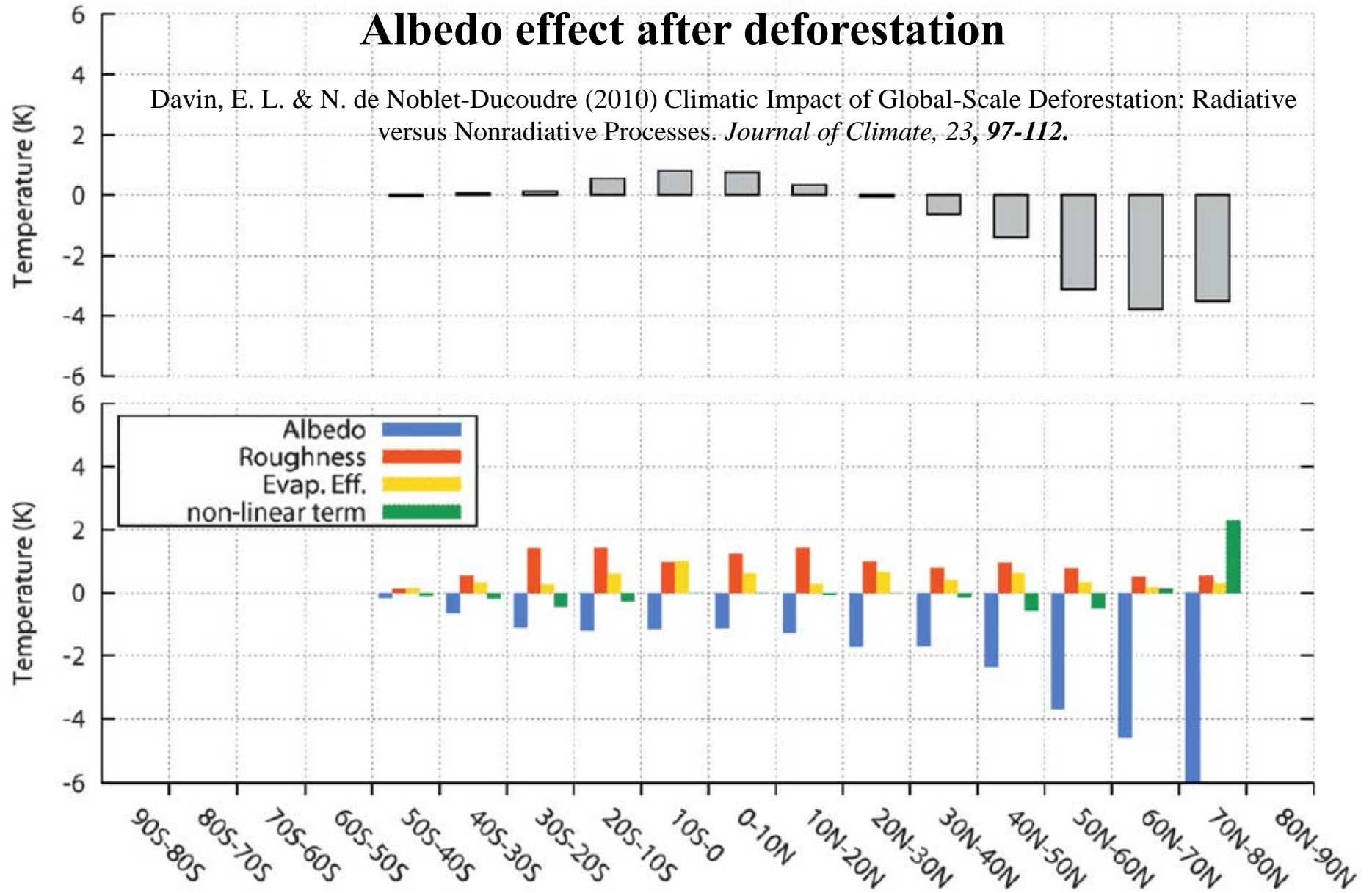


Outlines

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Albedo effect after deforestation

Davin, E. L. & N. de Noblet-Ducoudre (2010) Climatic Impact of Global-Scale Deforestation: Radiative versus Nonradiative Processes. *Journal of Climate*, 23, 97-112.



Annual mean change in surface temperature zonally averaged over deforested areas only. The bottom panel of each figure indicates the relative contribution of change



Albedo-based geoengineering solutions to offset CO₂

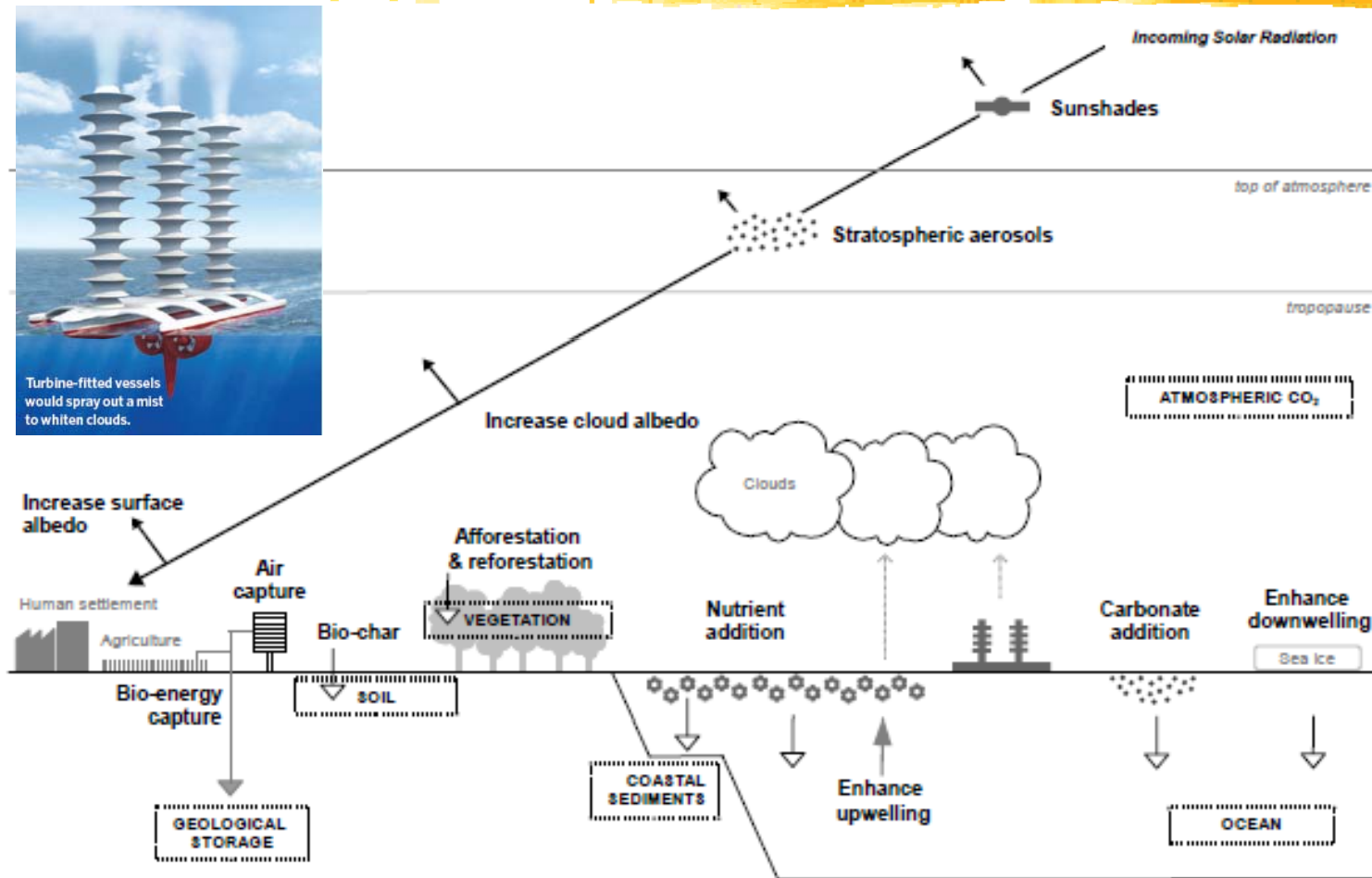


Fig. 1. Schematic overview of the climate geoengineering proposals considered. Black arrowheads indicate shortwave radiation, white arrowheads indicate enhancement of natural flows of carbon, grey downward arrow indicates engineered flow of carbon, grey upward arrow indicates engineered flow of water, dotted vertical arrows illustrate sources of cloud condensation nuclei, and dashed boxes indicate carbon stores. From Vaughan and Lenton (2009), not to scale.



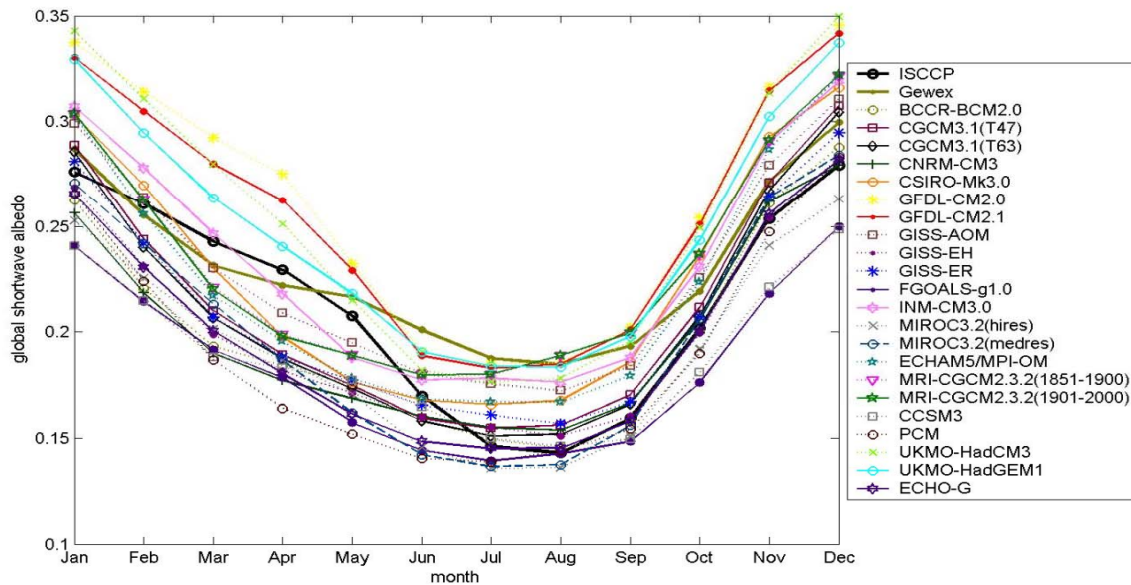
Table 1. Estimated radiative forcing potential of geoengineering options to alter planetary albedo. In common with the literature, for the first option, a back calculation is made of the albedo change required to counteract the radiative forcing due to doubling CO₂, whereas for the remaining options, forward calculations are made of the maximum potential radiative forcing. The latter are based on upper limit albedo changes and areas suggested in the literature (see text). Calculations were done at full precision but outputs are generally given to 2 significant figures commensurate with our first order approach.

Option	Area (m ²)	Fraction of Earth f_{Earth}	Albedo change within area $\Delta\alpha$	Scaled albedo change of layer	Transmittance factor f_a	Planetary albedo change $\Delta\alpha_p$	Solar radiation at TOA S_0 (Wm ⁻²)	Radiative forcing RF (Wm ⁻²)
<i>Increase marine cloud albedo</i>				$\Delta\alpha_a$				
Mechanical	8.9×10^{13}	0.175	0.074	0.013	0.84	0.011	345	-3.71
Biological	5.1×10^{13}	0.1	0.008	0.000067*	0.84	0.000056	345	-0.019
<i>Increase land surface albedo</i>				$\Delta\alpha_s$				
Desert	1.0×10^{13}	0.02	0.44	0.0088	0.73	0.0064	330	-2.12
Grassland	3.85×10^{13}	0.075	0.0425	0.0032	0.48	0.0015	330	-0.51
Cropland	1.4×10^{13}	0.028	0.08	0.0022	0.48	0.0011	330	-0.35
Settlements	3.25×10^{12}	0.0064	0.15	0.00096	0.48	0.00046	330	-0.15
Urban areas	1.5×10^{12}	0.0029	0.1	0.00029	0.48	0.00014	330	-0.047



Table 2. Estimated radiative forcing potential of carbon cycle geoengineering options. Effects are calculated relative to a strong mitigation scenario in which a total of 1000 PgC are emitted and atmospheric CO₂ (and corresponding radiative forcing) reaches 450 ppm (2.58 W m⁻²) in 2050, stabilises at 500 ppm (3.14 W m⁻²) in 2100 and then declines to 363 ppm (1.43 W m⁻²) on a millennial timescale.

Geoengineering Option	2050		2100		ΣC_{seq} (PgC)	3000	
	ΔCO_2 (ppm)	RF (W m ⁻²)	ΔCO_2 (ppm)	RF (W m ⁻²)		ΔCO_2 (ppm)	RF _{final} (W m ⁻²)
<i>Enhance land carbon sink</i>							
Afforestation	-41	-0.49	-34	-0.37	183	-16	-0.27
Bio-char production	-10	-0.12	-37	-0.40	399	-34	-0.52
Air capture and storage	-58	-0.74	-186	-2.5	>1000	> -85	> -1.43
<i>Enhance ocean carbon sink</i>							
Phosphorus addition	-5.9	-0.070	-12	-0.13	574	-52	-0.83
Nitrogen fertilisation	-4.5	-0.054	-9.3	-0.10	299	-25	-0.38
Iron fertilisation	-9.0	-0.11	-19	-0.20	227	-19	-0.29
Enhance upwelling	-0.1	-0.0017	-0.3	-0.0032	16*	-1.9	-0.028
Enhance downwelling	-0.08	-0.00095	-0.18	-0.0019	9*	-1.1	-0.016
Carbonate addition	-0.4	-0.0048	-2.3	-0.025	251*	-30	-0.46



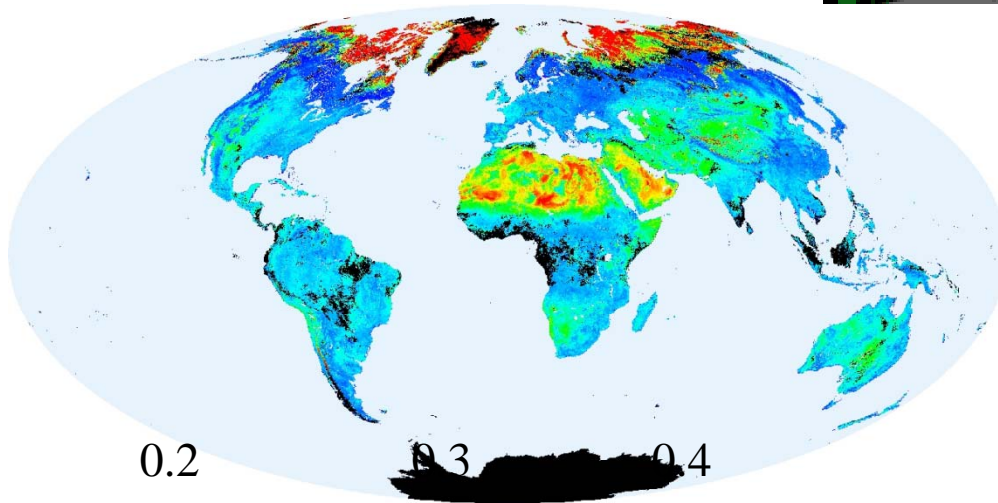
IPCC AR4 GCM model simulations and two satellite products

Land Surface Albedo

Model ID	Global	70°-90°N	50°-70°N	30°-50°N	10°-30°N	10°S-10°N	10°-30°S	30°-50°S	50°-70°S	70°-90°S
MODIS	0.240	0.526	0.230	0.208	0.266	0.148	0.160	0.155	0.716	0.747
ISCCP	0.211	0.579	0.261	0.183	0.201	0.093	0.117	0.104	0.701	0.773
GEWEX	0.228	0.520	0.217	0.187	0.241	0.153	0.161	0.157	0.692	0.767
BCCR-BCM2.0	0.194	0.541	0.255	0.191	0.190	0.106	0.139	0.119	0.598	0.752
CGCM3.1(T47)	0.205	0.445	0.278	0.180	0.155	0.125	0.238	0.234	0.796	0.786
CGCM3.1(T63)	0.202	0.495	0.297	0.204	0.181	0.129	0.252	0.218	0.836	0.782
CNRM-CM3	0.195	0.610	0.210	0.171	0.179	0.136	0.215	0.156	0.735	0.708
CSIRO-Mk3.0	0.221	0.659	0.252	0.271	0.201	0.126	0.167	0.194	0.822	0.799
GFDL-CM2.0	0.256	0.663	0.312	0.251	0.210	0.124	0.143	0.167	0.806	0.806
GFDL-CM2.1	0.251	0.561	0.268	0.203	0.193	0.112	0.136	0.076	0.378	0.756
GISS-AOM	0.222	0.465	0.248	0.208	0.178	0.127	0.172	0.177	0.770	0.750
GISS-EH	0.196	0.389	0.241	0.209	0.178	0.118	0.146	0.139	0.669	0.774
GISS-ER	0.203	0.486	0.262	0.215	0.178	0.123	0.151	0.147	0.726	0.787
FGOALS-g1.0	0.178	0.539	0.284	0.167	0.125	0.111	0.186	0.172	0.778	0.779
INM-CM3.0	0.227	0.492	0.302	0.237	0.199	0.142	0.203	0.227	0.755	0.782
MIROC3.2(hires)	0.187	0.479	0.272	0.176	0.134	0.106	0.175	0.163	0.767	0.740
MIROC3.2(medres)	0.195	0.513	0.282	0.181	0.135	0.107	0.178	0.155	0.768	0.712
ECHAM5/MPI-OM	0.217	0.761	0.274	0.224	0.209	0.108	0.161	0.194	0.866	0.865
MRI-CGCM2.3.2	0.225	0.448	0.207	0.211	0.182	0.148	0.254	0.234	0.791	0.784
CCSM3	0.186	0.745	0.286	0.205	0.164	0.097	0.111	0.129	0.929	0.828
PCM	0.185	0.524	0.242	0.149	0.175	0.126	0.181	0.124	0.824	0.803
UKMO-HadCM3	0.248	0.432	0.292	0.205	0.166	0.135	0.193	0.218	0.775	0.777
UKMO-HadGEM1	0.243	0.508	0.283	0.213	0.201	0.136	0.207	0.215	0.776	0.790
ECHO-G	0.193	0.413	0.284	0.192	0.174	0.124	0.164	0.139	0.744	0.755

