

atmospheric AEROSOL

*Assessment of Measurements and
of efforts in Global Modeling*

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Overview

- **aerosol**
 - a short introduction
- **available aerosol data-sets**
 - useful to modeling?
- **AERONET and new satellite sensor data**
 - the strength of synergetic approaches
- **aerosol global modeling**
 - a BIG can of worms
- **AeroCom**
 - an international effort to reduce uncertainties in aerosol global modeling

aerosol, **what is it ?**

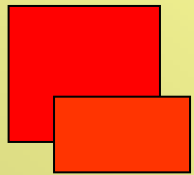
- **atmospheric particles smaller than cloud droplets**

aerosol, **where from?**

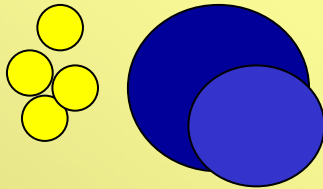
- **natural sources**
 - *(wind ⇒)* **dust, seasalt** *(lightning/fire ⇒)* **biomass**
- **anthropogenic sources**
 - *(industry ⇒)* **sulfate, nitrate, carbon** (fossil fuel)
 - *(land-use change ⇒)* **tropical biomass burning**

'primary' (as particles) - **'secondary'** (via the gas-phase)

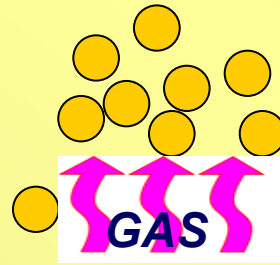
irregular shapes ... not just spheres
many sizes: nanometer to micrometer
strongly absorbing (soot) to non-absorbing



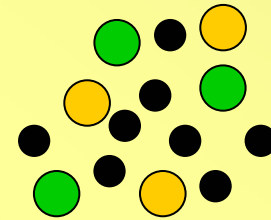
DUST



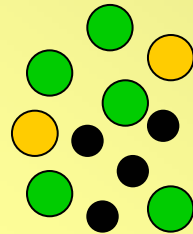
DMS, SEASALT



SULFATE



CARBONACEOUS



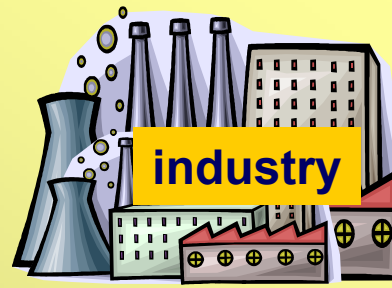
desert



ocean



volcano



industry



cities



forest

FIRE

aerosol, what properties?

- highly variable in space and time for

- **PROPERTY**

- amount
- composition
 - ⇒ absorption
- water uptake
- size

‘coarse’

($>1\mu\text{m}$ in size)

‘accumulation’

($.1-1\mu\text{m}$ in size)

‘Aitken’

($,01-.1\mu\text{m}$ in size)

‘nucleation’

(< $0.01\mu\text{m}$ in size)

- shape

- **OPTICAL PROPERTY**

‘aerosol optical thickness’

‘refractive index’, ‘ss-albedo’

‘hygroscopicity’ ...can change with time
orders of magnitude!

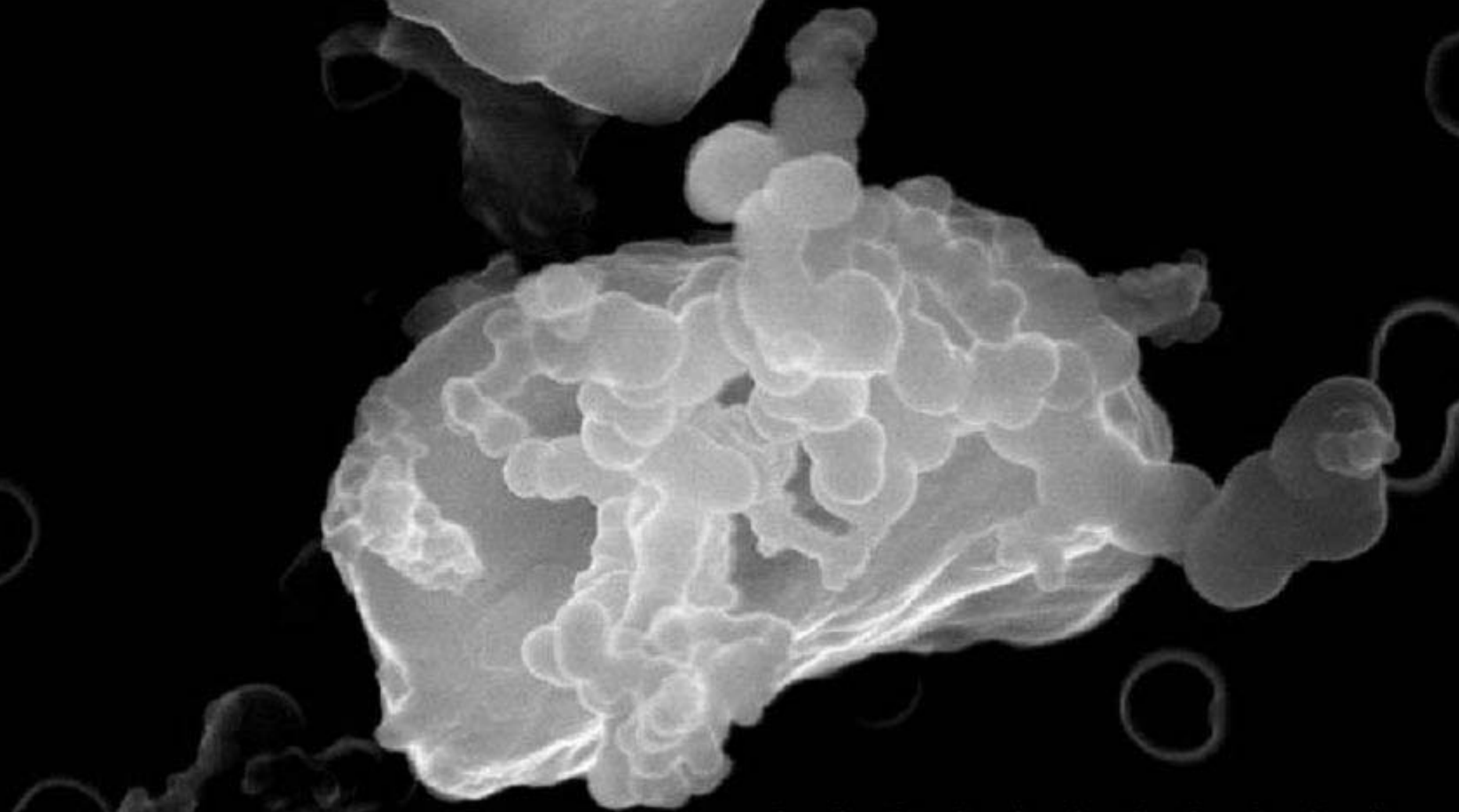
→ energy balance

→ solar energy balance

→ health

we prefer spheres (MIE) but...

ACE Asia: Supermicron Quartz with Soot



S4700 15.0kV 12.7mm x25.0k SE(U)

J. Anderson - Arizona State University

2.00um

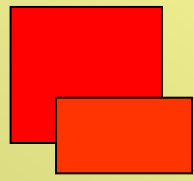
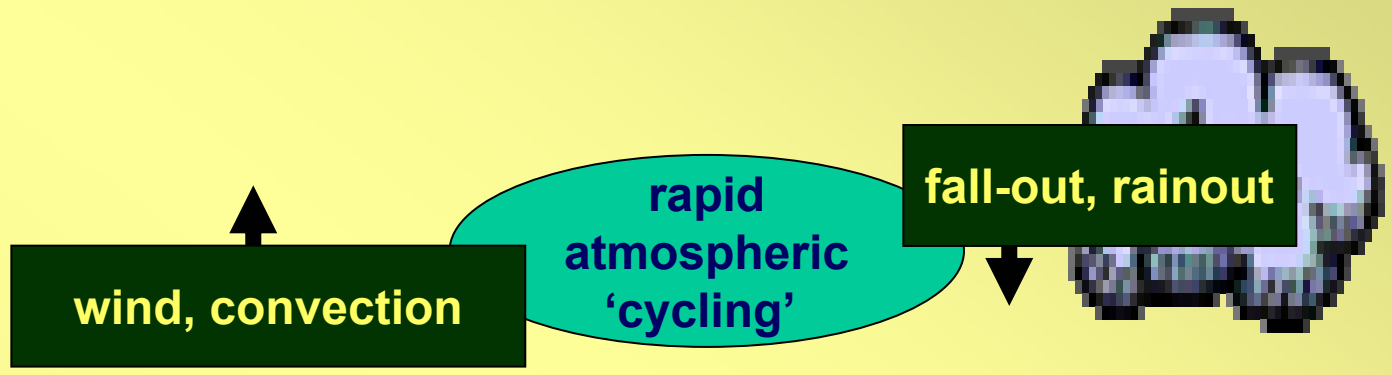
*...and we usually
assume spheres!*

components	DU	SS	SU	OC	BC
size-range (μm)	1-5	1-10	.1-1	.1-1	.1-.5
absorption	weak	no	no	weak	strong

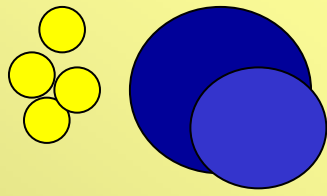
Aerosol

a modeler's nightmare

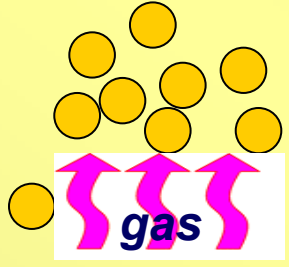
lifetime:
only a few days



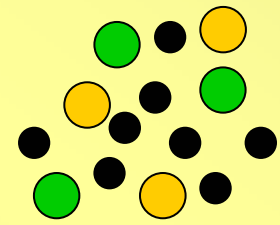
DUST



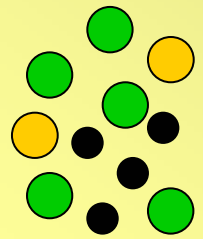
DMS, SEASALT



SULFATE



CARBONACEOUS



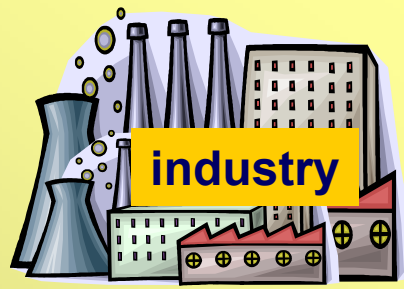
desert



ocean



volcano



industry



cities



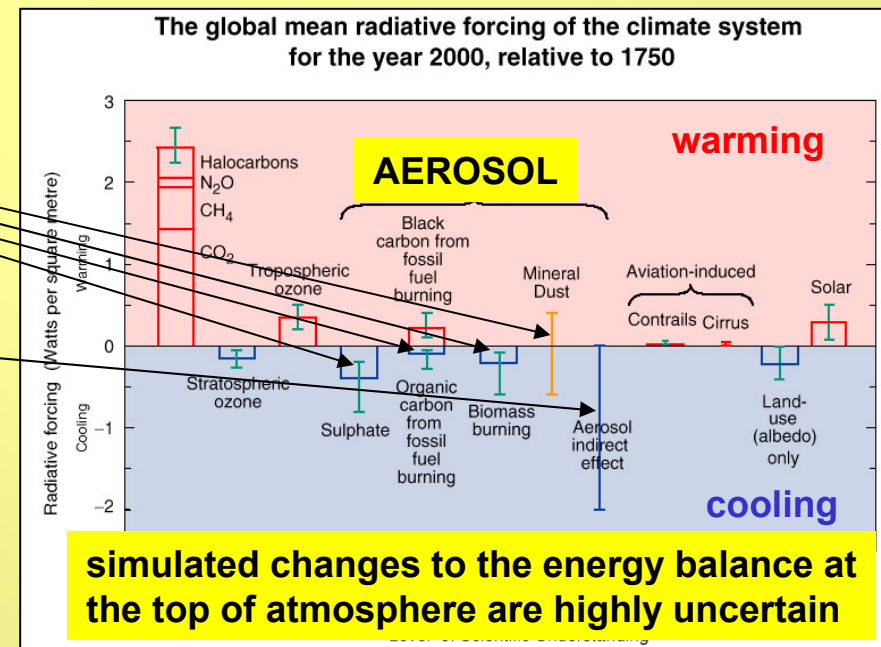
forest

FIRE

aerosol, why do we care ?

- effects on health (e.g. asthma cases)
- effects on quality of life (e.g. less clear days)
- effects on weather (e.g. precipitation)
- **effects on climate** (or the radiative energy balance)

- direct effects
from the presence of aerosol depends on the aerosol type
- indirect effects
through modifications of other atmos, parameters (e.g. clouds) many partially offsetting effects



aerosol, **good global data needed**

- **understanding of the aerosol climatic impact is based on MODELS**
 - **MODELS are as good as the data**
 - **MODELS need info on aerosol detail**

aerosol, **data of the past (1)**

- **simple climatologies**
 - **GADS was an initial attempt to define global aerosol based on local in-situ or remote sensing data**
 - **limited seasonality (Jan, Jul), 5deg lat/lon resolution**

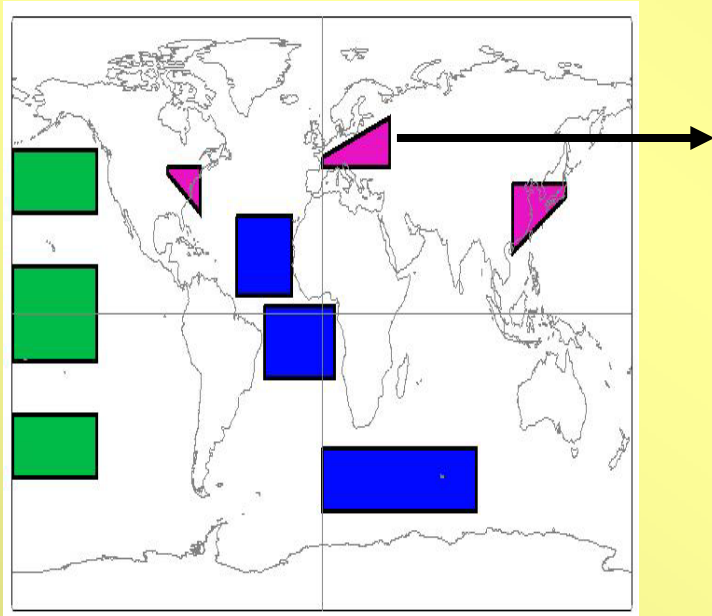
aerosol, data of the past (2)

- **25 years monitoring from space**
 - **AVHRR (visible / n-IR backscatter)**
 - amount: aerosol optical thickness (aot)
 - size: Angstrom parameter (α)
 - **TOMS (UV backscatter, 'molecular' contrast)**
 - amount: aerosol optical thickness (aot)
 - absorption: single scattering albedo (ω_0)
 - **SAGE/SAM (solar occultation)**
 - concentr./amount extinction and stratospheric aot
 - size: stratos. Angstrom parameter (α)

