

How is temperature forecast at ECMWF?

Annelize van Niekerk

... and the Physical Processes Team

Annelize.vanNiekerk@ecmwf.int

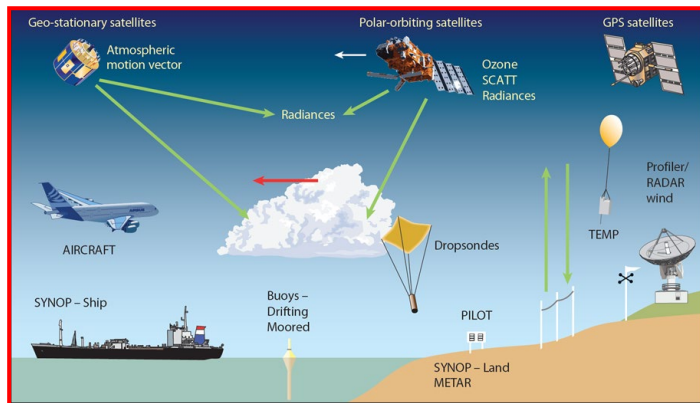
The ECMWF physical model

125 km

40 km

25 km 25 km

Data assimilation



Large-scale resolved dynamics

$$\frac{\partial T}{\partial t} + \frac{1}{a \cos^2 \theta} \left\{ U \frac{\partial T}{\partial \lambda} + V \cos \theta \frac{\partial T}{\partial \theta} \right\} + \eta \frac{\partial T}{\partial \eta} - \frac{\kappa T_v \omega}{(1 + (\delta - 1)q)p} = \boxed{P_T} + K_T$$

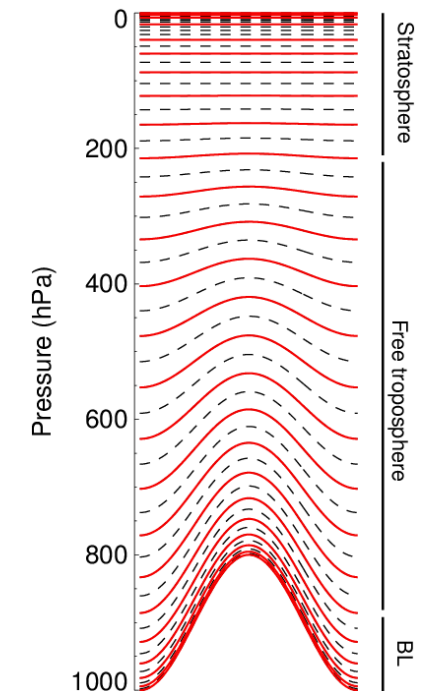
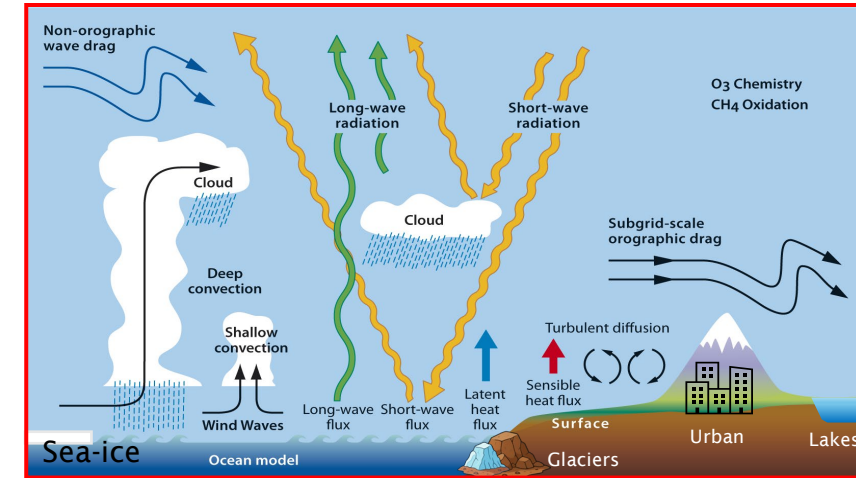
$$\frac{\partial q}{\partial t} + \frac{1}{a \cos^2 \theta} \left\{ U \frac{\partial q}{\partial \lambda} + V \cos \theta \frac{\partial q}{\partial \theta} \right\} + \eta \frac{\partial q}{\partial \eta} = \boxed{P_q} + K_q$$

$$\frac{\partial}{\partial t} \left(\frac{\partial p}{\partial \eta} \right) + \nabla \cdot \left(\mathbf{v}_H \frac{\partial p}{\partial \eta} \right) + \frac{\partial}{\partial \eta} \left(\eta \frac{\partial p}{\partial \eta} \right) = 0$$

$$\frac{\partial U}{\partial t} + \frac{1}{a \cos^2 \theta} \left\{ U \frac{\partial U}{\partial \lambda} + V \cos \theta \frac{\partial U}{\partial \theta} \right\} + \eta \frac{\partial U}{\partial \eta} - fV + \frac{1}{a} \left\{ \frac{\partial \phi}{\partial \lambda} + R_{dry} T_v \frac{\partial}{\partial \lambda} (\ln p) \right\} = \boxed{P_U} + K_U$$

$$\frac{\partial V}{\partial t} + \frac{1}{a \cos^2 \theta} \left\{ U \frac{\partial V}{\partial \lambda} + V \cos \theta \frac{\partial V}{\partial \theta} + \sin \theta (U^2 + V^2) \right\} + \eta \frac{\partial V}{\partial \eta} + fU + \frac{\cos \theta}{a} \left\{ \frac{\partial \phi}{\partial \theta} + R_{dry} T_v \frac{\partial}{\partial \theta} (\ln p) \right\} = \boxed{P_V} + K_V$$

Parameterized sub-grid processes



We live on the surface of Earth...



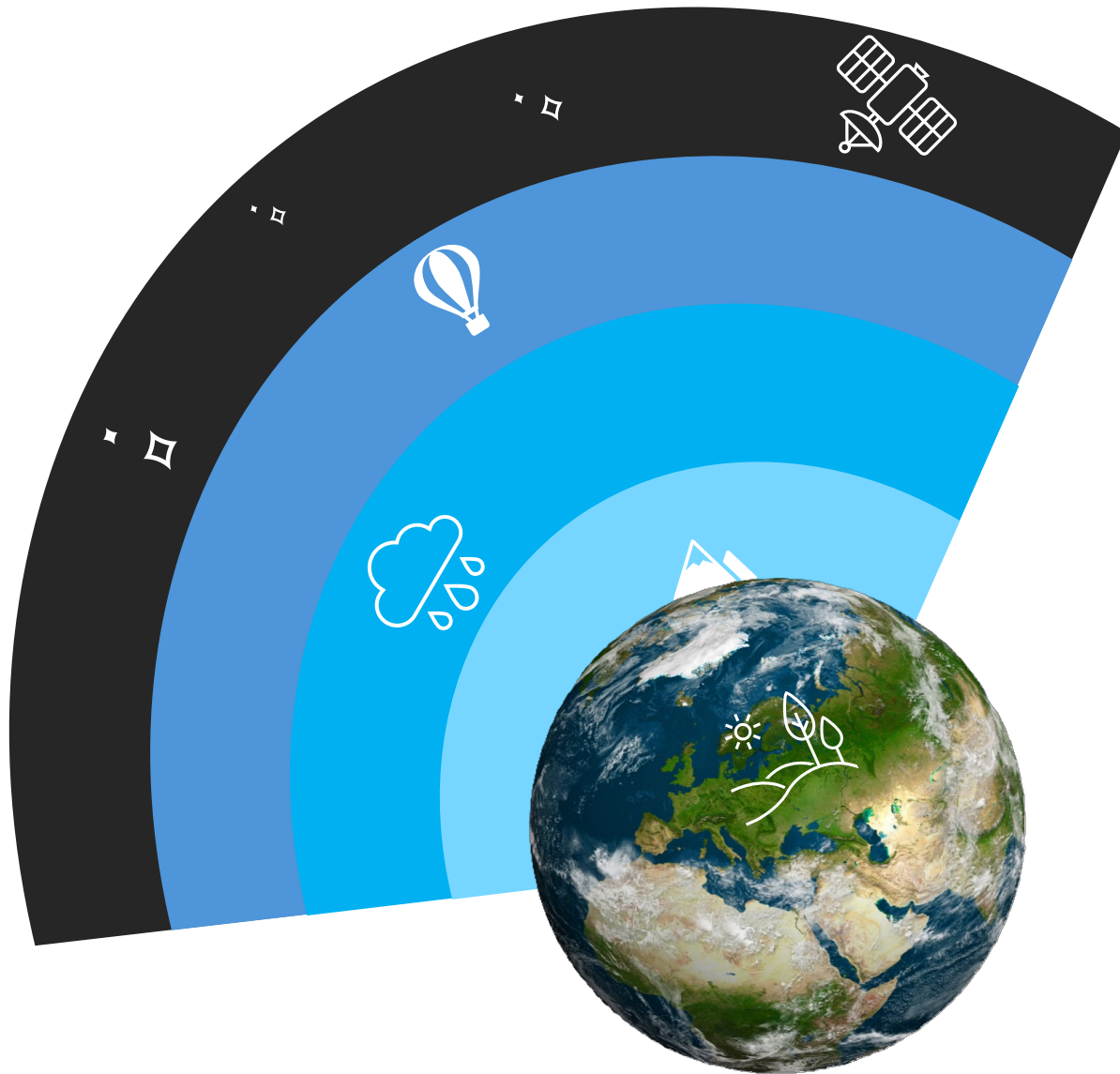
...2m temperatures
are, therefore, the
most important
forecast variable for
what we 'feel'

We live on the surface of Earth...

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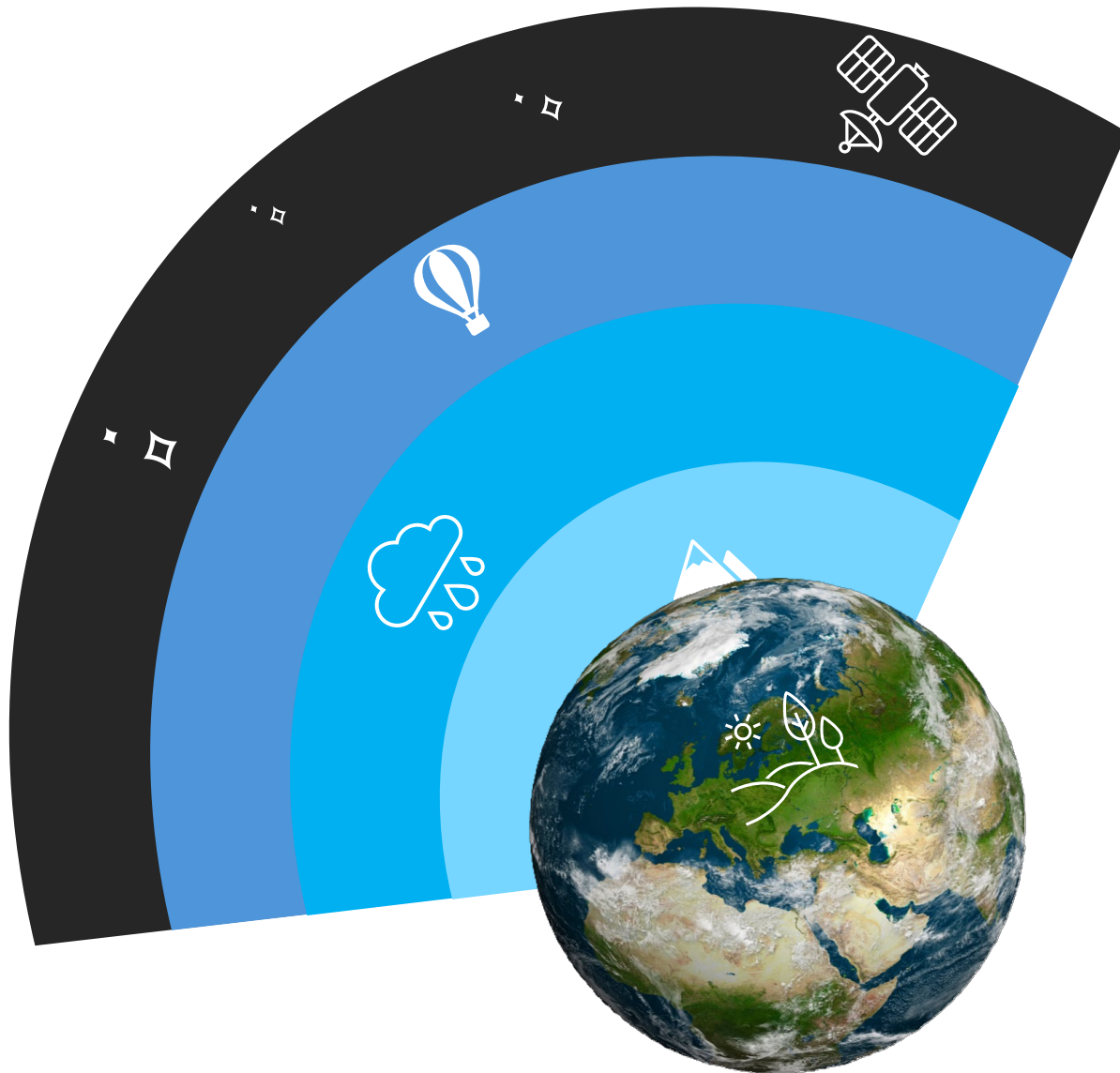
We live on the surface of Earth...



...2m temperatures are, therefore, the most important forecast variable for what we 'feel'

Why then do we forecast from 'soil-to-space'?

We live on the surface of Earth...

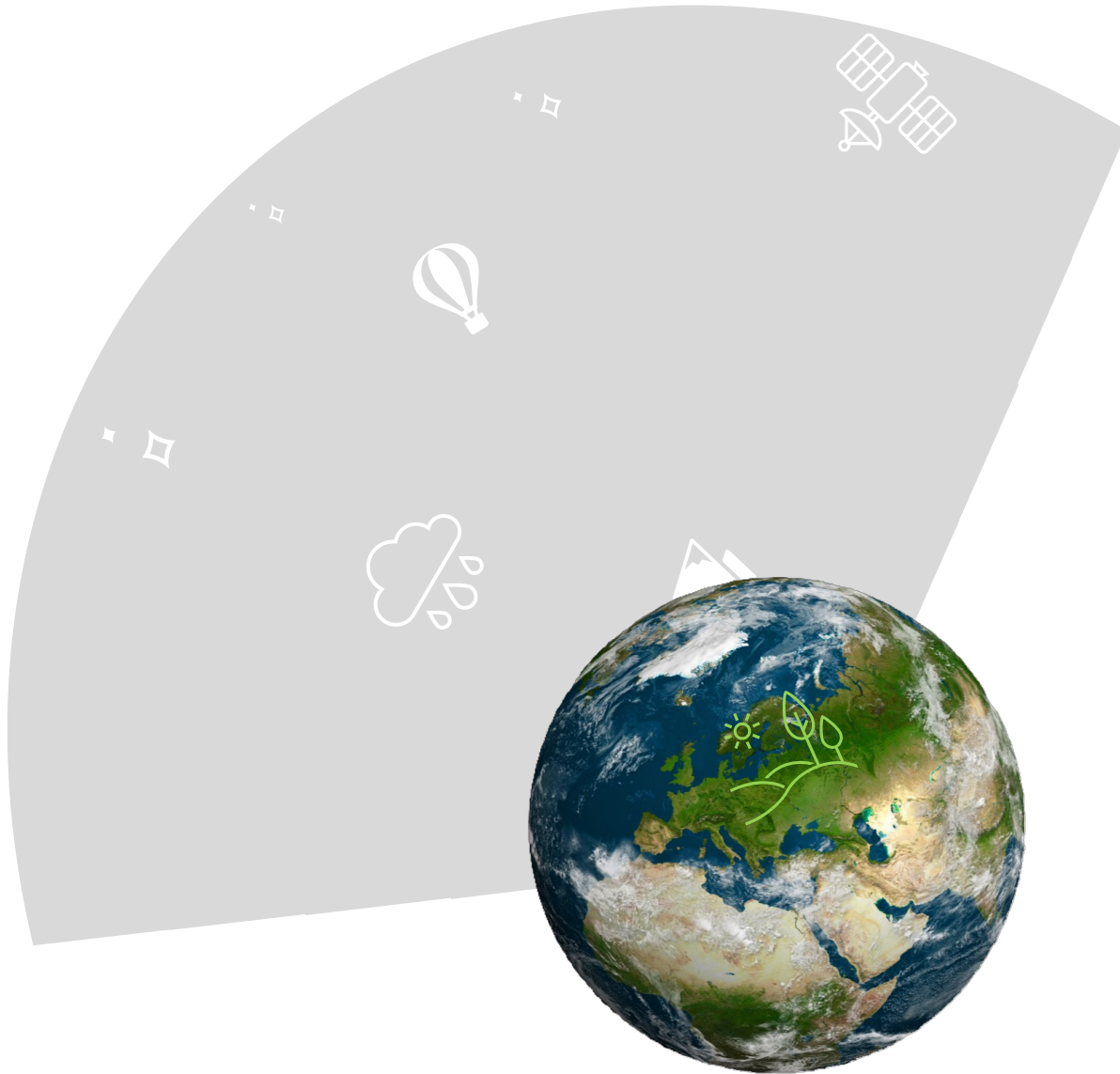


...2m temperatures are, therefore, the most important forecast variable for what we 'feel'

Why then do we forecast from 'soil-to-space'?

There many **processes** throughout the atmosphere that impact surface temperatures

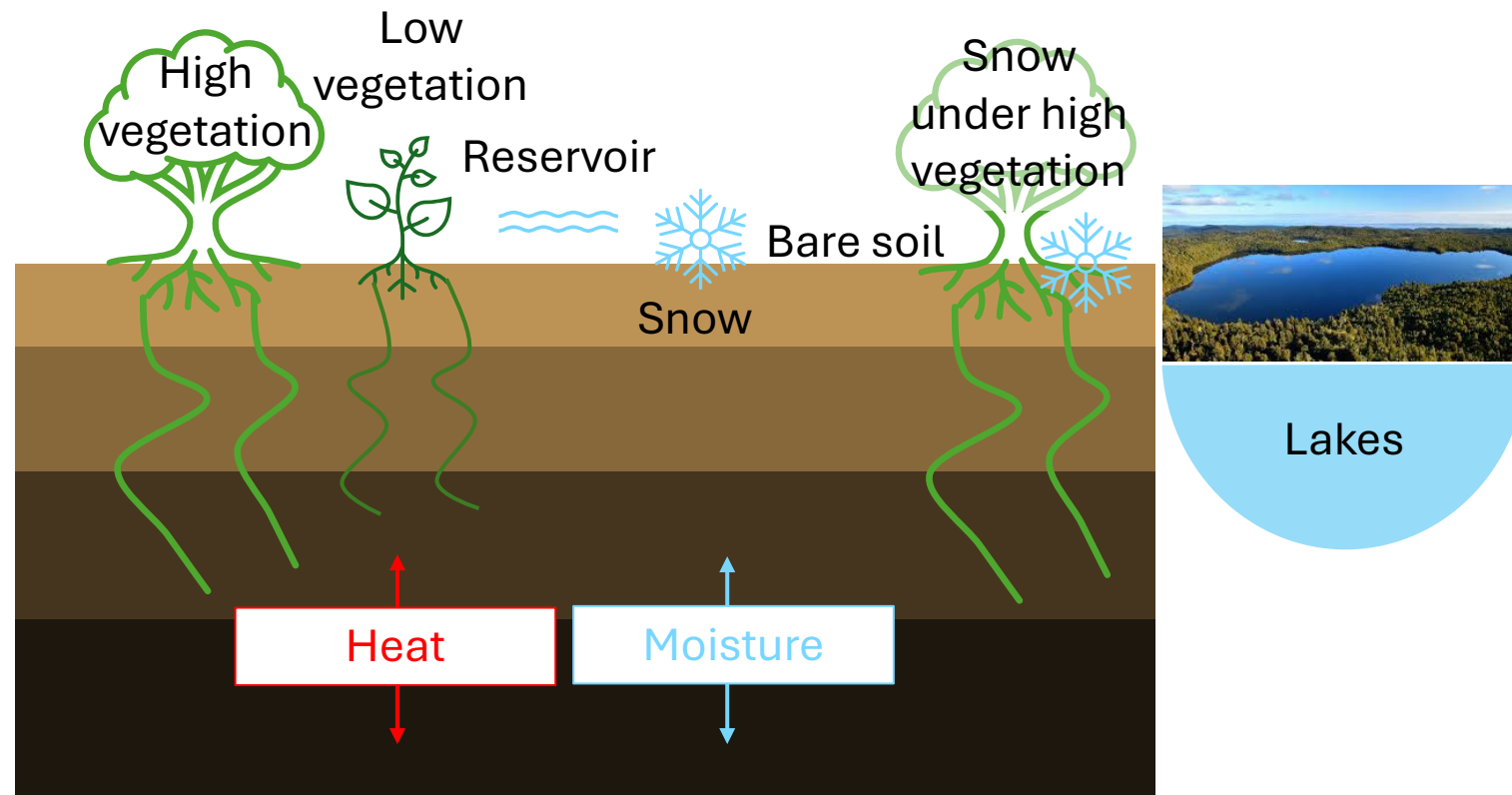
Processes that impact forecast temperature



Land surface and ocean

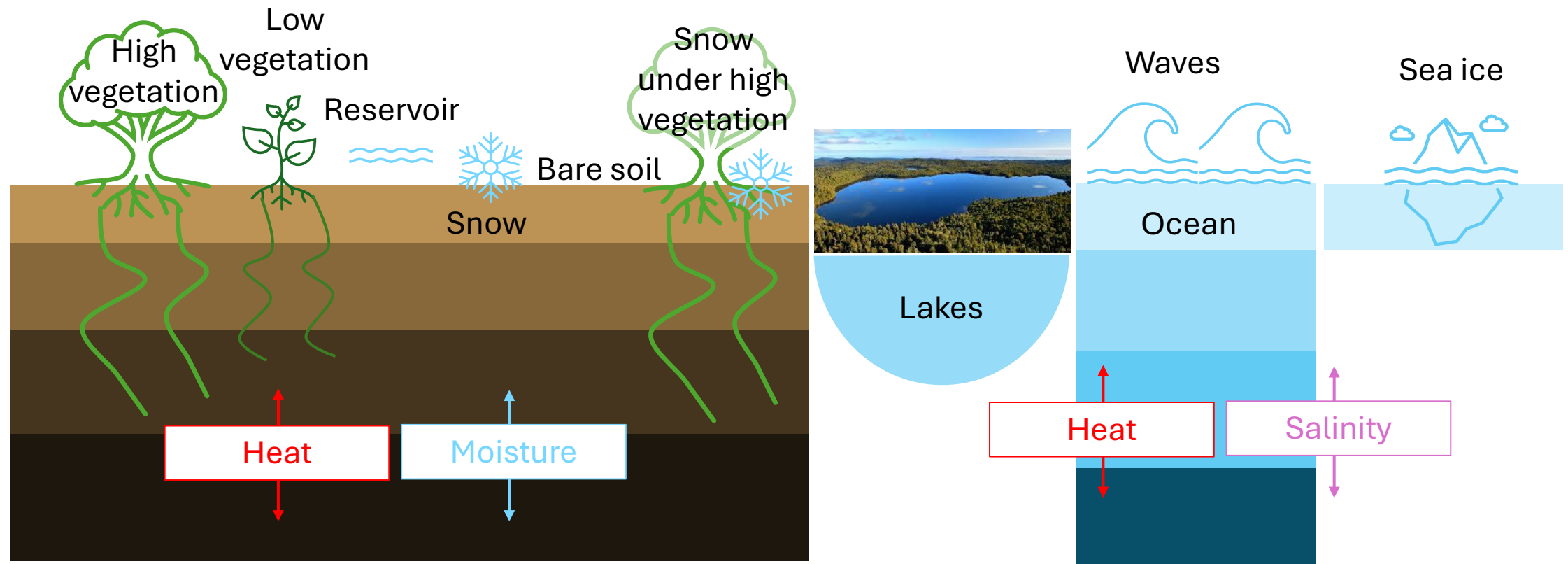
Land surface, waves and ocean

First model level ~10m above surface



Land surface, waves and ocean

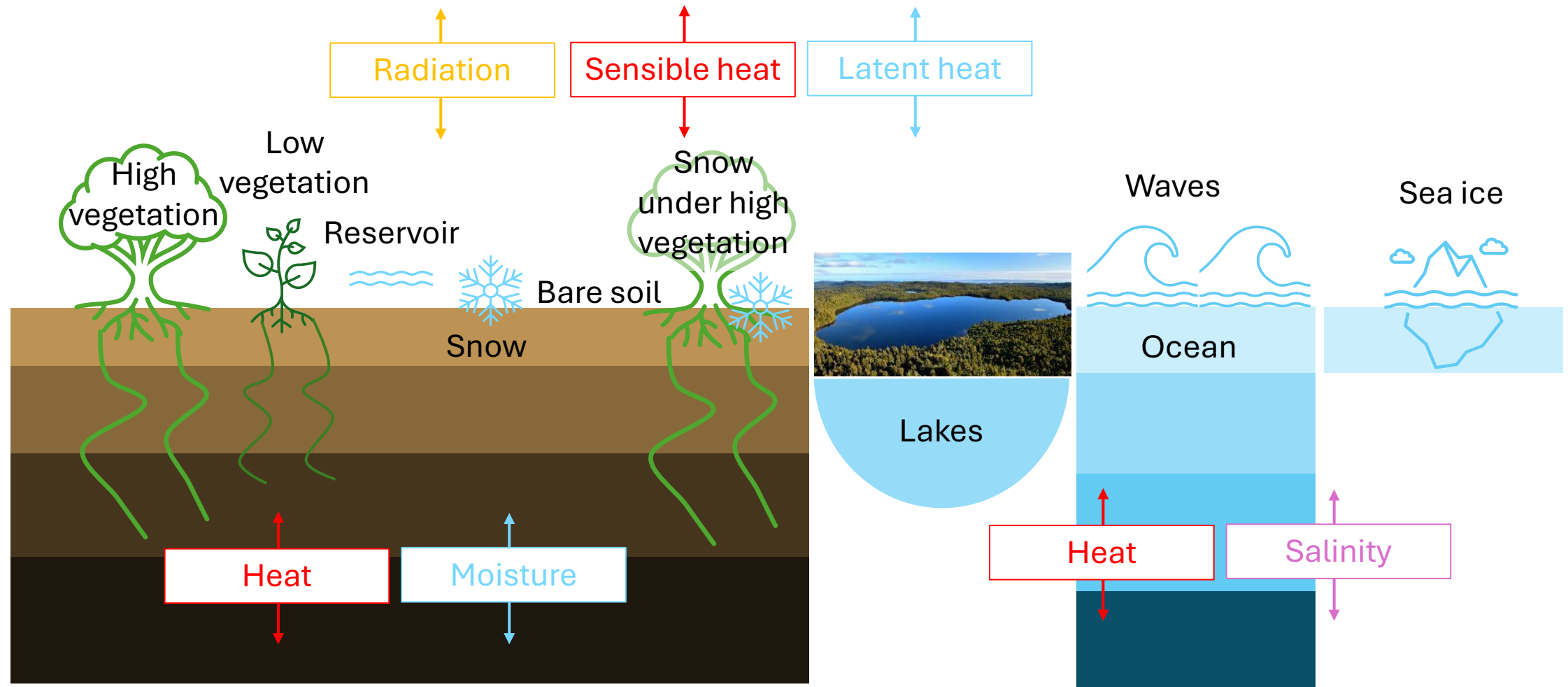
First model level ~10m above surface



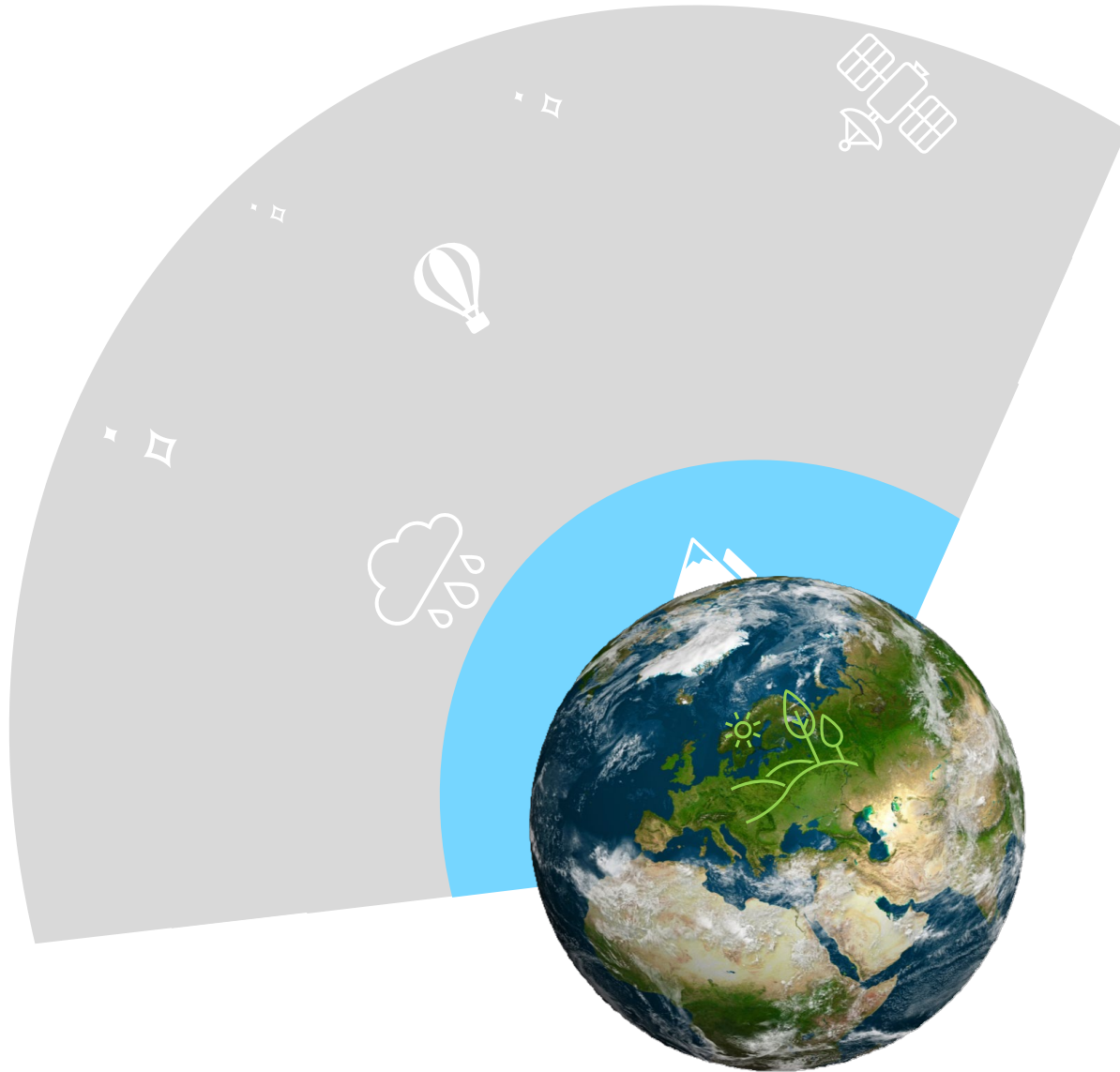
Land surface, waves and ocean

$$R_{LW} + R_{SW} + SH + LH = \Lambda(T_{10m} - T_{surf})$$

First model level ~10m above surface



Processes that impact forecast temperature

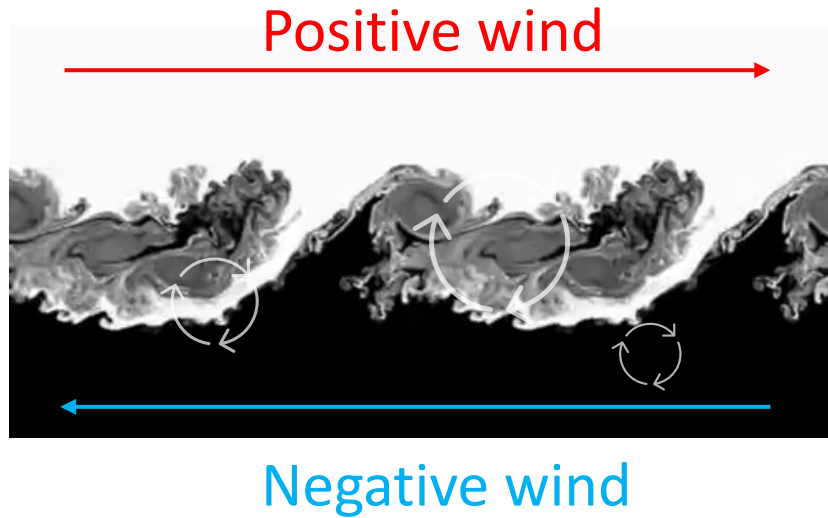


Boundary layer turbulence

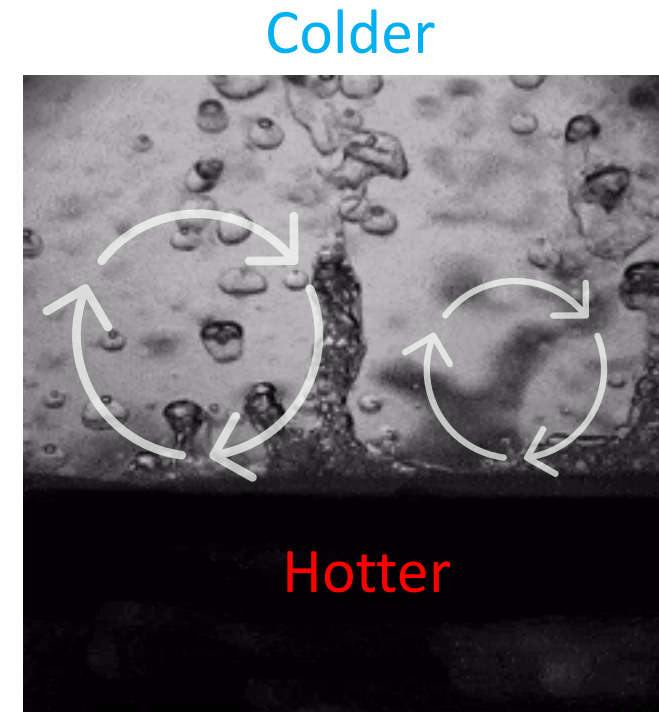
Land surface and ocean

Types of turbulence

Shear turbulence:

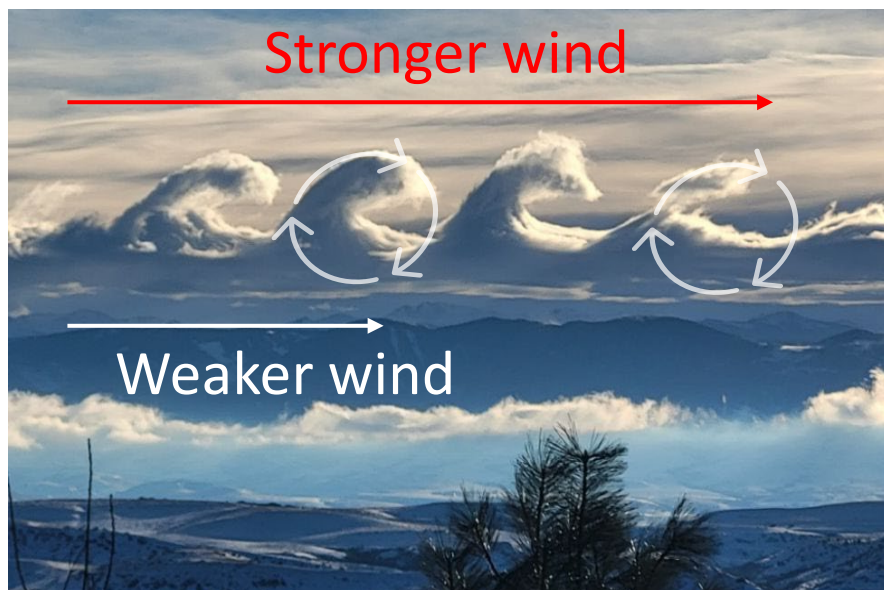


Convective turbulence:

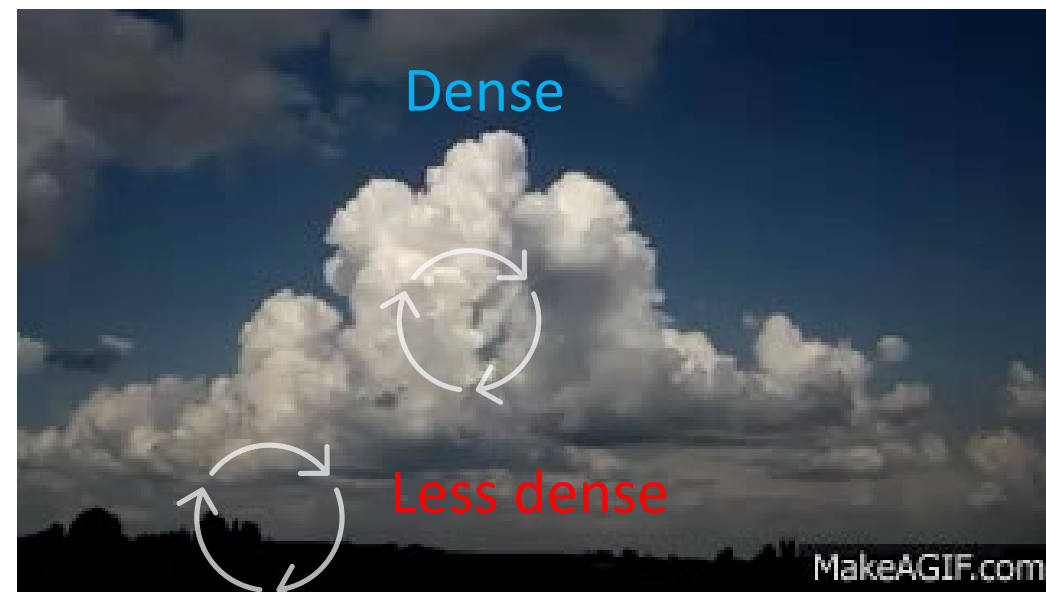


Types of turbulence

Shear turbulence:



Convective turbulence:



Turbulence acts to homogenise the fluid by mixing:
momentum, heat, moisture, tracers and clouds

Processes that impact forecast temperature

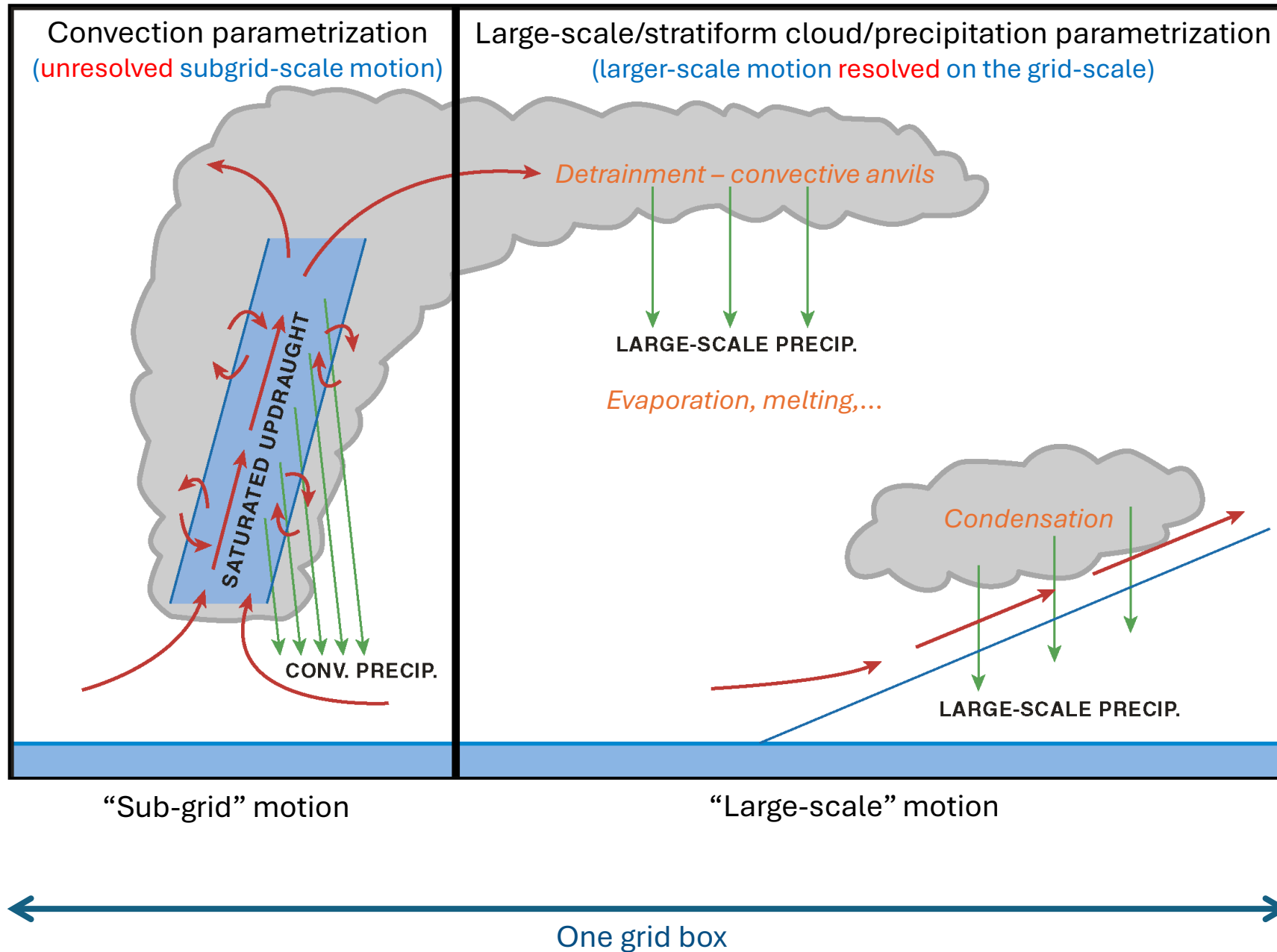


Clouds and convection

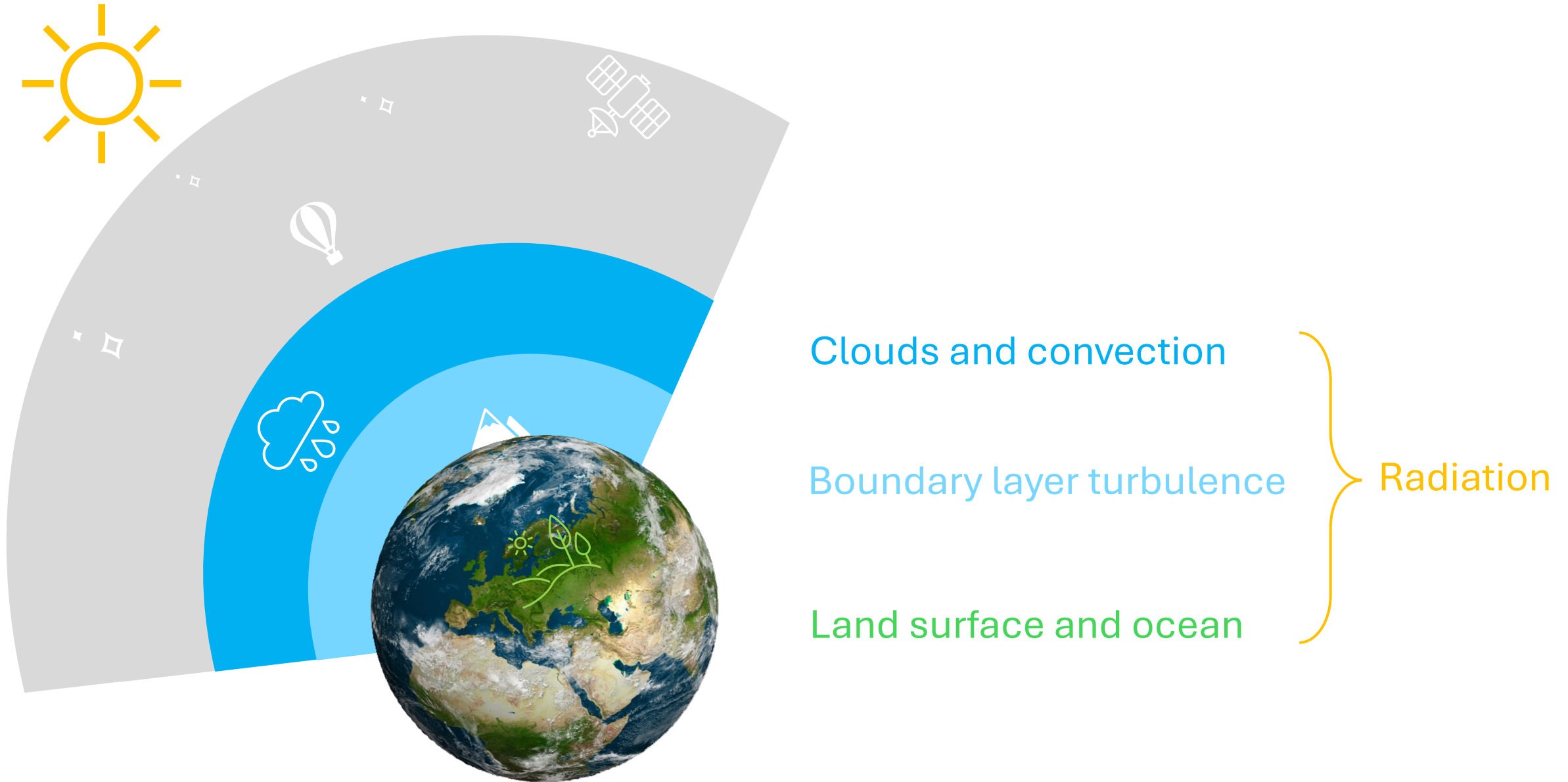
Boundary layer turbulence

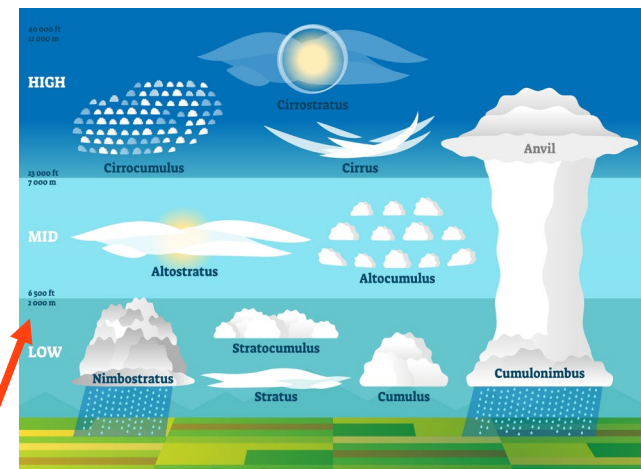
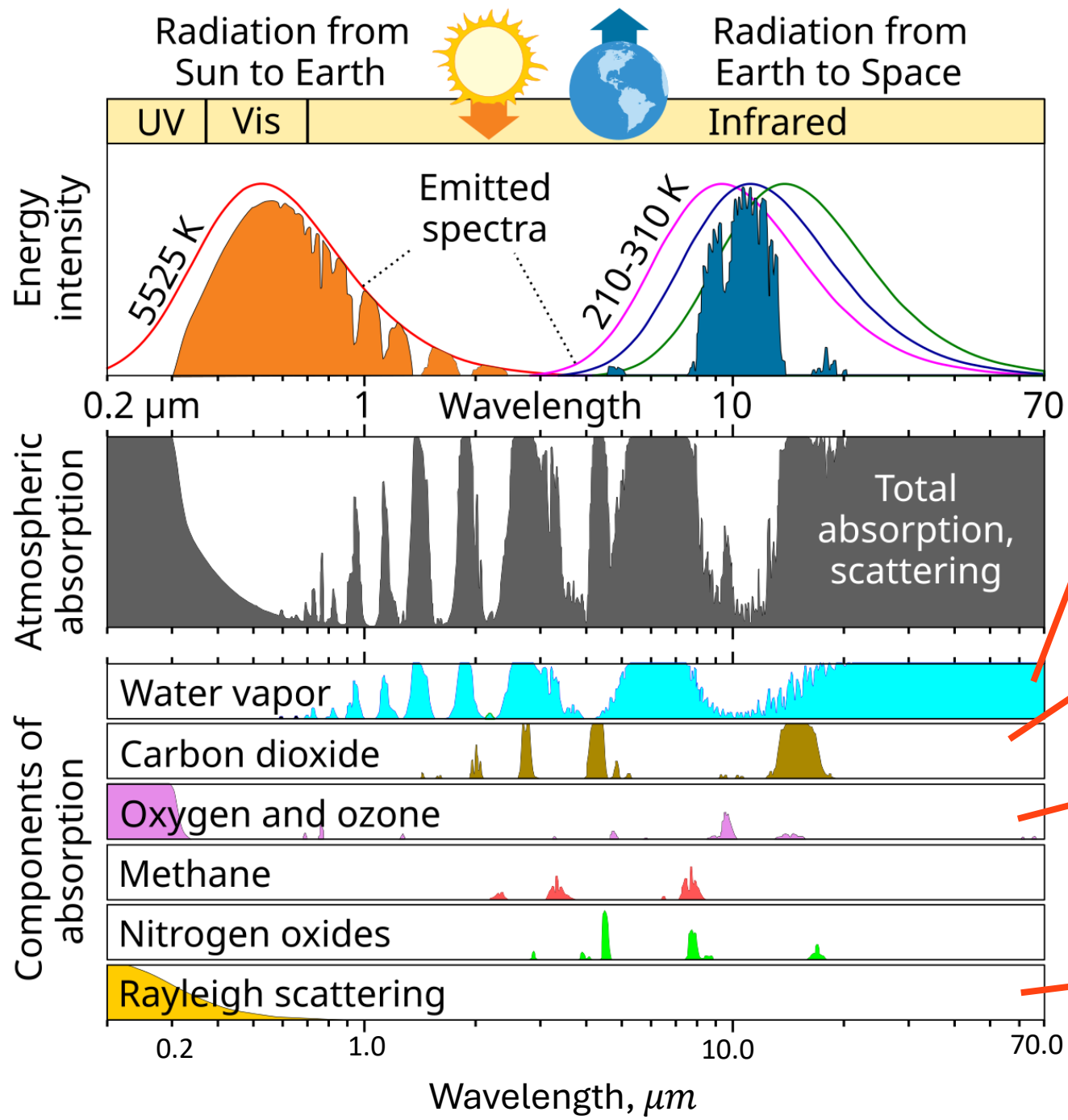
Land surface and ocean

Large-scale and sub-grid convection

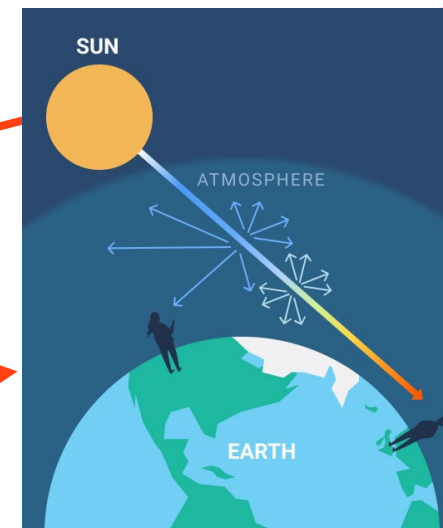
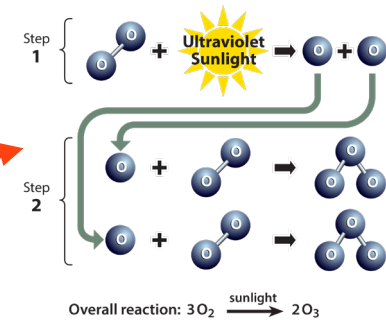


Processes that impact forecast temperature

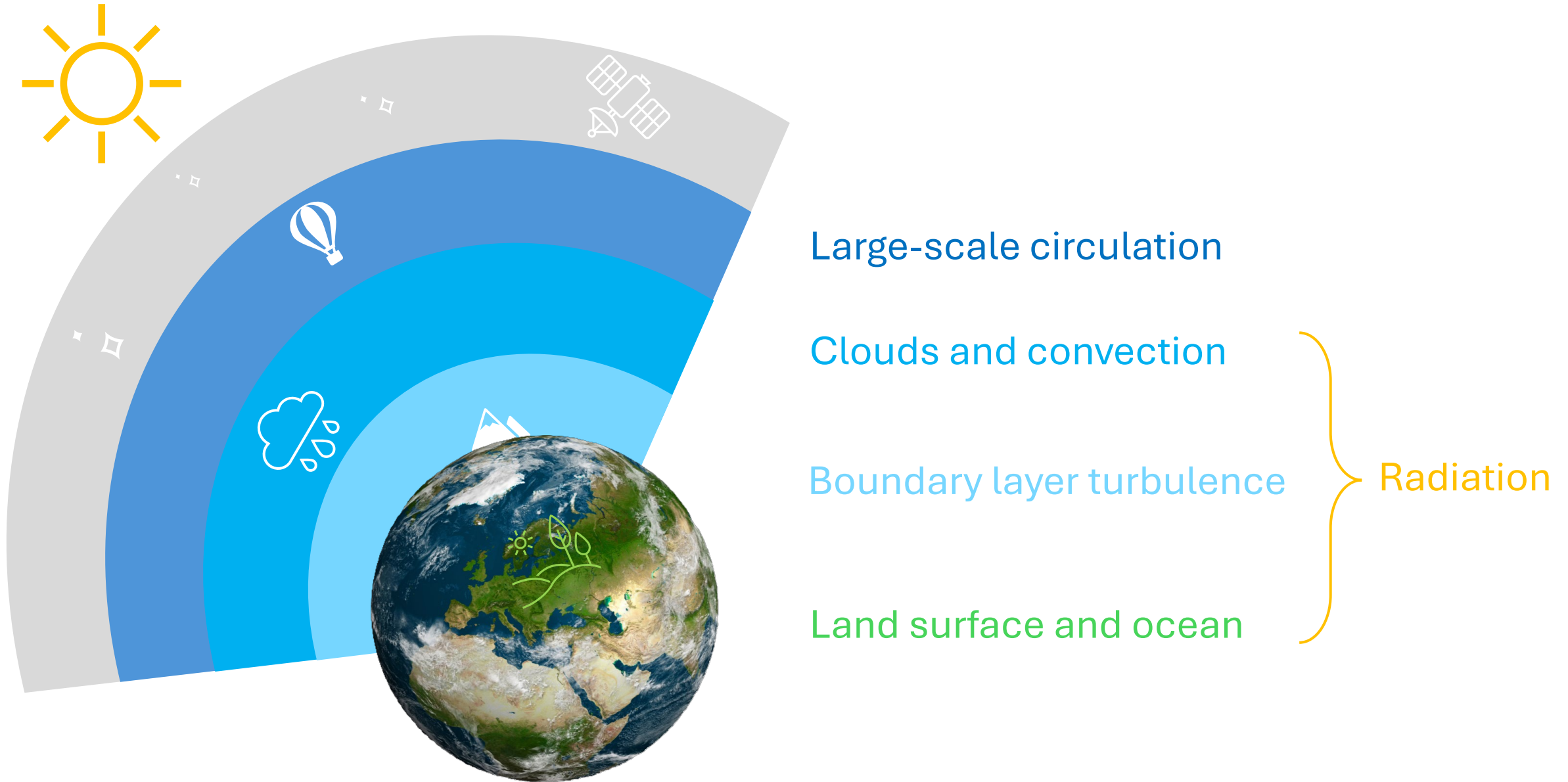




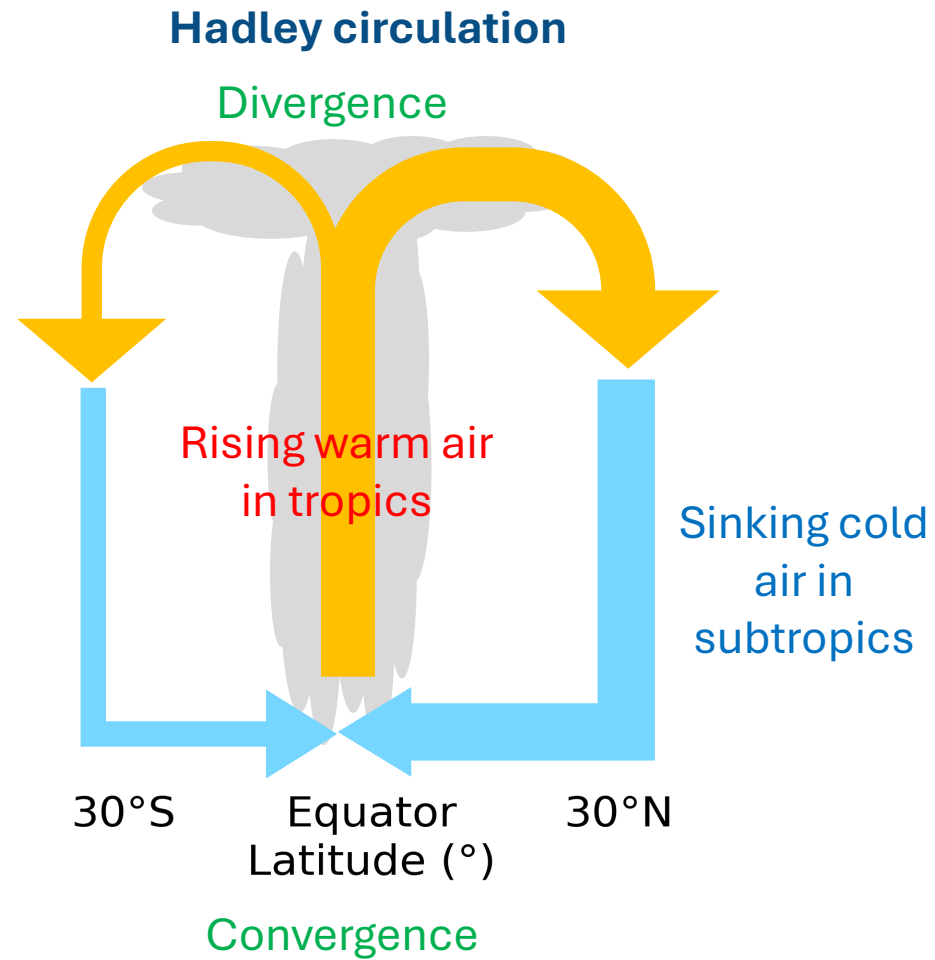
Stratospheric Ozone Production



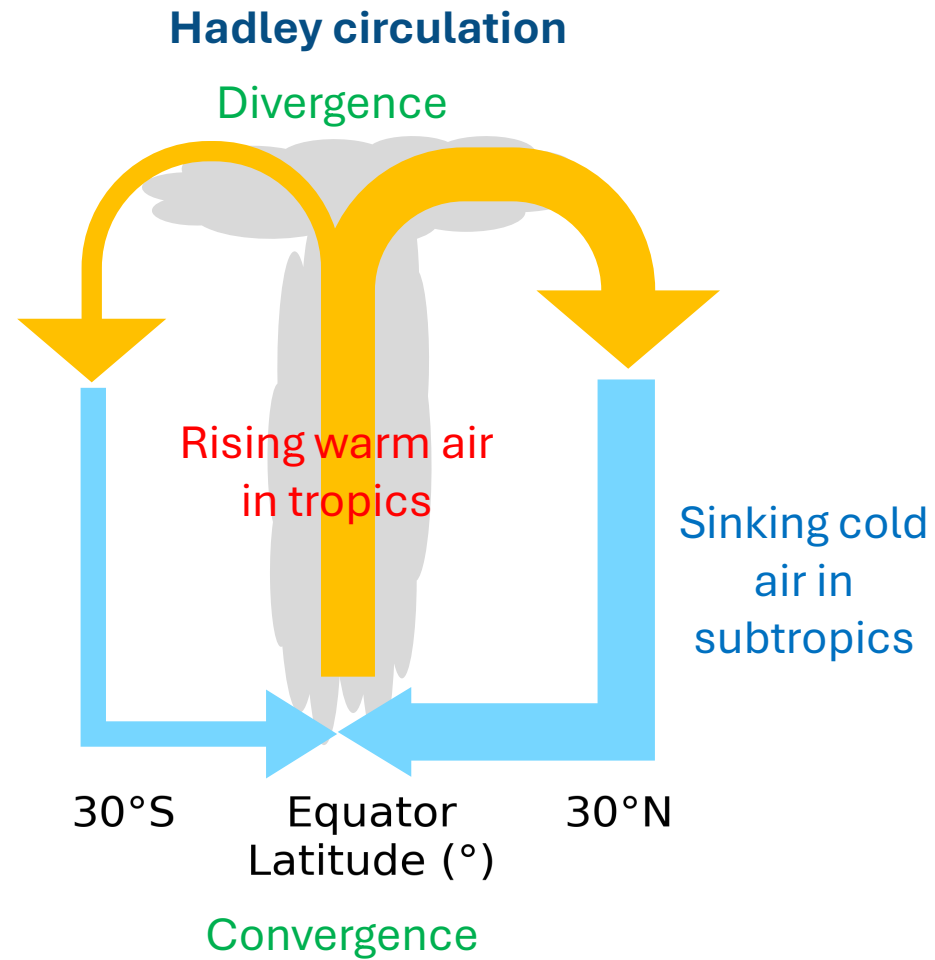
Processes that impact forecast temperature



