

Modeling Tropospheric Chemistry: 0D to 3D models

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Fundamentals of Atmospheric Chemistry and Aerosol Modeling
Aug 13-15, 2018

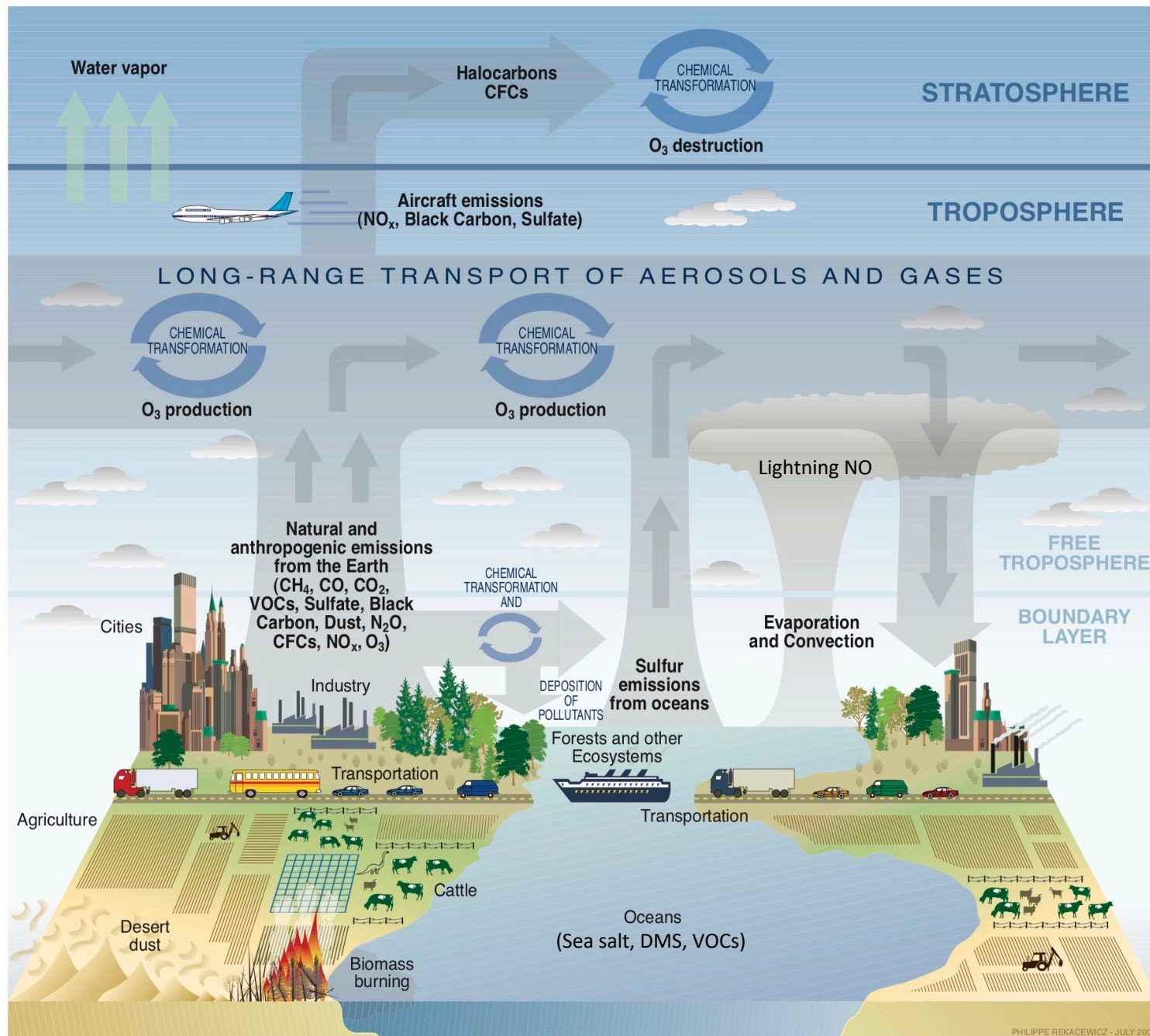


Fig. 3-1 of Climate Change Science Program Strategic Plan, 2003

Some regions, e.g. the Arctic, require unique considerations ...

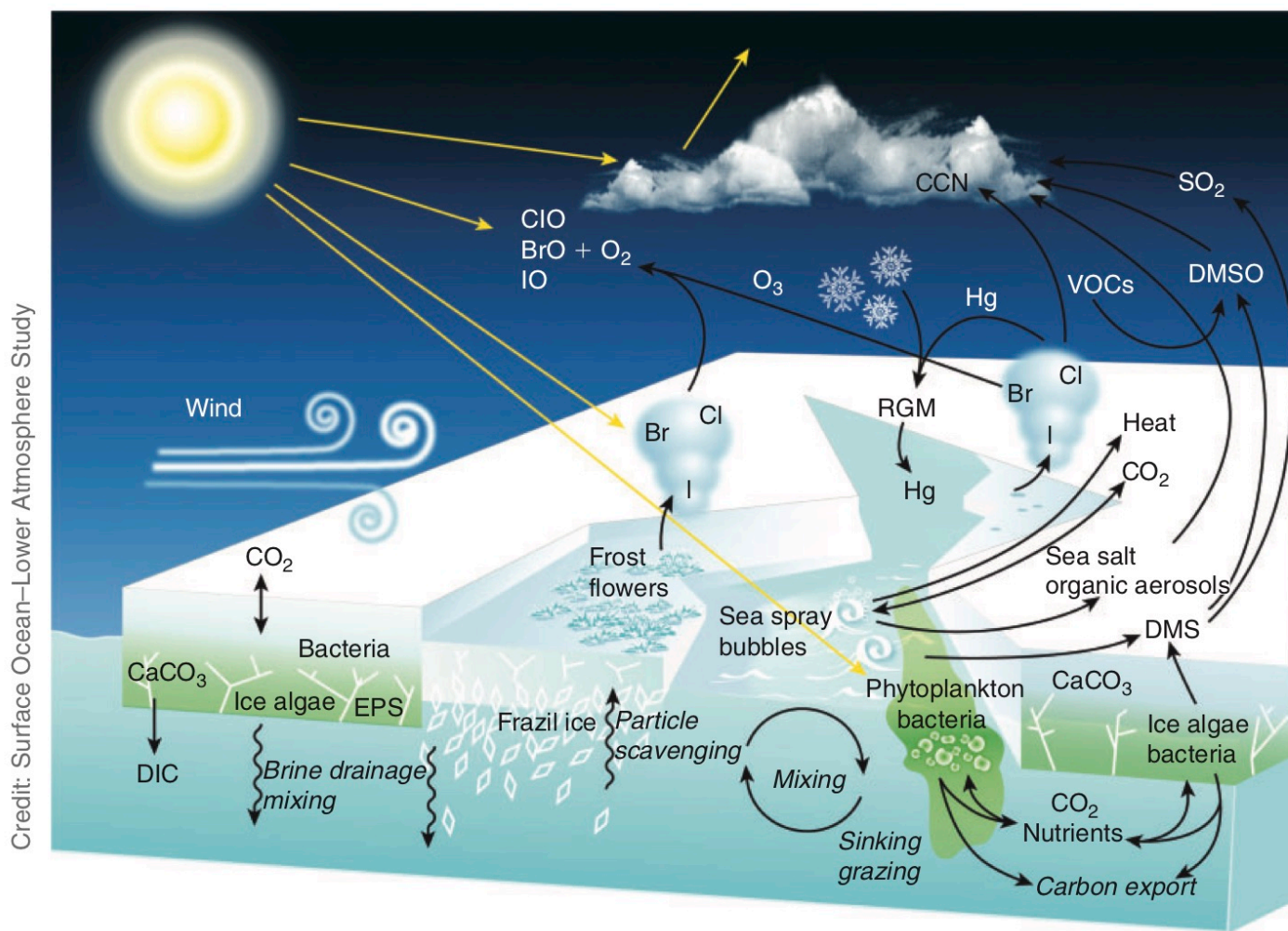
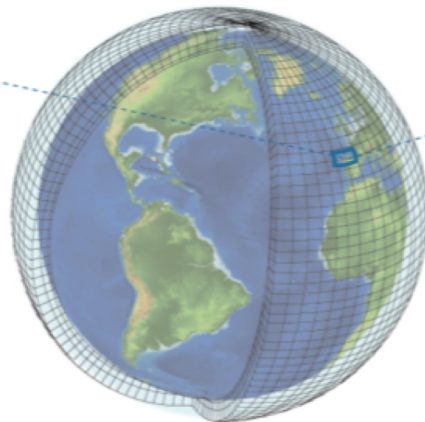
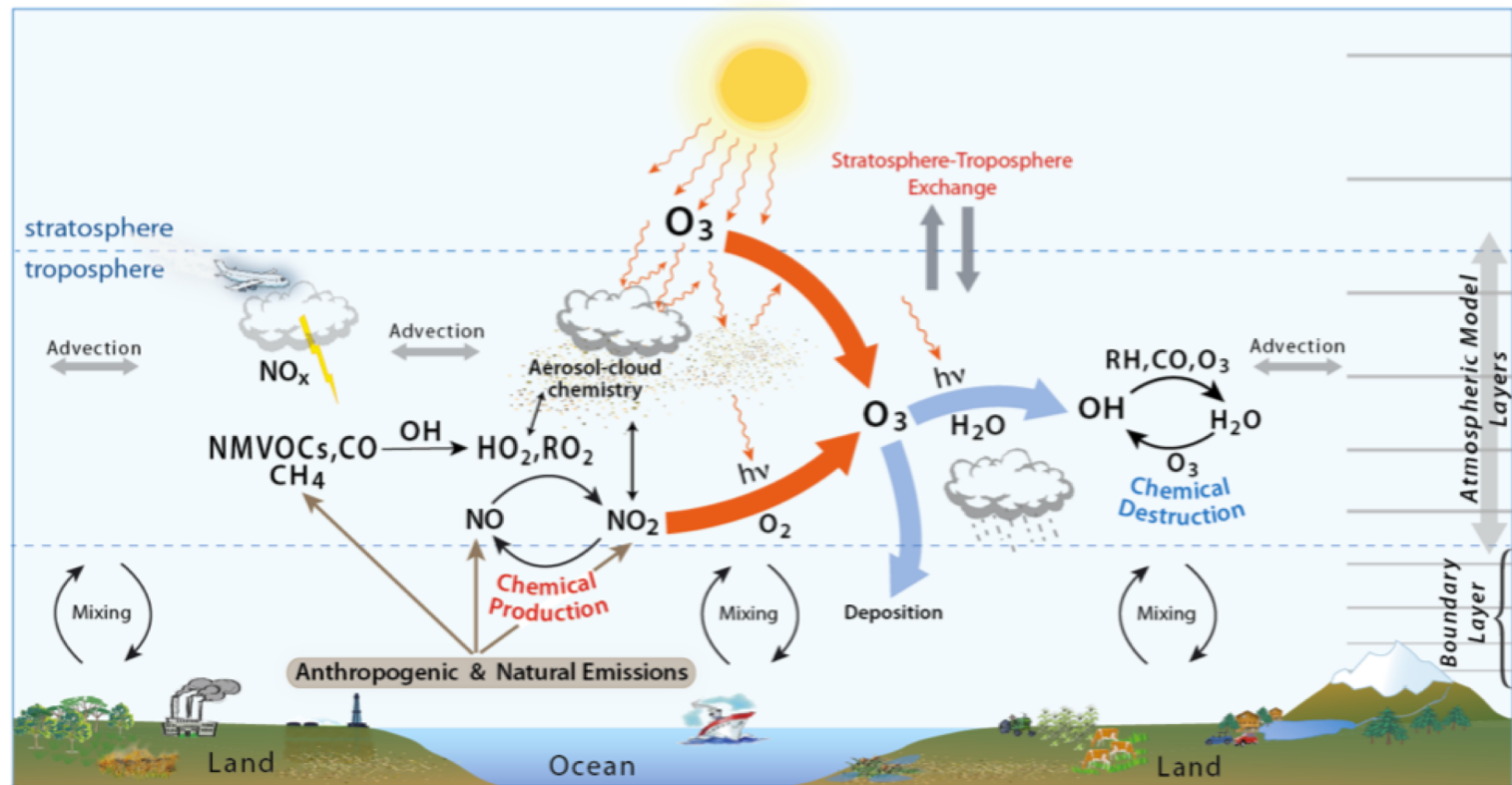


Fig. 3. Conceptual diagram illustrating the main issues, processes and species relating to the SOLAS Sea Ice strategy (after Shepson et al.^[54]). Processes are indicated in *italic*. (DIC, dissolved inorganic carbon; EPS, extracellular polymeric substances; CCN, cloud condensation nuclei; RGM, reactive gaseous mercury; VOCs, volatile organic compounds; DMS, dimethyl sulfide; DMSO, dimethyl sulfoxide.)