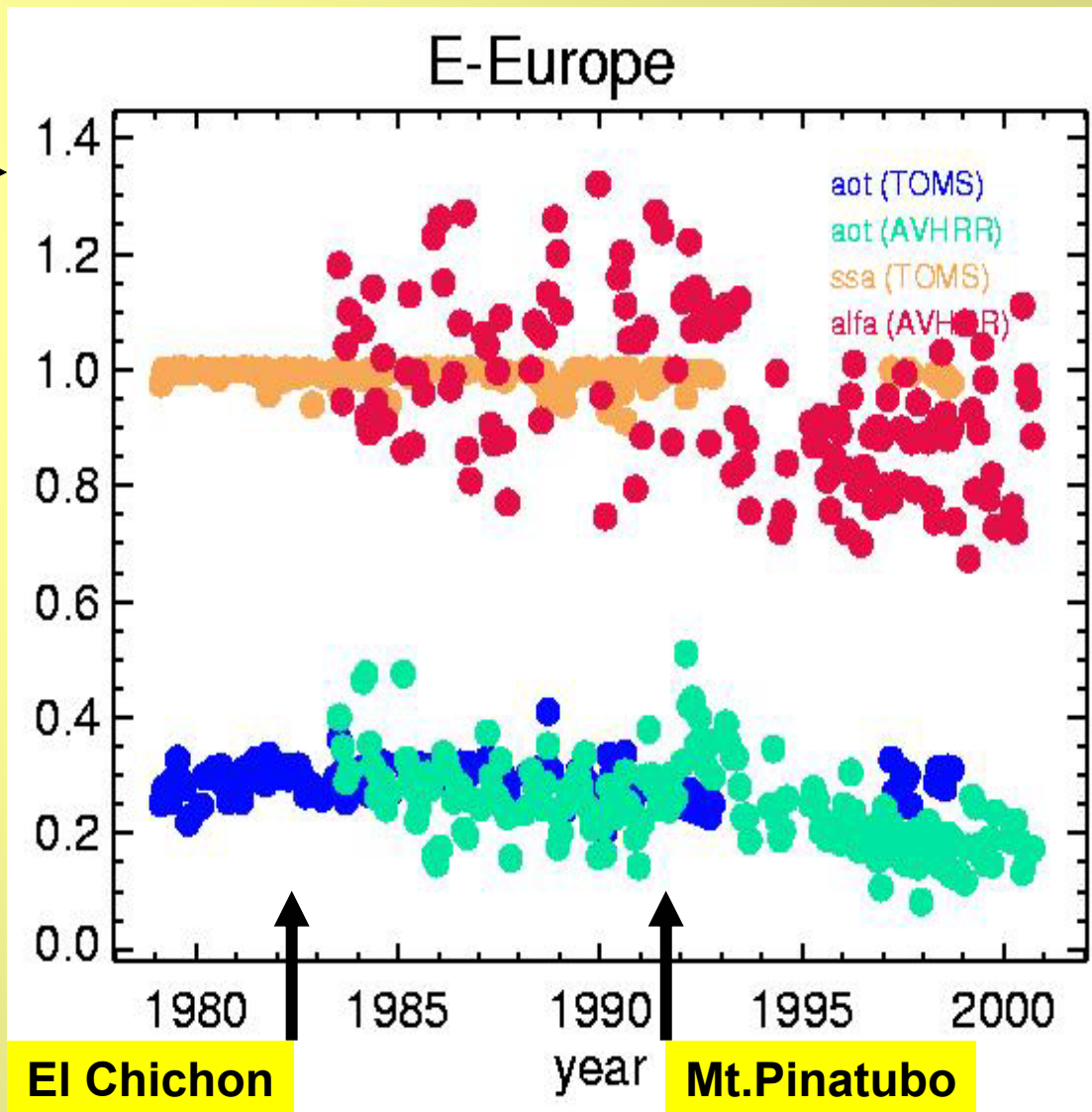
aerosol, retrieval limitations

- **a-priori assumptions are necessary**
 - at best two measurements ... many unknowns!
- **good cloud detection is essential** ⇨ fine pixels
 - false aerosol identification (sub-pixel clouds)
- **accurate (land) surface contributions are needed**
 - false aerosol identification (plancton, sub-pixel snow)
- **sensor and platform problems**
 - sensor drifts, overpass drifts, calibration issues
 - sensor data from different platforms
- **poor temporal resolution of polar-orbiters**
 - 'am' (TOMS) not necessarily agree with 'pm' data (AVHRR)

a regional example



- AVHRR: aot α
- TOMS: aot ssa

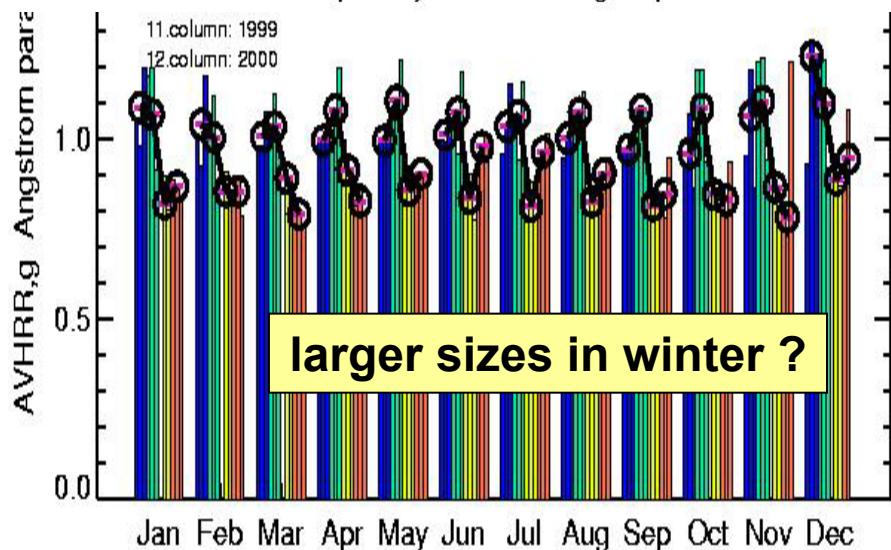
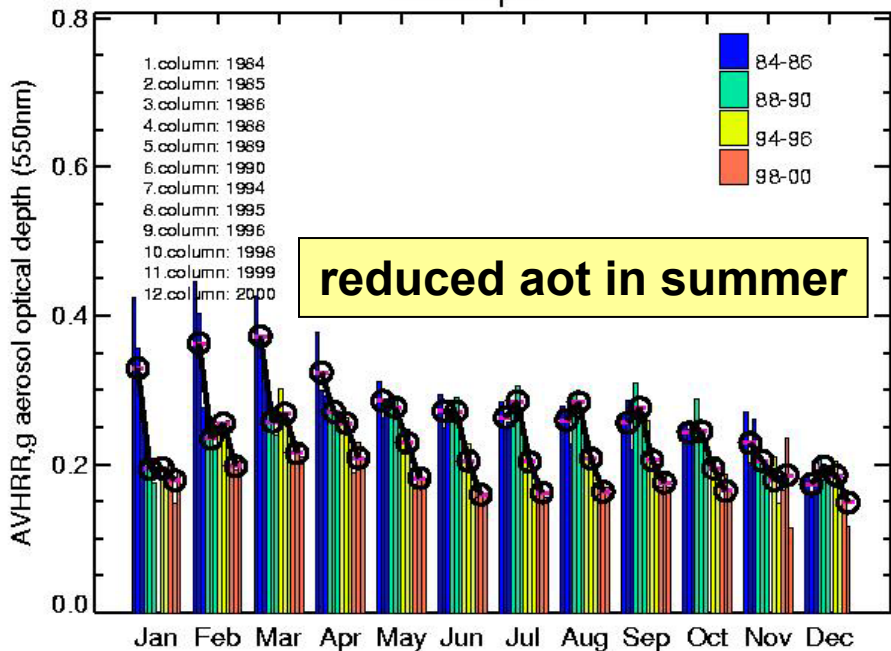


AVHRR

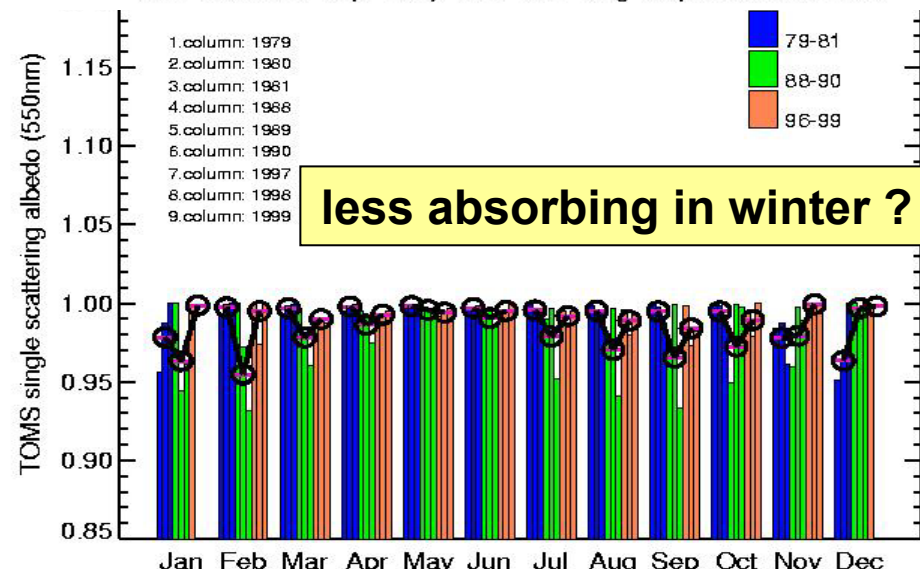
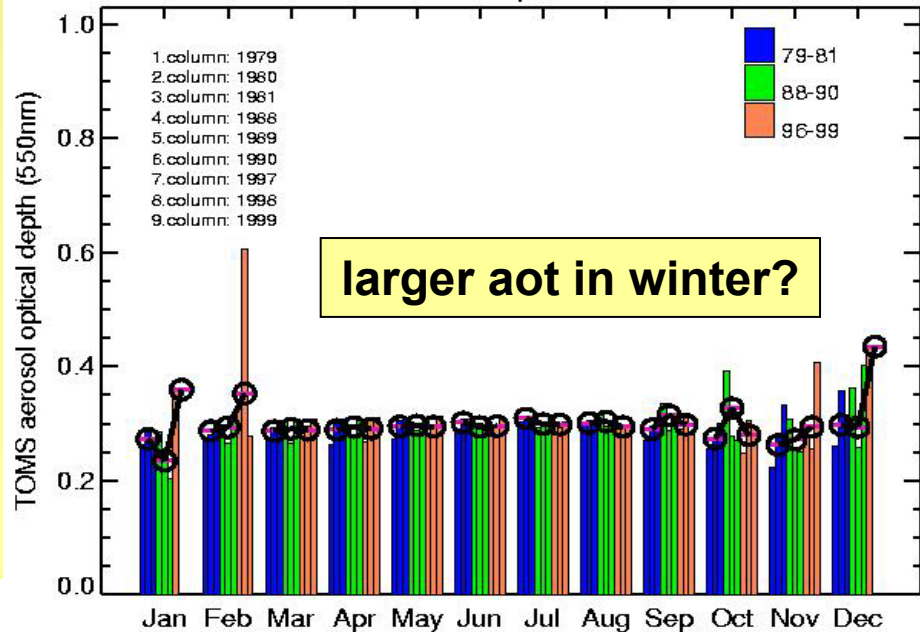
European trends ?

TOMS

E-Europe



E-Europe

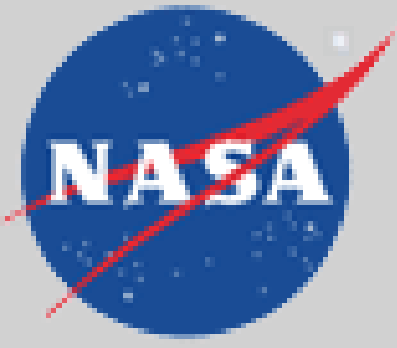


aerosol, **new data have arrived !**

- **new and improved satellite sensors**
 - MODIS, MISR, POLDER, MERIS, SAGEIII, Sciamachy, ...
 - **ground-based monitoring networks**
 - AERONET, SKYNET, IMPROVE, EMEP, EARLINET,...
- note: individual data-sets have individual strength*

aerosol, **in search of quality data**

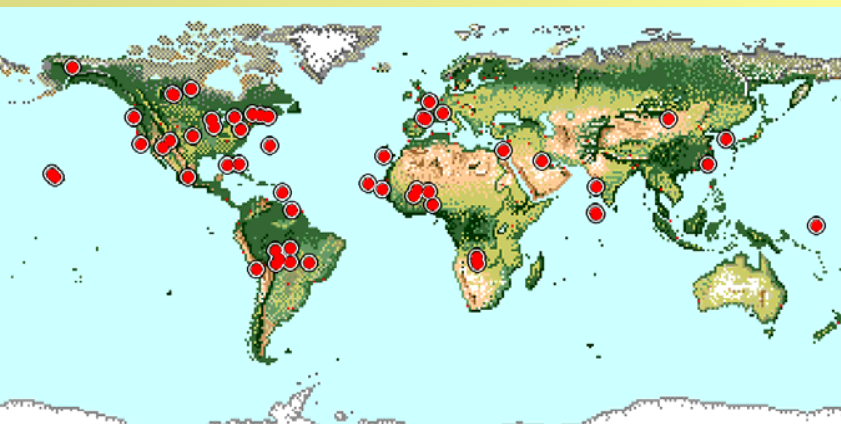
- **better data reference to global modeling**
require synergetic approaches: ***combine new information to coherent data-sets***
... thus a demonstration with **AERONET** data



AERONET



- worldwide network of robotic sun/sky-photometers (with satellite data transmission)
- supervised and maintained at NASA-Goddard
- many spectral bands
 - standard: 34, .38, .44, .50, .67, .87, .94, 1.02 μm
 - polarization: 5 bands + polarization at .87 μm
 - new instrument: added bands at 1.6 and 2.2 μm
- sampling: 1/hr (sky-mode), 4/hr (sun-mode)
- retrieved aerosol properties



- optical depth
- Angstrom parameter
- size-distribution (22 bins from .5-15 μm)
- refractive index \Rightarrow (ss-albedo)
- non-sphericity

AERONET statistics

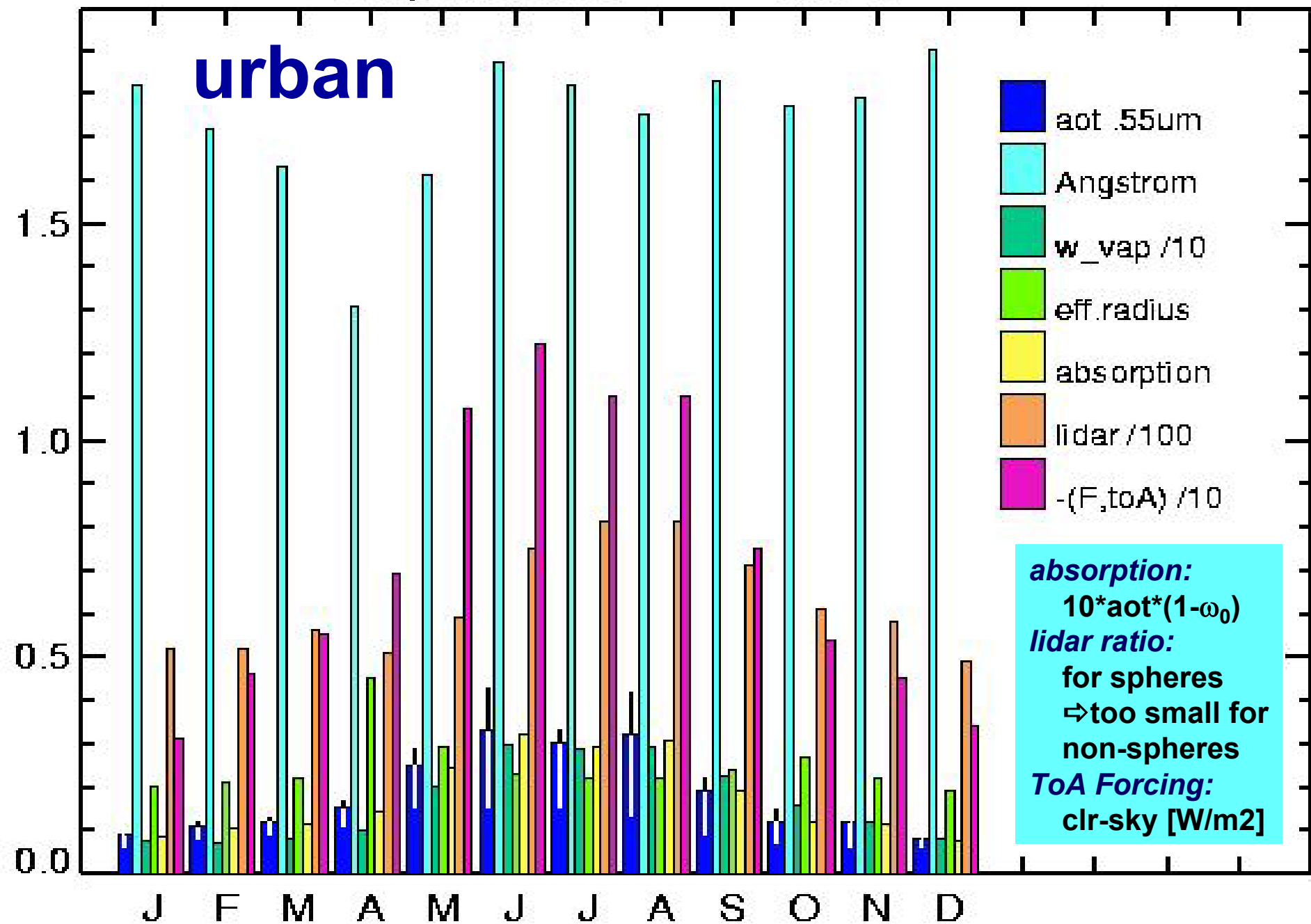
monthly average properties

- **a sampler for three sites:**
 - **GSFC (near Washington DC) ‘urban’**
 - **Mongu (Zambia) (JUL-NOV) ‘biomass’**
 - **Cape Verde (west of Sahara) ‘dust’**
- **measured properties: (aot, Angstrom)**
- **derived properties: (absorption, size)**
- **value-added properties: (forcing, lidar ratio)**
- **locally – aerosol is completely defined !**
 - **limitation: column data, no info on ‘components’**

