



A research institute
of the ETH Domain

Comprehensive Expansion: Advancements in Downscaling Algorithms Beyond Temperature and Precipitation for the CHELSA Dataset

Dirk Nikolaus Karger

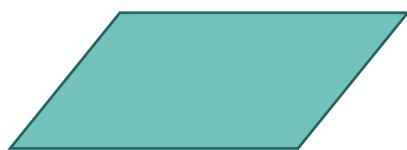
Swiss Federal Institute for Forest, Snow
and Landscape Research WSL

CHELSA – W5E5 variables in ISIMIP

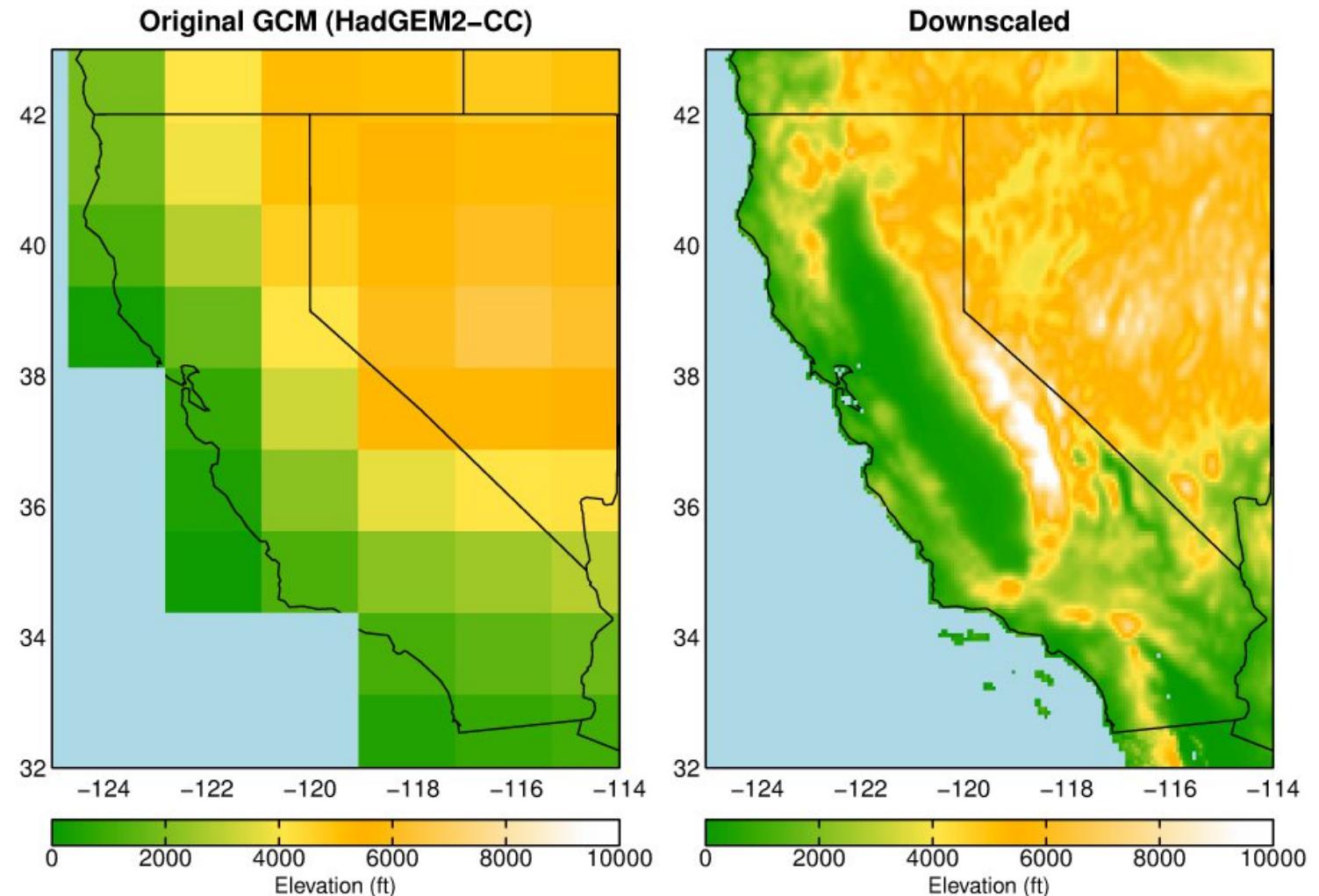
shortname	description	CF Standard Name	levels	frequency
tas	daily-mean near-surface (2 meter) air temperature	air temperature	surface	day
tasmax	daily-maximum near-surface (2 meter) air temperature	air temperature	surface	day
tasmin	daily-minimum near-surface (2 meter) air temperature	air temperature	surface	day
rsds	total downwelling shortwave solar radiation	surface downwelling shortwave flux in air	surface	day
pr	daily precipitation flux (liquid and solid)	precipitation flux	surface	day

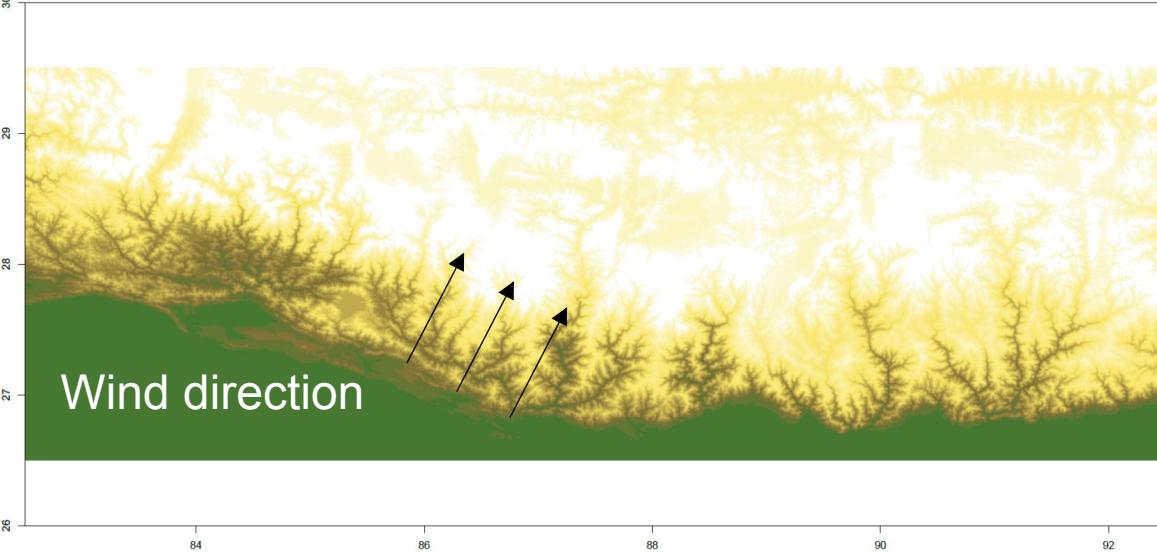
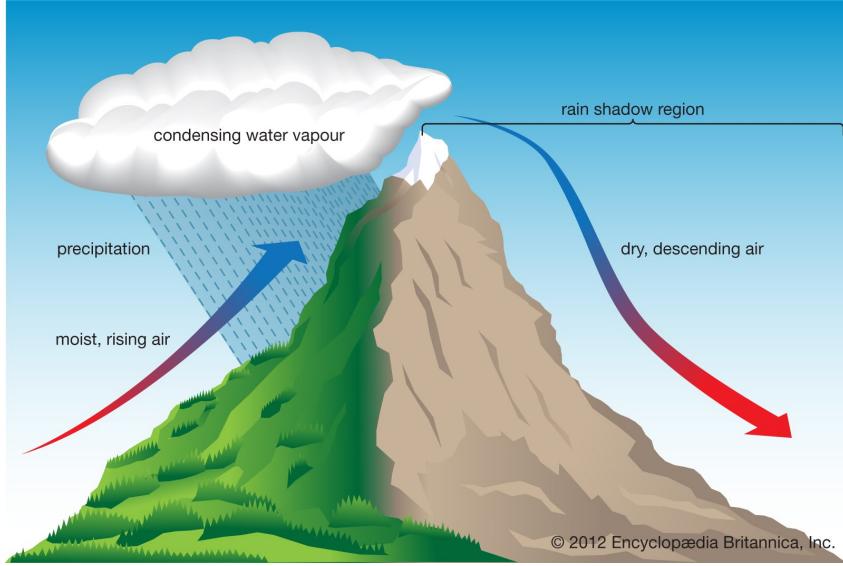
CHELSA uses gridded coarse resolution data as input

Coarse resolution



downscaling



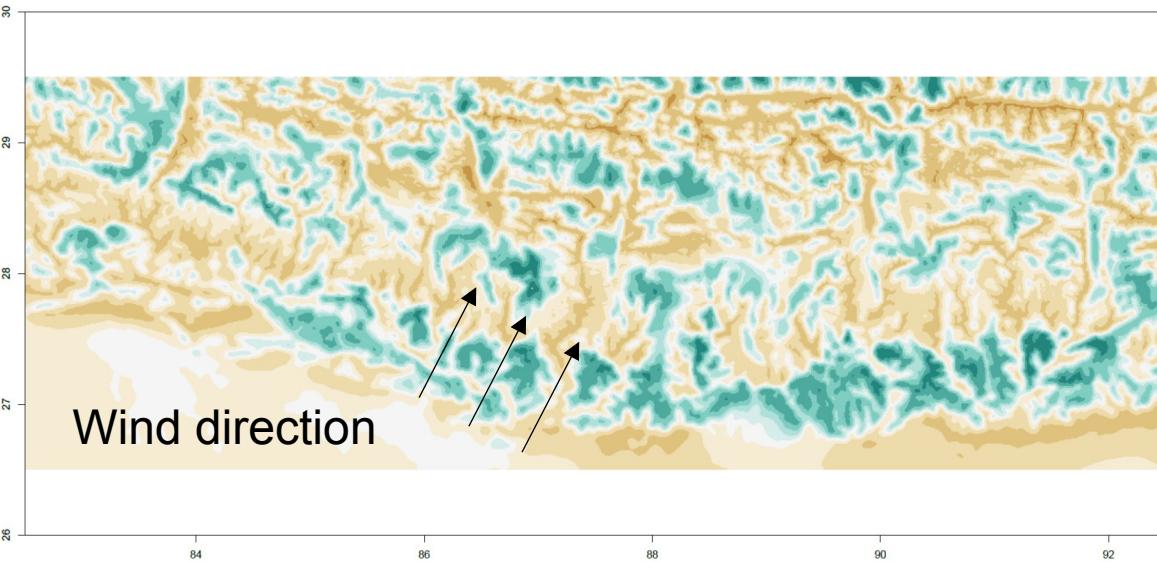


Wind direction and condensation levels
are taken into account

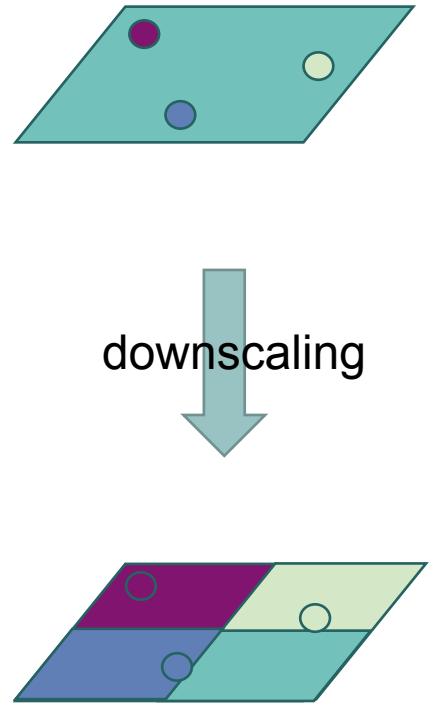
$$H_W = \frac{\sum_{i=1}^n \frac{1}{d_{WHi}} \tan^{-1}(\frac{d_{WZi}}{d_{WHi}})}{\sum_{i=1}^n \frac{1}{d_{LHi}}} + \frac{\sum_{i=1}^n \frac{1}{d_{LHi}} \tan^{-1}(\frac{d_{LZi}}{d_{LHi}})}{\sum_{i=1}^n \frac{1}{d_{LHi}}}$$

$$H_L = \frac{\sum_{i=1}^n \frac{1}{d_{WHi}} \tan^{-1}(\frac{d_{LZi}}{d_{WHi}})}{\sum_{i=1}^n \frac{1}{d_{LHi}}}$$

$$pr_h = S(pr_c) * \frac{\frac{1}{n} \sum_{i=1}^n H_n}{\frac{1}{n} \sum_{i=1}^n \sum_{i=1}^n H_n}$$



Principle of the precipitation downscaling in CHELSA



orography

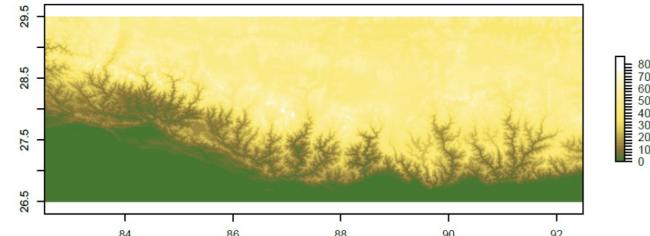
$S(pr_c)$

B-spline

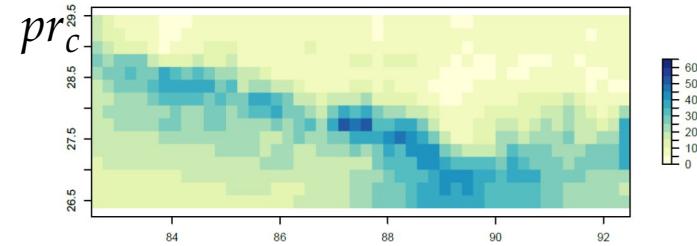
$$H_W = \frac{\sum_{i=1}^n \frac{1}{d_{WHi}} \tan^{-1} \left(\frac{d_{WZi}}{d_{WHi}} \right)}{\sum_{i=1}^n \frac{1}{d_{LHi}}} + \frac{\sum_{i=1}^n \frac{1}{d_{LHi}} \tan^{-1} \left(\frac{d_{LZi}}{d_{LHi}} \right)}{\sum_{i=1}^n \frac{1}{d_{LHi}}}$$