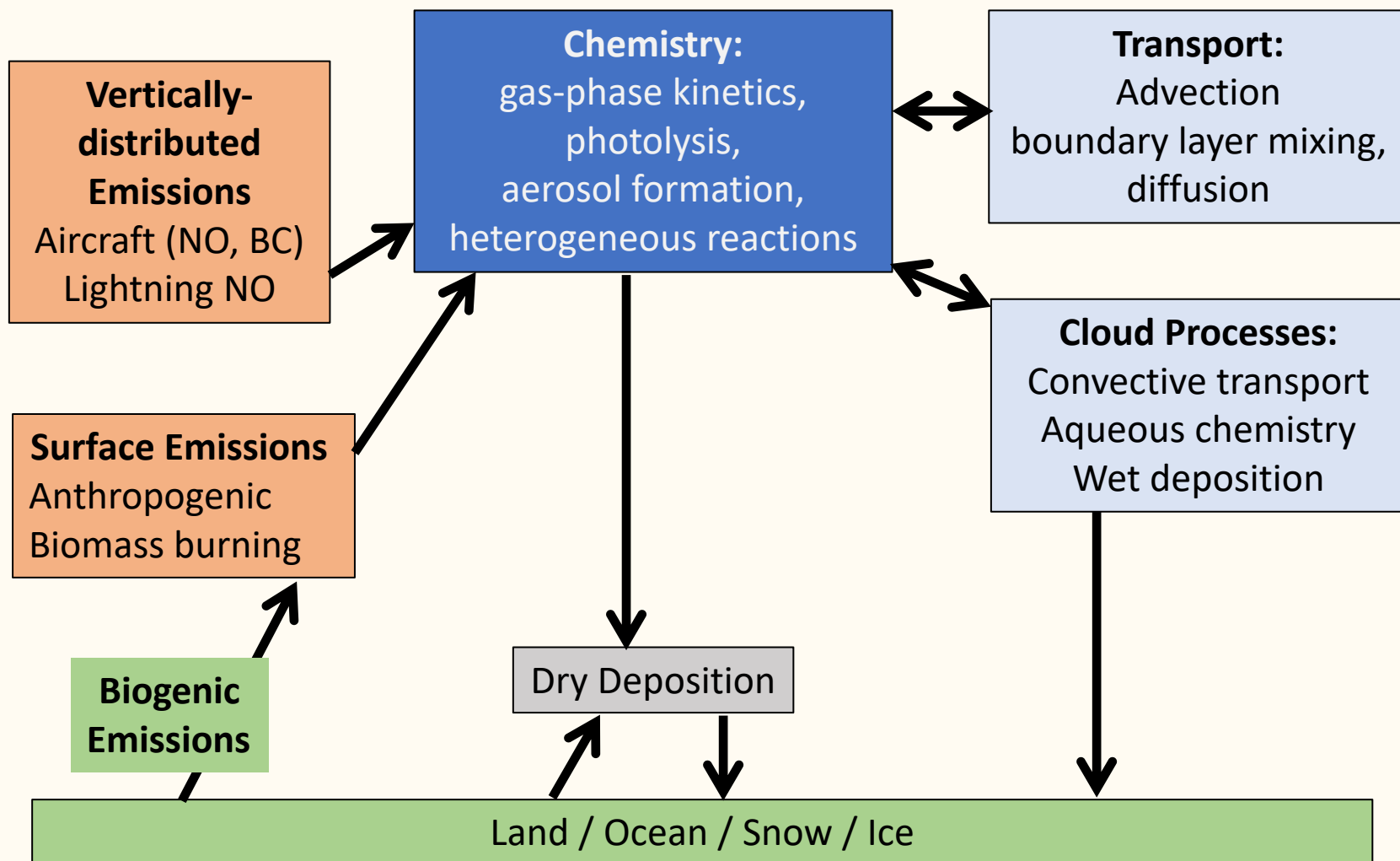
C.S. Law et al., Environ. Chem., 2013

Simulating tropospheric ozone



P. Young et al., *Elementa*, 2017

Computer models treat each process as a separate module



Solution for each chemical species i

$$\frac{\partial c(i)}{\partial t} = \text{Production}(i) - \text{Loss}(i) = E_i + C_i + A_i + T_i + W_i + D_i$$

For each compound, at each timestep, the change in concentration is the sum of the change in concentration for each process:

E_i : Emissions

C_i : Gas-phase-Chemistry

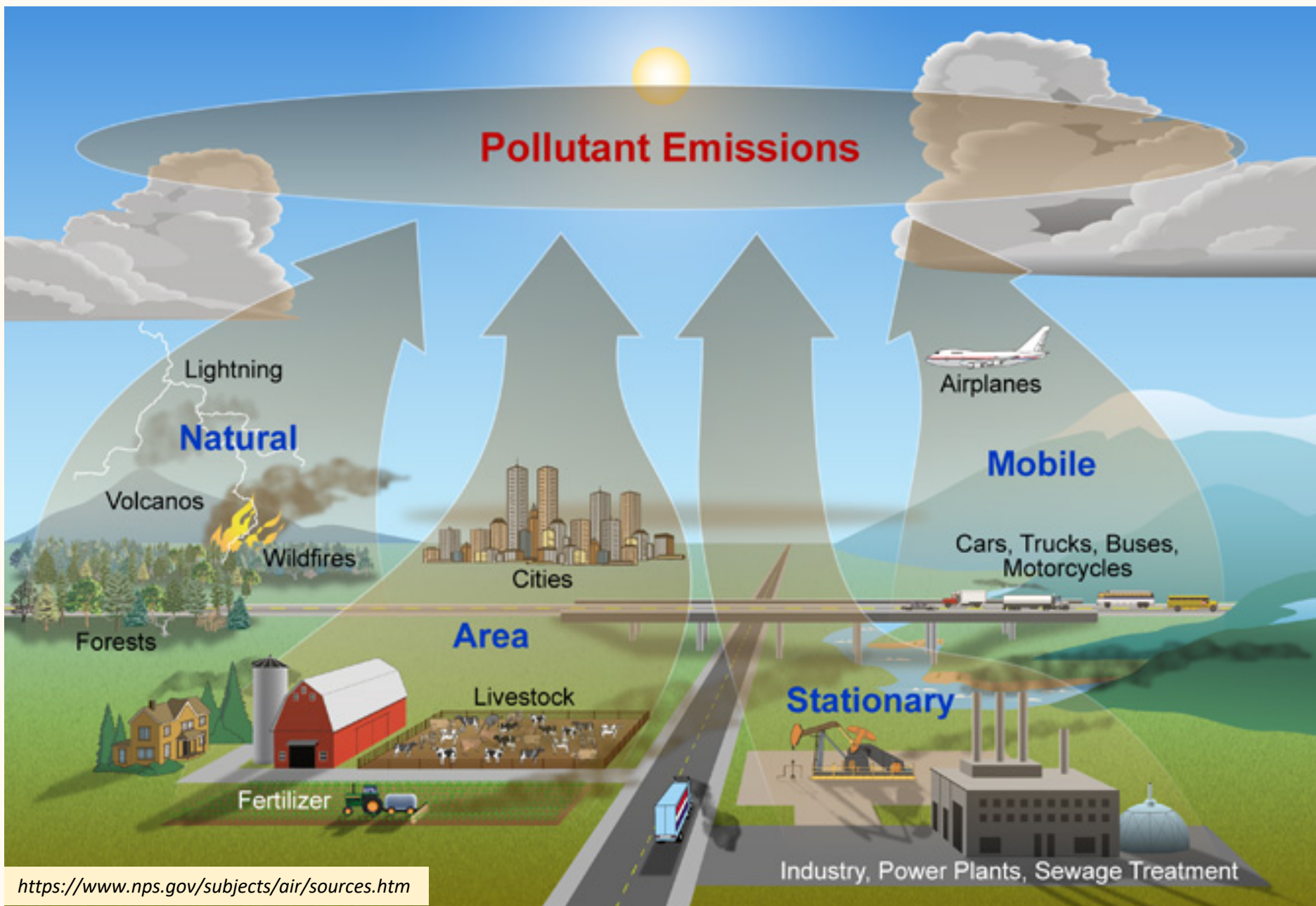
A_i : Aerosol-processes

T_i : Advection + Diffusion

W_i : Cloud-processes (wet deposition)

D_i : Dry deposition

For compounds with short lifetimes the order of operators can affect results



<https://www.nps.gov/subjects/air/sources.htm>

Anthropogenic emissions

- Power plants (NO_x , SO_2)
- Industry (solvents – VOCs)
- Traffic (NO_x , CO, PM)
- Residential heating (CO, VOCs, PM)
- Agriculture (NH_3)
- Shipping (NO_x , HCs)
- Aviation (NO_x)

Many emissions inventories at: <http://eccad.aeris-data.fr/>

ECCAD: Emissions of atmospheric Compounds and Compilation of Ancillary Data

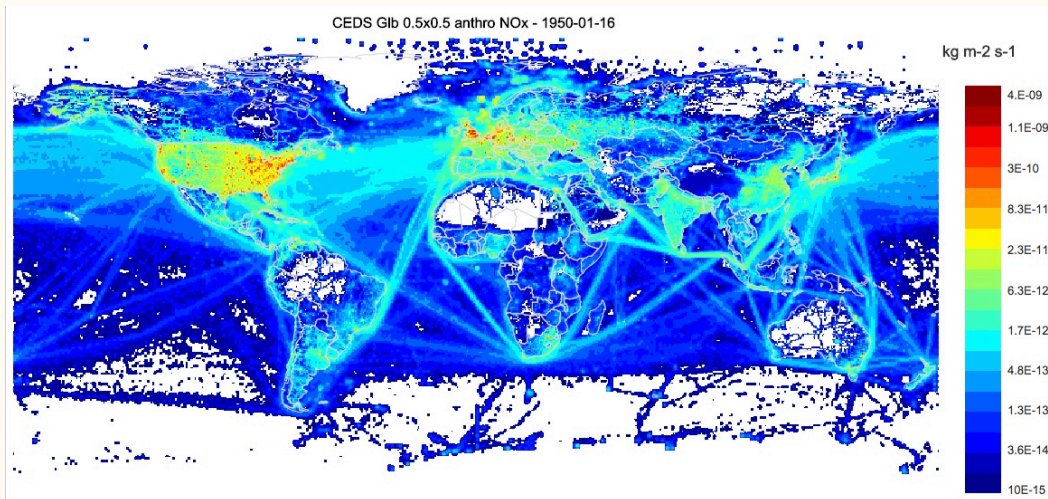
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Sign in

