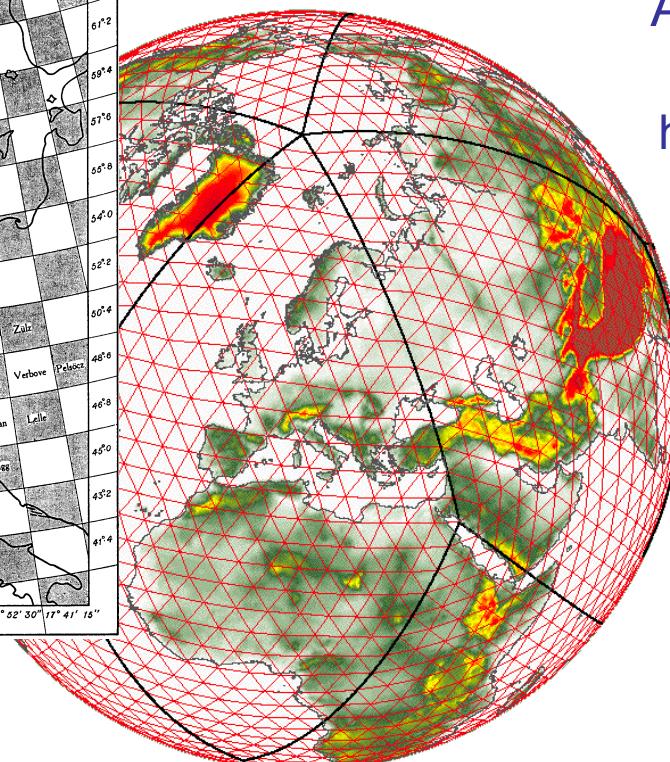
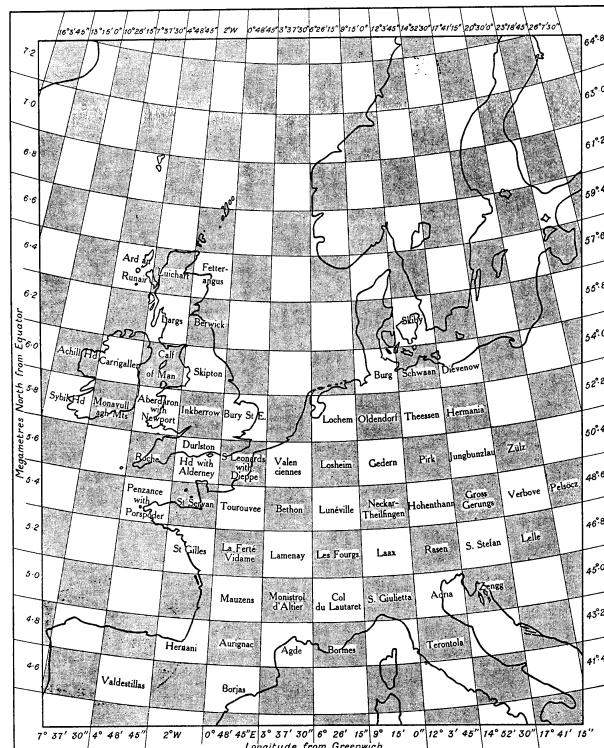


Chapter 1

Introduction to Weather and Climate Models



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February 2015

Handouts

Numerical Modelling of Weather and Climate

Christoph Schär and Ulrike Lohmann

http://www.iac.ethz.ch/edu/courses/master/modules/numerical_modelling_of_weather_and_climate

Outline

Historical perspective

Atmospheric scales

Governing equations

Parameterizations

Initial conditions

Climate models

Computational aspects

Billow Couds (Gemsfarenstock, March 11, 2003)



Picture: Thomas Schumann

Sometimes the atmosphere exhibits physical principles in a textbook fashion,
....

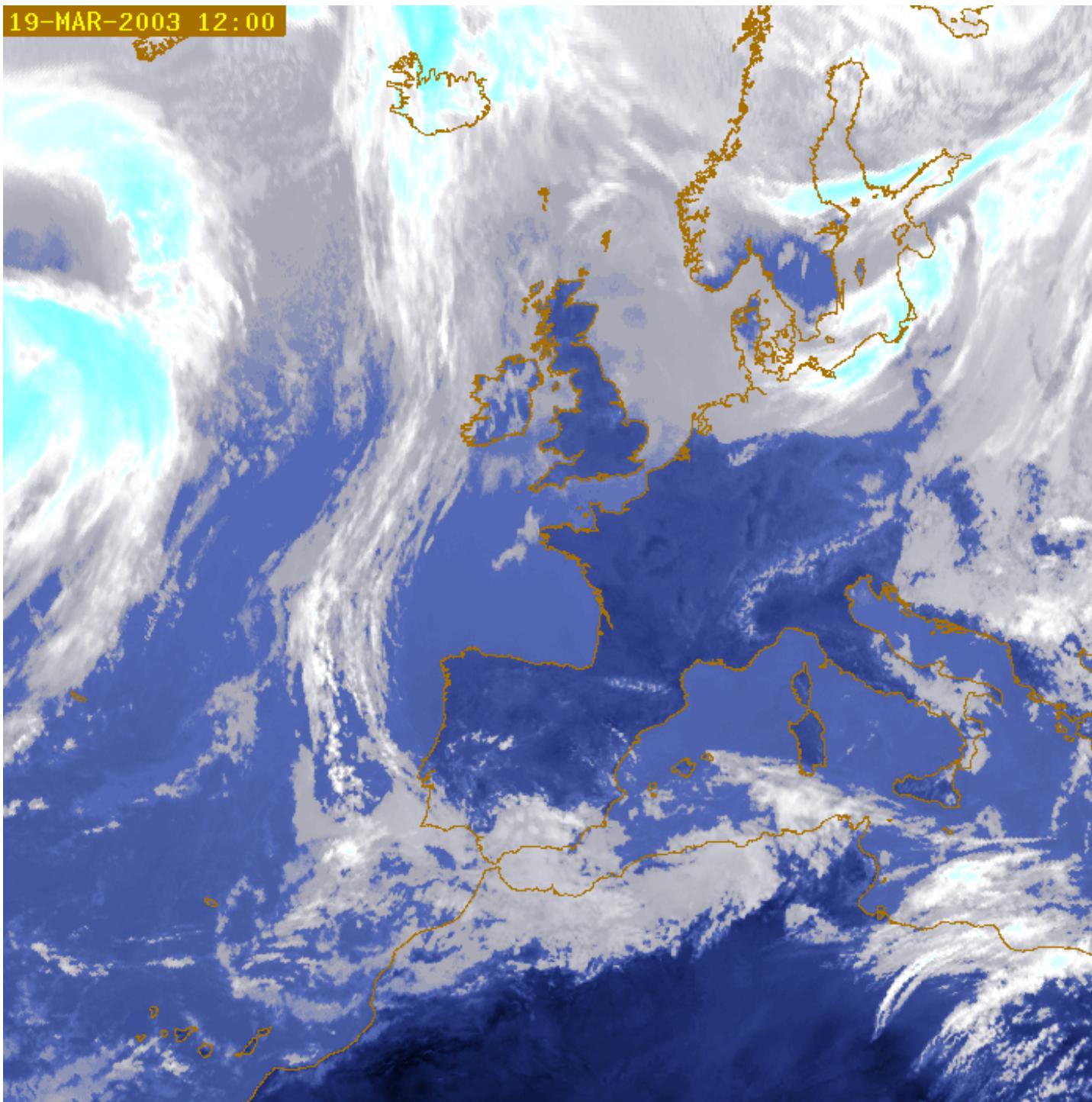
Richardson-Number Criteria:

$$Ri = \left(\frac{-g}{\rho} \frac{\partial \rho}{\partial z} \right)^{1/2} \left| \frac{\partial \mathbf{v}}{\partial z} \right|^{-1} \leq \frac{1}{4}$$



Laboratory-Experiment of Kelvin-Helmholtz Instability

19-MAR-2003 12:00



.... more generally,
however, the
atmosphere looks
very complicated, and
a direct link to simple
physical principles is
not evident.

IR Satellite Picture
MeteoSwiss, Zürich

Vilhelm Bjerknes (1862-1951)



In 1904, Bjerknes proposed that „weather forecasting should be considered as an *initial value problem* of mathematical physics.“

His proposal was motivated by his new circulation theorem, that overcame obvious limitations of Kelvin's earlier theorem.

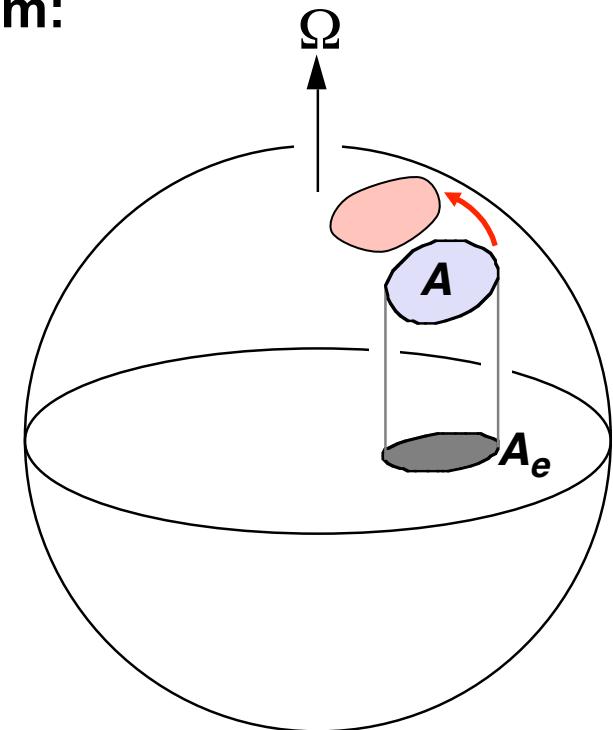
Circulation Theorem:

Circulation:

$$C = \oint \mathbf{v} \cdot d\mathbf{s}$$

Kelvin's Theorem:

$$\frac{dC}{dt} = 0$$



Bjerknes' Theorem:

$$\frac{dC}{dt} = -2\Omega \frac{dA_e}{dt} - \oint \rho^{-1} dp$$

Earth's

rotation

baroclinic

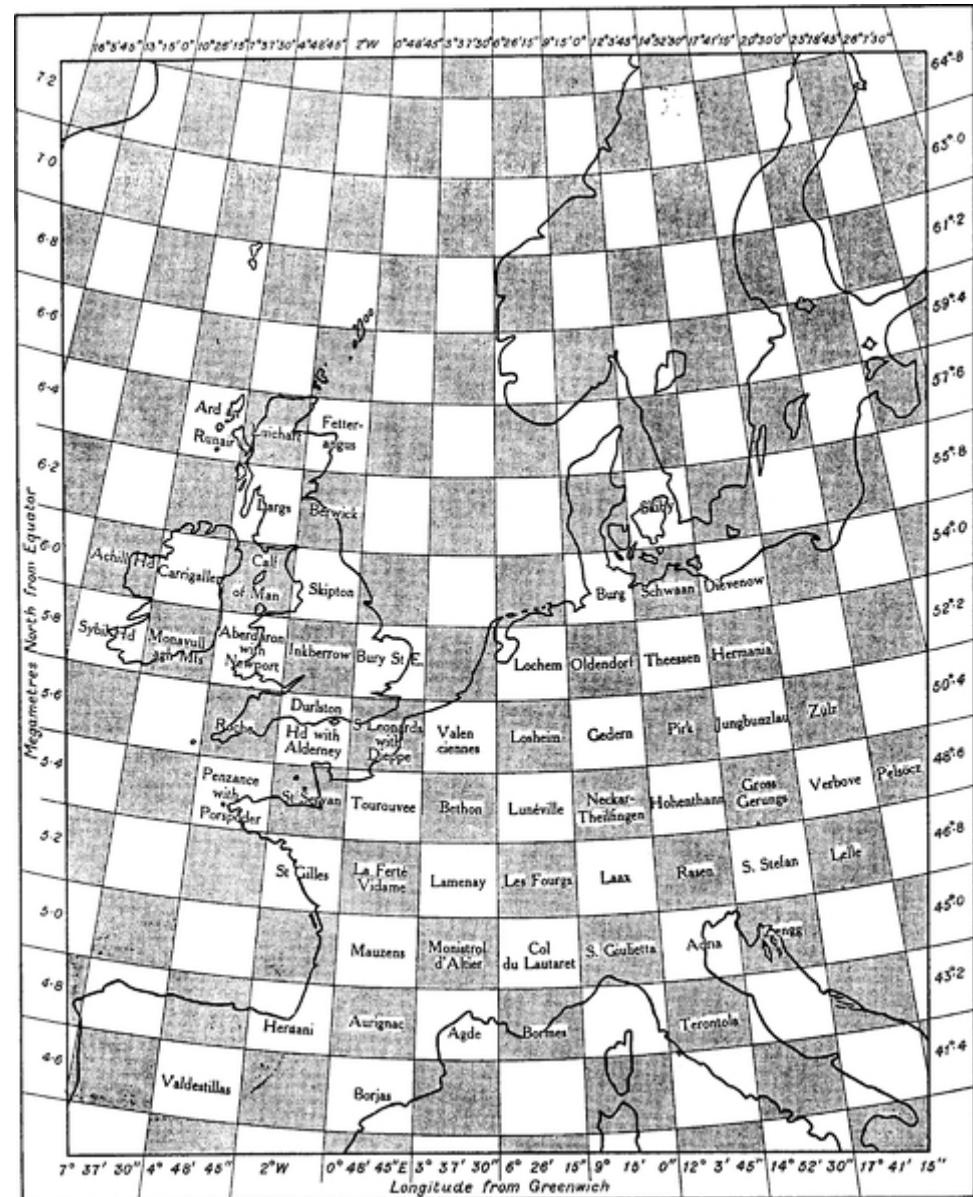
effects

Lewis F. Richardson (1881-1953)



In 1922, Richardson provided the first formulation of the atmospheric equations on a computational grid.

"If the coordinate chequer were 200 km square in plan, ... 64,000 computers would be needed to race the weather. In any case, the organisation indicated is a central forecast-factory."



Extracts from Richardson's Book

«Individual computers are *specialising on the separate equations*. Let us hope for their sake that they are moved on from time to time to new operations.»

Parallelization

«It took me the best part of six weeks to draw up the *computing forms*»

Distributed Memory

«The work of each region is coordinated by an official of higher rank. Numerous little "night signs" display the instantaneous values so *their neighbouring computers can read them*.»

Communication

«From the floor a tall pillar rises to half the height of the hall. It carries a large pulpit on its top. In this sits the man in charge of the whole theatre. One of his duties is to *Maintain an uniform speed of progress*. He turns a beam of rosy light upon any region that is running ahead of the rest, and a beam of blue light upon those who are behindhand.»

Synchronization

«In a neighbouring building there is a research department. Outside are playing fields, houses, mountains and lakes, for it was thought that those who compute the weather should breathe it freely.»

“Cooling”

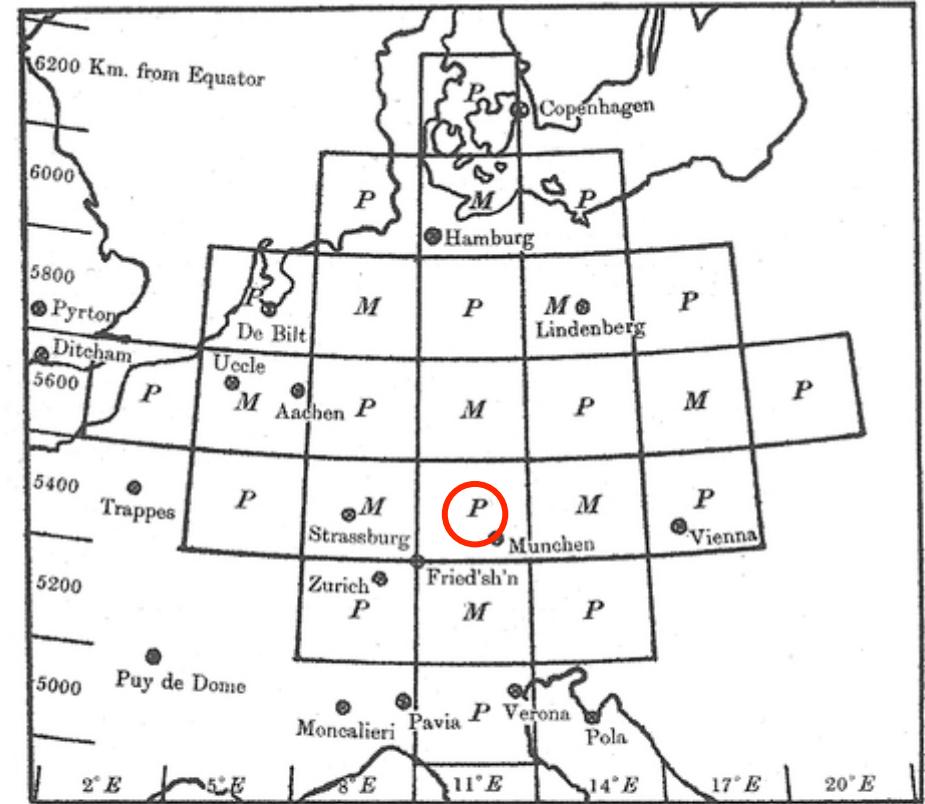
Richardson's forecast (proof of concept)

Richardson conducted one 6-hour forecast for a single grid-point to the North of Munich.

He conducted only one single time step, so the main result was the pressure tendency at this point.

The forecast was published 14 years after the event:

- Initialized: May 20, 1910, 4 UTC
- Executed: during 6 weeks in 1916
- Published: 1922 (in his book)



Richardson used a staggered grid (P = pressure, M = momentum).
Red circle shows targeted grid point.
Locations with radio- or pilot-balloon ascents are also indicated.