Zhang, et al., *IEEE JSTARS*, 2010

Part of MODIS
algorithm for global
production



**Global White-sky
(bihemispherical) Albedo for the
period 30 Sep - 13 Oct, 2002**



bidirectional reflectance factor (BRF) $R(\theta_i, \phi_i, \theta_v, \phi_v)$, which is numerically equivalent to BRDF multiplied by π :

$$R(\theta_i, \phi_i, \theta_v, \phi_v) = \pi f(\theta_i, \phi_i, \theta_v, \phi_v) \quad \text{Equation 1-9}$$

we need surface *directional-hemispherical reflectance* (DHR), which can be integrated from BRF over all reflected directions:

$$r(-\mu_i, \phi_i) = \frac{1}{\pi} \int_0^{2\pi} \int_0^1 R(\mu_i, \phi_i, \mu, \phi) \mu d\mu d\phi \quad \text{Equation 1-10}$$

It is often called local or planar *albedo*, but is called black-sky albedo in the MODIS products of the NASA Earth Observing System (EOS) program. The *bihemispherical reflectance* (BHR) is a further integration of DHR over all illumination directions:

$$r_0 = 2 \int_0^1 r(-\mu_i, \phi_i) \mu_i d\mu_i \quad \text{Equation 1-11}$$

which is also called global or spherical albedo, or bright-sky albedo in the MODIS



Albedometer measurements

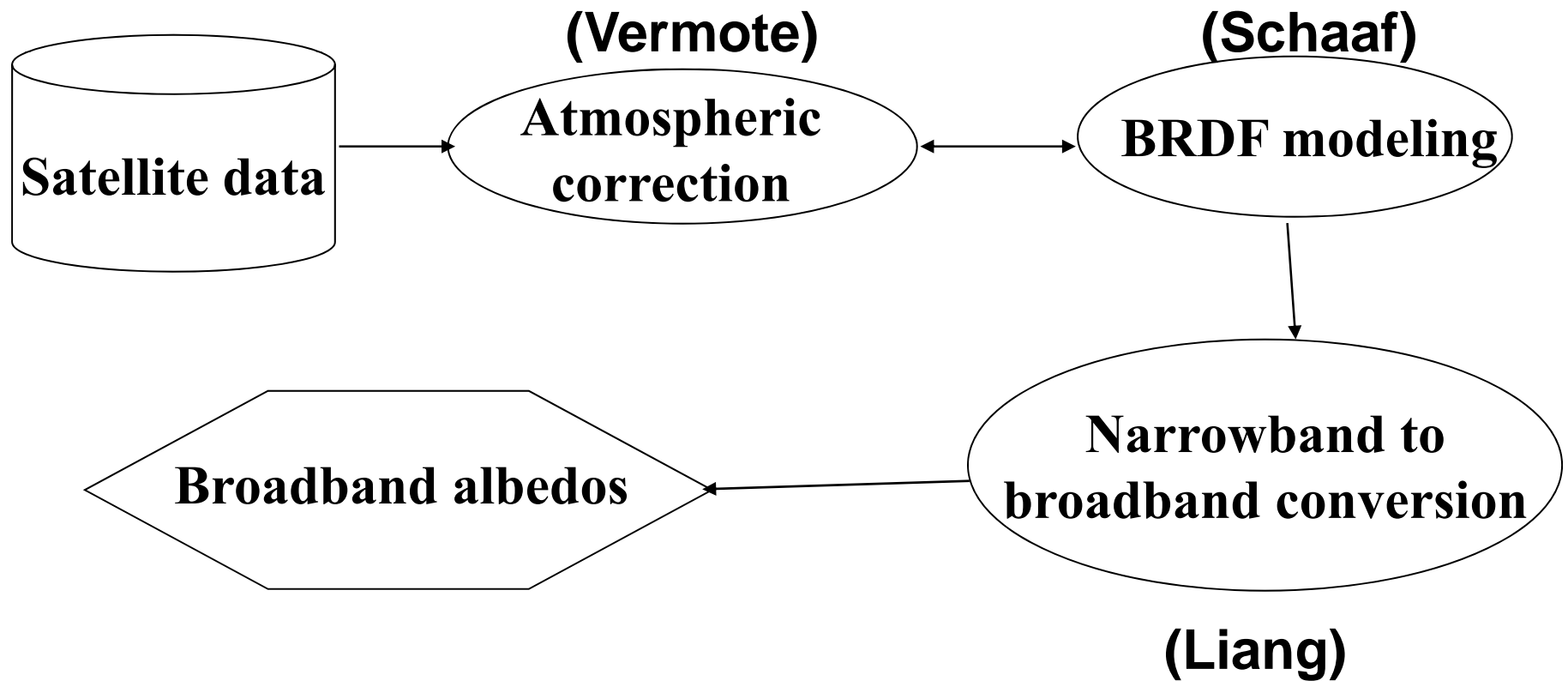
$$\alpha_s = \frac{E_s^\uparrow}{E_s^\downarrow} = \frac{\int_0^{2\pi} \int_0^1 I_s^\uparrow(\theta_s, \theta_v, \phi) \cdot \mu \cdot d\mu \cdot d\phi}{\int_0^{2\pi} \int_0^1 I_s^\downarrow(\theta_s, \theta_v, \phi) \cdot \mu^* \cdot d\mu^* \cdot d\phi}$$

$$I_s^\uparrow = \frac{1}{\pi} \int_0^{2\pi} \int_0^1 r_s(\theta_s, \theta_v, \phi) I_s^\downarrow(\theta_s, \theta_v, \phi) \mu^* d\mu^* d\phi$$

$$\alpha_s = (1 - s) \cdot \alpha_{black} + s \cdot \alpha_{white}$$



MODIS albedo algorithm





MODIS aerosol over the land

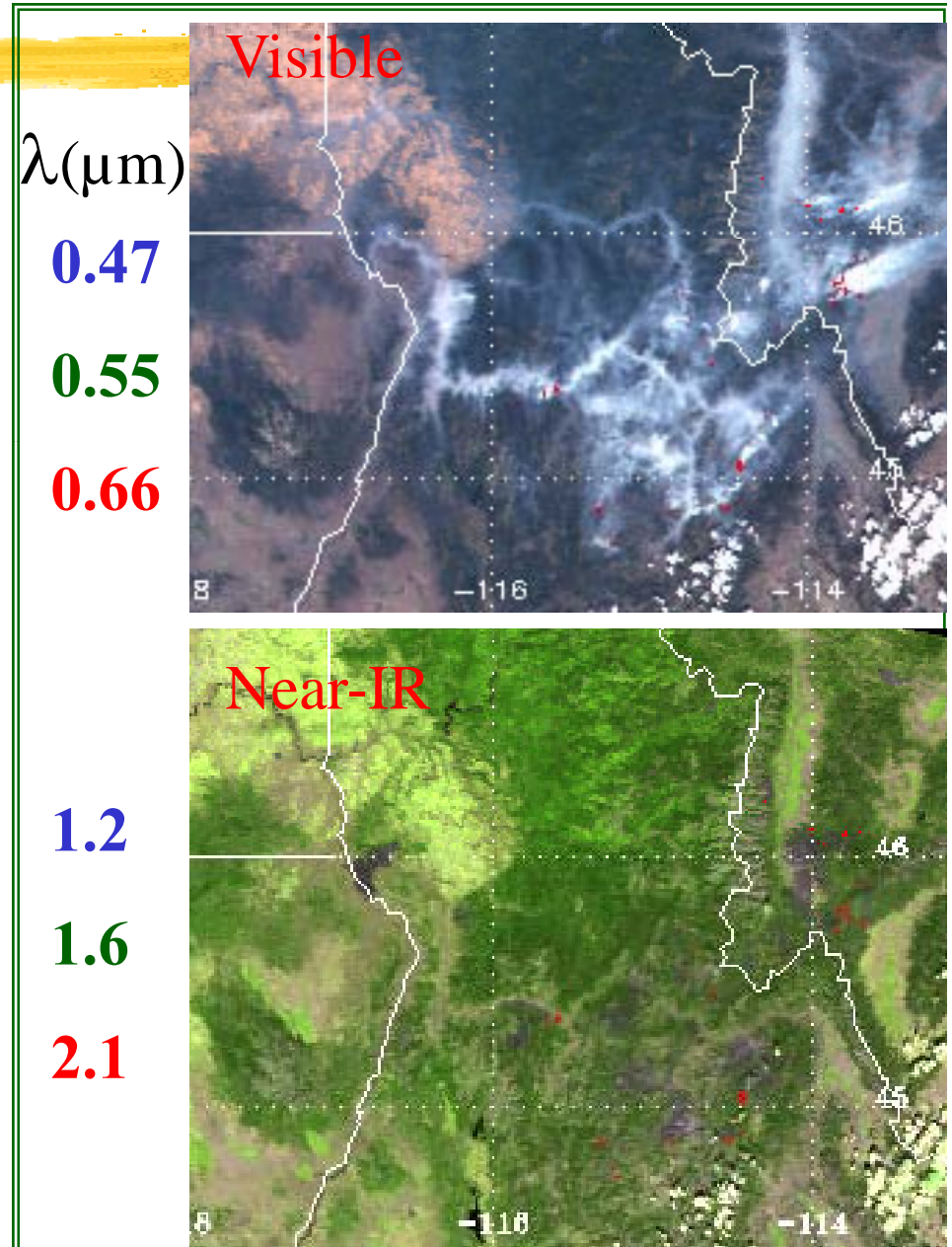
- Derive apparent reflectance (normalized radiance at the top of the atmosphere) at $2.2 \mu\text{m}$, $\rho_{2.2}^*$
- Assume that aerosol is transparent at $2.2 \mu\text{m}$, thus: $\rho_{2.2}^s = \rho_{2.2}^*$ where $\rho_{2.2}^s$ is the surface reflectance.
- Determine surface reflectance at $0.47_{\mu\text{m}}$ and $0.66 \mu\text{m}$:

$$\rho_{0.47}^s = 0.5 \rho_{0.66}^s = 0.25 \rho_{2.2}^*$$
- Determine the aerosol type (e.g. smoke or dust) from the excess radiance:

$$\Delta \rho_{0.47}^* = \rho_{0.47}^* - \rho_{0.47}^s \quad \Delta \rho_{0.66}^* = \rho_{0.66}^* - \rho_{0.66}^s$$

$$\Delta \rho_{0.47}^* / \Delta \rho_{0.66}^* \implies \text{aerosol type}$$
- Determine the optical thickness from the excess radiance:

$$\Delta \rho_{0.47}^* \implies \tau_{0.47}, \quad \Delta \rho_{0.66}^* \implies \tau_{0.66}$$



The modified three-parameter linear surface BRF kernel model:

$$r_s(\theta_s, \theta_v, \phi) = f_{iso} + f_{vol} \cdot K_{vol}(\theta_s, \theta_v, \phi) + f_{geo} \cdot K_{geo}(\theta_s, \theta_v, \phi)$$

Where: the volumetric and geometrical kernel function has the following form:

$$K_{vol} = K_{RT} = \frac{(\pi/2 - \xi) \cos \xi + \sin \xi}{\cos \theta + \cos \vartheta} \left(1 + \frac{1}{1 + \frac{\xi}{\xi_0}} \right) - \frac{\pi}{4}$$

$$K_{geo} = K_{LSR} = O(\theta_s, \theta_v, \phi) - \sec \theta' - \sec \vartheta' + 0.5(1 + \cos \xi') \sec \theta' \sec \vartheta'$$

$$O = (t - \sin t \cos t)(\sec \theta' + \sec \vartheta') / \pi$$

$$\cos t = \frac{h \sqrt{D^2 + (\tan \theta' \tan \nu' \sin \phi)^2}}{b(\sec \theta' + \sec \vartheta')}$$

$$D = \sqrt{\tan^2 \theta' + \tan^2 \nu' - 2 \tan \theta' \tan \nu' \cos \phi}$$

$$\cos \xi' = \cos \theta' \cos \nu' + \sin \theta' \sin \nu' \cos \phi$$

$$\theta' = \tan^{-1} \left(\frac{b}{r} \tan \theta_s \right)$$

$$\vartheta' = \tan^{-1} \left(\frac{b}{r} \tan \theta_v \right)$$

$$\xi_0 = 1.5^\circ, \frac{h}{b} = 2.0 \text{ and } \frac{b}{r} = 1.0$$

Liang, S., (2001), Narrowband to Broadband Conversion of Land Surface Albedo. I. Algorithms, *Remote Sensing of Environment*, 76:213-238.

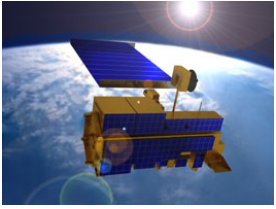
$$\alpha_{\text{short}} = 0.160\alpha_1 + 0.291\alpha_2 + 0.243\alpha_3 + 0.116\alpha_4 \\ + 0.112\alpha_5 + 0.081\alpha_7 - 0.0015$$

$$\alpha_{\text{visible}} = 0.331\alpha_1 + 0.424\alpha_3 + 0.246\alpha_4$$

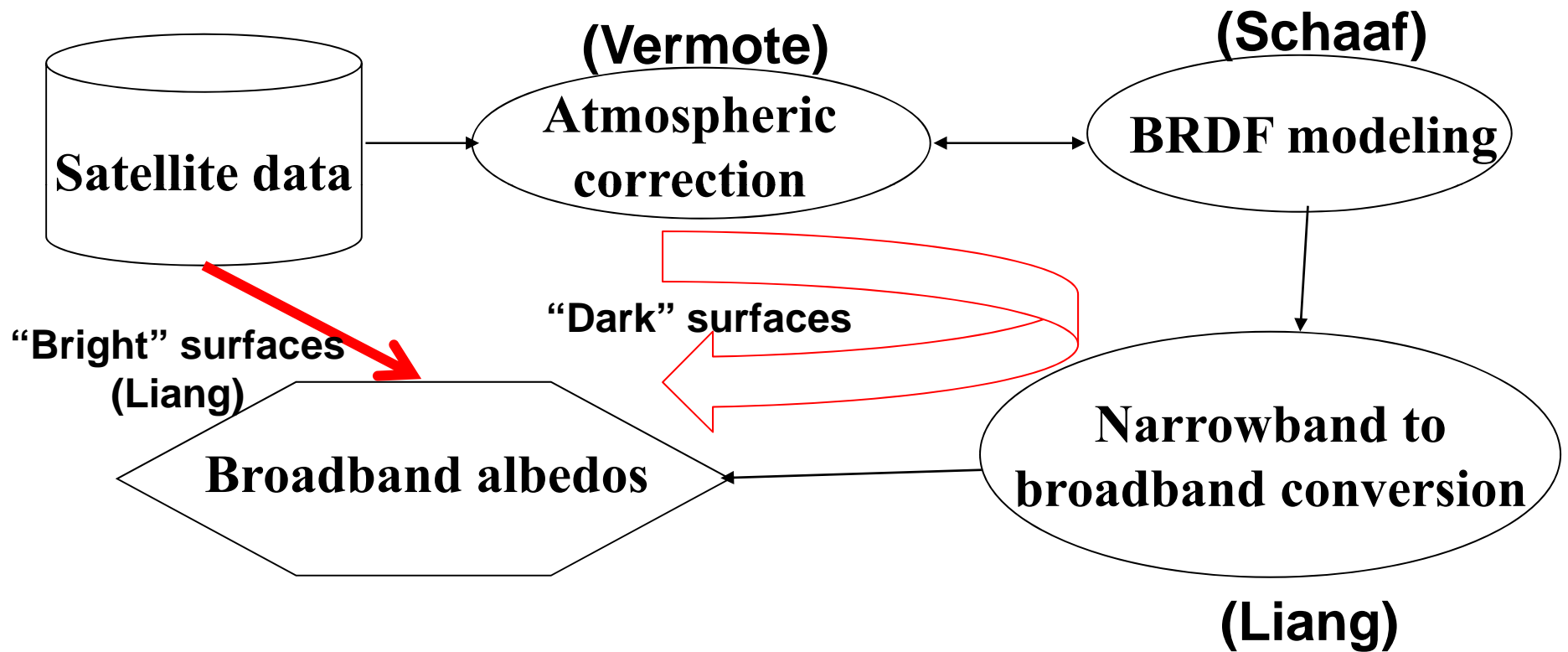
$$\alpha_{\text{diffuse-visible}} = 0.246\alpha_1 + 0.528\alpha_3 + 0.226\alpha_4 - 0.0013$$

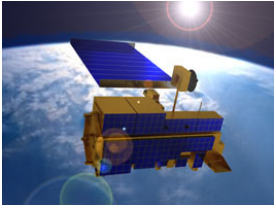
$$\alpha_{\text{direct-visible}} = 0.369\alpha_1 + 0.374\alpha_3 + 0.257\alpha_4$$

$$\alpha_{\text{NIR}} = 0.039\alpha_1 + 0.504\alpha_2 - 0.071\alpha_3 + 0.105\alpha_4 \\ + 0.252\alpha_5 + 0.069\alpha_6 + 0.101\alpha_7$$