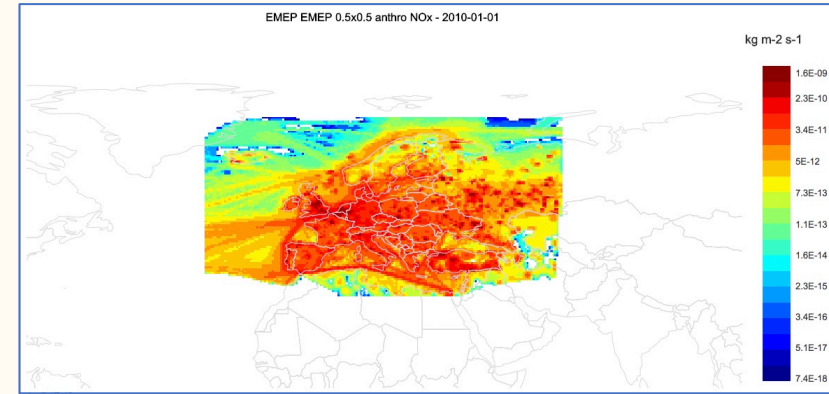
Title▲▼	Categories▲▼	Temporal coverage▲▼	Time resolution▲▼	Grid size▲▼	Provider(s)			Sign in
MACCity Global	HYDE1.3 Global	Anthropogenic	1890-1990	Monthly	1°		EDGAR	
RCPs Global	POET Global	Anthropogenic Biomass burning Biogenic Oceanic	1990-2000 1990-2000 1990-2000 1990-2000	Yearly Monthly Monthly Yearly	1°		JRC	
ACCMIP Global								
CAMS-GLOB-ANT Global		Anthropogenic Biomass burning	1984-2018 1984-1990	Yearly Yearly				
CAMS-GLOB-BIO Global	GEIA Global	Biogenic Oceanic Volcanic Lightning	2000-2000 1990-2000 2000-2000 1990-1990	Yearly Yearly Yearly Monthly	1°		GEIA	
CAMS-GLOB-SHIP Global -Pole								
CAMS-GLOB-OCE Global	EDGARv4.tox1 Global	Anthropogenic	1970-2008	Yearly	0.1°		EDGAR	
CAMS-REG-GHG TNO Europe	GFASv1.3 Global	Biomass burning	2003-2016	Daily	0.1°		mac	
CAMS-REG-AP TNO Europe	GFASv1.2 Global	Biomass burning	2003-2015	Daily	0.1°		mac	
HTAPv2 Global	GFED3 Global	Biomass burning	1997-2010	Monthly	0.5°		GFED	
EDGARv4.3.2 Global	GFED4 Global	Biomass burning	1997-2015	Monthly	0.25°		GFED	
EDGARv4.3.1 Global	IS4FIRES Global	Biomass burning	2000-2011	Monthly	0.5°			
EDGARv4.2 Global	DECISO-NOx China South Africa Middle East	EMEP EMEP Europe	Anthropogenic	1980-2020	Yearly	0.5°	emep	
EDGARv4.3.2-monthly Global		REAS1.1 East Asia	Anthropogenic	1980-2020	Monthly	0.5°	REAS	
EDGARv4.3.1-monthly Global	PKU05 Global	MPI-CNRS East Asia 2	Anthropogenic	2008-2016	Yearly	0.25°	MPI-CNRS	
ECLIPSE-GAINS-4a Global	PKU Global	CR2-MMA Chile	Anthropogenic	2016-2016	Yearly	0.01°	(CR)²	
ECLIPSE-GAINS-V5a Global	GICC Global	REAS2.1 East Asia	Anthropogenic	2000-2008	Monthly	0.25°	REAS	
CEDS Global	GUESS-ES Global	IASB-TD-OMI Global	Biomass burning Biogenic	2005-2014 2005-2014	Monthly Monthly	0.5°	IASB-TD-OMI	
RETRO Global	AMMABB Global	IASB-TD-OMI-NCP North China Plain	Anthropogenic	2007-2012	Monthly	0.25°	IASB-TD-OMI	
	MEGAN-MACC Global	MarcoPoloKNMI China	Anthropogenic	2007-2013	Yearly	0.25°	MarcoPoloKNMI	
Junker-Liousse Global	MEGANv2 Global	L14-Africa Africa	Anthropogenic	2005-2030	Decadal	0.25°	L14-Africa	
Andres-CO2-v2016 Global	APIFLAME Charmex	SAFAR-India India	Anthropogenic	1991-2011	Monthly	1°	SAFAR-India	
	TNO-MACCII TNO Europe	DACCIWA Africa	Anthropogenic	1990-2015	Yearly	0.125°	DACCIWA	
		DACCIWA-flaring Africa	Anthropogenic	1994-2015	Yearly	0.125°	DACCIWA	

Anthropogenic NO_x Emissions – various inventories

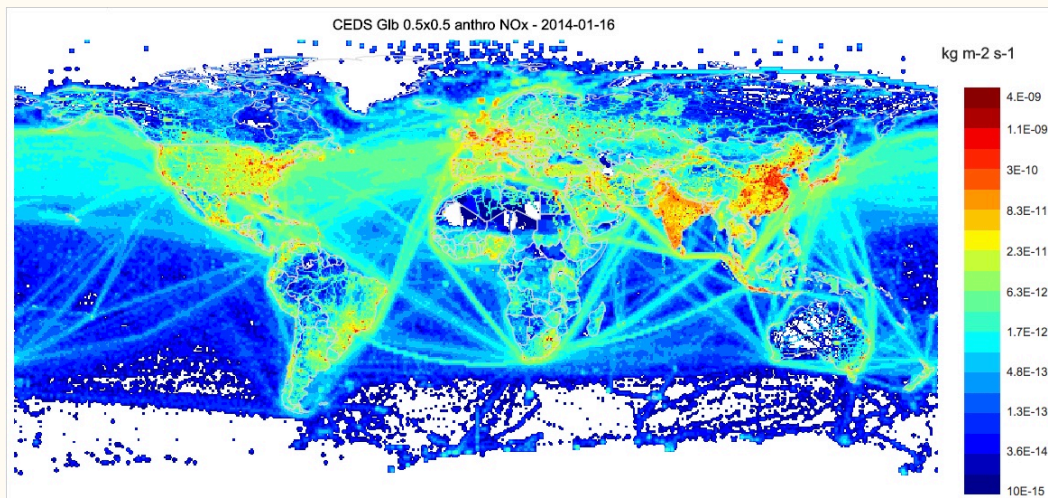
CEDS (0.5deg) - 1950



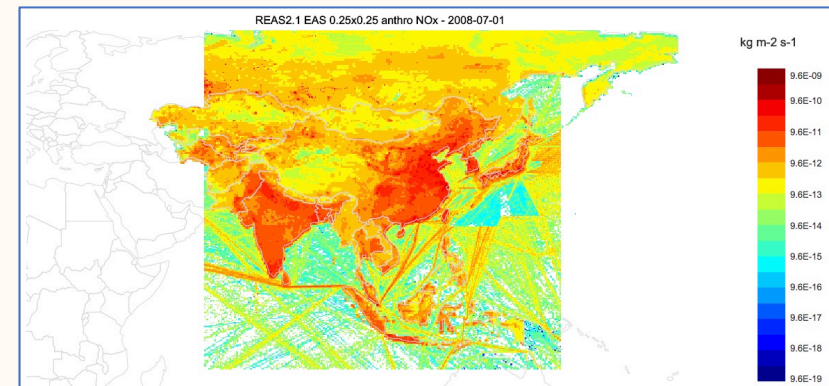
EMEP Europe (0.5 deg) - 2010



CEDS - 2014



REAS2.1 Asia (0.25deg) - 2008



Images from: <http://eccad.aeris-data.fr/>

Large differences between inventories

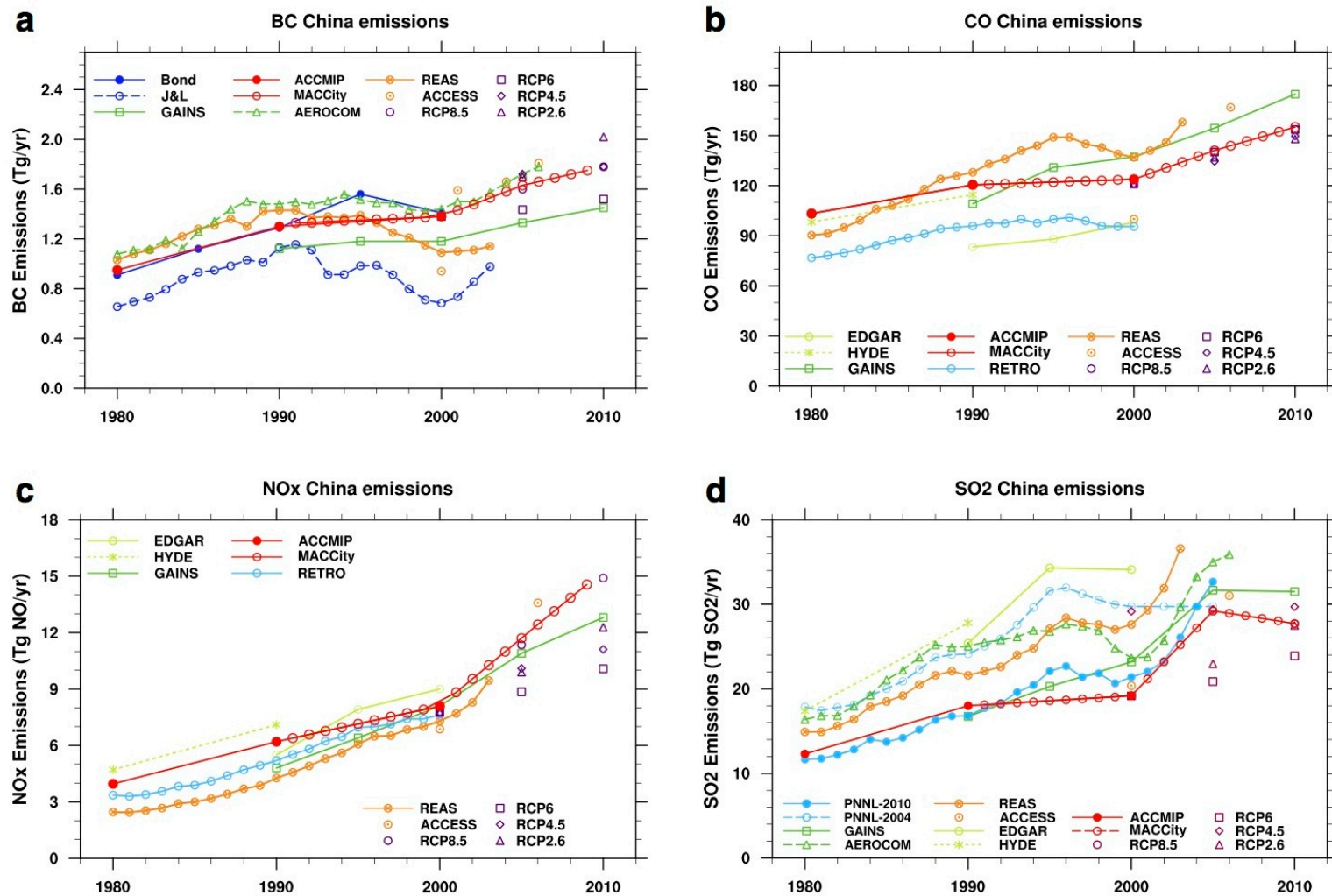


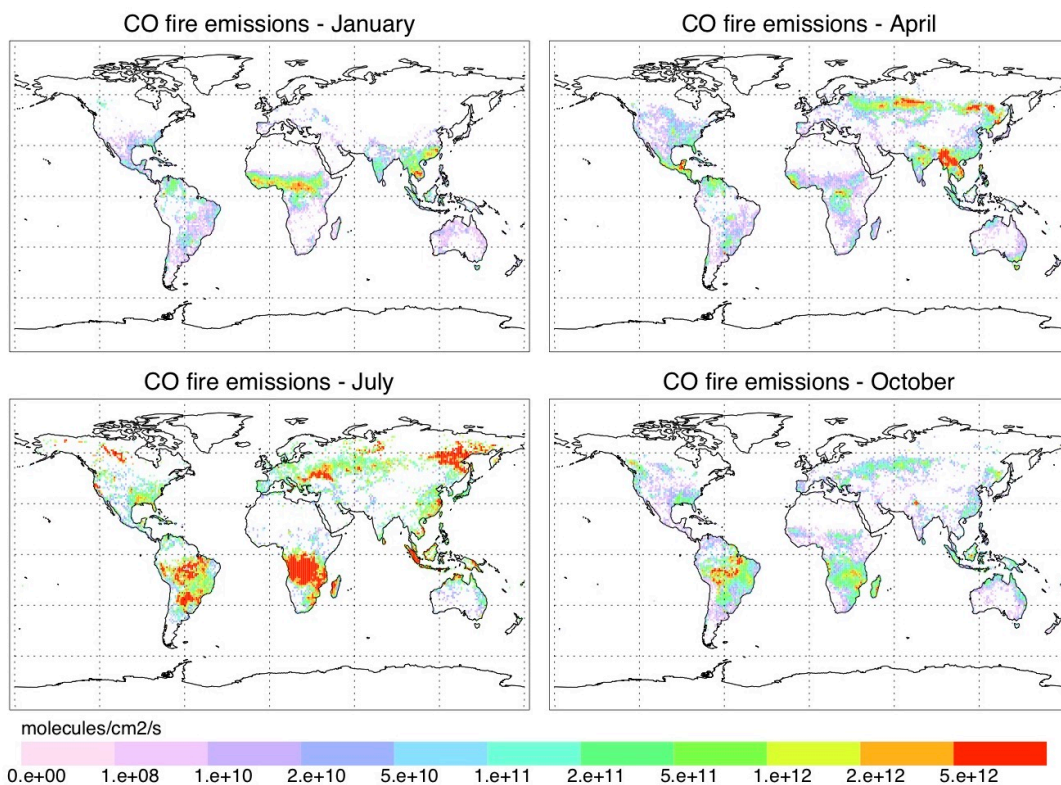
Fig. 4 Emissions of BC (*top left*), CO (*to right*), NO_x (*bottom left*) and SO₂ (*bottom right*) in China from 1980 to 2010