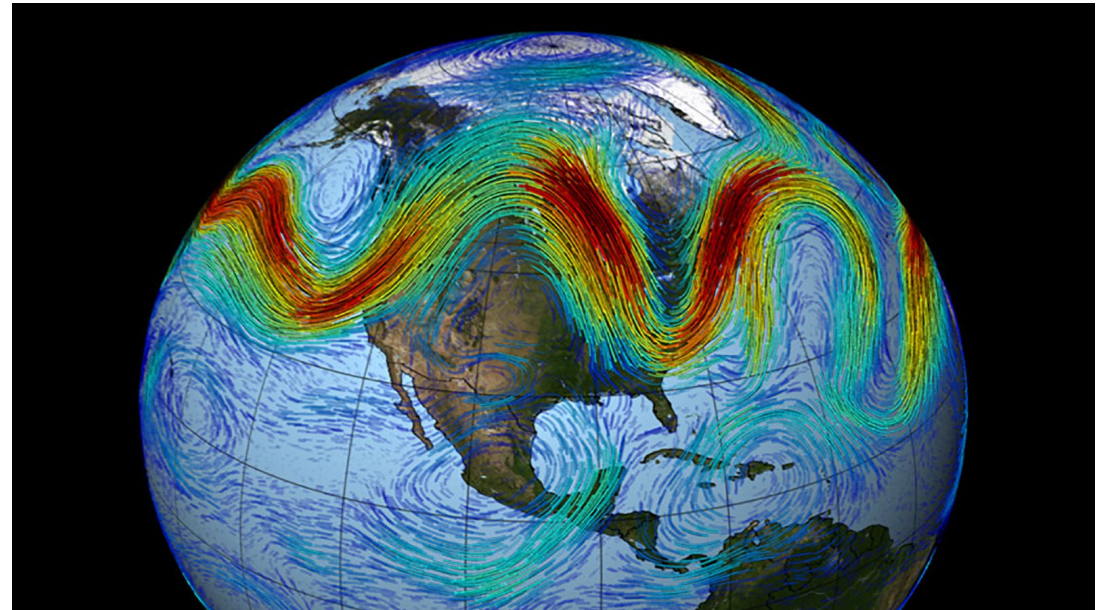
Large-scale circulation



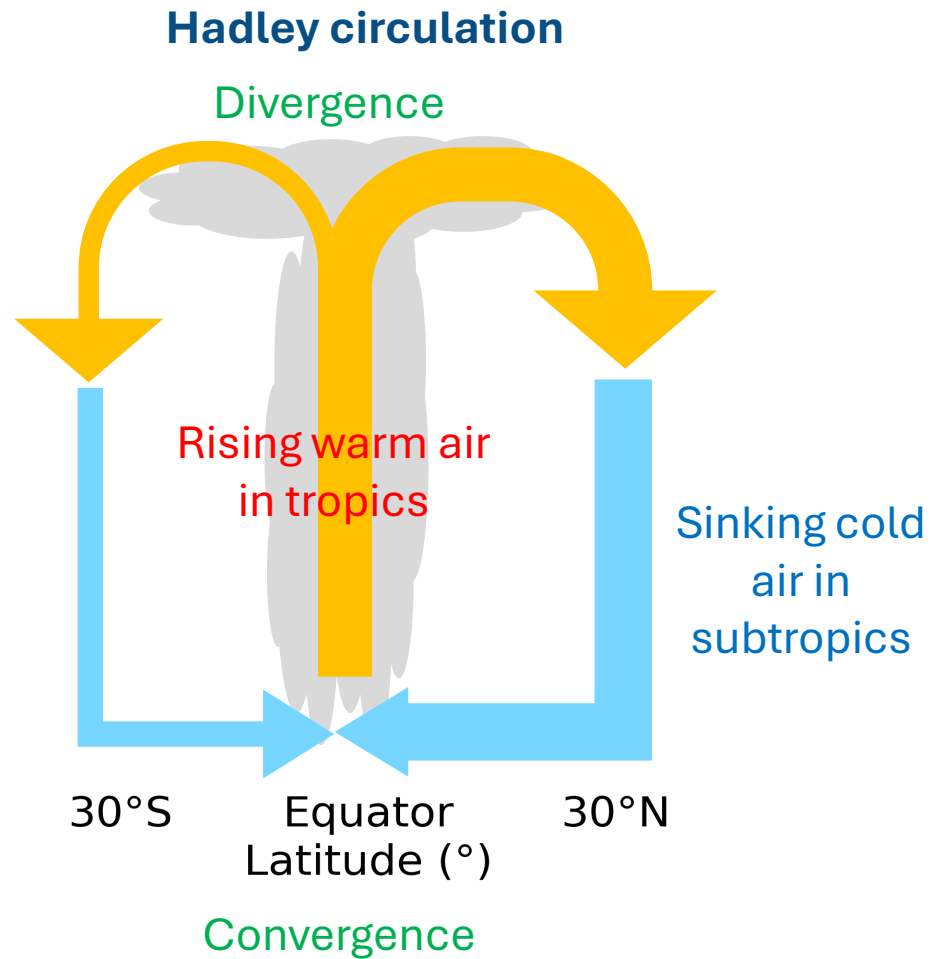
Large-scale circulation



Planetary waves



Large-scale circulation



Large-scale resolved dynamics

$$\frac{\partial T}{\partial t} + \frac{1}{a \cos^2 \theta} \left\{ U \frac{\partial T}{\partial \lambda} + V \cos \theta \frac{\partial T}{\partial \theta} \right\} + \dot{\eta} \frac{\partial T}{\partial \eta} - \frac{\kappa T_v \omega}{(1 + (\delta - 1)q)p} = \boxed{P_T} + K_T$$

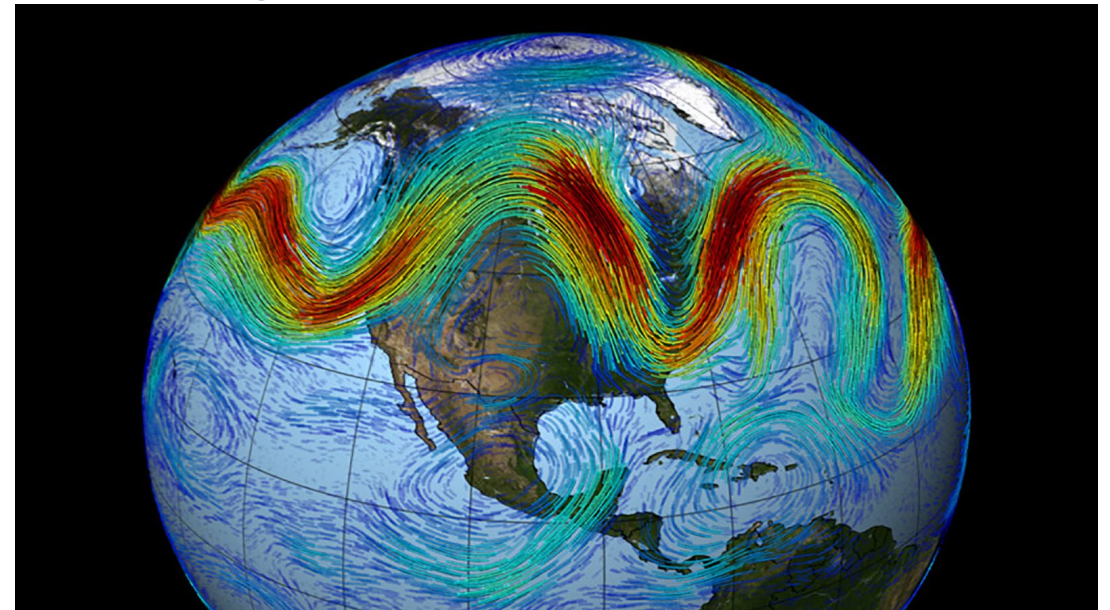
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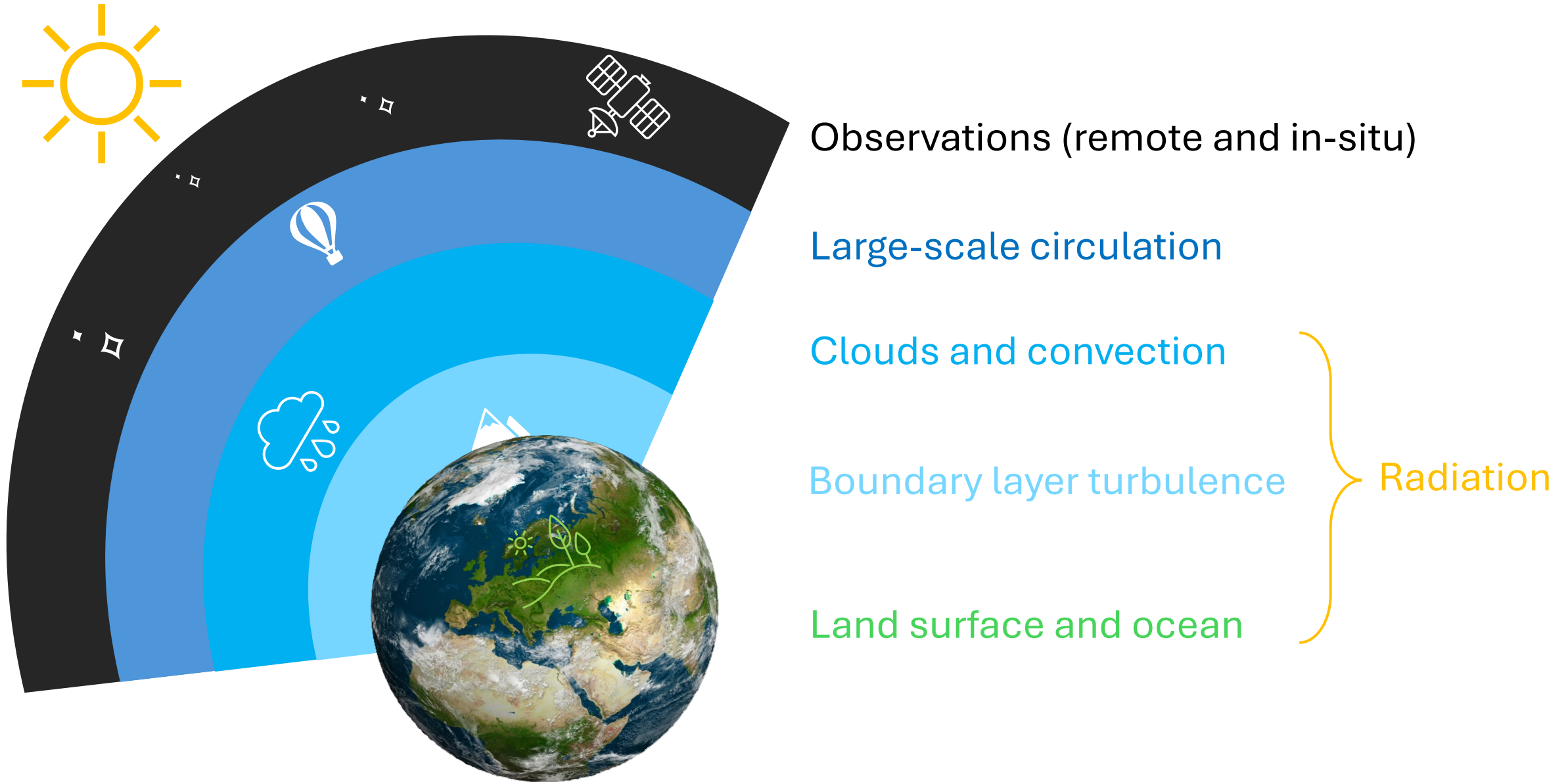
$$\frac{\partial U}{\partial t} + \frac{1}{a \cos^2 \theta} \left\{ U \frac{\partial U}{\partial \lambda} + V \cos \theta \frac{\partial U}{\partial \theta} \right\} + \dot{\eta} \frac{\partial U}{\partial \eta} - fV + \frac{1}{a} \left\{ \frac{\partial \phi}{\partial \lambda} + R_{\text{dry}} T_v \frac{\partial}{\partial \lambda} (\ln p) \right\} = \boxed{P_U} + K_U$$

$$\frac{\partial V}{\partial t} + \frac{1}{a \cos^2 \theta} \left\{ U \frac{\partial V}{\partial \lambda} + V \cos \theta \frac{\partial V}{\partial \theta} + \sin \theta (U^2 + V^2) \right\} + \dot{\eta} \frac{\partial V}{\partial \eta} + fU + \frac{\cos \theta}{a} \left\{ \frac{\partial \phi}{\partial \theta} + R_{\text{dry}} T_v \frac{\partial}{\partial \theta} (\ln p) \right\} = \boxed{P_V} + K_V$$

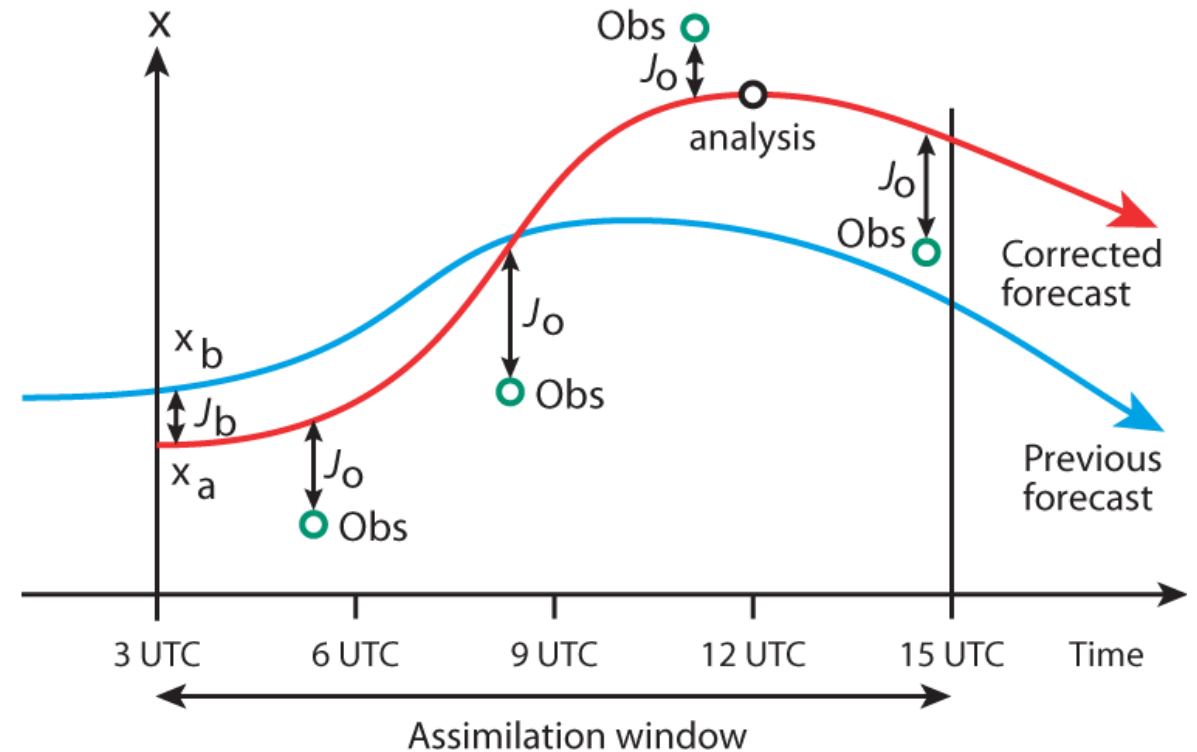
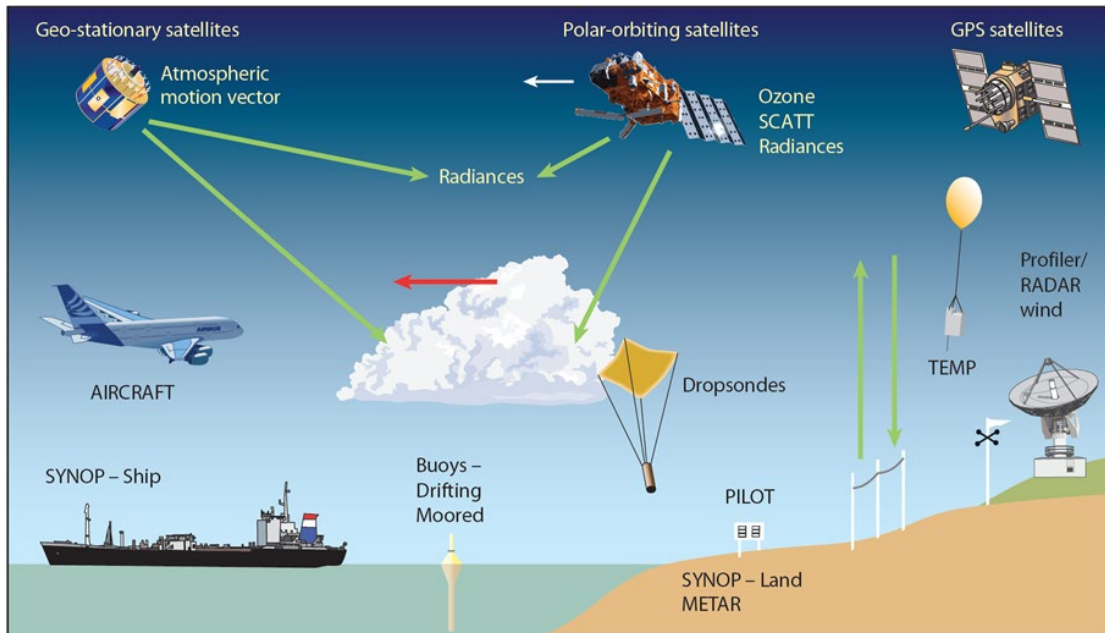
Planetary waves



Processes that impact forecast temperature

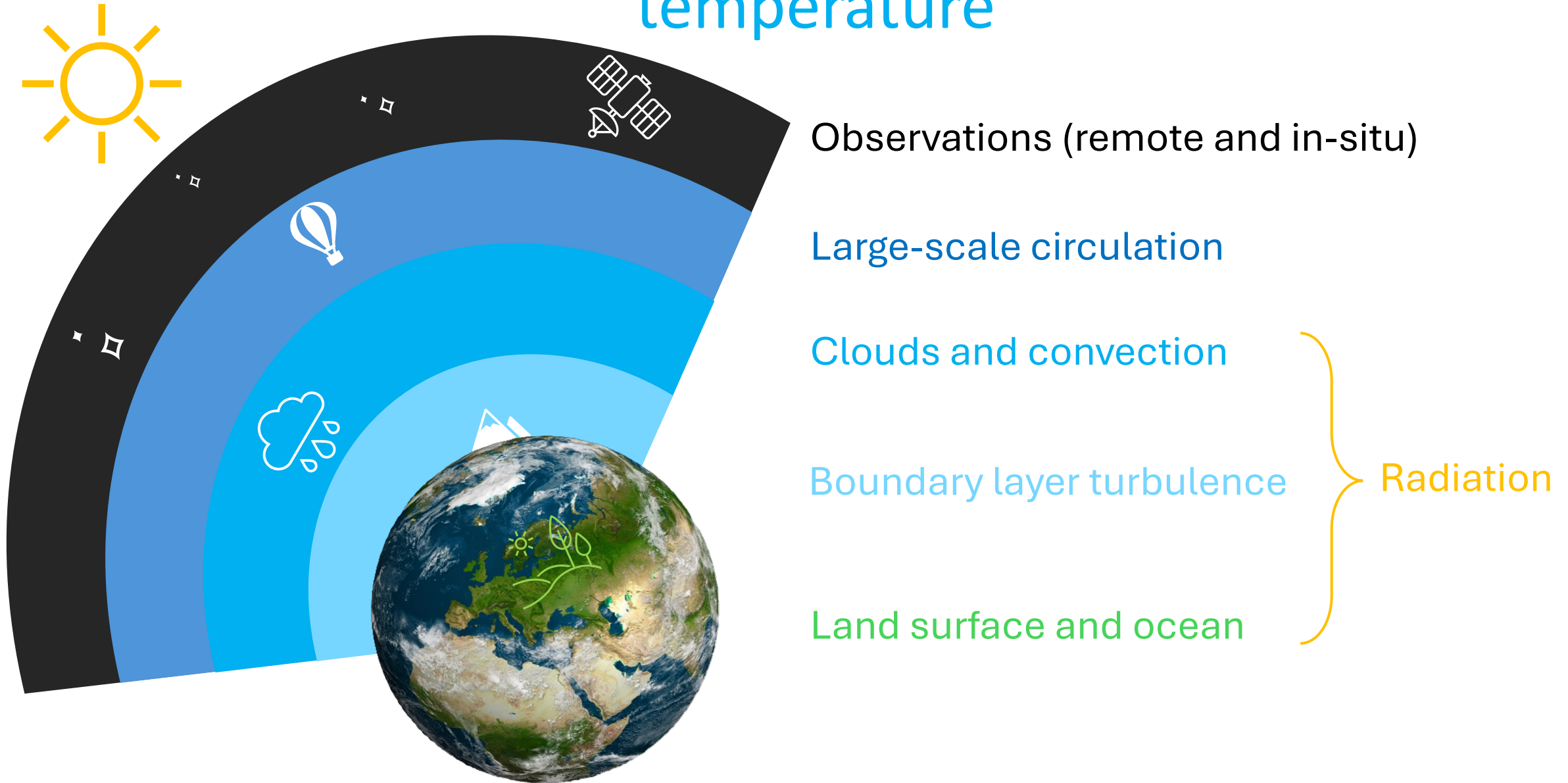


4D-Variational data assimilation

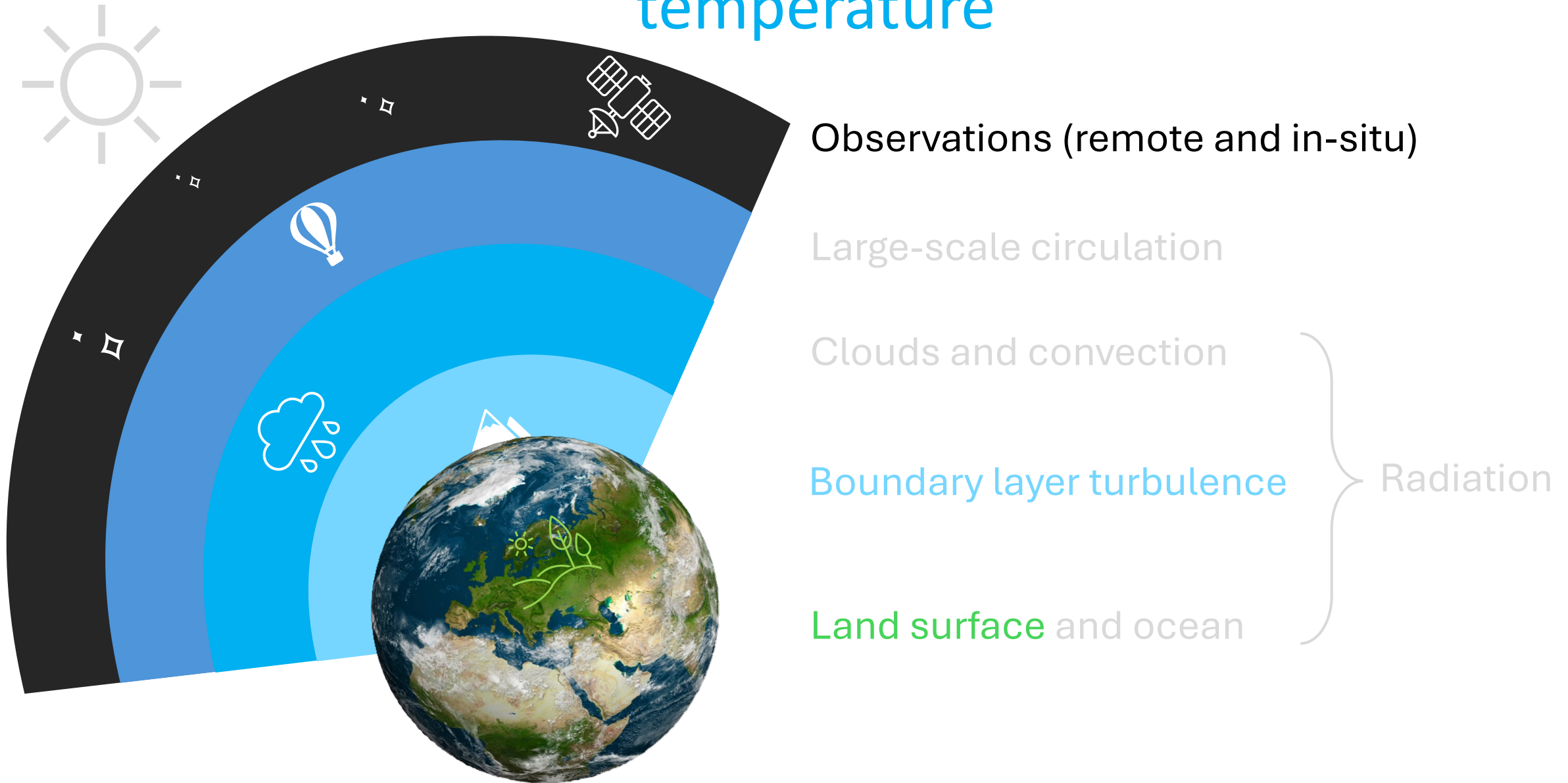


Many (millions) of observations are assimilated to get an accurate initial condition for the model, including many temperature observations

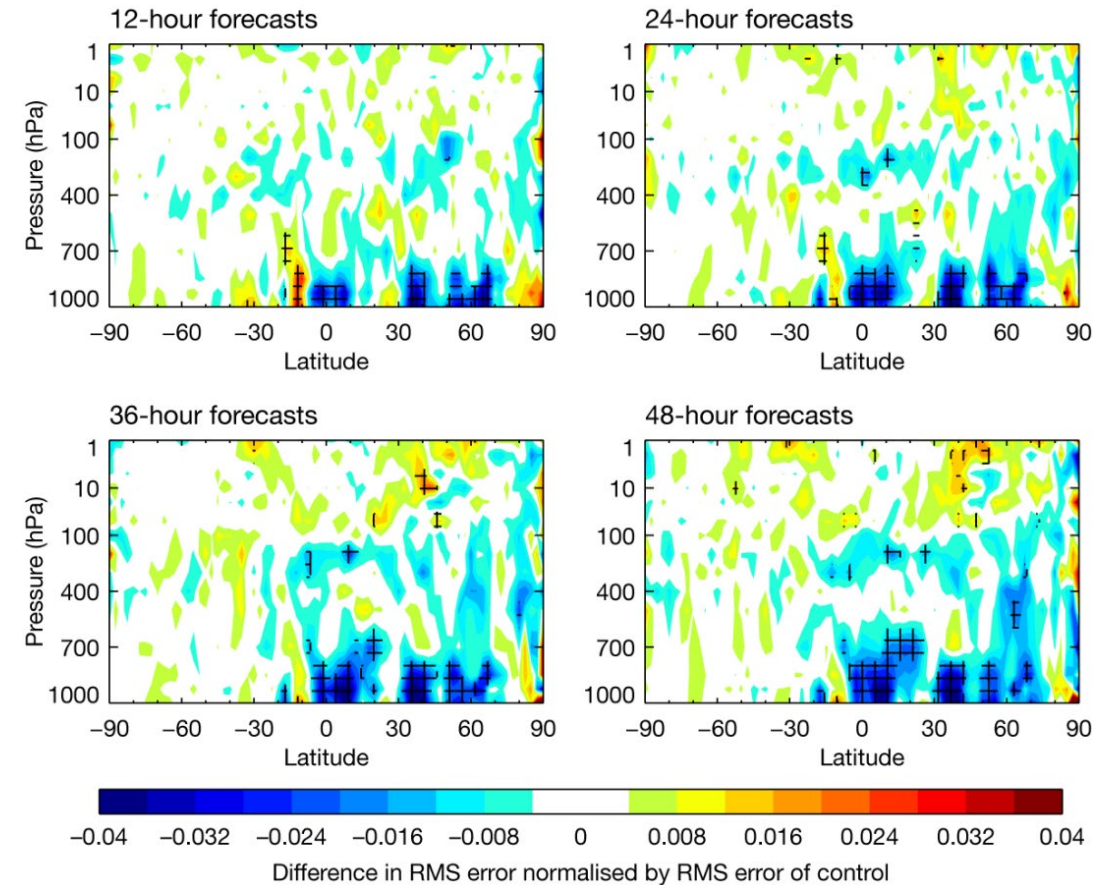
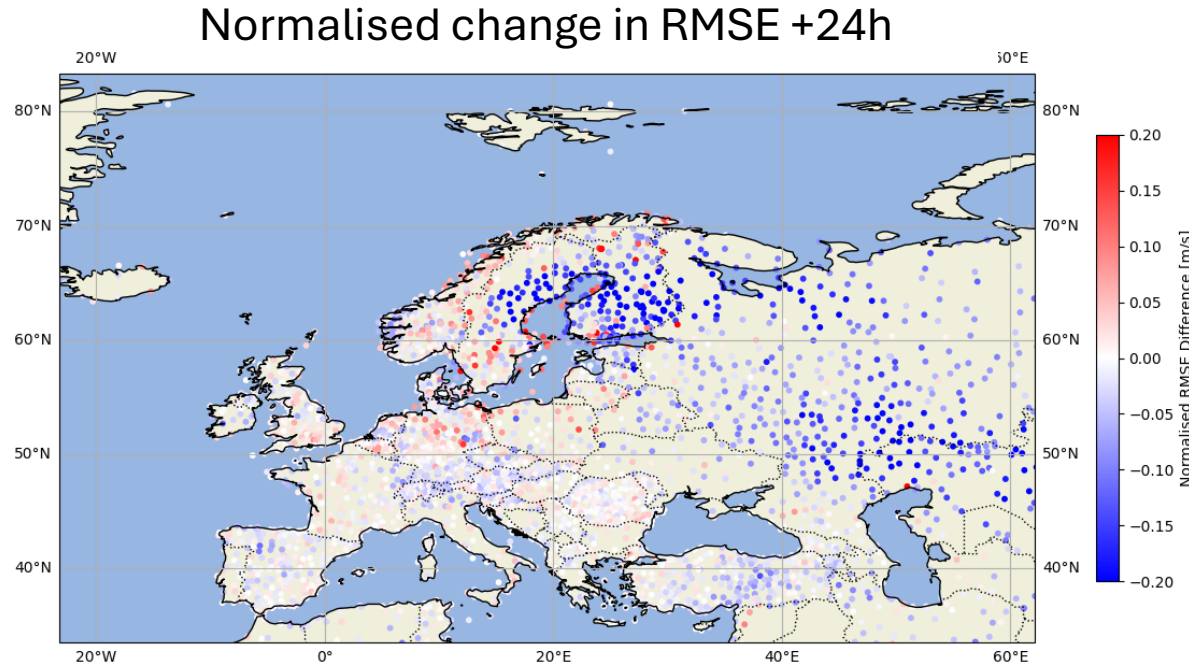
Interaction of processes that impact forecast temperature