Divergence of horizontal momentum-per-area. Increase of pressure

COMPUTING FORM P.XIII.

$$\text{The equation is typified by: } -\frac{\partial R_s}{\partial t} = \frac{\partial M_{ss}}{\partial s} + \frac{\partial M_{sn}}{\partial n} - M_{ss} \frac{\tan \phi}{a} + m_{ss} - m_{sn}^* + \frac{2}{a} M_{sn}. \text{ (See Ch. 4/2 #5.)}$$

* In the equation for the lowest stratum the corresponding term $-m_{sn}^*$ does not appear

Longitude 11° East $\delta s = 441 \times 10^6$			Latitude 5400 km North $\delta n = 400 \times 10^6$			Instant 1910 May 20 th 7 th G.M.T. Interval, δt 6 hours $a^{-1} \cdot \tan \phi = 1.78 \times 10^{-6}$ $a = 6.36 \times 10^6$					
Ref.:—	$\frac{\partial M_s}{\partial s}$	$\frac{\partial M_s}{\partial n}$	previous 3 columns	previous column		Form P.vi	Form P.vii	equation above	previous column	previous column	previous column
h_1	$10^{-6} s$	$10^{-6} s$	$-\frac{M_s \tan \phi}{a}$	$\text{div}'_{ss} M$	$-g\delta t \text{div}'_{ss} M$	m_s	$\frac{3M_s}{a}$	$-\frac{\partial R}{\partial t}$	$+\frac{\partial R}{\partial t} \delta t$	$g \frac{\partial R}{\partial t} \delta t$	$\frac{\partial p}{\partial t} \delta t$
	$10^{-6} s$	$10^{-6} s$	$10^{-6} s$	$10^{-6} s$	$100 s$	$10^{-6} s$	$10^{-6} s$		$100 s$	$100 s$	$100 s$
h_2	-61	-245	-6	-312	656	0					0
h_3	367	-257	2	112	-236	-83		-229	49.5	483	483
h_4	93	-303	-16	-226	478	0.06		-136	29.4	287	287
h_5	32	-55	-12	-35	74	165					770
h_6	-256	88	-8	-226	479						1032
	Note: $\text{div}'_{ss} M$ is a contraction for $\frac{\partial M_s}{\partial s} + \frac{\partial M_s}{\partial n} - M_s \frac{\tan \phi}{a}$				SUM = 1461 $= \frac{\partial p_s}{\partial t} \delta t$	Last row of Form P.vi					

the subsequent columns to be filled up after
vertical velocity has been computed on
last row of Form P.vi

$$\frac{\partial p_s}{\partial t} = 145.1 \text{ hPa/6h}$$

check by
 $\Sigma g\delta t \text{div}'_{ss} M$

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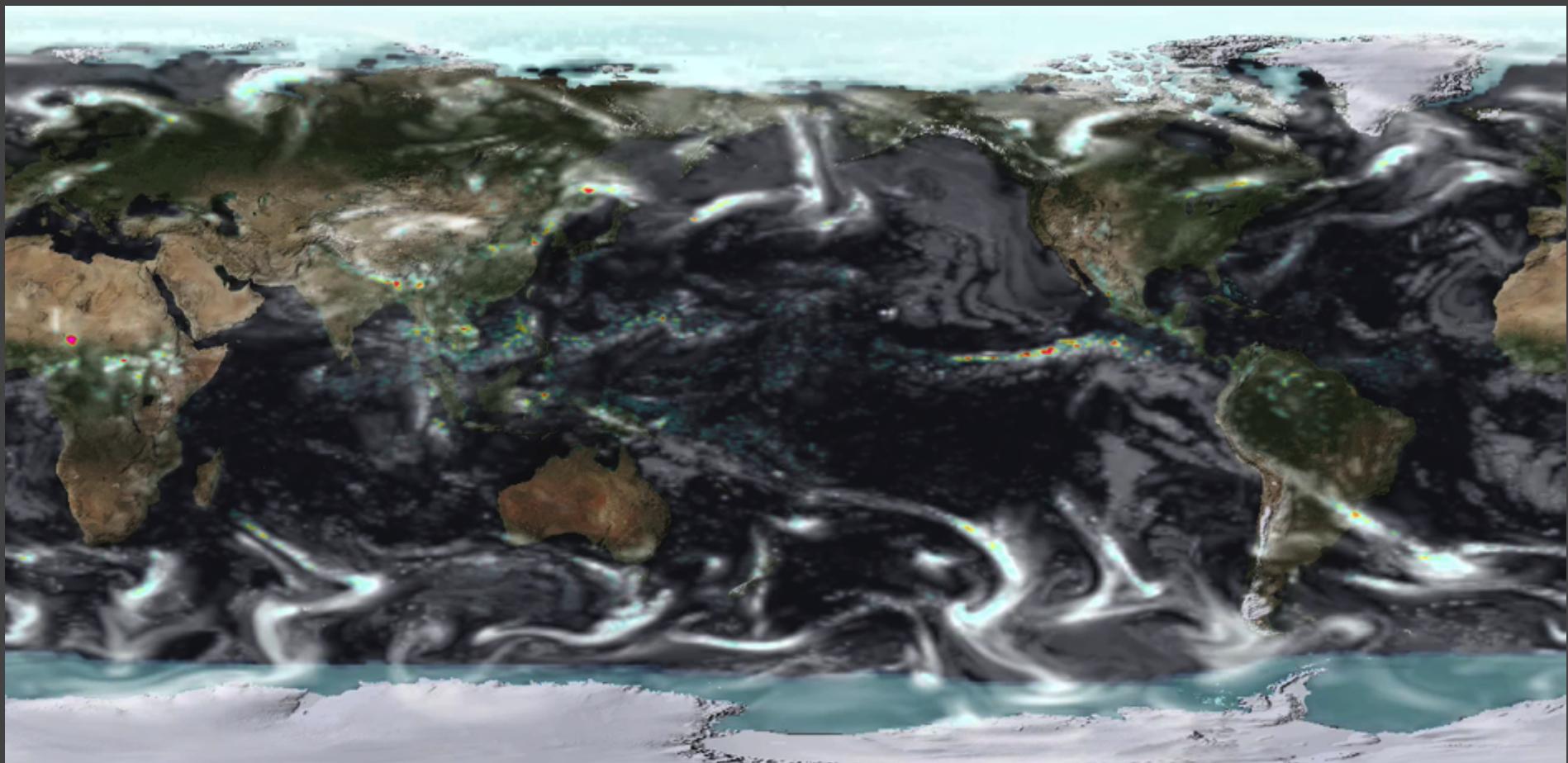
Parameterizations

Initial conditions

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Global



NUGAM (N216 HadGAM1a)

Model by the UJCC Team and UKMO/NCAS collaborators: <http://www.earthsimulator.org.uk>
Movie by: R. Stöckli (NASA Earth Observatory, USA) and P.L. Vidale (NCAS, UK)

1 AUG 1978 01h UTC

UK-Japan Climate Collaboration

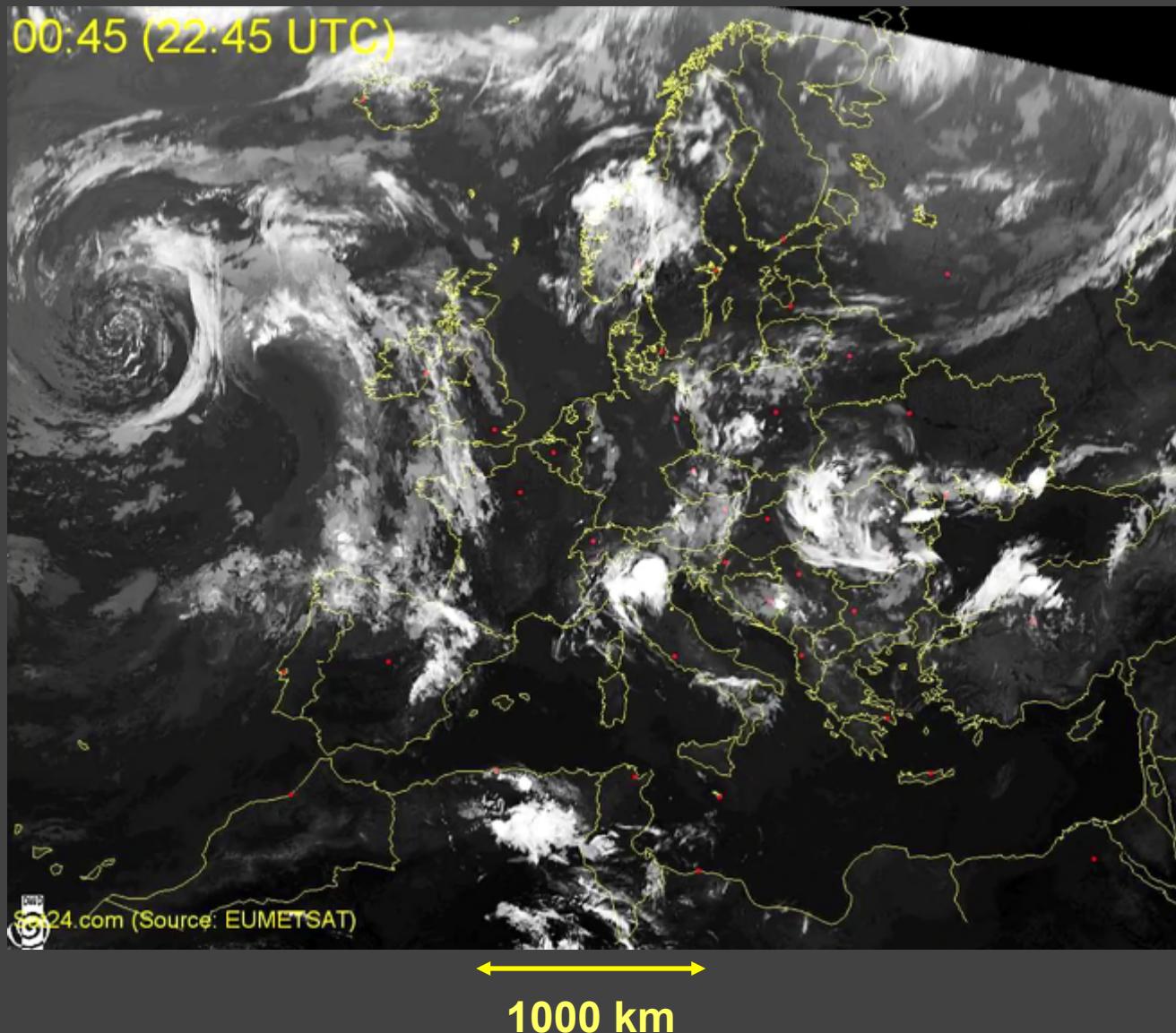


↔

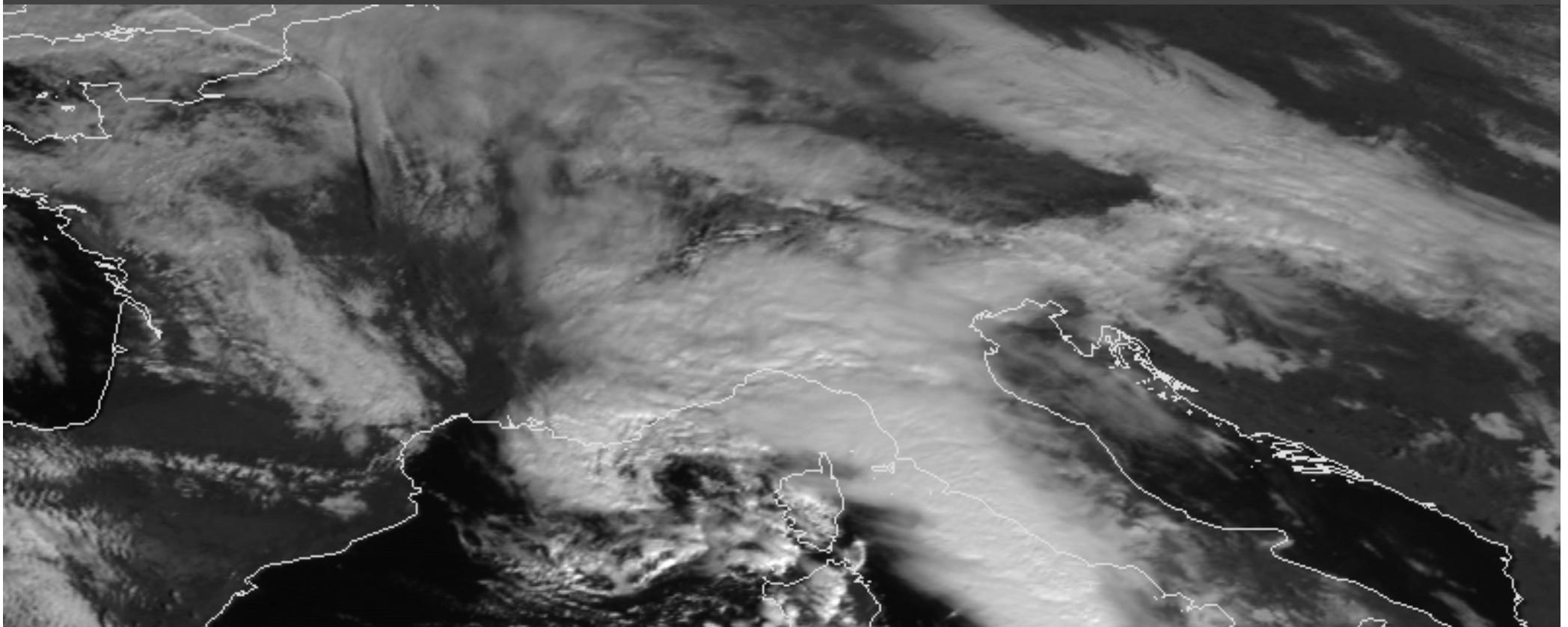
10,000 km

(High-resolution global model)

Continental



Regional / Mesoscale

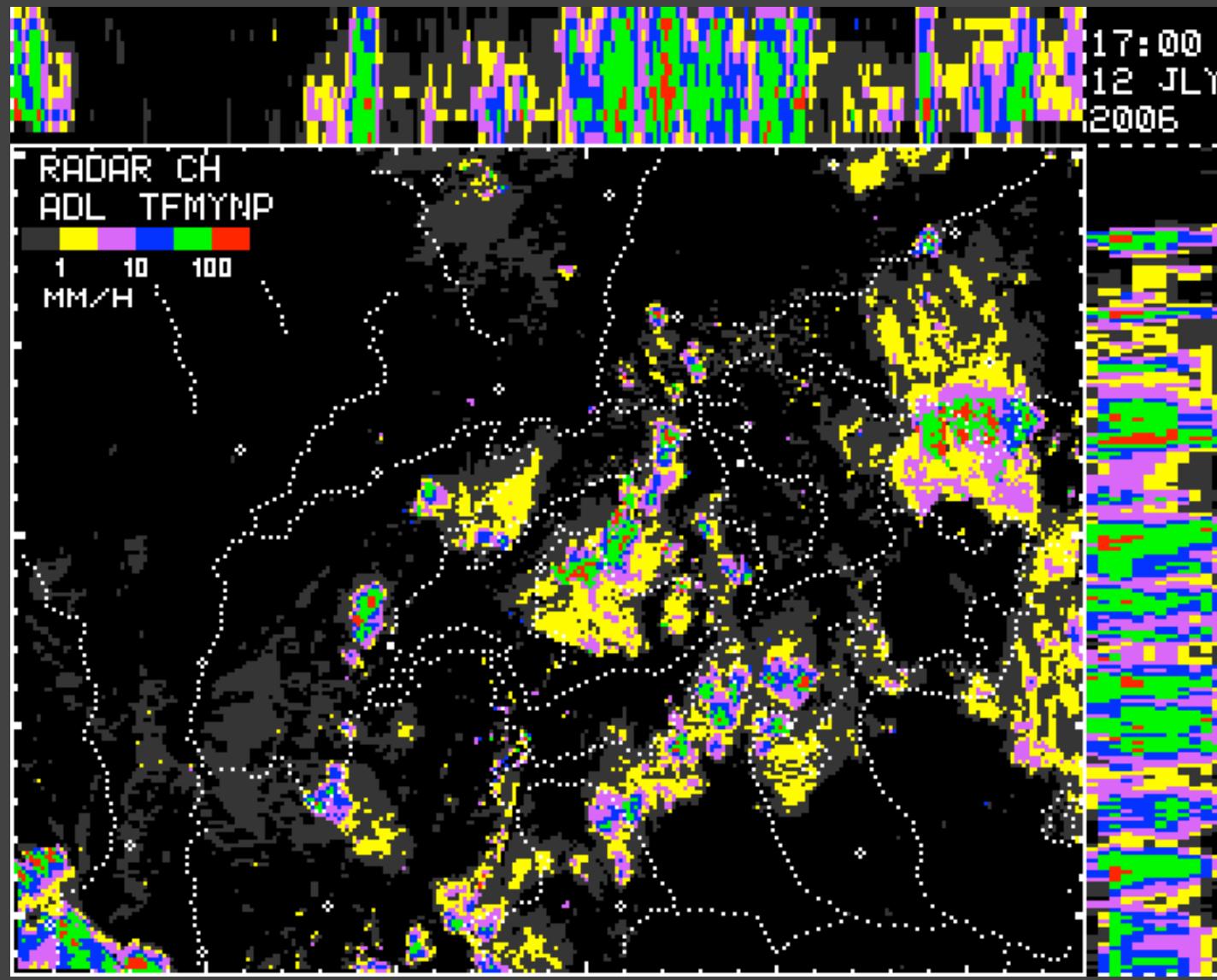


500 km

21. October 1999

(MAP, Meteosat Rapid Scan, VIS)

Regional / Mesoscale



Weather radar, 12 July 2006

100 km

(MeteoSwiss)