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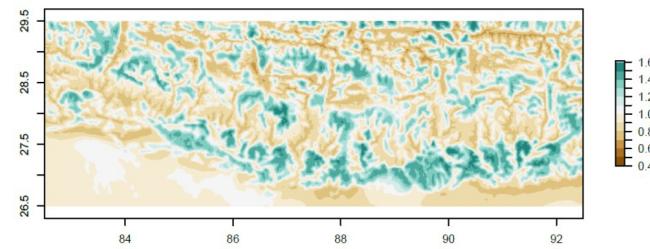
$$pr_h = S(pr_c) * \frac{\frac{1}{n} \sum_{i=1}^n H_n}{\frac{1}{n} \sum_{i=1}^n \sum_{i=1}^n H_n}$$



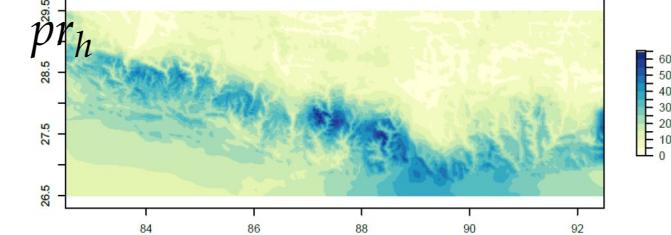
Coarse resolution precipitation



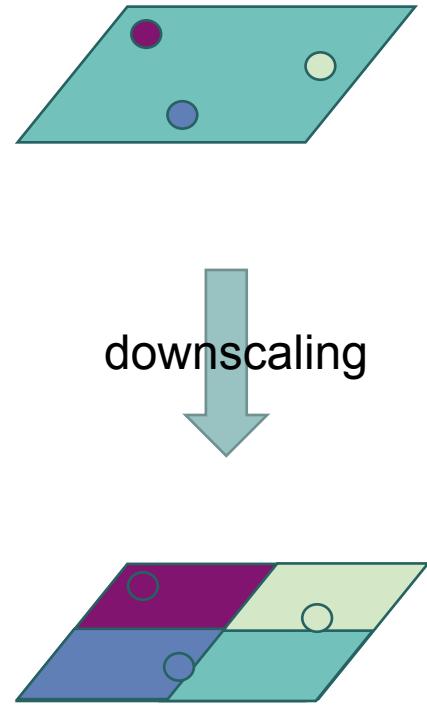
Windward leeward index H



High resolution precipitation



Principle of the precipitation downscaling in CHELSA



orography

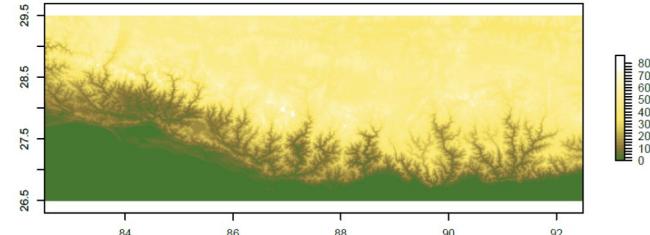
$S(pr_c)$

B-spline

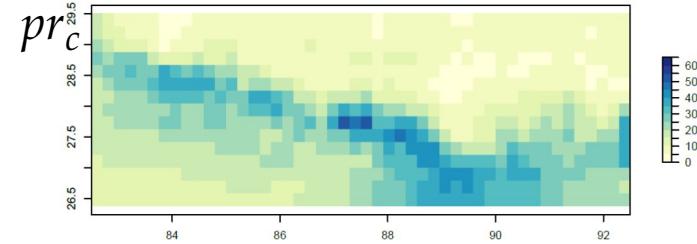
$$H_W = \frac{\sum_{i=1}^n \frac{1}{d_{WHi}} \tan^{-1} \left(\frac{d_{WZi}}{d_{WHi}} \right)}{\sum_{i=1}^n \frac{1}{d_{LHi}}} + \frac{\sum_{i=1}^n \frac{1}{d_{LHi}} \tan^{-1} \left(\frac{d_{LZi}}{d_{LHi}} \right)}{\sum_{i=1}^n \frac{1}{d_{LHi}}}$$

$$H_L = \frac{\sum_{i=1}^n \frac{1}{d_{WHi}} \tan^{-1} \left(\frac{d_{LZi}}{d_{WHi}} \right)}{\sum_{i=1}^n \frac{1}{d_{LHi}}}$$

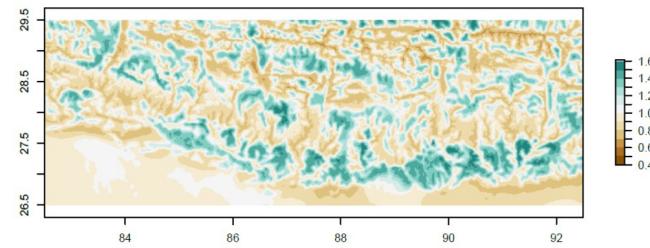
Physical constraint
(mass preservation)



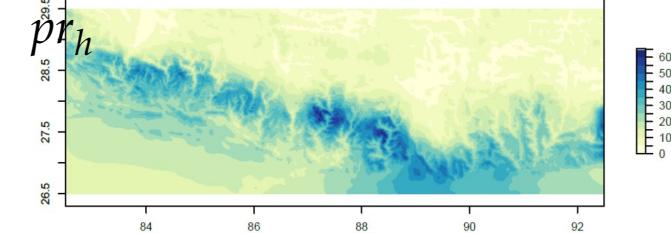
Coarse resolution precipitation



Windward leeward index H



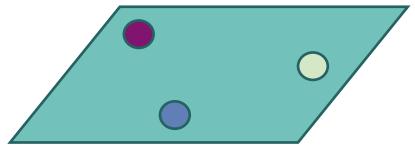
High resolution precipitation



Absolute Bias reduction from coarse to high resolution



Coarse resolution



downscaling

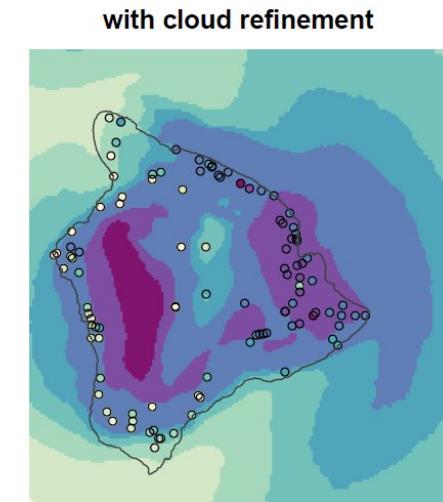
High resolution



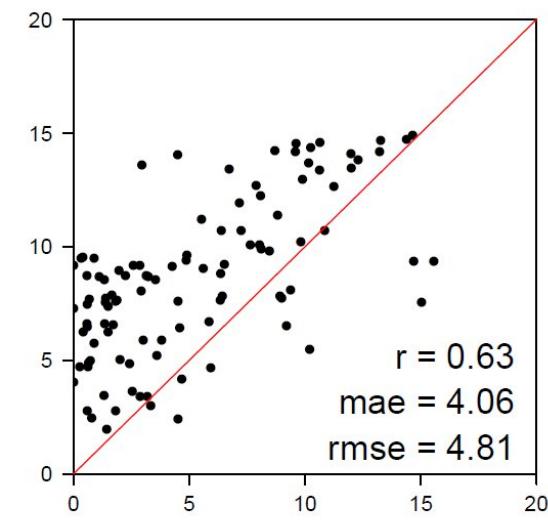
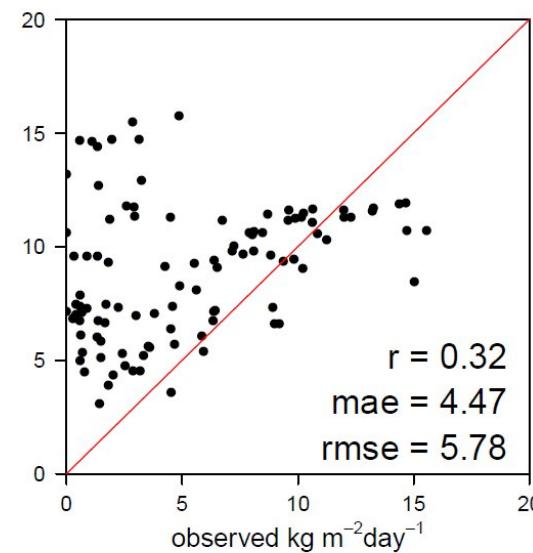
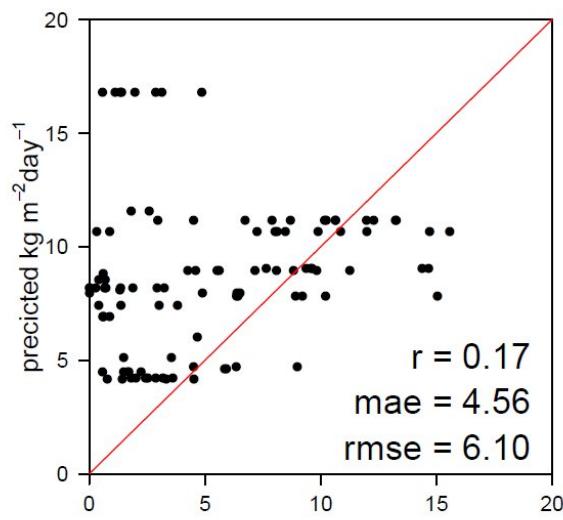
without downscaling