Properties at

GSFC

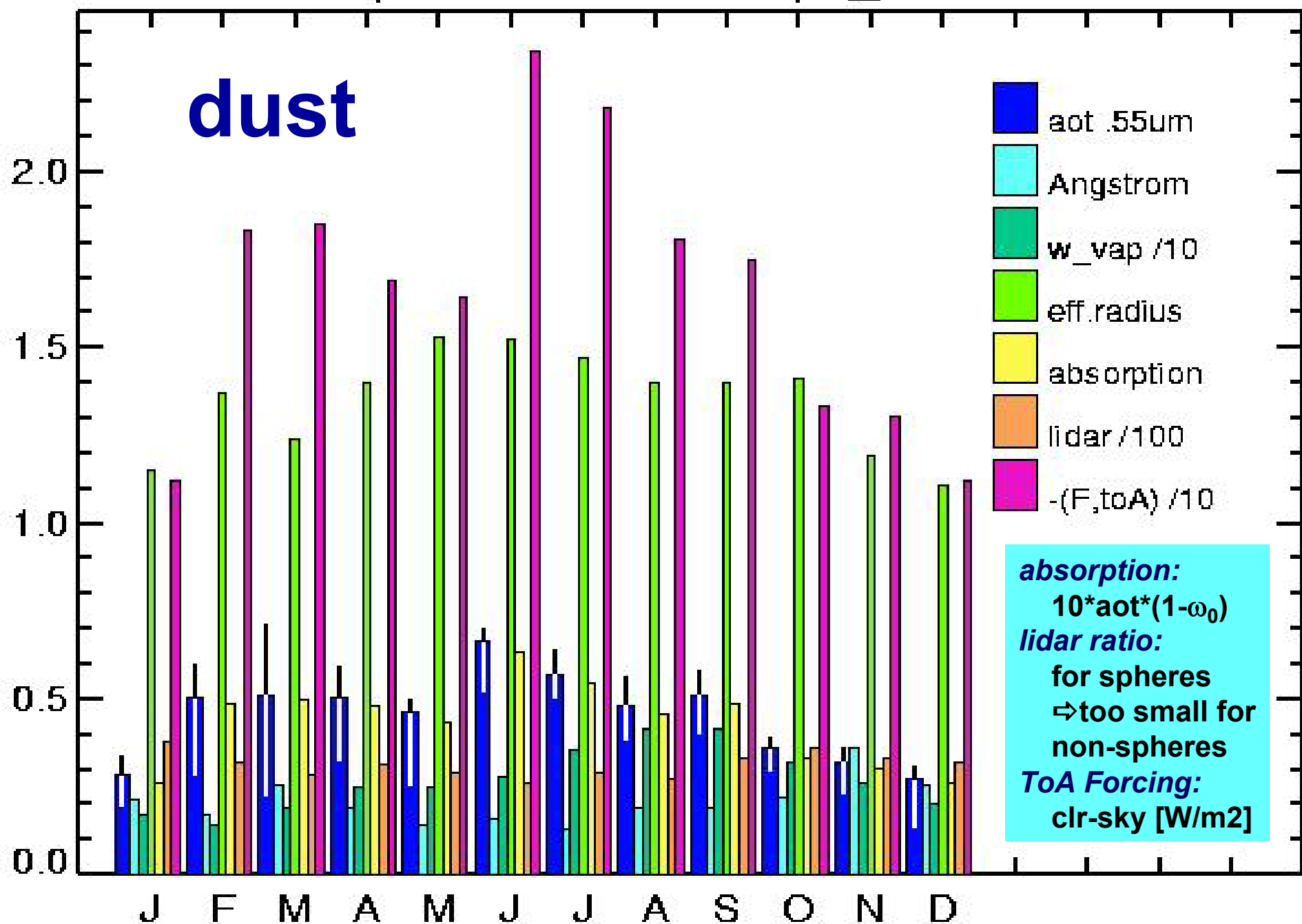
urban



Properties at

Cape_Verde

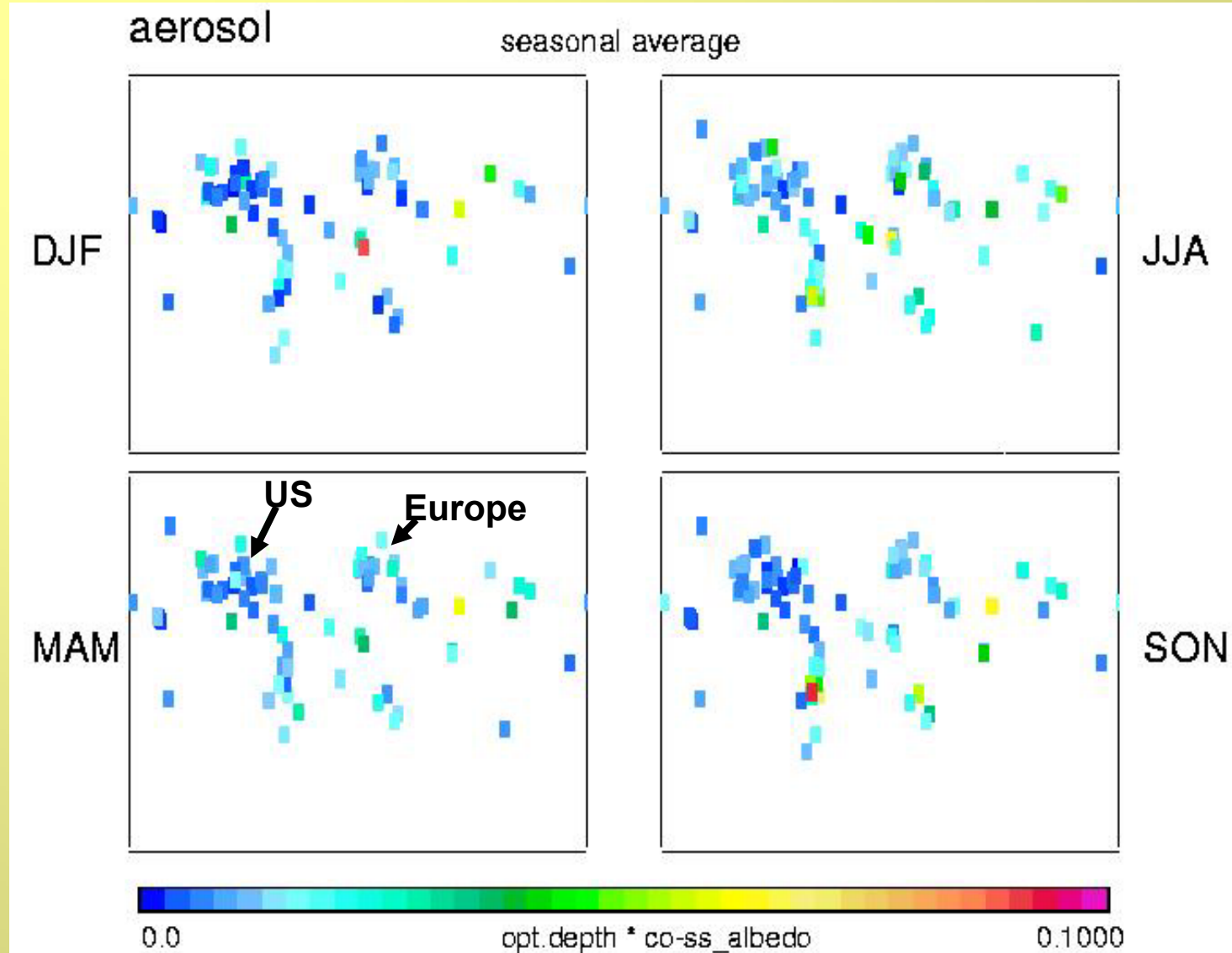
dust



the consistency among data allows combination for global assessment

this example:
seasonal avg.
for aerosol
absorption
 $[\tau * (1-\omega_0)]$

interesting...
... AERONET
indicates more
absorption by
aerosol over
Europe than
over east- US



data harmonization

in search of the **best** possible data

best

- in terms of accuracy
 - in terms of detail
 - In terms of consistency (different aerosol properties)
 - in terms of resolution (temporal and spatial)
-
- **AERONET and satellite retrievals**
(similarly AERONET and modeling)
 - **examples on**
 - how AERONET data can 'help' (\Rightarrow)
 - how AERONET data can 'learn' (\Leftarrow)

new generation of sensors

- **high spatial resolution**
 - as good as 1km * 1km
- **more spectral detail**
 - MODIS, MERIS (aerosol and cloud data!)
- **multi directional**
 - MISR, AATSR (land-retrievals, altitude info)
- **polarization**
 - POLDER (polarized signal for land retrieval)
- **vertical profiling with lidar and radar**
 - CALIPSO, CLOUDSAT (to be launched next year)

AERONET ⇨ satellite data

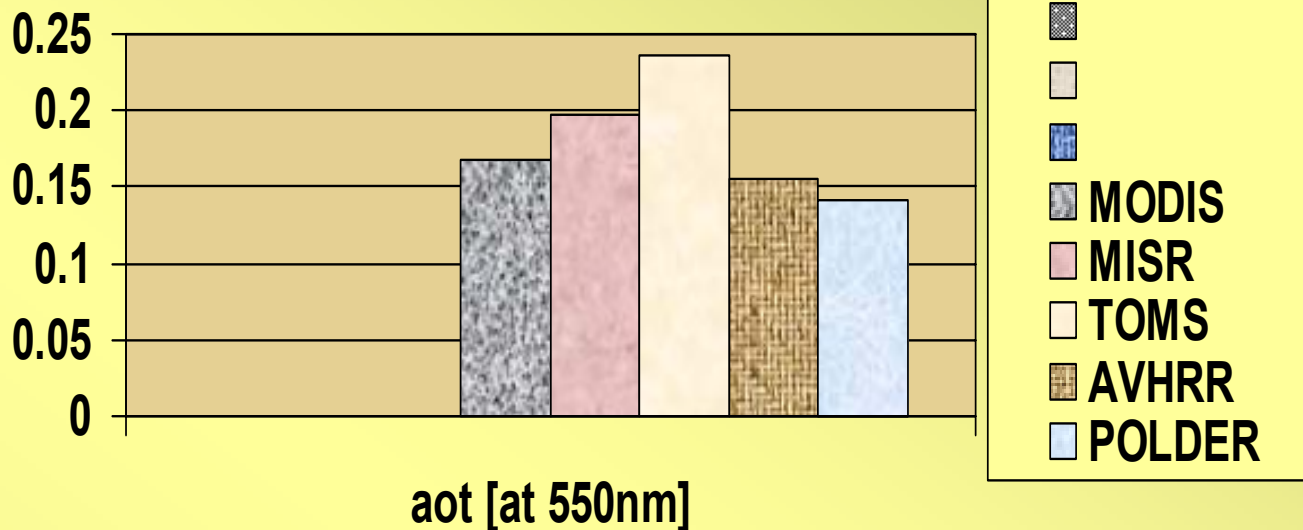
satellite aot data (*aerosol optical thickness or depth*)

- **what is available ? what is best?**

<i>Satellite</i>	<i>Advantage</i>	<i>Disadvantage</i>
AVHRR TOMS	<i>historic record</i> <i>historic record</i>	calibration, not over land large (50km) pixel height or absorption assumed
MODIS MISR POLDER	<i>small pixel</i> <i>altitude info</i>	failure over deserts temporally sparse short record, over land: less sensitive to large sizes
SEAWIFS GOES or MSG	<i>high temporal resolution</i>	not over land, no IR channels lack of detail with broad bands very limited over land