Granier et al., Climatic Change, 2011

Biomass burning

Natural and Anthropogenic

- Wildfires
- Prescribed burns (forest understory)
- Agricultural burning
- Trash burning
- Residential burning of biofuel



Large seasonal variability

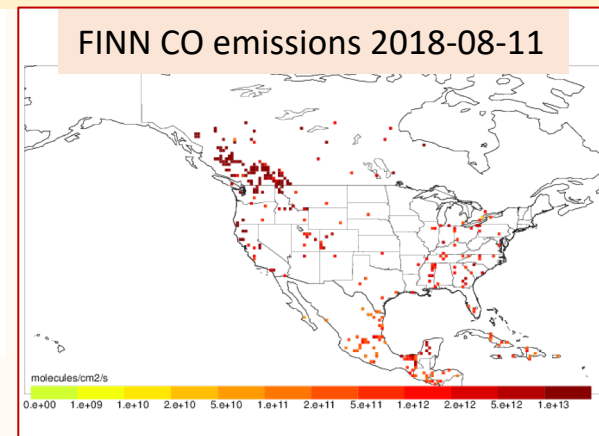
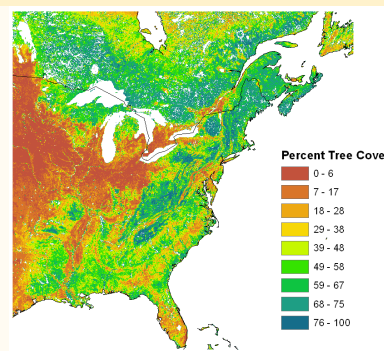
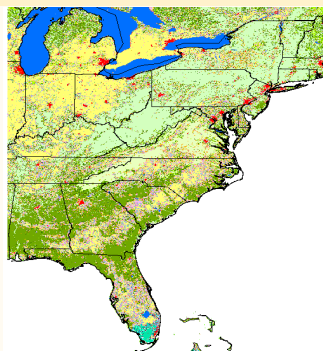
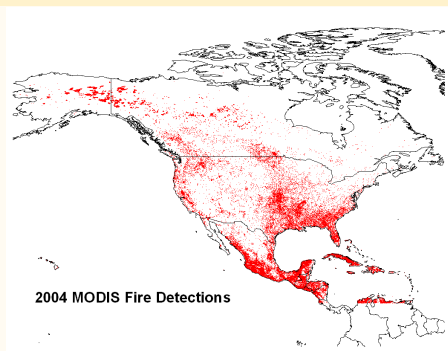
- Natural wet/dry season
- Agricultural activity

Large inter-annual variability
due to climate

Large uncertainties in

- What is burned
- How much is burned
- Emission factors

Fire emissions models (e.g., FINN)



$$\text{Emissions}(i) = f(A(x,t), B(x,t), E_f(i))$$

For compound i , location x , time t :

- $A(x,t)$: Area burned
- $B(x,t)$: Biomass burned (kg-biomass-burned/area)
- E_f : Emission factor for each chemical compound (mass-emission-species/kg-biomass-burned)
 - type of vegetation (ecology)
 - fuel characteristics (amounts of each type of biomass)
 - fuel condition (moisture content)

Input data sets

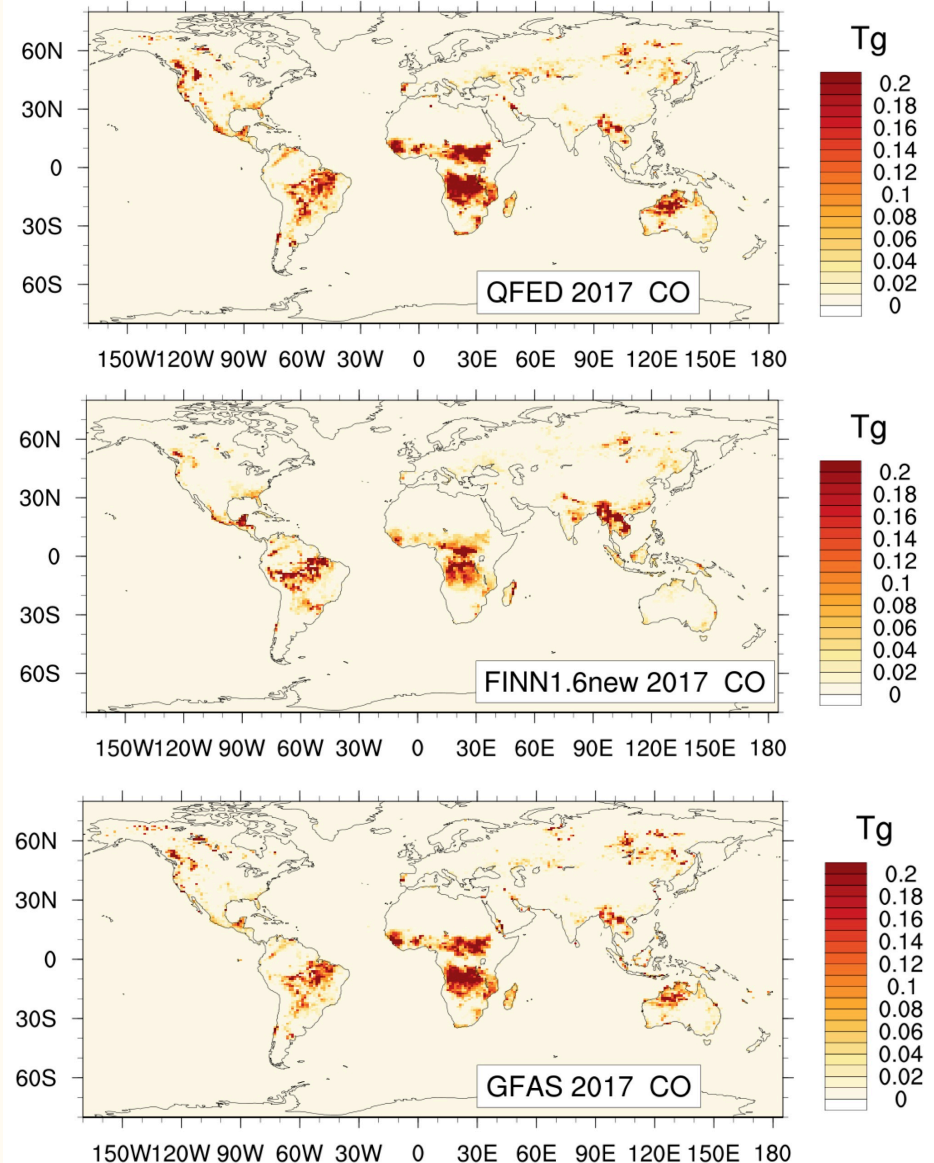
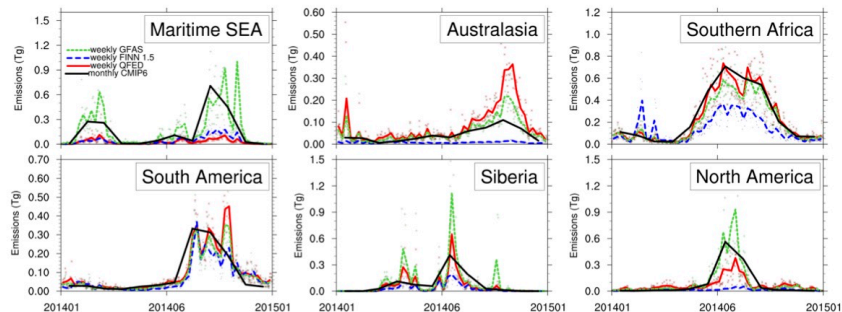
- MODIS Thermal Anomalies (Fire counts)
- Land cover maps (from MODIS and SPOT)

Other inventories use similar algorithms
Some use Fire Radiative Power satellite obs
Different landcover maps, emission factors, etc.

Differences in fire inventories

- FINN- Fire INventory from NCAR (Wiedinmyer)
- QFED - Quick Fire Emissions Dataset (NASA)
- GFAS – Global Fire Assimilation System (CAMS/ECMWF)
- GFED – Global Fire Emissions Database (van der Werf et al.)
- For CMIP6 (1750-2100)

Monthly fire emissions - 2014



Natural emissions

Natural terrestrial emissions

- VOCs from vegetation (isoprene, terpenes, methanol, etc.)
- Soil NO
- Volcanoes (SO_2)

Lightning NO

Ocean emissions – DMS, CO, VOCs

Dust

Sea salt

A biogenic emissions model: MEGANv2.1

Inputs:
model-simulated data
or
observations

LAI: Leaf Area Index
PFT: Plant Functional Type

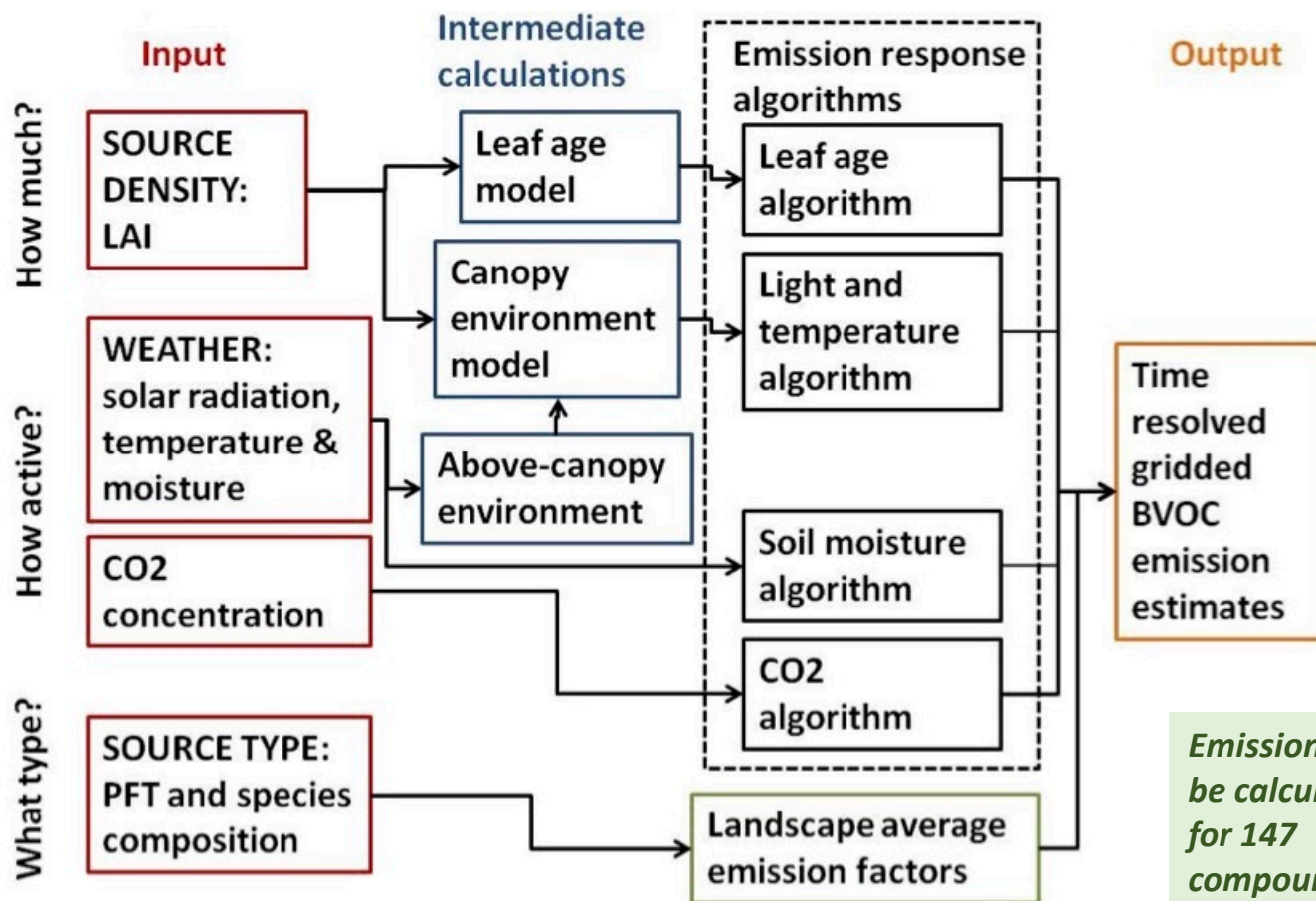


Fig. 1. Schematic of MEGAN2.1 model components and driving variables.