without cloud refinement

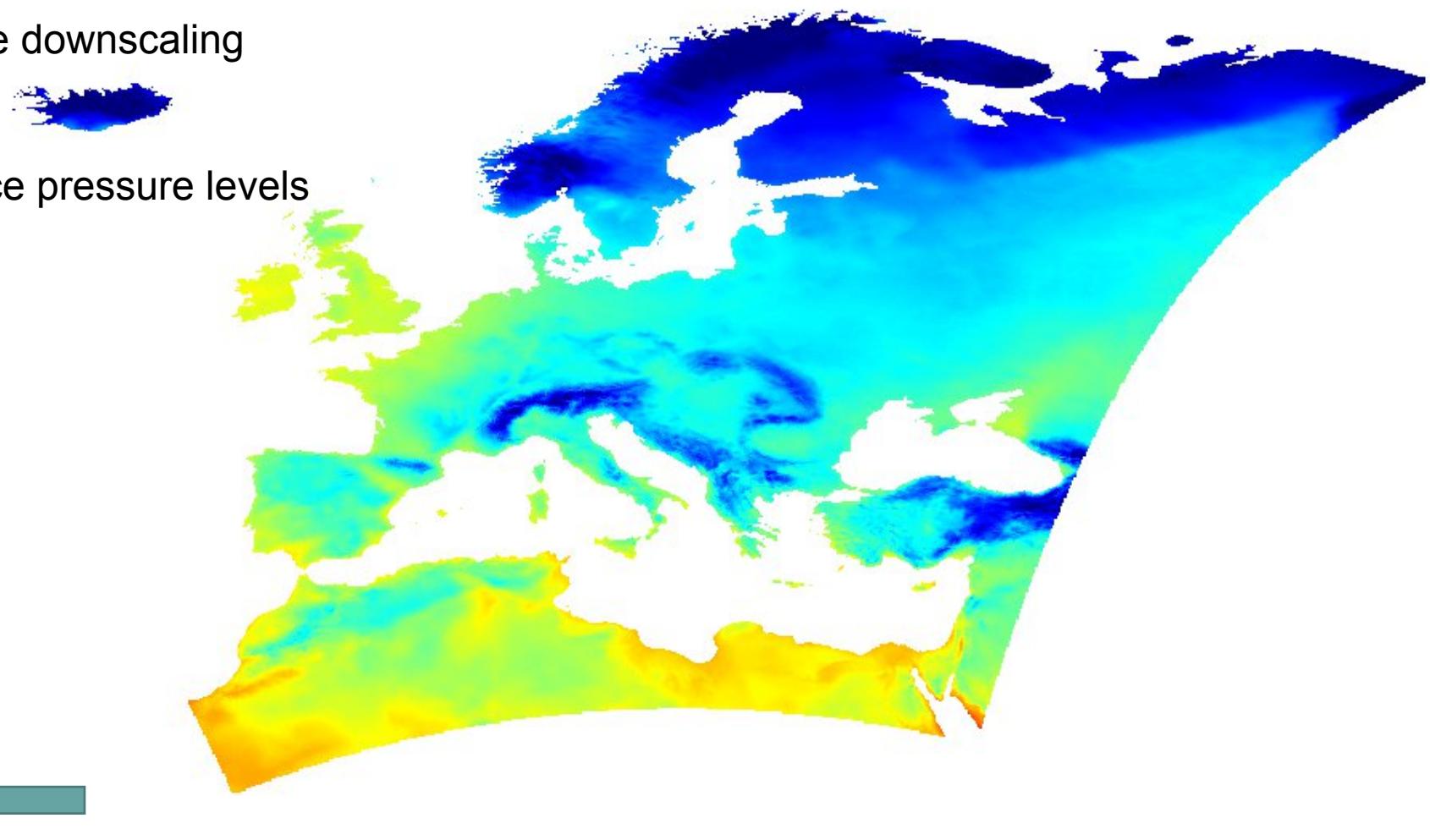


with cloud refinement

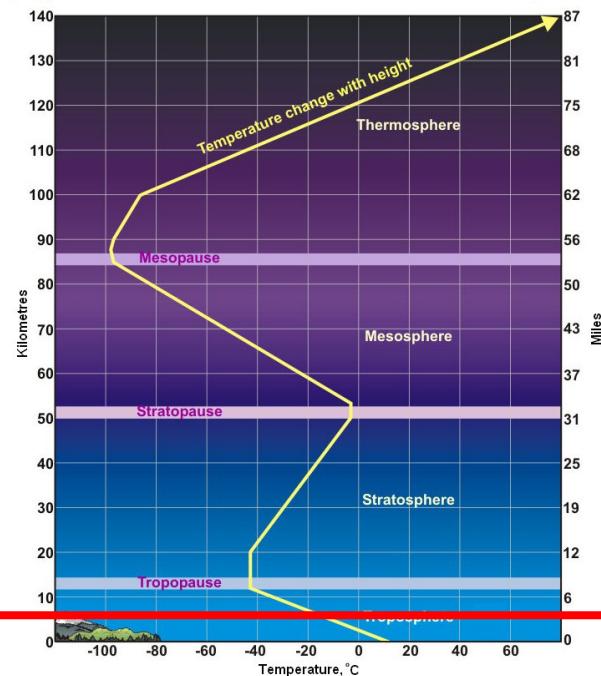


Temperature

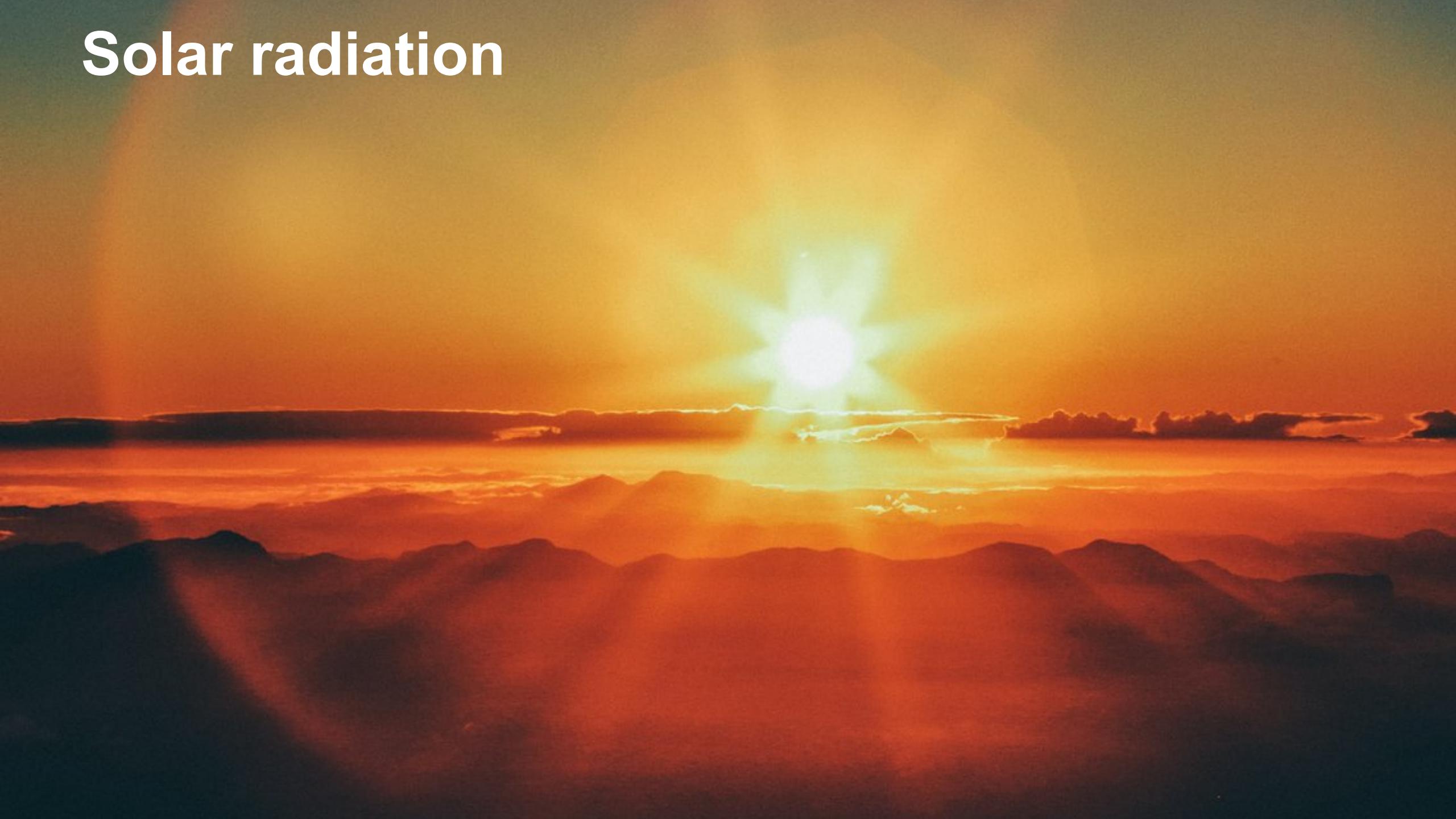
Lapse rate based temperature downscaling