Interaction of processes that impact forecast temperature



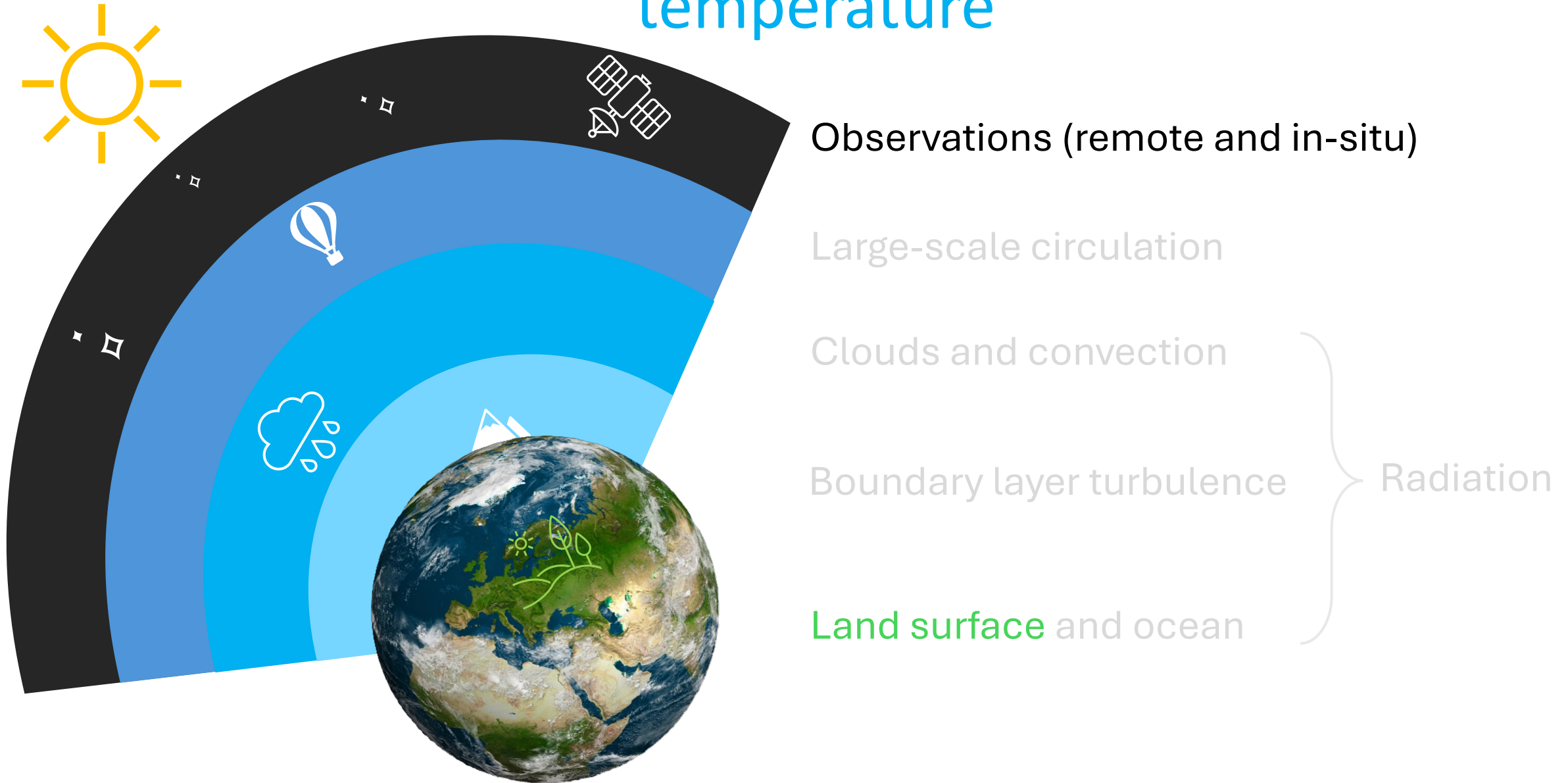
Impact of assimilating in-situ 2m temperature observations



Assimilation of 2m temperatures over land
has a lasting impact (+48 hours) on the
temperature throughout the troposphere

Ingleby et al, ECMWF newsletter,
Winter 2024

Interaction of processes that impact forecast temperature



Soil moisture analysis

For each grid point, analysed soil moisture state vector $\mathbf{x}_a$:

$$\mathbf{x}_a = \mathbf{x}_b + \mathbf{K} (\mathbf{y} - \mathcal{H}[\mathbf{x}_b])$$

$\mathbf{x}$ background soil moisture state vector,

$\mathcal{H}$ non linear observation operator

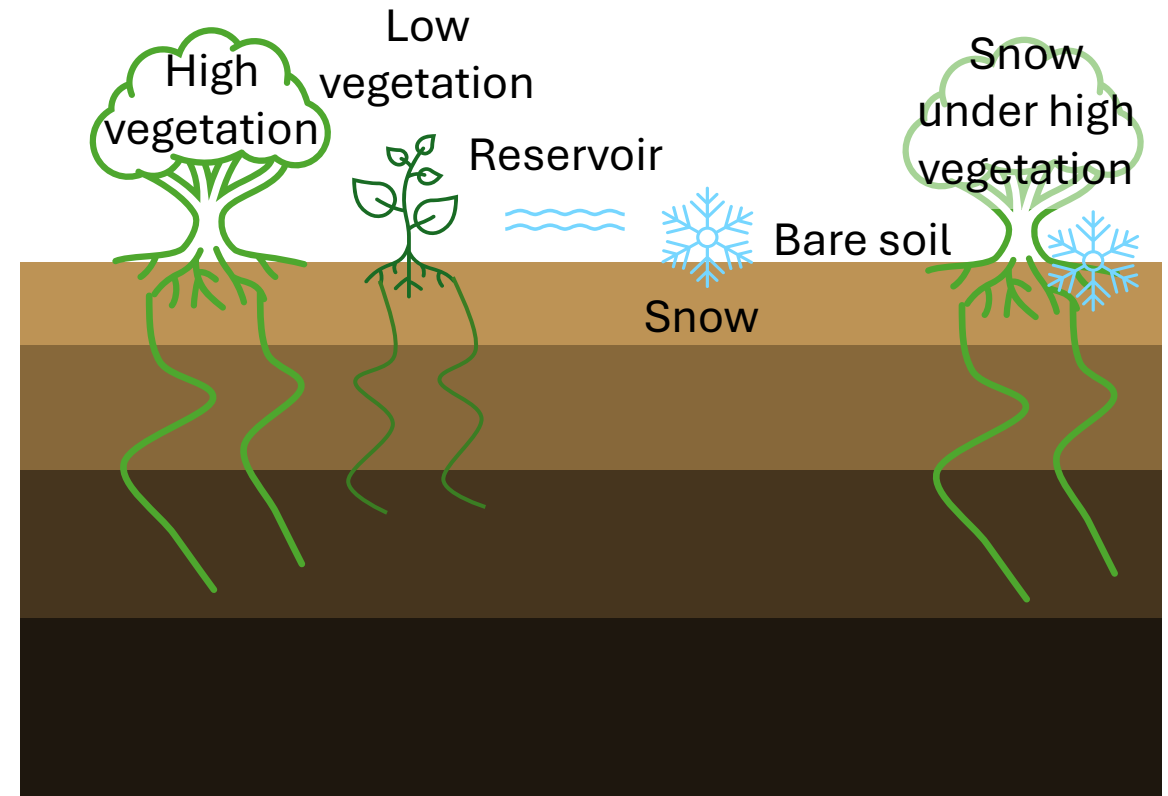
$\mathbf{y}$ observation vector

$\mathbf{K}$ Kalman gain matrix, function of $\mathbf{H}$ (linearisation of $\mathcal{H}$), $\mathbf{P}$ and $\mathbf{R}$ (covariance matrices of background and observation errors).

Used at ECMWF (NWP, ERA5, ERA6), DWD, UKMO

Drusch et al., GRL, 2009
de Rosnay et al., QJRMS, 2013
Fairbairn et al., JHM 2019
Munoz-Sabater et al QJRMS, 2019
Rodriguez-Fernandez et al, RS 2019

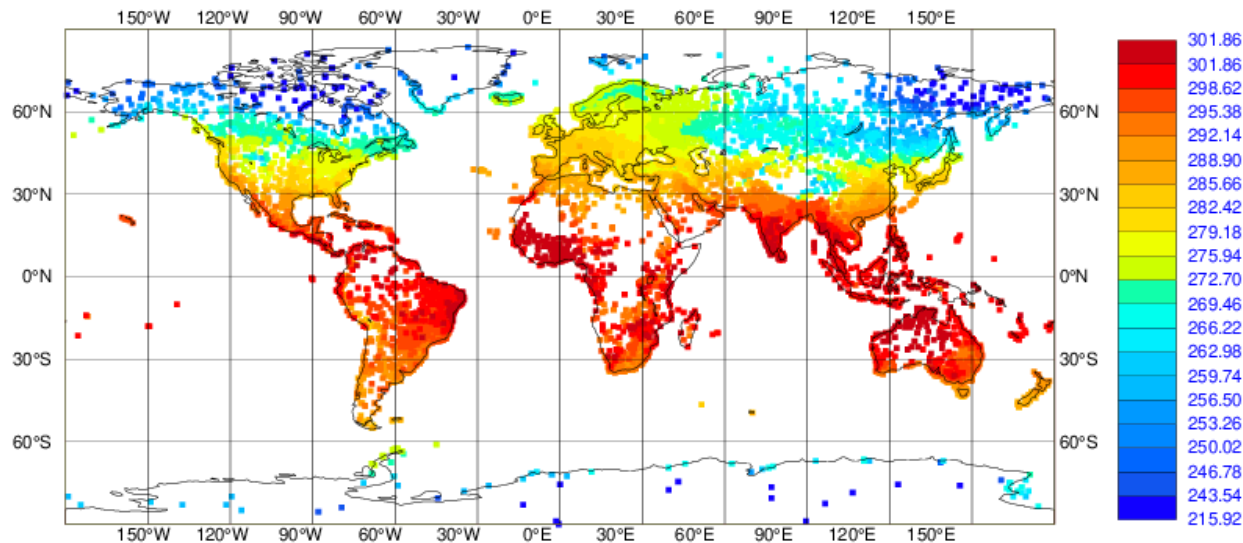
Assimilation of temperatures and humidity is used to corrects the soil moisture trajectory of Land Surface Model



SYNOP T2m, RH2m in situ data assimilated in land data assimilation

T2M FROM SYNOP
OBSERVED VALUE [K] (USED)
DATA PERIOD = 2020-02-22 09 - 2020-02-24 09
EXP = 0001, CHANNEL = 1

Min: 219.163 Max: 312.000 Mean: 282.548
GRID: 0.50x 0.50



Screen level observations are:

- T2m, two meter temperature
- RH2m, relative humidity (RH2m)

Diversity of Report types:

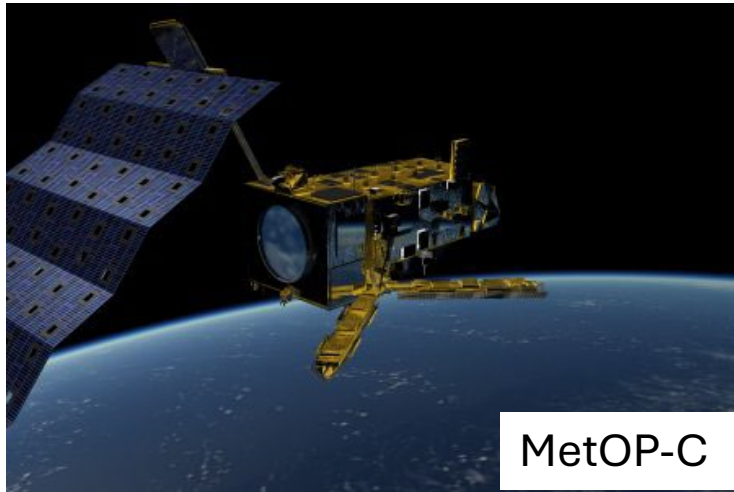
Automatic and manual SYNOP stations, METAR (METeorological Airport Reports), etc...

c/o Patricia de Rosnay

Soil moisture satellite observations assimilated operationally

Active microwave data:

ASCAT: Advanced Scatterometer
On MetOP-A (2006-2021),
MetOP-B (2012-), MetOP-C (2018-)
C-band (5.6GHz) backscattering coefficient
EUMETSAT Operational mission



Scatterometer soil moisture also used in ERA5
(ERS-SCAT, Metop/ASCAT)

Passive microwave data:

SMOS: Soil Moisture & Ocean Salinity (2009-)
L-band (1.4 GHz) Brightness Temperature
ESA Earth Explorer, dedicated soil moisture mission
(Munoz-Sabater et al., 2020, Rodriguez-Fernandez et al., 2019)



c/o Patricia de Rosnay

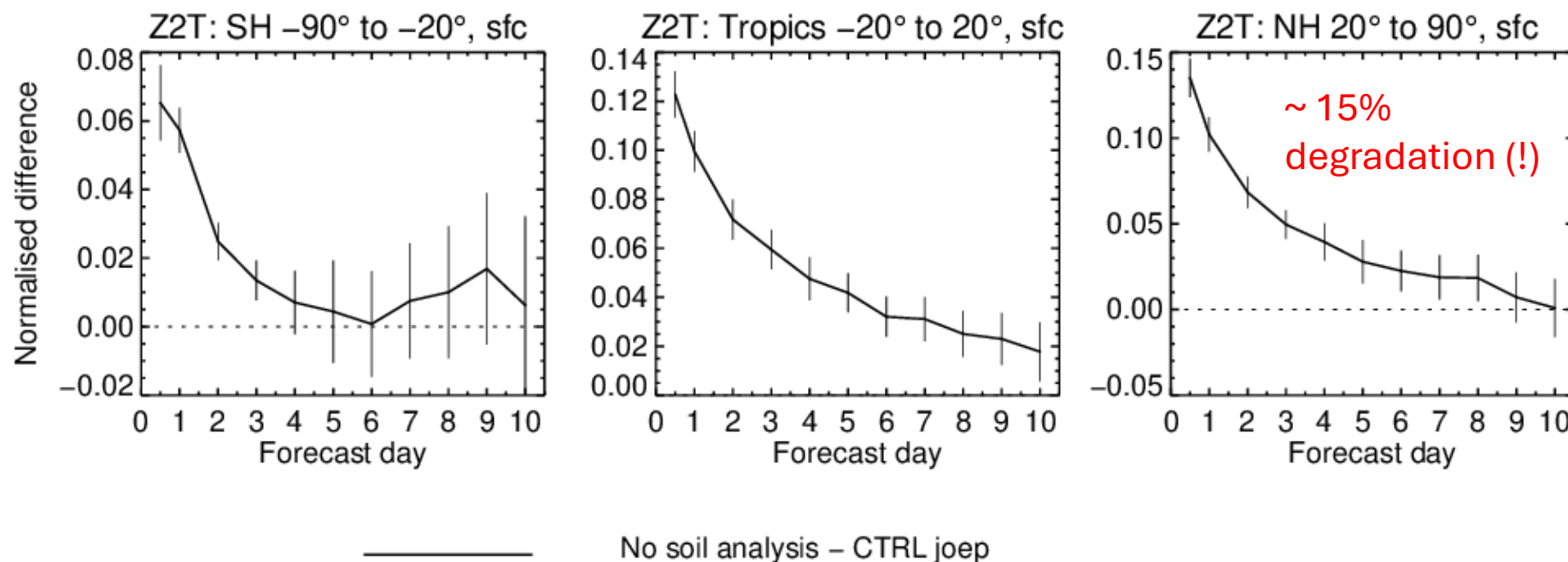
Soil analysis: impact on the atmospheric forecast

Impact of **removing** soil moisture and temperature analysis on 2m T RMSE

1-Jun-2024 to 19-Aug-2024 from 140 to 159 samples. Verified against own-analysis.

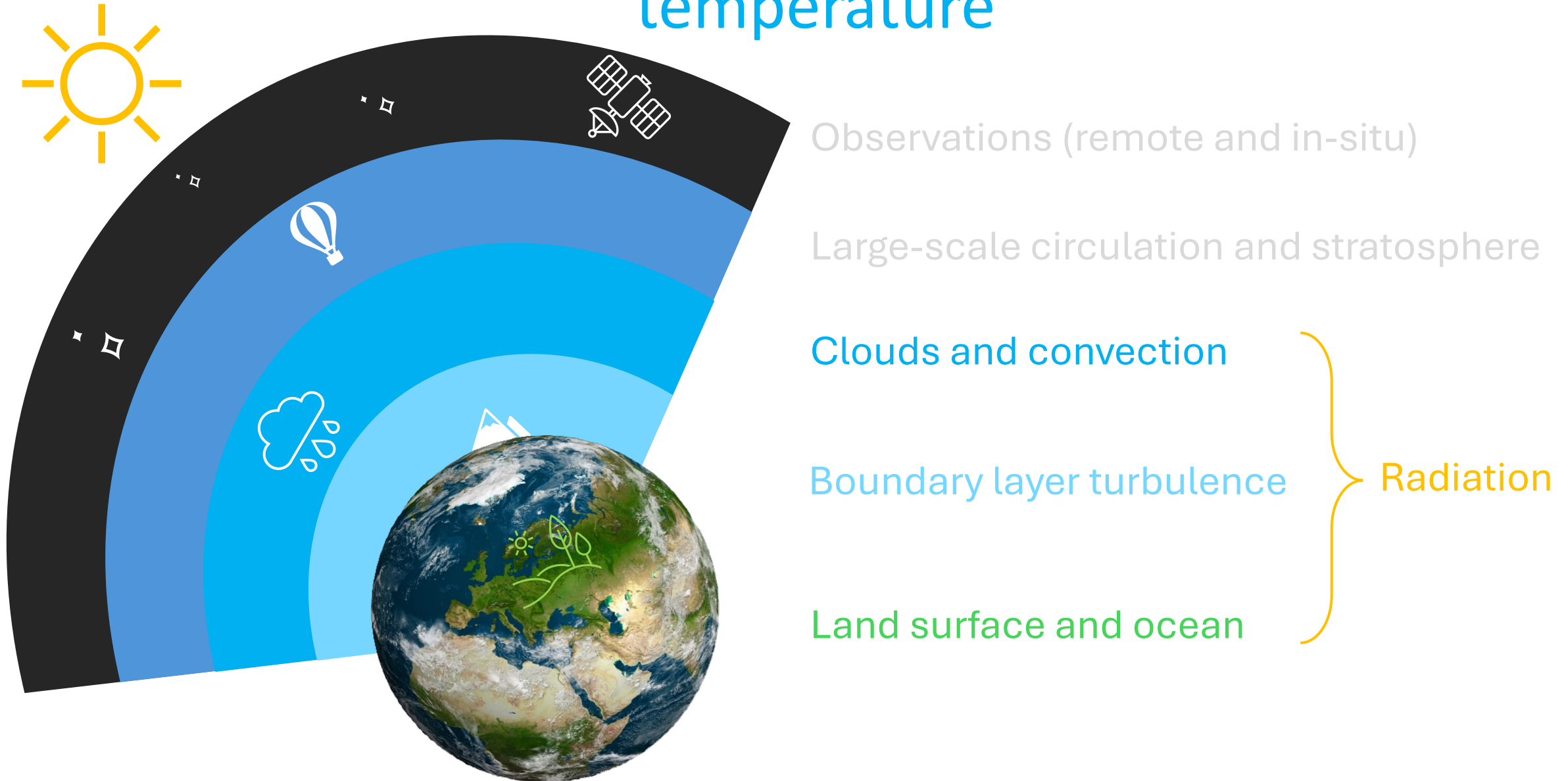
Confidence range 95% with AR(2) inflation and Sidak correction for 4 independent tests.

JJA 2024
IFS cycle 50r1



Significant positive impact of soil DA on low level atmospheric temperature forecasts

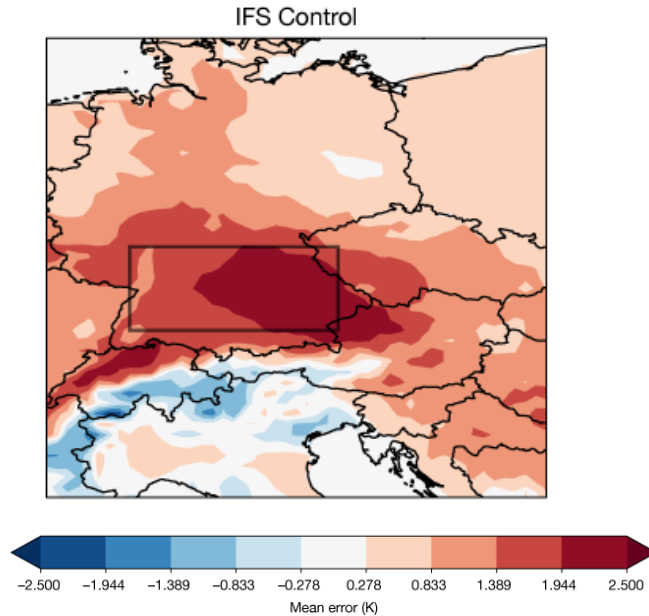
Interaction of processes that impact forecast temperature



Example: Low cloud cover and temperature errors

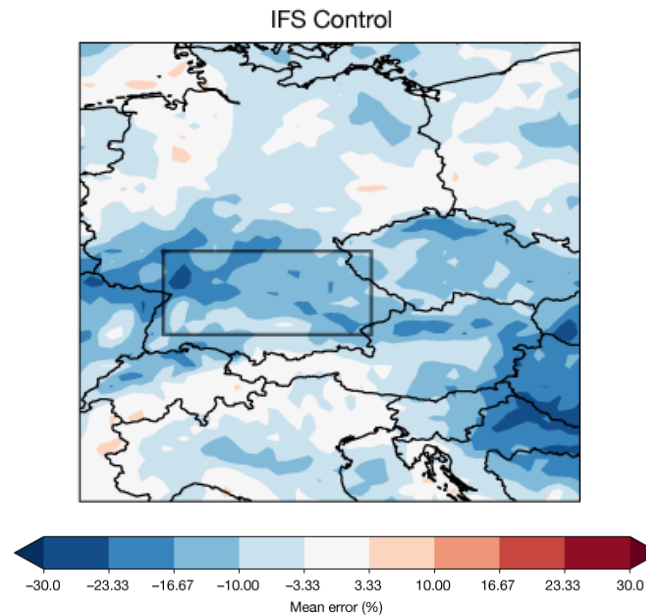
Errors in anticyclonic conditions

2m T error



Too warm 2m
temperatures

Low cloud cover error

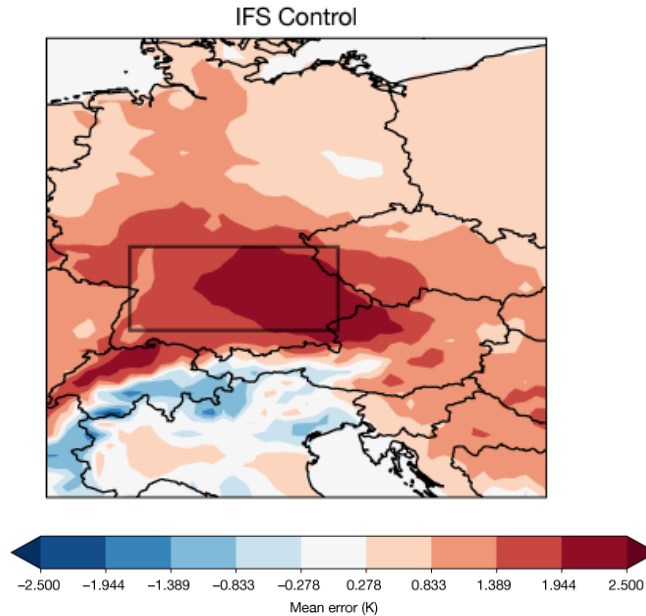


...because of too little low
cloud

Example: Low cloud cover and temperature errors

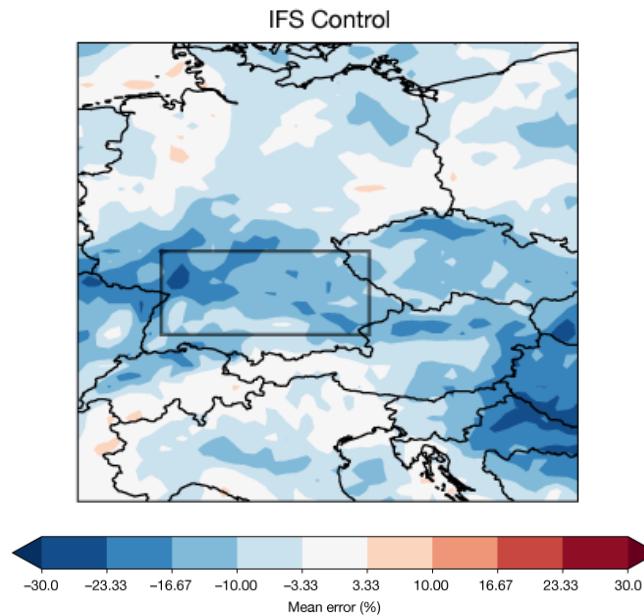
Errors in anticyclonic conditions

2m T error



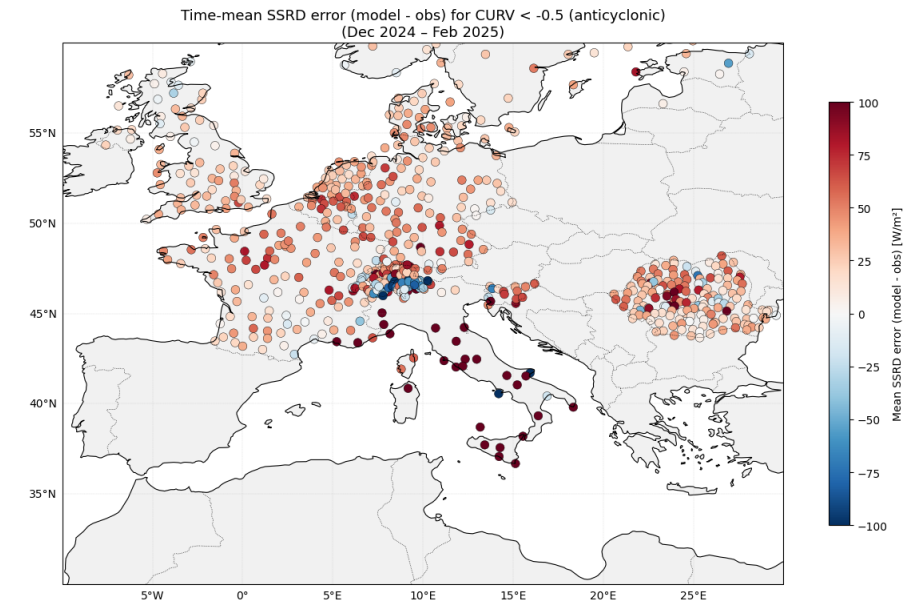
Too warm 2m
temperatures

Low cloud cover error



...because of too little low
cloud

Surface SW radiation error

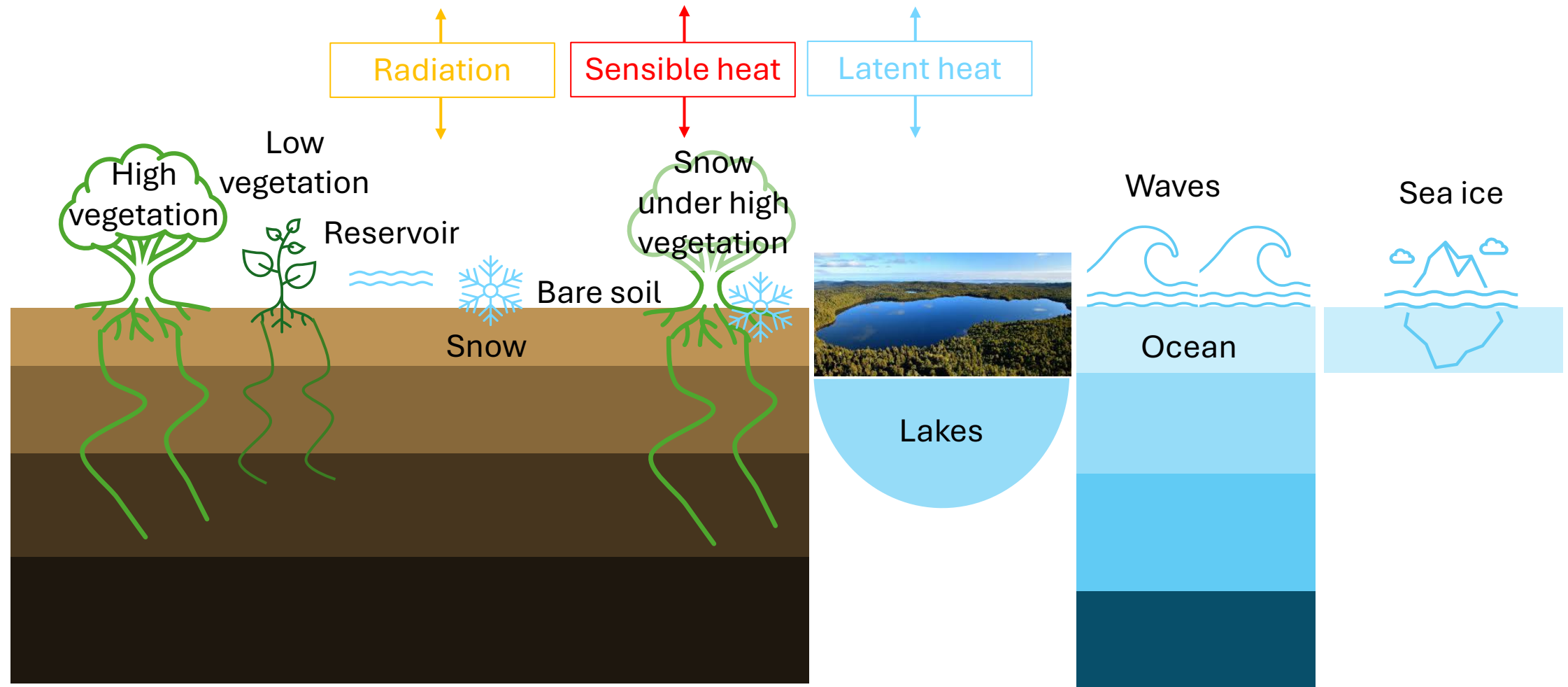


Too much SW radiation
reaching the surface,
heating the surface and
increasing fluxes