Guenther et al., GMD, 2012

MEGAN algorithm in a 3D model

The MEGAN-v2.1 algorithm is online in many models

Emissions for species i : $F_i = \gamma_i \sum \epsilon_{i,j} \chi_j$

where

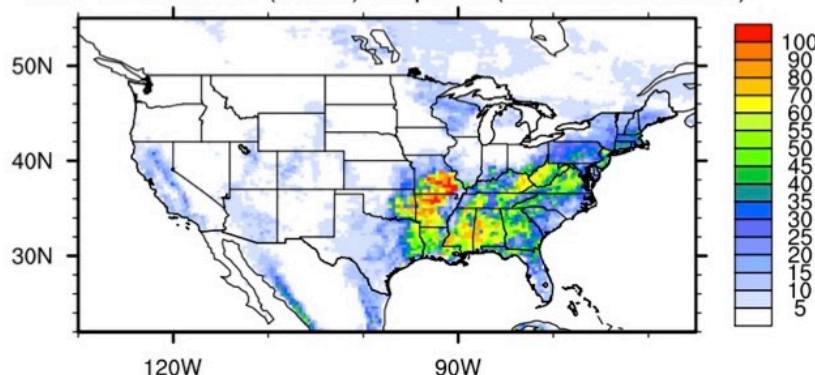
γ_i : emission activity factor, depends on leaf area index (LAI), meteorology (T, solar radiation), leaf age, soil moisture, with separate light-dependent and light-independent factors

$\epsilon_{i,j}$: emission factor at standard conditions for vegetation type (PFT) j

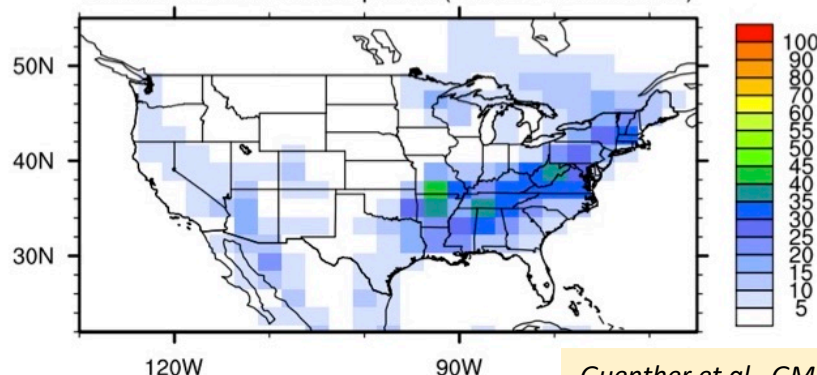
χ_j : fractional area of PFT j

Isoprene emissions in 2 different models differ greatly due to meteorology and vegetation maps

WRF-MEGANv2.1 (offline): Isoprene (micro-moles/m²/hr)



CLM-MEGANv2.1: Isoprene (micro-moles/m²/hr)



Guenther et al., GMD, 2012

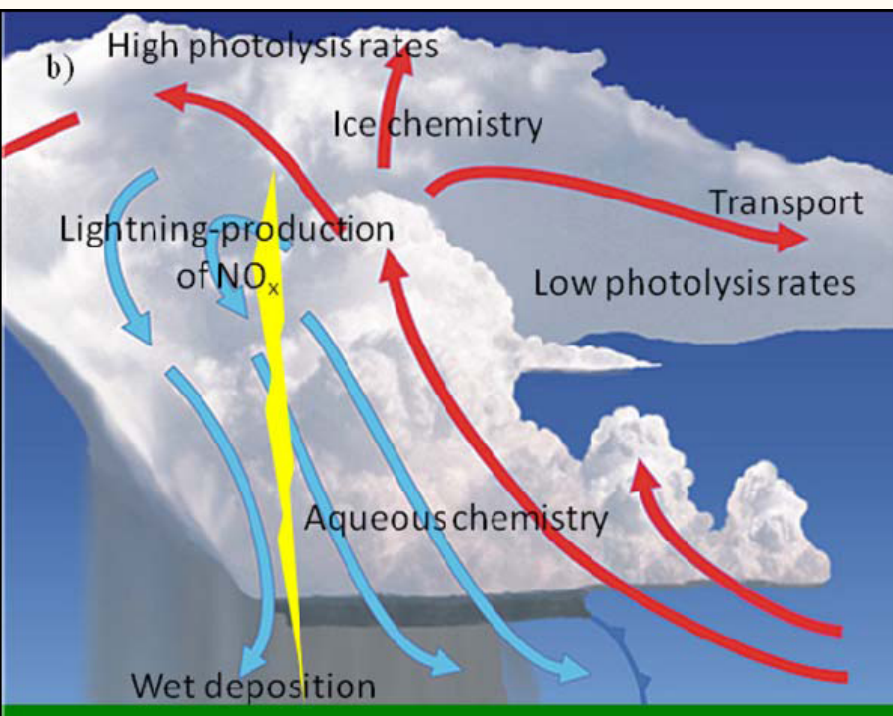
Lightning NO emissions

Lightning is produced in deep convective storms, creating a plasma leading to thermolysis of O_2 resulting in large NO emissions (1-10 Tg/yr globally)

Often parameterized based on cloud top height

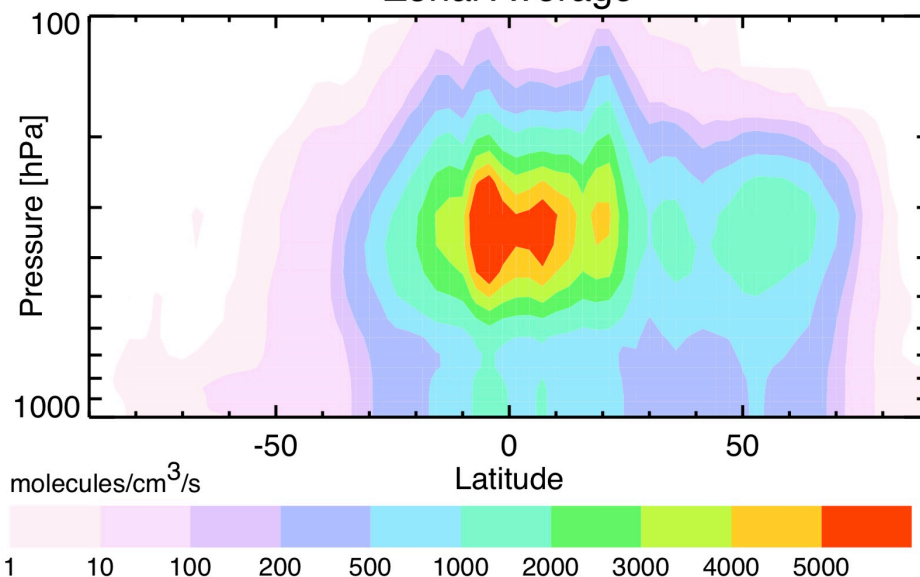
Cloud resolving models might use other approaches (e.g., updraft volume, ice mass flux, etc.)

Can also use observations of lightning flashes

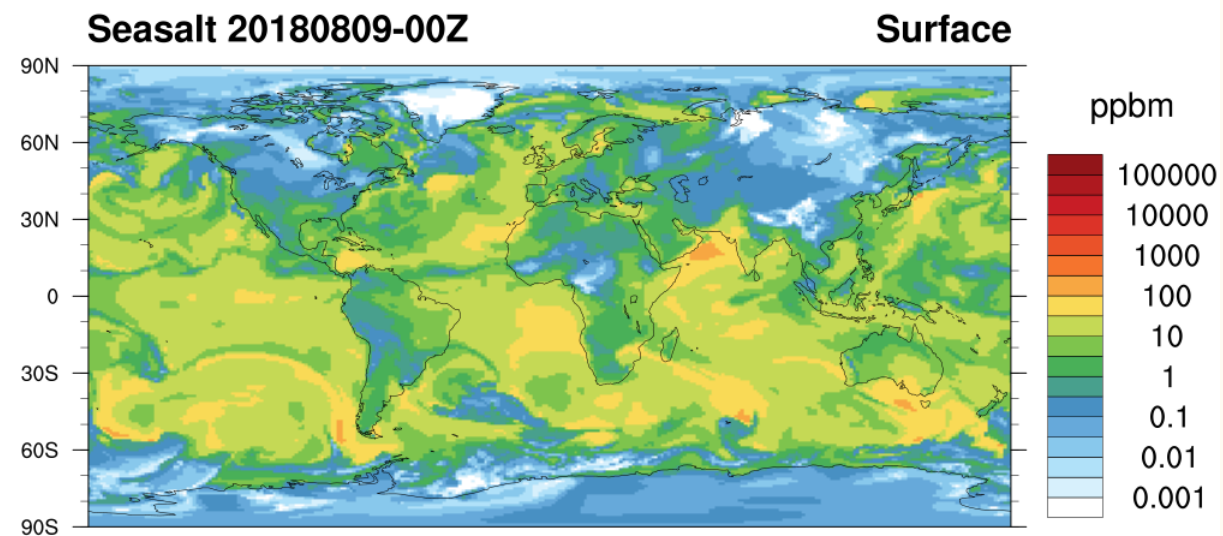
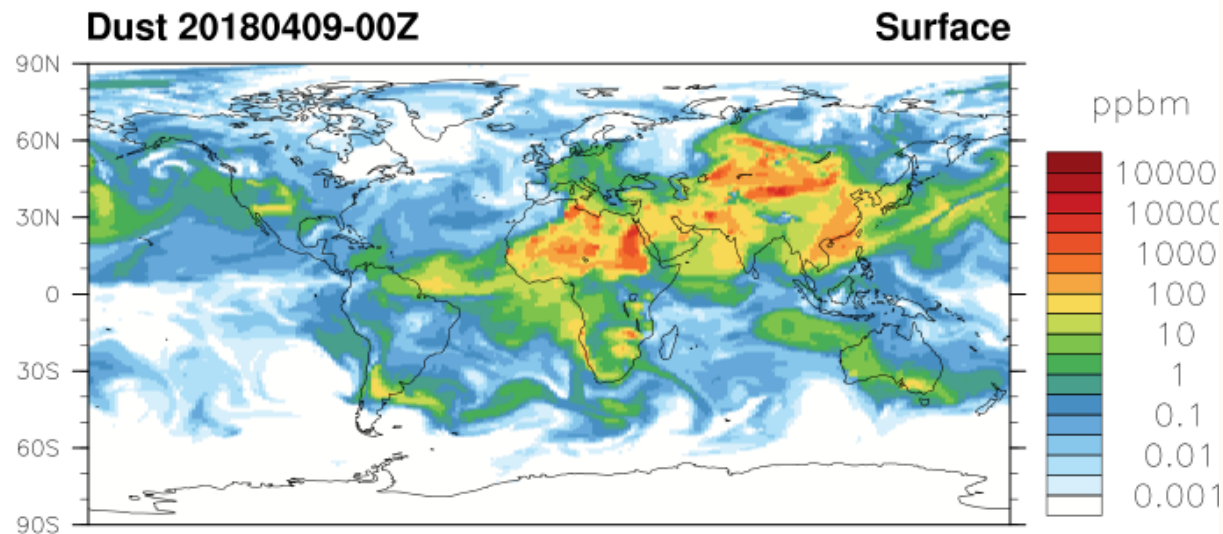