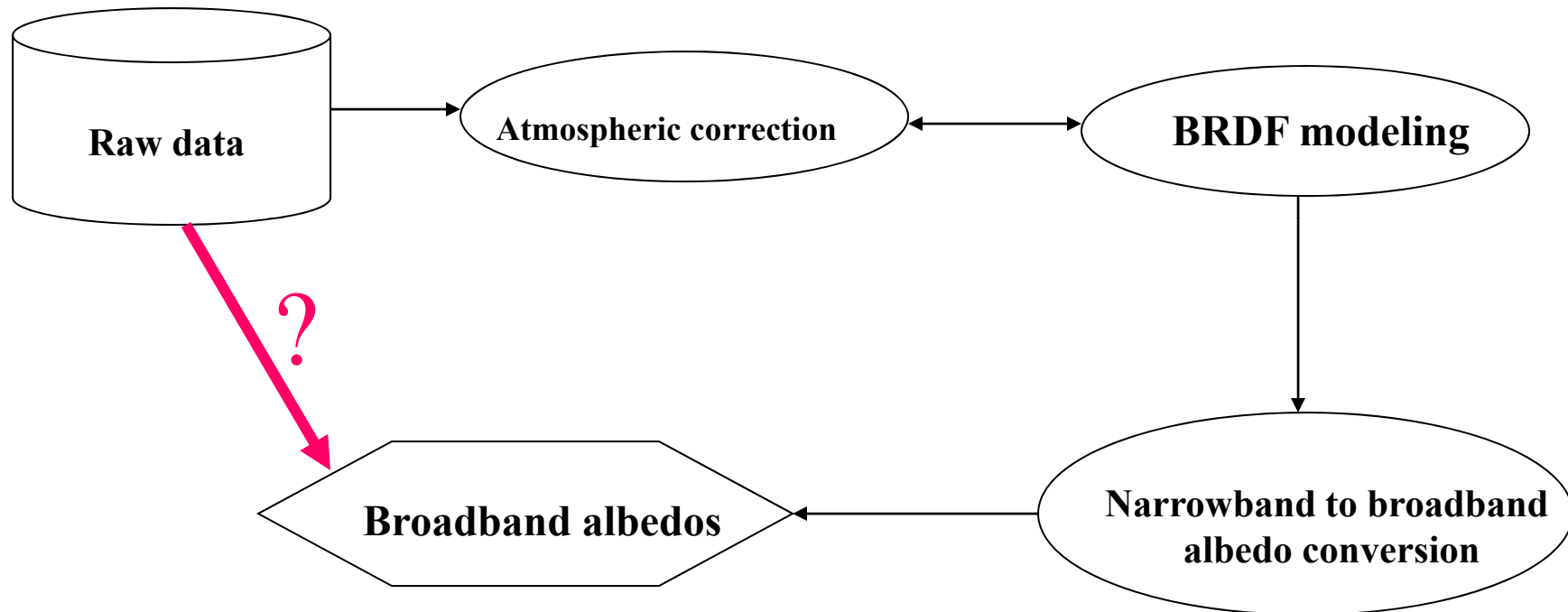


JPSS (VIIRS) two albedo algorithms



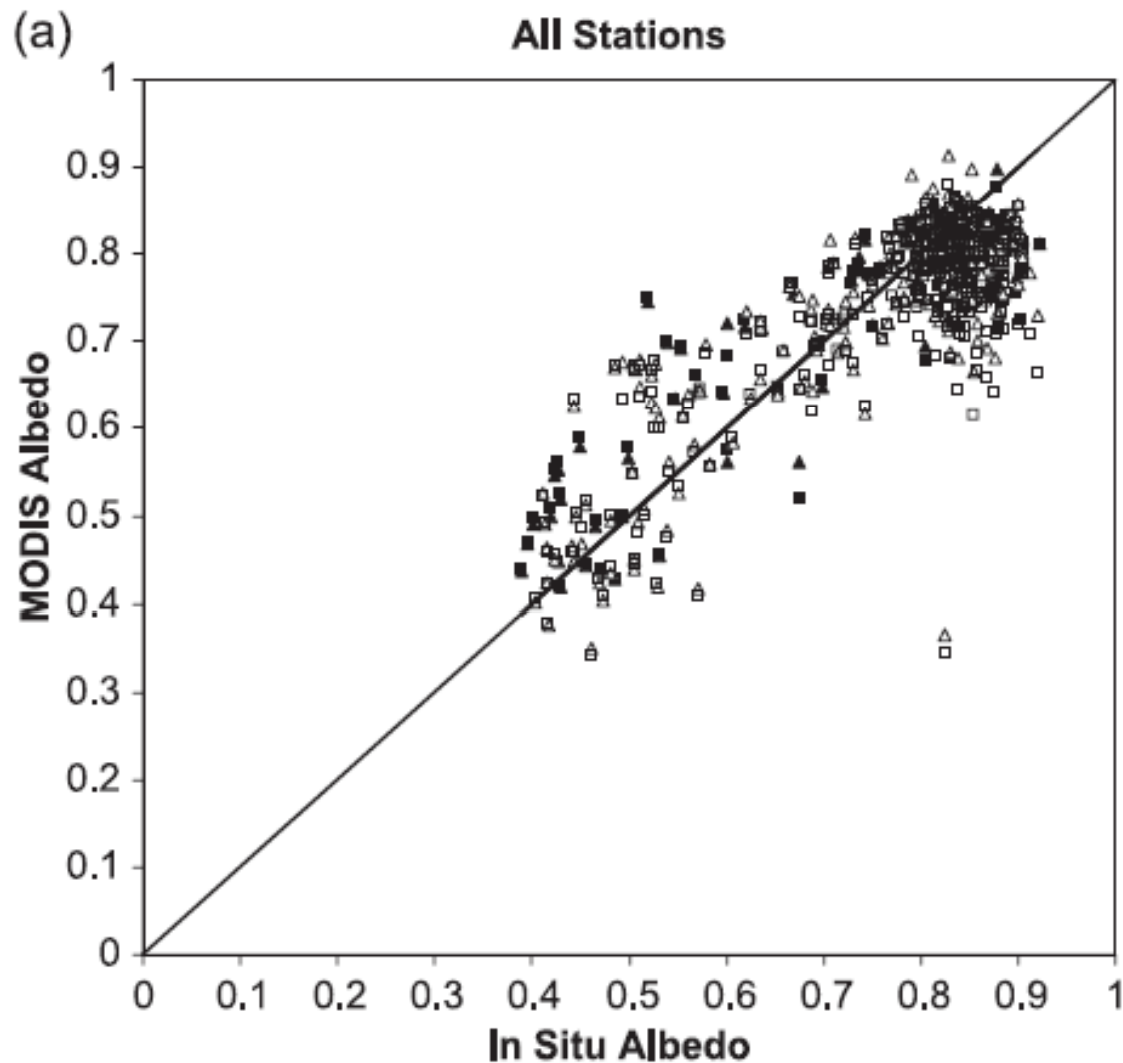


Broadband albedo estimation Joint Polar Satellite System (JPSS)



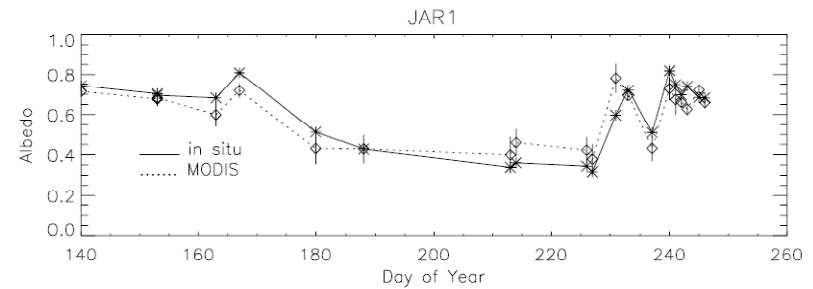
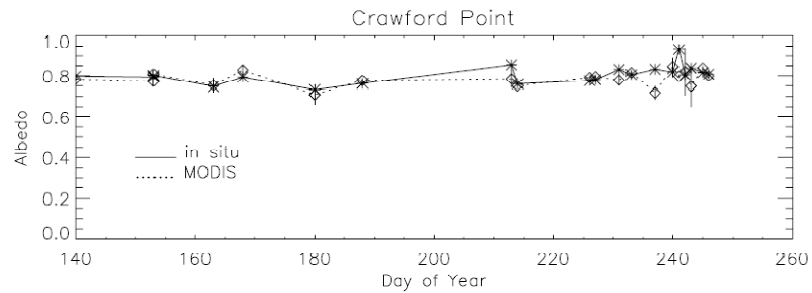
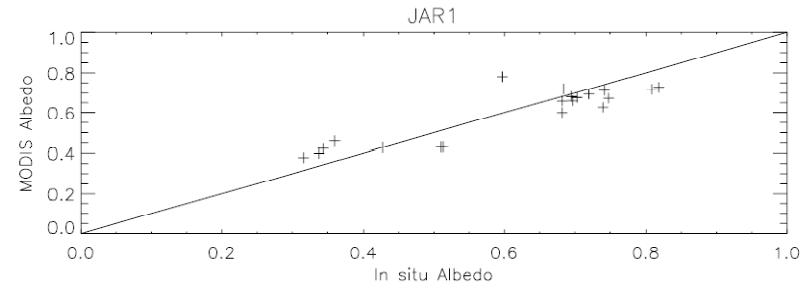
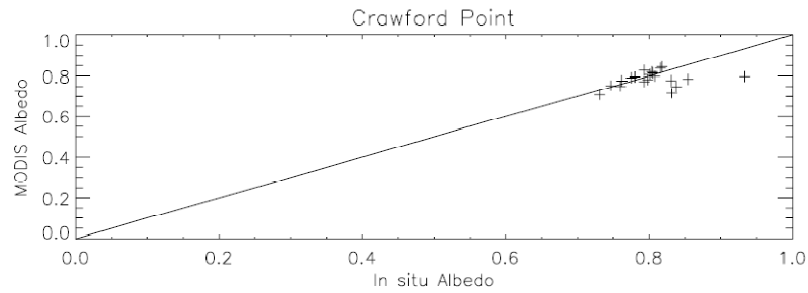
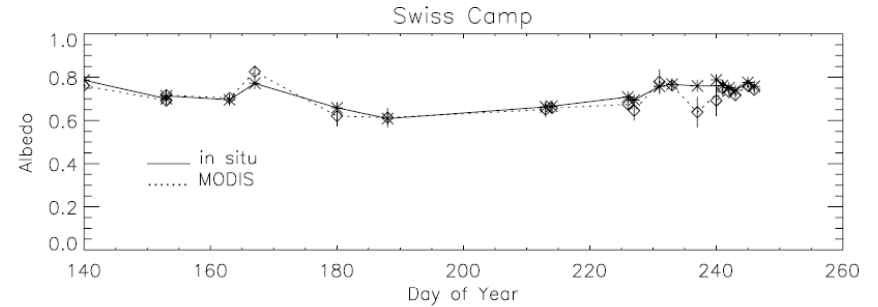
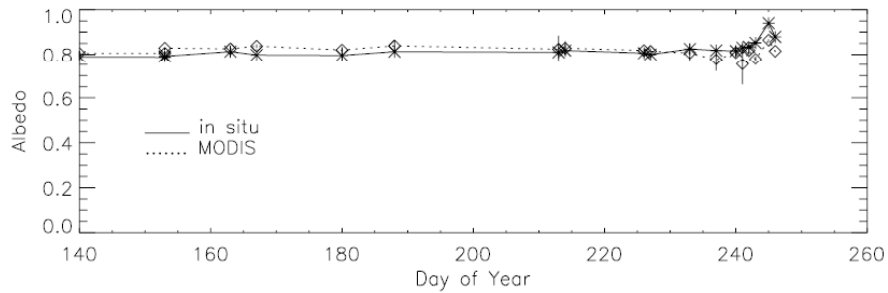
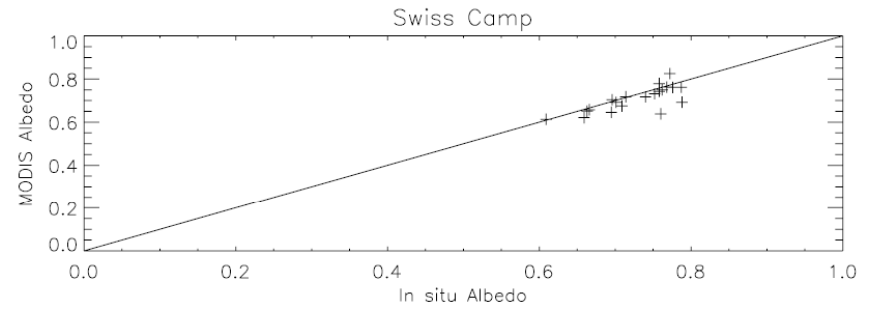
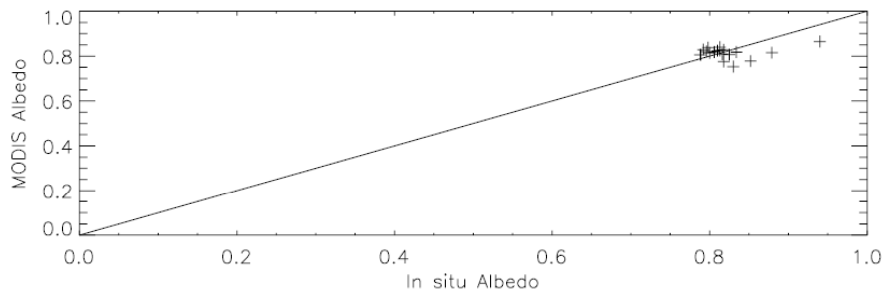
Liang, S., (2003), A direct algorithm for estimating land surface broadband albedos from MODIS imagery, *IEEE Trans. Geosci. Remote Sen.*, 41(1):136-145;

Liang, S., J. Stroeve and J. Box, (2005), Mapping daily snow shortwave broadband albedo from MODIS: The improved direct estimation algorithm and validation, *Journal of Geophysical Research.* 110 (D10): Art. No. D10109.



Validation of MODIS
shortwave albedo product
over Greenland

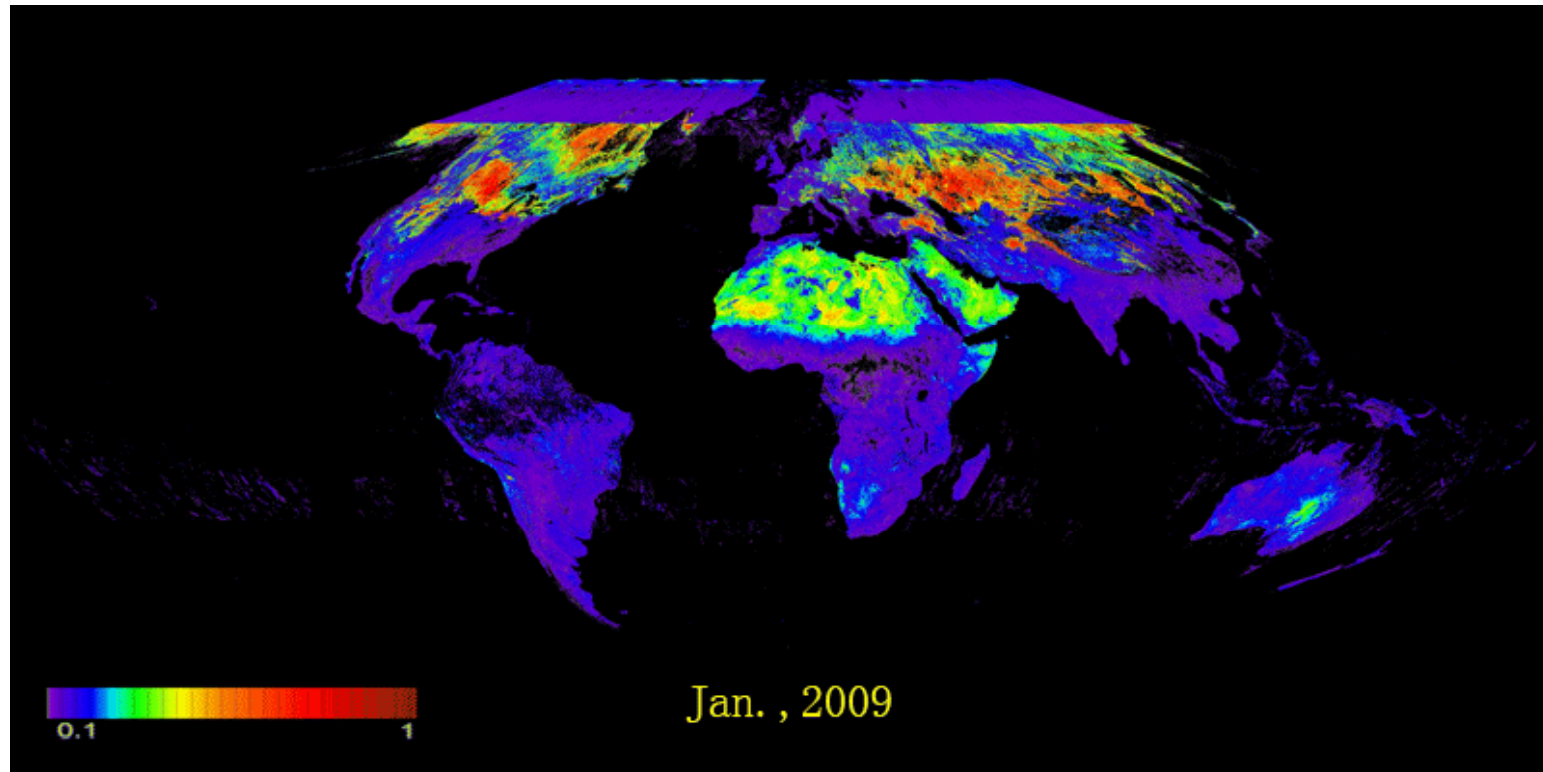
Stroeve, J., J. Box, F. Gao, S., Liang, A., Nolin, and C. Schaaf, (2005), *Accuracy assessment of the MODIS 16-day snow albedo product: Comparisons with Greenland in situ measurements*, *Remote Sensing of Environment*, 94(1):46-60.



Liang, S., J. Stroeve and J. Box, (2005), *Mapping daily snow shortwave broadband albedo from MODIS: The improved direct estimation algorithm and validation*, *Journal of Geophysical Research*. 110 (D10): Art. No. D10109.