Land surface, waves and ocean

$$R_{LW} + R_{SW} + SH + LH = \Lambda(T_{10m} - T_{surf})$$

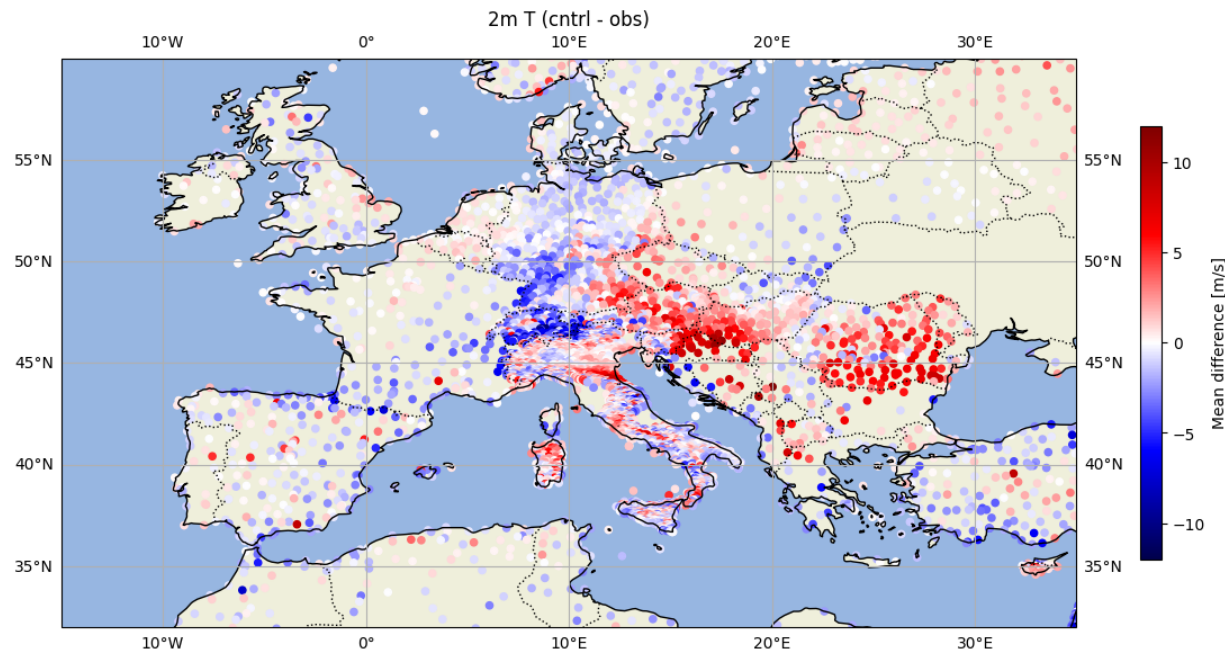
First model level ~10m above surface



Example: case of 2m T errors in anticyclonic condition

Model 2m temperature error T+72 h

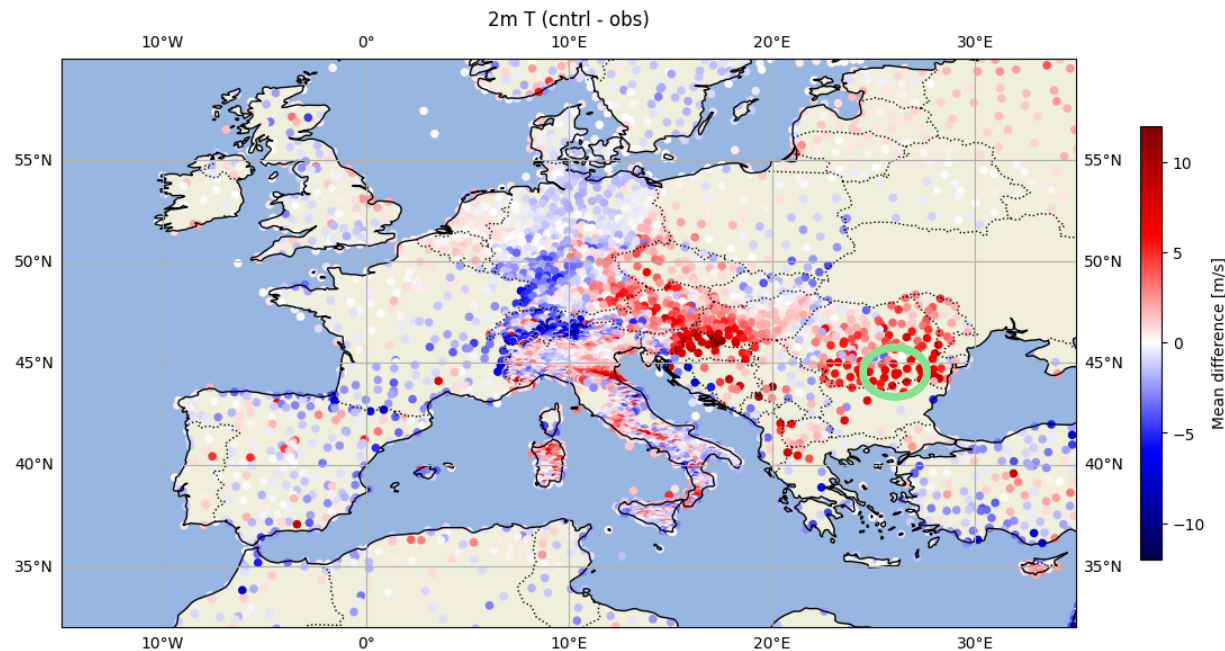
21 Dec 2022



Large +ve errors ~8K

Example: case of 2m T errors in anticyclonic condition

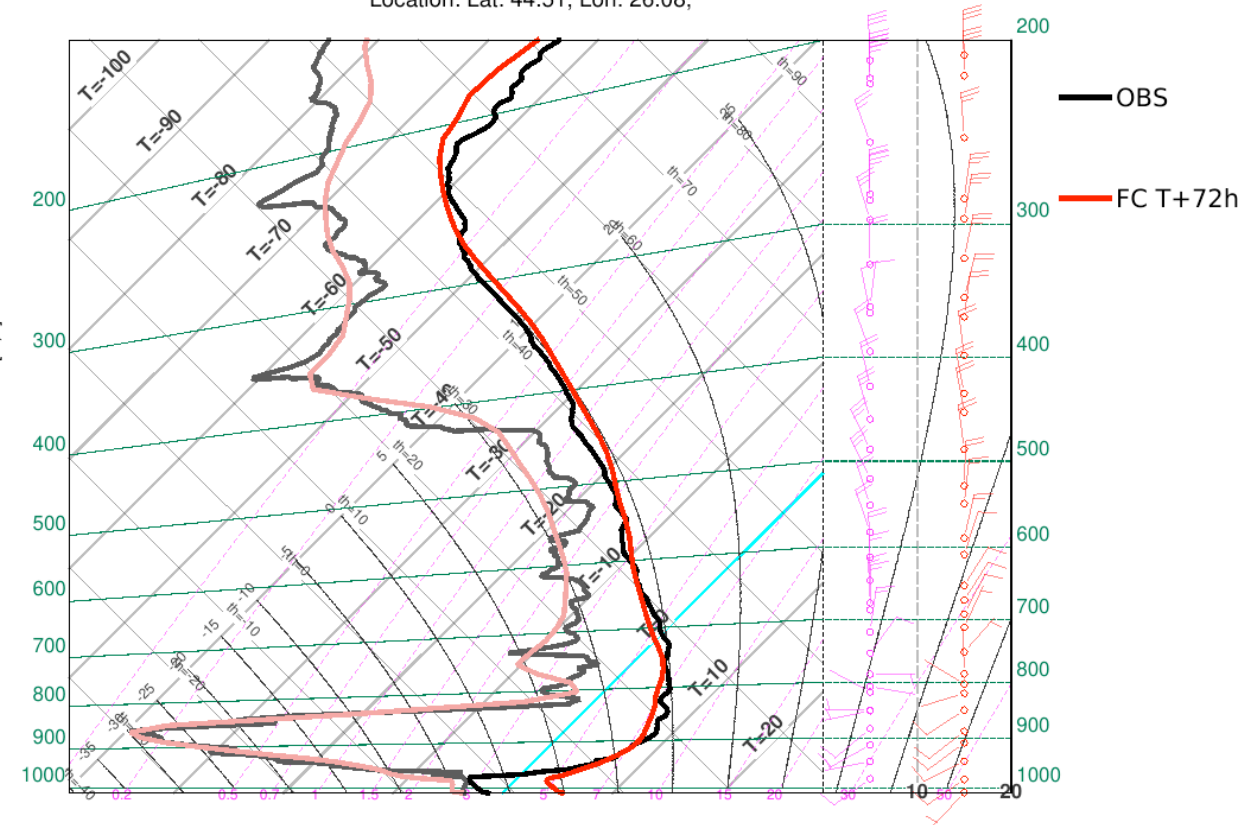
Model 2m temperature error T+72 h
21 Dec 2022



Large +ve errors ~8K

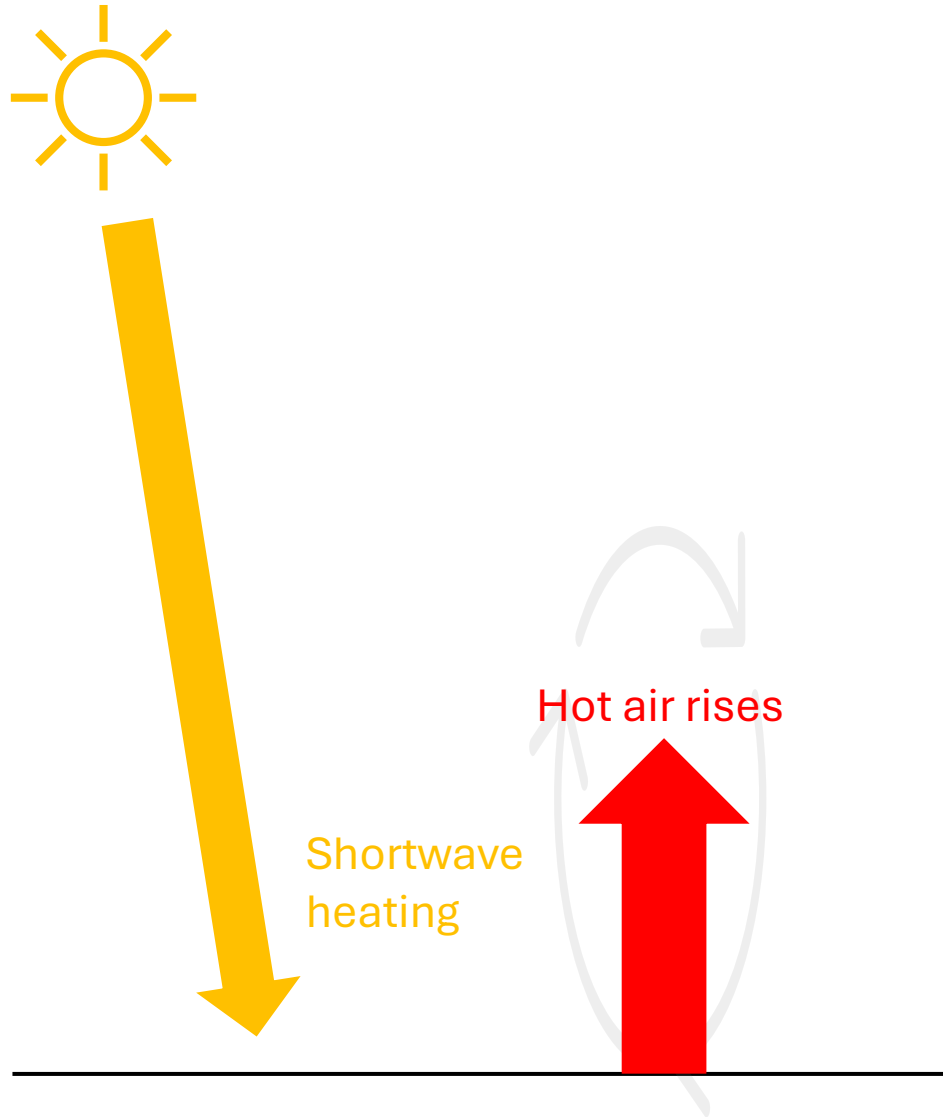
Radiosonde and model at
Bucharest

21/12/2022 11:03 UTC; WMO id: 15420, station height: --- m
Location: Lat: 44.51, Lon: 26.08,

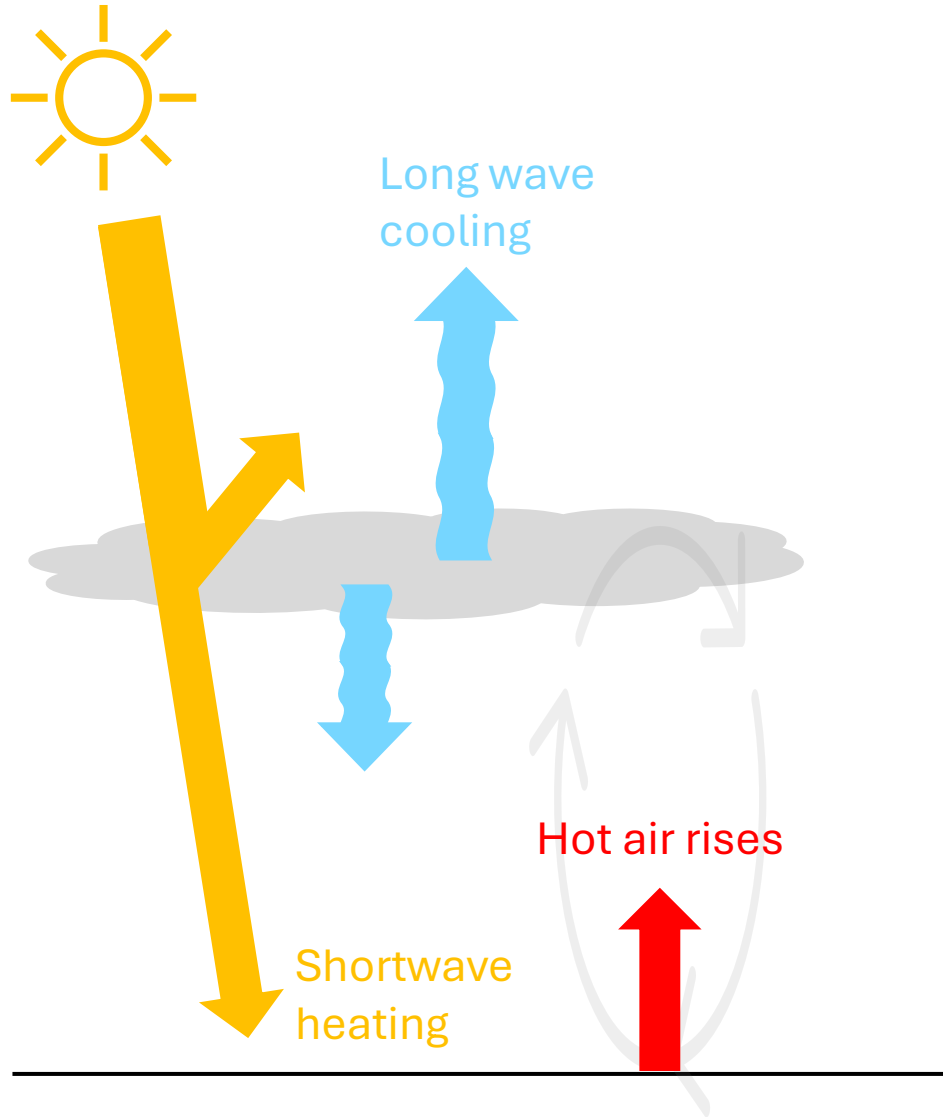


Strong inversion and low cloud
not capture in model

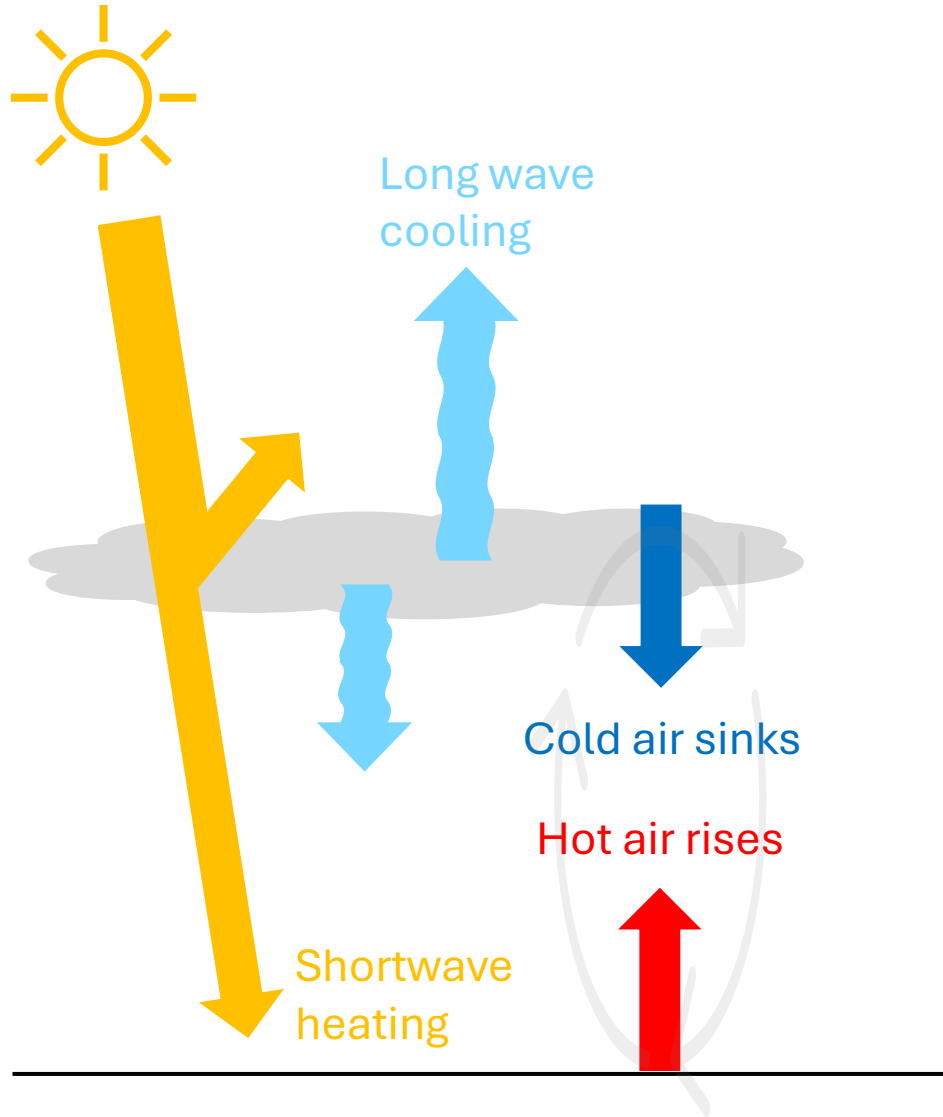
Example: Cloud-top driven mixing



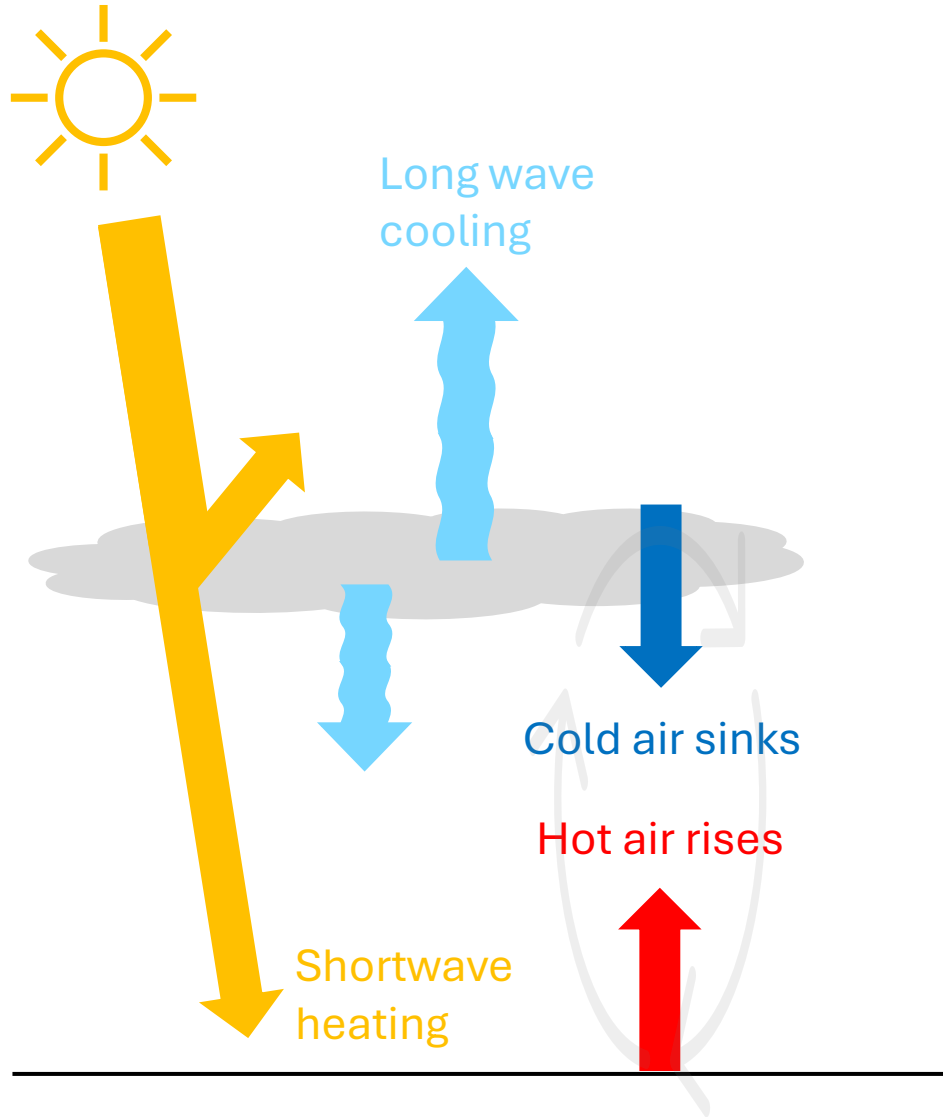
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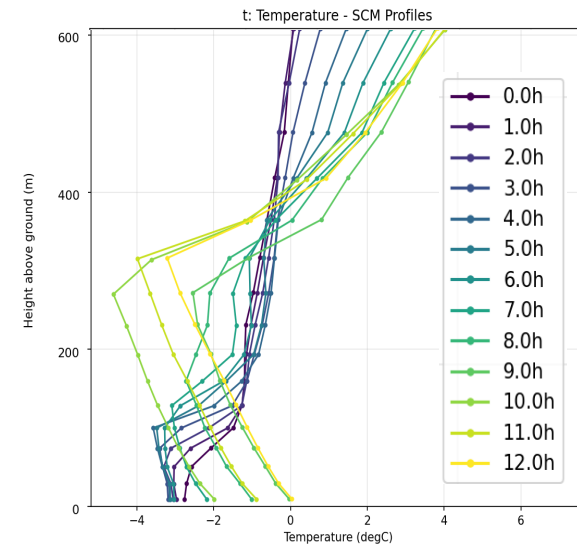
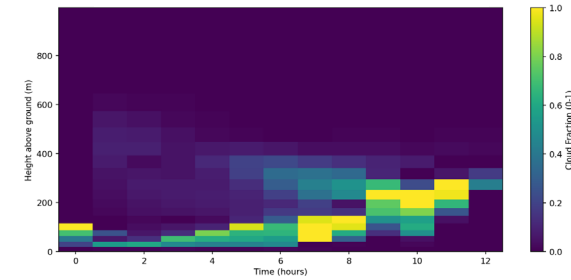
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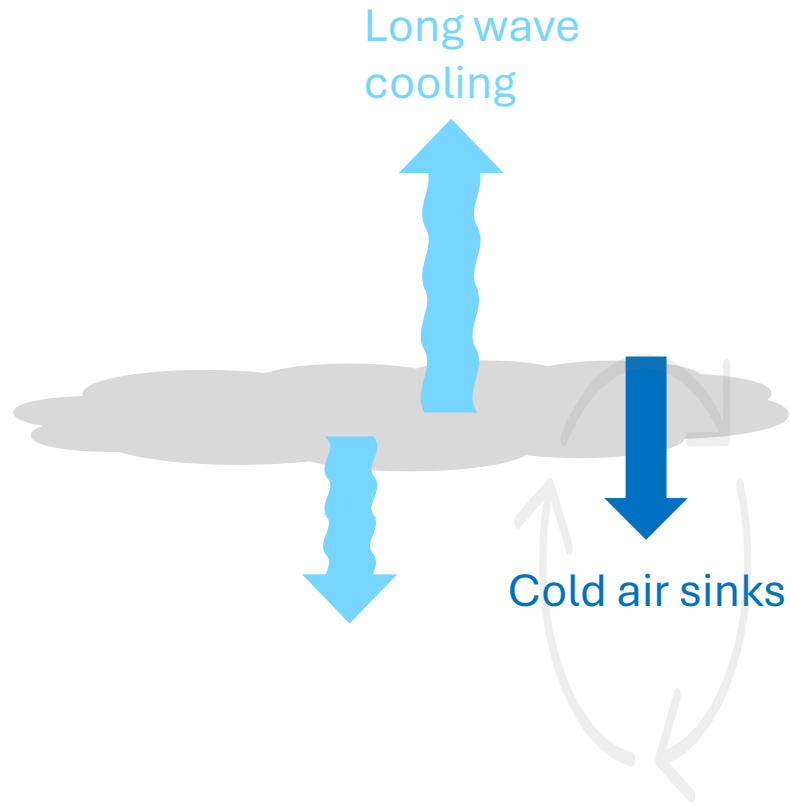
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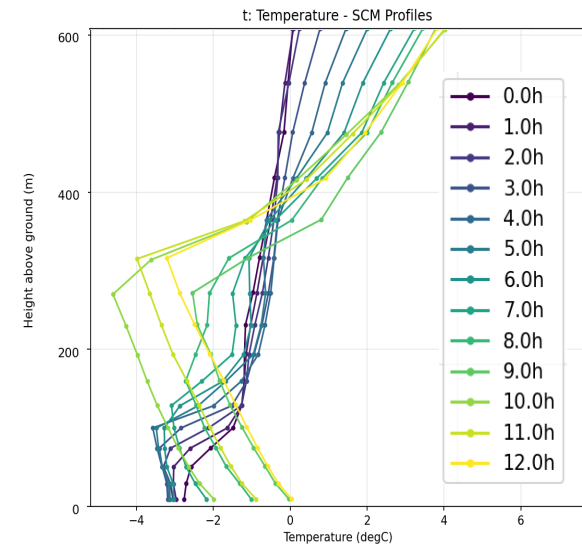
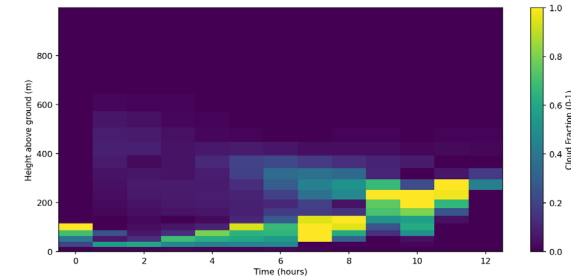
Analysis
Cloud fraction