Annual Total Lightning NO Emissions - 2006

Zonal Average



Dust and Sea salt



Parameterizations using
surface wind speeds
Calculated online in
model

Dust: based on soil
erodibility maps,
tuned to AOD
observations

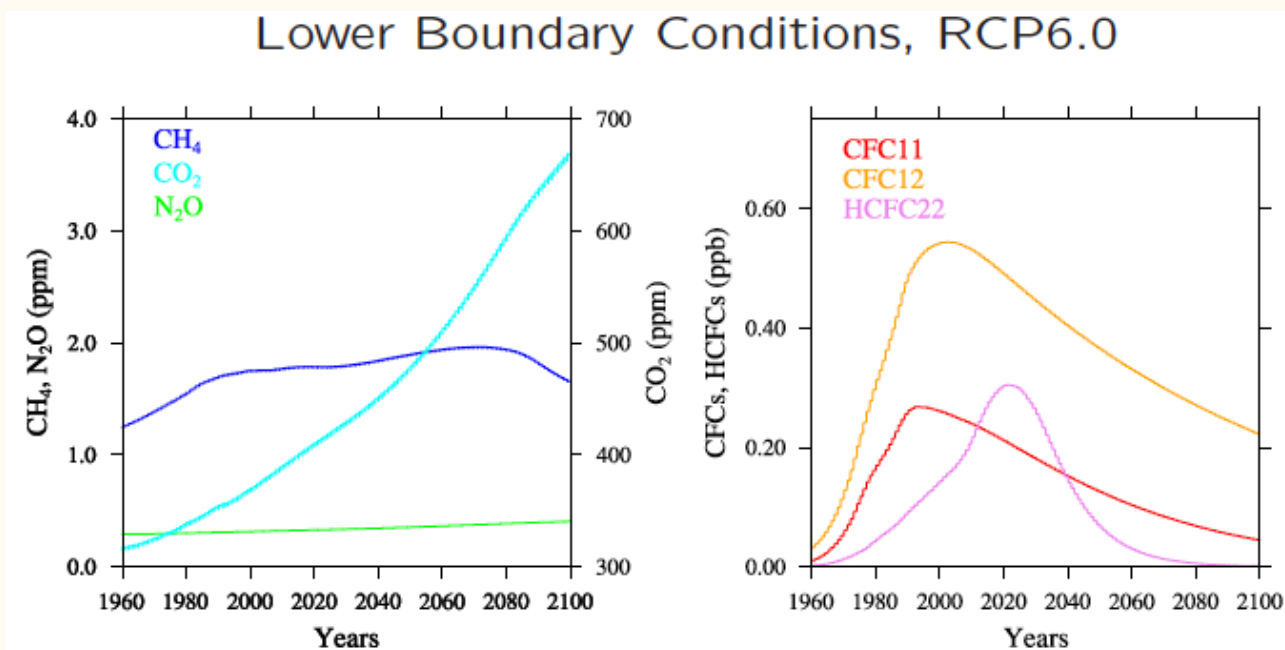
Images from CESM/WACCM:
<https://www.acom.ucar.edu/wacm/forecast/>

Other Boundary Conditions

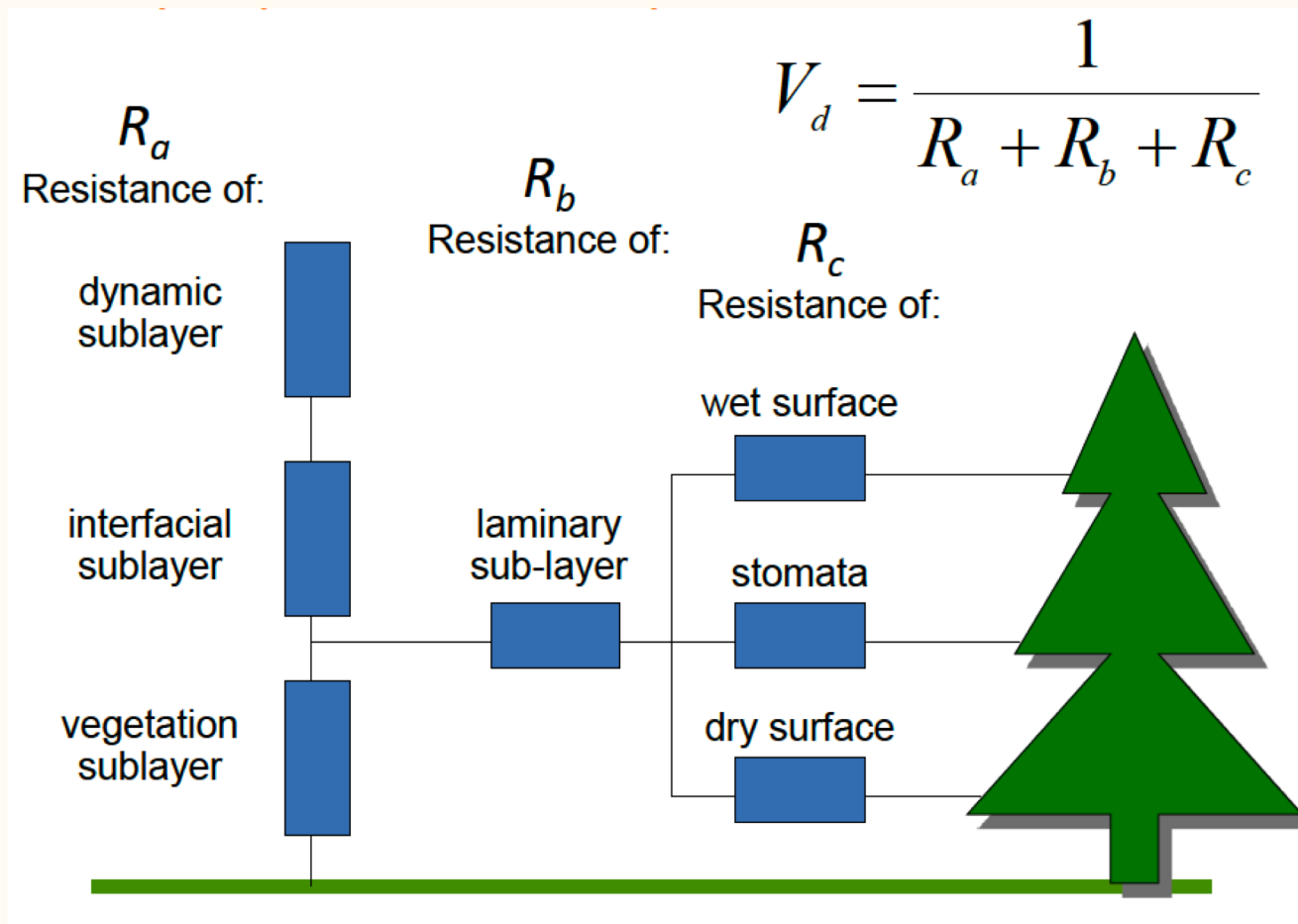
Long-lived species - for tropospheric chemistry often better to prescribe surface mixing ratios when sources and sinks are uncertain, such as for CH_4

Lateral boundary conditions for regional models

Stratosphere – ozone and other long-lived species (HNO_3 , NO_x) are important source to troposphere



Dry Deposition Velocity



Varies with surface type (vegetation, ocean, etc.)

Key component of ozone budget

Important for sticky and soluble gases: HNO_3 , CO, OVOCs, etc.

Wet Deposition

Wet deposition

- Scavenging in convective updrafts
- Rainout: in-cloud scavenging
- Washout: below-cloud scavenging

Highly soluble gases (HNO_3) and aerosols - kinetic process limited by mass transfer

Moderately soluble gases: dependent on effective Henry's Law constant

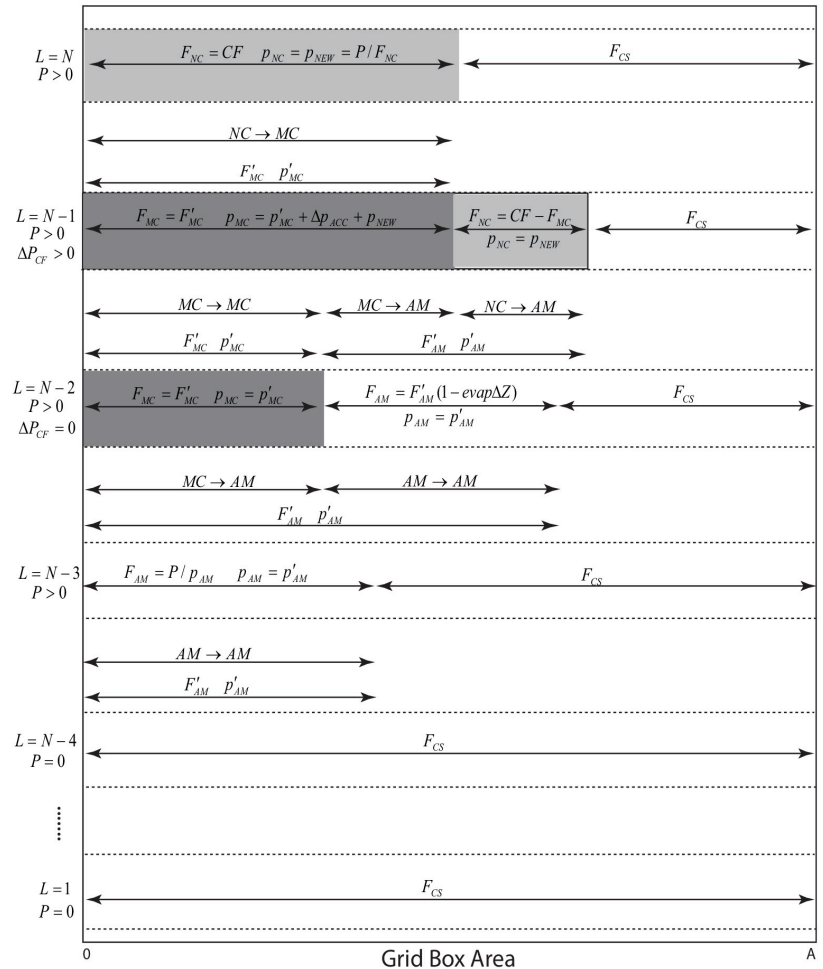
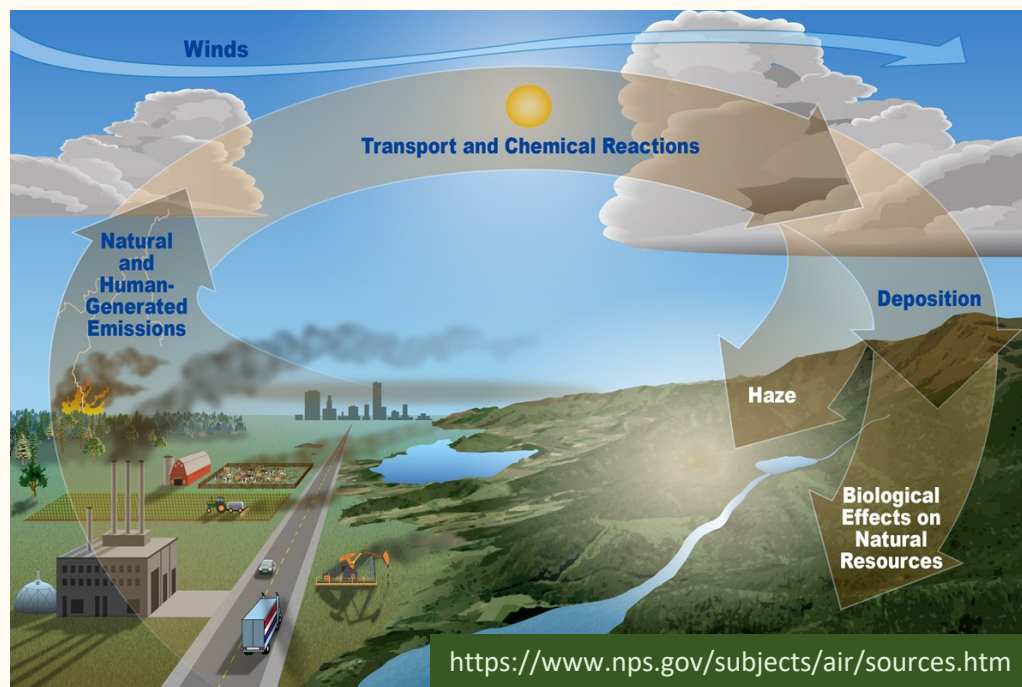