Global Broadband Albedo



The GLASS project retrieves the daily 1km global albedo from MODIS surface reflectance products (1984-2010, 5km from 1984-1999; 1km from 2000-2010)



Developing GOSE-R algorithm as a NOAA
GOES-R Land Science Team member

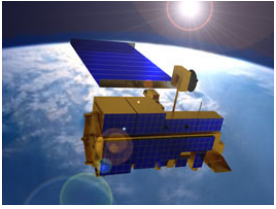
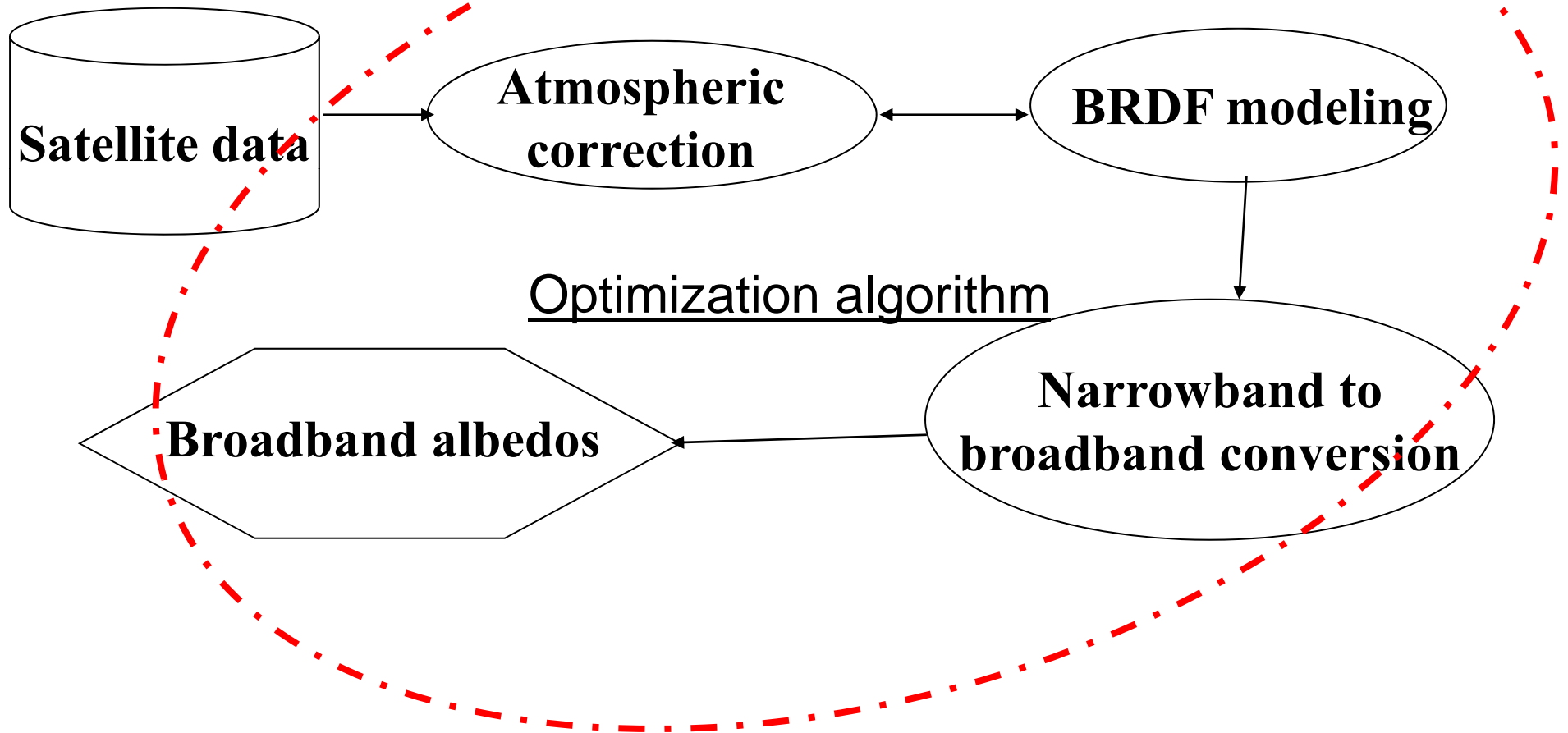


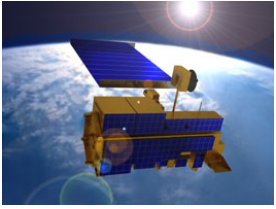
Illustration of the proposed GOES-R ABI albedo algorithms



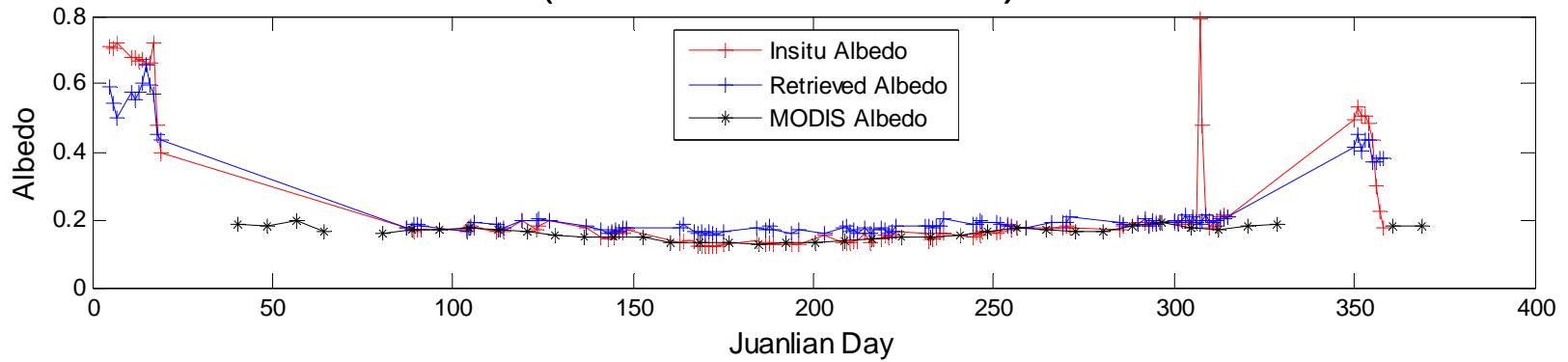
Minimizing the cost function:

$$J(\mathbf{x}) = \sum_{i=1}^{NO} \left(A_i^{Clm} - A_i^{Est}(\mathbf{x}) \right)^2 + \sum_{i=1}^{NO} \sum_{j=1}^{NB} \left(R_{i,j}^{Obs} - R_{i,j}^{Est}(\mathbf{x}) \right)^2$$
$$+ J_c = J_b + J_o + J_c$$

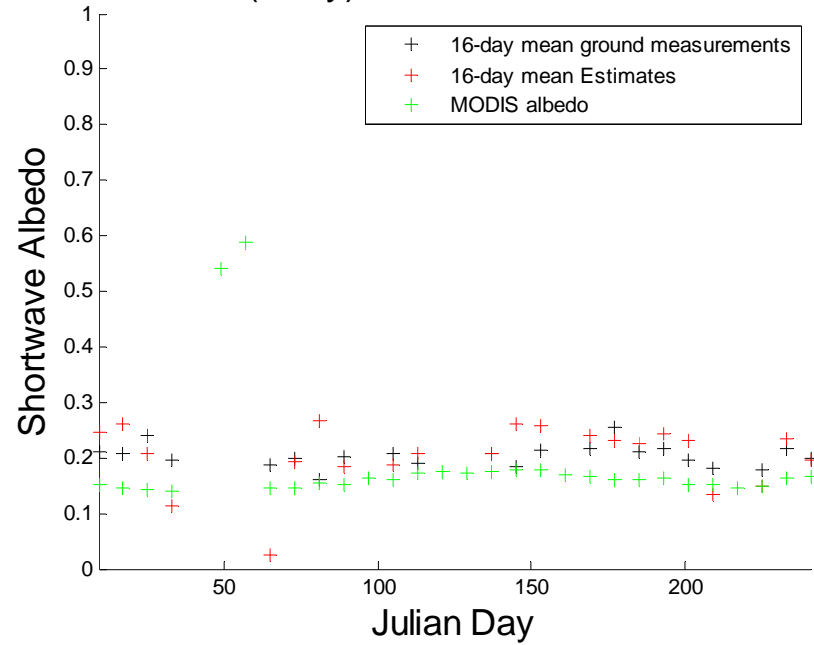
Where \mathbf{x} are the coefficients of the surface BRDF model and aerosol optical depth (AOD), $A_i^{Est}(\mathbf{x})$ is the calculated surface albedo using the BRDF model, A_i^{Clm} are the background values of albedo from albedo climatology, $R_{i,j}^{Obs}$ and $R_{i,j}^{Est}$ refer to the observed and modeled top-of-atmosphere (TOA) reflectance. J_c is the cost function to account for various constrains (physical meanings of BRDF parameters and AODs, etc.)



Fort Peck (48.31N,105.10W) Valid Obs=4



DE-Geb(Daily) Lat:51.1001 Lon:10.9143



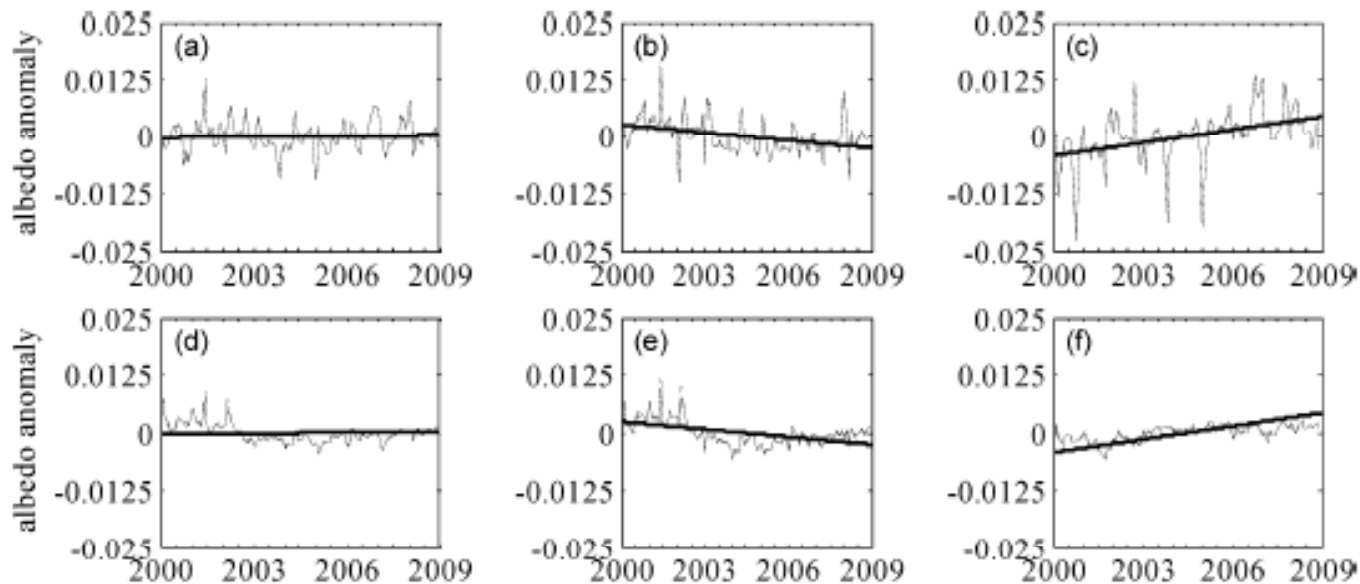
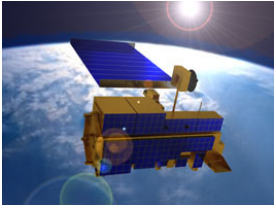
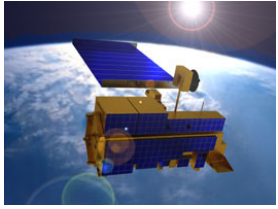


Fig. 2. Global, Northern Hemisphere, Southern Hemisphere land surface shortwave albedo anomalies under snow-covered and snow-free conditions. (a) Global albedo anomaly under snow-covered condition. (b) NH albedo anomaly under snow-covered condition. (c) SH albedo anomaly under snow-covered condition. (d) Global albedo anomaly under snow-free condition. (e) NH albedo anomaly under snow-free condition. (f) SH albedo anomaly under snow-free condition.

Zhang, X. S. Liang, K. Wang, L. Li, and S. Gui, (2010), A climatological analysis of global land surface shortwave broadband albedo from MODIS, *IEEE Journal of Special Topics in Applied Earth Observations and Remote Sensing*, 3:296-305

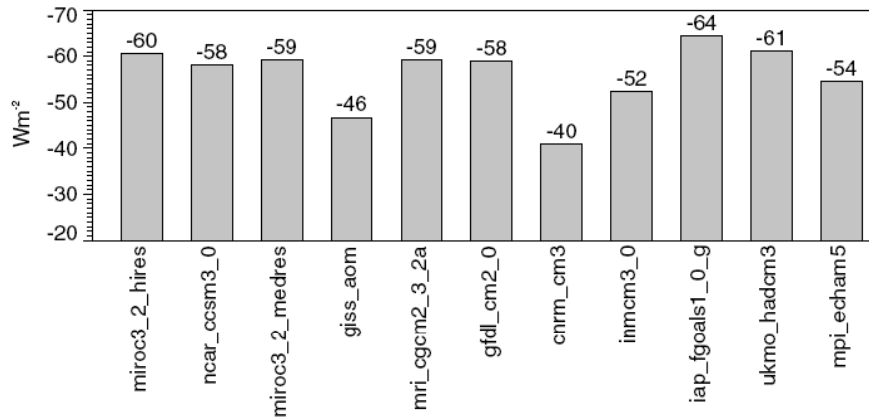


Outlines

- ♣ Background
- ♣ Insolation/PAR
- ♣ Albedo
- ♣ Longwave net radiation
- ♣ Evapotranspiration
- ♣ Summary



IPCC AR4 : SURFACE NET LW ALL SKY



Radiation budgets by IPCC

IPCC AR4 : LONGWAVE DOWNWARD RADIATION ALL SKY

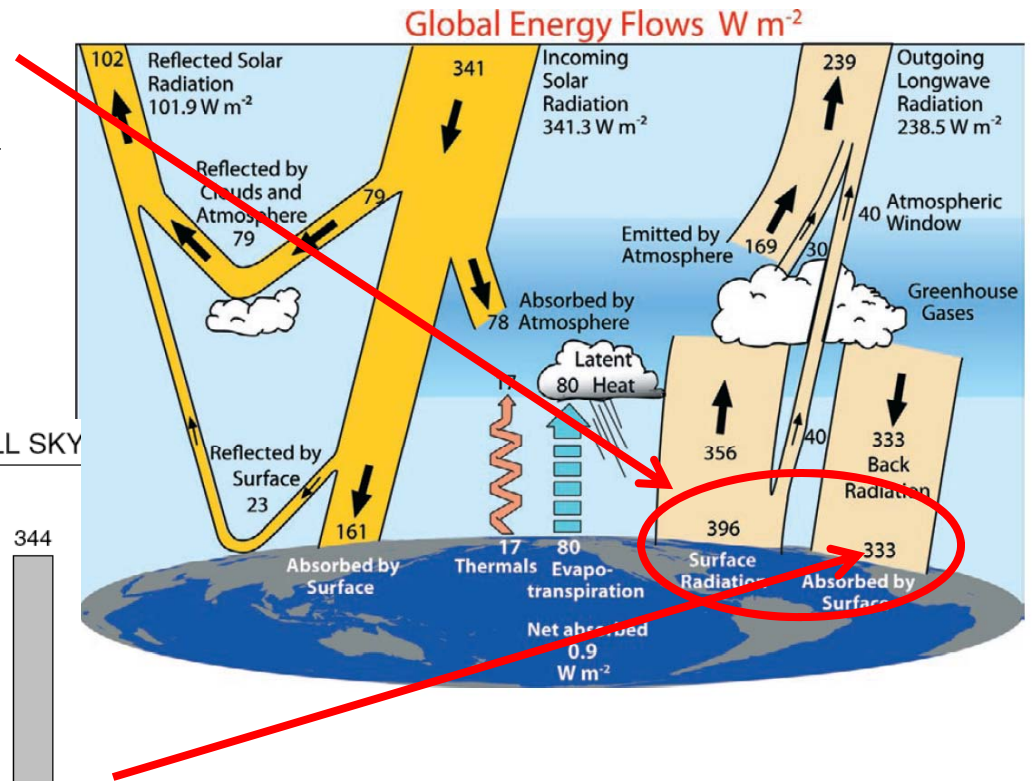
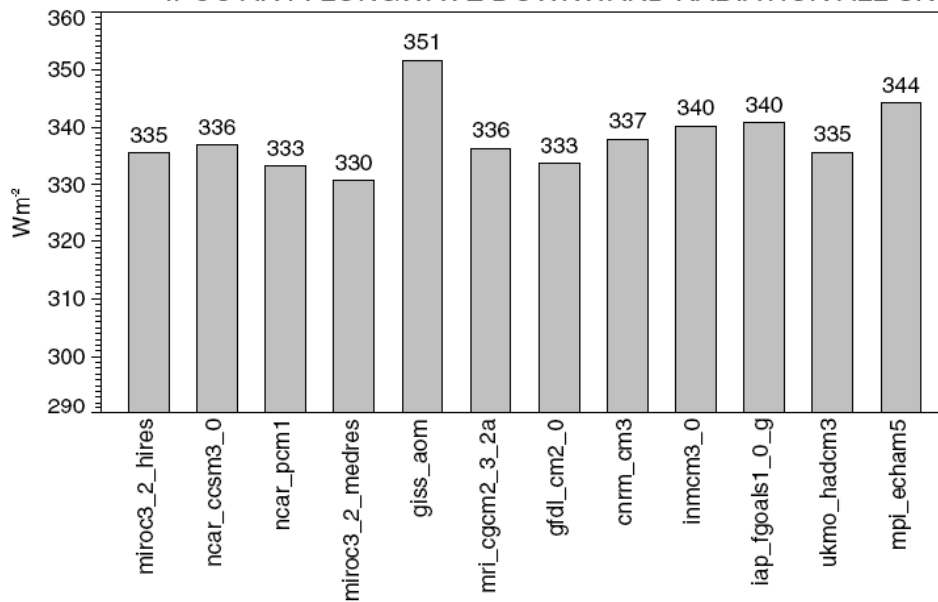
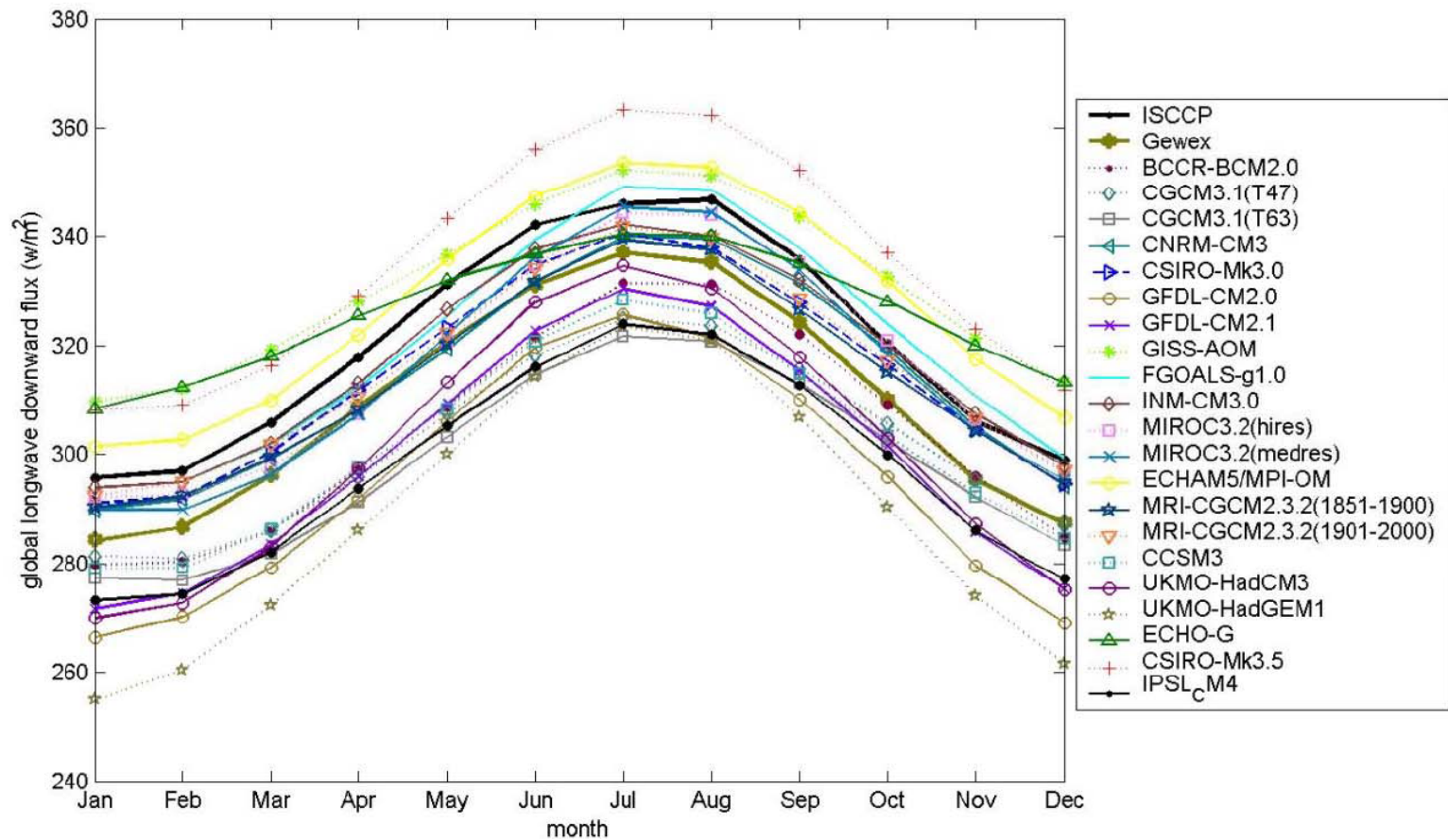




Fig.1 Monthly averages of longwave downward radiation from two satellite products (ISCCP and GEWEX) and different GCMs in the IPCC AR4.