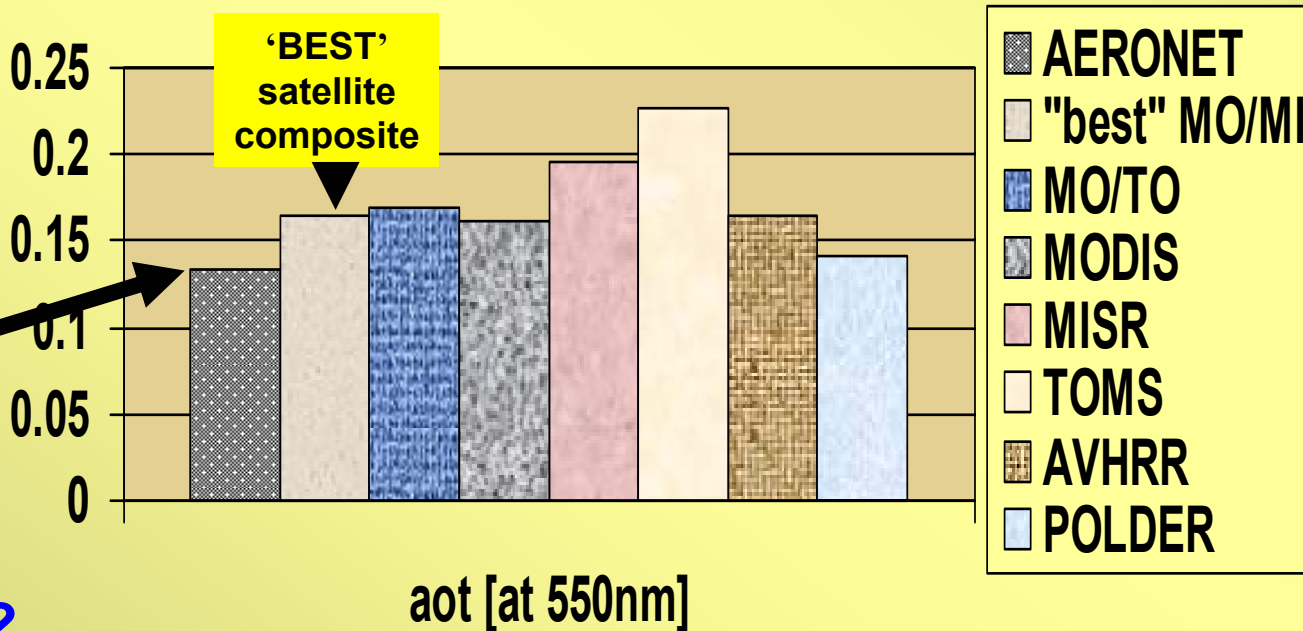
aot - global yearly averages

averaging
over all
available ⇒
data



normalized
by model
to offset
sampling
biases ⇒

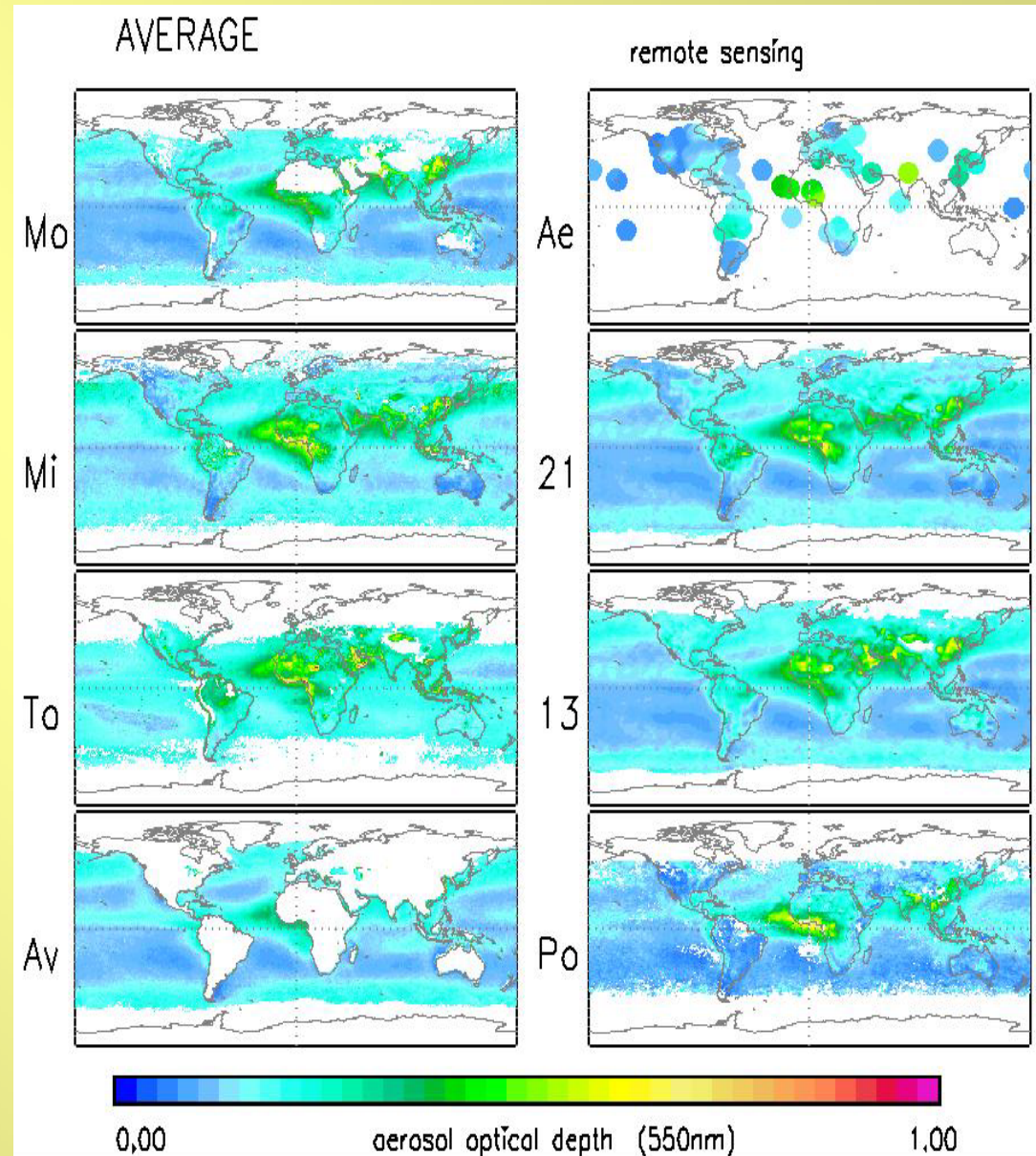
AERONET
has lower aots
than satellites
a clear-sky bias?



comparisons or annual pattern

Mo: MODIS composites:
Mi: MISR 12:Mo,Mi
To: TOMS 13:Mo,To
Av: AVHRR
Po: POLDER Ae:Aeronet

difficult to depict a
best global retrieval
⇒ composite needed
a MODIS (ocean) MISR
(land) combination
seems promising ...
...but differences to
AERONET still exist

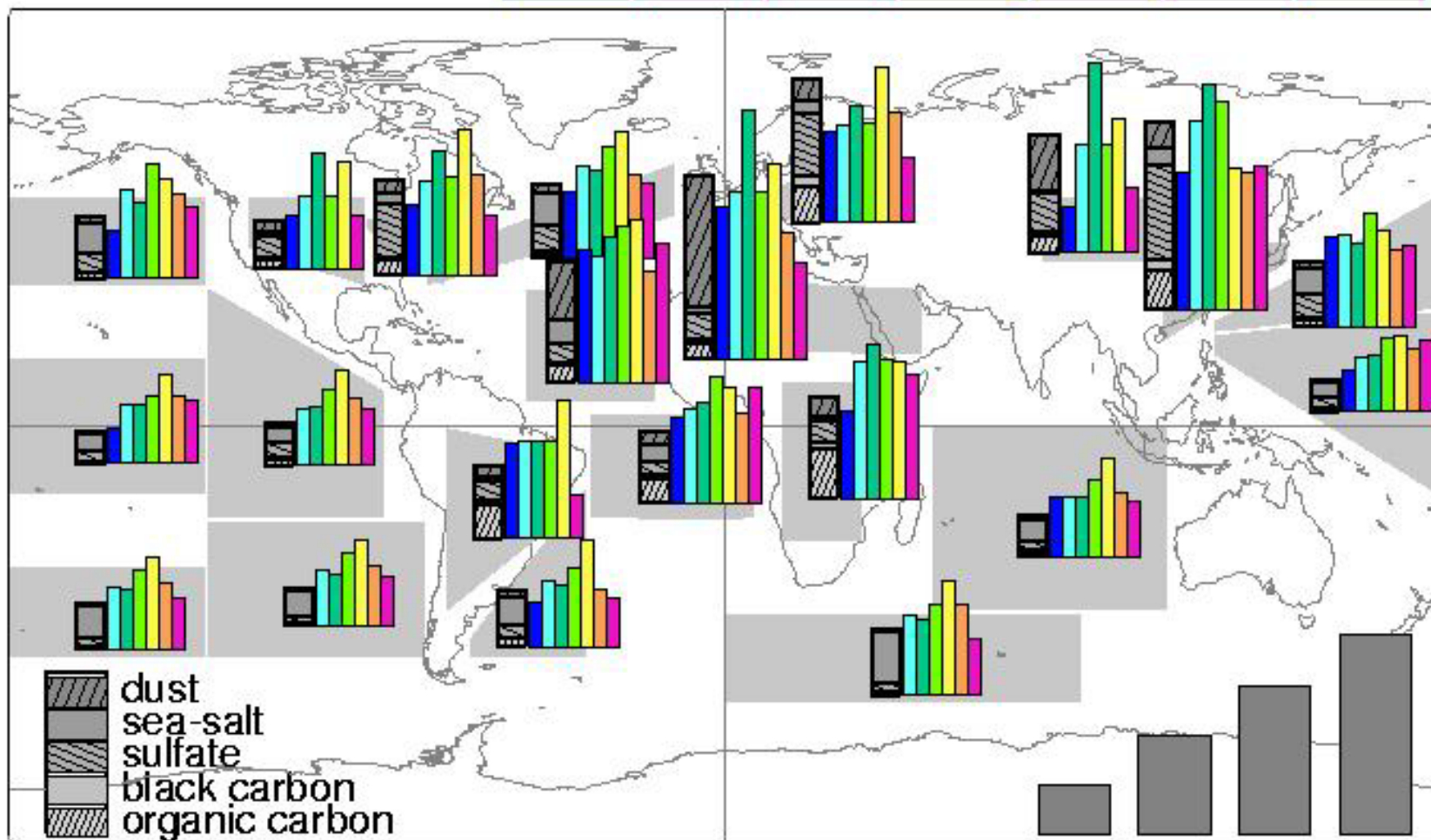


regional avgs highlight differences

avg.model

and measurements AOT

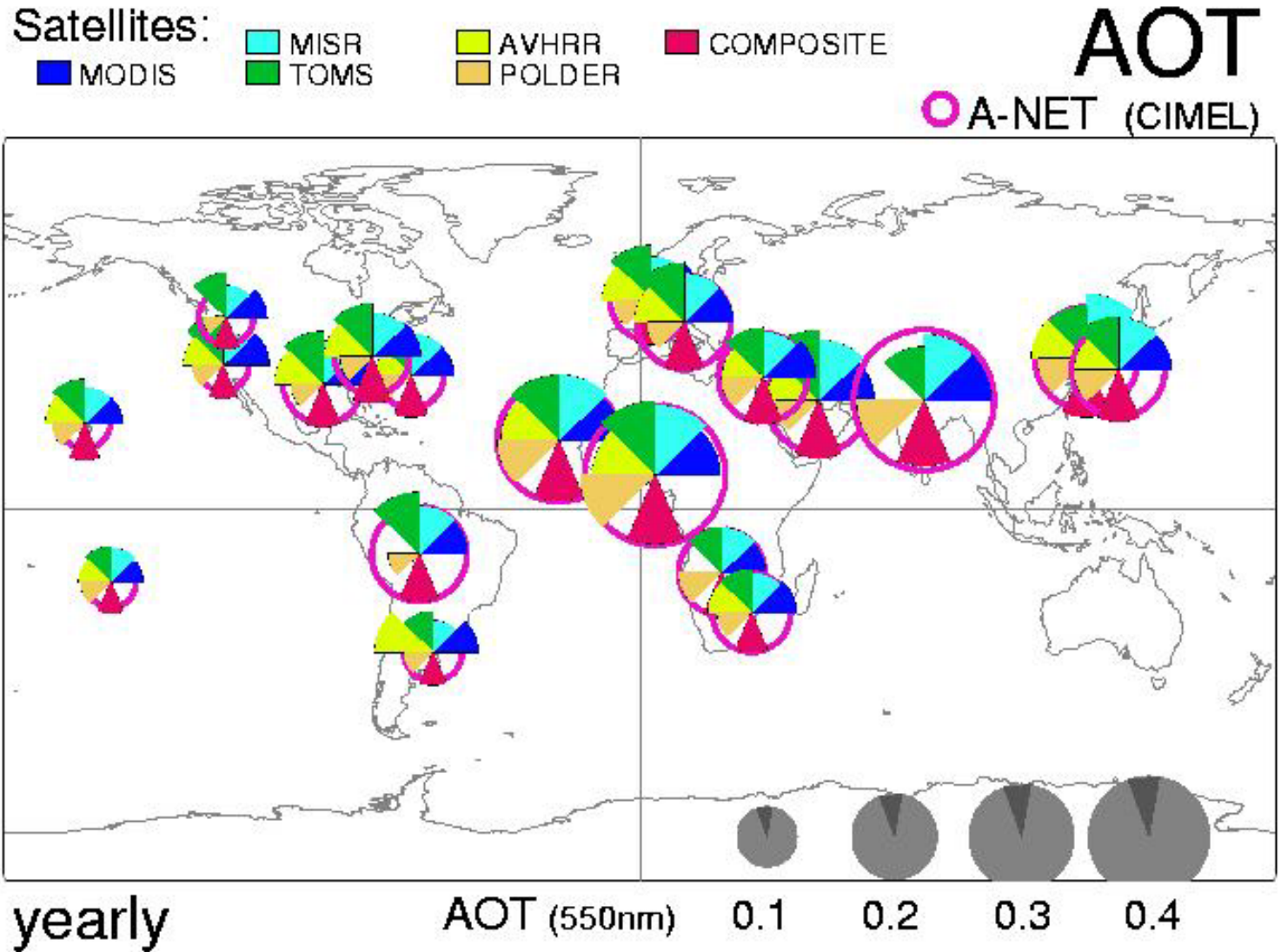
A-NET S-MIX MODIS MISR TOMS AVHRR POLDER



yearly

AOT (550nm) 0.1 0.2 0.3 0.4

local comparisons to AERONET



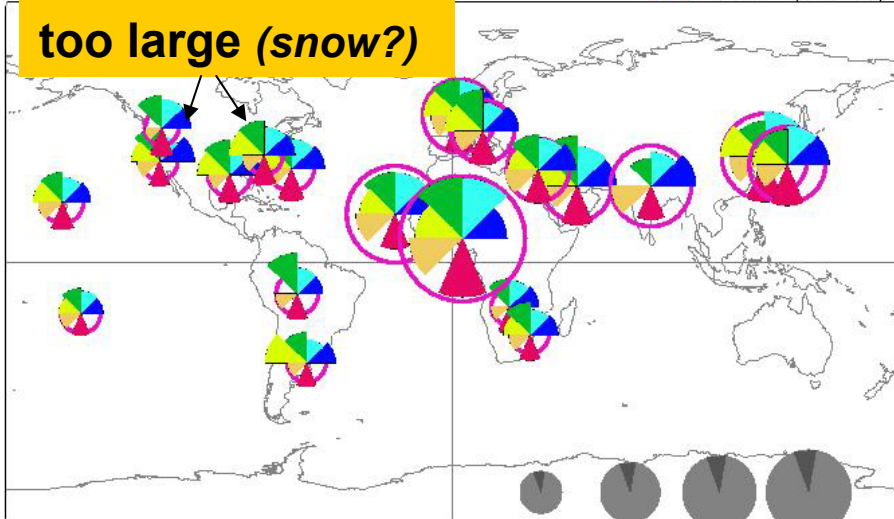
composite:
MISR /
MODIS

... still
satellite
data are

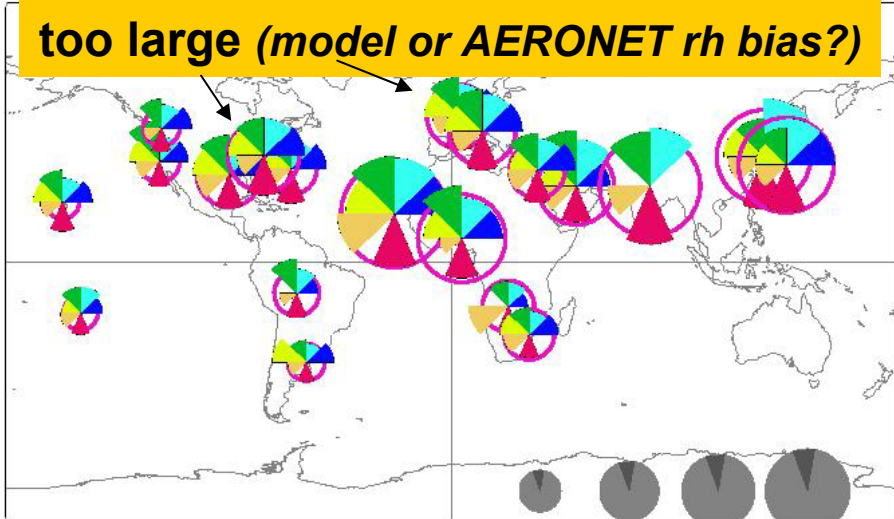
generally larger than AERONET, particular in urban regions

seasonal comparisons at AERONET

Satellites: MISR, AVHRR, COMPOSITE, MODIS, TOMS, POLDER, A-NET (CIMEL)