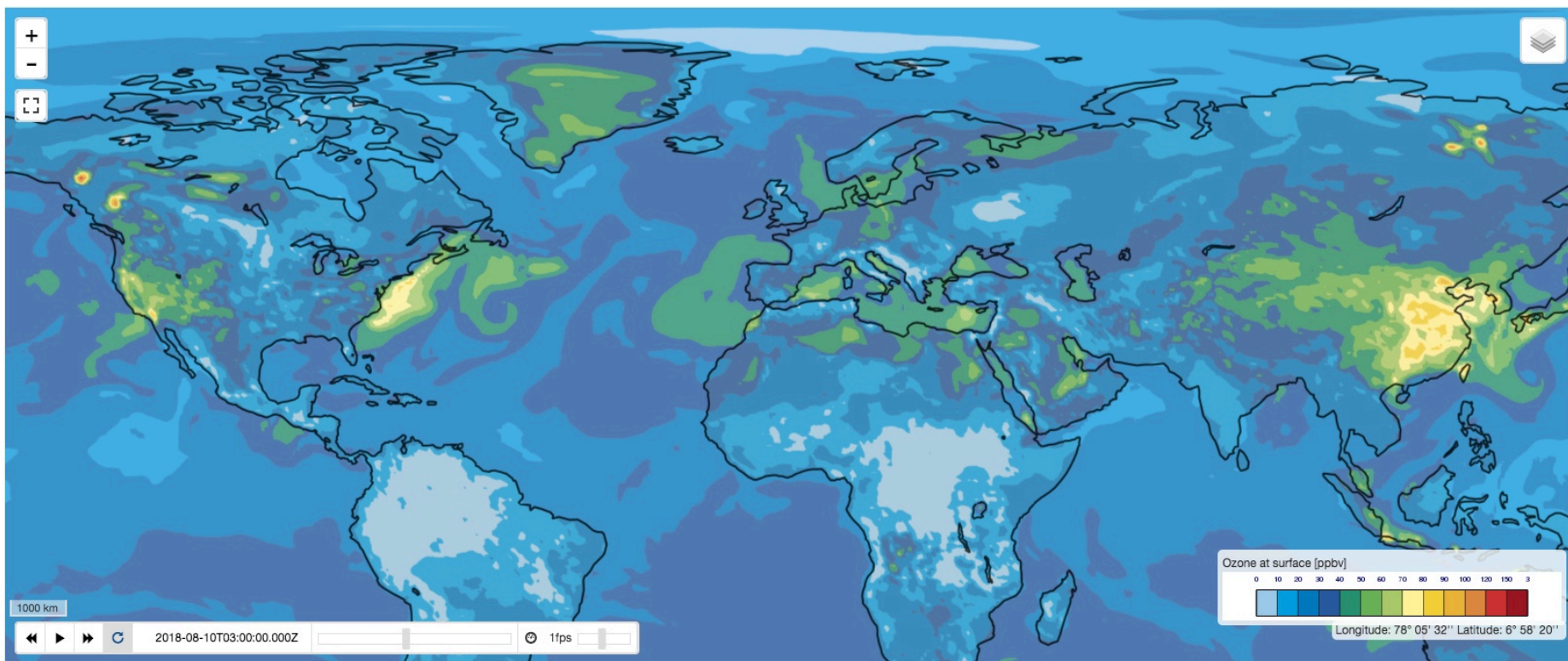
Fig. A1. Schematic of an idealized grid box with N levels, indicated by the horizontal dotted lines. The space between the levels is for illustrative purposes only. Levels N to $N-2$ contain clouds (grey shaded areas) and Levels N to $N-3$ have $P(L) > 0$. All gridbox fractions and precipitation rates are defined in the text and are shown for each level. The terms $XX \rightarrow YY$ indicate the gridbox fraction corresponding to $\hat{f}_{XX \rightarrow YY}$ as defined in the text (Eqs. A3 to A8). Level N is the first precipitating level, and the precipitating fraction and rate are given by Eqs. (A1) and (A21). Level $N-1$ provides an example with $CF(L) > CF(L+1)$, $\Delta P_{CF} > 0$ (Eq. A18), and $p_{NEW} > 0$ (p_{MC} and p_{NC} given by Eqs. A27 and A28). Level $N-2$ has $CF(L+1) > CF(L)$ and $\Delta P_{CF} = 0$, with the standard evaporation in the ambient region sufficient to account for the decrease in P from level $N-1$ to $N-2$ (Eqs. A16 and A17). Level $N-3$ has $CF(L) = 0$ and the decrease in P from level $N-2$ to $N-3$ exceeds the standard evaporation rate (Eq. A17). There is full evaporation of the precipitation in level $N-4$ ($P(L) = 0$).

Applications

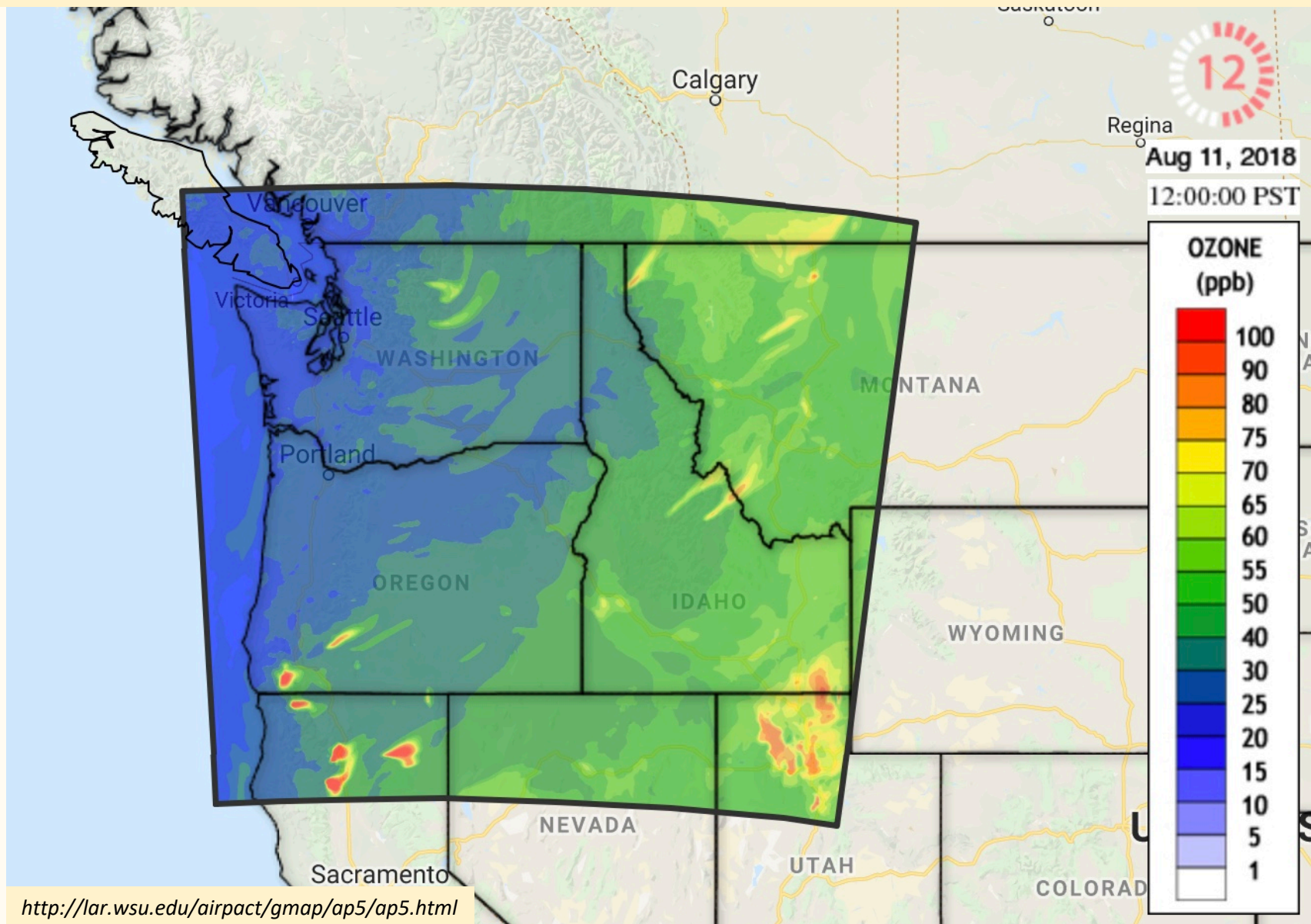
- Chemical & air quality forecasts
- Source attribution through “tagging”
- Analysis of observations and understanding processes
- Chemistry-climate coupling



Global surface forecasts

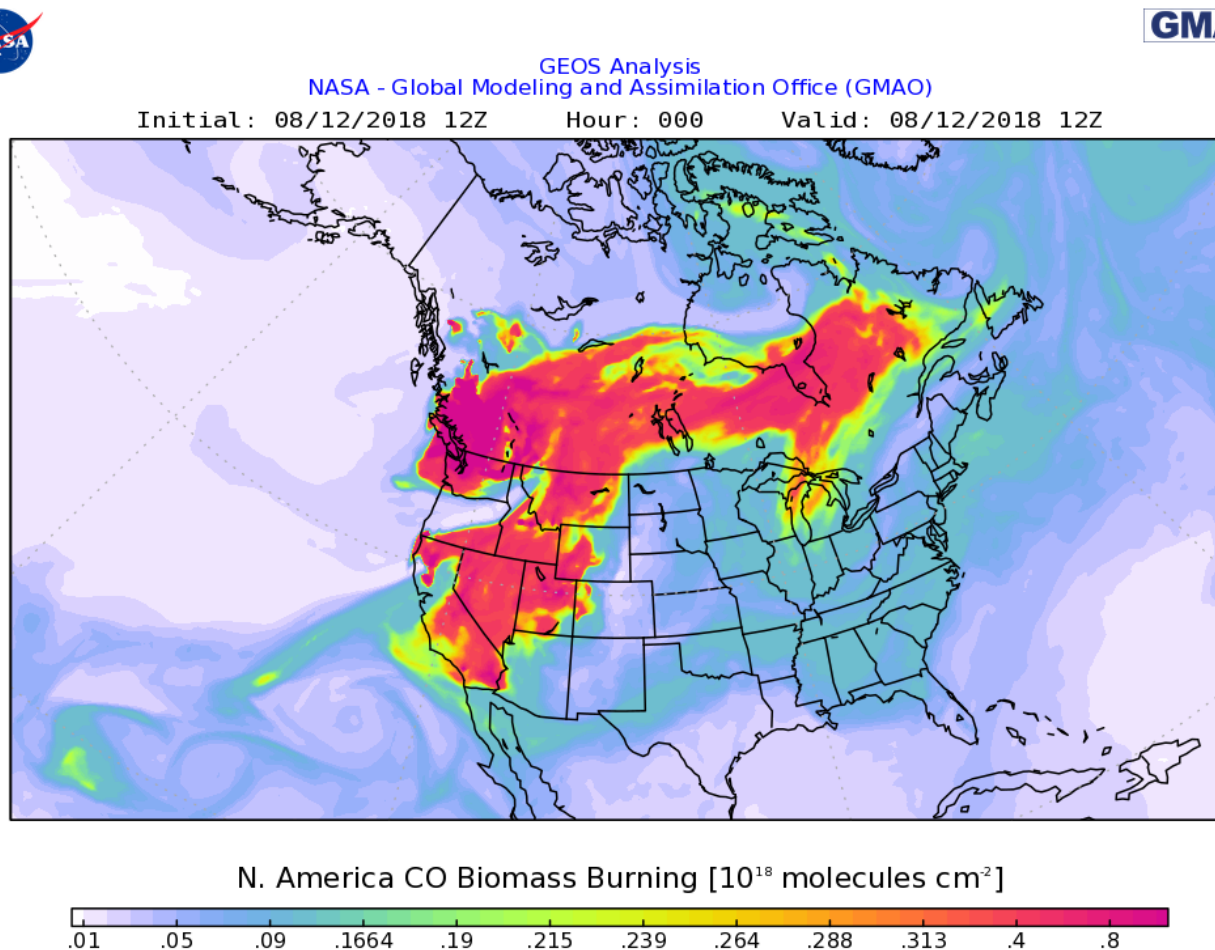


Regional AQ forecasts (e.g., AIRPACT, WSU)