Estimation of downward longwave radiation

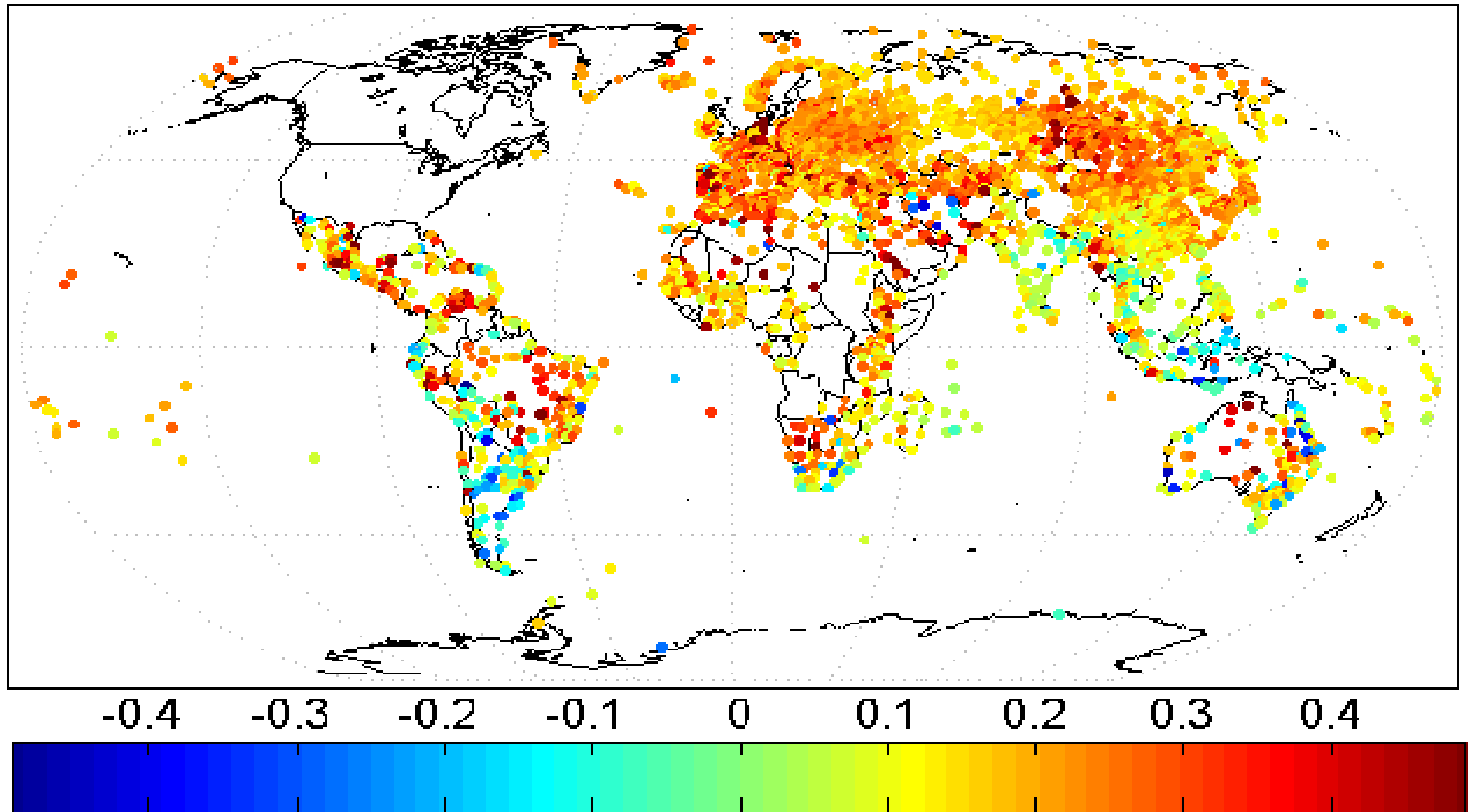
- 1). Empirical methods
- 2). Calculating downward flux using atmospheric profiles
- 3). Calculating downward flux from TOA radiance directly

Wang, K., and **S. Liang**, (2009), Global atmospheric downward longwave radiation under all-sky conditions from 1973 to 2008, *Journal of Geophysical Research*, 114, D19101, doi:19110.11029/12009JD011800

Wang, W. & **S. Liang**, (2009), Estimating High-Spatial Resolution Clear-Sky Land Surface Downwelling and Net Longwave Radiation from MODIS Data, *Remote Sensing of Environment*, 113:745-754

Wang, W., & **S. Liang**, (2010). A Method for Estimating Clear-sky Instantaneous Land Surface Longwave Radiation from GOES Sounder and GOES-R ABI Data. *IEEE Geoscience and Remote Sensing Letters*, 7, 708-712

Trend in Downward Longwave Radiation ($W m^{-2} ya^{-1}$)



Linear trend of daily (L_d) over 3200 global weather stations where data are available for at least 300 months (25 years) during the period of 1973-2008.

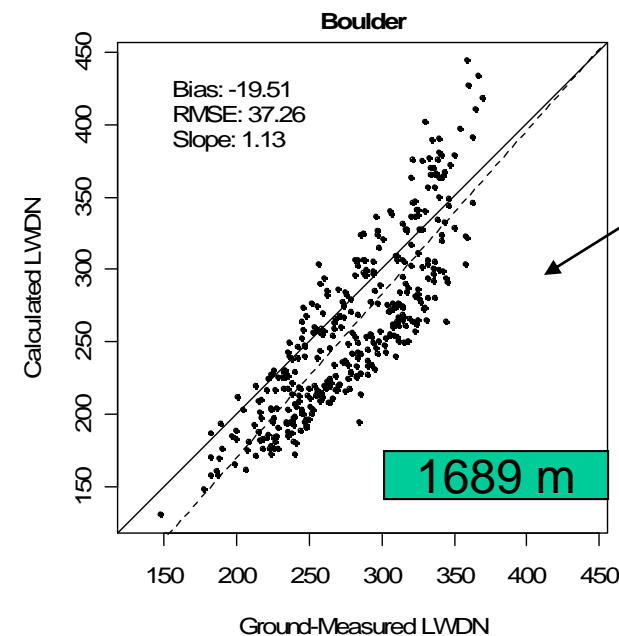
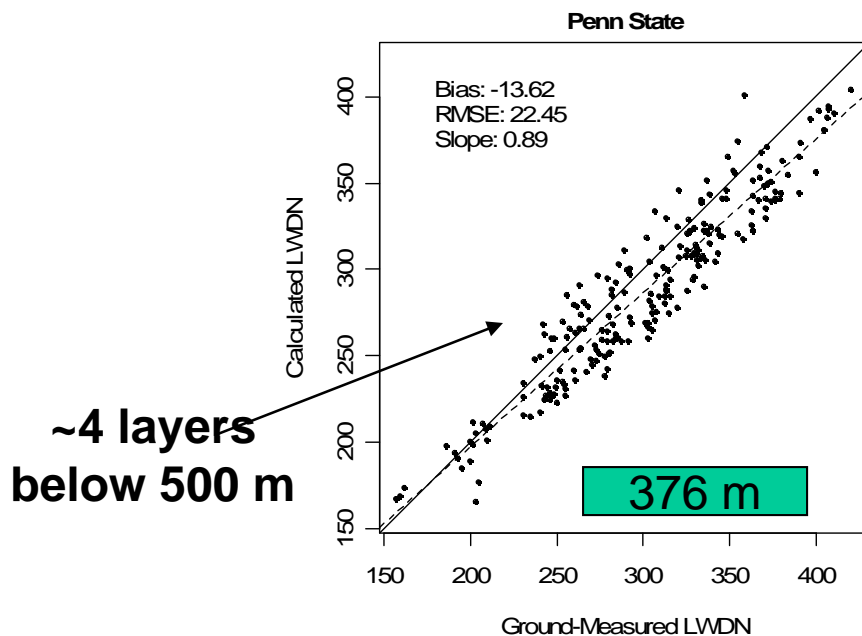
Wang, K., and S. Liang, (2009), Global atmospheric downward longwave radiation under all-sky conditions from 1973 to 2008. *Journal of Geophysical Research*. 114. D19101. doi:10.1029/2009JD011800

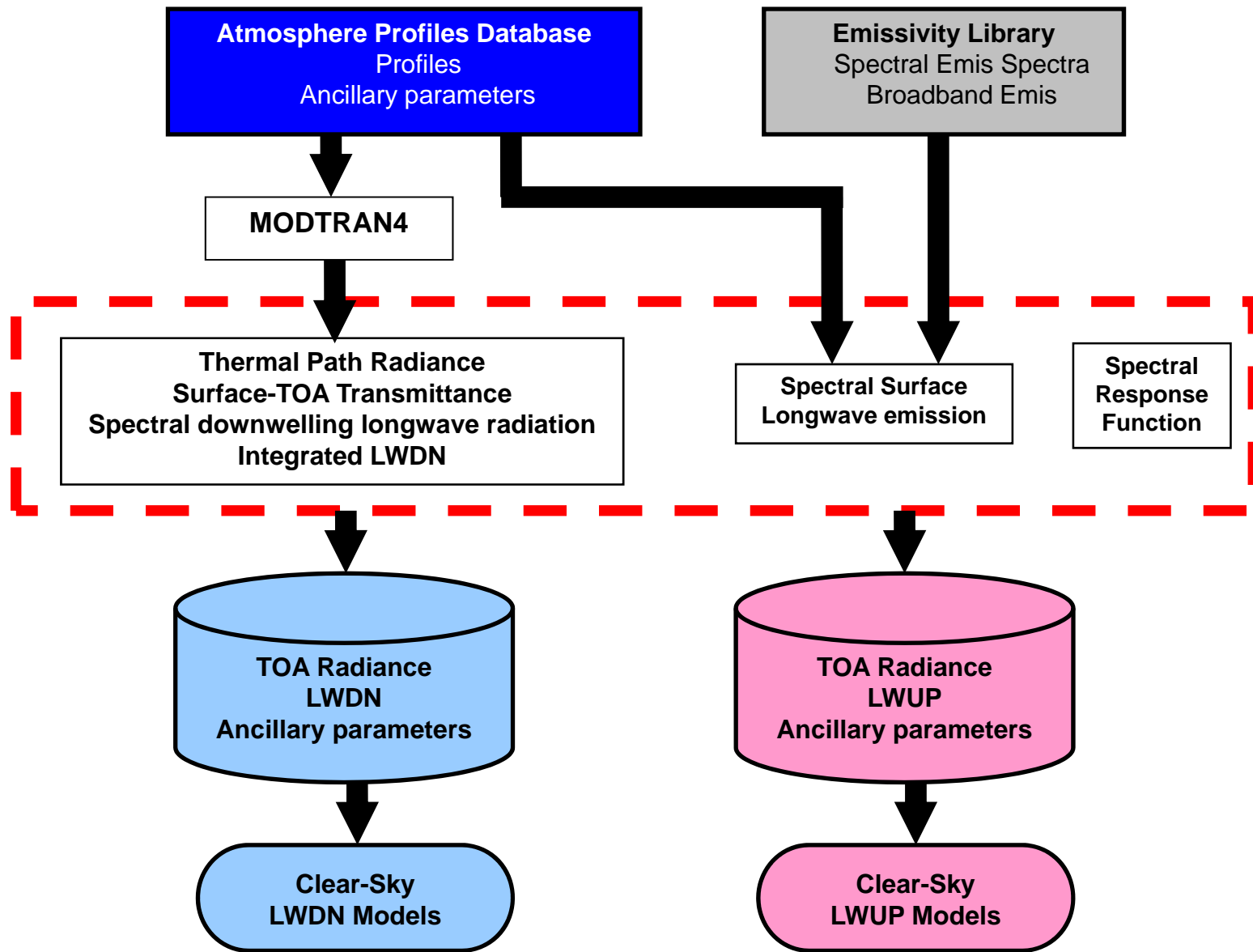


♣ Using MODIS Profiles & MODTRAN4

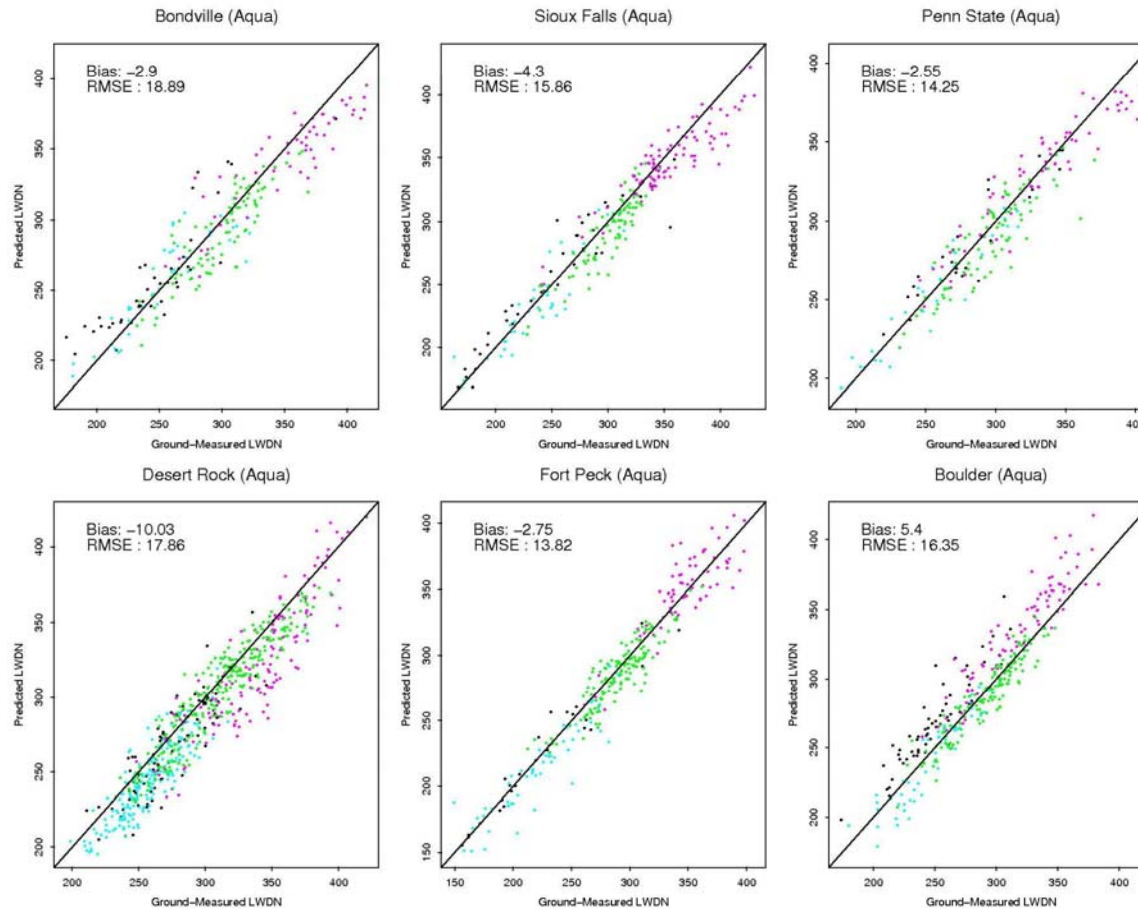
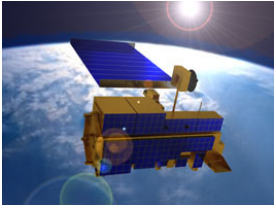
♣ Problems:

- LWDN dominated by near surface temp. & moisture
- MODIS profiles are coarse (20 levels)
 - 1000, 950, 920, 850, 800, 700, 620, 500, 400, 300, 250, 200, 150, 100, 70, 50, 30, 20, 10, 5 hPa
- Large errors, especially over high elevation sites