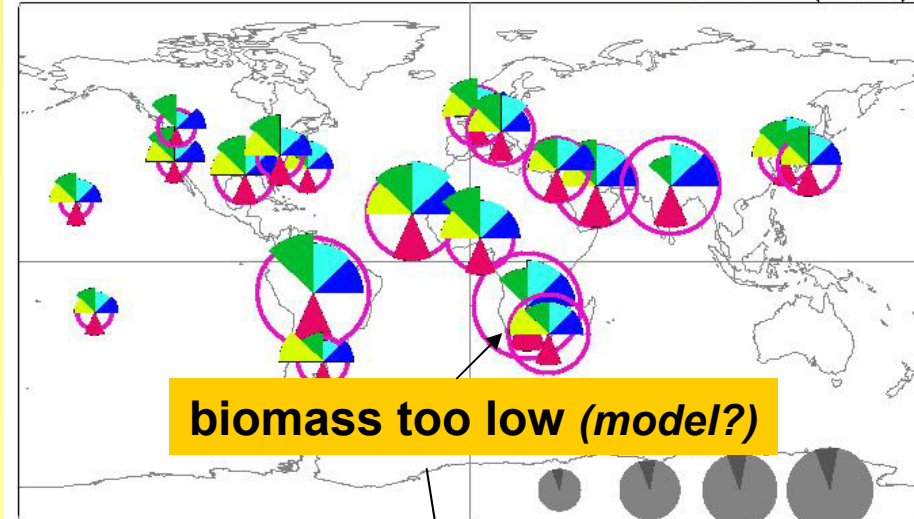


Mar AOT (550nm) 0.1 0.2 0.3 0.4

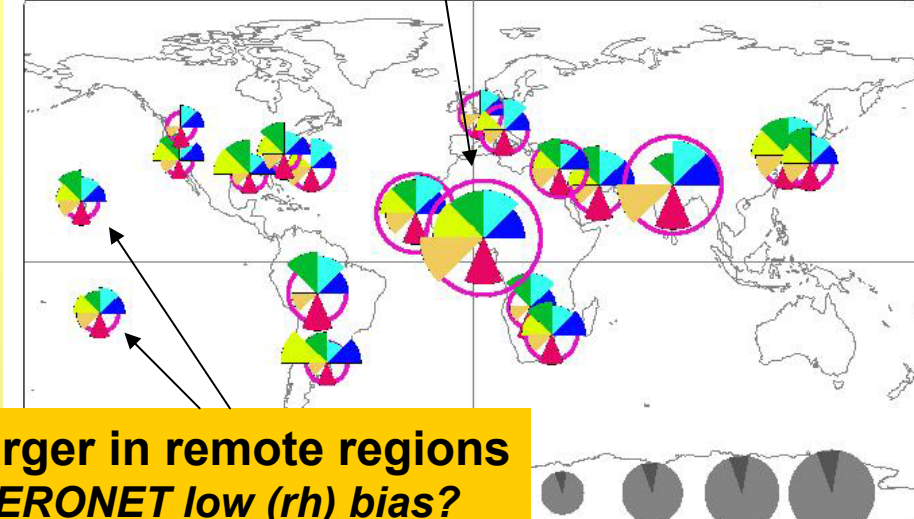


Jun AOT (550nm) 0.1 0.2 0.3 0.4

Satellites: MISR, AVHRR, COMPOSITE, MODIS, TOMS, POLDER, A-NET (CIMEL)



Sep AOT (550nm) 0.1 0.2 0.3 0.4



Dec AOT (550nm) 0.1 0.2 0.3 0.4

first impressions

- **MODIS** best choice over the oceans ... but too low in dust outflow regions (*high aot 'filtered as' clouds*)
- **MISR** most complete land cover ... while biased high over oceans (*poor temporal sampling at ca 1/week*)
- **MODIS (ocean) / MISR (land) combination** the 'best' satellite product is generally larger than AERONET ... but too low during the biomass burning season

open issues:

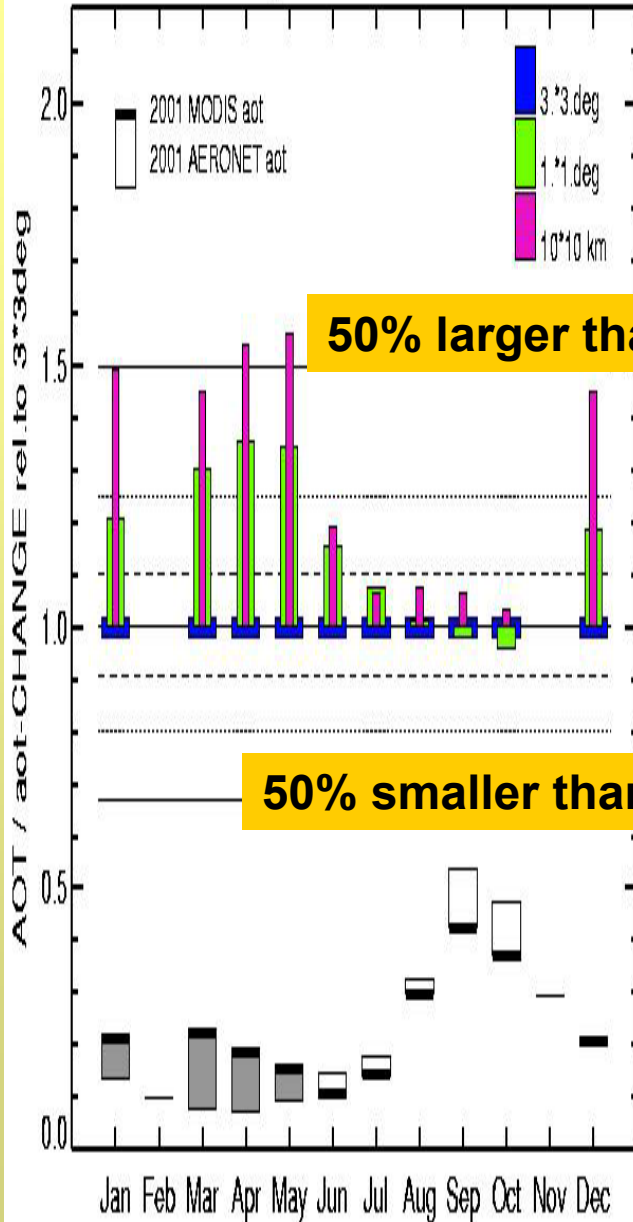
- are AERONET aot smaller due to a clear-sky bias?
- what can be said about the quality of retrievals of low aot in remote regions (of no AERONET sites?)
- is it 'fair' to compare point data with regional data?

satellite data ⇨ AERONET

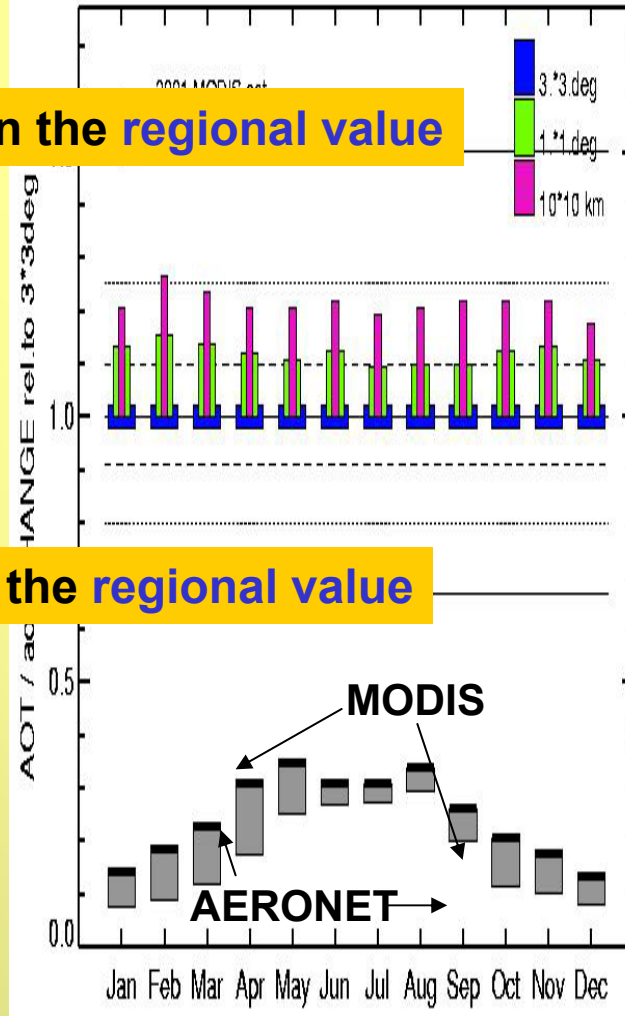
- **use spatial information of satellite data**
 - **to relate local measurement detail to**
 - coarse gridded data-sets
 - coarse resolution data in global modeling
- **how ?**
 - **compare averages for different scales**
 - agreement ... indicates a 'useful' site
 - bias: 'useful' site after a bias adjustment
 - highly variable (season/years) : leave off comparison ... unless secondary data exist

“scaling”

Mongu -15.25 23.15



GSFC 39.02 283.13



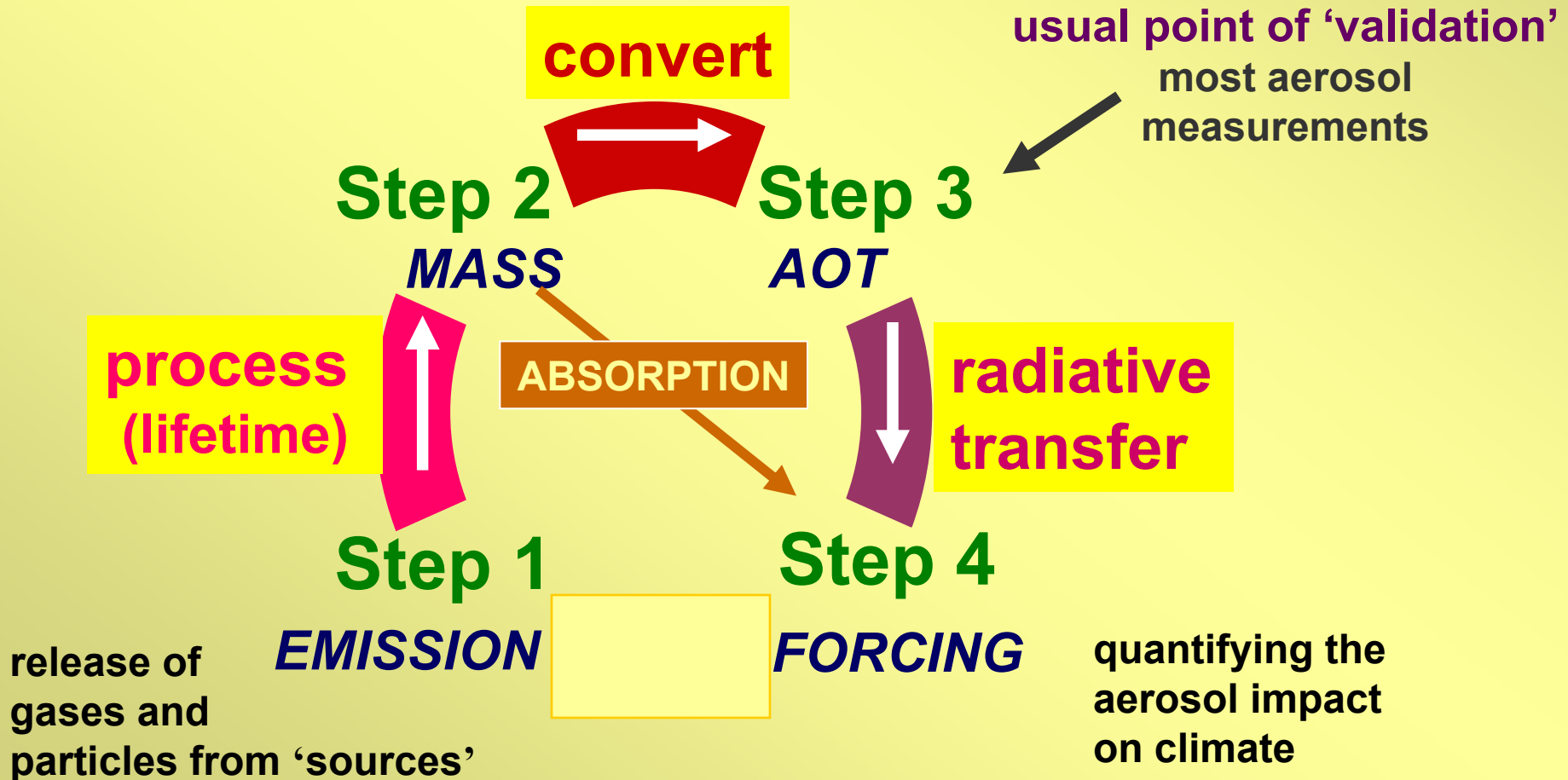
- Comparison of
 - 300*300km data
 - 100*100km data
 - 10*10km data
 - GSFC (urban)
 - 20% above the regional average
 - Mongu (biomass)
 - good match for the biomass season (Jul-Nov)
- ↩ at the bottom are AERONET-MODIS comparisons (2001)
note: MODIS statistics are very poor!

needed scaling activities

- for different spatial domains a data-base of simultaneous satellite retrievals over AERONET sites is needed
- satellite requirements:
 - small (~1km) pixel retrievals at regional coverage
 - sufficient data (for seasonal /annual dependence)
 - coverage of all AERONET sites (incl. desert sites)

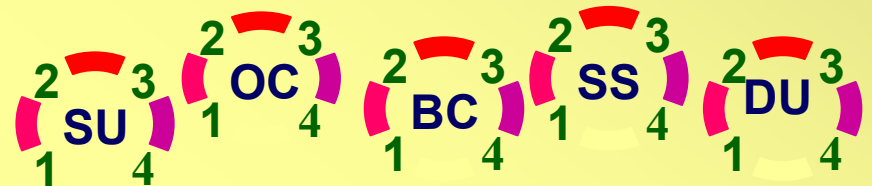
MODIS and MISR data are a start ... although their smallest pixels size at 10.0 and 17.6 km is too large to represent 'truly' local characteristics

Aerosol (in global) modeling a 4 Step process



Tuning opportunities !

- better aerosol modules in all major climate models distinguish SU, SS, DU, OC, BC
 - many processes and assumptions (⇒ new errors ?)

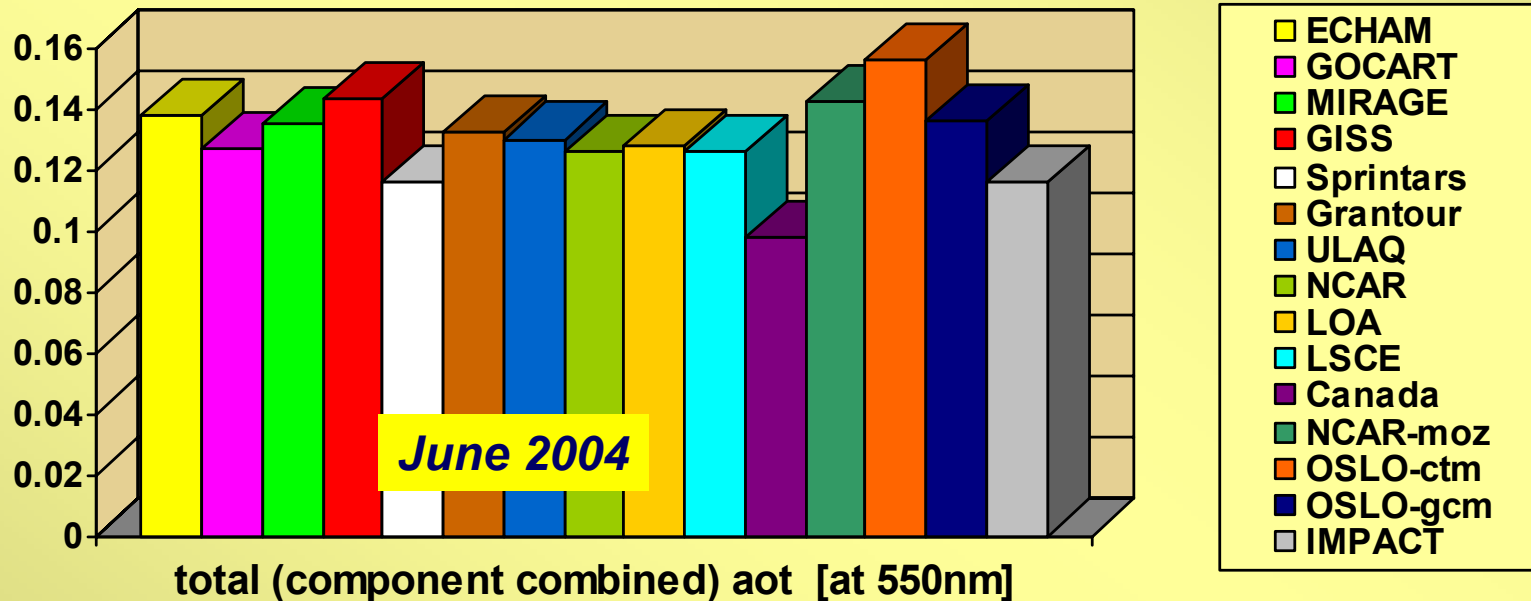
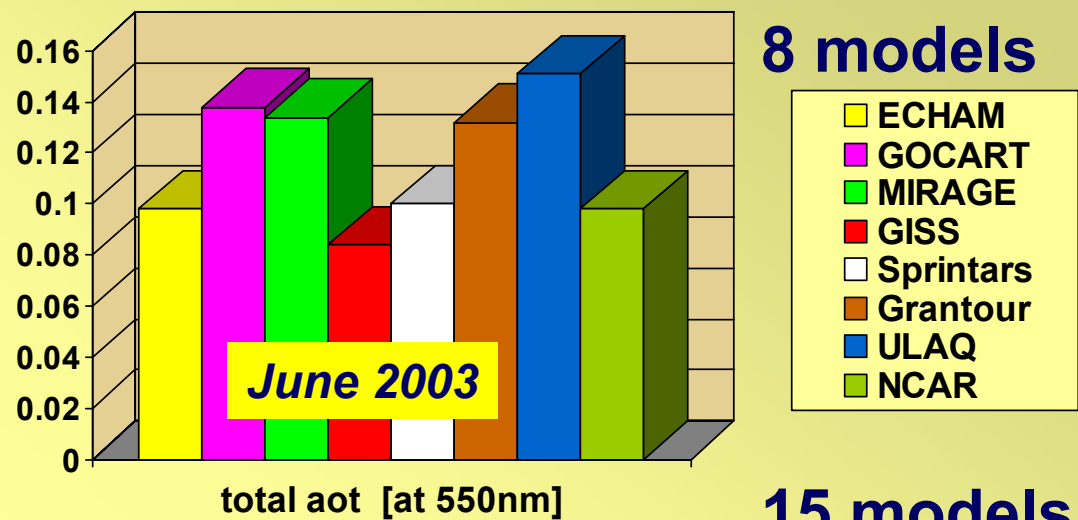


su-sulfate, ss-seasalt, du-dust, oc-org.carb, bc-soot

- one bad error is sufficient to destroy a good effort
- there are always way to ‘adjust’ to the globally (annual) averaged aot of satellites