Local



↔

Meter to kilometer

Sub-millimeter

Large
cloud drop
 $r = 50 \mu\text{m}$

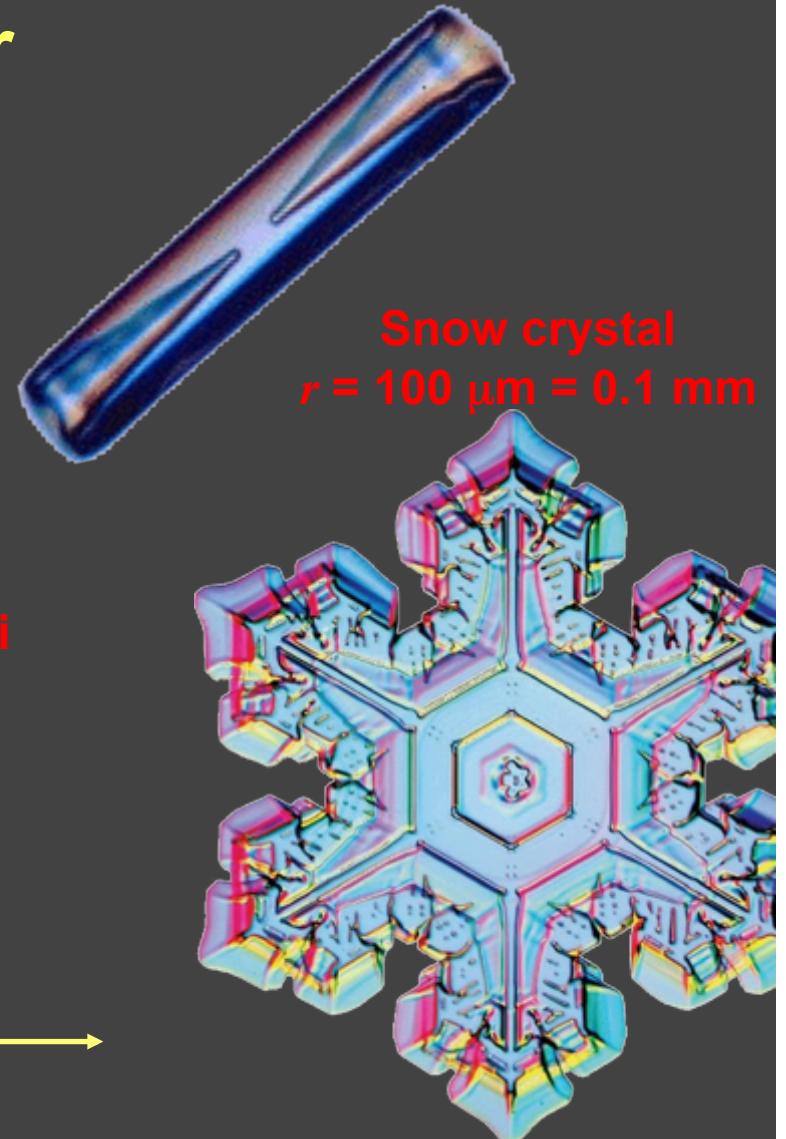
$$r = 10 \mu\text{m}$$

Small
raindrop
 $r = 100 \mu\text{m} = 0.1 \text{ mm}$

$$0.1 \text{ mm}$$

Typical raindrop
 $r = 1000 \mu\text{m} = 1 \text{ mm}$

Snow crystal
 $r = 100 \mu\text{m} = 0.1 \text{ mm}$



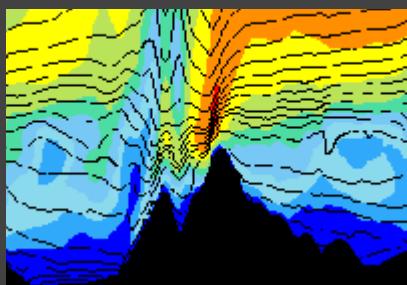
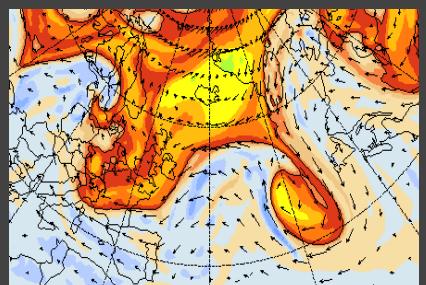


Lewis F. Richardson (1881-1953)

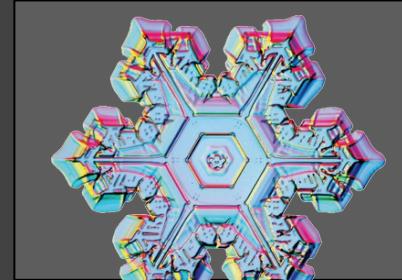
**Big whirls have little whirls,
little whirls have lesser whirls.**

“down-scale”

$$10^6 \text{ m} = 1 \text{ Mm}$$



$$10^{-6} \text{ m} = 1 \mu\text{m}$$



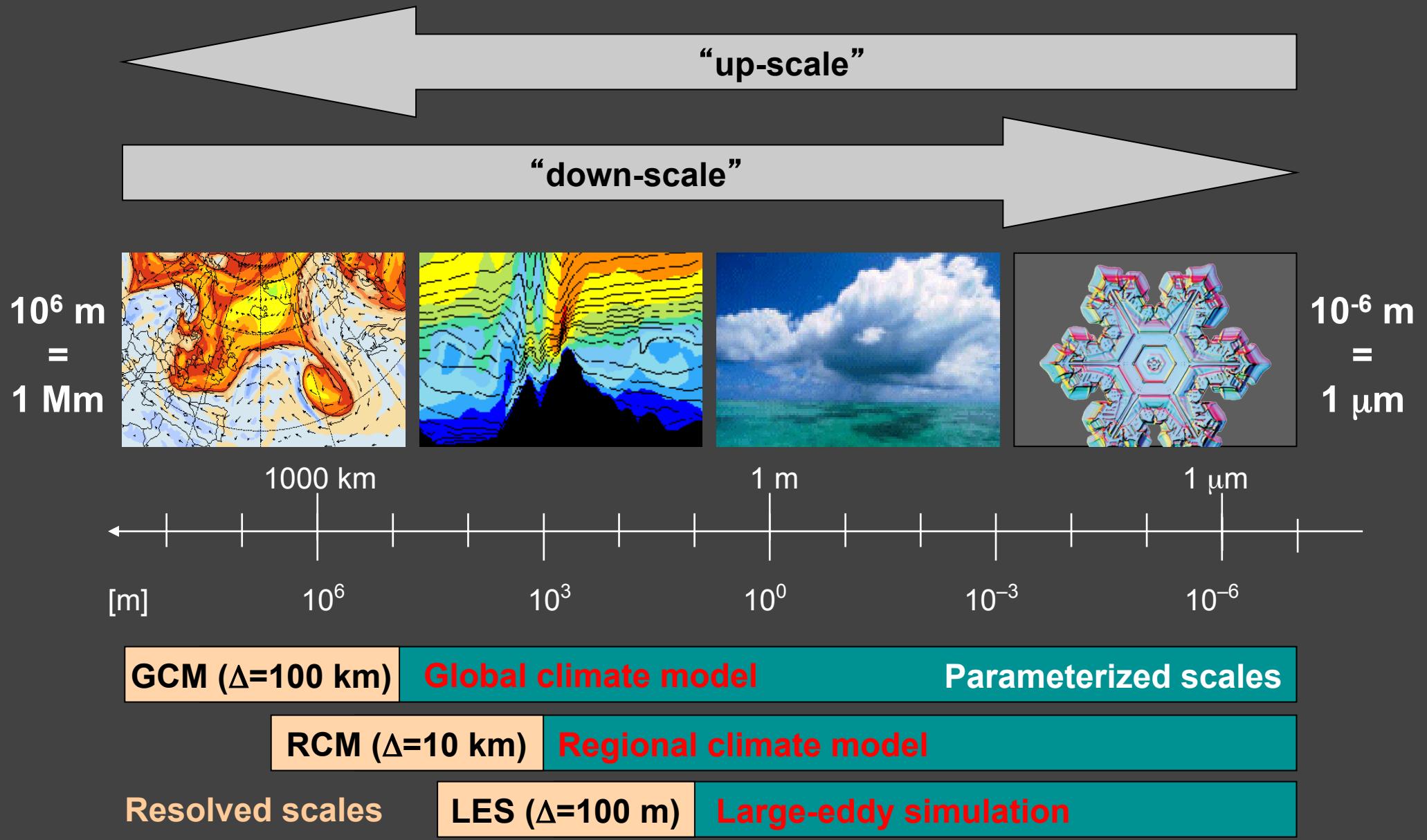
“up-scale”

Edward N. Lorenz (1917-2008)

**Does the flap of a butterfly's wings in Brazil
set off a tornado in Texas?**



Multi-scale interactions in the climate system



Outline

Historical perspective

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Ingredients to Deterministic Weather Forecasting

1. Governing set of equations
 - Equations describing the motion and thermodynamics of the atmosphere (“**dynamics**”).
 - Equations describing the interaction of the atmospheric flow with a wide range of physical processes (**parameterizations**, “**physics**”). This includes: radiation, boundary layer processes, turbulence, cloud microphysics, soil hydrology, etc.
2. Discretized form of equations on a computer (**model**)
3. Initial conditions to start integration
 - Sufficient measurements to initialize model (**observing system**)
 - Preparation of the observations on a computational grid (**data assimilation**)

“Euler Equations” in Cartesian Coordinates

Momentum equations

$$\frac{Du}{Dt} - fv = -\frac{1}{\rho} \frac{\partial p}{\partial x} + F_x \quad \frac{Dv}{Dt} + fu = -\frac{1}{\rho} \frac{\partial p}{\partial y} + F_y$$

$$\frac{Dw}{Dt} = -\frac{1}{\rho} \frac{\partial p}{\partial z} - g + F_z$$

with $\frac{D}{Dt} = \frac{\partial}{\partial t} + u \frac{\partial}{\partial x} + v \frac{\partial}{\partial y} + w \frac{\partial}{\partial z}$

Equation of state

$$p = \rho R T$$

Thermodynamic equation

$$\frac{DT}{Dt} - \frac{1}{c_p \rho} \frac{Dp}{Dt} = H$$

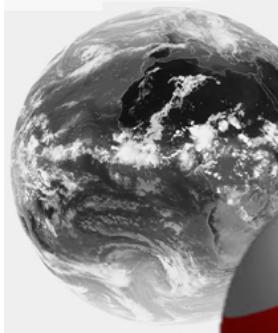
Continuity equation

$$\frac{\partial \rho}{\partial t} + \frac{\partial(u\rho)}{\partial x} + \frac{\partial(v\rho)}{\partial y} + \frac{\partial(w\rho)}{\partial z} = 0$$

Why not the full “Navier-Stokes Equations”?

Reynolds-number in atmospheric flows is far too large!

Anisotropy of Atmosphere



Diameter: 12,700 km

Depth of troposphere: 10 km (90% of atmosphere's mass)



Scaled to a Billiard ball:

Diameter: about 5 cm

Depth of troposphere: 0.04 mm

=> Implies $\Delta z \ll \Delta x, \Delta y$

Critical velocities

~300 m/s sound propagation

~100 m/s horizontal wind velocity

~20 m/s vertical gravity-wave (buoyancy-wave) propagation

Numerics: Courant-Friedrichs-Levy (CFL) stability criterion

$$\left| \frac{U \Delta t}{\Delta z} \right| \leq 1 \quad \text{where } U \text{ denotes largest velocity in system}$$

would require $\Delta t \leq 0.1$ s

Approach for large-scale models: Hydrostatic Approximation

Vertical momentum equation:

$$\cancel{\frac{d\psi}{dt}} = -\frac{1}{\rho} \frac{\partial p}{\partial z} - g + \cancel{F_z}$$

**Balance between pressure-force and gravity,
neglect vertical acceleration**

Implications:

- Suppresses vertical sound propagation
- w must be diagnosed from continuity equation (diagnostic variable)
- much easier to maintain time-step criterion
- BUT: only valid for $\Delta x > \sim 10$ km

Discretization on the Sphere

The simplest model grid on the sphere is a regular latitude / longitude grid.

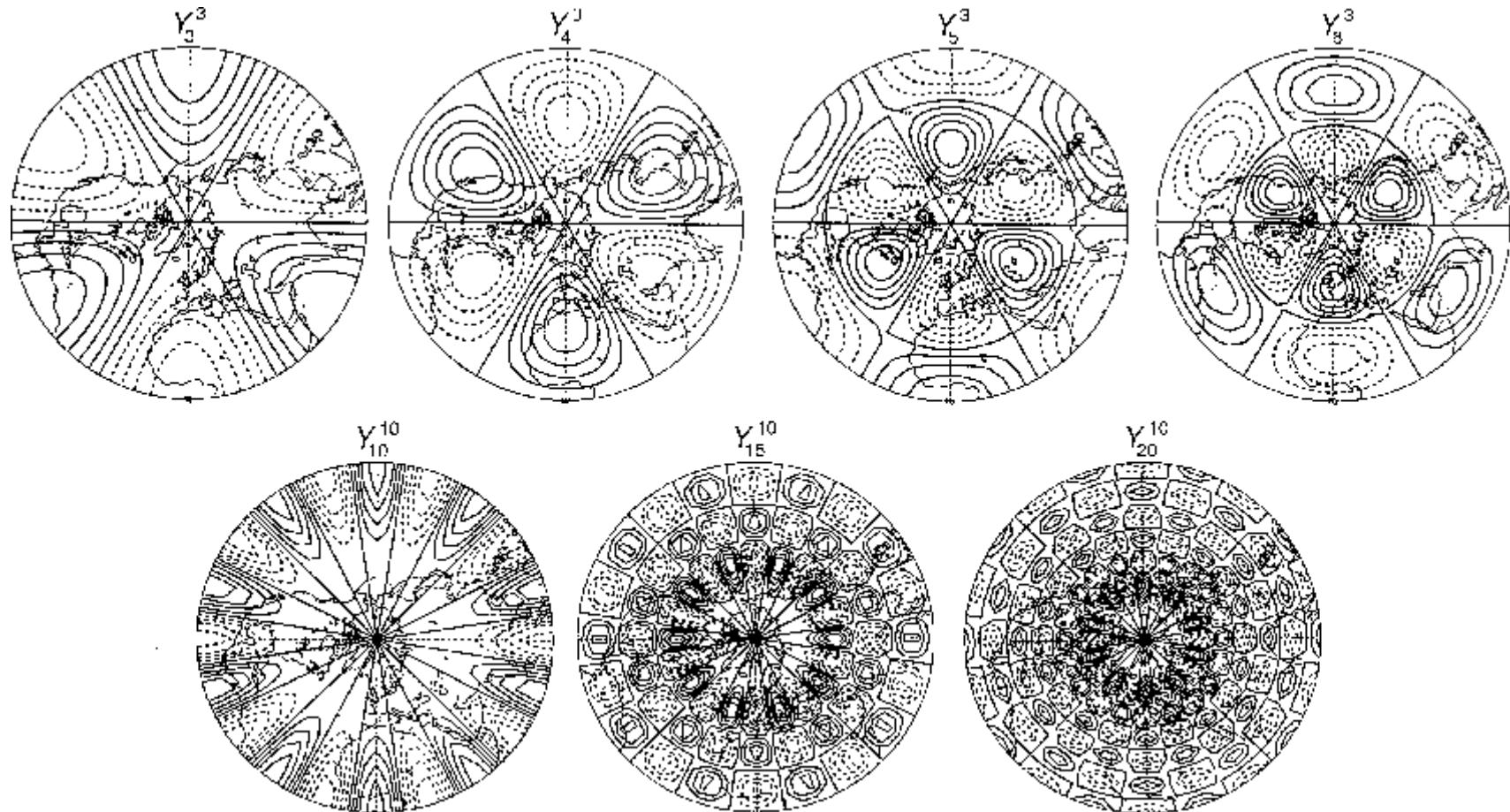
It has some difficult side effects (e.g. pole problem).