Framework of Hybrid Methods



Nonlinear Models

- ♣ Results similar to Terra
- ♣ Smaller RMSEs in Aqua-derived LWDN
 - Smaller systematic errors in Aqua (*Liu et al, 2006*)
 - *Diff. overpass times*
 - *diff. atmospheric conditions*

Avg. RMSE: 17.60 W/m²
Avg. Bias: -0.40 W/m²

day/fallwinter
 night/fallwinter
 day/springsummer
 night/springsummer



Estimating longwave upwelling radiation (LWUP)

(1) Temperature-Emissivity Method

$$F_u = \varepsilon \int_{\lambda_1}^{\lambda_2} \pi B(T_s) d\lambda + (1 - \varepsilon) F_d$$

T_s MODIS LST (MOD11_L2)

ε Broadband emis (derived from MOD11B1)

(2) Hybrid Method

♣ Following the framework for hybrid methods

→ Emissivity Effect

- UCSB Emissivity Library (59 spectra)
- ~2000 MODIS Profile

♣ Statistical Analysis

→ Linear SULR Models ($R^2: 0.990$, $RMSE < 5.42 \text{ W/m}^2$)

$$F_u = a_0 + a_1 L_{29} + a_2 L_{31} + a_3 L_{32}$$

→ Artificial Neural Network (ANN) Models

($R^2: 0.996$ $RMSEs < 3.7 \text{ W/m}^2$)

5 Models in total

θ	Model
0°	
15°	
30°	
45°	
60°	

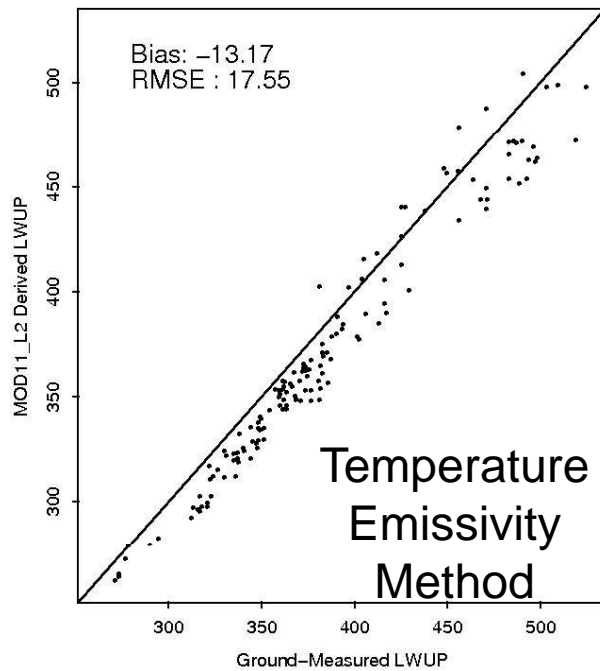
Representative thermal space-borne RS systems

sensors	Wavebands (μm)	# of bands	Spatial resolution
AVHRR	3.55 – 3.93 10.30 – 12.50	1 2	1.1km
MODIS	3.66 – 4.08 8.400 – 13.48	3 4	1km
ATSR/ATSR-2/AASTR	3.55 – 3.93 10.4 – 12.5	1 2	1km
ASTER	8.125 – 11.65	5	90m
TM/ETM+	10.00 – 12.90	1	120m/60m
MSG-SEVIRI	8.30–13.0	3	3km at nadir
FY-3	3.7 – 4 9.59 – 13.49	2 4	1.1km
GOES	10.2 – 12.5	2	4km at nadir
GMS(Geostationary Meteorological Satellite)	3.5 – 4.0 10.3 – 12.5	1 1	4km at nadir
HJ-1B	3.50 – 3.90 10.5 – 12.5	1 1	150m 300m
METOP-IASI	3.2–15.5 (645–2760 cm^{-1})	8461	12km

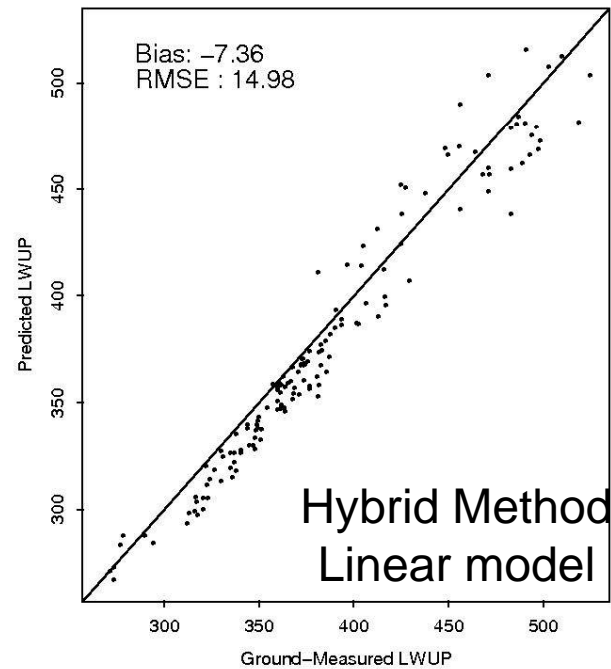


Estimating LWUP: validation

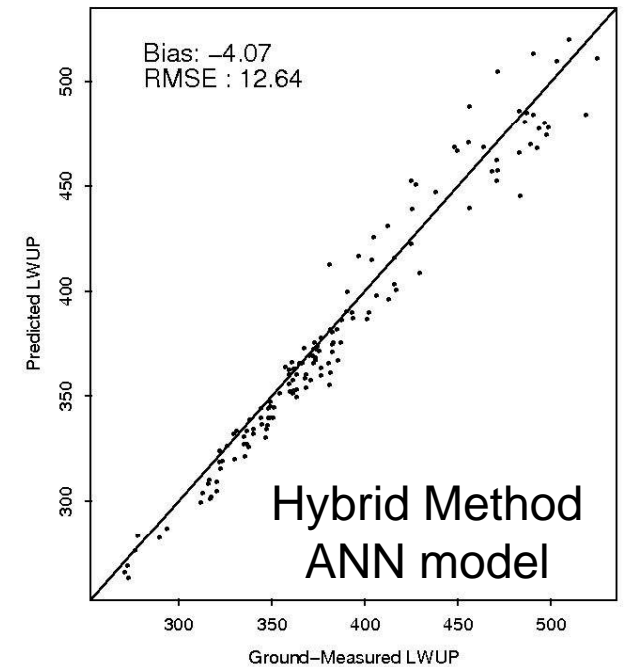
Bondville, IL (Aqua)



Bondville, IL (Aqua)



Bondville, IL (Aqua)



Bondville, IL (cropland, elevation 213 m)

- Smaller RMSEs in Aqua
- Hybrid method outperforms temperature-emissivity method
- ANN model outperforms linear model

Wang, W., S. Liang & J. A. Augustine, (2009), Estimating Clear-Sky Land Surface Upwelling Longwave Radiation from MODIS Data. *IEEE Trans. Geosci. and Remote Sens.* 47(5):1555-1575

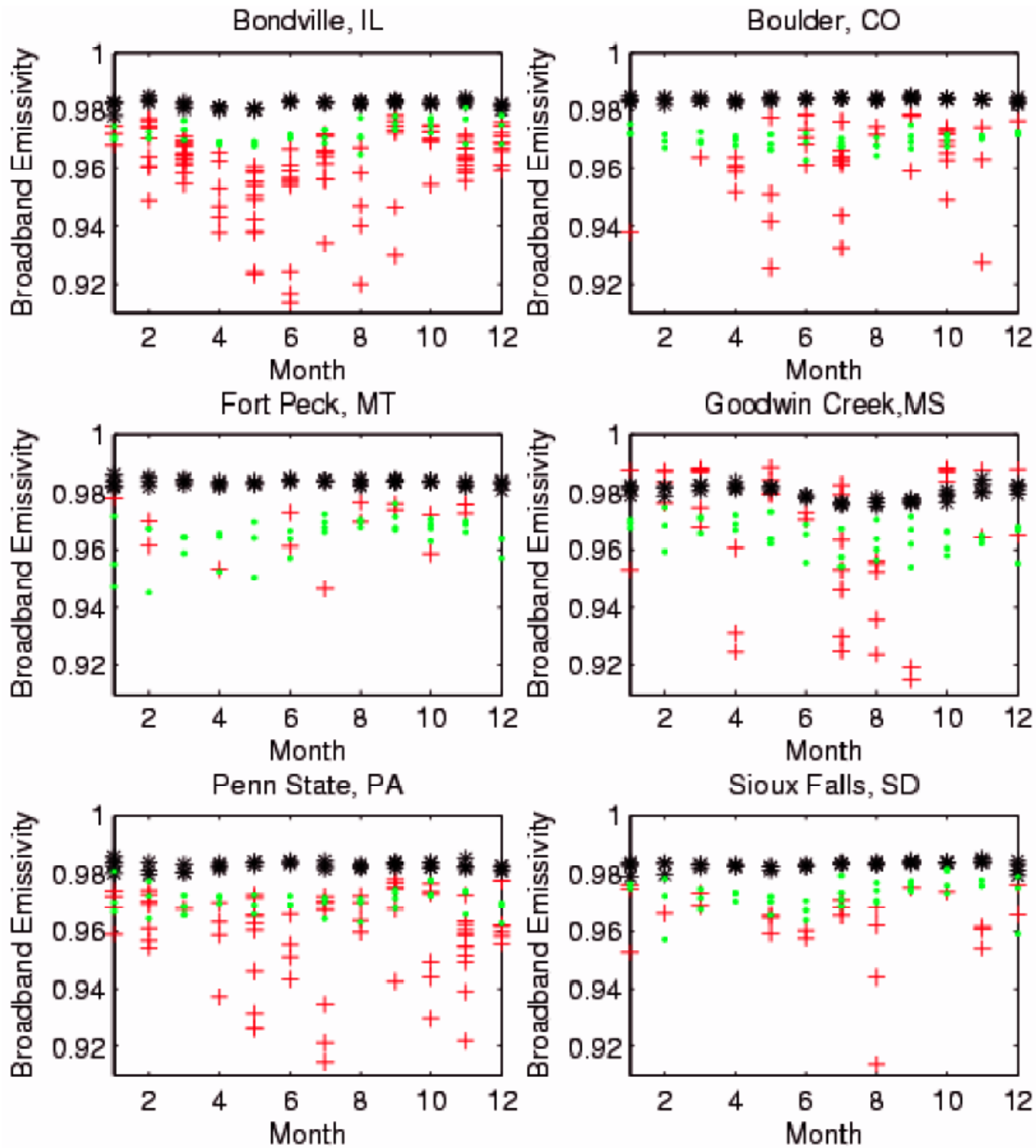


MODIS LST validation

Summary of validation results

Site	MOD11_L2 (°C)		MOD07_L2 (°C)	
	Bias (MOD-GT)	RMSE	Bias (MOD-GT)	RMSE
Brookings	0.62	1.63	1.30	1.97
Audubon	0.72	1.31	2.98	3.74
Canaan Valley	0.04	1.42	1.20	2.08
Black Hills	0.15	1.48	3.14	4.10
Fort Peck	-2.19	2.51	0.34	2.70
Hainich	-2.21	2.51	-2.12	2.58
Tharandt	-3.23	3.44	-3.38	3.73
Bondville	-3.09	3.41	-0.16	2.50

Wang, W., **S. Liang**, and T. Meyer, (2008), Validating MODIS land surface temperature products, *Remote Sensing of Environment*, 112:623-635

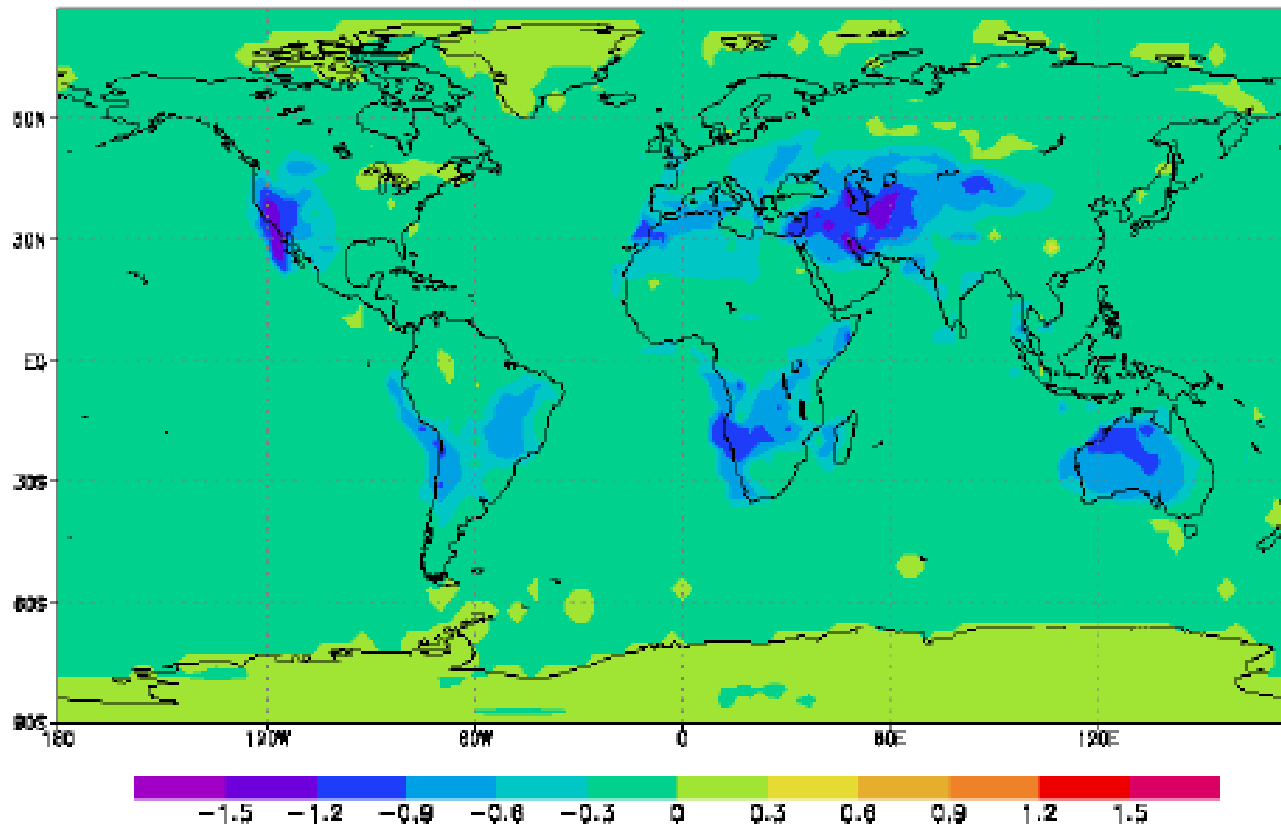


Broadband emissivity calculated from MODIS Collection 4 (green dot) and Collection 5 (black star) monthly emissivity products and ASTER daily emissivity products (red plus sign) at a resolution of 0.05° at six SURFRAD sites.

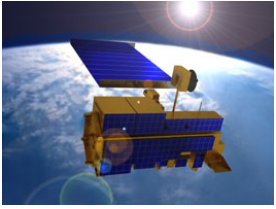
Wang, K., and S. Liang, (2009), Evaluation of ASTER and MODIS land surface temperature and emissivity products using surface longwave radiation observations at SURFRAD sites, *Remote Sensing of Environment*, 113:745-754

The emissivity impact on ground temperature in the coupled CAM2/CLM2 model, control run minus sensitivity run. Unit is K.

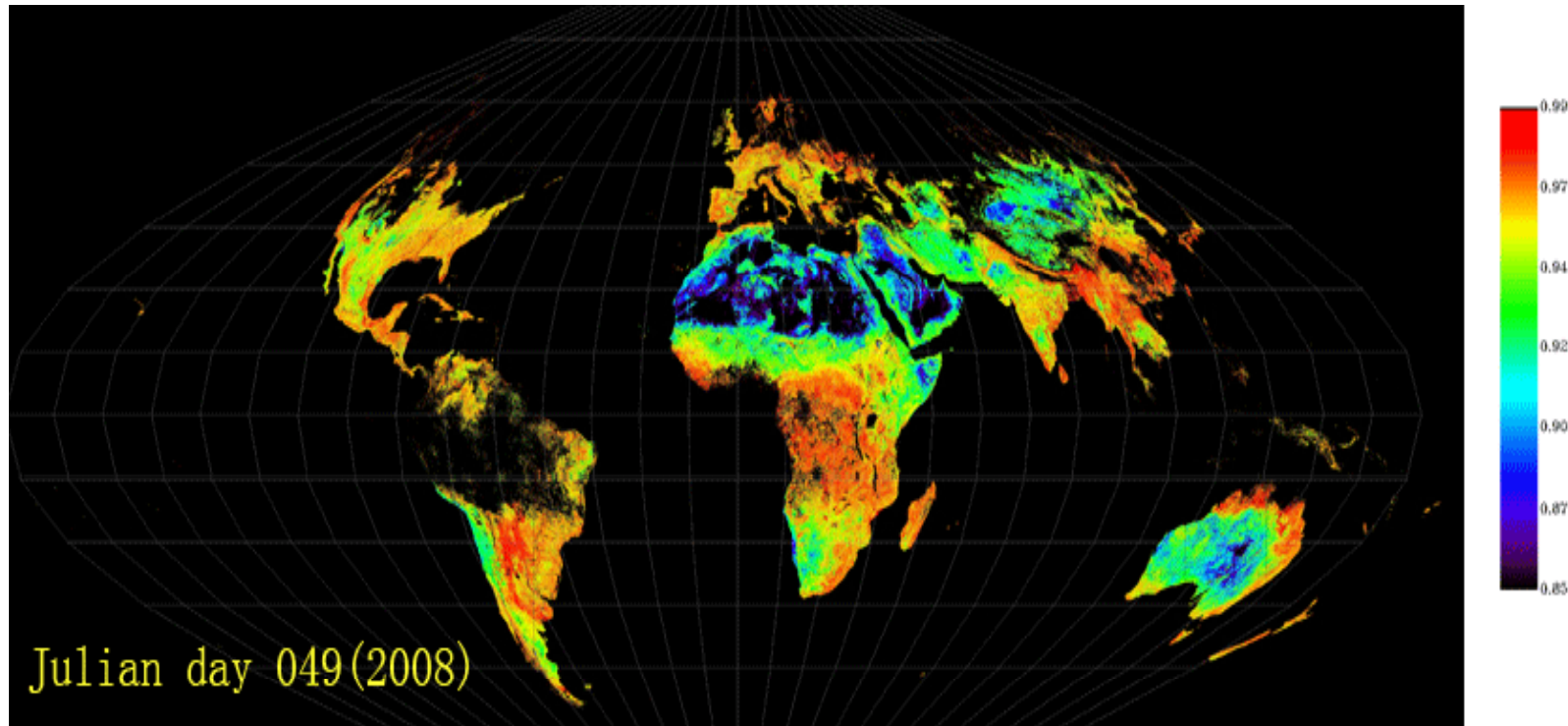
CAM2/CLM2, Daily averaged $T_g - T_g @em=0.86$, Sep.