AOT

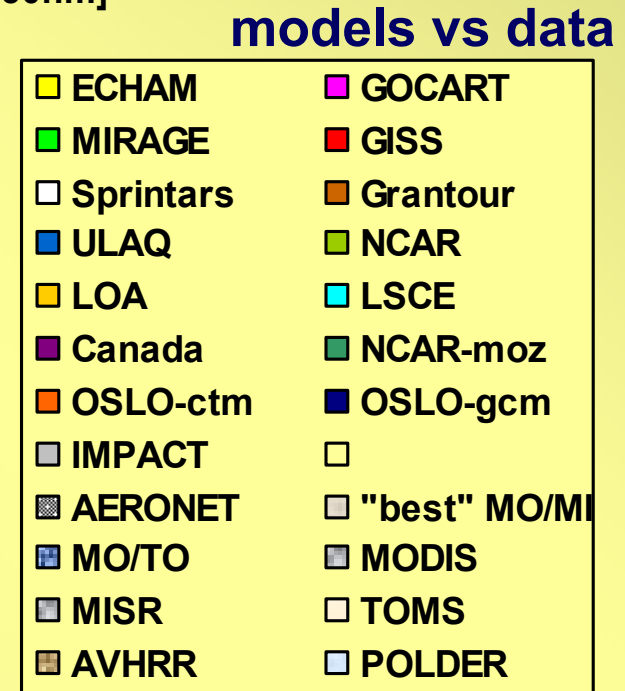
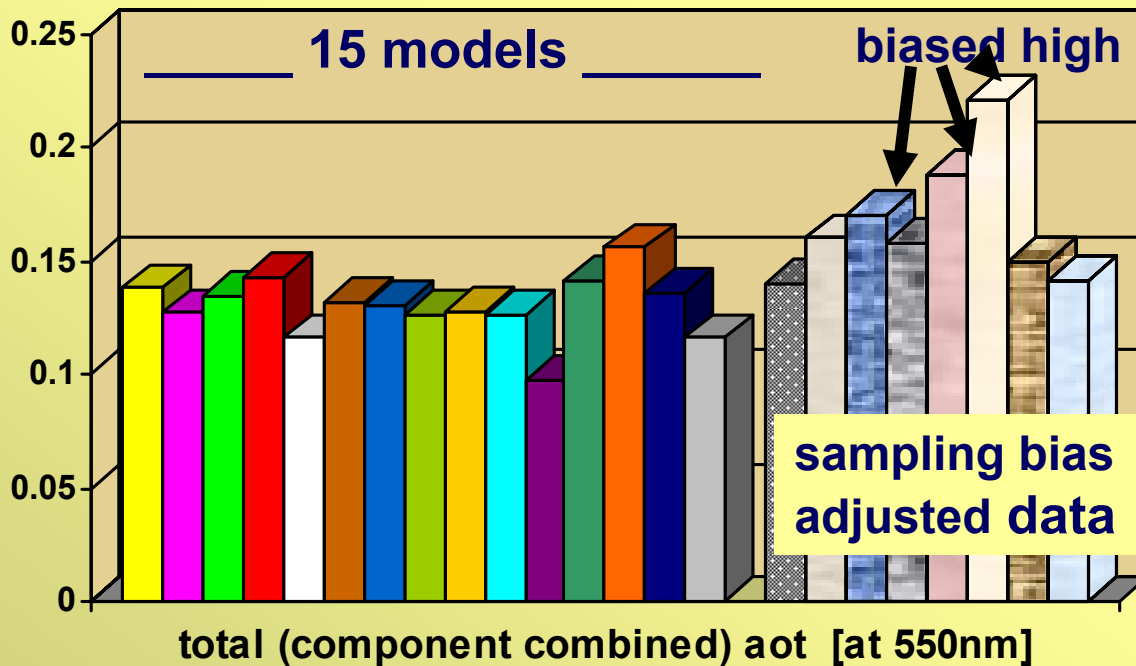
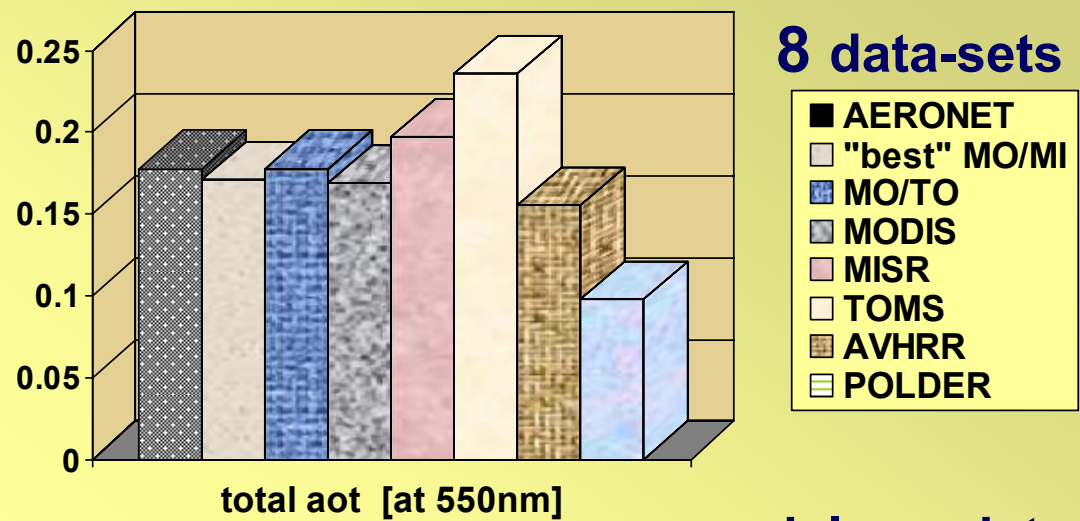
annual global average



since the last year: more component models have appeared
 models seem to converge towards one annual global average

AOT

annual global average



this agreement is encouraging – are we making progress?

quantify global uncertainty

- max/min factors of **15** (* 13 'no extremes') models with aerosol component modules
- different min/max factors for aot and mass demonstrate MEE-differences
- these are still global annual averages!

	mass max/min	mass* max/min	aot max/min	aot * max/min
SU	2.4	1.9	3.8	2.1
BC	3.3	2.1	11	3.2
OC	3.5	1.5	4.0	2.1
DU	14	5.5	8.8	4.1
SS	6.0	2.6	7.4	3.6
TO	2.6	1.9	2.6	1.9
ffrac*	3.0	2.1	2.3	1.5

* ffrac: fine mode (sizes >1 μ m) fraction

dust and sea-salt are associated with largest disagreements
 good agreement for OC surprises ("if uncertain, look what others do")

dry mass

optical depth

mass ext. eff.

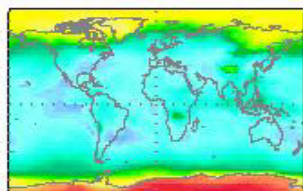
Max/Min factors

Uncertainty

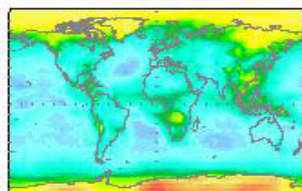
no extremes

SU

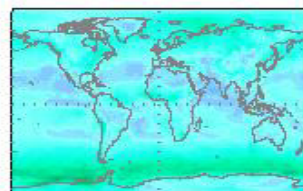
Ms



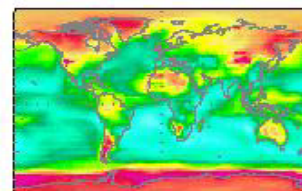
As



Rs



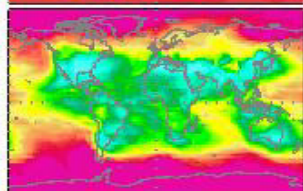
Mf



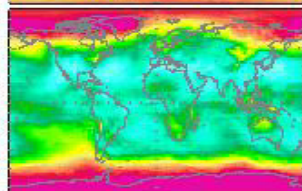
mass fract. >1 μm aot

OC

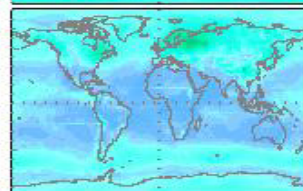
Mo



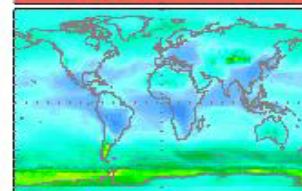
Ao



Ro

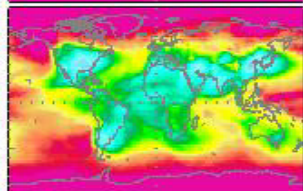


Af

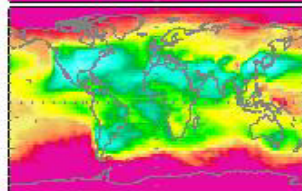


BC

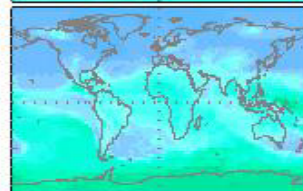
Mb



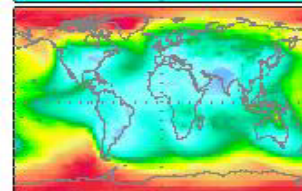
Ab



Rb



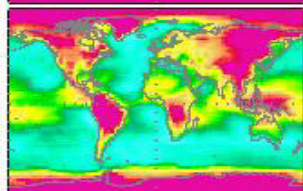
Mr



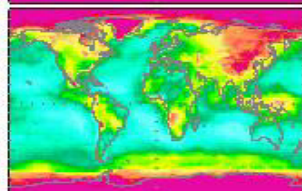
mass BC/OC

SS

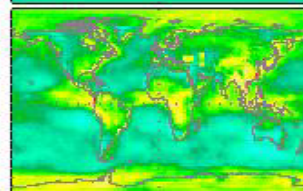
Mn



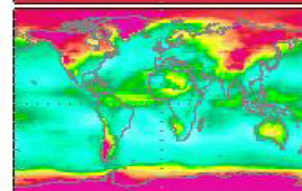
An



Rn



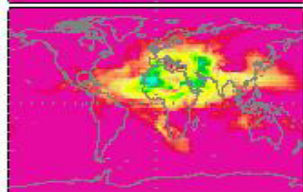
Mt



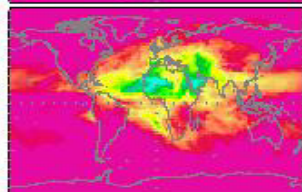
mass total aot

DU

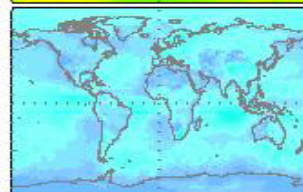
Md



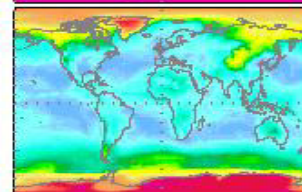
Ad



Rd



At



0,00

max/mín factor (15 models)

20,00

aerosol modeling uncertainty (of 13 models, without the 2 extremes) illustrate the need for regional (and seasonal) assessments

let us take a break

- **large differences among models discourage**
 - who is going to believe any aerosol module?
- **are there any data that can identify skill?**
 - generally not at the required detail
 - if yes - how accurate are the data?
 - **aerosol optical depth**
 - compare on a regional basis to a MODIS / MISR retrieval composite (*possibly currently the most accurate data-set*)
 - **aerosol absorption**
 - compare mass weighted imaginary parts (a measure of absorption) to results of AERONET sky-data inversions

Models:

- ECHAM5
- Grantour
- MIRAGE

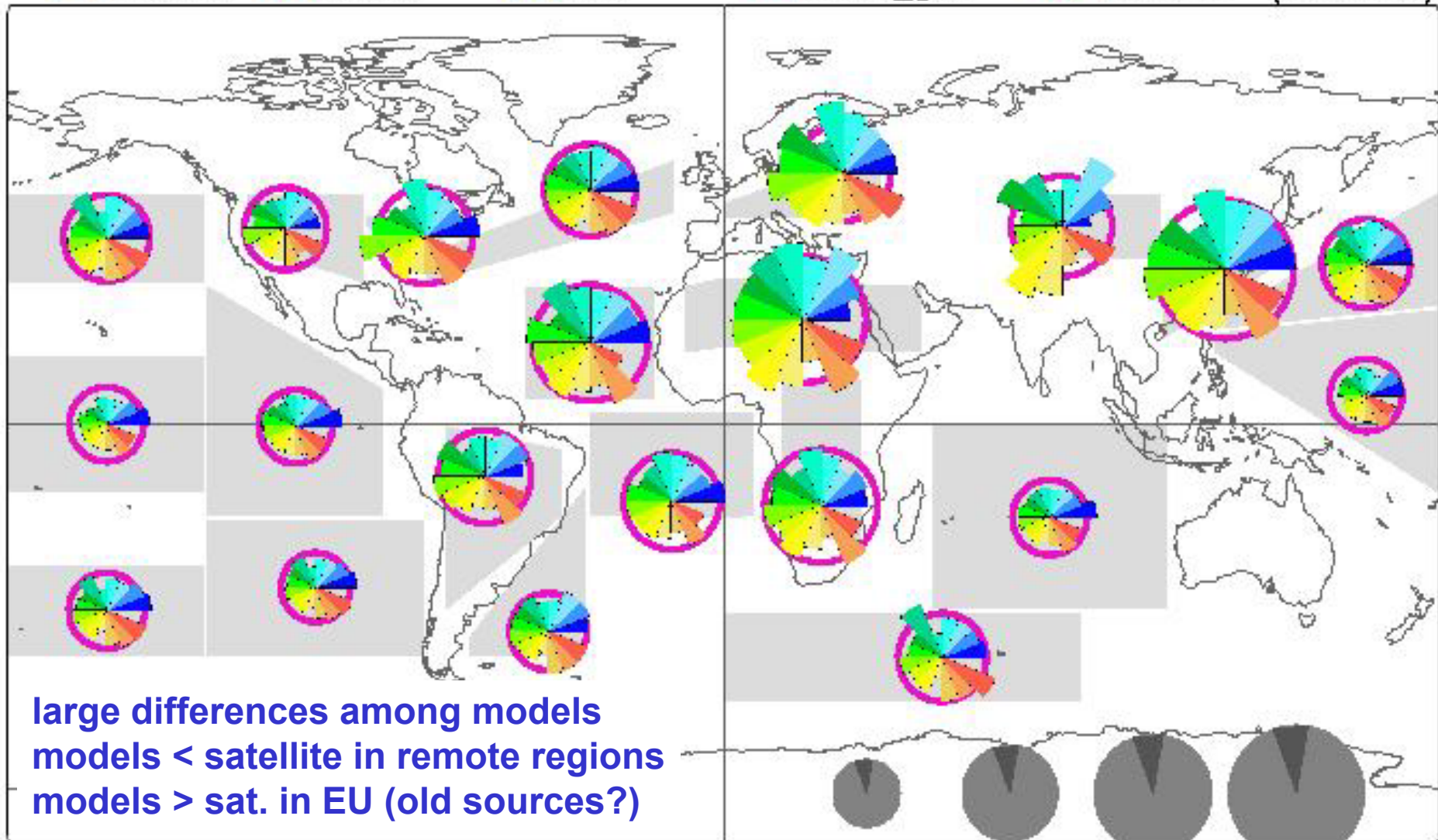
- GOCART
- SPRINT
- GISS
- ULAQ

- NCAR_mat
- NCAR_moz
- IMPACT
- LOA

- LSCE
- CANADA
- OSLO_ctm
- OSLO_gcm

AOT

MO/MI (satellite)



yearly

AOT (550nm)