Global Spectral Models

Represent a two-dimensional field ϕ on the sphere as an expansion using spherical harmonics $Y_n^m(\lambda, \varphi)$:

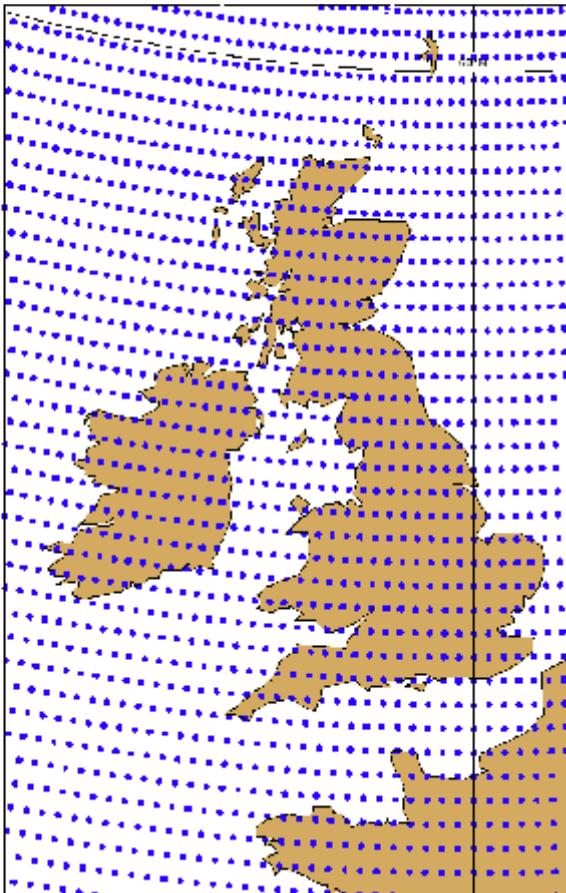
$$\phi(\lambda, \varphi, t) = \sum_{m=-M}^M \sum_{n=|m|}^{N(m)} \psi_n^m(t) Y_n^m(\lambda, \varphi)$$



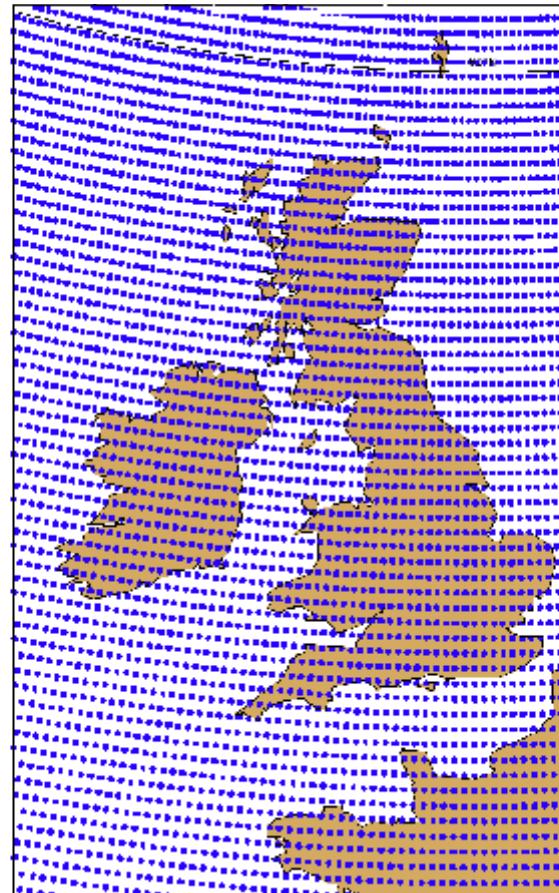
Example: ECMWF Model

Resolution Upgrades February 2006 and January 2010

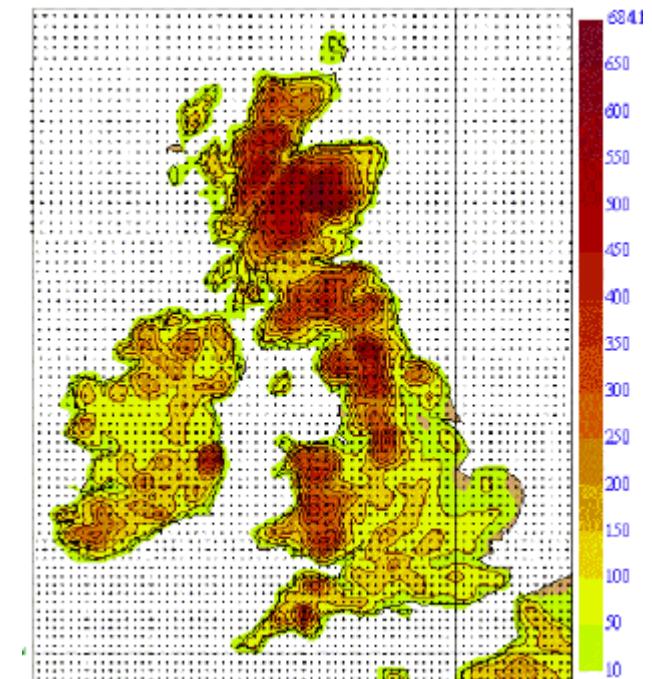
$T_L511 / N256 \sim 40 \text{ km}$
 (until Feb. 2006)
 348,528 grid points



$T_L799 / N400 \sim 25 \text{ km}$
 (since Feb. 2006)
 843,490 grid points



$T_L1279 / N640 \sim 16 \text{ km}$
 (since Jan. 2010)
 2,140,704 grid points



Global Models with variable Resolution

Action de
Recherche
Petite
Echelle
Grande
Echelle

ARPEGE (Meteo France)

Representation on the globe with the spectral approach, but using

- grid stretching
- pole rotation

to enhance resolution over the region of interest

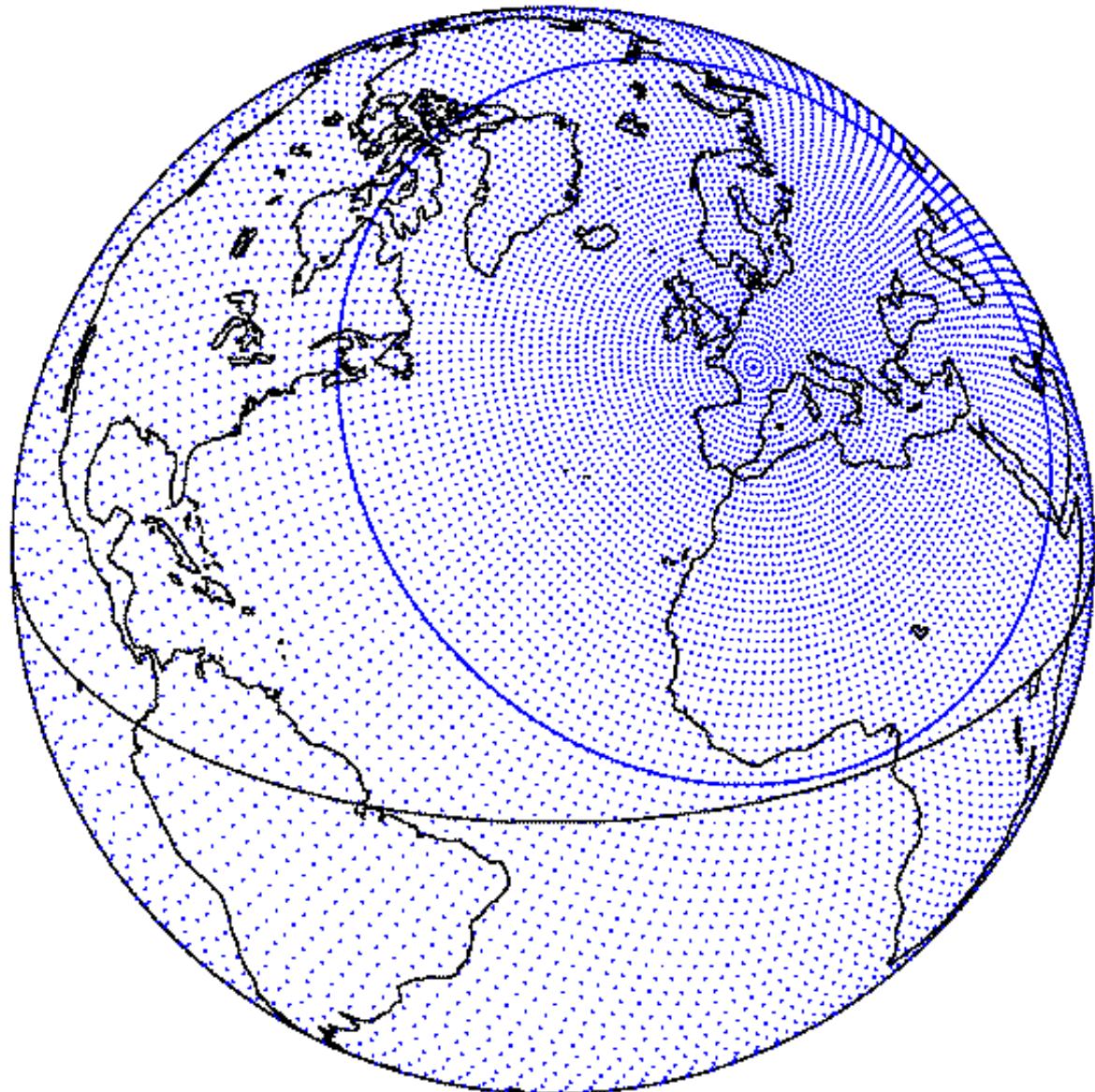
For example:

spectral resolution: $T_L 358$
stretching factor: C2.4

Resulting grid:

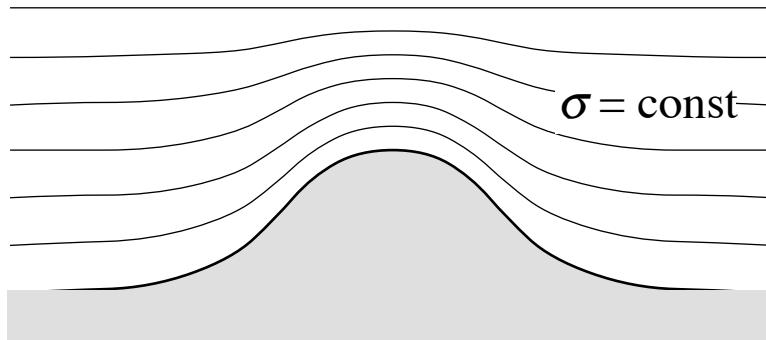
$\Delta = 23 \text{ km (France)}$

$\Delta = 133 \text{ km (antipodes)}$



Vertical Discretization: Terrain-Following Coordinates

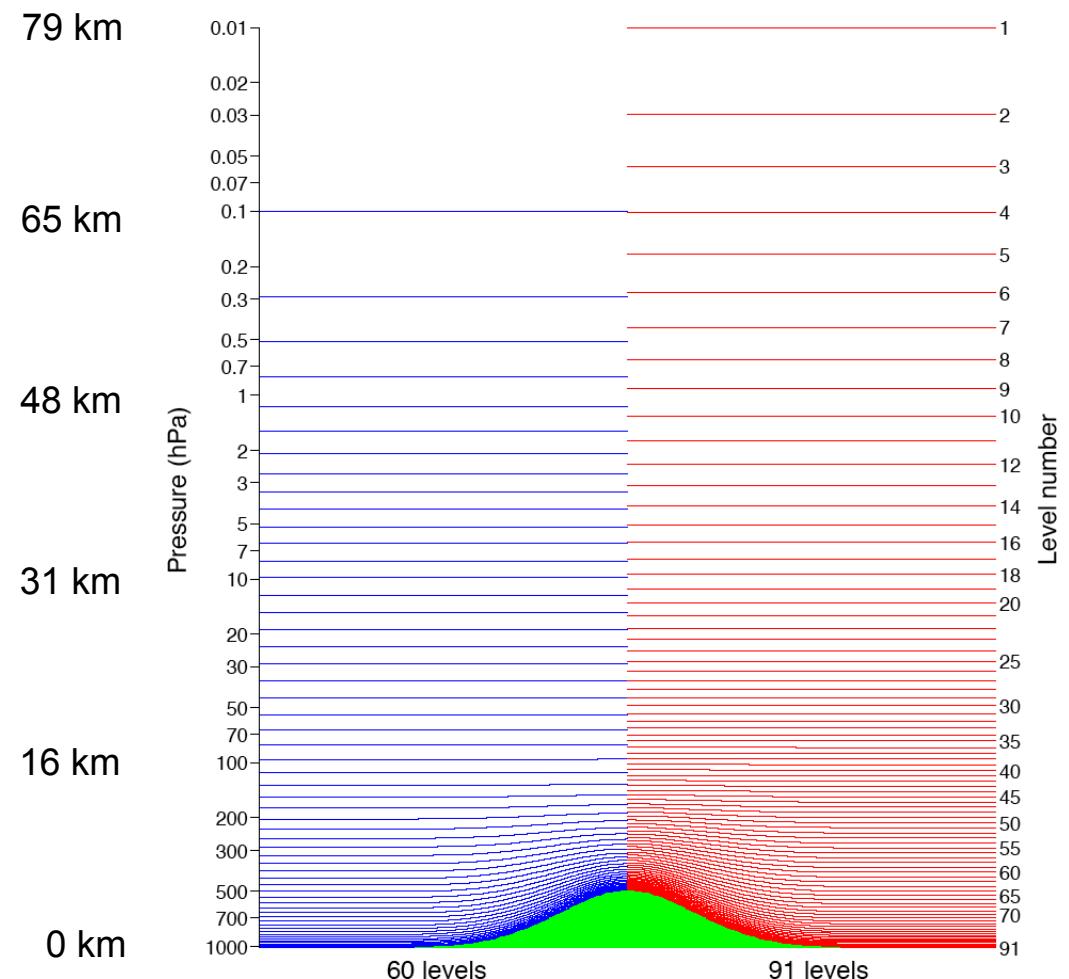
44



Typical vertical resolution in climate and numerical weather prediction models:
20-150 levels

Hydrostatic models use a pressure-based coordinate system

Current resolution at ECMWF since June 2013:
137 levels with top at 0.01 hPa (ca 79 km)



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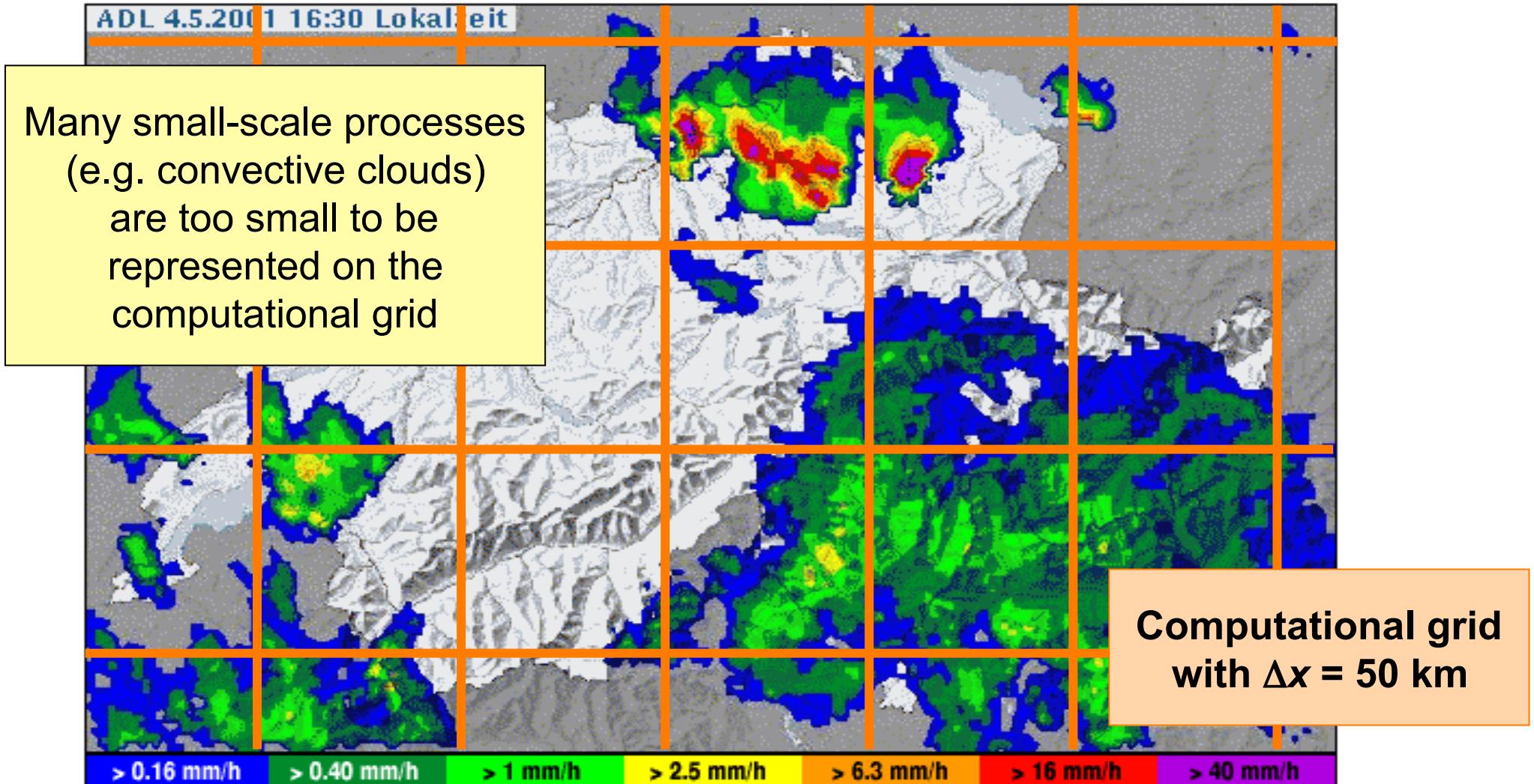
Parameterizations

Initial conditions

Climate models

Computational aspects

The Parameterization Problem



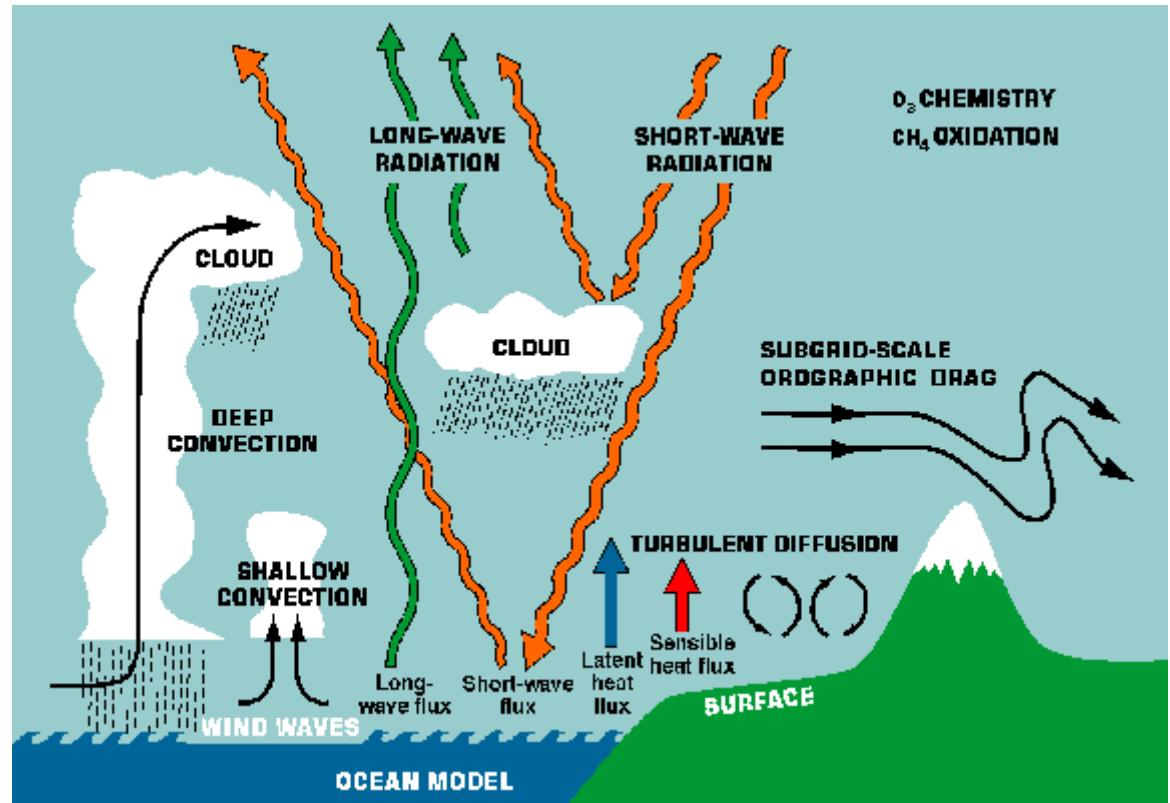
Parameterized Processes

Typical atmospheric models have

$$10 \text{ km} \leq \Delta x \leq 200 \text{ km}$$

Processes that are not explicitly represented at these resolutions are "**parameterized**" instead, using physical understanding of the underlying processes, or semi-empirical relations.