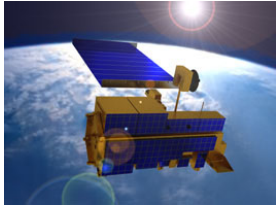
Jin, M., and S. Liang, (2006), Impacts of the MODIS broadband emissivity on GCM simulation, *J. Climate*, 19:2867-2881.



Chinese GLASS Broadband Emissivity Product



**Global eight-day 1-Km land surface broadband emissivity retrieved by
our new algorithm using MODIS albedos.
(5km from 1984-1999, 1km 2000-2010, 8day)**

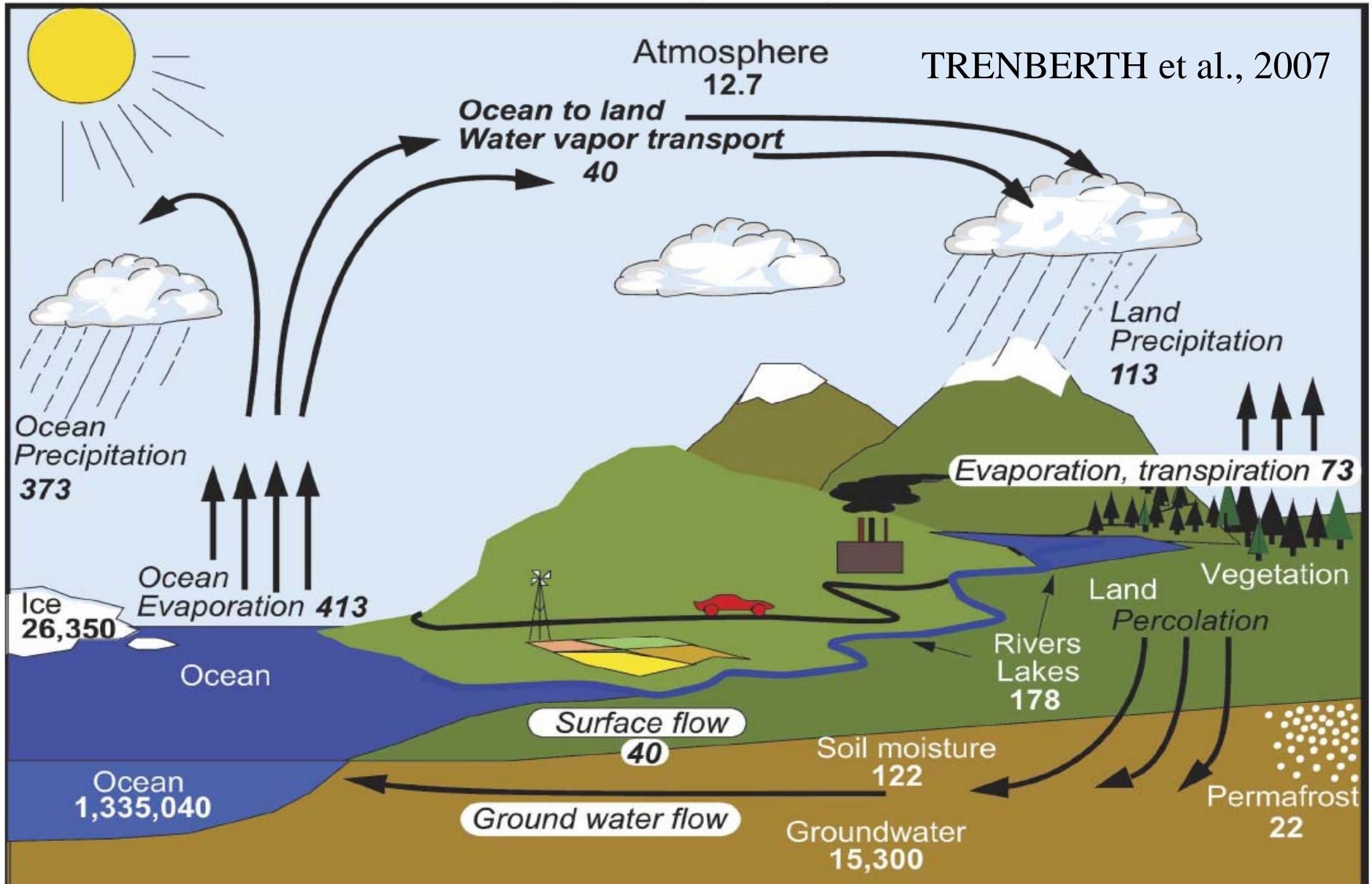


Outlines

- ♣ Background
- ♣ Insolation/PAR
- ♣ Albedo
- ♣ Longwave net radiation
- ♣ Evapotranspiration
- ♣ Summary

Hydrological Cycle

TRENBERTH et al., 2007

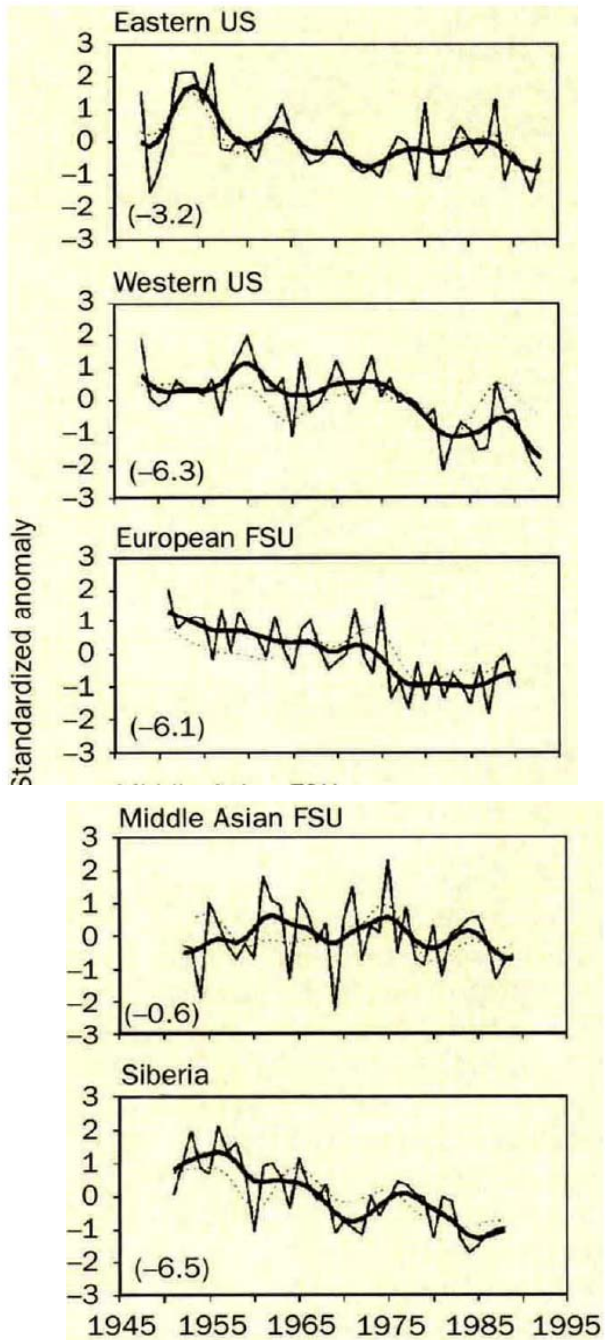


Units: Thousand cubic km for storage, and *thousand cubic km/yr* for exchanges

FIG. 1. The hydrological cycle. Estimates of the main water reservoirs, given in plain font in 10^3 km^3 , and the flow of moisture through the system, given in slant font ($10^3 \text{ km}^3 \text{ yr}^{-1}$), equivalent to Eg (10^{18} g) yr^{-1} .



- ♣ Qin, J., **S. Liang**, R. Liu, H. Zhang, and B. Hu, (2007), A Weak-Constraint Based Data Assimilation Scheme for Estimating Surface Turbulent Fluxes, *IEEE Geoscience and Remote Sensing Letters*, 4(4):649-653
- ♣ Wang, K. and **Liang, S.** (2008), An improved method to estimate evapotranspiration from a combination of net radiation, vegetation index and temperature: Influence of soil moisture. *Journal of Hydrometeorology* 9, 712-727
- ♣ Yiao, Y. **S. Liang**, Q. Qin, and K. Wang, (2010), Monitoring Drought over the Conterminous United States from MODIS and NCEP Reanalysis-2 Data, *Journal of Applied Meteorology and Climatology*, 49(8), 1665-1680
- ♣ Yiao, Y. **S. Liang**, Q. Qin, and K. Wang, S. Zhao, (2010), Monitoring Global Land Surface Drought Based on an Improved Evapotranspiration Model, *International Journal of Applied Earth Observation and Geoinformation*, 12(6): S266-S276
- ♣ Galleguillos, M., F. Jacob, L. Prévot, P. Lagacherie, **S. Liang**, (2010), Spatializing and validating daily evapotranspiration over a Mediterranean vineyard watershed, *IEEE Geoscience and Remote Sensing Letters*, in press
- ♣ Wang, K., B. Dickinson, M. Wild, **S. Liang**, (2010), Evidence for Decadal Variation in Global Terrestrial Evapotranspiration between 1982 and 2002, Part 1: Model Development", *Journal of Geophysical Research - Atmospheres* , 115
- ♣ Wang, K., B. Dickinson, M. Wild, **S. Liang**, (2010), Evidence for Decadal Variation in Global Terrestrial Evapotranspiration between 1982 and 2002, Part 2: Results", *Journal of Geophysical Research – Atmospheres* , 115
- ♣ Xu, T., S. Liu, **S. Liang**, & J. Qin, (2010), Improving Predictions of Water and Heat Fluxes by Assimilating MODIS Land Surface Temperature Products into Common Land Model, *Journal of Hydrometeorology* , in press



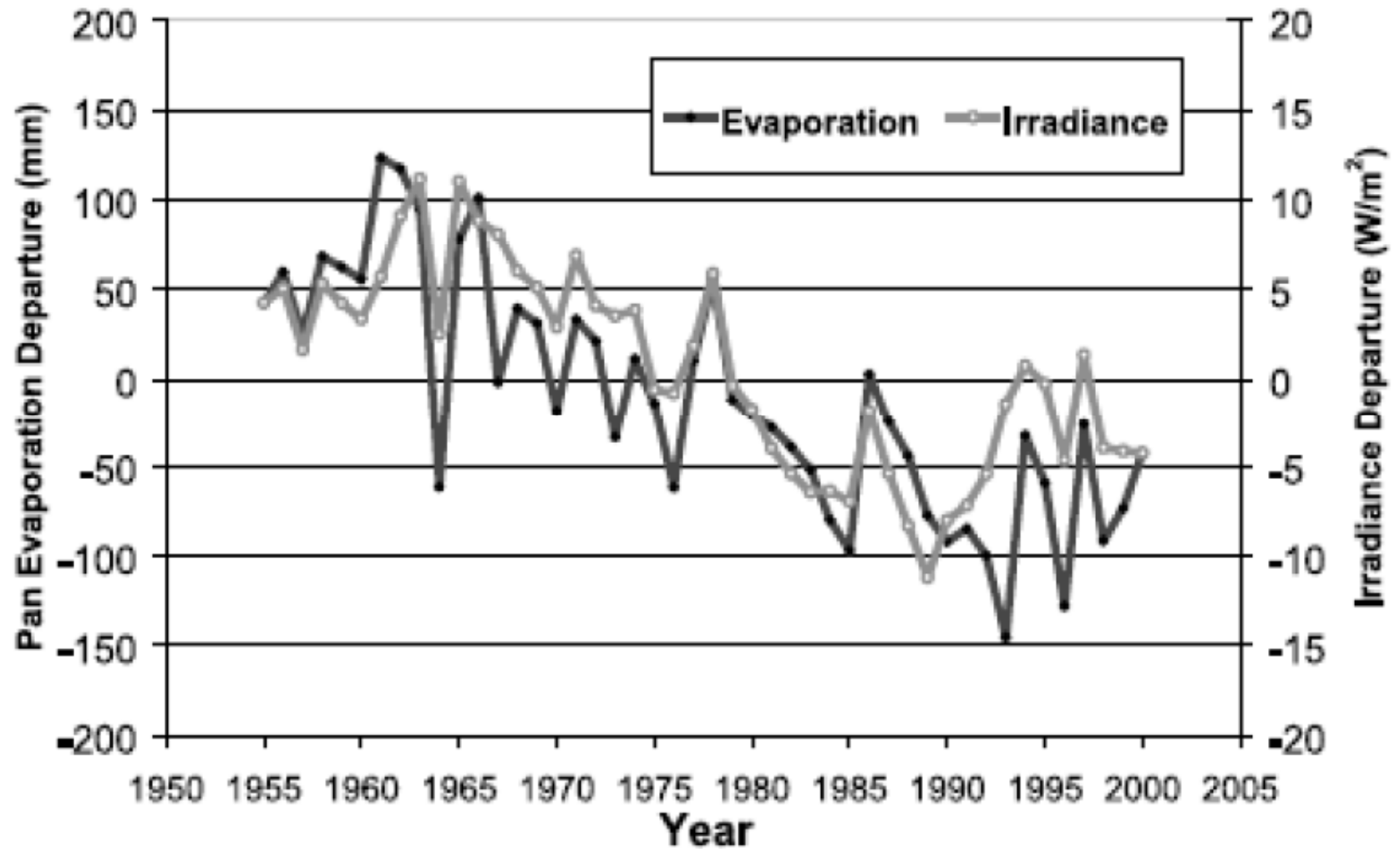
Peterson, T.C., Golubev, V.S. and Groisman, P.Y.
1995: EVAPORATION LOSING ITS STRENGTH.

Nature 377, 687-688.

Farquhar 2002 *Science*;

Ohmura and Wild 2002 *Science*

85 sites in China



Qian et al. 2006 GRL

Impact on pan evaporation

$$ET_E = \frac{\Delta}{\Delta + \gamma} \cdot R_s \cdot [a_1 + a_2 \cdot VI + RHD \cdot (a_3 + a_4 \cdot VI)], \quad (8)$$

$$ET_A = \frac{\gamma}{\Delta + \gamma} \cdot WS \cdot VPD \cdot [a_5 + RHD \cdot (a_6 + a_7 VI)], \quad (9)$$

$$ET = a_8 \cdot (ET_E + ET_A) + a_9 \cdot (ET_E + ET_A)^2, \quad (10)$$

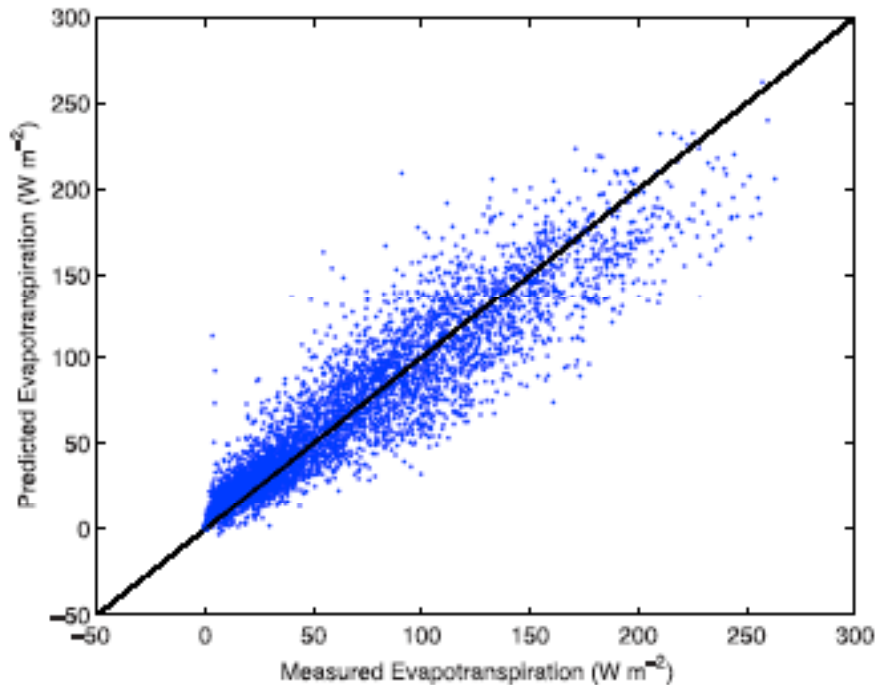


Figure 3. Comparison of the 16 day average predicted and ground-measured ET collected at all 64 sites shown in Table 1 when EVI is used. We used a 16 day average because the MODIS EVI data are available for a 16 day interval.

Wang, K., R. E. Dickinson, M. Wild, S. Liang (2010). "Evidence for decadal variation in global terrestrial evapotranspiration between 1982 and 2002: 1. Model development." Journal of Geophysical Research-Atmospheres **115**.

Wang, K., R. E. Dickinson, M. Wild, S. Liang (2010). "Evidence for decadal variation in global terrestrial evapotranspiration between 1982 and 2002: 2. results" Journal of Geophysical Research-Atmospheres **115**.