Using two surface near surface pressure levels

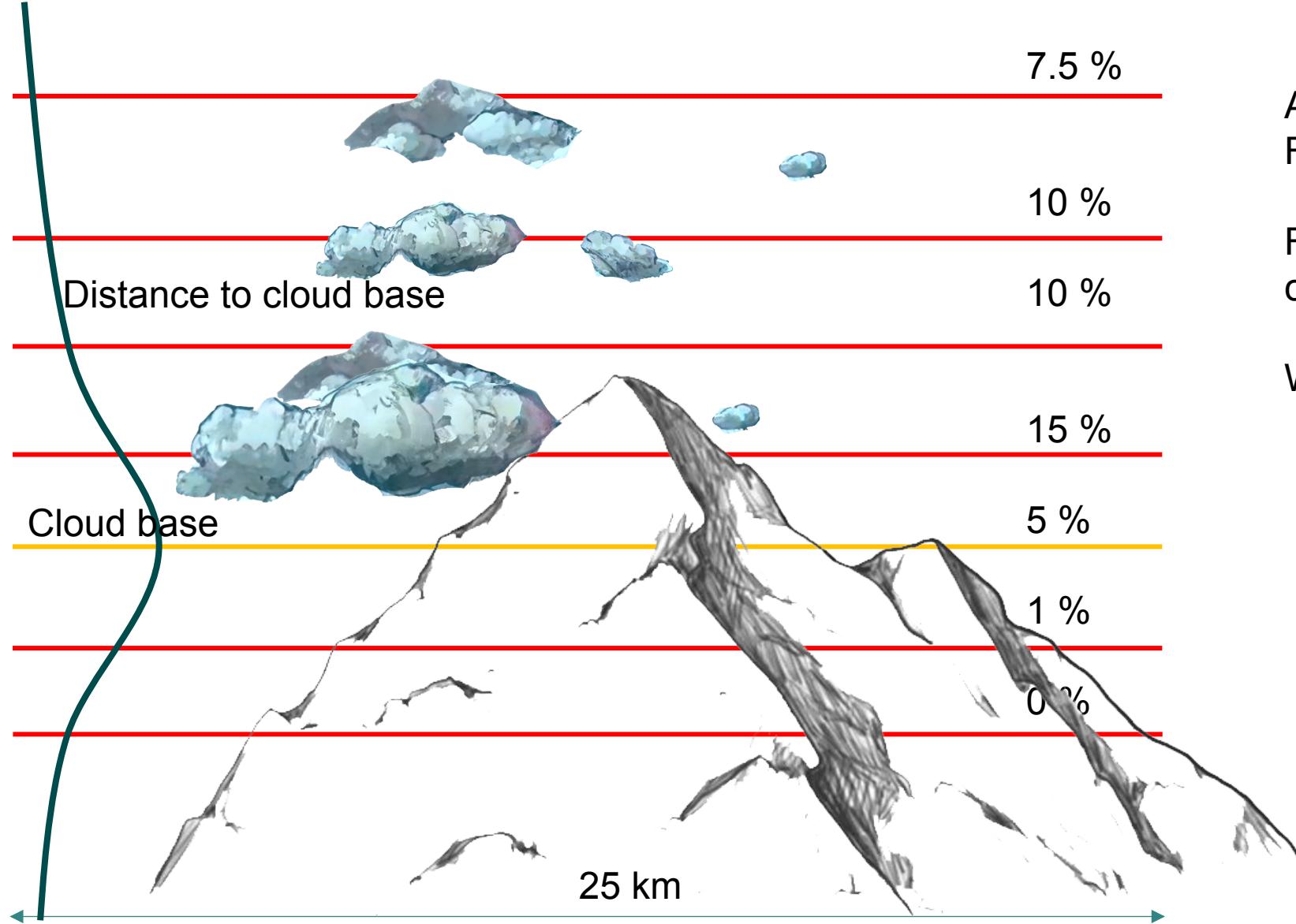


Solar radiation





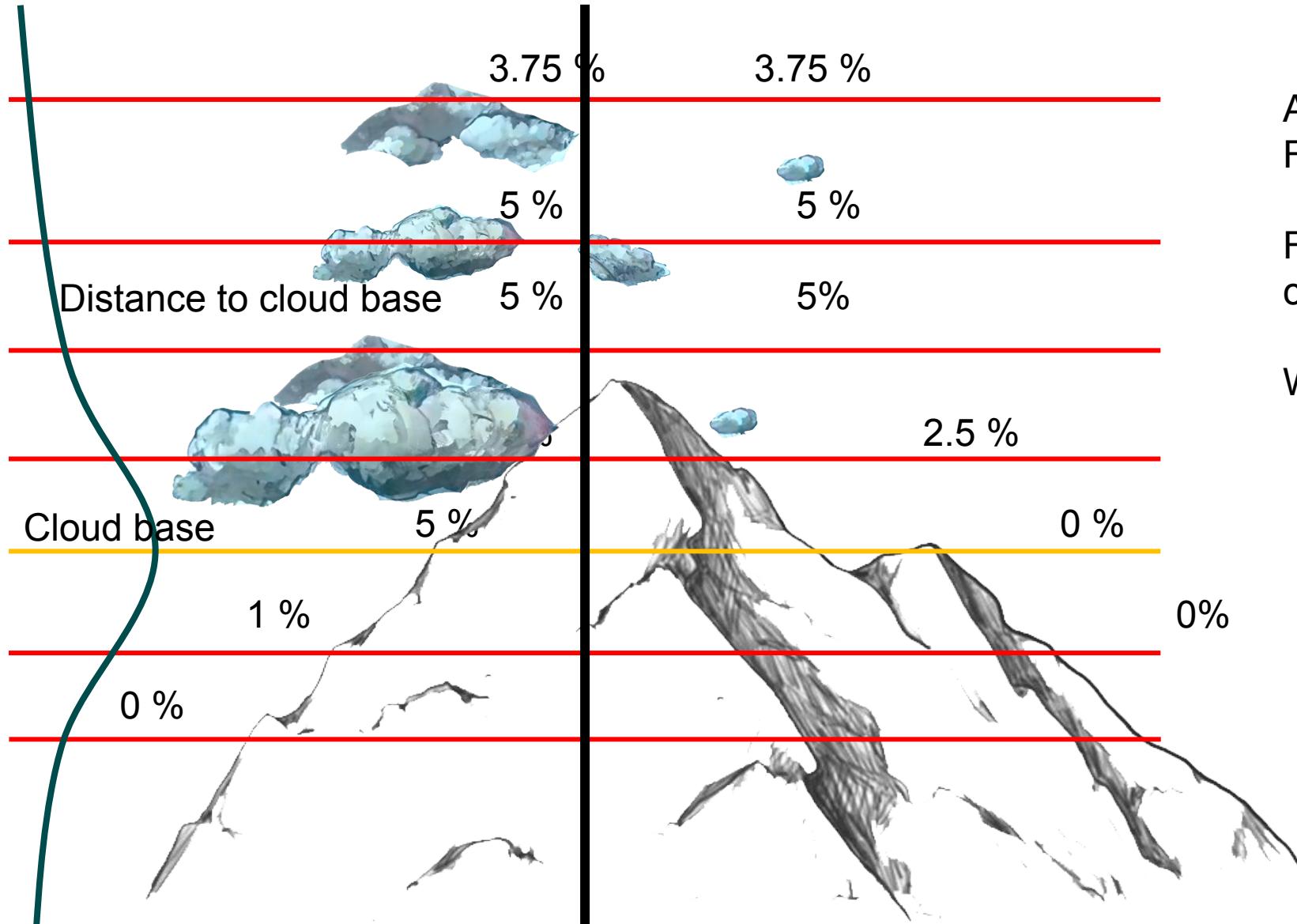
Orographic clouds



Algorithm to downscale cloud cover
From atmospheric cloud cover fractions

Function describing distance from
cloud base

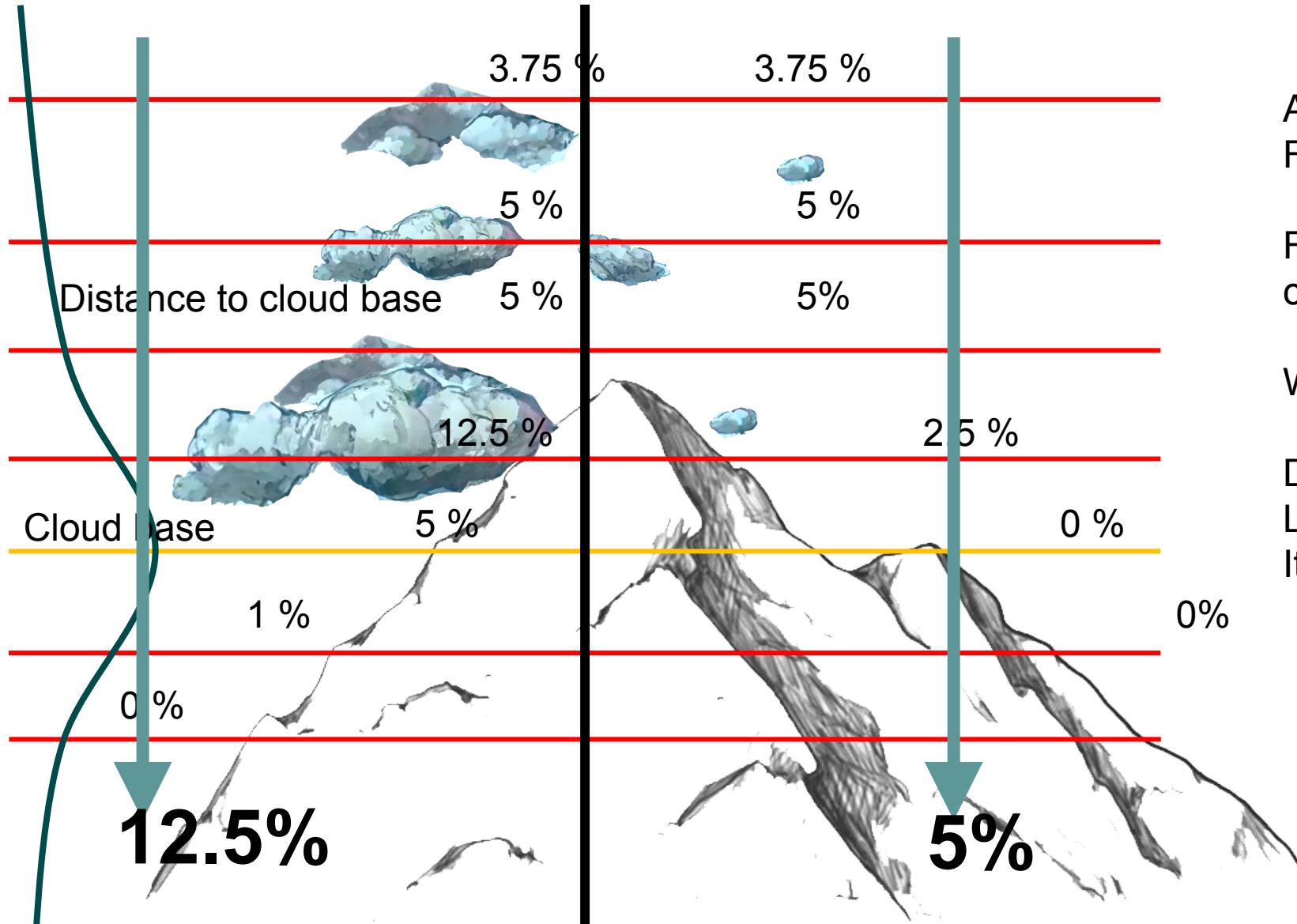
Windward leeward index



Algorithm to downscale cloud cover
From atmospheric cloud cover fractions

Function describing distance from
cloud base

Windward leeward index



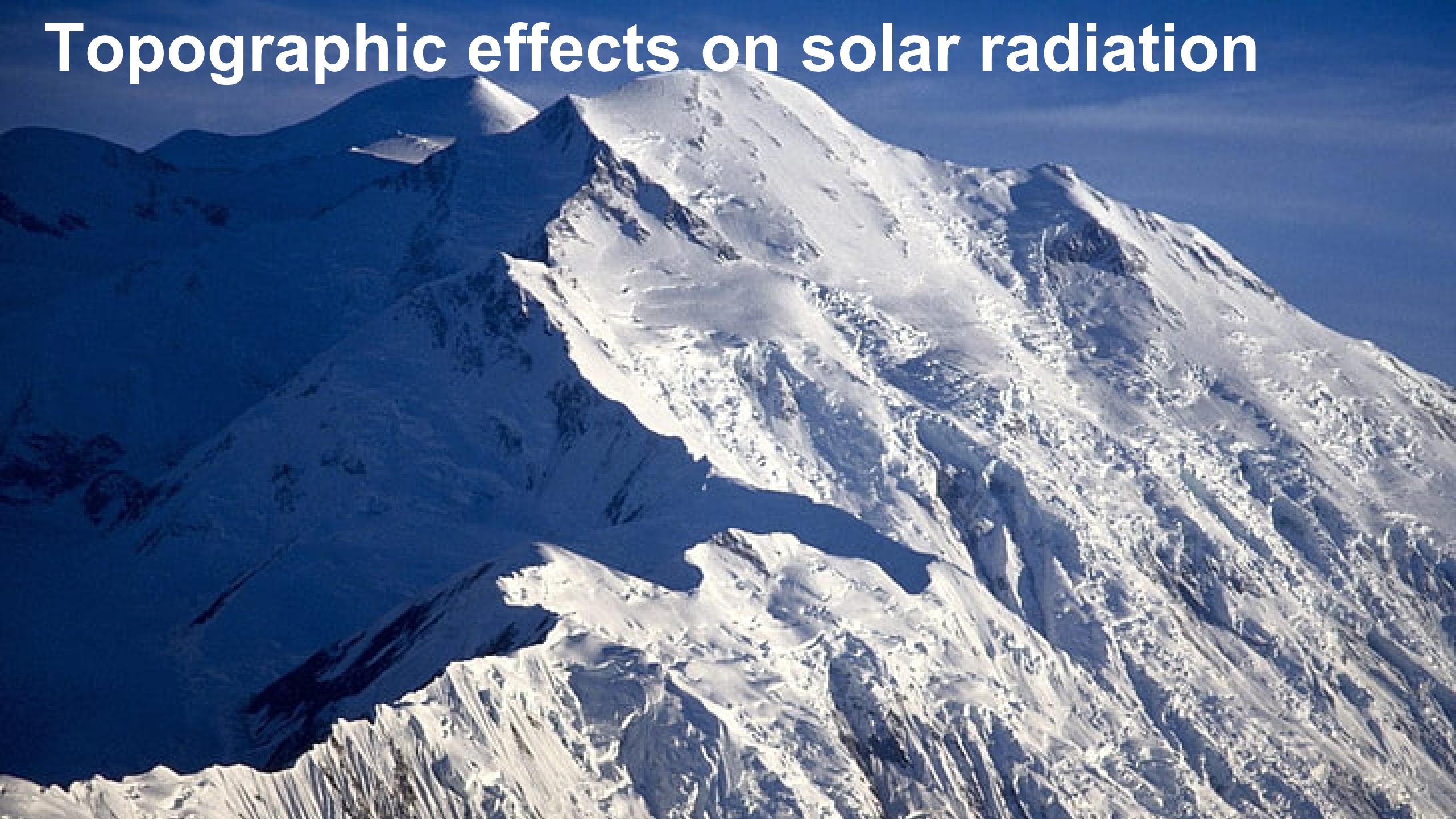
Algorithm to downscale cloud cover
From atmospheric cloud cover fractions

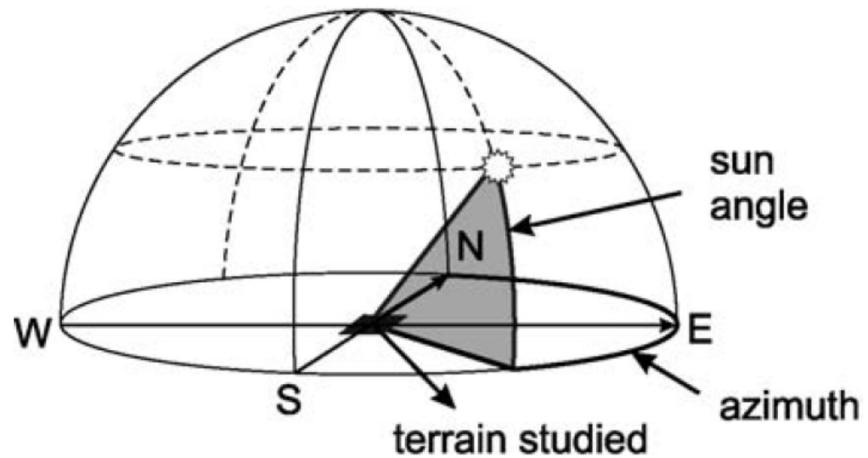
Function describing distance from
cloud base

Windward leeward index

Distribute cloud cover based on windward
Leeward position of a pressure level and
Its distance to cloud base height

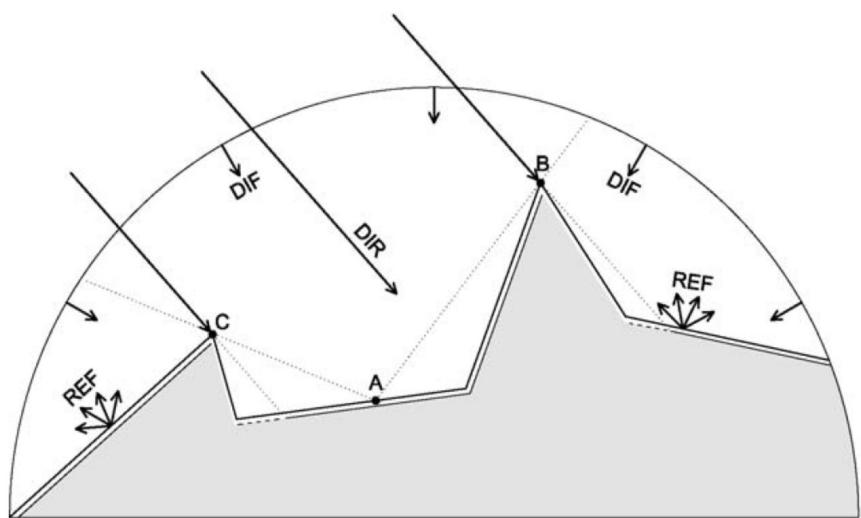
Topographic effects on solar radiation





Solar radiation based on terrain exposition and sun angle

Böhner, J., & Antonic, O. (2009). Land-Surface Parameters Specific to Topo-Climatology. In T. Hengl, & H. I. Reuter (Eds.), GEOMORPHOMETRY: CONCEPTS, SOFTWARE, APPLICATIONS (pp. 195-226). Elsevier Science., in: in T. Hengl, & H. I. Reuter (eds.) Geomorphometry: Concepts, Software, Applications, Elsevier Science, 195–226, 2009.



- Clear sky conditions
- Direct and diffuse radiation

Solar radiation based on Terrain exposition and sun angle

First approximation

Combine with cloud cover
To get total incoming
solar radiation at surface

$$R = R_0(1 - 0.75n^{3.4})$$

where:

R = solar radiation
 n = cloud cover (0.0 - 1.0)

and

$$R_0 = 990 \sin \phi - 30$$

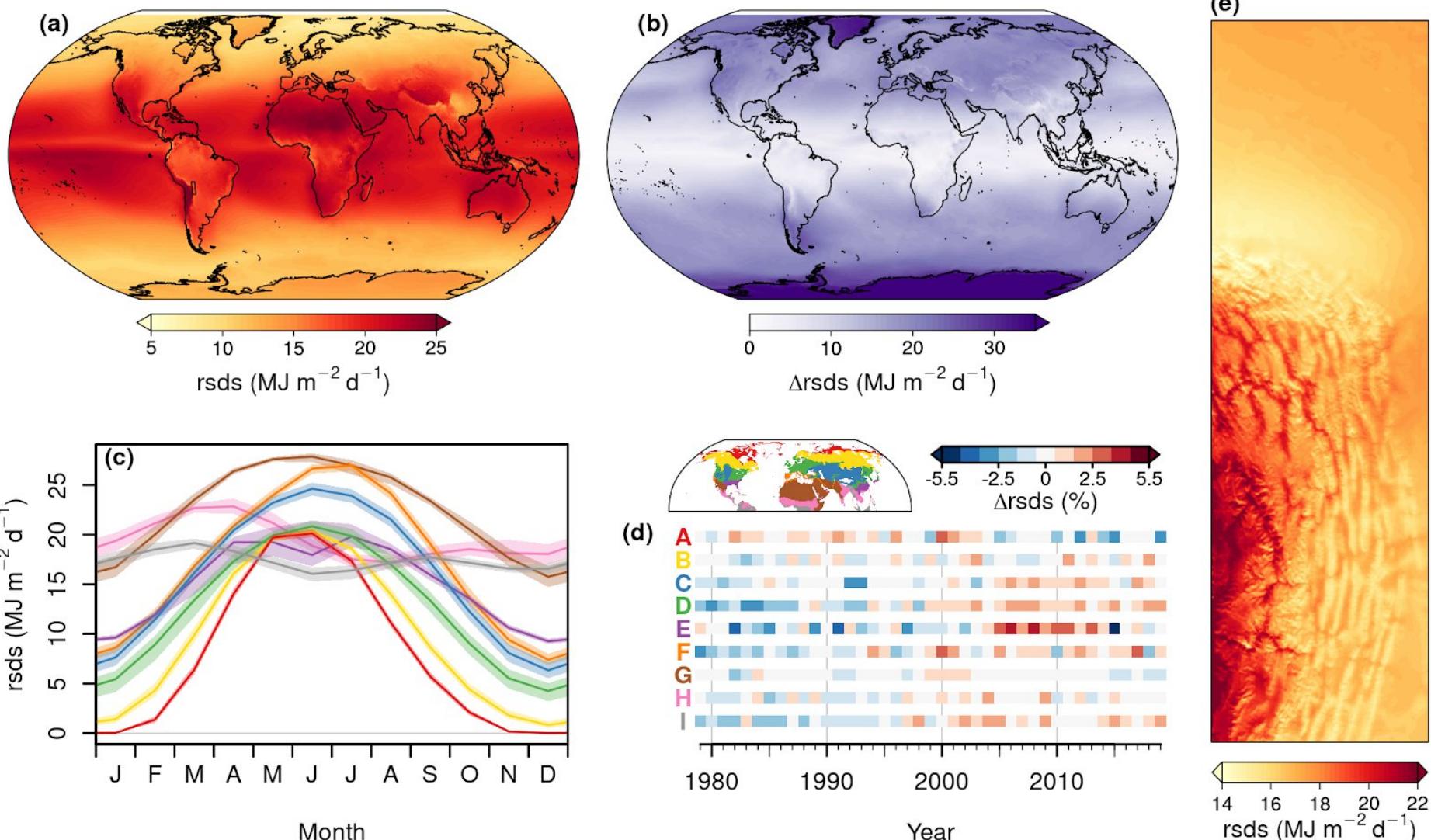
where:

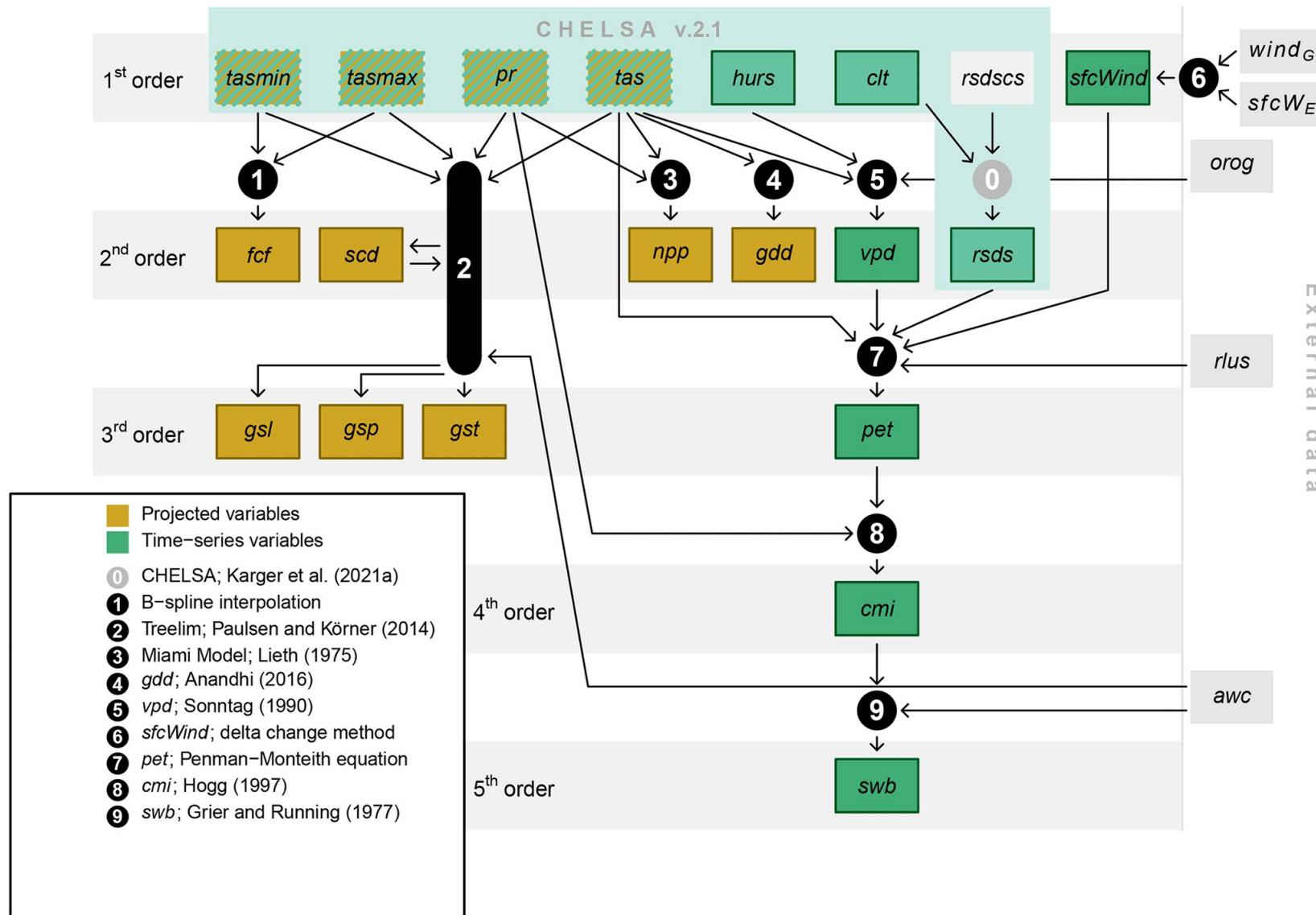
$$\phi = \frac{\phi_{tp} + \phi_p}{2}$$

ϕ_{tp} = previous hour solar elevation angle

ϕ_p = current hour solar elevation angle

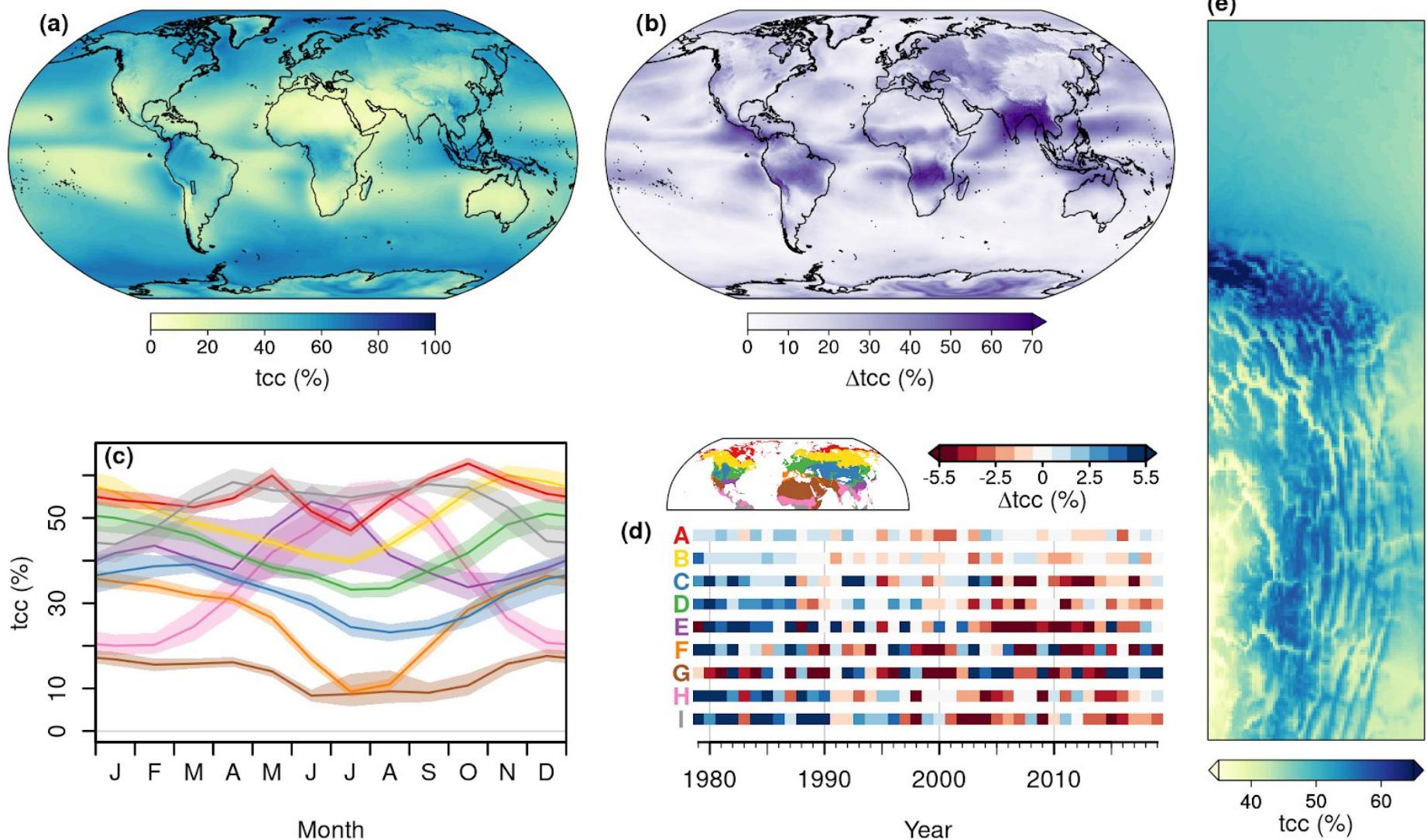
R_0 = clear sky insolation ($\text{W} \cdot \text{m}^{-2}$)





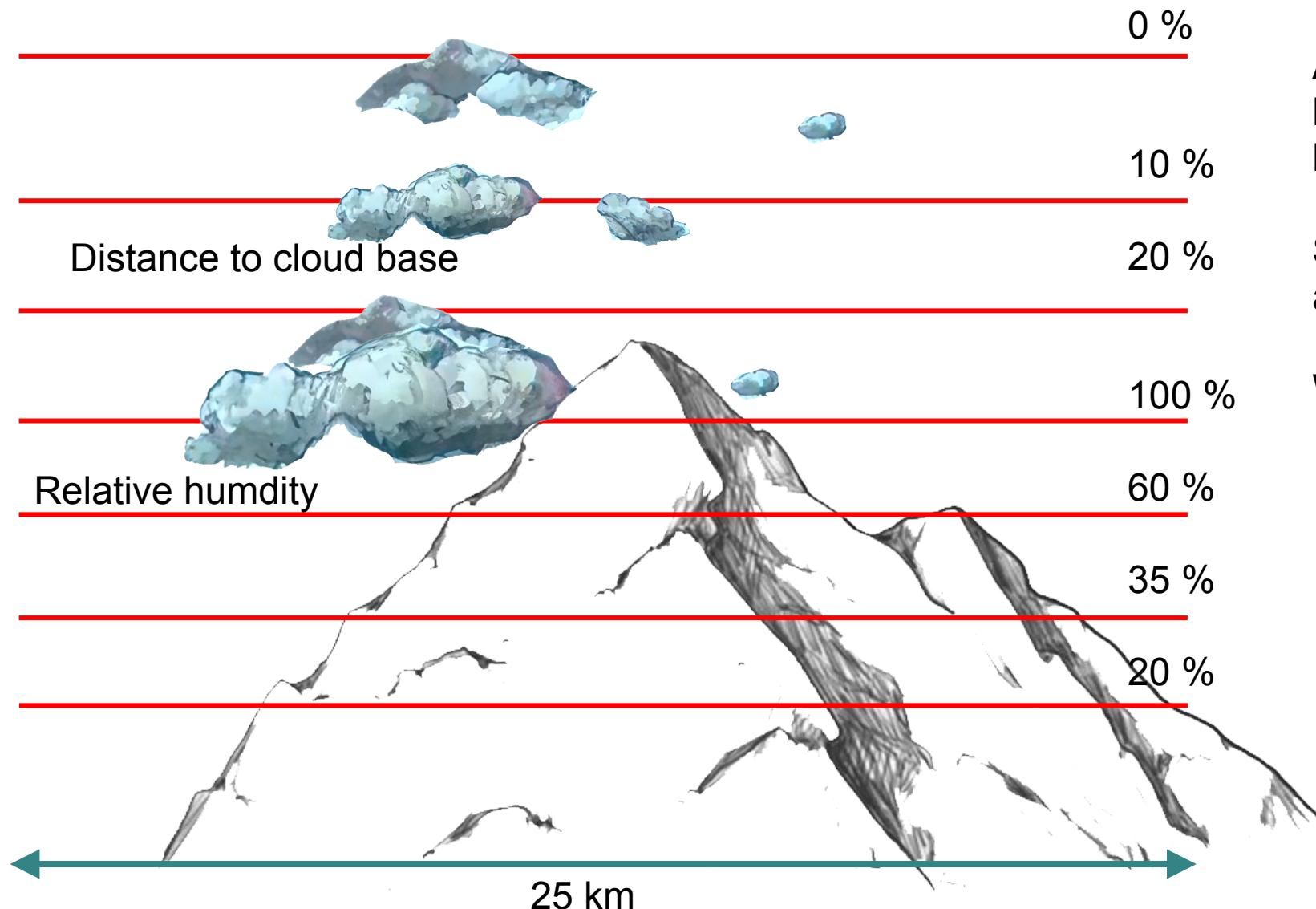
Philipp Brun

Cloud cover based on maximal overlap assumption



Problems with cloud cover

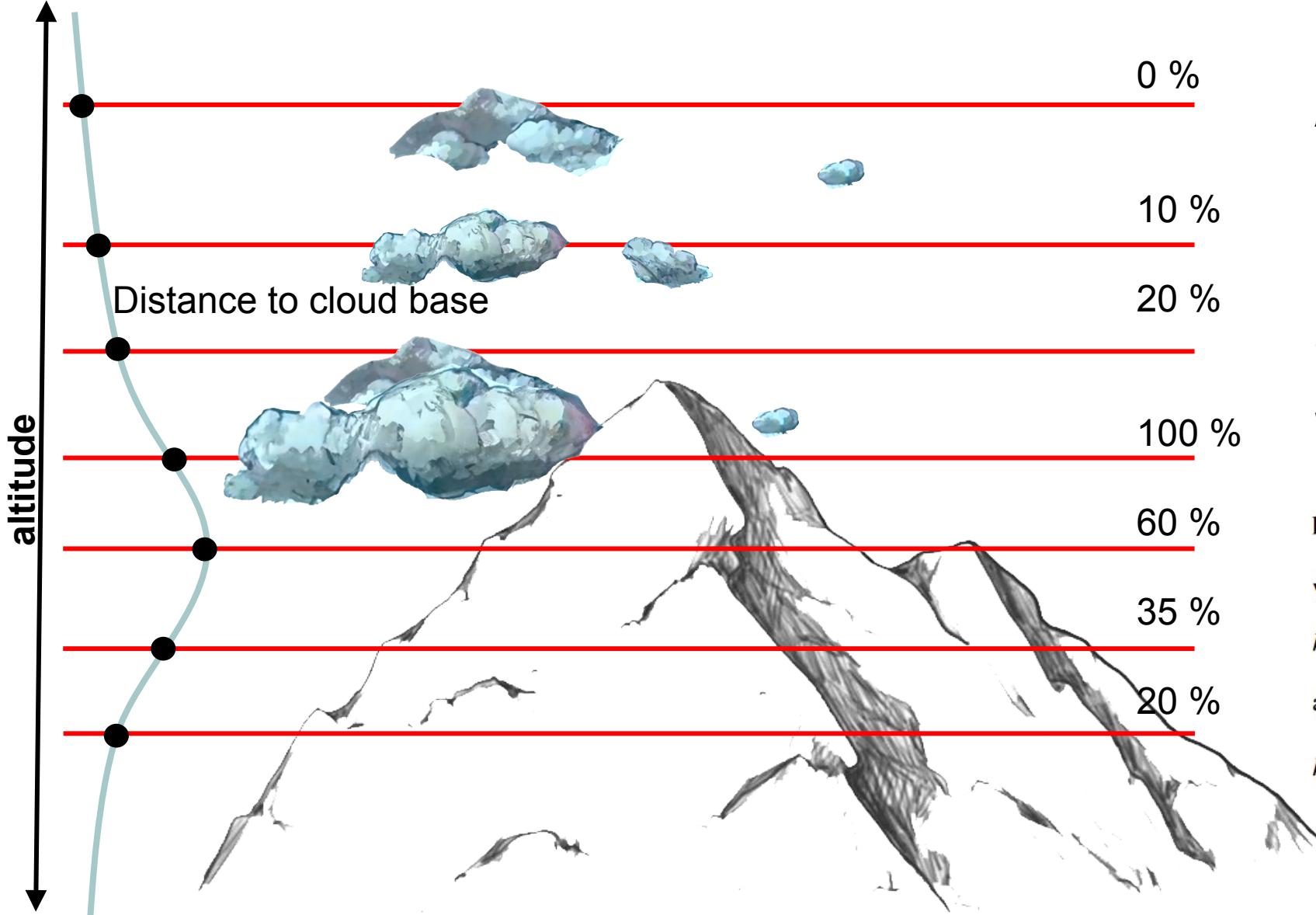
- No convection included
- Still huge biases in the forcing data (e.g. ERA5) compared to observations -> maybe a bias correction of the ERA5 data is needed



Algorithm to downscale relative humidity (*hurs*) from atmospheric humidity (*hur*) on pressure levels

Spline interpolation between *hur* and altitude

Windward leeward index



Algorithm to downscale relative humidity (*hurs*) from atmospheric humidity (*hur*) on pressure levels