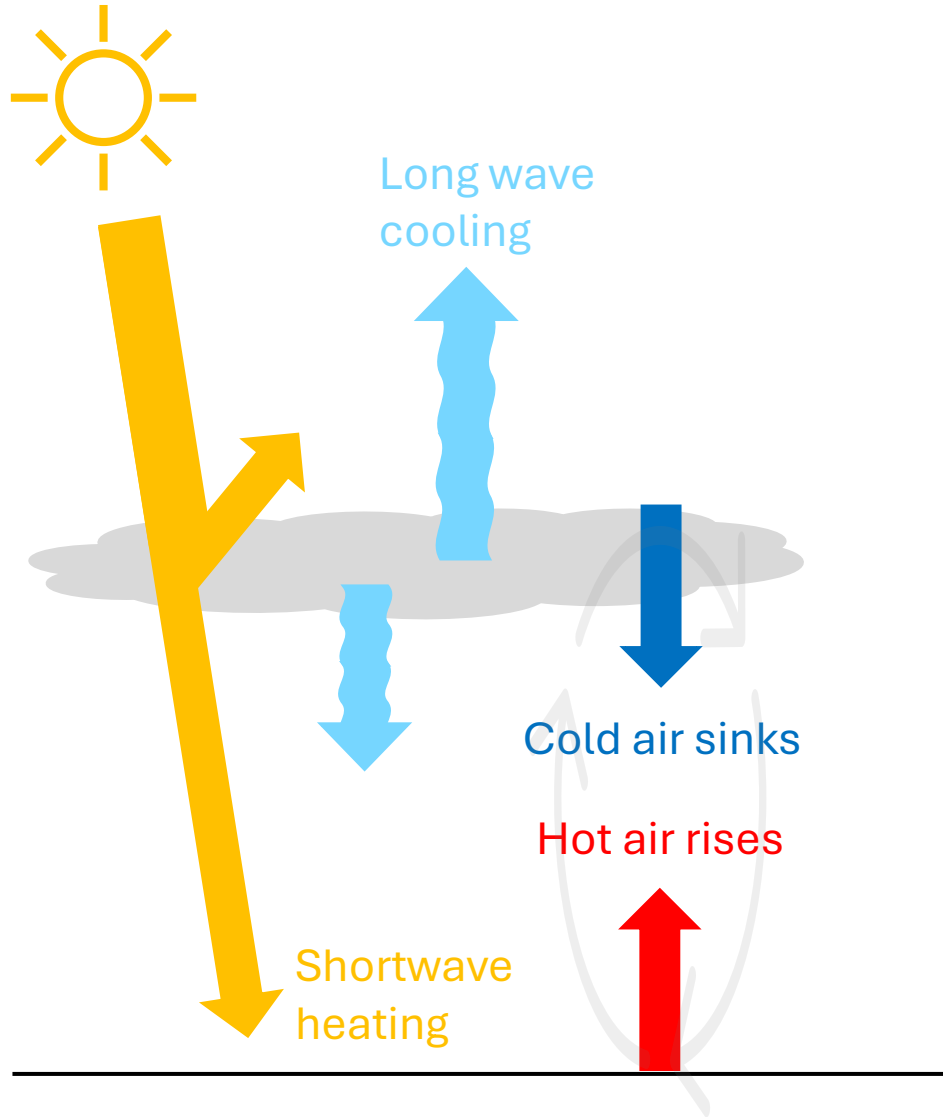
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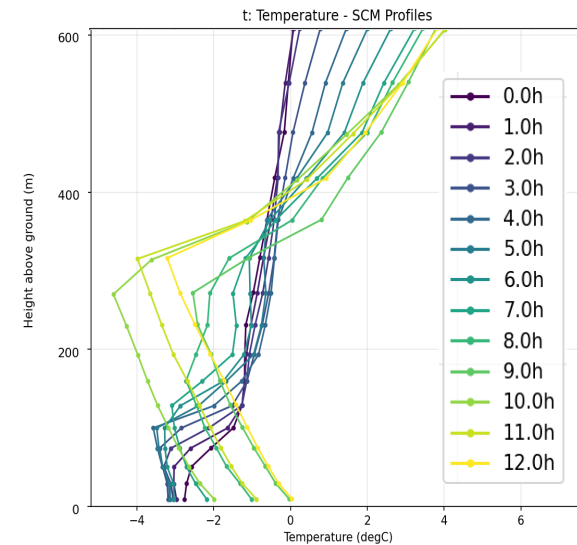
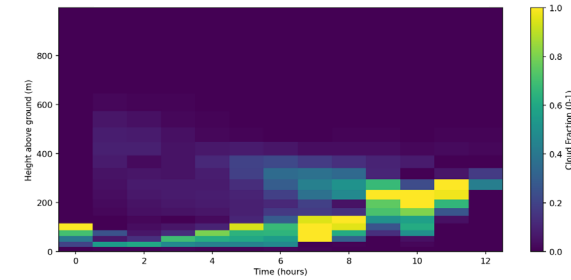
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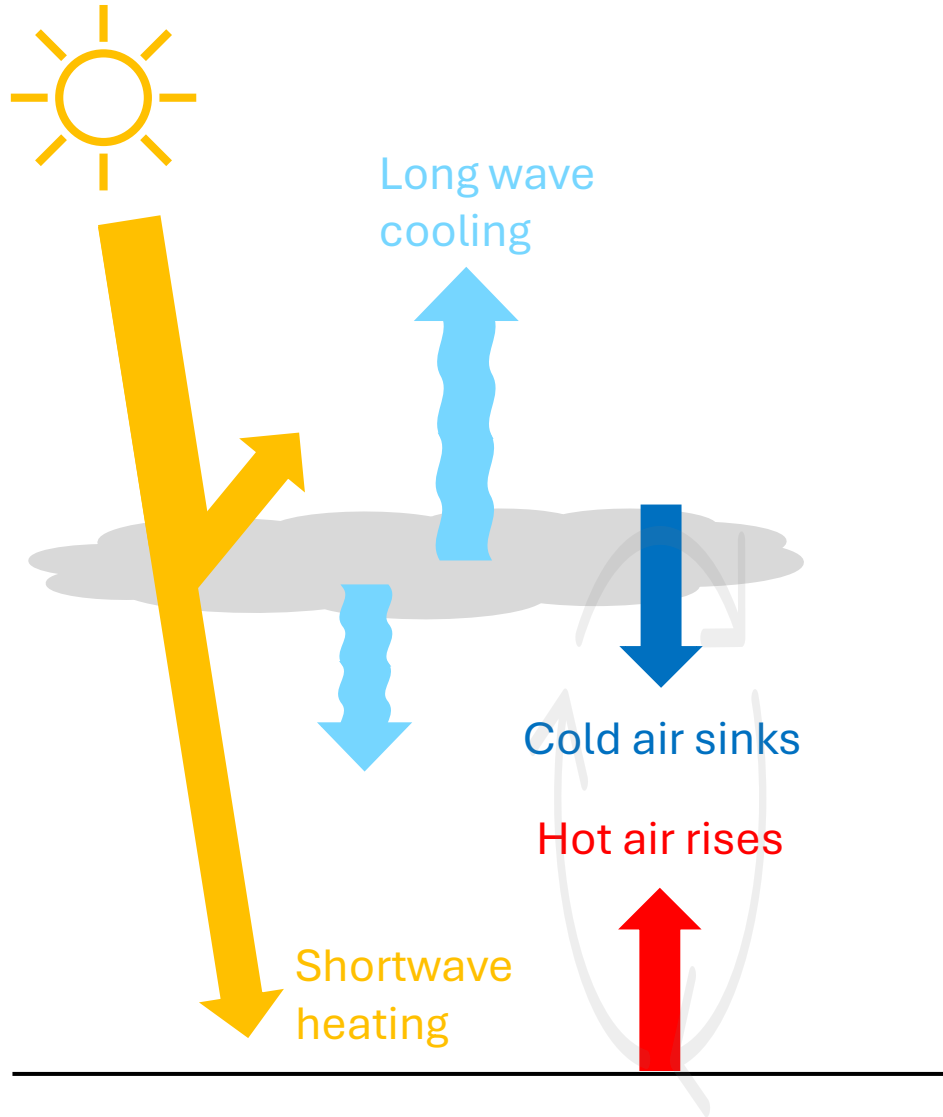
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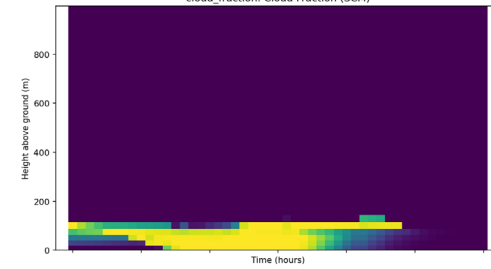
Analysis
Cloud fraction



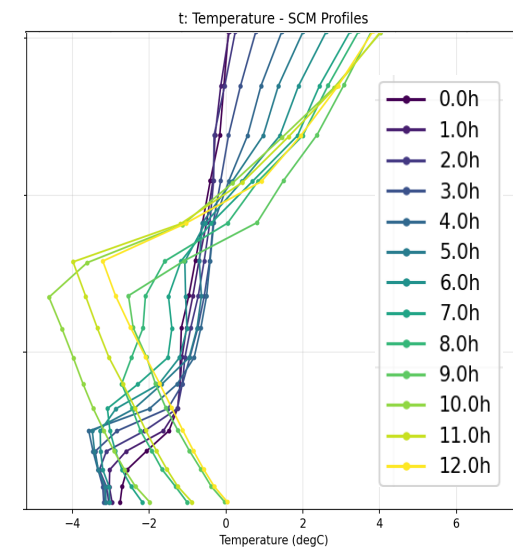
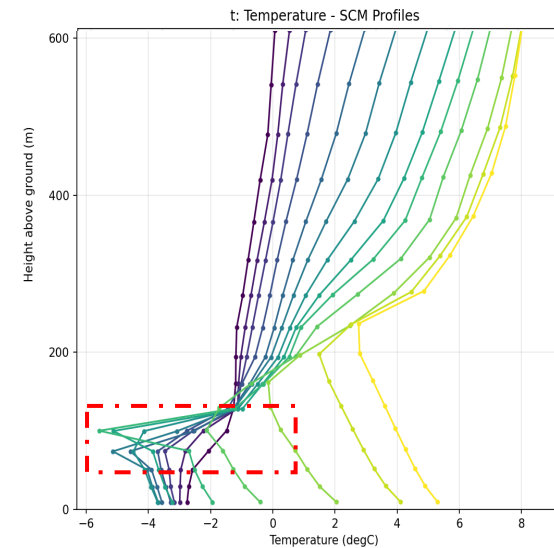
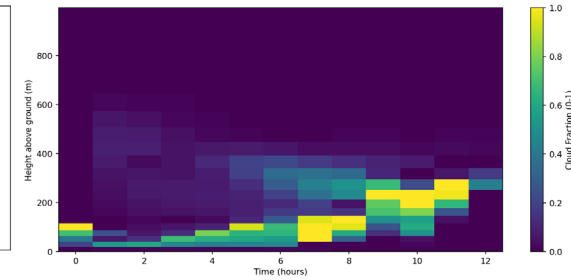
Example: Cloud-top driven mixing



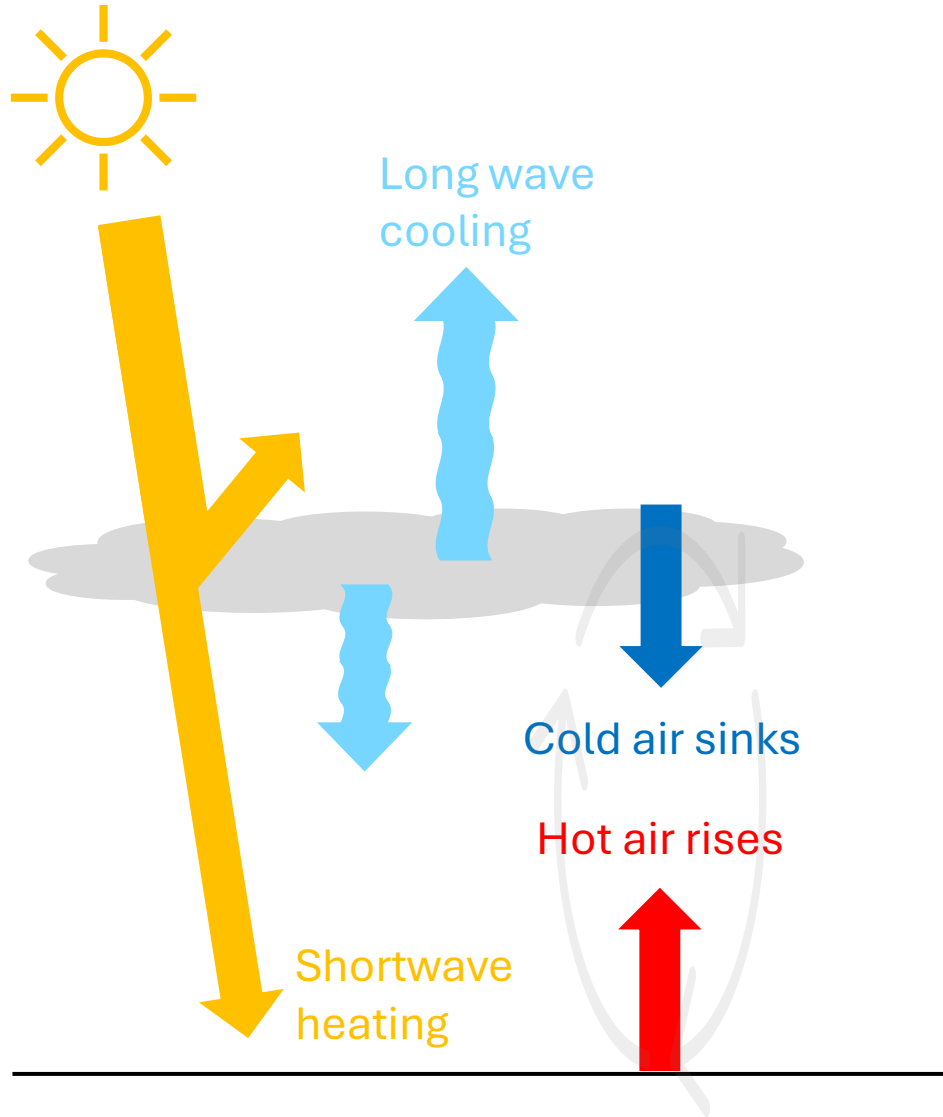
Control
Cloud fraction



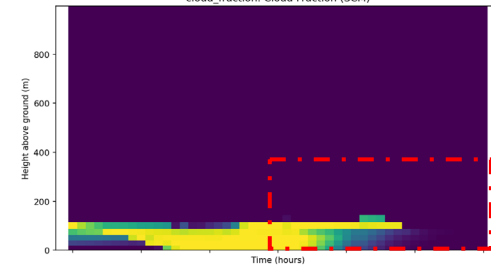
Analysis
Cloud fraction



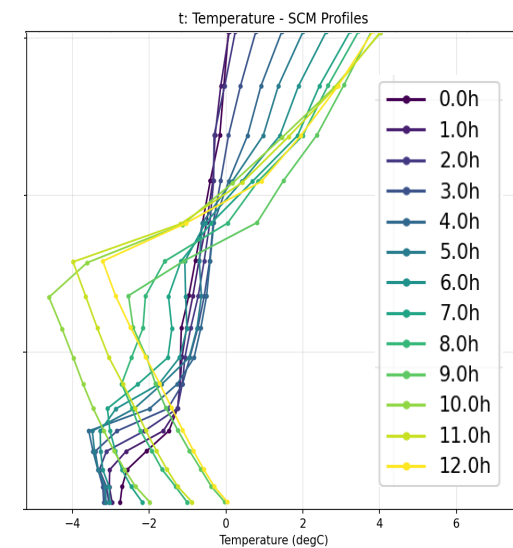
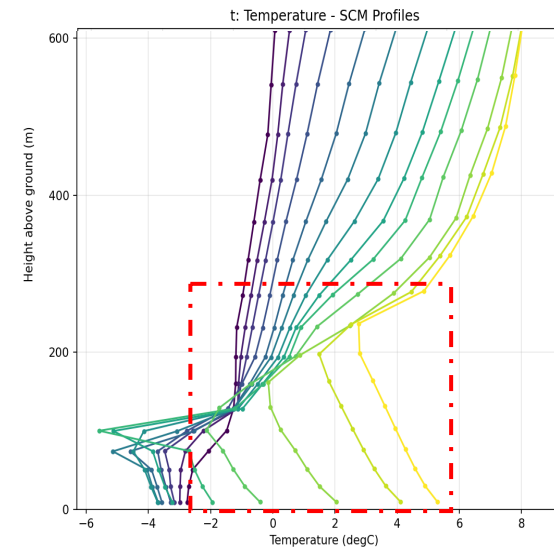
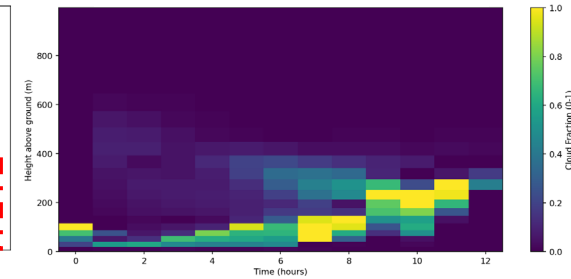
Example: Cloud-top driven mixing



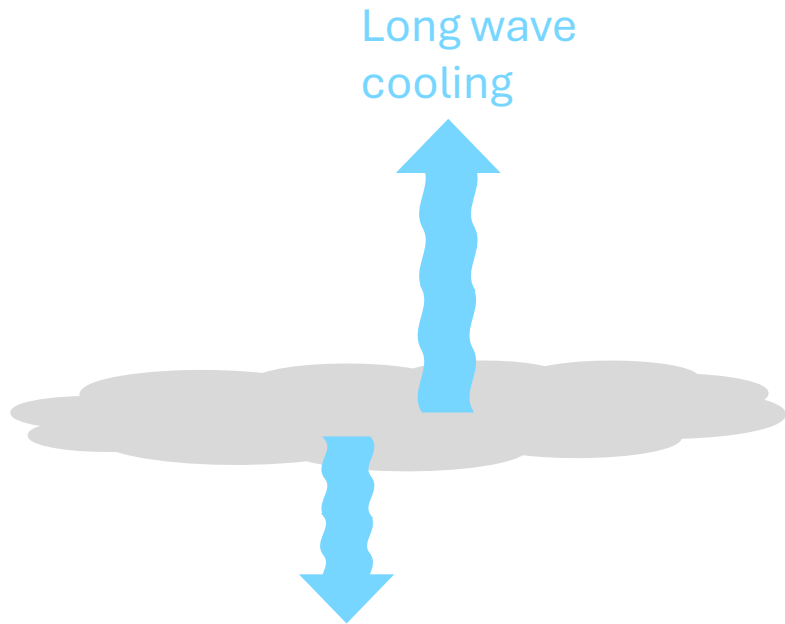
Control
Cloud fraction



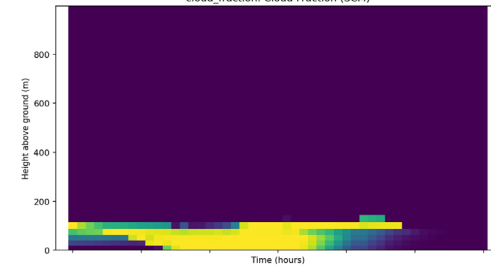
Analysis
Cloud fraction



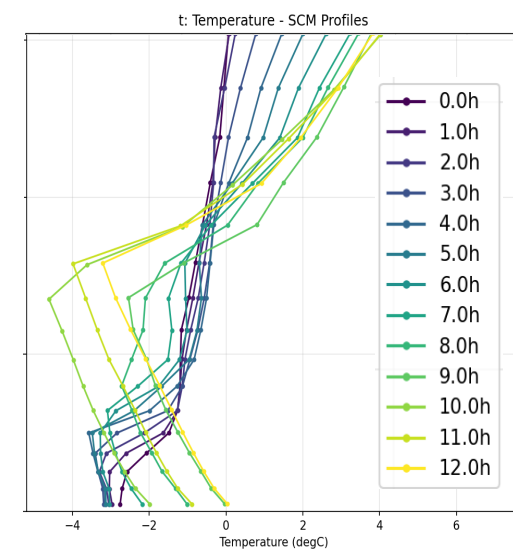
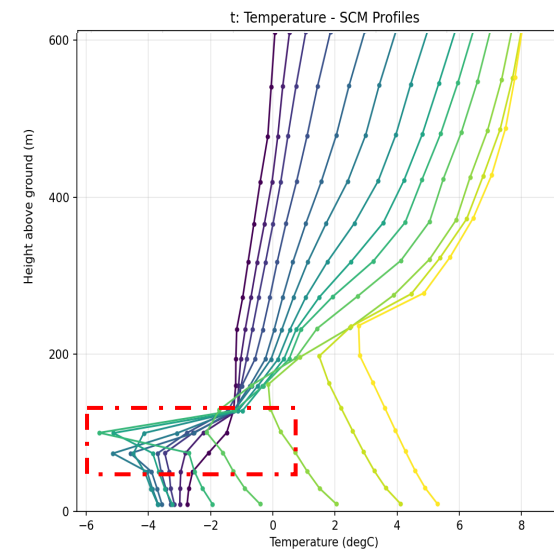
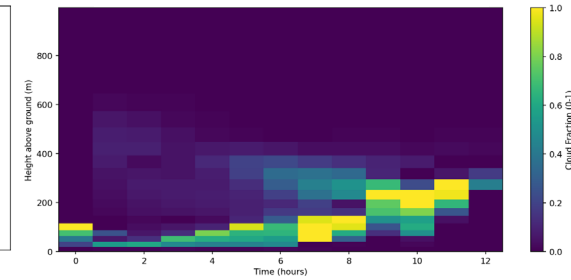
Example: Cloud-top driven mixing



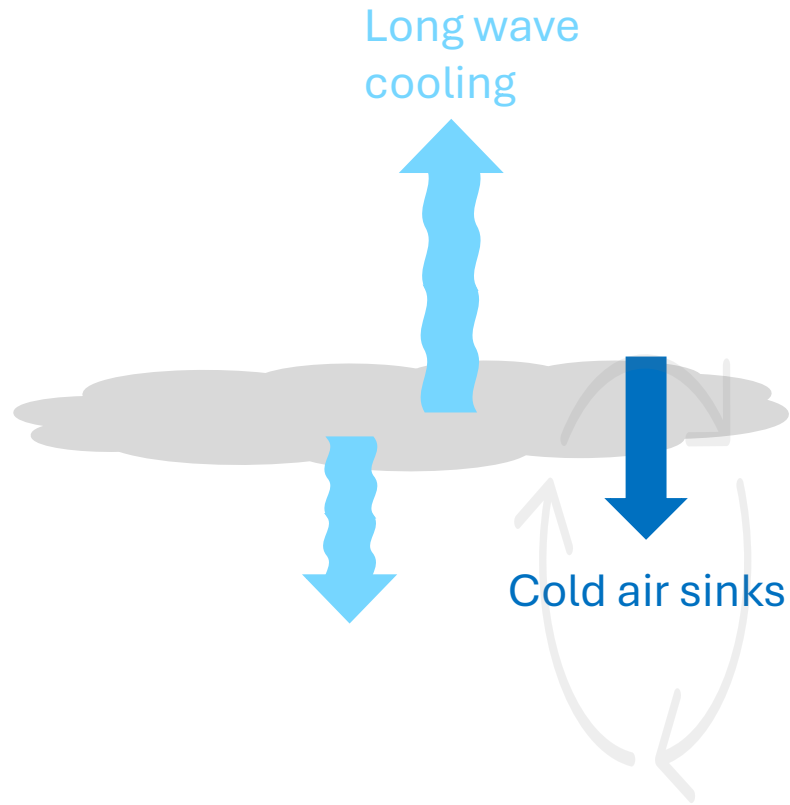
Control
Cloud fraction



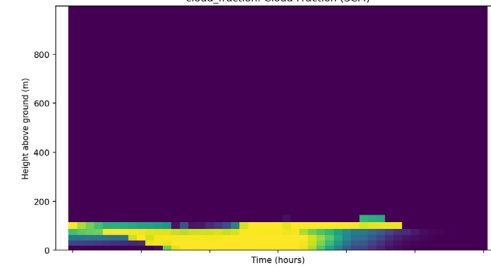
Analysis
Cloud fraction



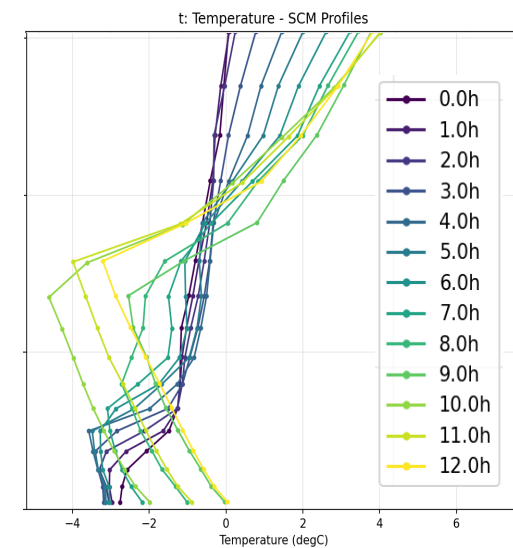
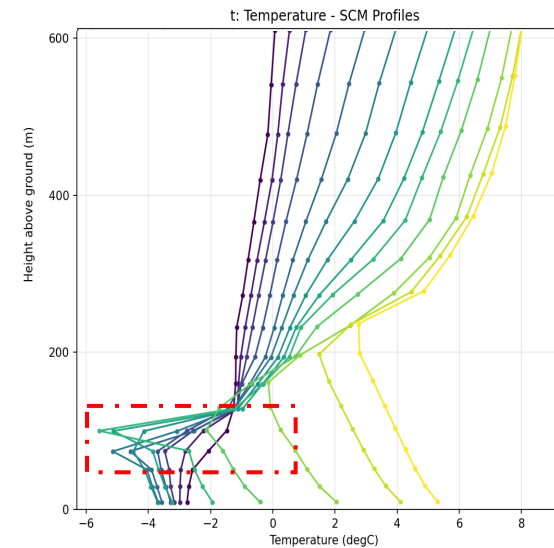
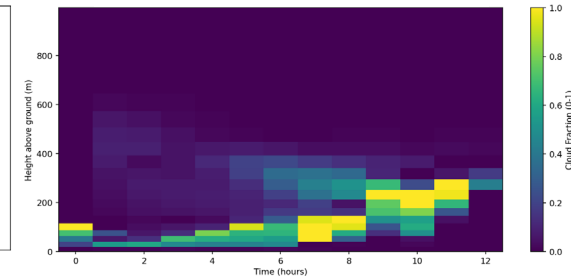
Example: Cloud-top driven mixing



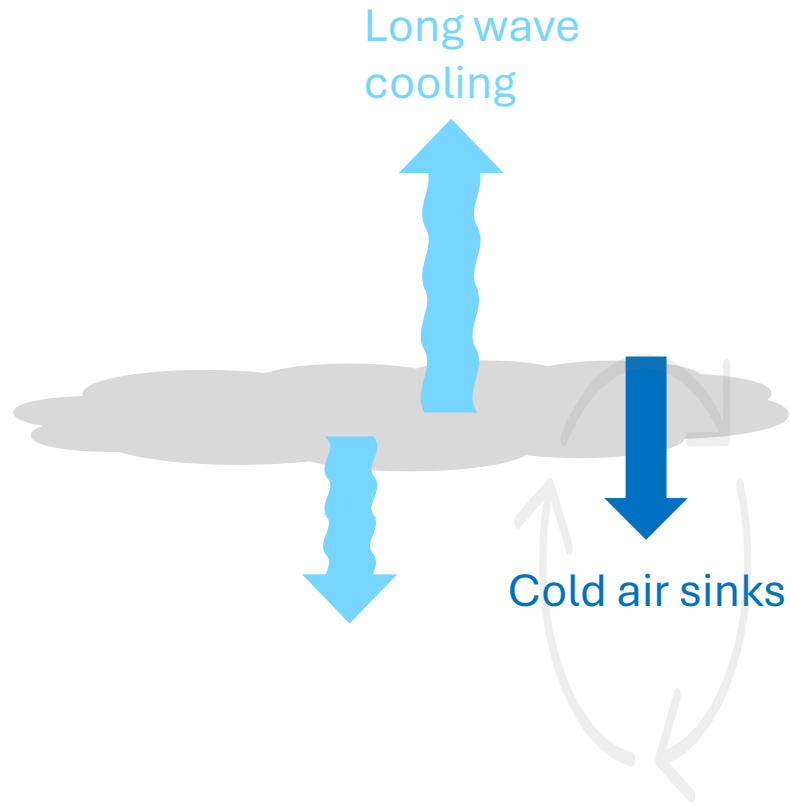
Control
Cloud fraction



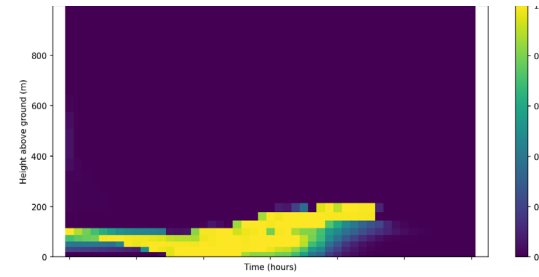
Analysis
Cloud fraction



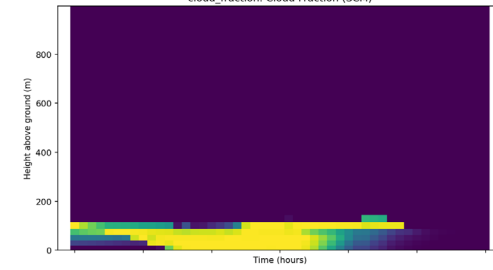
Example: Cloud-top driven mixing



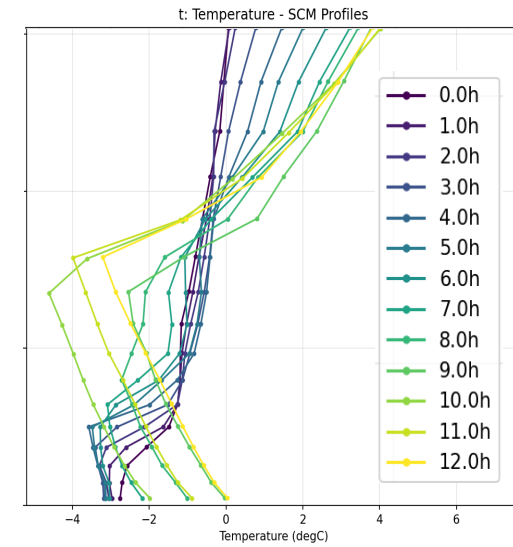
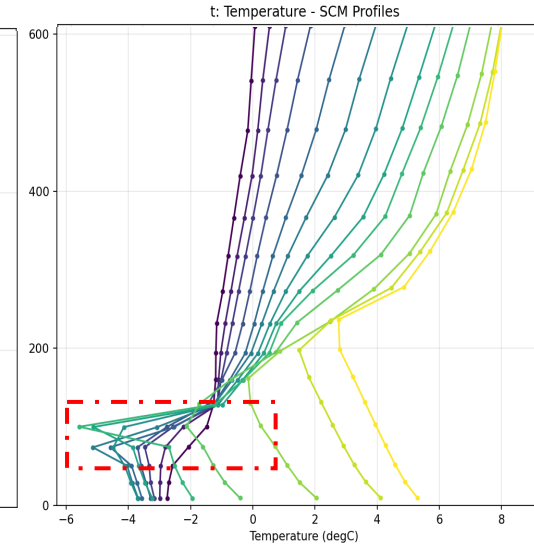
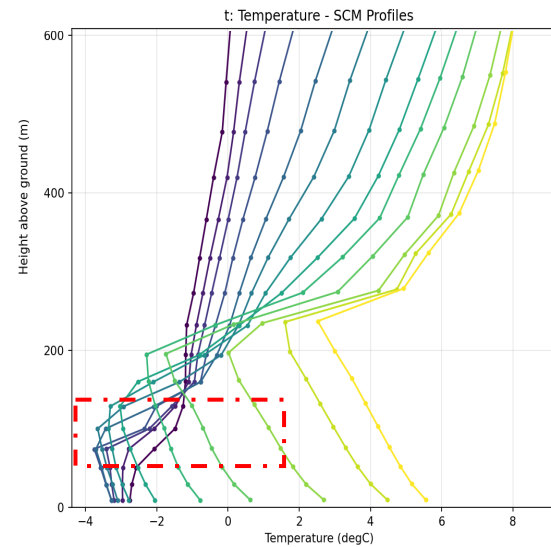
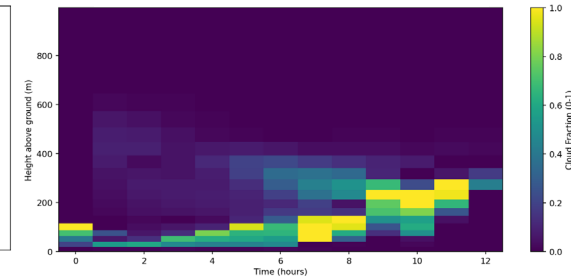
With cloud-top mixing