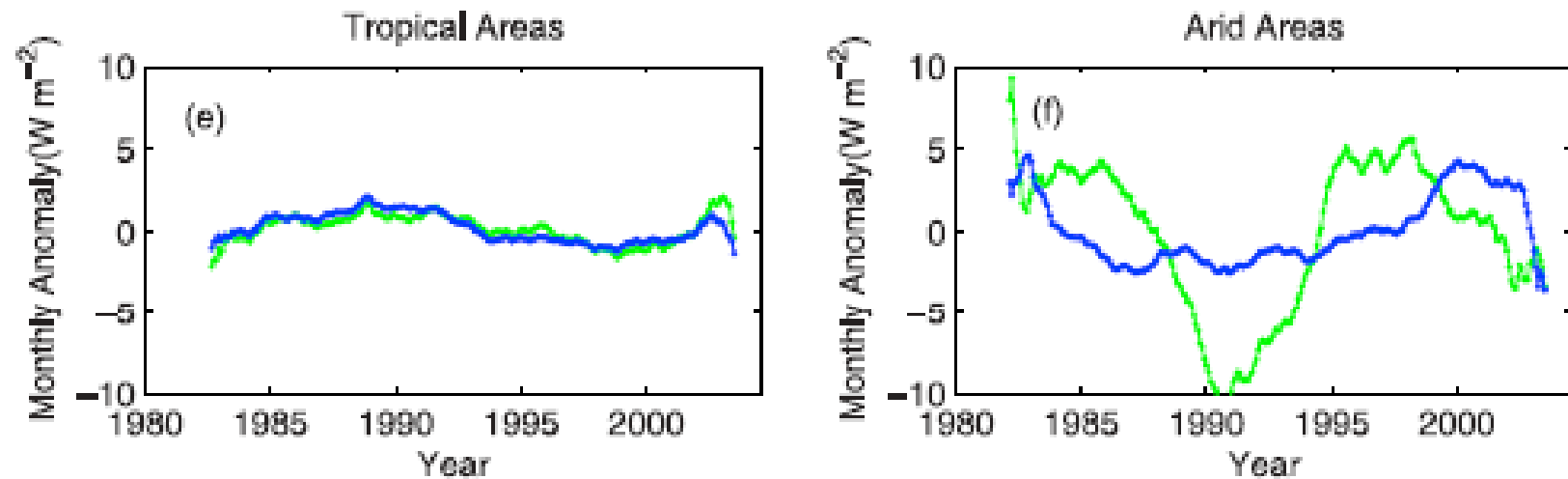
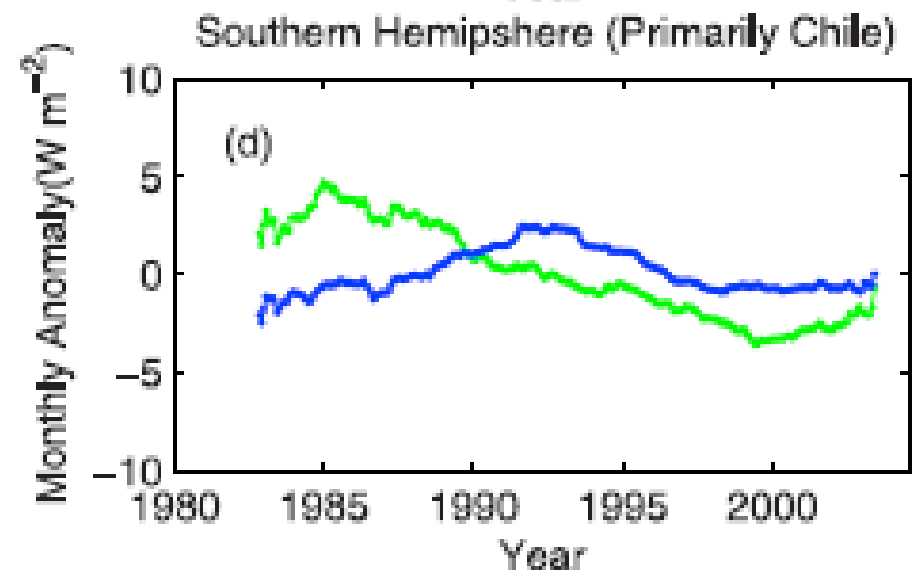
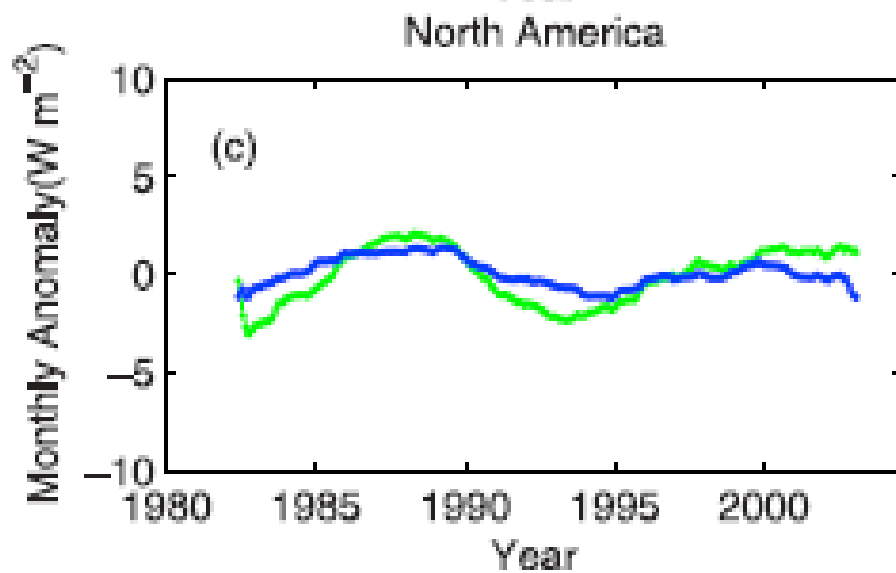
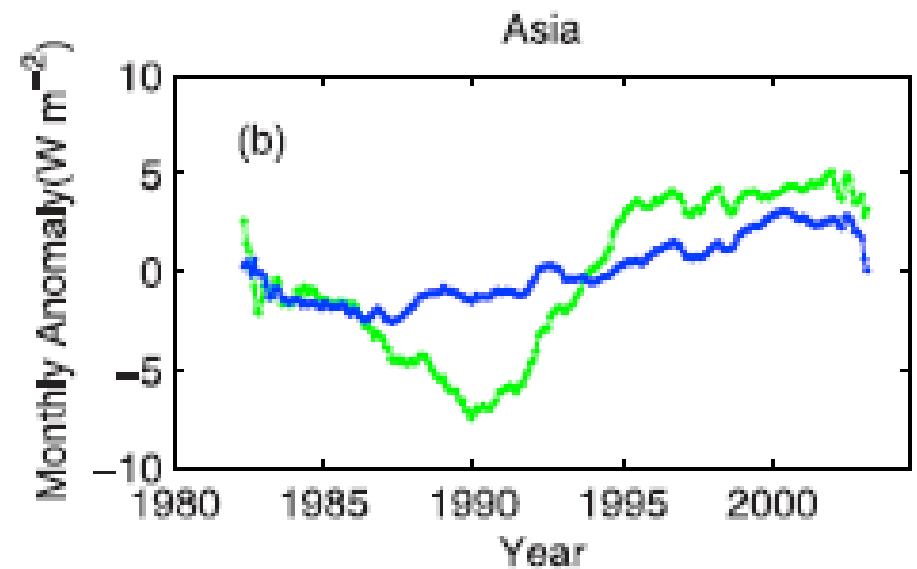
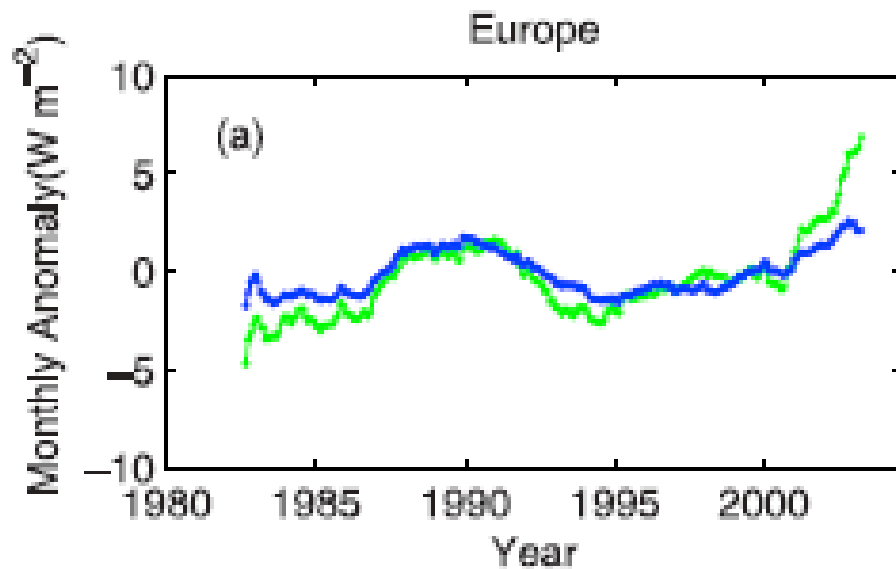
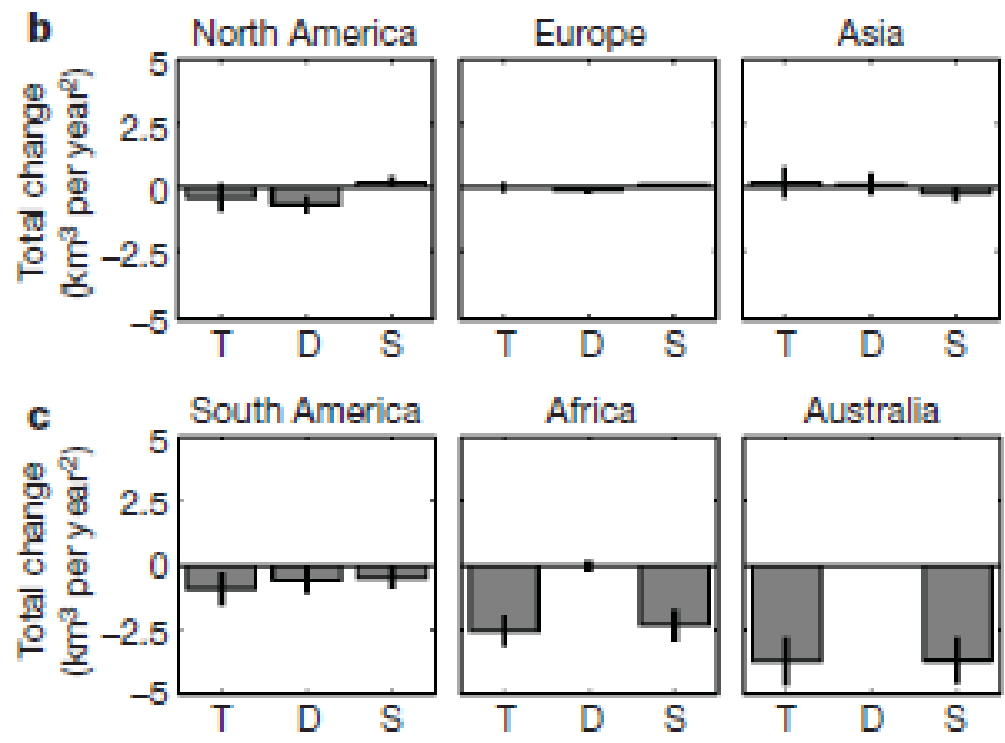
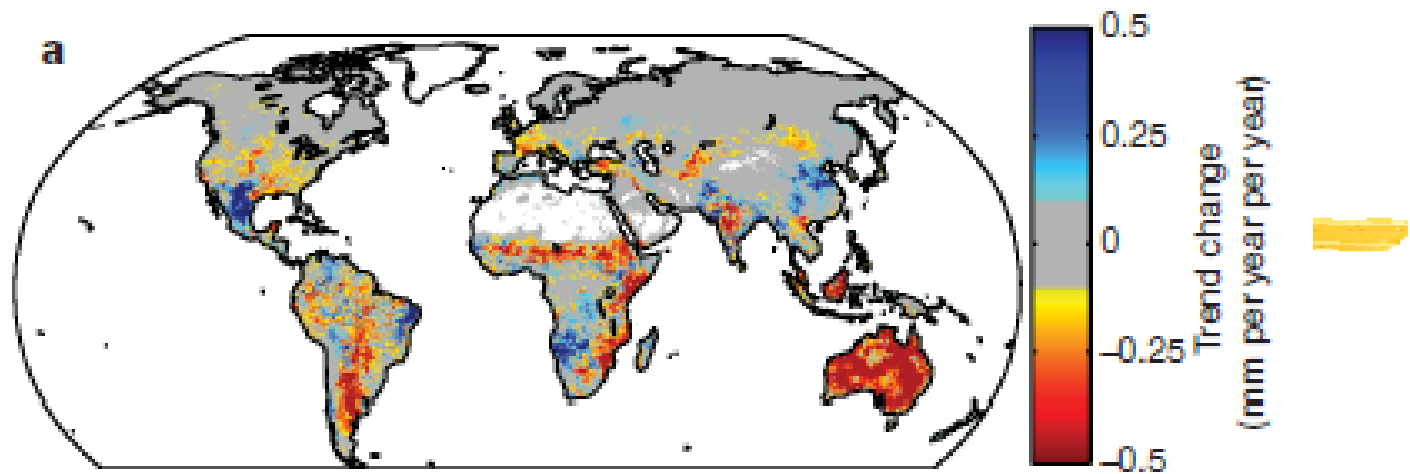


Figure 3. The times series of 5 year smoothed ET (blue line) and R_s (green line) anomalies for (a) Europe, (b) Asia, (c) North America, (d) the Southern Hemisphere, (e) tropical areas, and (f) arid areas. Tropical areas refer to regions where latitude is between $20^{\circ}S-20^{\circ}N$, and arid areas refer to regions where multiyear average relative humidity is less than 50%. ET and R_s demonstrate significant decadal variation in all the regions.



Global ET over land increased by $0.6 W m^{-2}$ per decade equal to $1.2 W m^{-2}$ (about 2.2% in relative value) or 15 mm yr⁻¹ in water flux from 1982-2002



Jung, et al.,
Nature, 2010

Figure 3 | ET trend changes. a, Map of the change in ET trend between 1982–1997 and 1998–2008 in millimetres per year per year. Small trend changes of ± 0.1 mm per year per year are shown in grey to enhance clarity. Total

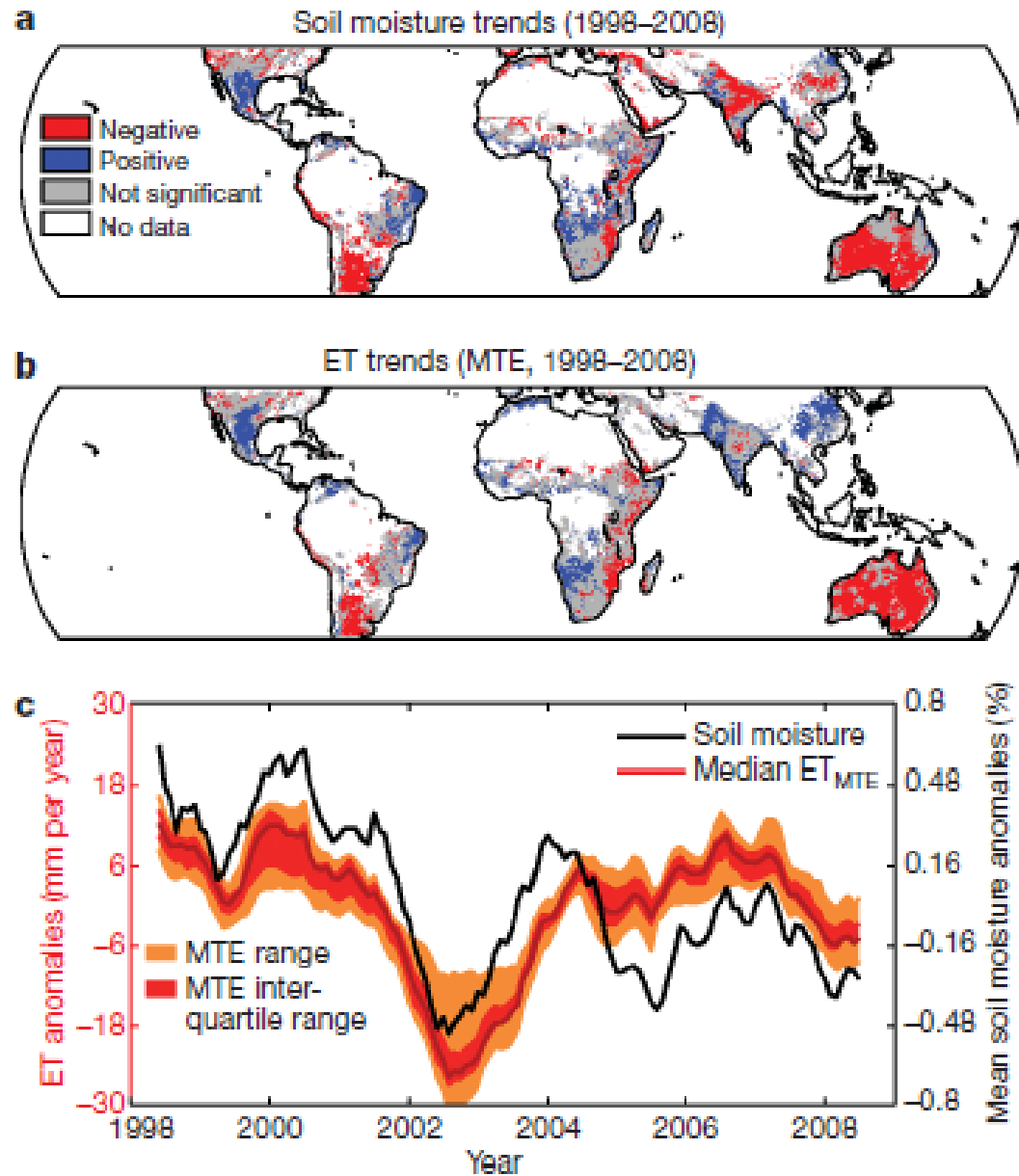
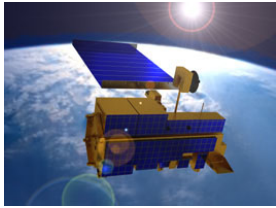
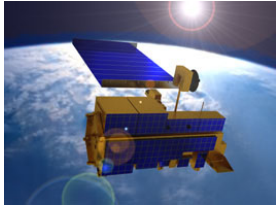


Figure 4 | Soil-moisture and ET trends. Significant ($P < 0.1$) soil-moisture trends derived from TRMM (a), significant ($P < 0.1$) ET trends from MTE (b) and mean ET and soil-moisture anomalies (seasonal cycle subtracted and filtered with an 11-month running mean) of all valid pixels of the TRMM



Summary

- ♣ Climate change and human activities have greatly impacted land surface radiation and energy budgets
- ♣ Many feedbacks and impacts have not been fully understood
- ♣ Satellite remote sensing can accurately map high-resolution land surface radiation and energy budgets and thus be greatly valuable for various applications



Thank  you !