Spline interpolation between *hur* and altitude

Windward leeward index

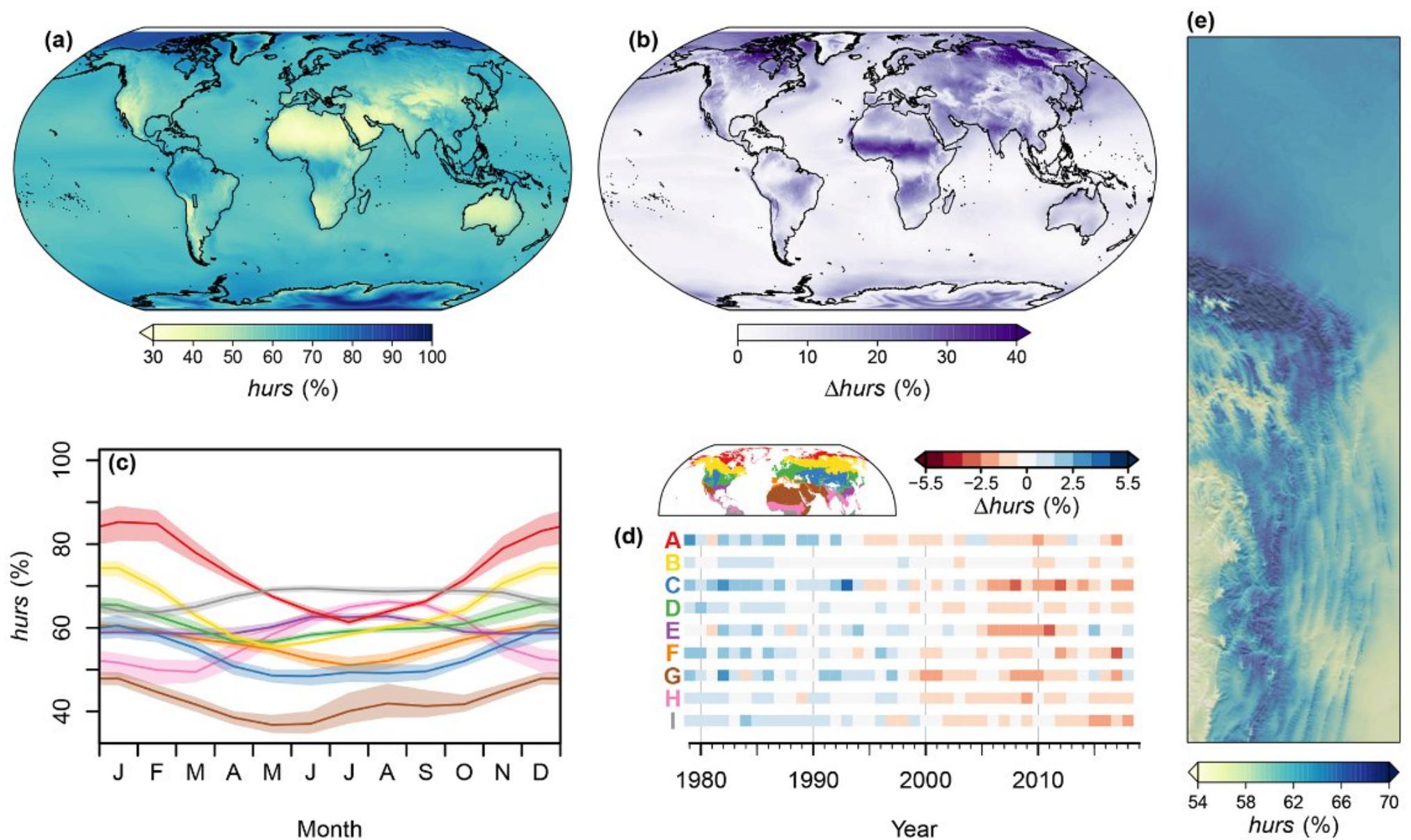
$$hurs = \frac{1}{(1 + \exp(-1 \cdot h))},$$

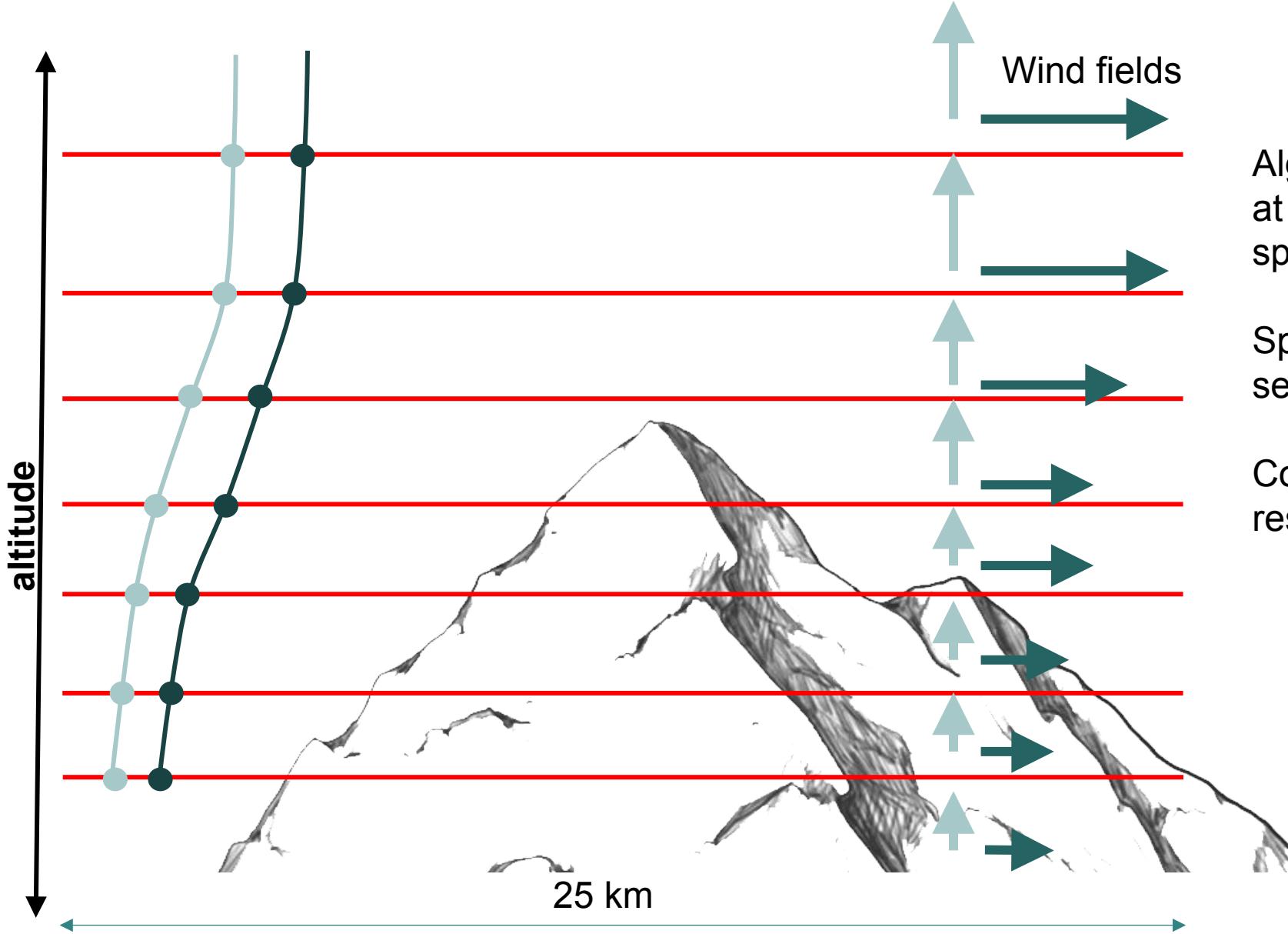
with

$$h = \frac{h_t \cdot (H + (H_c - H)(1 - H_c))}{H_c}$$

and h_t being the logit-transformed version of $hurs_{\text{orog}}$:

$$h_t = \log \left(\frac{hurs_{\text{orog}}}{1 - hurs_{\text{orog}}} \right),$$



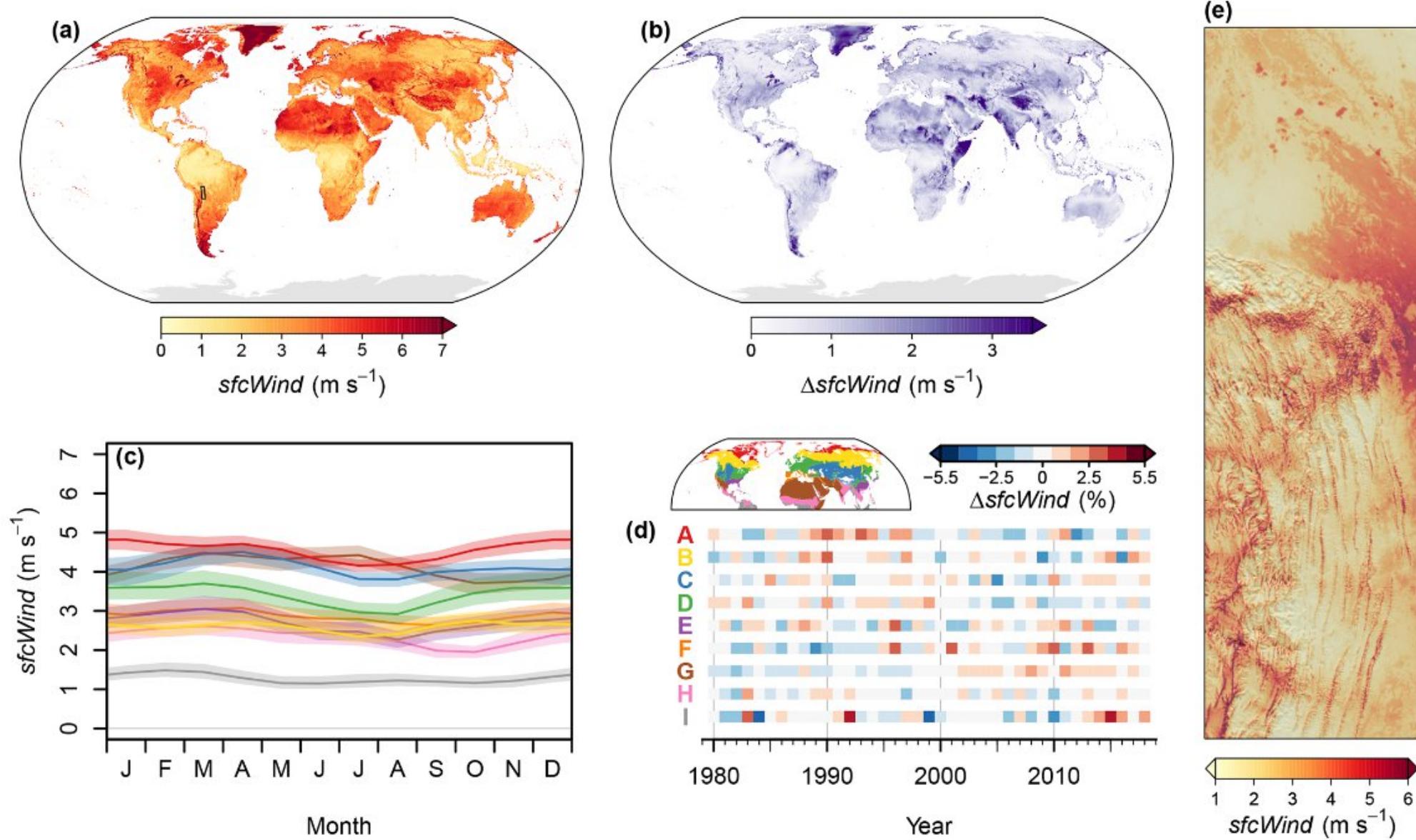


Wind fields

Algorithm to downscale wind speed
at pressure levels to surface wind
speed (*sfcWind*)

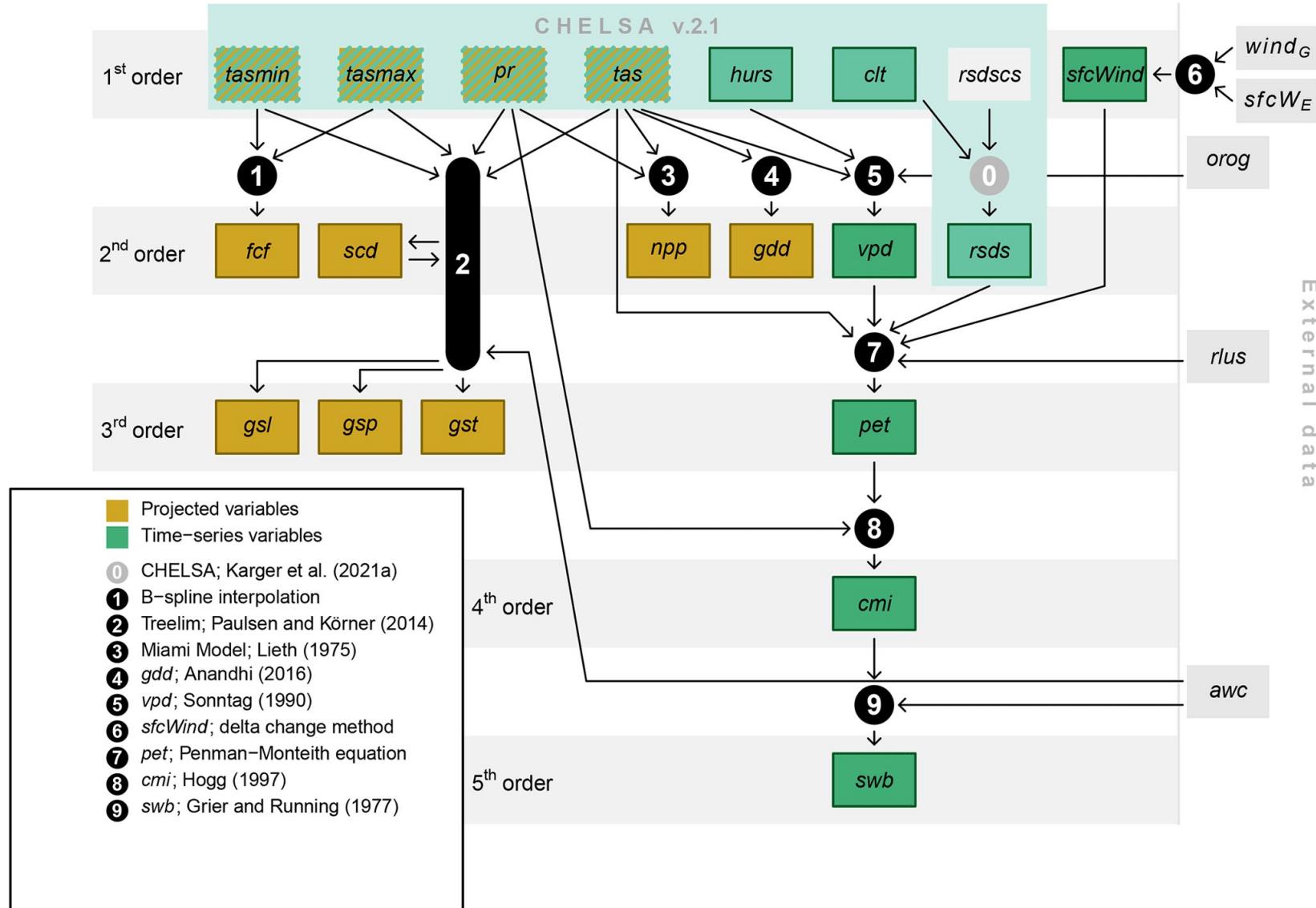
Spline interpolation between
seperately for wind components (u, v)