Parameterized processes contribute substantially to uncertainties in weather forecasting and climate models.



Parameterized Processes

Momentum equations

$$\frac{Du}{Dt} - fv = -\frac{1}{\rho} \frac{\partial p}{\partial x} + F_x \quad \frac{Dv}{Dt} + fu = -\frac{1}{\rho} \frac{\partial p}{\partial y} + F_y$$

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Continuity equation

$$\frac{\partial \rho}{\partial t} + \frac{\partial(u\rho)}{\partial x} + \frac{\partial(v\rho)}{\partial y} + \frac{\partial(w\rho)}{\partial z} = 0$$

Parameterized terms

$(F_x, F_y)/\rho$ non-conservative forces

H/c_p diabatic heating rate

Additional equations

$$\frac{Dq_{vap}}{Dt} = S_{vap} \quad \frac{Dq_{cld}}{Dt} = S_{cld}$$

Equations for specific water vapour and cloud water content

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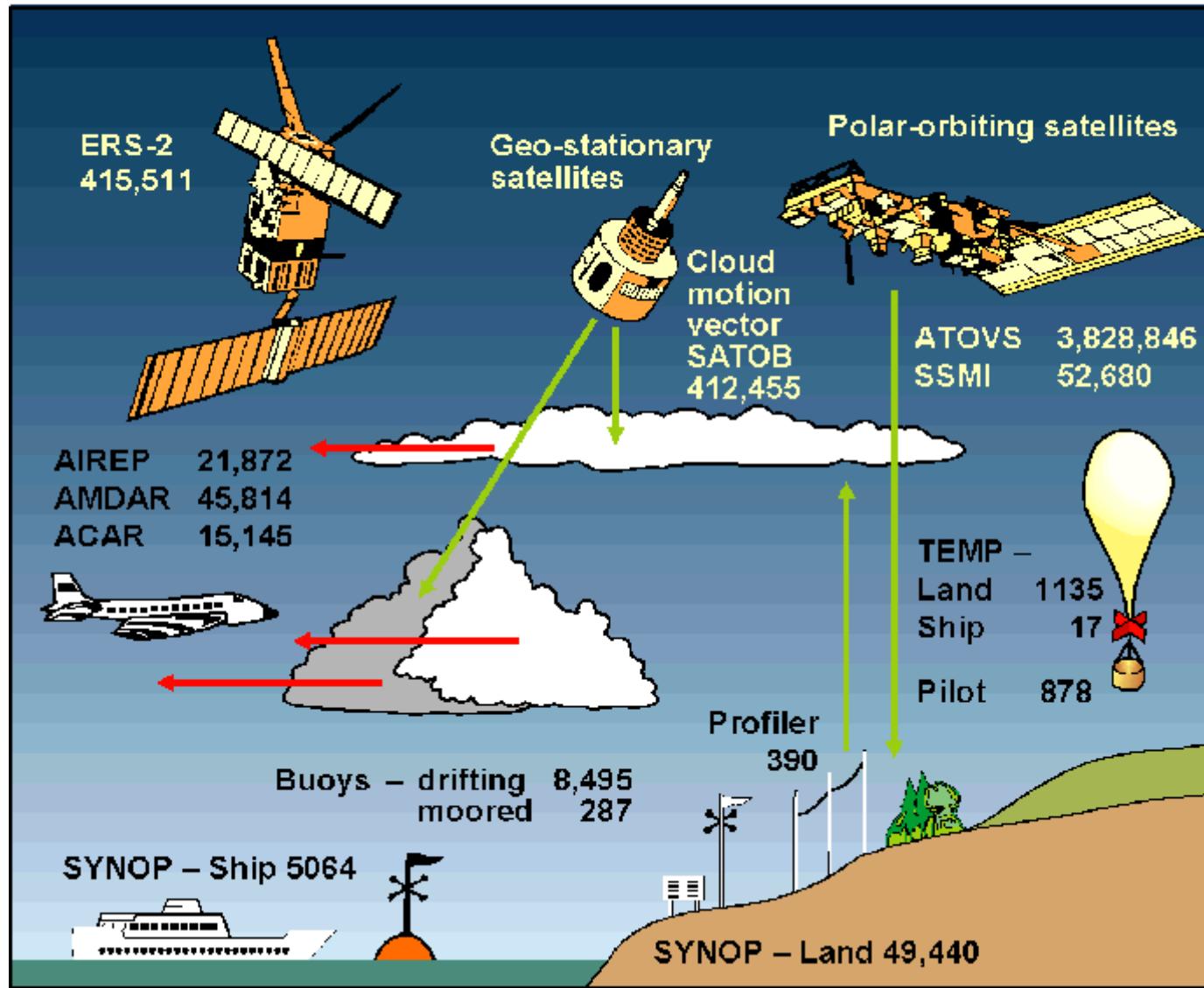
Climate models

Computational aspects

Initial conditions are obtained from observations

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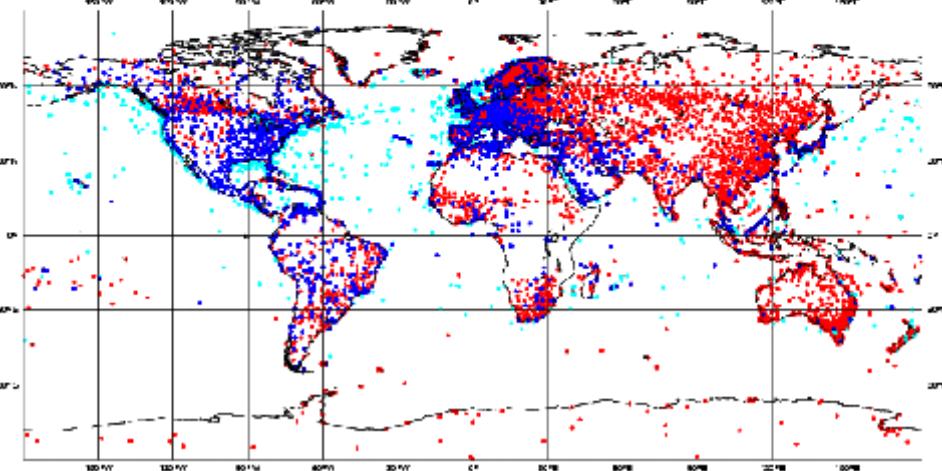
24 hour summary of observations received at ECMWF, 18 March 2000



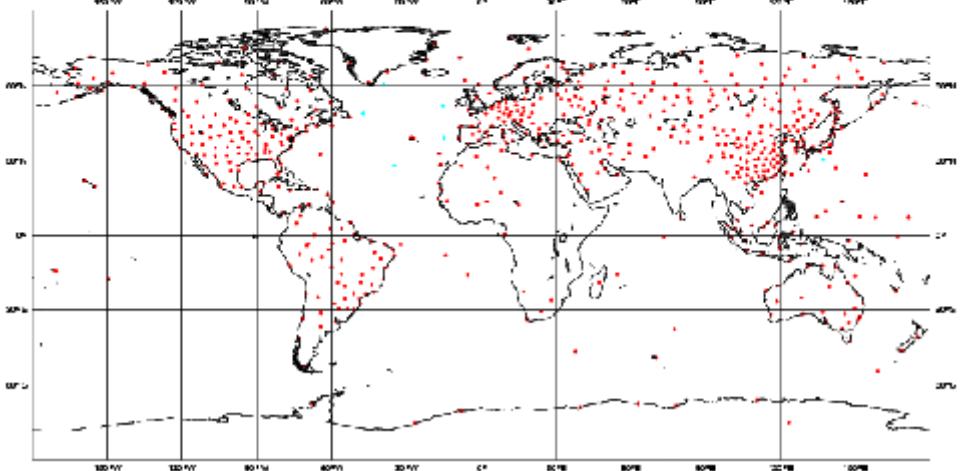
(ECMWF 2001)

Global observing system

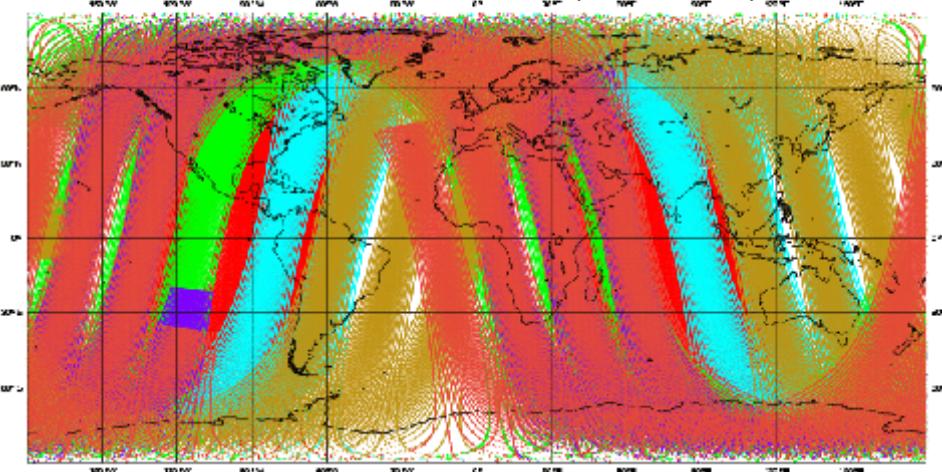
Surface observations (33,904)



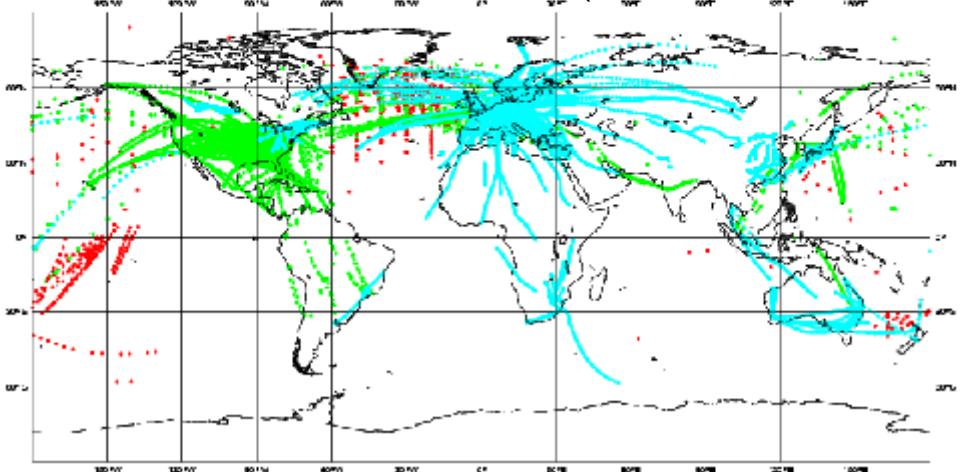
Radiosonde observations (641)



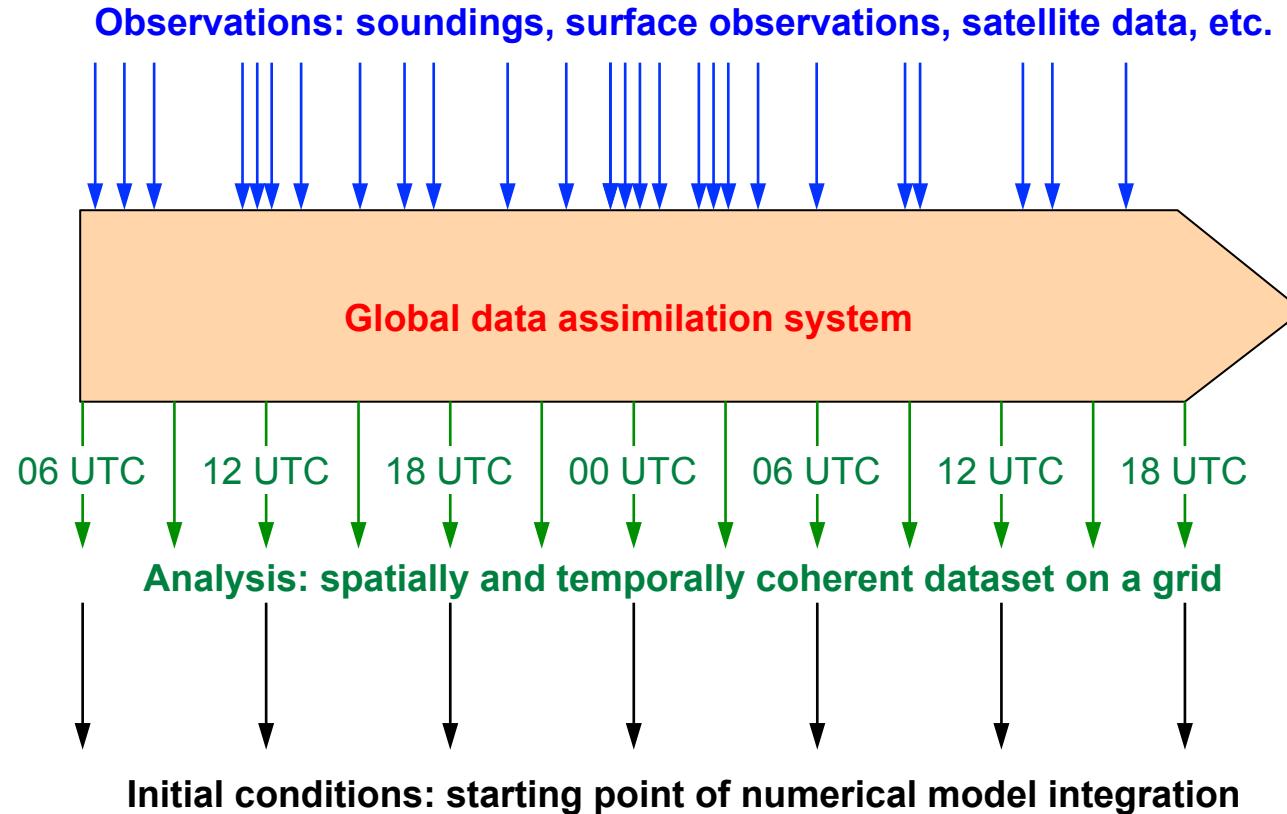
MW / IR satellite data (1,343,053)



Aircraft observations (46,037)



Global Data Assimilation



Global data assimilation systems ingest a wide range of data from various instruments and observation times. They run a general circulation model (GCM) in hindcast mode. The resulting analysis is a spatially and temporally coherent description of the actual state of the atmosphere. In data sparse regions, where few observations are available, these systems in essence provide a mixture between a short-range (e.g. 6 h) forecast and the available observations.

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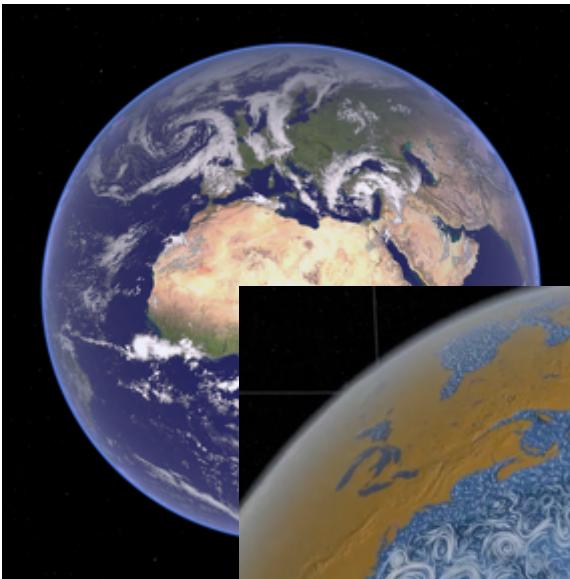
Parameterizations

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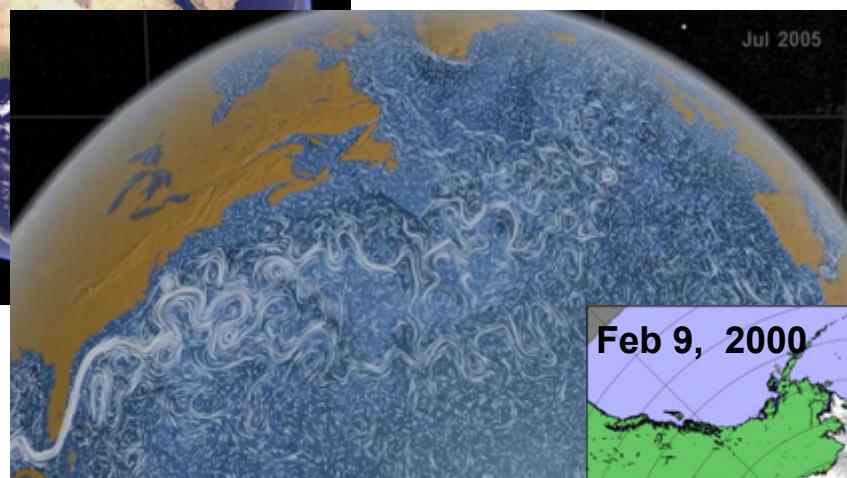
Climate models