Source Attribution

"Tagging" of source type or region allows identification of source contributions to observations



Frequently used
in flight planning
for aircraft
experiments

Adding tagging to chemical mechanisms

Artificial tracers:

CO_fire_10day -> CO₂ : $k=1/10\text{days}$, CO emissions from fires

SO₂_volcano -> {} : $k=1/x$, SO₂ emissions from active volcano

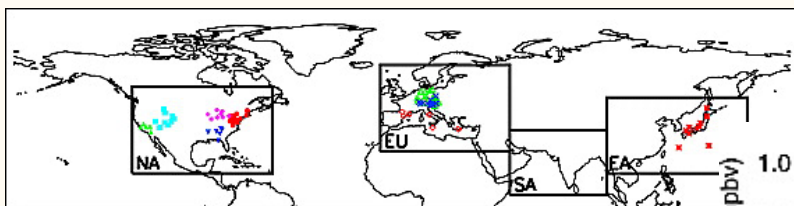
"Tagged" CO, runs parallel to complete chemistry without affecting it:

CO_fire + OH -> OH : $k=k(\text{CO}+\text{OH})$

Other compounds with simple chemistry can be tagged and added to standard chemistry (BC)

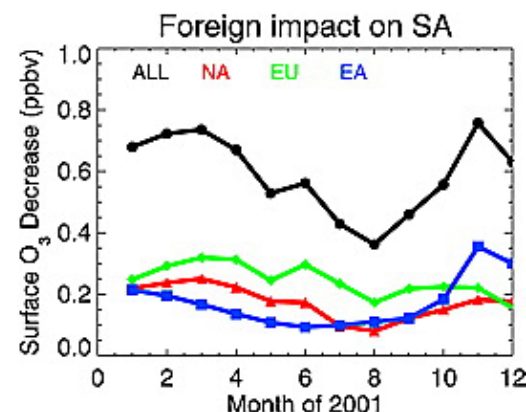
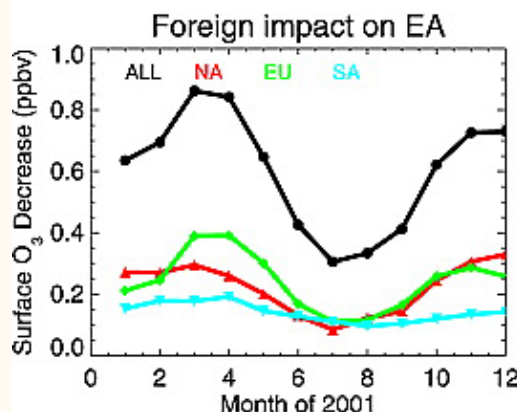
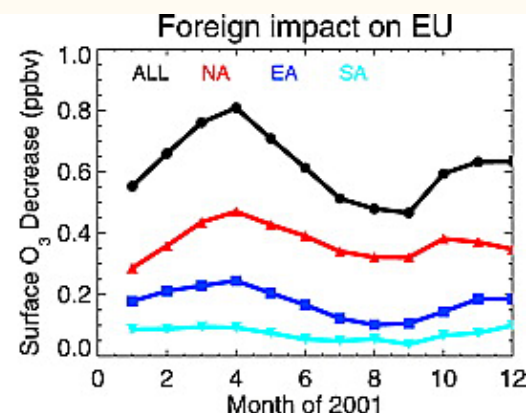
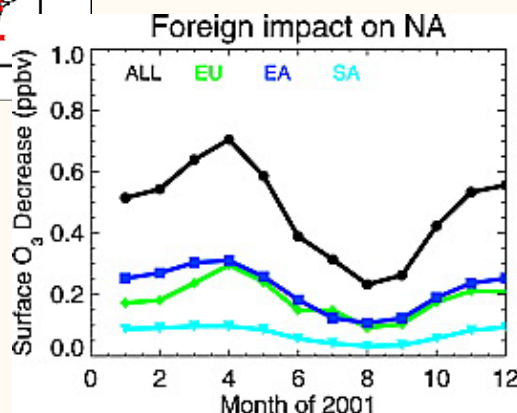
Emissions perturbations

- For studying impacts of actually changing the emissions
- To estimate source contributions from various source types or regions

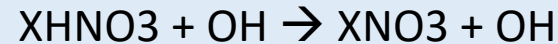
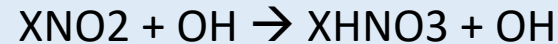
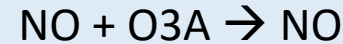
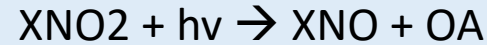
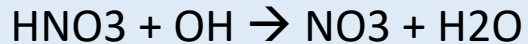
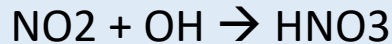
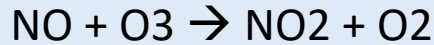
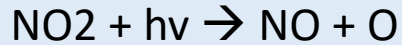


HTAP model intercomparison

Decreased emissions in each region by 20% and found decrease in ozone



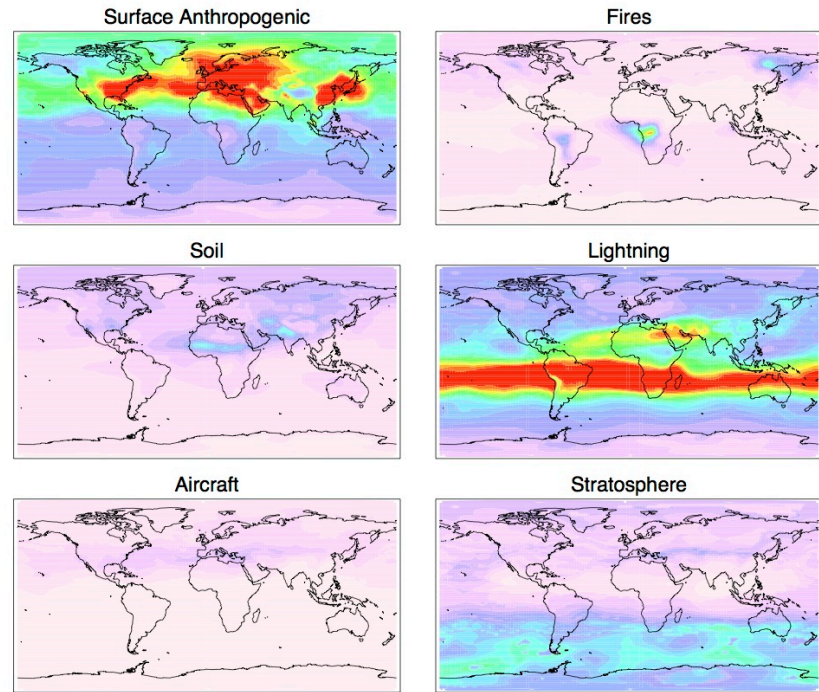
Tracking ozone from NO emissions



Any region or source type of NO emissions can be tagged, to quantify contribution of ozone from that source

Similar tagging of VOCs -> ozone also possible (Butler, GMD, 2018)

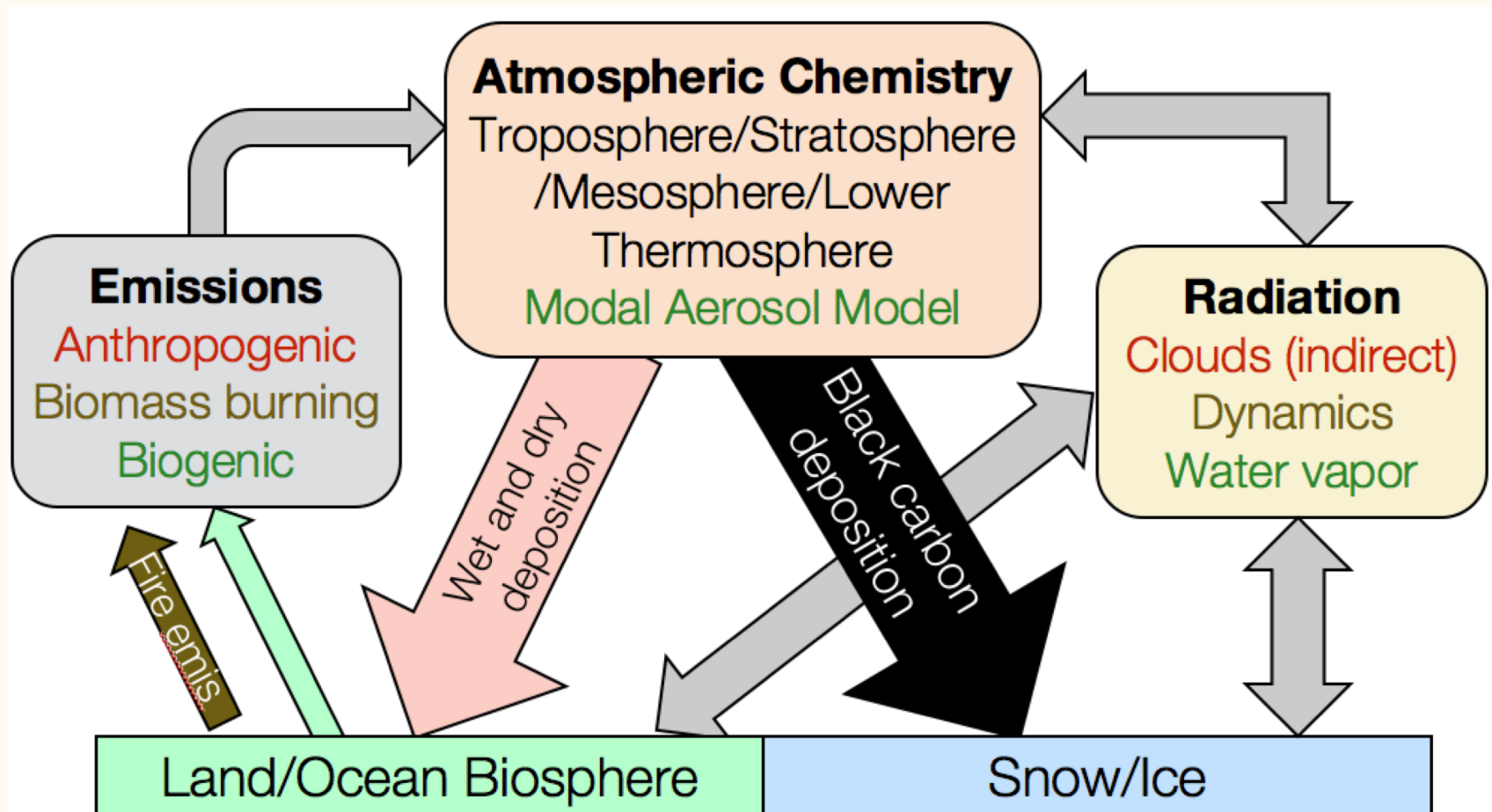
Tagged O₃ Column (O₃ < 150 ppbv) - Jul



Aircraft experiments: forecasting and analysis



Chemistry-Climate Interactions



Conclusions

- Models are tools to explore atmospheric processes
- Emissions are very uncertain and should be appropriately selected and evaluated for each study
- Uncertainties in other processes (deposition, chemistry, clouds, meteorology) must be kept in mind