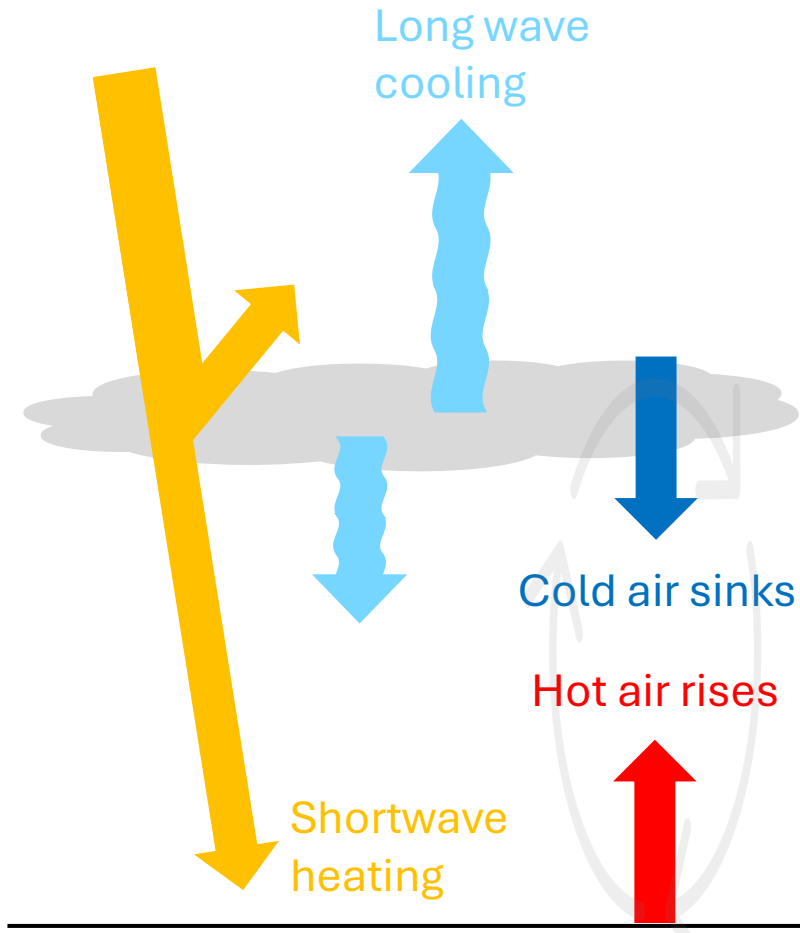
Control
Cloud fraction



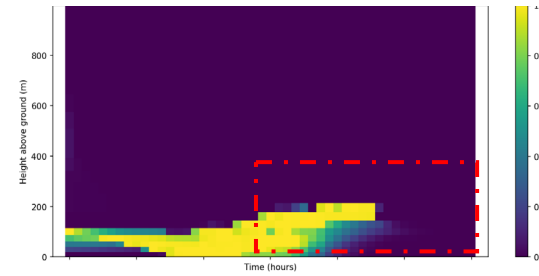
Analysis
Cloud fraction



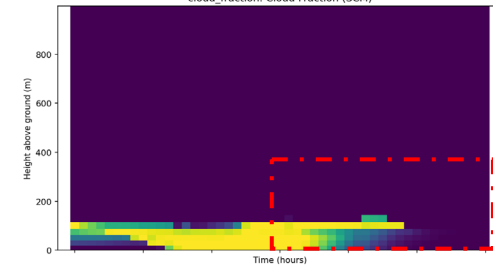
Example: Cloud-top driven mixing



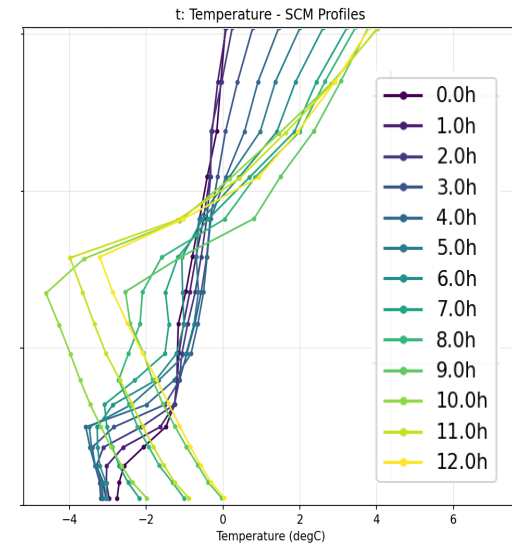
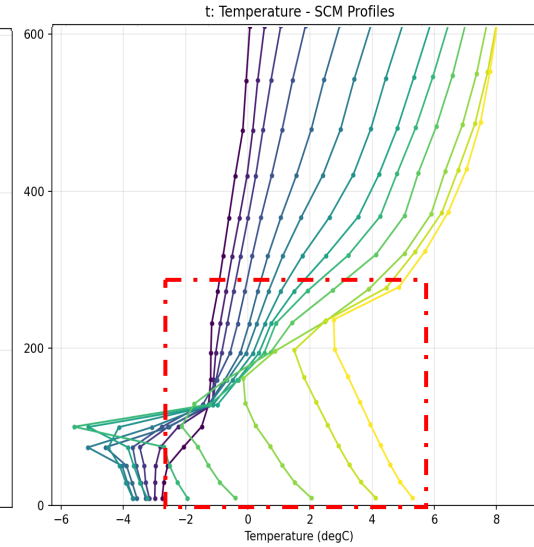
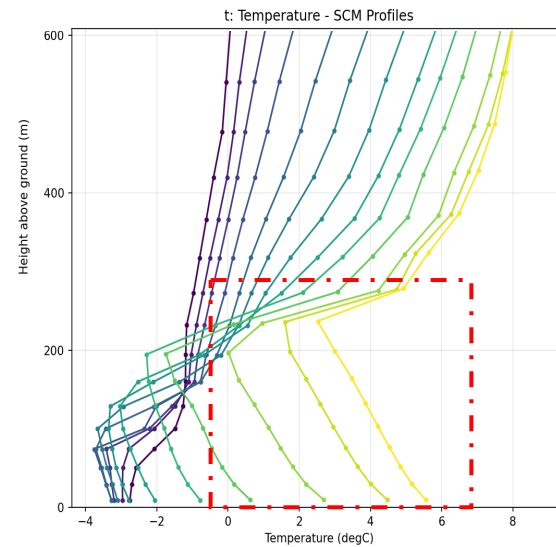
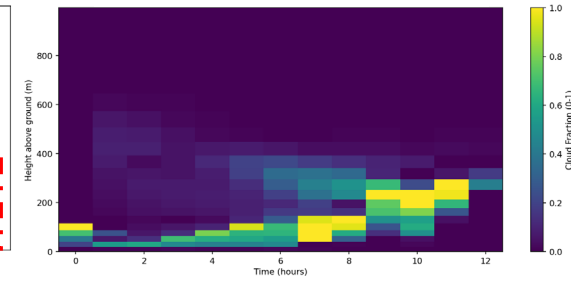
With cloud-top mixing



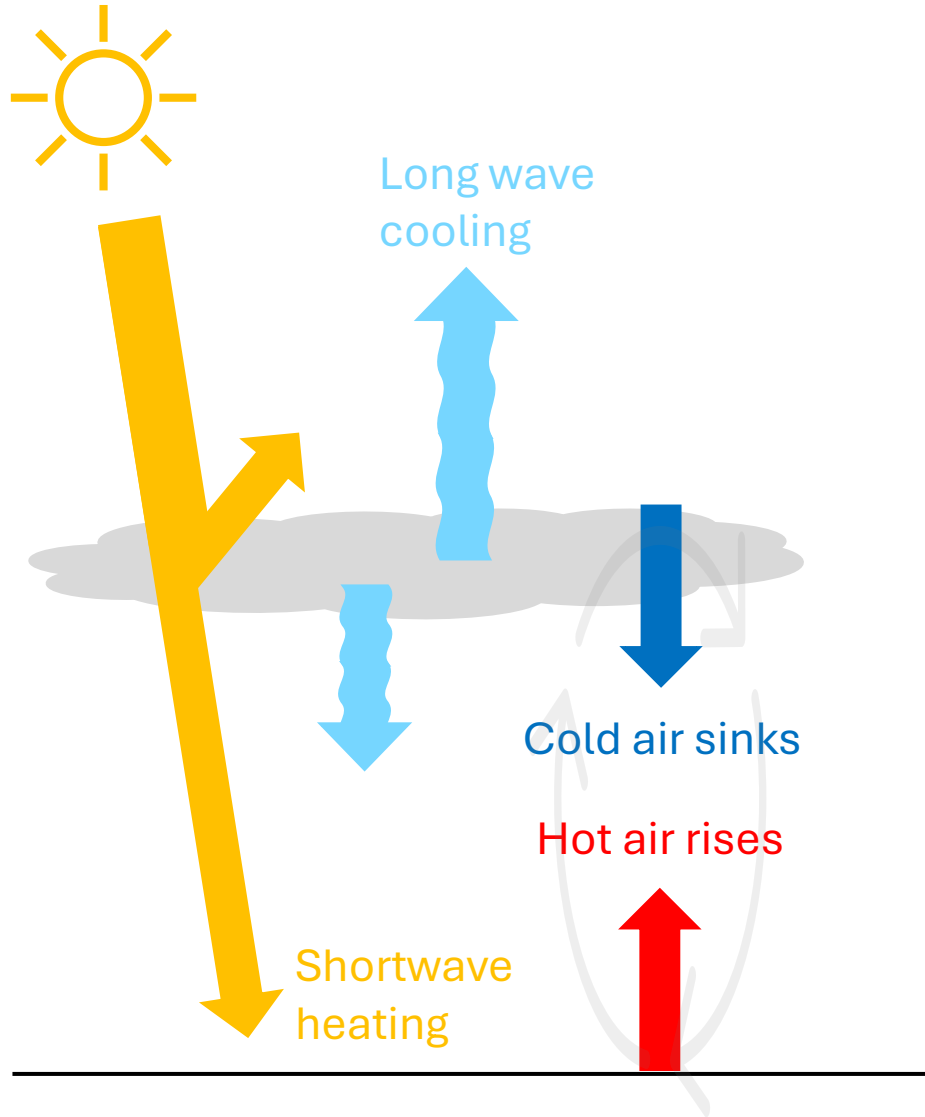
Control
Cloud fraction



Analysis
Cloud fraction

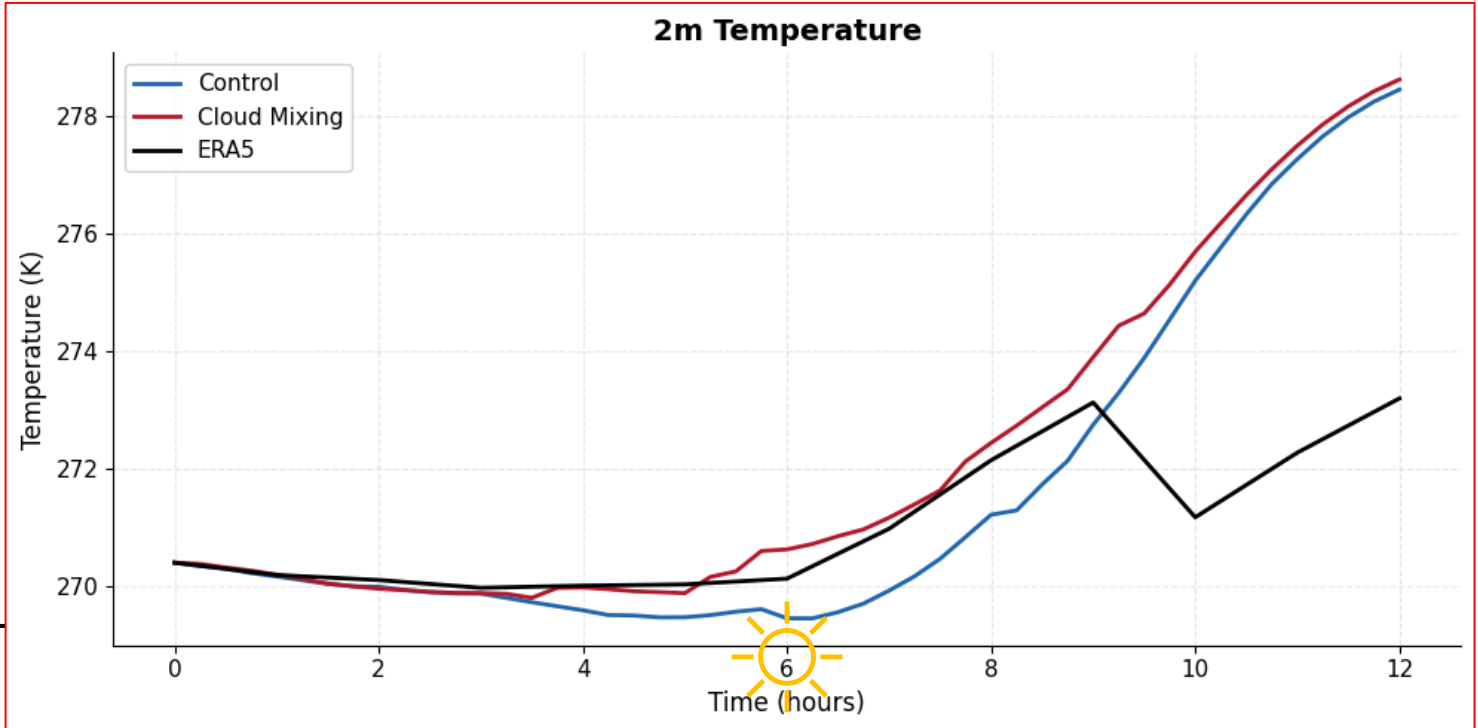


Example: Cloud-top driven mixing

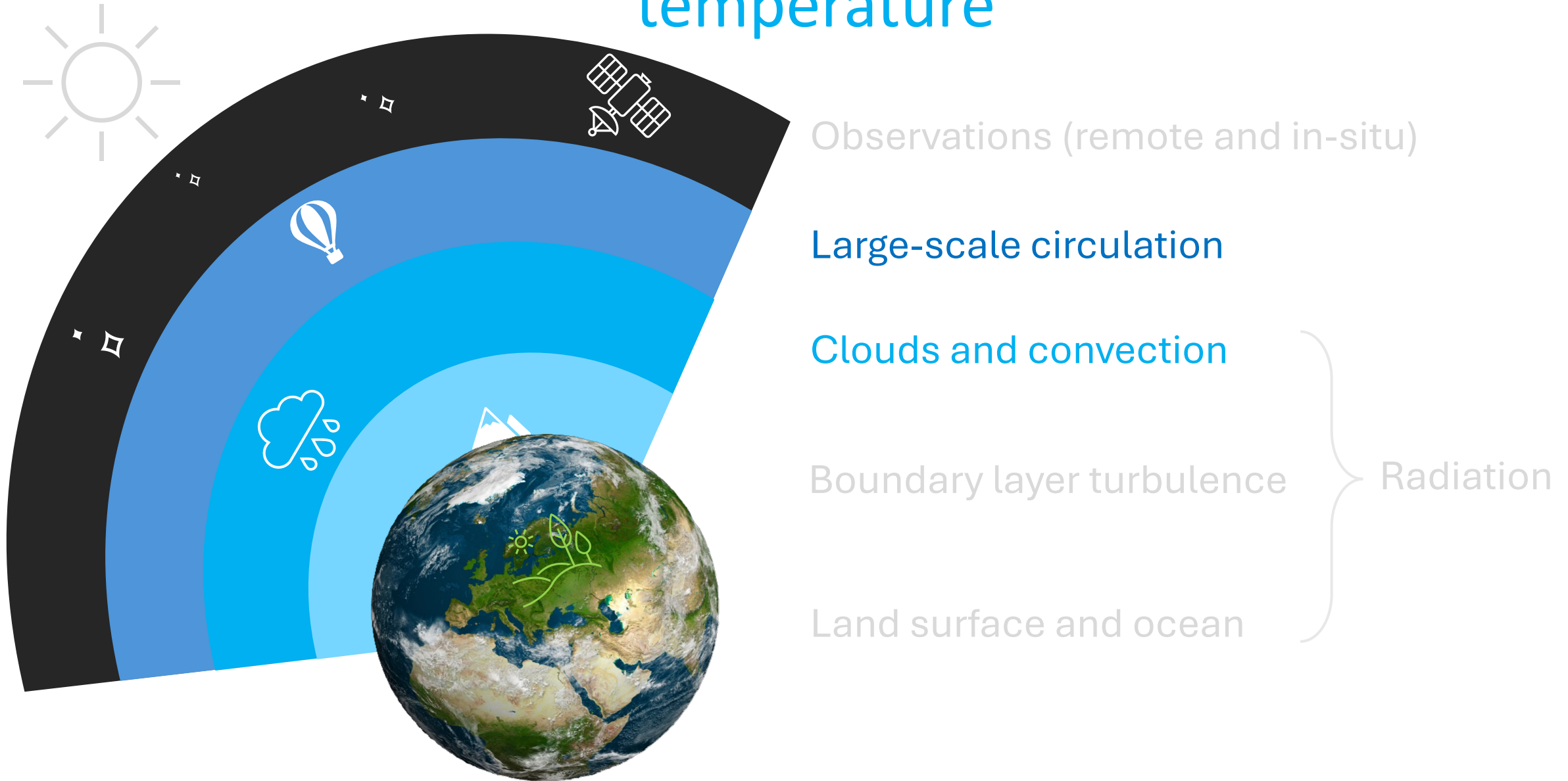


Possible causes:

Too much **turbulent diffusion** in daytime (removes cloud)
Incorrect subgrid **cloud** formation (e.g. too shallow a layer)
Insufficient surface moisture transport (e.g. not enough **moisture flux from surface**)
...others

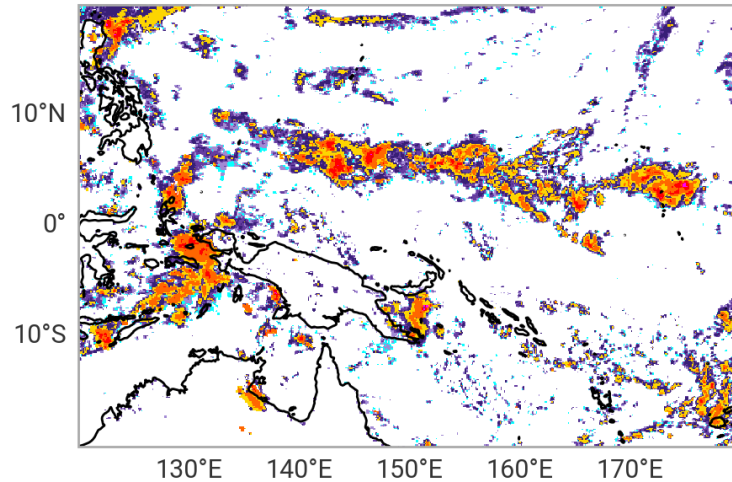


Interaction of processes that impact forecast temperature

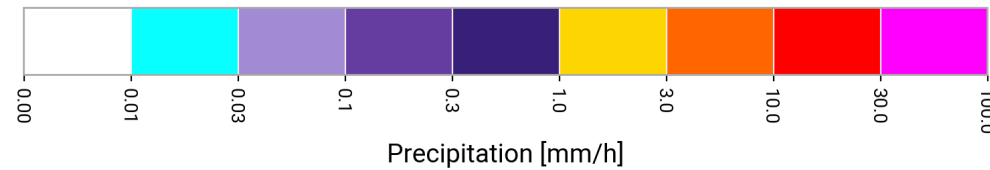
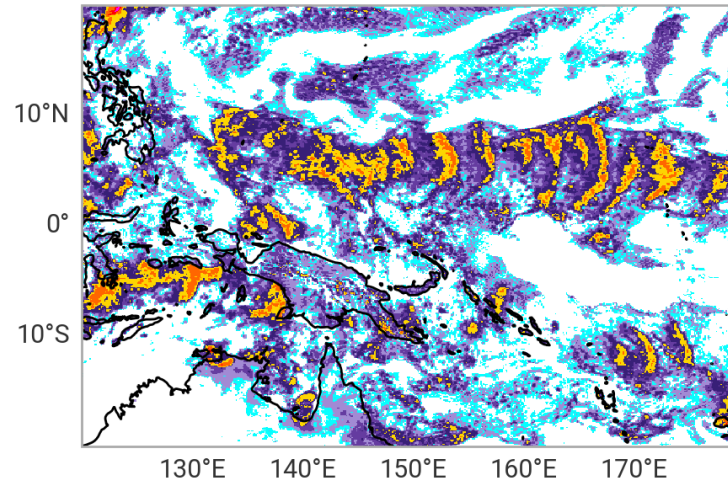


Convective organization in tropics

Observations
IMERG



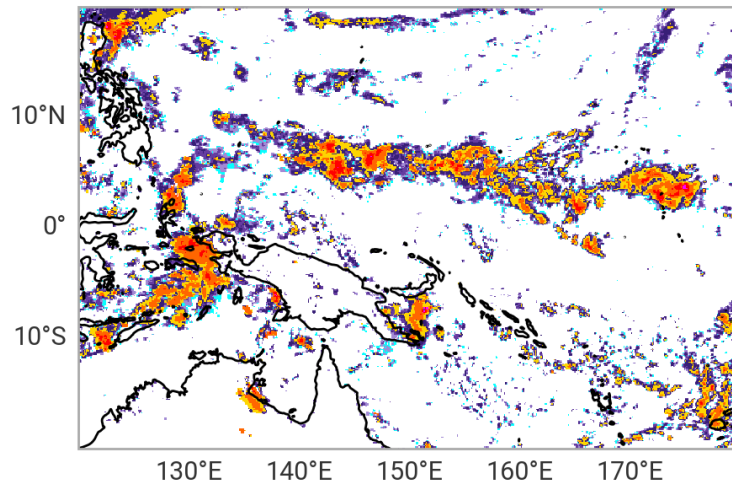
ECMWF IFS Cy50r1



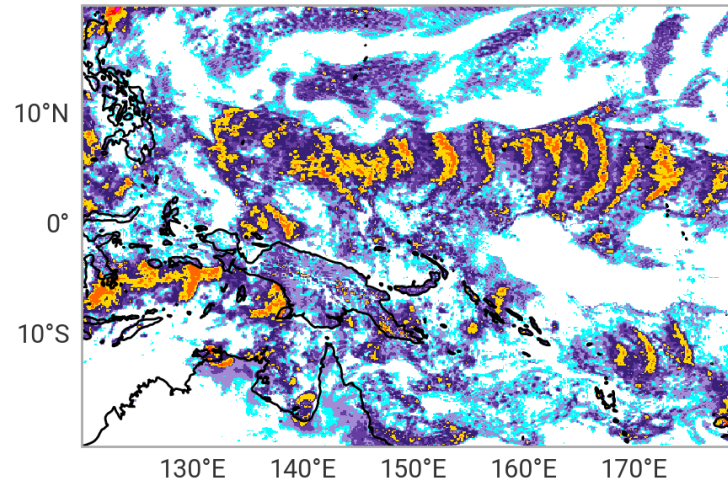
IFS at 9 km resolutions
does not exhibit enough
convective organization

Convective organization in tropics

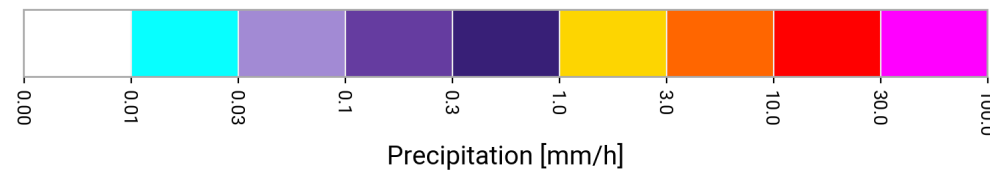
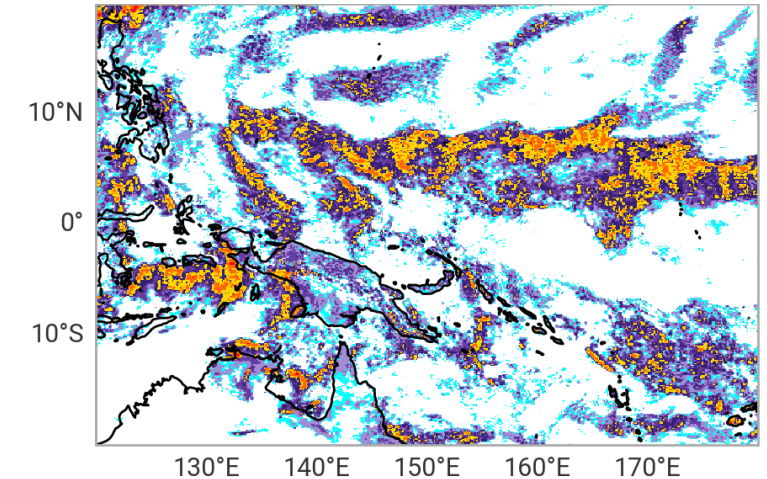
Observations
IMERG



ECMWF IFS Cy50r1



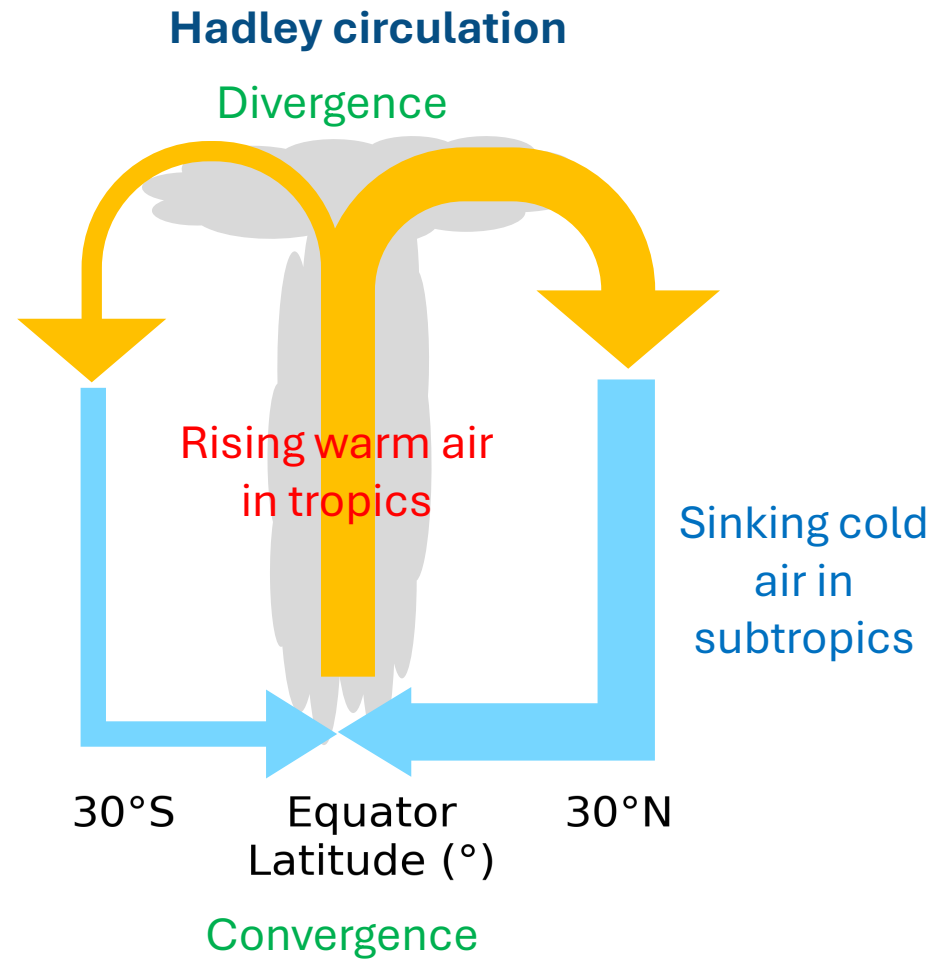
ECMWF IFS Cy50r1 +
improvements to organisation