Computational aspects

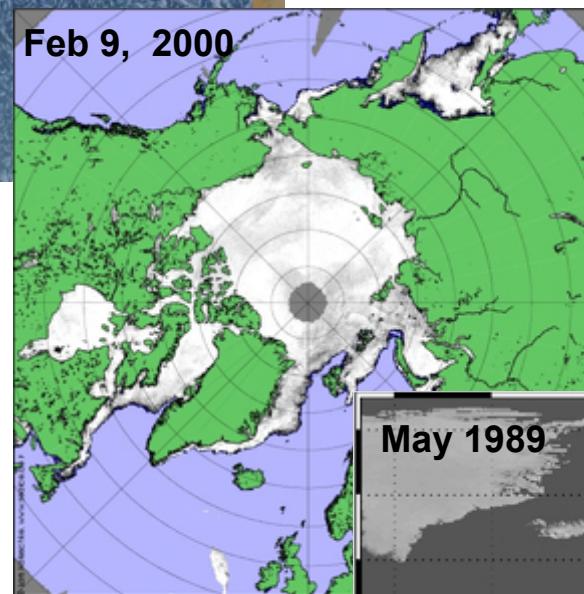
The Coupled Climate System



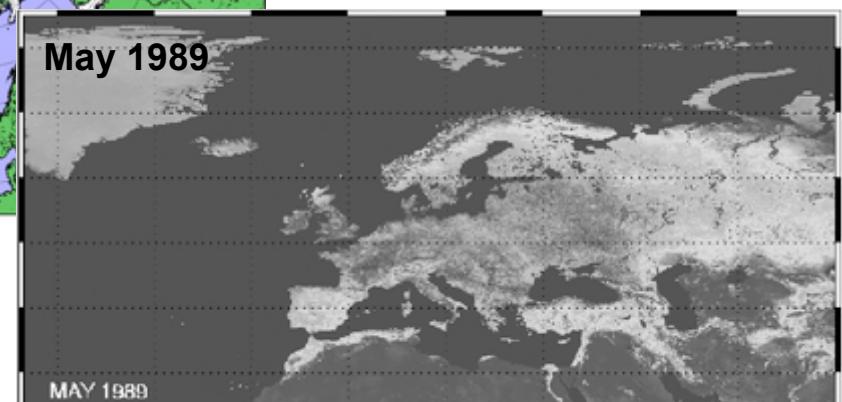
Atmosphere



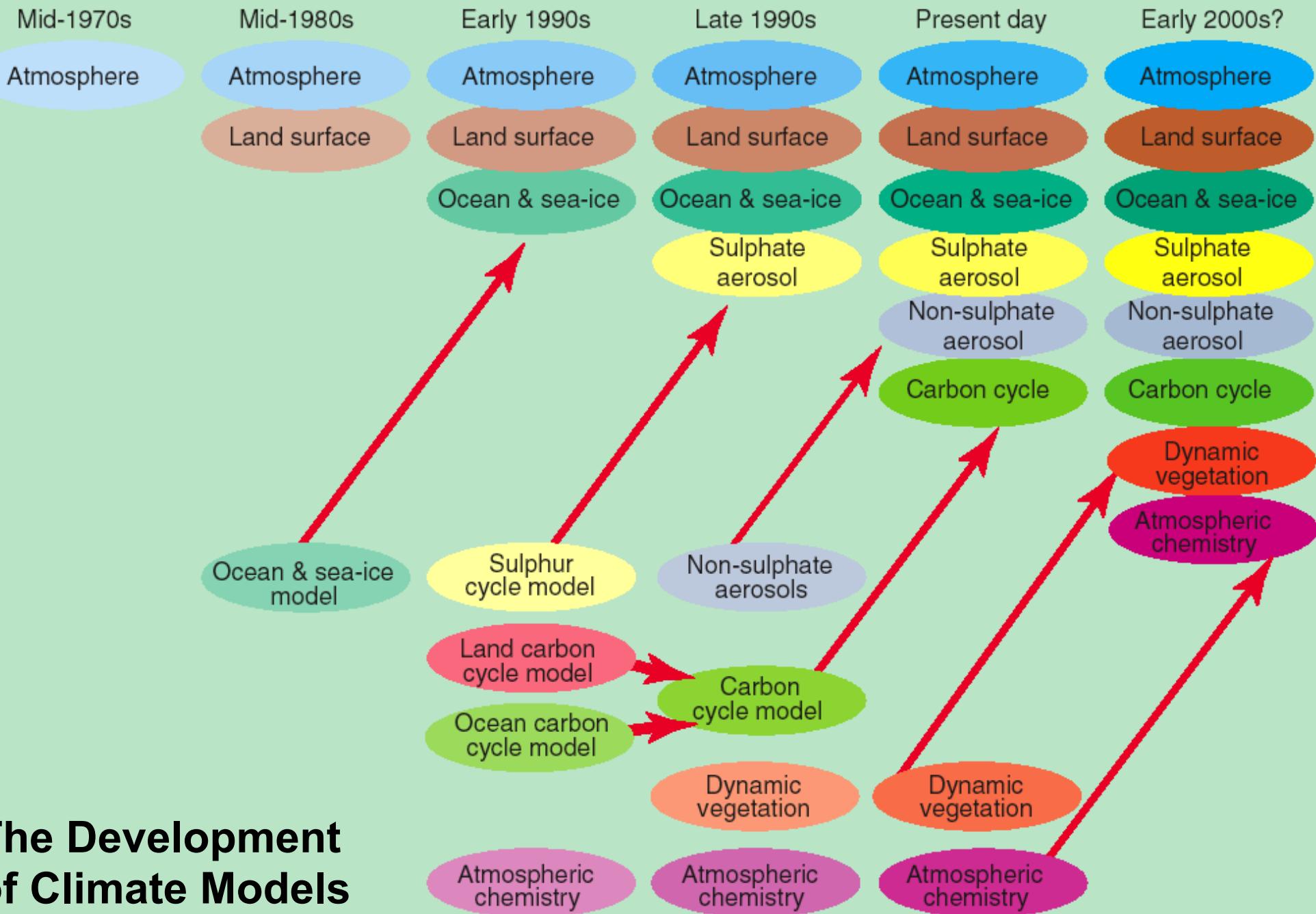
Ocean



Sea Ice



Land Surfaces

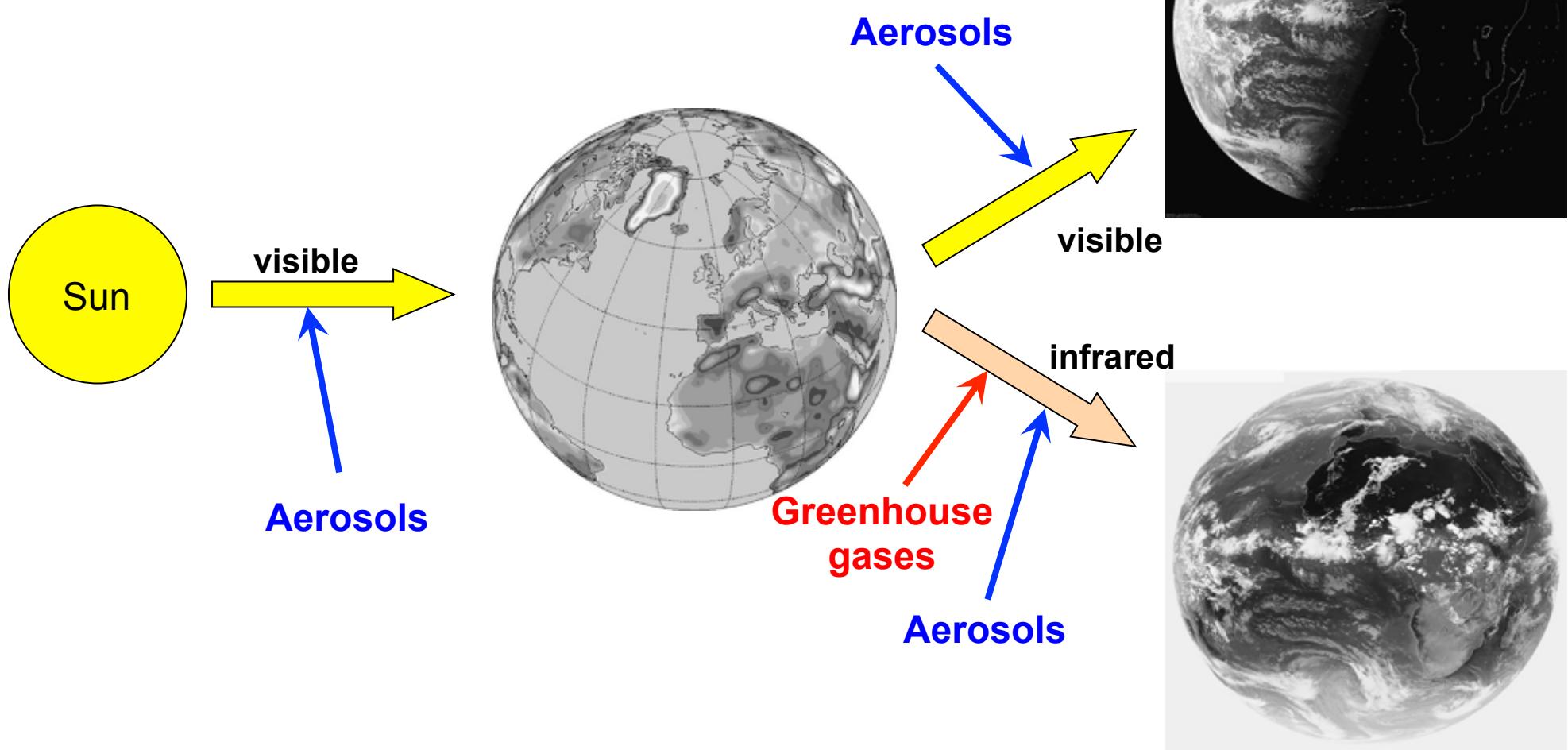


The Development of Climate Models

The Global Energy Balance

Energy Input = Energy Output

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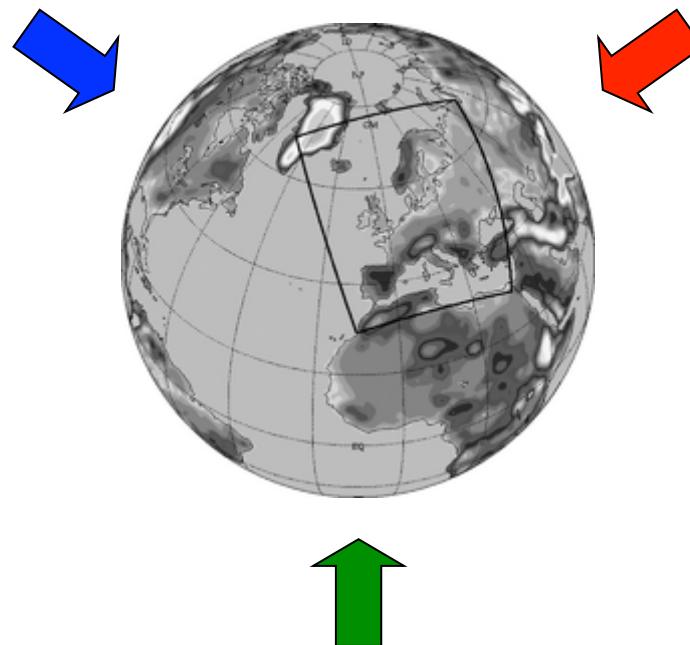
Sources of climate variations

External forcing

- Solar variability •
- Volcanic forcing •
- Orbital changes •

Anthropogenic forcing

- Emissions of greenhouse gases •
- Emissions of aerosol •
- Emissions of reactive gases (e.g. CFCs) •
- Land-surface changes •



Internal variability

- internal oscillation within the climate system •

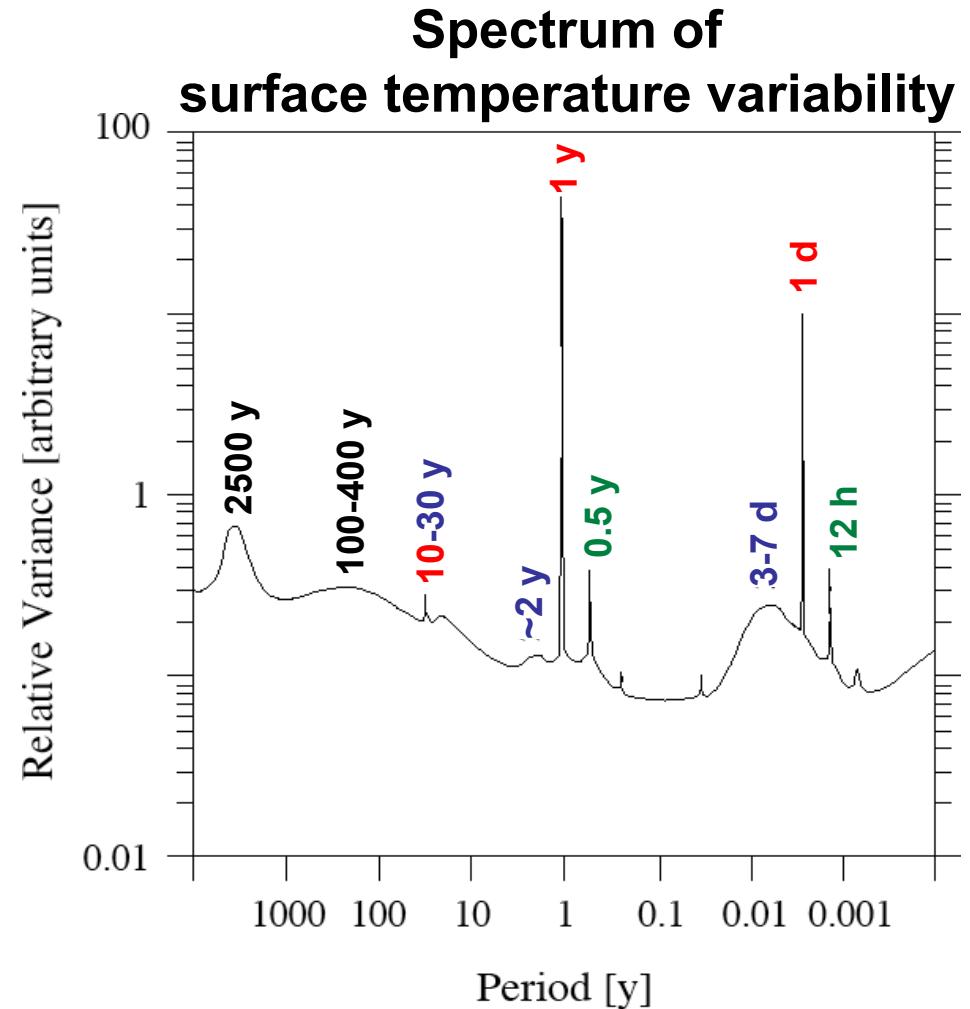
Internal variability

Observed climate records include

- response to forcings (natural and anthropogenic)
- internal variations (e.g. due to atmospheric variability)

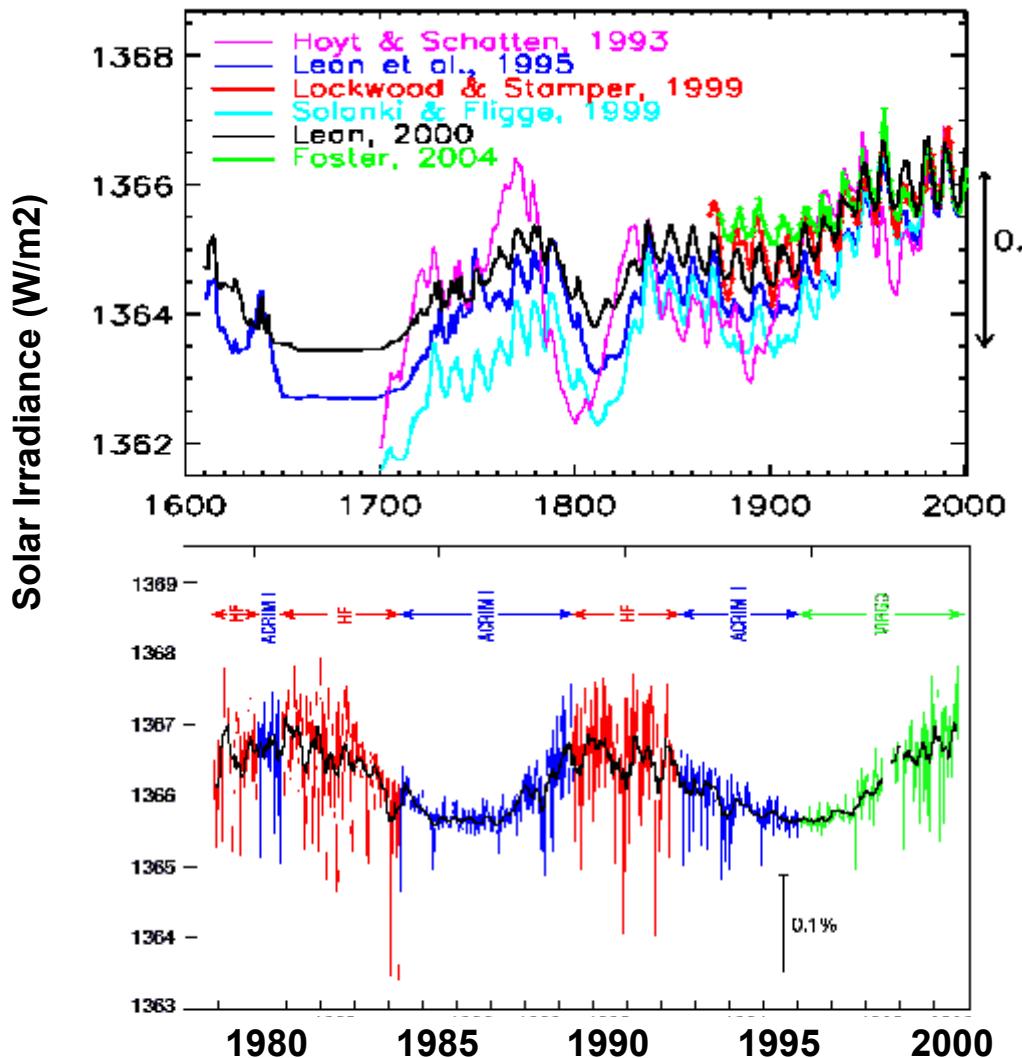
Some of the observed variations are due to:

- external forcings
- harmonics of external forcings
- internal variations

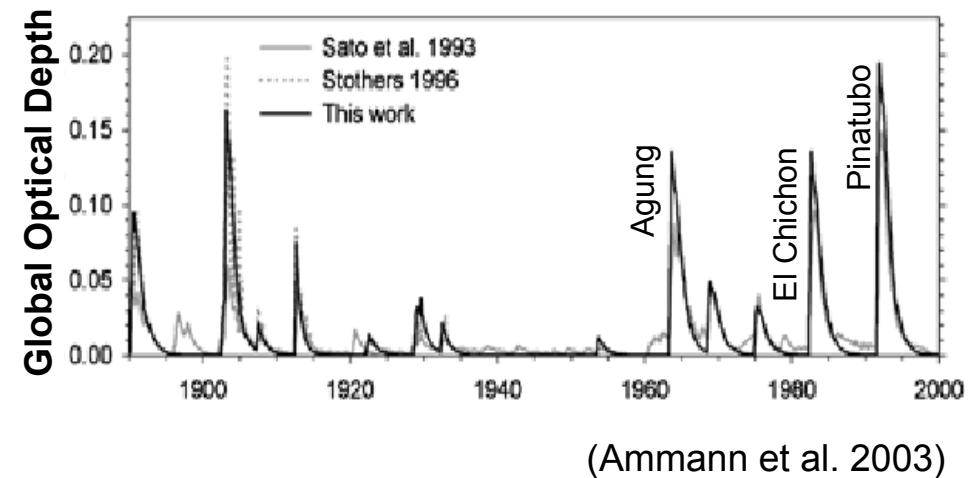


Natural Climate Forcings

Solar Forcing



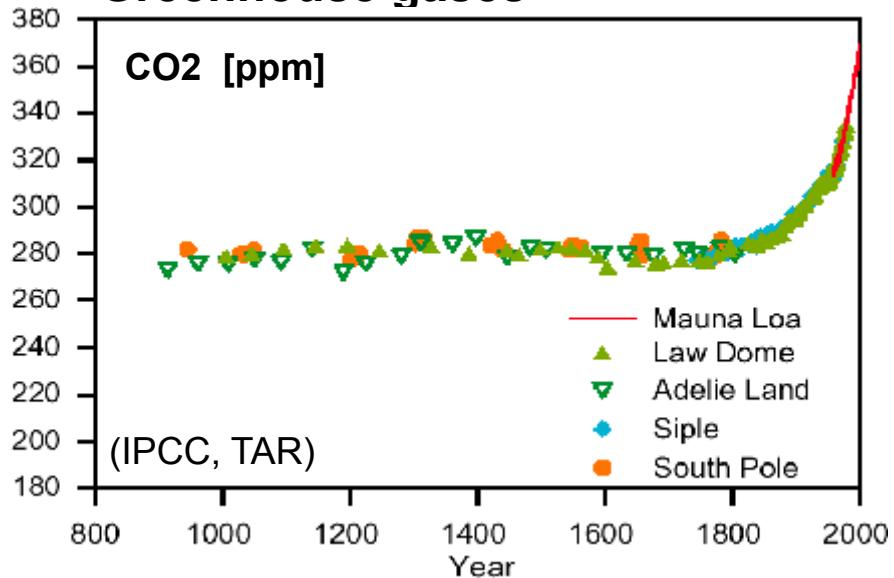
Volcanic Forcing



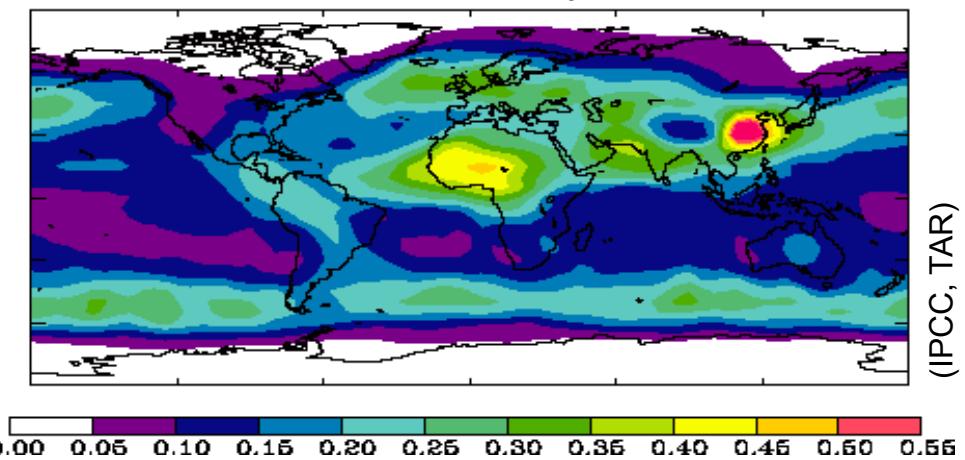
(Ammann et al. 2003)

Anthropogenic Climate Forcing

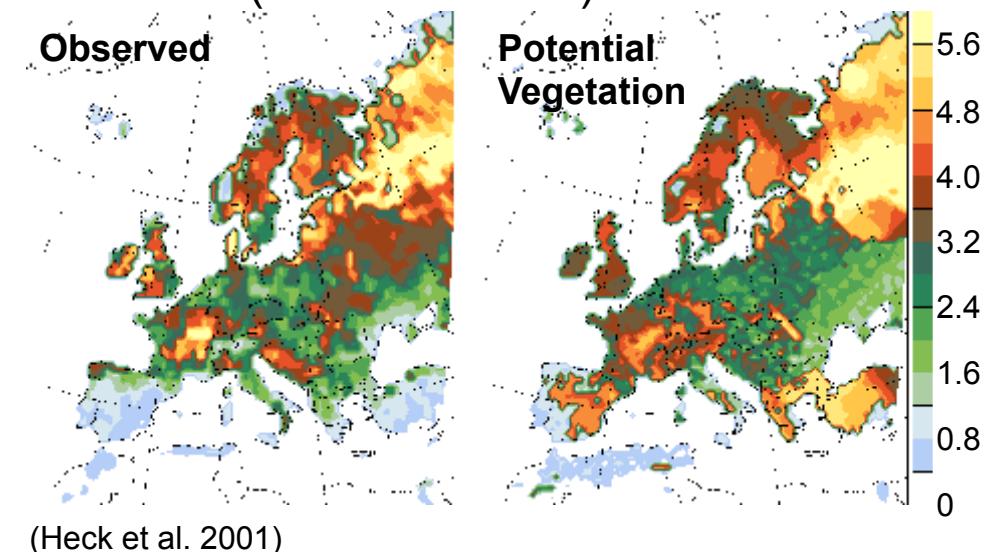
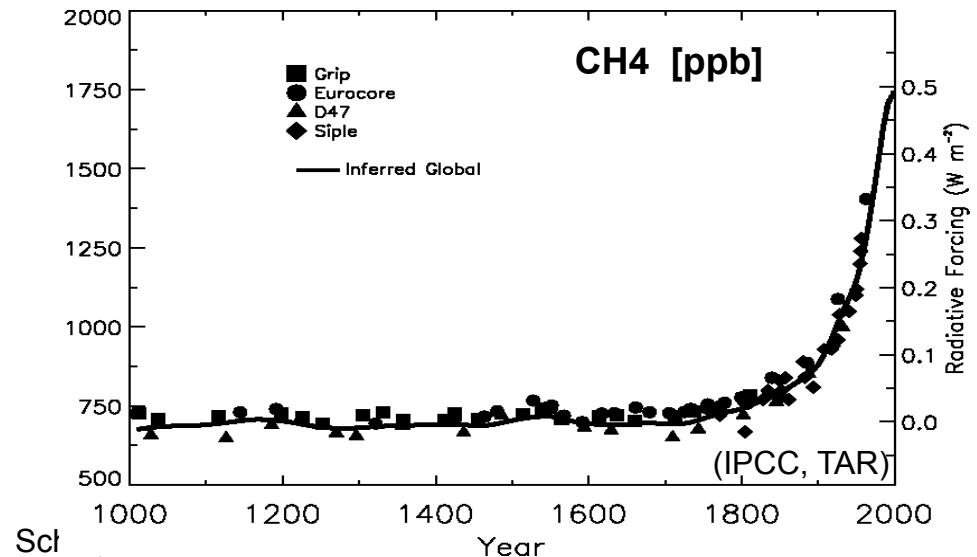
Greenhouse gases



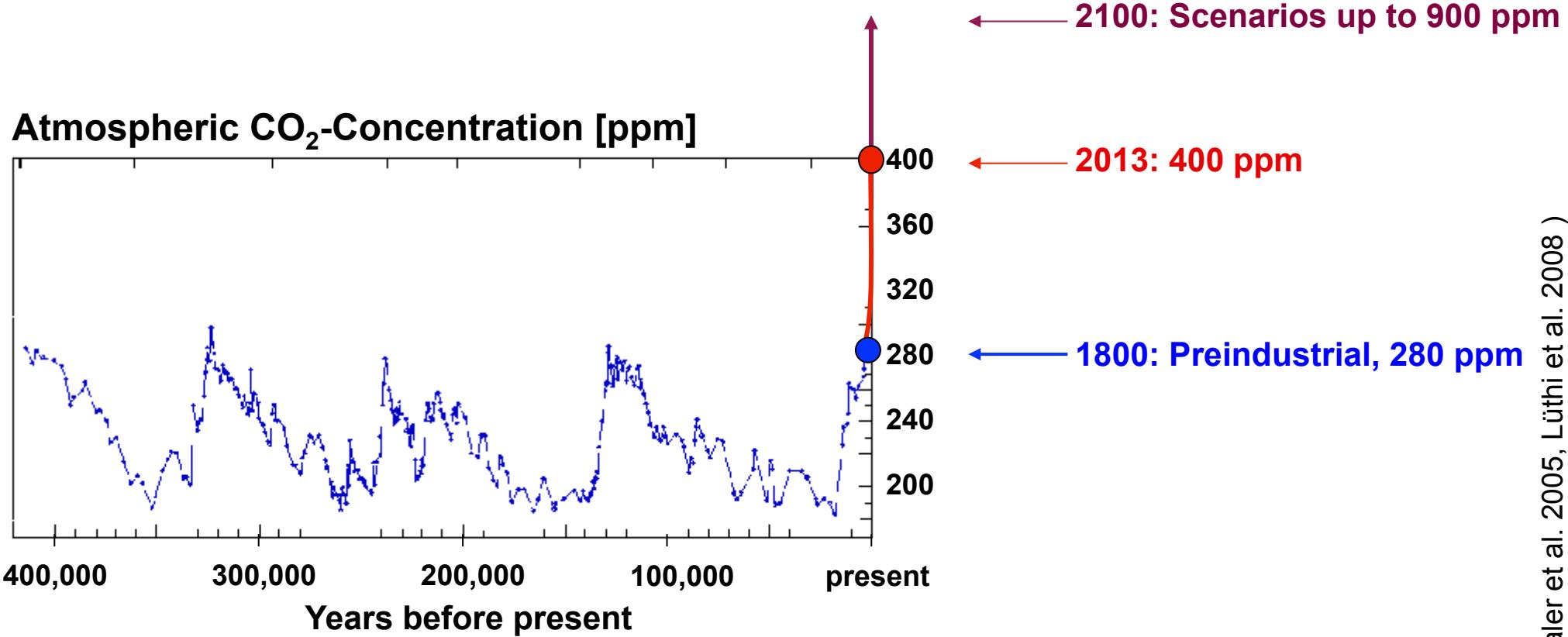
Aerosols (optical depth, year 2000)



Land use (leaf area index)

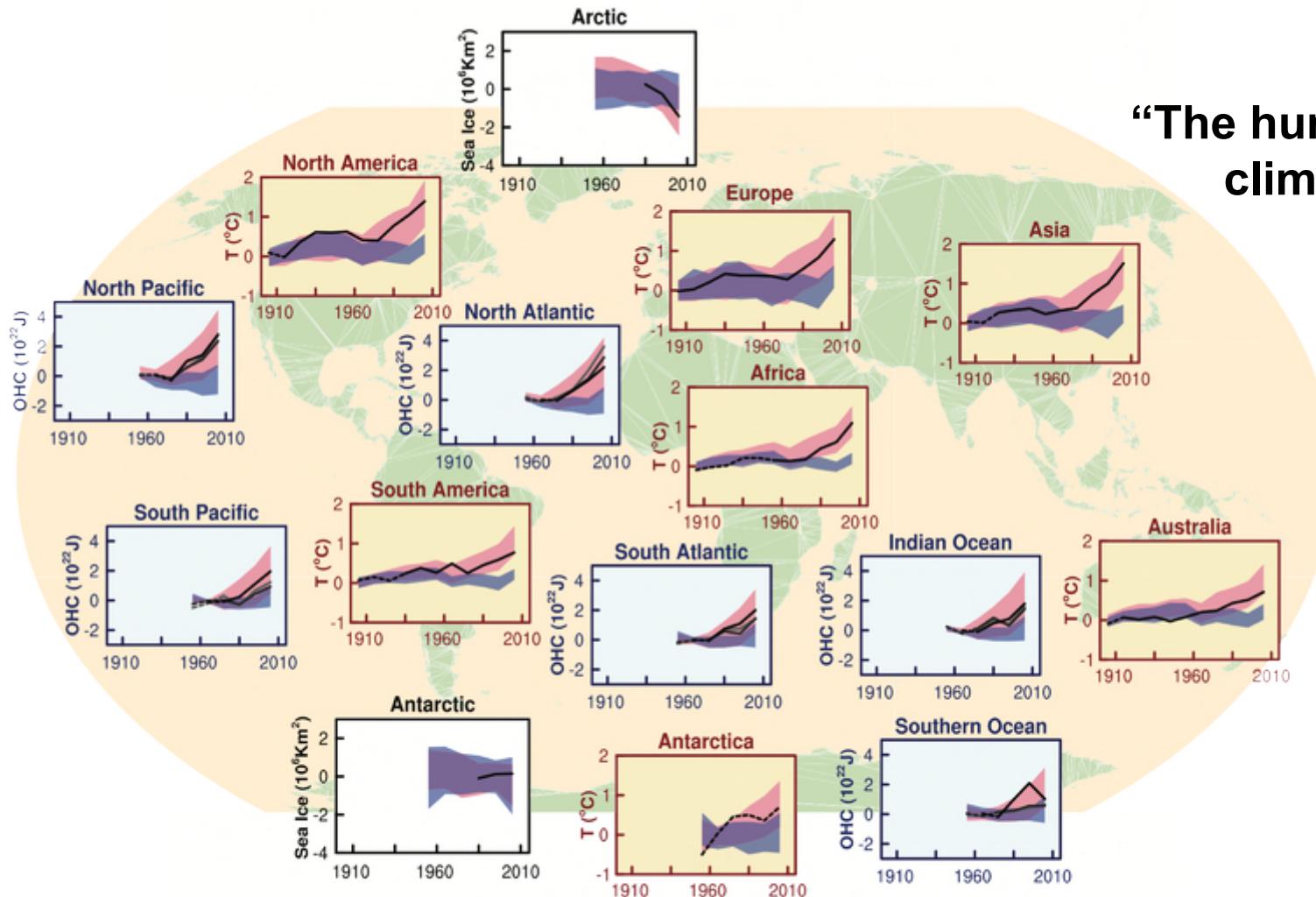


Carbon Dioxide (CO_2)



Current CO_2 concentrations are higher than ever in the last 800,000 years

Models with and without anthropogenic effects



“The human influence on the climate system is clear.”
IPCC AR5 (2013)