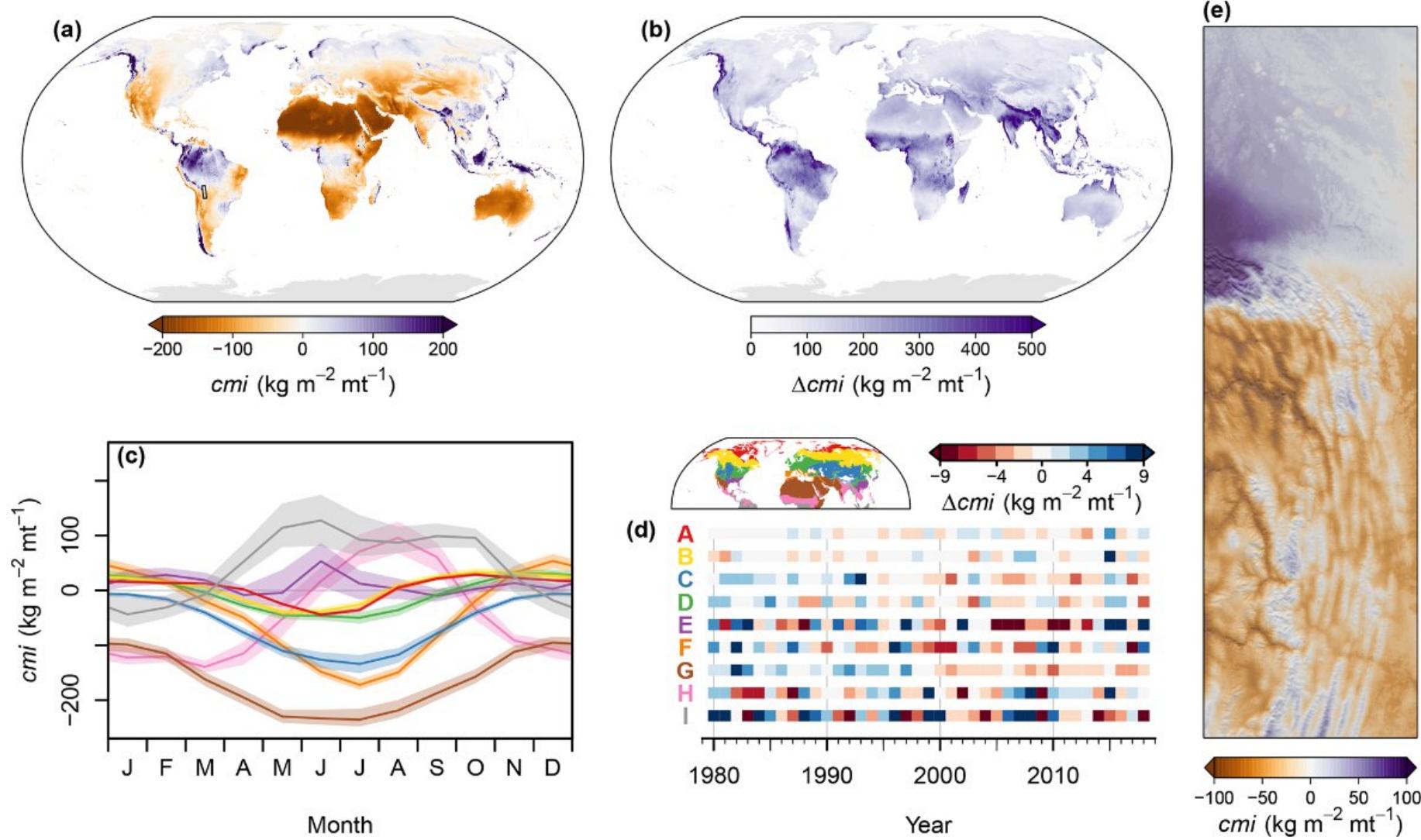
Combine wind vectors at high
resolution to wind speed

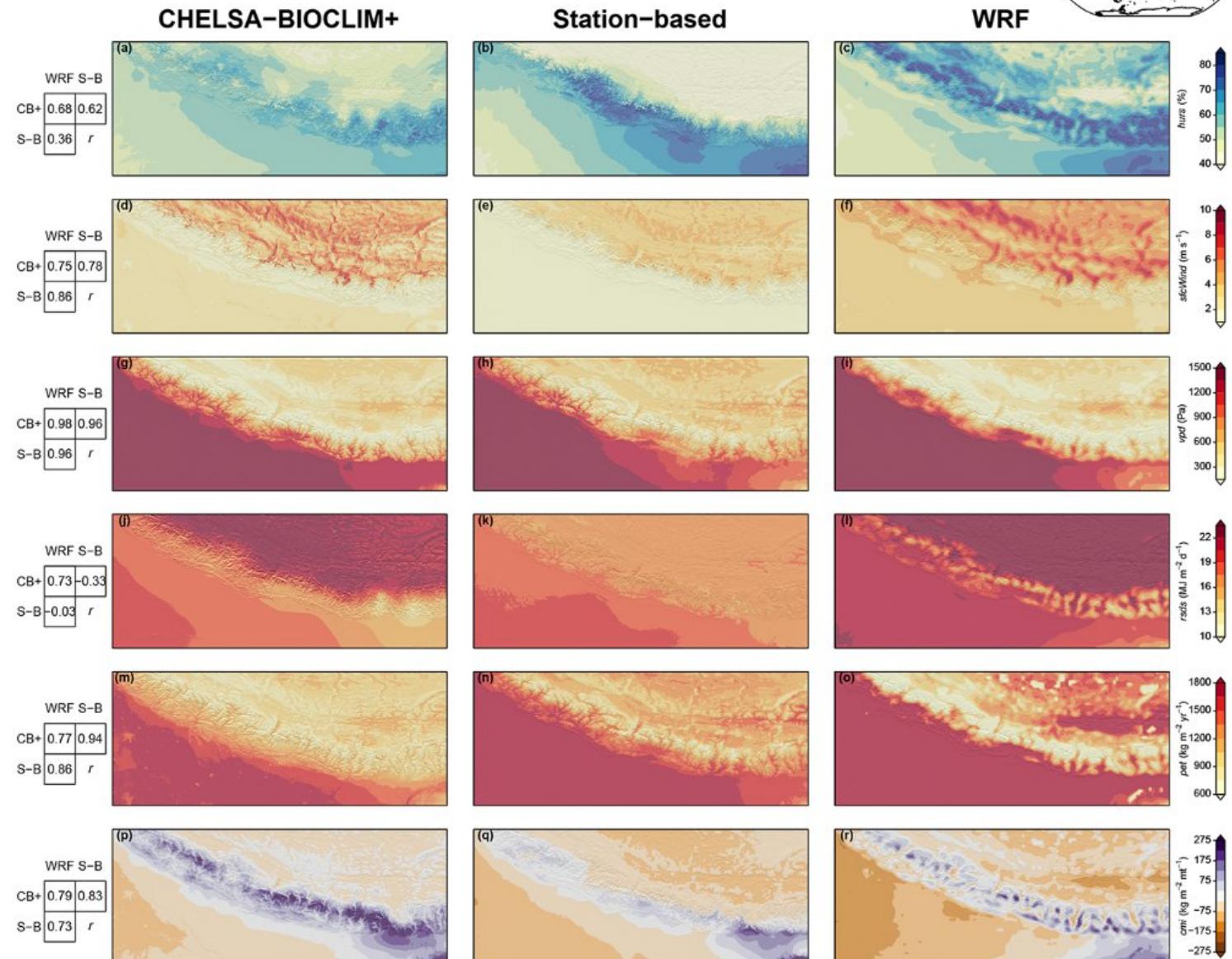
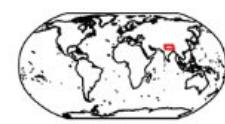


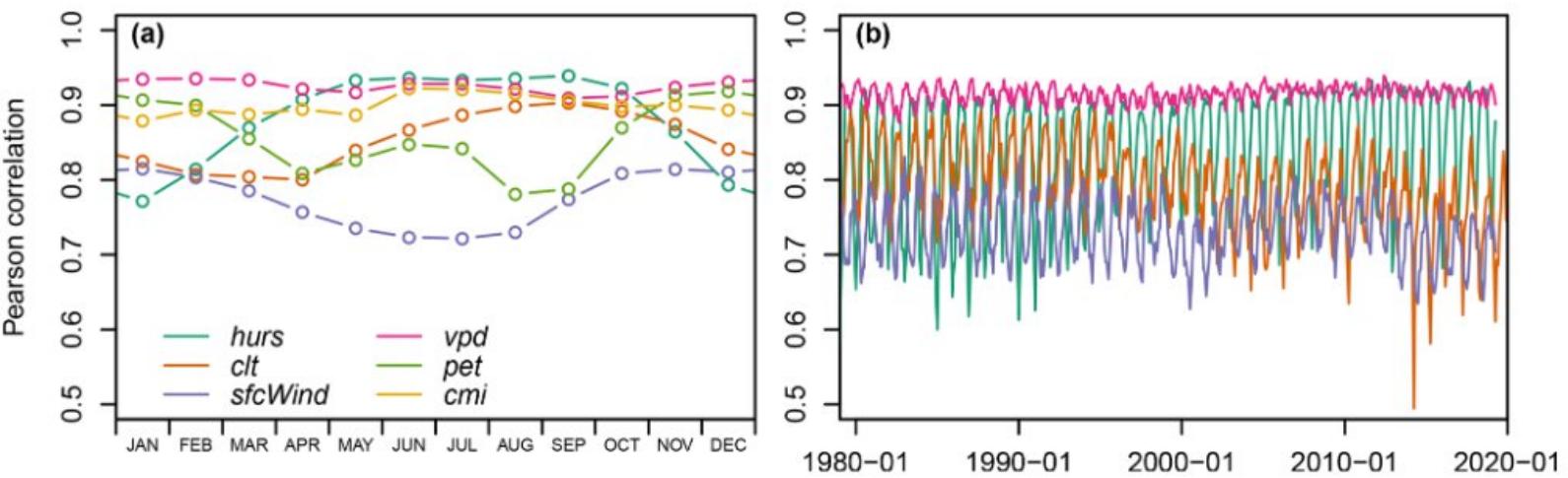
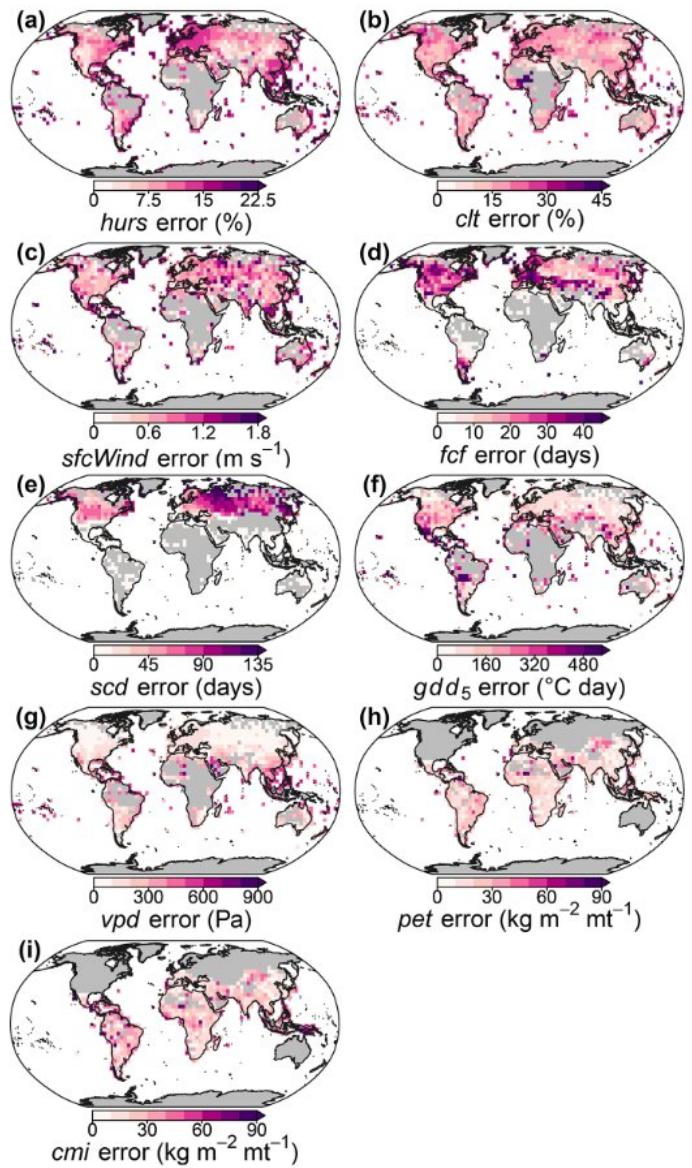
CHELSA – variables

shortname	description	CF Standard Name	levels	frequency
tas	daily-mean near-surface (2 meter) air temperature	air temperature	surface	day
tasmax	daily-maximum near-surface (2 meter) air temperature	air temperature	surface	day
tasmin	daily-minimum near-surface (2 meter) air temperature	air temperature	surface	day
rsds	total downwelling shortwave solar radiation	surface downwelling shortwave flux in air	surface	day
rlds	total downwelling longwave solar radiation	surface downwelling longwave flux in air	surface	day
hurs	daily-mean near-surface (2 meter) relative humidity	relative humidity	surface	day
pr	daily precipitation flux (liquid and solid)	precipitation flux	surface	day
ps	daily mean near-surface (2 meter) air pressure	surface air pressure	surface	day
clt	surface total cloud area fraction	cloud area fraction	surface	day
tz	near surface temperature lapse rate	-	surface	day
sfcWind	daily-mean 10m wind speed	wind speed	surface	day









Correlations with observations high for:

vpd

pet
cni

lower for:

clt
sfcWind
hurs

Good news: Low error propagation !

Where are we going with this?

- Monthly versions are online
- Daily version is in production (delayed due to CDS server switch)
- R package to access data on gitlabext.wsl.ch (beta version)
- Aim is: to produce a comprehensive dataset for 1940-today
- No CMIP planned yet (storage and compute problems)



Acknowledgements



Funders & Partner institutions



Thanks....

Bias compared to stations
In the continental United States over 2003-2016.