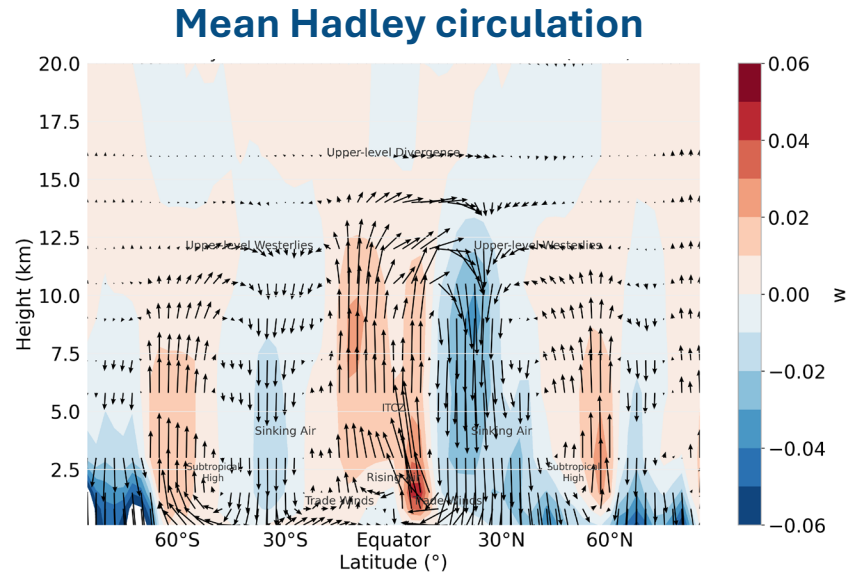
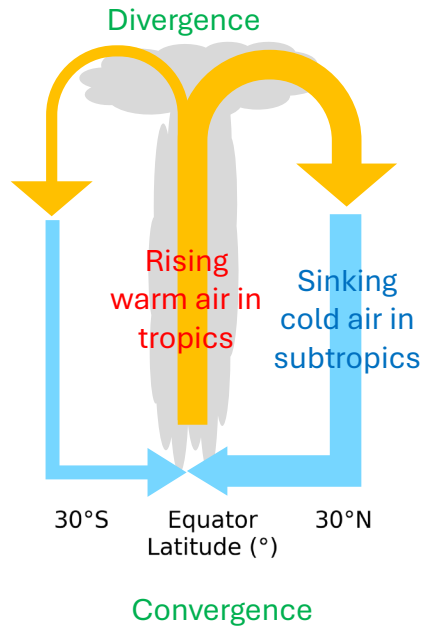
IFS at 9 km resolutions
does not exhibit enough
convective organization

Changes to convection parametrization
(reducing downdrafts) improves
convective organisation

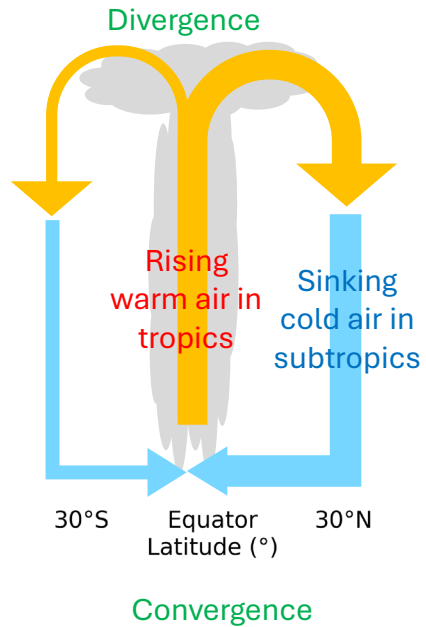
Convection in the tropics and largescale transport



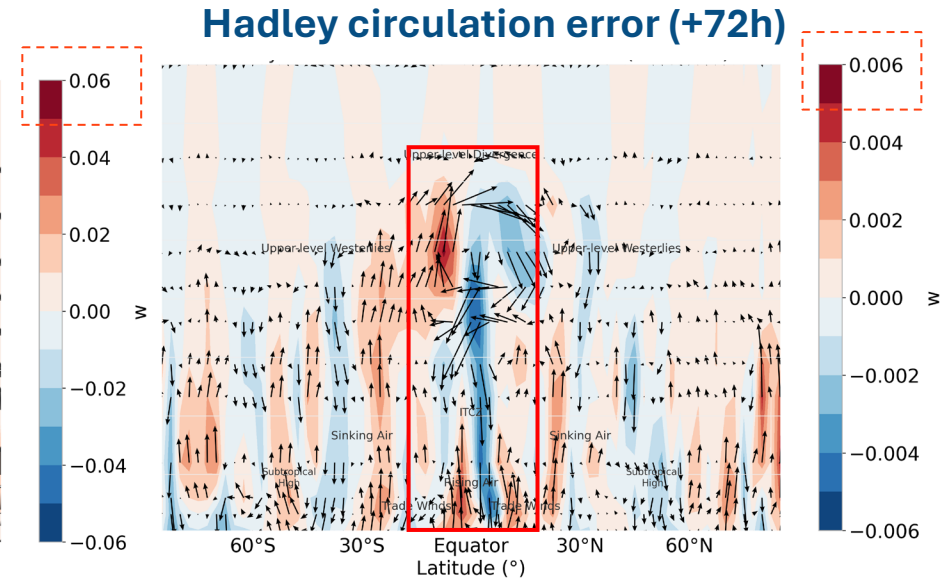
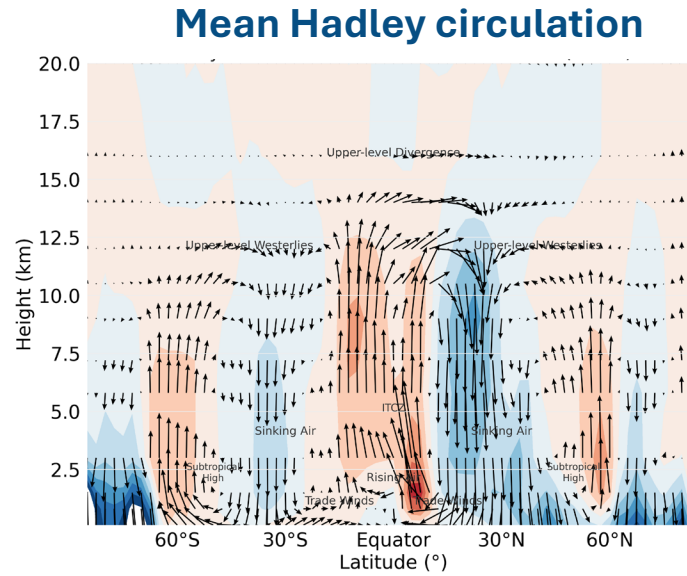
Convection in the tropics and largescale transport



Convection in the tropics and largescale transport



The model currently exhibits a too-weak Hadley circulation



Convection in the tropics and largescale transport



The model currently exhibits a too-weak Hadley circulation

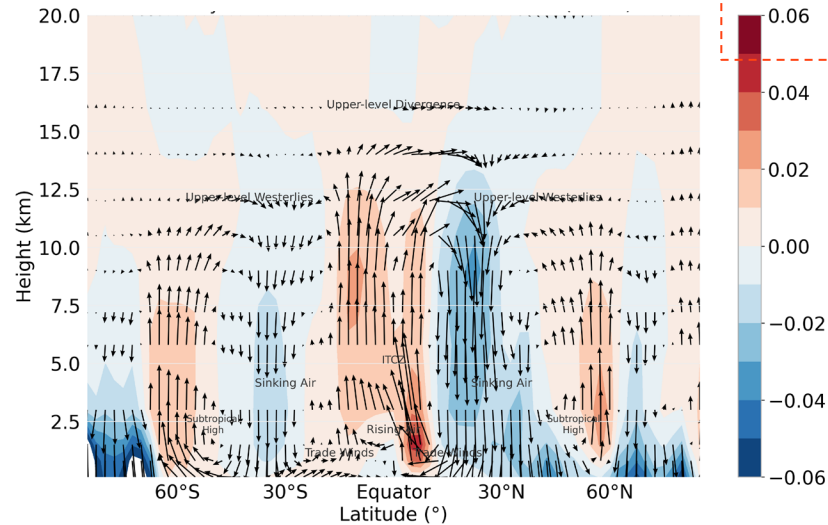
Convergence

Impact of improving convective organization in the tropics

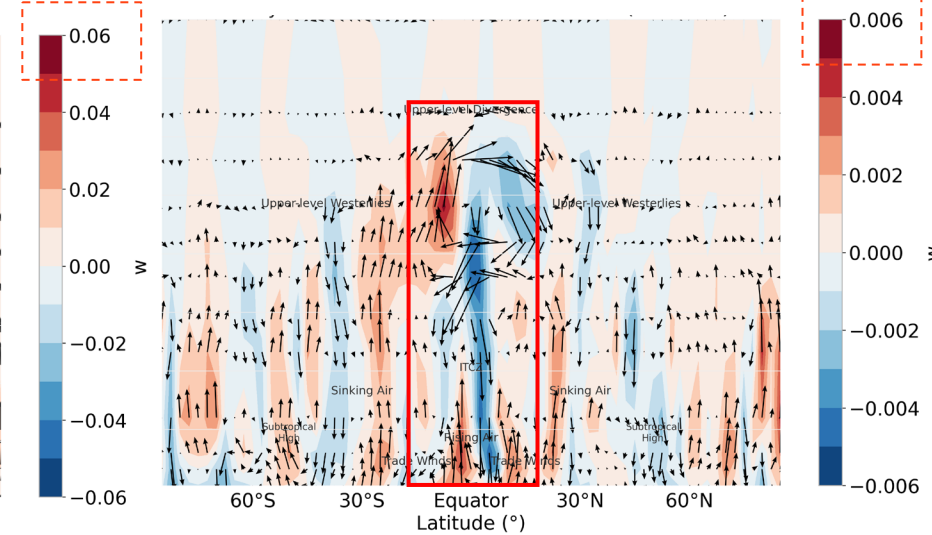


Tobias Becker, Benoit Vanniere

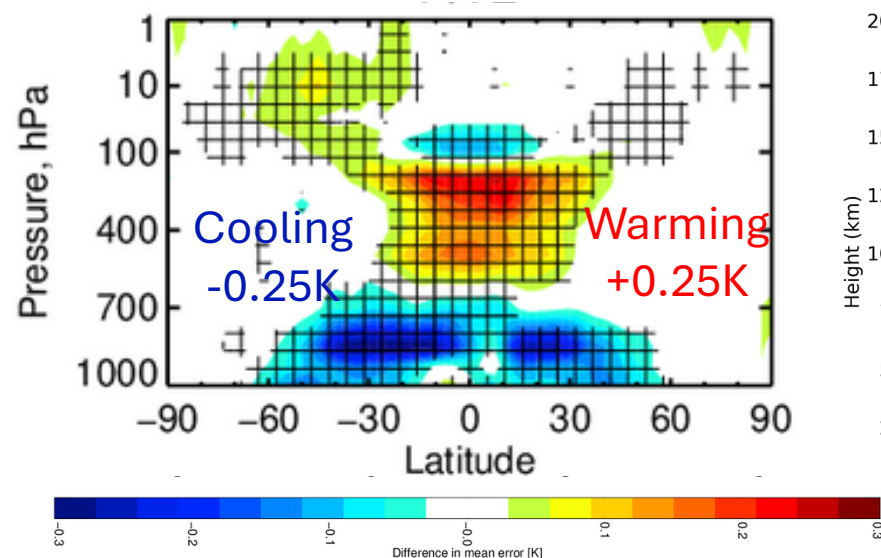
Mean Hadley circulation



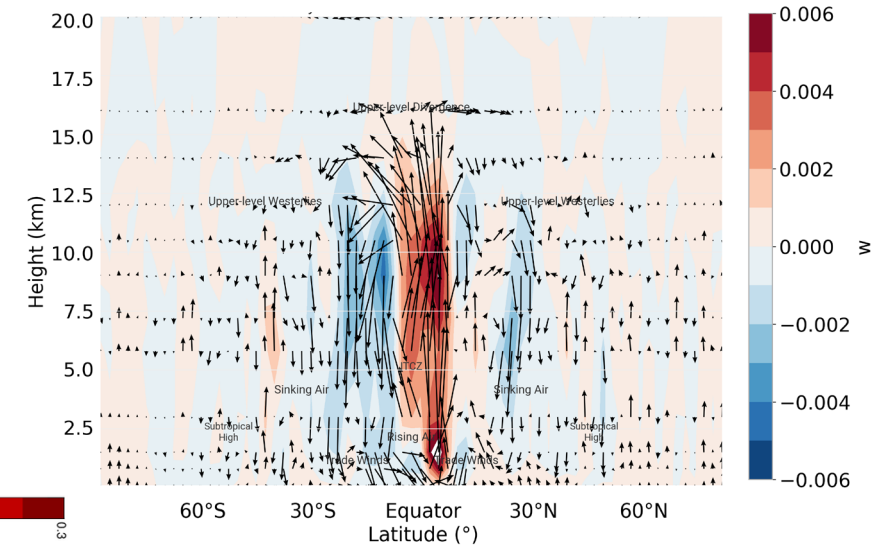
Hadley circulation error (+72h)



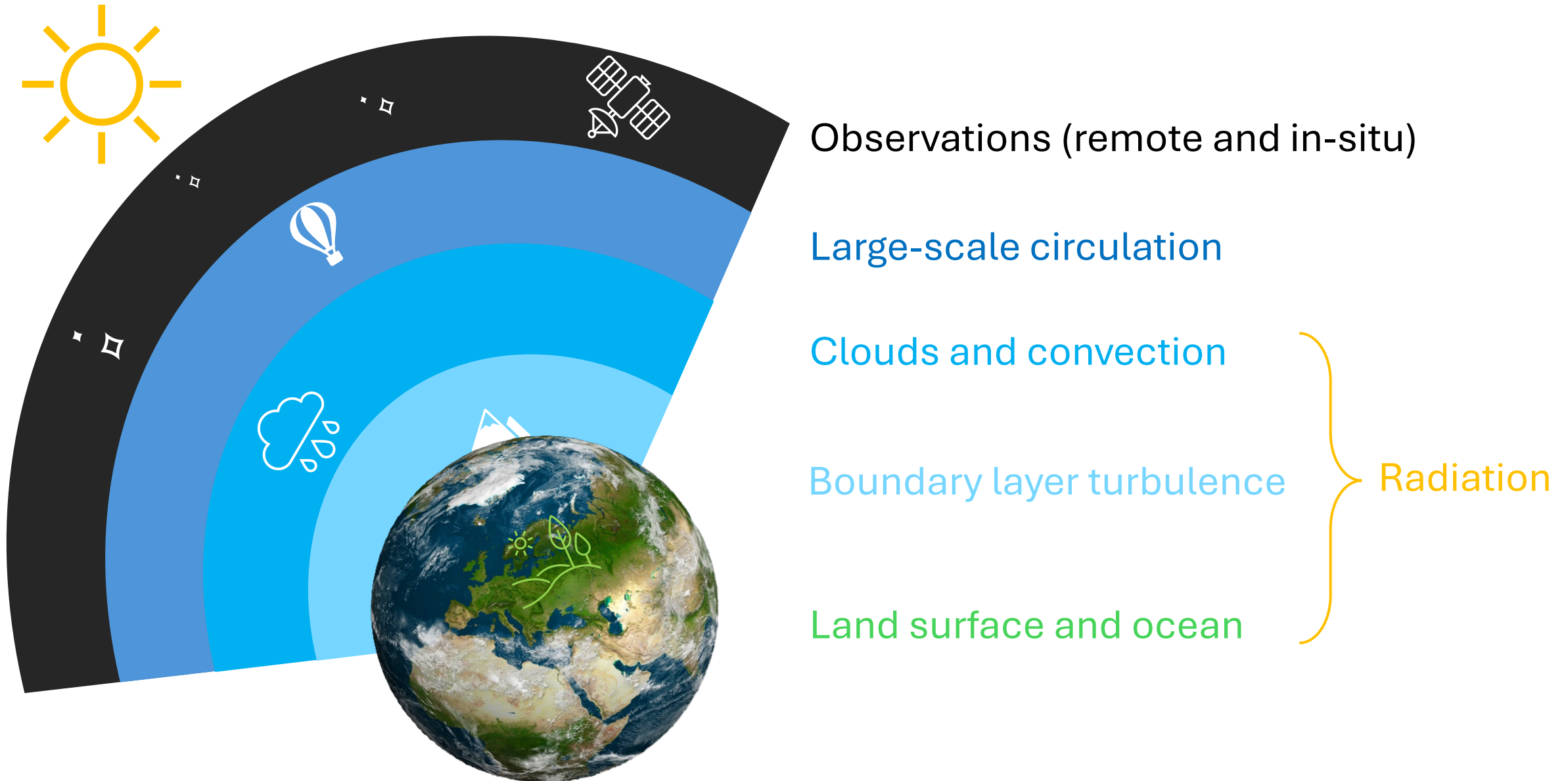
Mean temperature change (+72h)



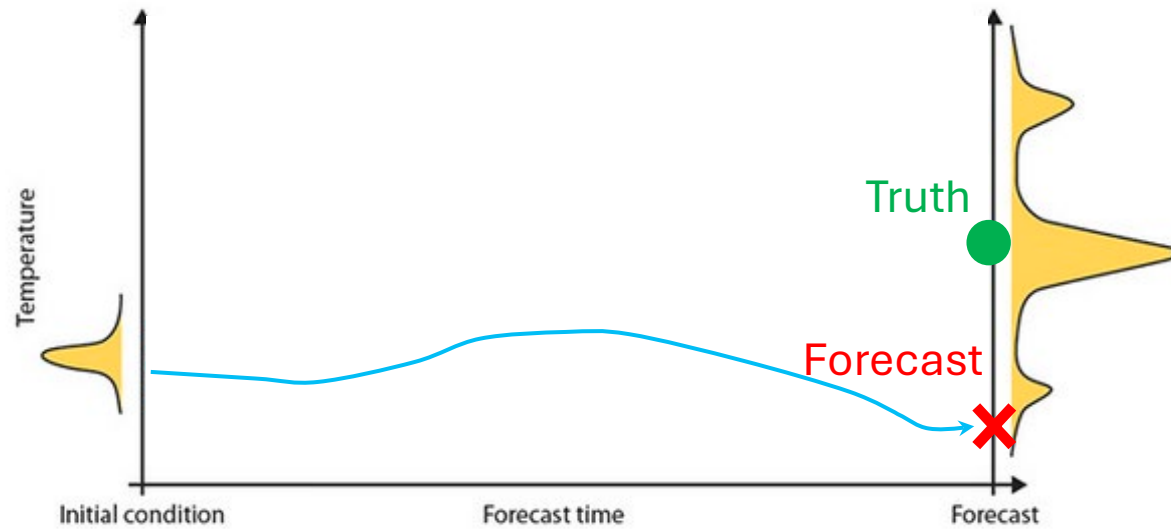
Mean Hadley circulation change (+72h)



Perturbing processes that impact forecast temperature

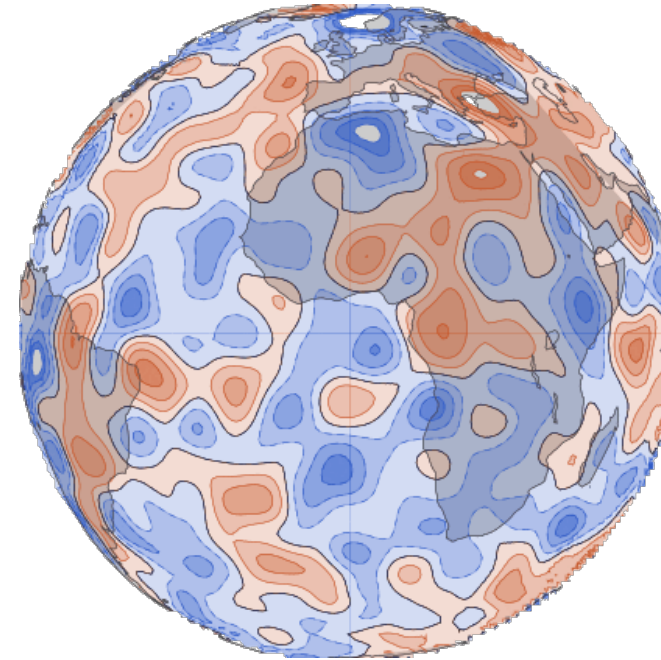
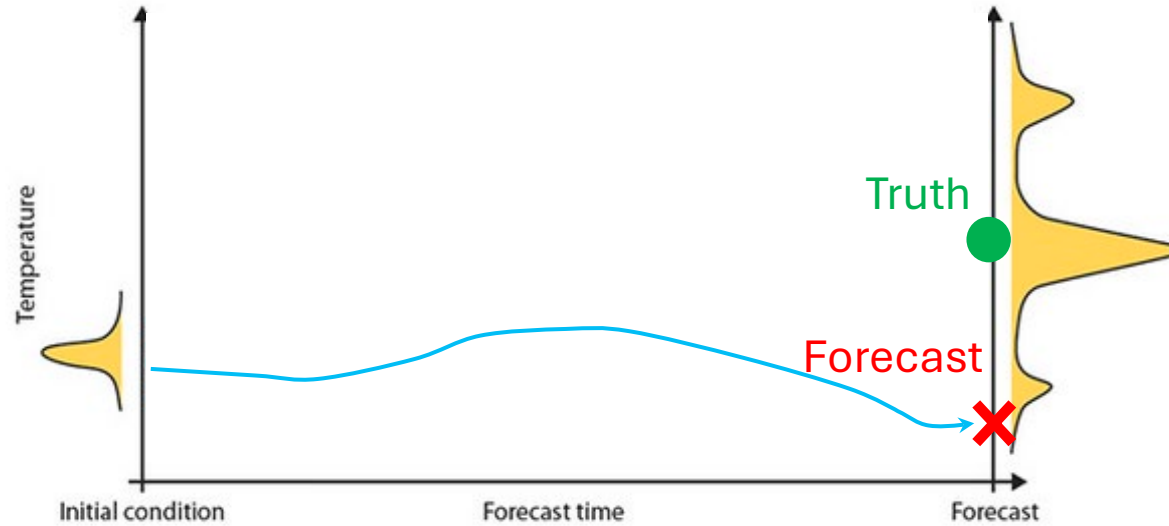


Ensemble forecasting as a means of accounting for process uncertainty

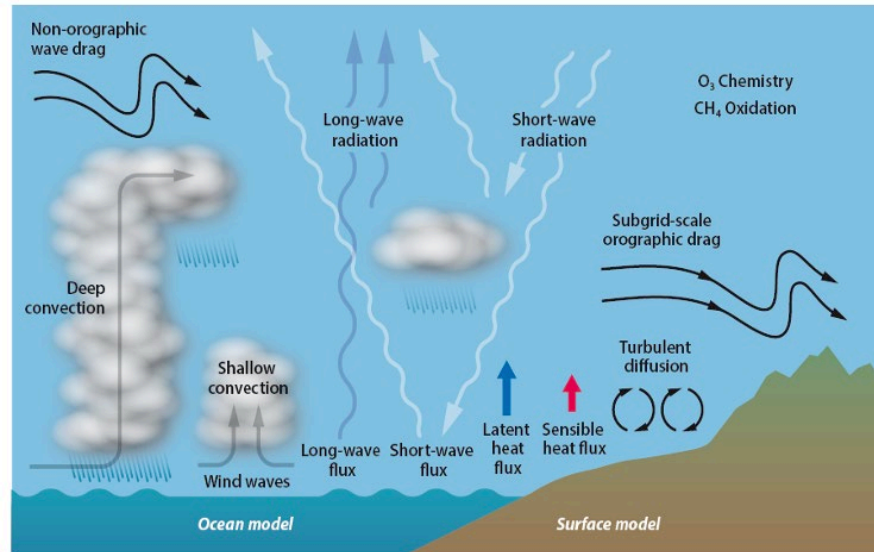


Ensemble forecasting as a means of accounting for process uncertainty

SPP perturbation pattern example



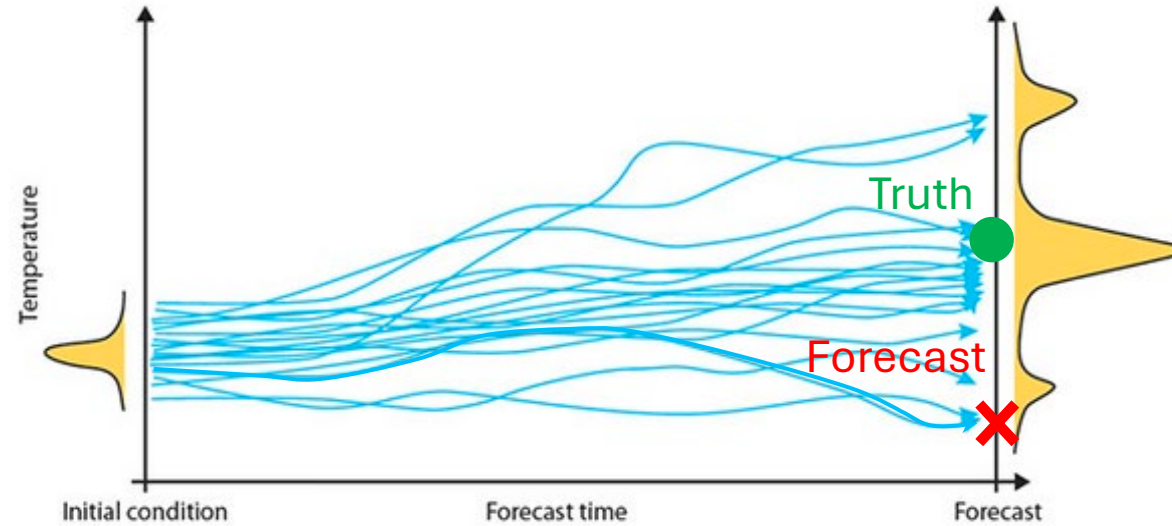
Red:
+0.5 perturbation
Blue:
-0.5 perturbation



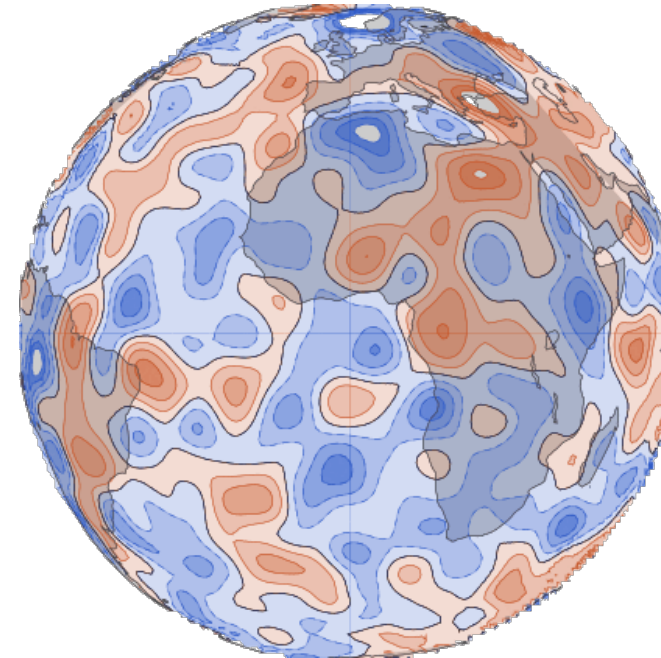
A stochastically perturbed parametrisations scheme (SPP) is used to overcome uncertainties in the sub-grid physical processes (an their interactions)

This allows for a probability of outcomes to be produced...

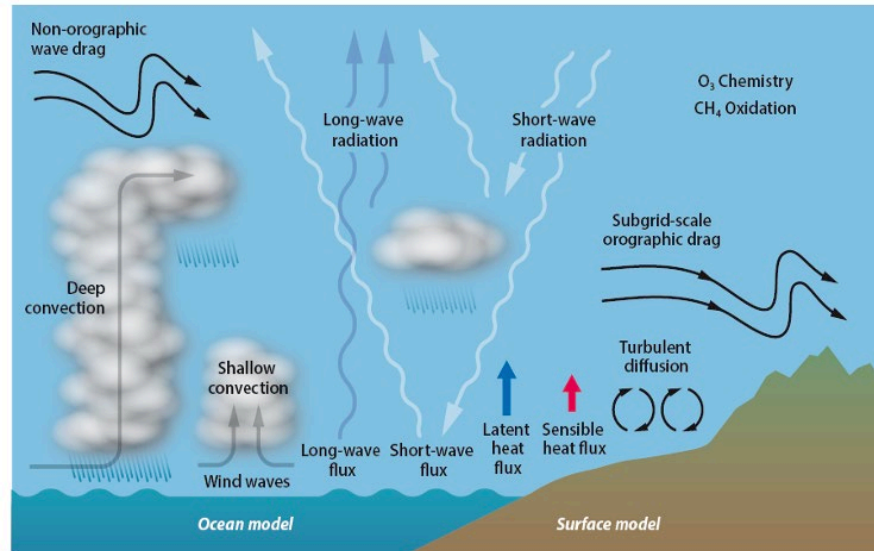
Ensemble forecasting as a means of accounting for process uncertainty



SPP perturbation pattern example



Red:
+0.5 perturbation
Blue:
-0.5 perturbation



A stochastically perturbed parametrisations scheme (SPP) is used to overcome uncertainties in the sub-grid physical processes (an their interactions)

This allows for a probability of outcomes to be produced...

Conclusions

- The ECMWF model represents many physical processes: from local to global scale
- Each process plays a role in the prediction of near surface temperature
- But, processes interact strongly and, in certain cases, a delicate balance between them is required to capture correct temperature
- Ensemble forecasting, namely SPP, helps to account for uncertainty in the processes and their interactions
- Continued understanding of the insufficiencies in the representation of processes is needed for accurate temperature forecasts