Outline

Historical perspective

Atmospheric scales

Governing equations

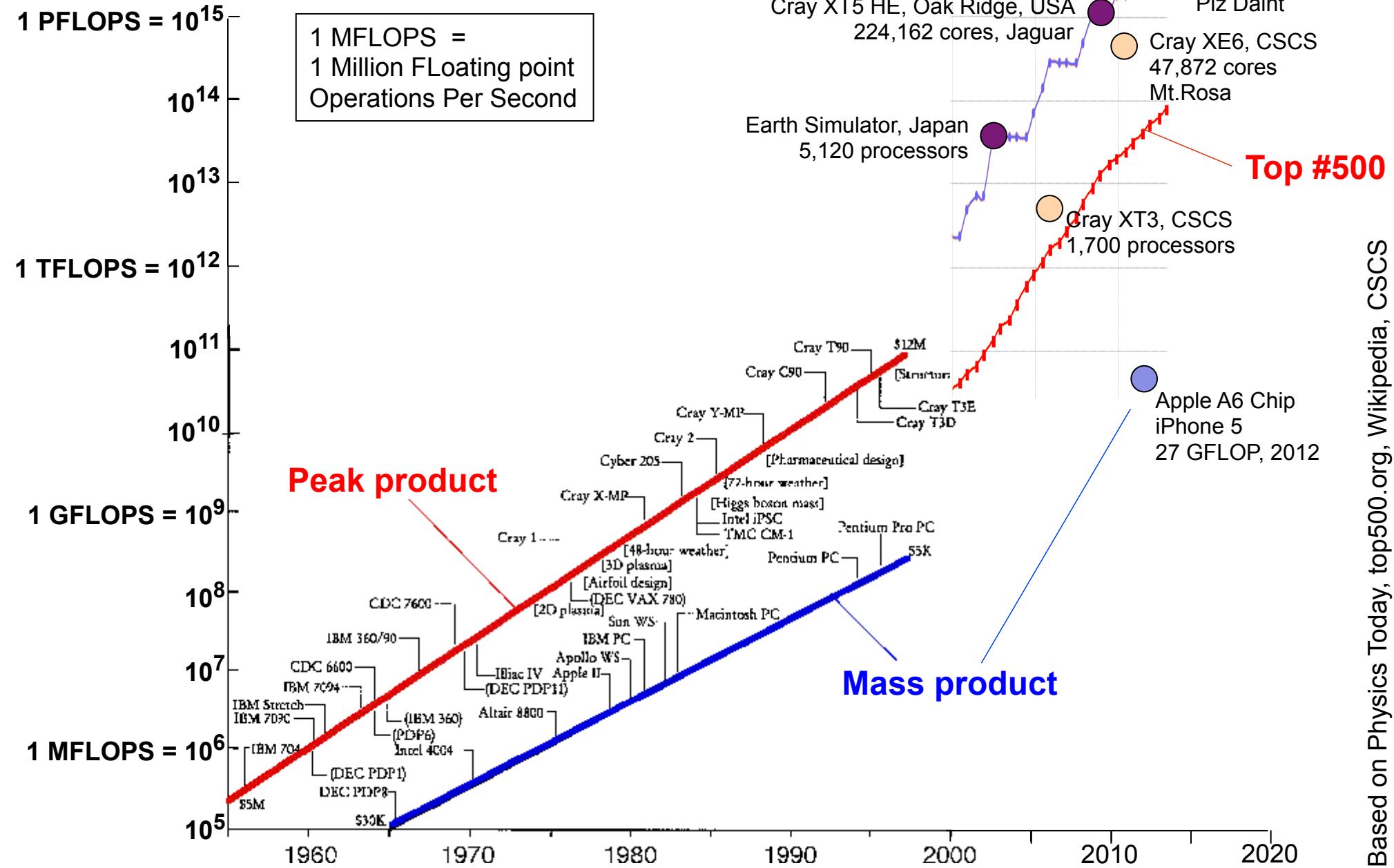
Parameterizations

Initial conditions

Climate models

Computational aspects

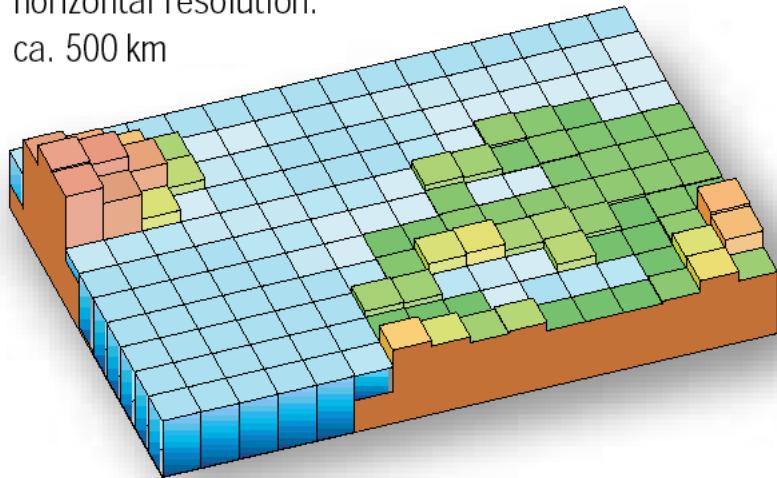
Development of CPU power



Horizontal Resolution

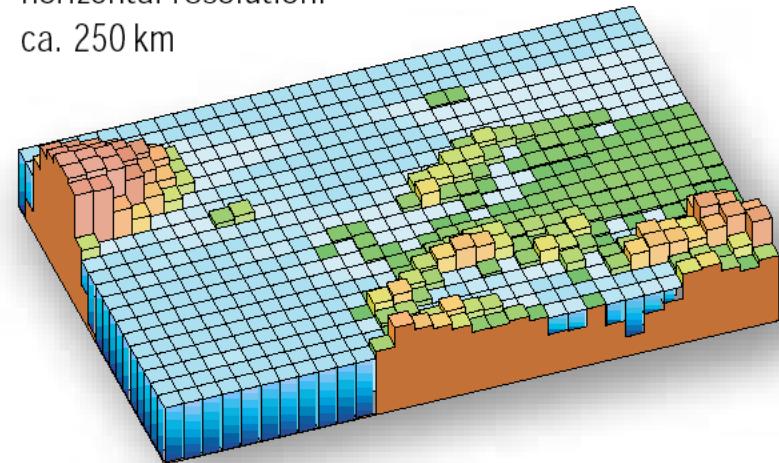
Model T21

horizontal resolution:
ca. 500 km



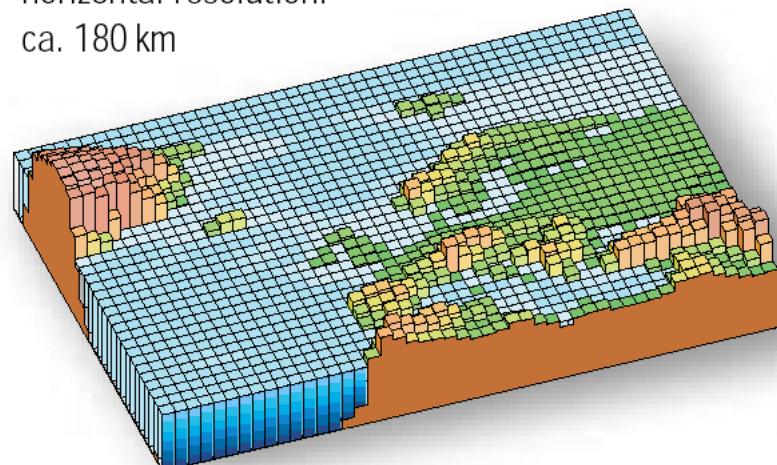
Model T42

horizontal resolution:
ca. 250 km



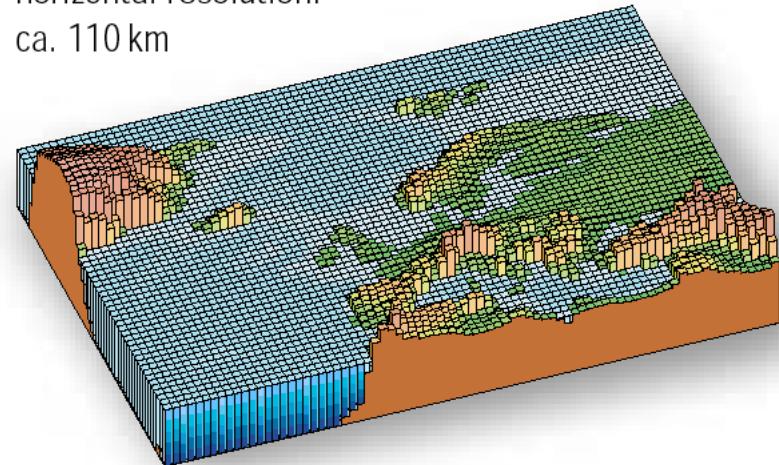
Model T63

horizontal resolution:
ca. 180 km



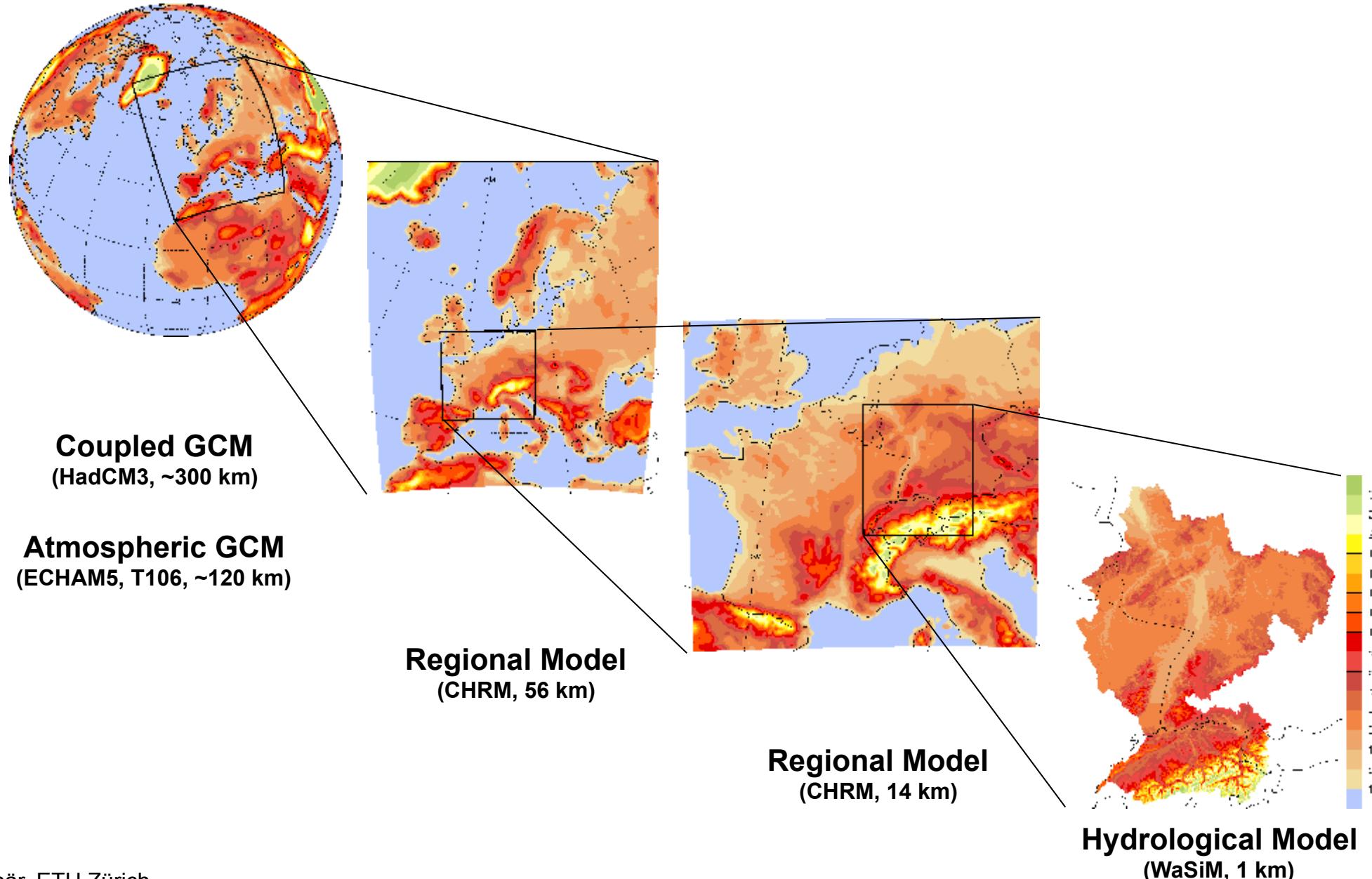
Model T106

horizontal resolution:
ca. 110 km



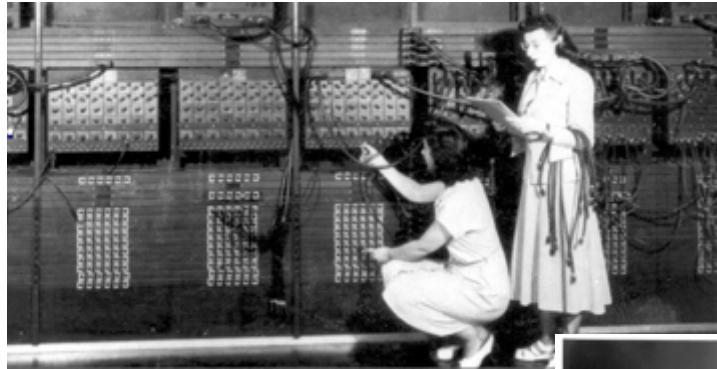
Model Chain for Climate Change Impact Study

95



Computational power matters

1946: Eniac



A few hundreds operations per second:

- plus / minus: 0.2 ms
- multiplication: 2.8 ms
- square root: 300 ms

Speed around 10^3 floating point operations per second (1 kFlop/s)



John von Neumann, 1903-1957
(studied chemistry at ETH)

An Apple iPhone 5 has about 25 GFlop/s = 25×10^9 Flops/s

Schär, ETH Zürich

Today: Cray XC30 at CSCS (Lugano)



Linpack benchmark: 6×10^{15} Floating point operations per second (6 PFlop/s).

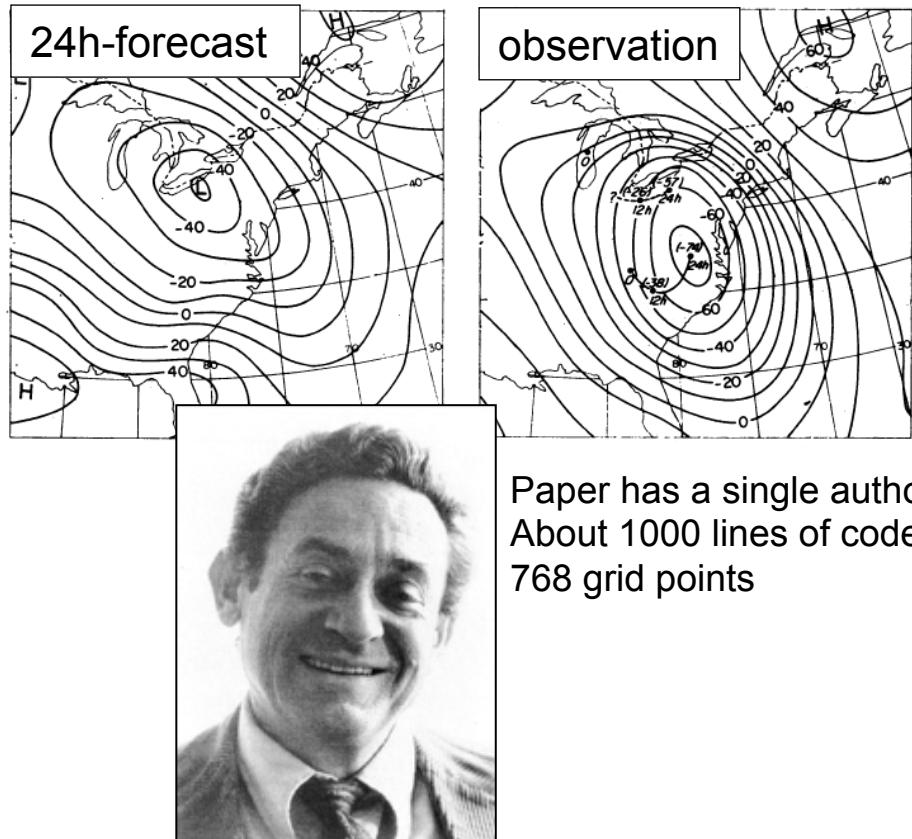
Currently fastest computer (Tianhe-2 “MilkyWay-2”), located in Guangzhou, China, is about 5 times faster.

“I think there is a world market for maybe five computers.”

*Thomas Watson, 1943
Chairman of IBM*

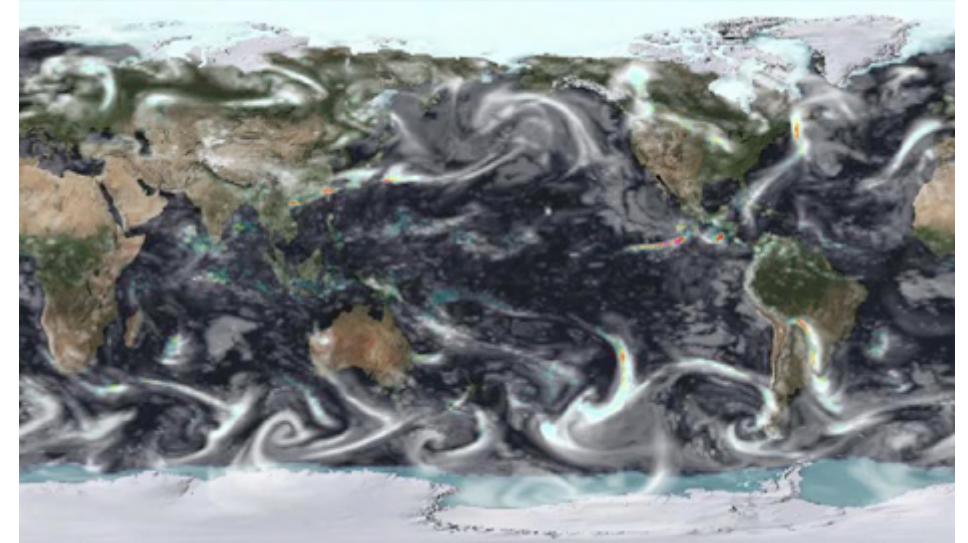
Model complexity matters

1954: first numerical weather forecast



Jule G. Charney, 1917-1981

Today: high-resolution climate



(Vidale et al., UK-Japanese collaboration)

Code of a few hundred authors,
about 1 million lines of source code,
about 120 million grid points

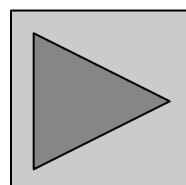
European-scale cloud-resolving simulations

- Enabled by GPU-capable COSMO prototype, has rewritten dynamical core, able to run on heterogeneous many-core architectures.

Collaborative effort between MeteoSwiss (Oliver Fuhrer), ETH / C2SM and CSCS



- Use for climate simulations in PhD of David Leutwyler:
 $\Delta=2.2$ km, 1500 x 1500 x 60 grid points, >10 years of integration
- Code runs since January 2014 on PizDaint (Cray hybrid XC30) at CSCS (Lugano)
 Implementation uses 144 nodes, each with 1 CPU and 1 GPU (Kepler K20X)
 Able to run 1 day in 18 minutes wall-clock time (1 year in about 5 days)
- Current test case: Storm Kyrill
 initialized January 16, 2007
 7-day simulations



Summary

The simulation of weather and climate has come a long way towards realizing the visions of V. Bjerknes (1904) and L.F. Richardson (1922).

Some key issues:

- Small-scale processes are important for large scale features:
=> High spatial resolution desirable (computational challenge)
- Representation of sub-gridscale processes:
=> Improved parameterizations and understanding needed
=> Some parameterizations may be replaced by explicit treatment (convection)
- Nonlinearities imply intrinsic predictability limitations:
=> Probabilistic methods
- Weather versus climate
=> Weather models can be validated against observations
=> Only limited observations on decadal climate variations available