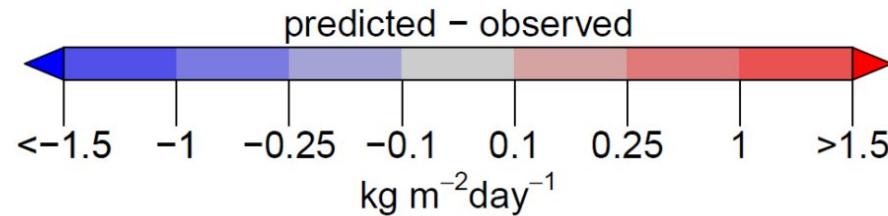
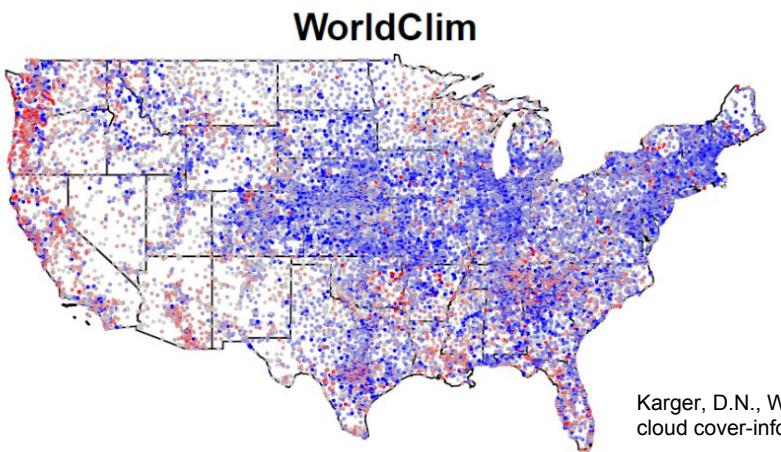
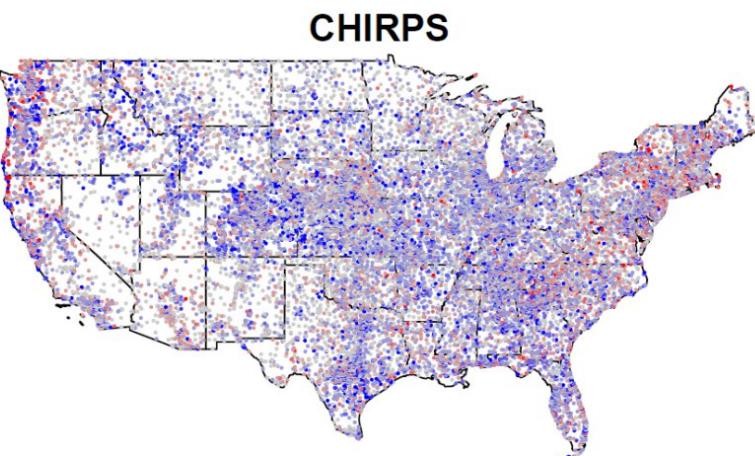
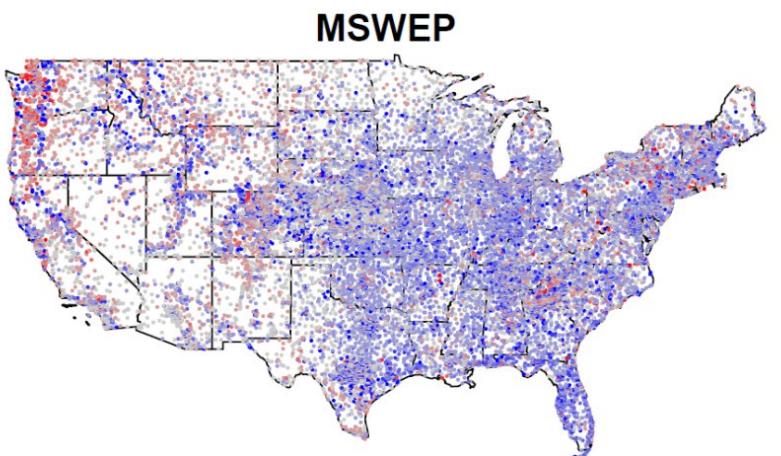
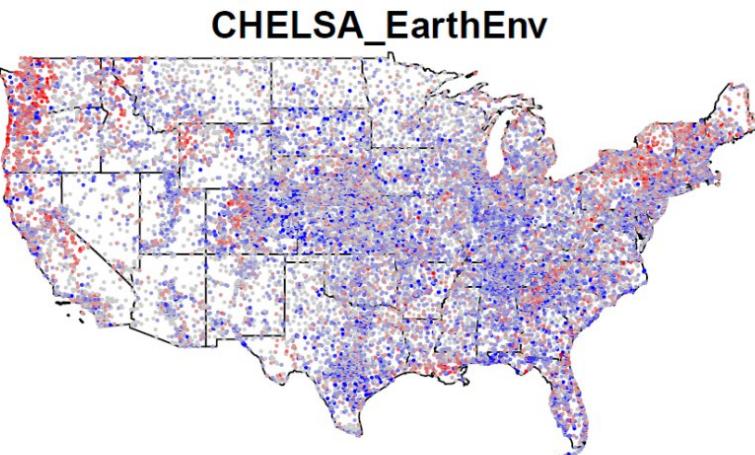
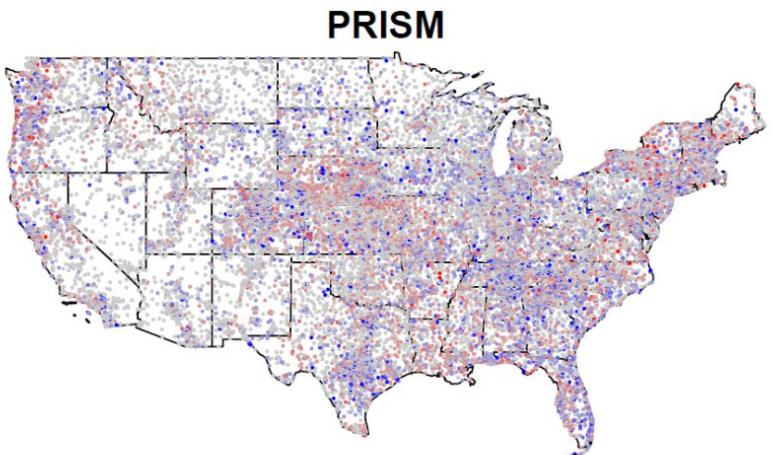
PRISM has lowest bias

- Only available in US
- Uses most of these stations

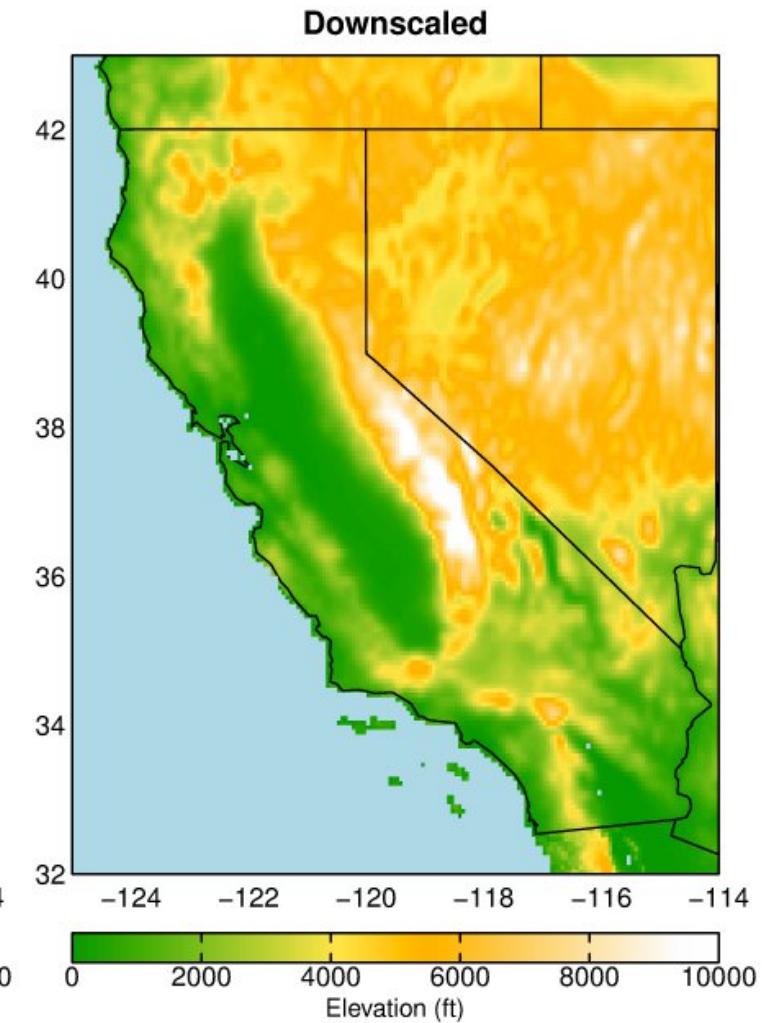
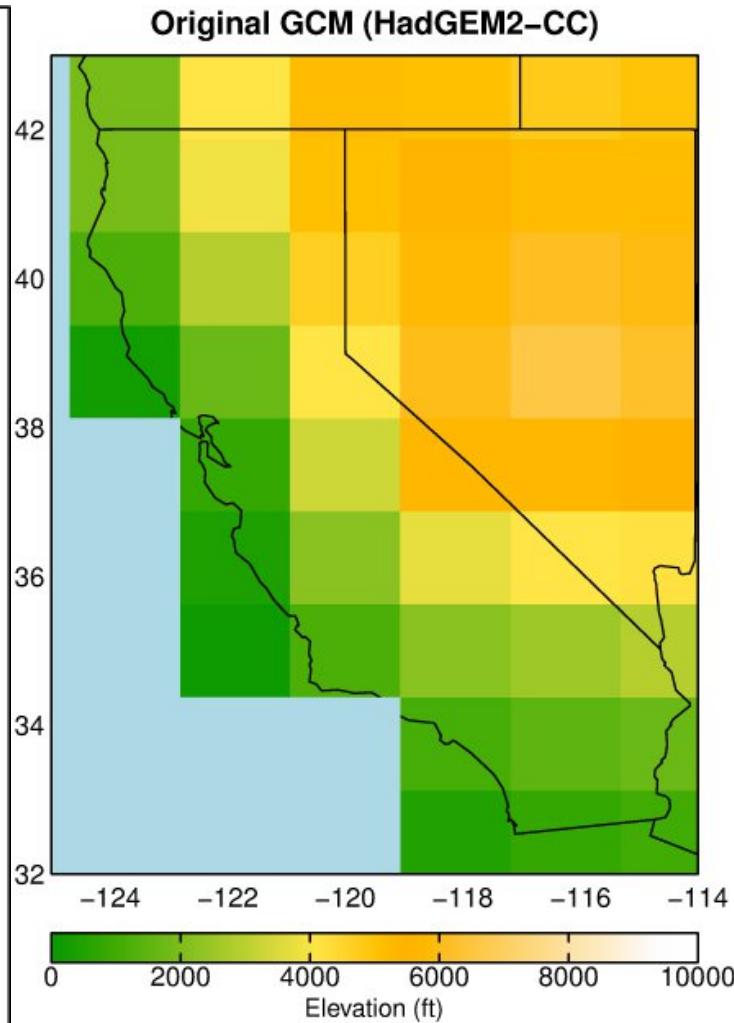
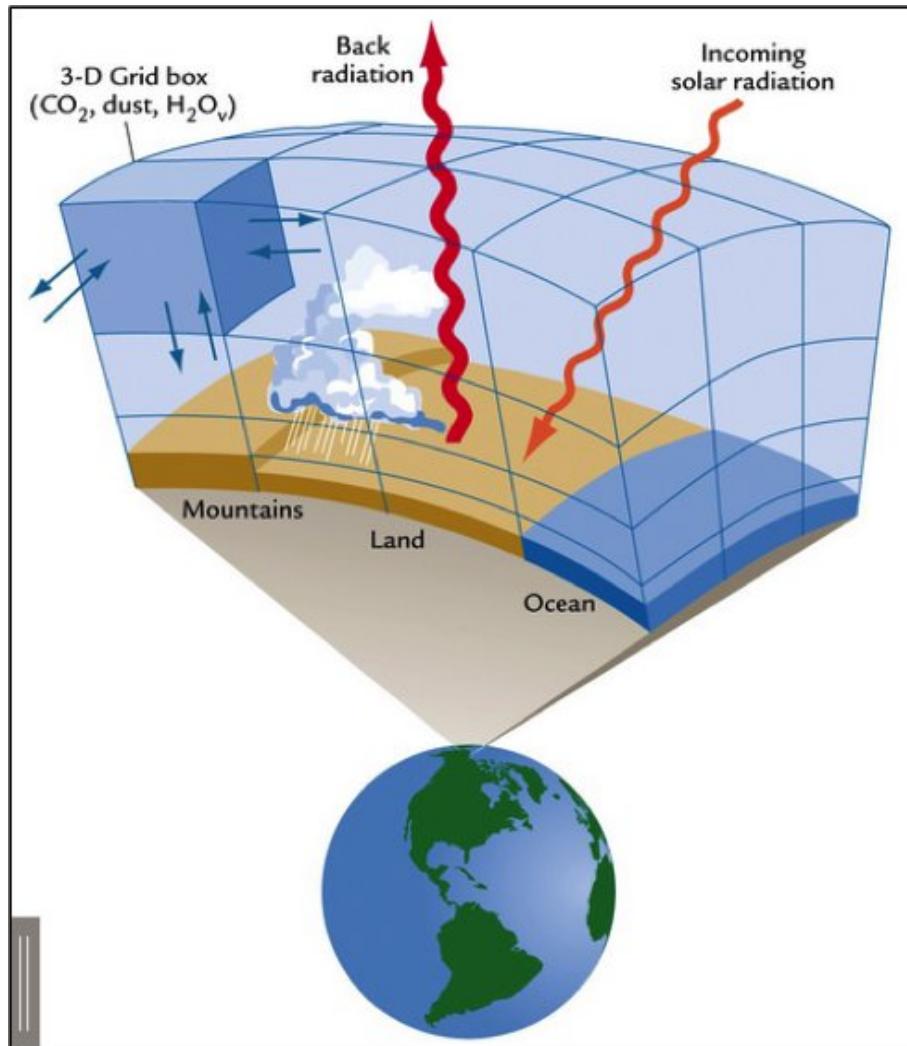
CHELSA, MSWEP,
CHIRPS have moderate biases

WorldClim has large bias

- Version using CRU TS stations used here



Global climate model data is usually too coarse for most ecological applications



Only a few models (e.g. ICON) are currently able to run at kilometer scale globally

They are computationally extremely intensive

Run over short time periods only

Extremely large amount of data

Not available for years to come