0.1 0.2 0.3 0.4

Models:

- ECHAM5
- Grantour
- MIRAGE

- GOCART
- SPRINT
- GISS
- ULAQ

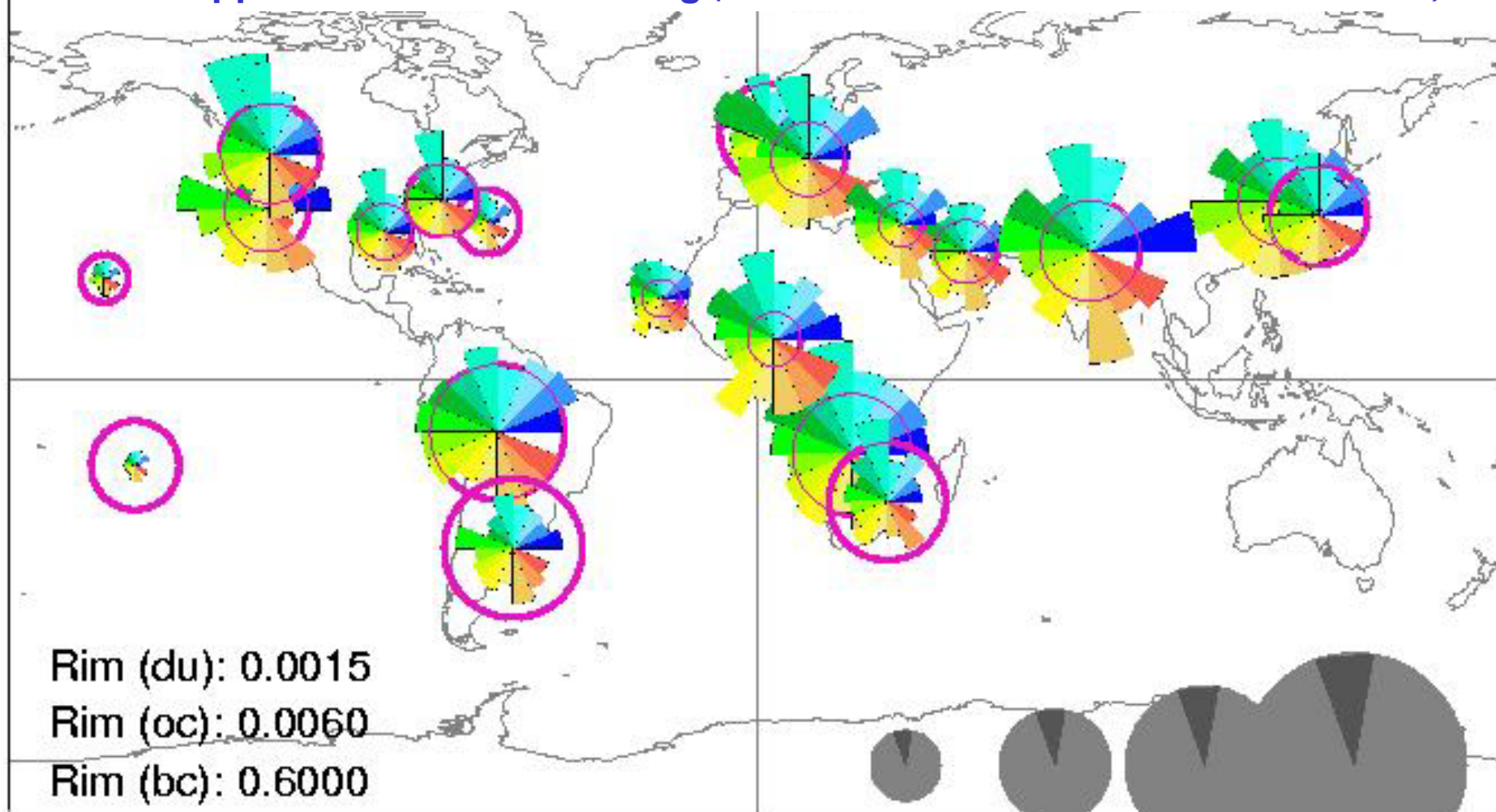
- NCAR_mat
- NCAR_moz
- IMPACT
- LOA

- LSCE
- CANADA
- OSLO_ctm
- OSLO_gcm

Ri wet

○ A-NET (CIMEL)

models appear to be too absorbing (here ECHAM water mass was added to all models)



Rim (du): 0.0015
 Rim (oc): 0.0060
 Rim (bc): 0.6000

yearly

Ri wet (abs)

.004 .010 .020 .040

we have a modeling problem !

- **why these differences ?**
 - input (*emission data, meteorology*)
 - aerosol processing! (*clouds, chemistry, transport*)
 - assumptions (*size, water uptake*) ... lack of data
- **what to do?**
 - acquire quality data (*determine data accuracy*)
 - diagnose models (*comparisons to data*)
 - assure comparability (*same input*)

... in short **AeroCom**

AeroCom

an initiative of MPI and LSCE

- **AeroCom** <http://nansen.ipsl.jussieu.fr/Aerocom>
 - **validate against quality data!**
 - surface concentrations (IMPROVE, EMEP, GAW)
 - surface remote sensing (AERONET, EARLINET)
 - remote sensing from space (MODIS, MISR)
 - **15+ groups participate so far**
 - A: ‘best as you can’ – simulation
 - B: yr 2000 simulation with prescribed emissions
 - C: yr 2000 simulation with pre-industrial emissions – to address anthropogenic ‘forcing’

AeroCom activities

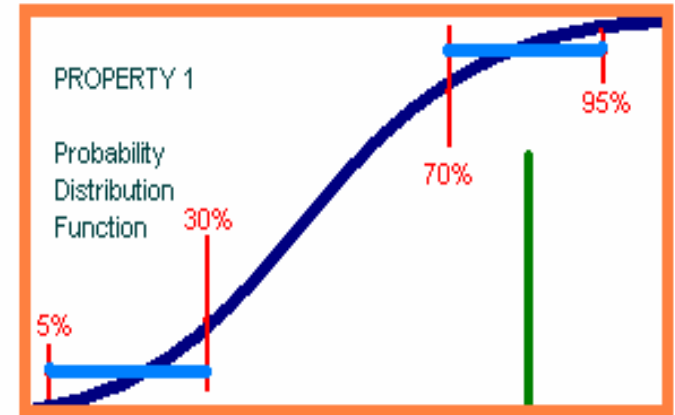
- **acquire and establish quality data ref.**
- **diagnose models** (eliminate weak components) **for a more harmonized model behavior**
- **test modeled cloud-aerosol interactions** (processing, indirect) **to observed correlations**
- **provide ‘more certain’ forcings for IPCC**
- **enhance model and data group contacts**
 - **regular meetings: NY-Dec04, Oslo-Jun05, ...**

2 way- correlations

aerosol and cloud interactions

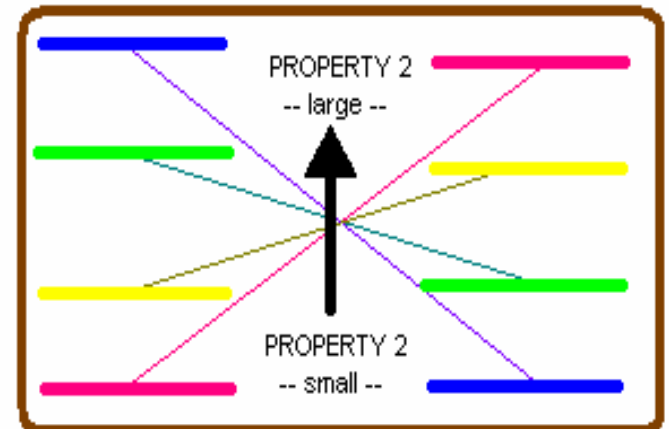
- A. pick a pair of co-located data-sets
- B. rank data of the reference property
- C. determine data averages of the reference property falling into the 5-30% and 70-95% PDF ranges
- C. determine range associated data averages of the second property
- D. determine correlation:
+ slopes **agree**, - slopes **disagree**
- E. determine correlation strength:
use normalized slope steepness
- F. repeat - by exchanging properties

cumulative PDF of reference property



collect property 2 values
associated with property 1

then ... compare PDF-bin
associated averages

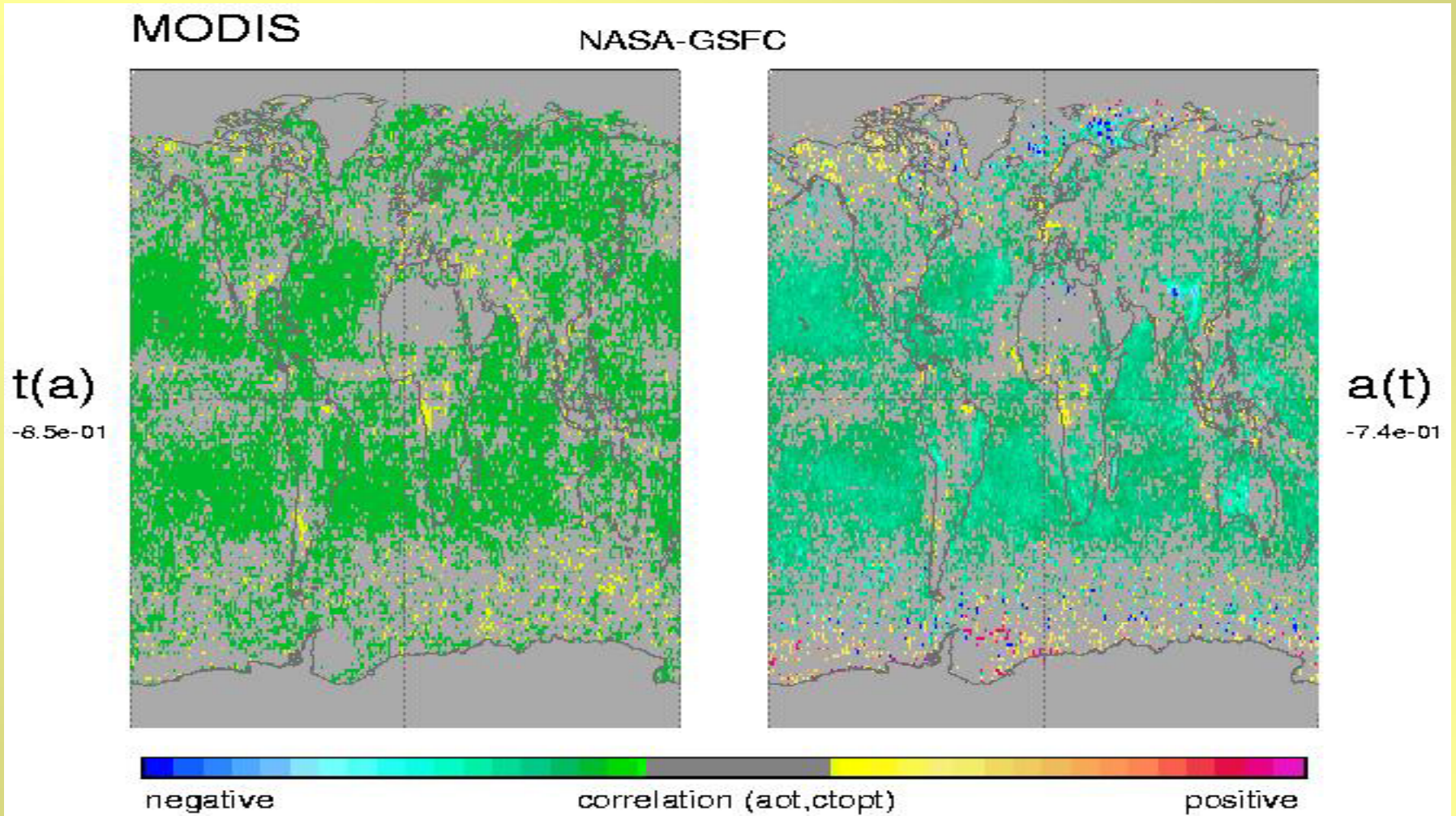


CORRELATION

strong negative
weak negative

strong positive
weak positive

aerosol - cloud

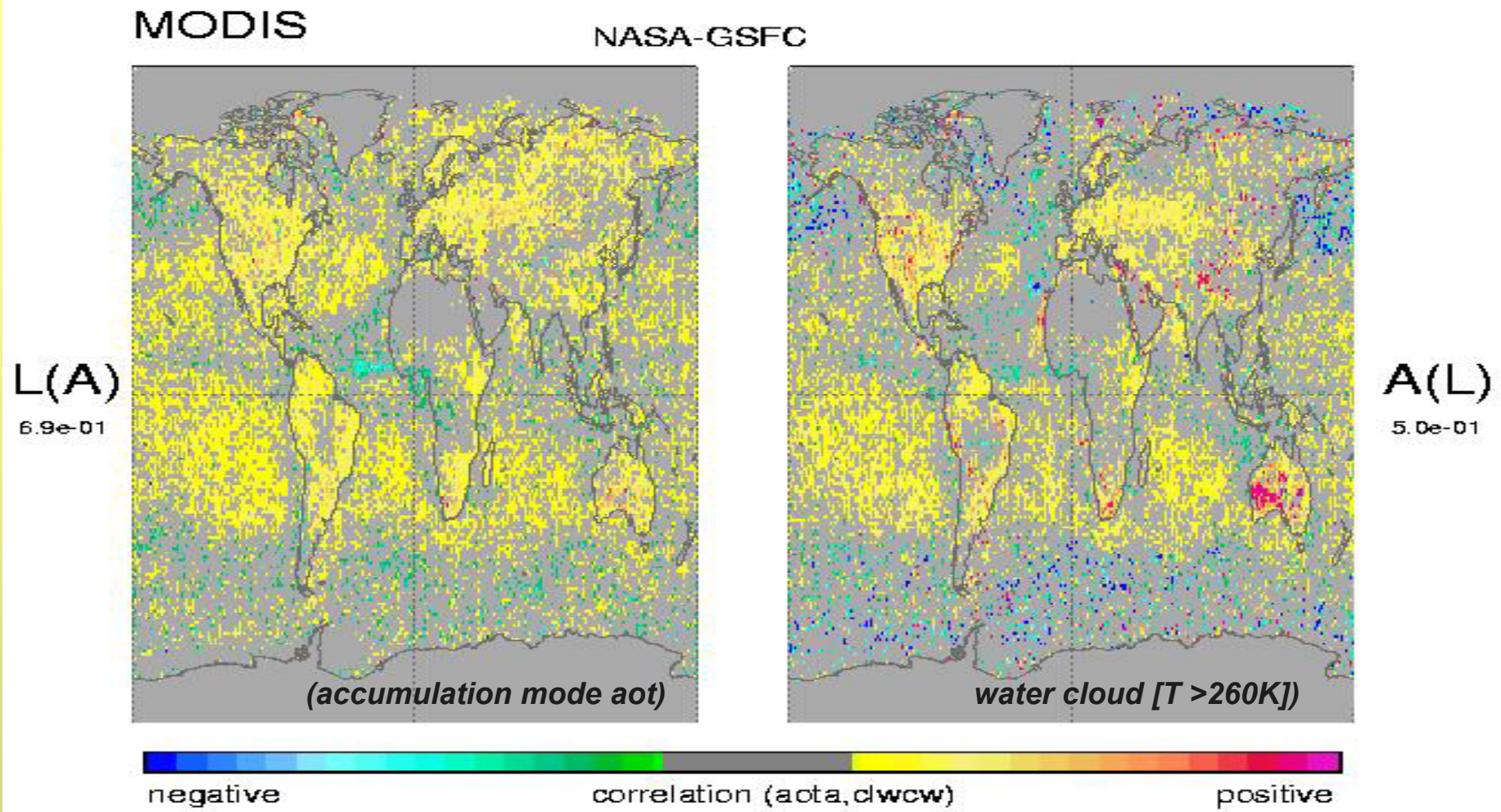


- aerosol optical depth (A) – cloud top temp. (t)

predominantly anti-correlated *as higher cloud top reduce (IR) radiation losses to space ... the expected solar albedo losses are to be partly compensated!*

anti-correlation is stronger with respect to changes in top temperature (right)

aerosol - cloud

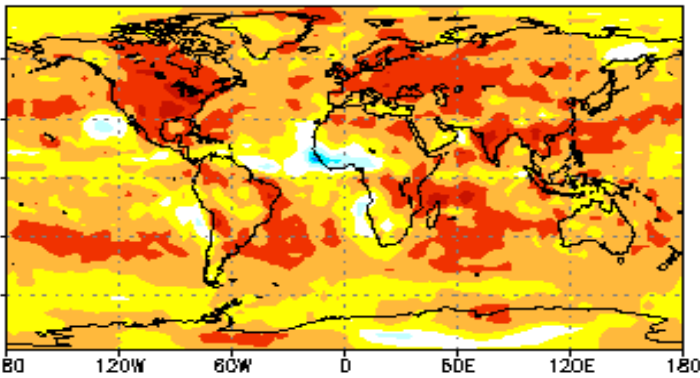


- **aerosol optical depth (A) – cloud liquid water (L)**
more specific choices can lead to stronger signals at the expense, explanations will remain a challenge and reasons for (anti-) and correlations are offered:

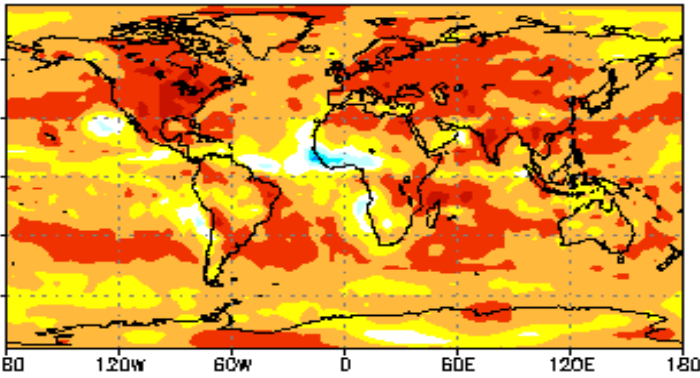
higher altitude dust signal disappears

land signal increases (+ lifetime?)

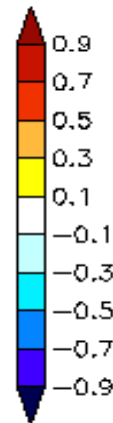
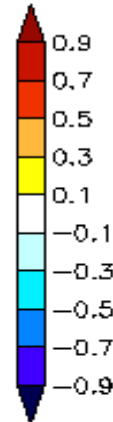
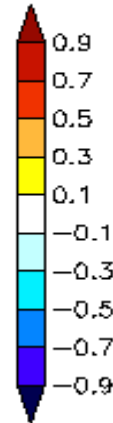
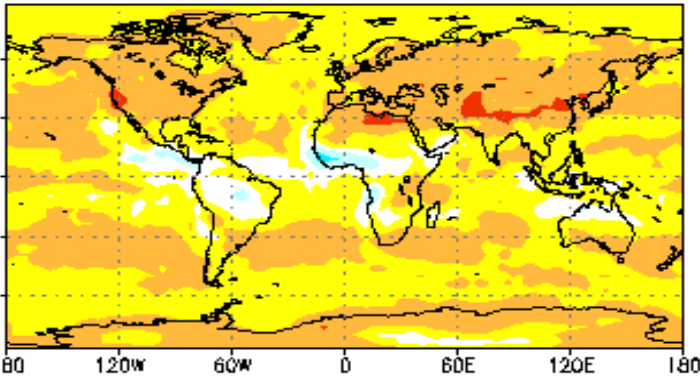
Correlation AOT-LWP ($r=0.37$)



Correlation AOT-TWP ($r=0.40$)

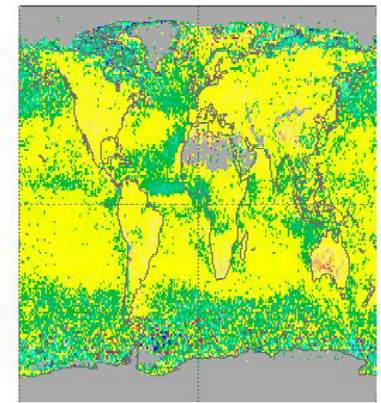
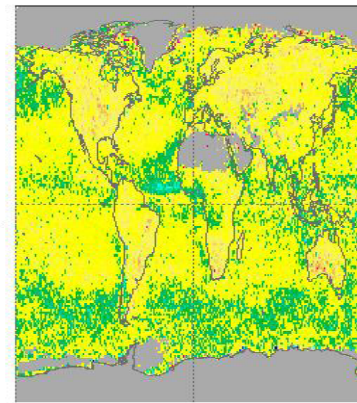


Correlation AOT-CC ($r=0.26$)



MODIS

NASA-GSFC



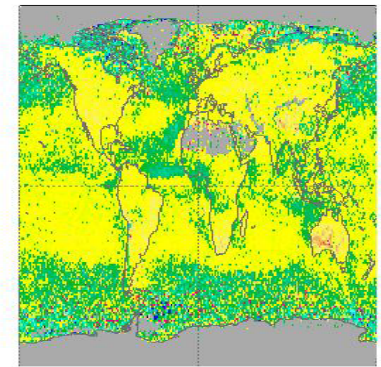
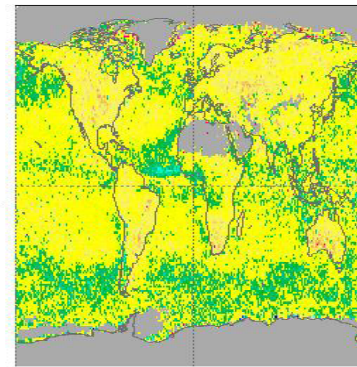
$I(a)$
 $5.5e-01$

$a(l)$
 $3.0e-01$

negative correlation (aot,clwc) positive

MODIS

NASA-GSFC

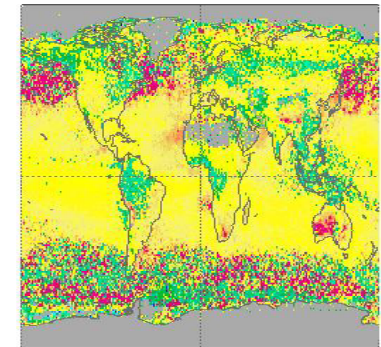
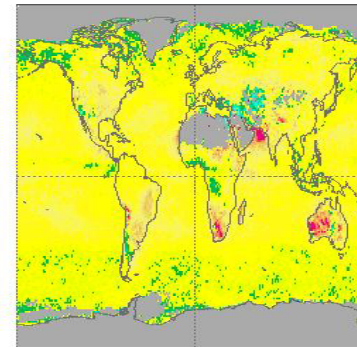


$I(a)$
 $5.5e-01$

$a(l)$
 $3.0e-01$

MODIS

NASA-GSFC



$f(a)$
 $6.7e-01$

$a(f)$
 $5.6e-01$

negative correlation (aot,clrac) positive

Message

- **anthropogenic impact of aerosol on climate needs to be better quantified** (reduce uncertainties)
- **uncertainties in aerosol forcing** (the end product in modeling) **do not represent ‘actual’ uncertainties**
- **model differences at intermediate processing steps and on different scales are much larger**
- **quality data** (*e.g. AERONET*) **can provide at least a few constraints – data synergy helps**
...in turn data can benefit from modeling

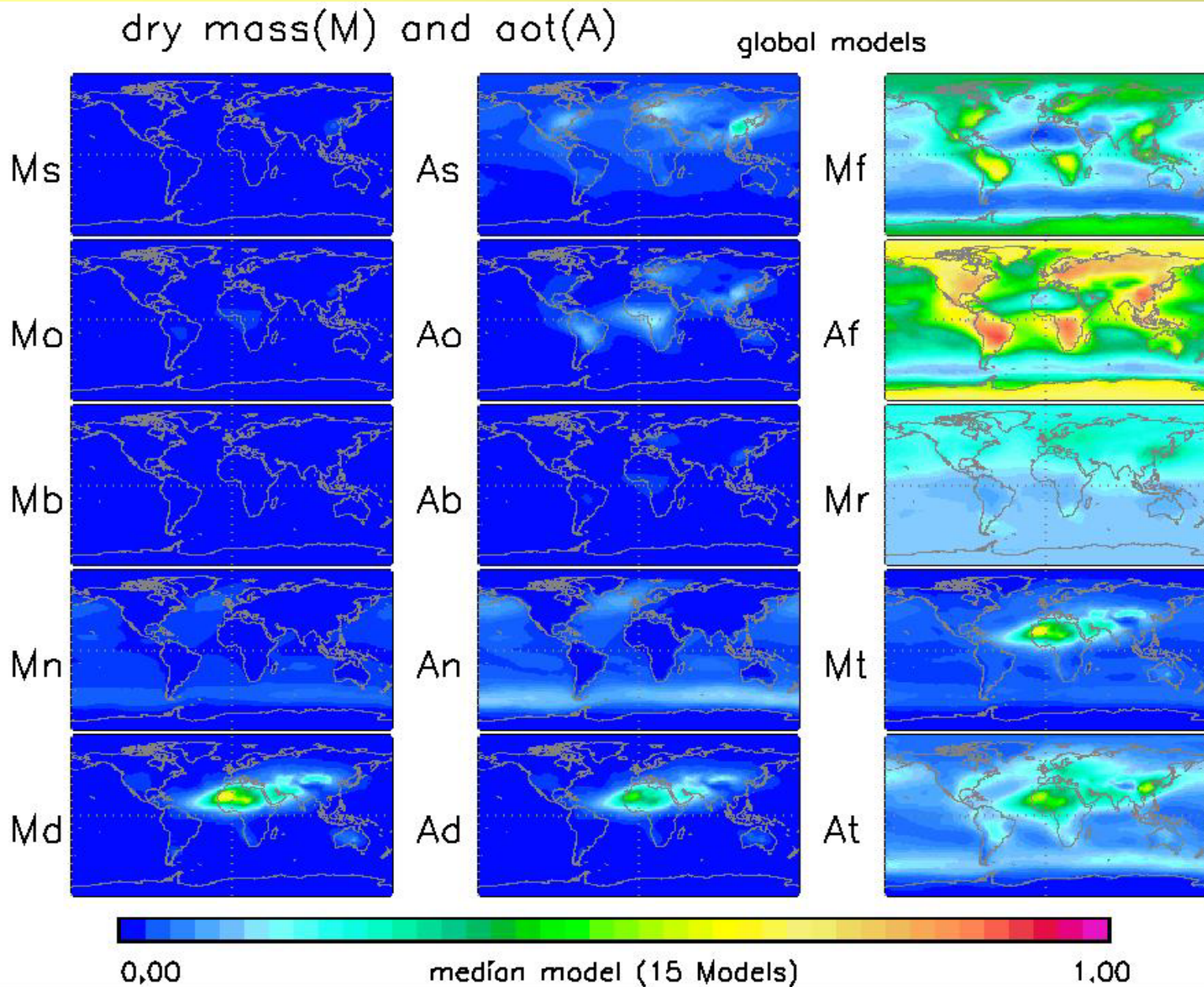
a reference? 'median model'

modeling
in return
can help
complete
data-sets

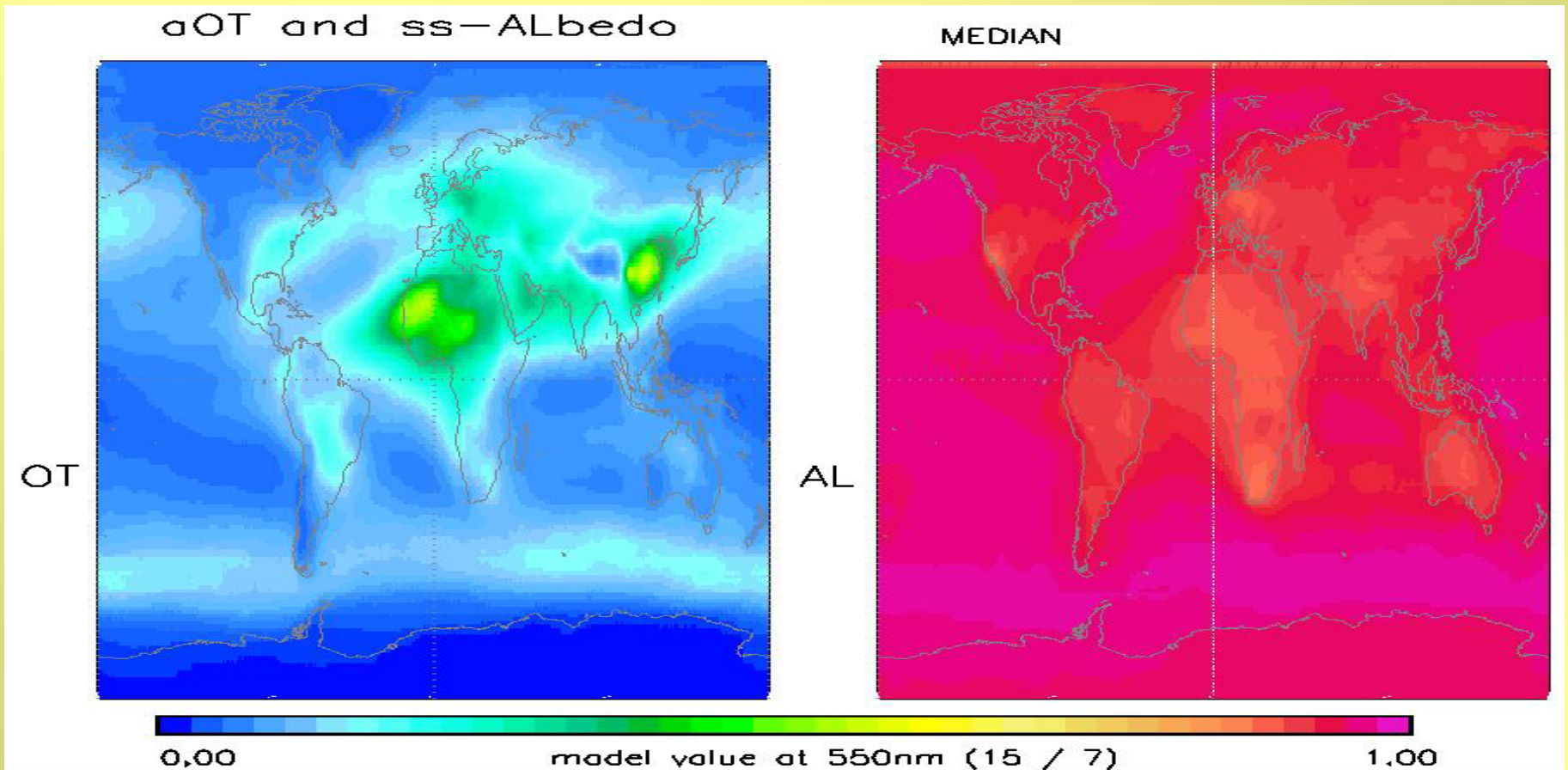
M_ mass
A_ aot

_s sulfate
_o org.carb
_b black c.
_n seasalt
_d dust

-t (s,o,b,n,d)
-f (s,o,b)
-r o/b-ratio



median *aot* (aerosol optical depth) and
 median *SSa* (single scattering albedo)



essential aerosol optical properties when determining the aerosol forcing