



**Barcelona  
Supercomputing  
Center**

*Centro Nacional de Supercomputación*

# R user meeting

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02/06/2022

# Agenda

1. Ice-breaker
2. News
  - s2dv
  - ClimProjDiags
  - General R
3. User presentation: CSDownscale package [Jaume]
4. Q&A
  -

# Ice-breaker

# No ice-breaker today :(



IT periodic talk yesterday (1st June)

[https://earth.bsc.es/wiki/lib/exe/fetch.php?media=computing:it\\_periodic\\_talks\\_hpc\\_01-06-2022.pdf](https://earth.bsc.es/wiki/lib/exe/fetch.php?media=computing:it_periodic_talks_hpc_01-06-2022.pdf)

s2dv

# MeanDims() efficiency improvement

s2dv::MeanDims is improved to have better efficiency.

Status: Master branch

!! Note that apply() could be very slow when operation is over large dimensions !!

```
exp <- array(rnorm(10000), dim = c(dat = 2, memb = 4, sdate = 120, lat = 90, lon = 60))
MeanDims(exp, c('dat', 'memb'))
```

new s2dv::MeanDims

Code	File	Memory (MB)		Time (ms)
▼ MeanDims2	<expr>	0	0	80
aperm	MeanDims2.R	0	0	80

old s2dv::MeanDims

Code	File	Memory (MB)		Time (ms)
▼ MeanDims1	<expr>	-133.4	155.9	6180
▼ apply	MeanDims1.R	-133.4	155.9	6180
► FUN		-86.0	78.8	3770
array		-24.3	45.1	1840
aperm		0	0	80











# MeanDims() efficiency improvement

The functions that use MeanDims() inside are more efficient now.







E.g., ACC()

```
exp <- array(rnorm(100000), dim = c(dataset = 1, lat = 30, lon = 20, member = 10, sdate = 120))
obs <- array(rnorm(100000), dim = c(dataset = 1, lat = 30, lon = 20, member = 1, sdate = 120))
```

**new s2dv::ACC**

Code	File	Memory (MB)	Time (ms)
▼ ACC	<expr>	0  38.2	480 
▼ compiler::tryCmpfun	<expr>	0  3.2	200 
▶ tryCatch		0  3.2	200 
▶ Apply		0  32.4	200 
▶ MeanDims		0  2.6	80 

**old s2dv::ACC**

Code	File	Memory (MB)	Time (ms)
▼ ACC	<expr>	-49.4  69.1	1370 
▶ MeanDims		-49.4  54.2	1170 
▶ Apply		0  14.9	200 

# PlotEquiMap new parameters

New parameters ``xlabels``, ``ylabels``, ``xlonshft``, ``ylatshft`` to improve the tick labels, especially when longitudes and latitudes are shifted.

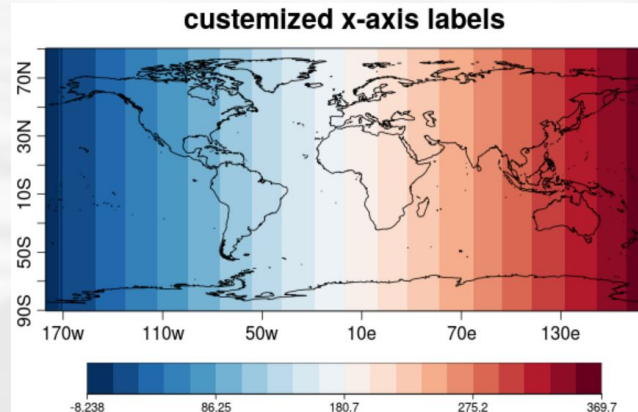
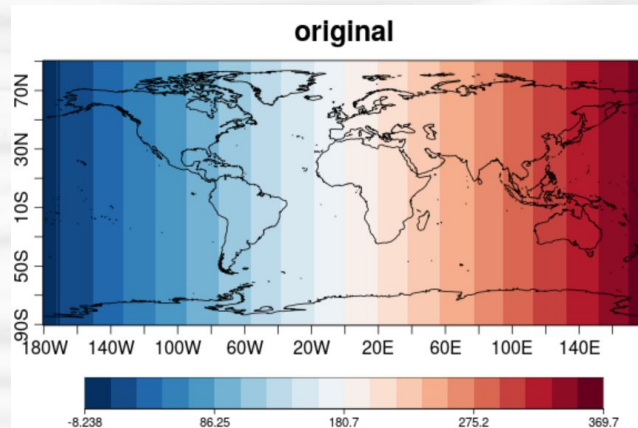
The default sets the first x tick label at 0 (left margin), and ``xlonshft`` shifts the ticks by the degrees specified, e.g. ``xlonshft = 10`` will shift the ticks by 10 degrees.

With `'xlabels'`, you can customize the value of longitude labels. The number of values depend on the parameter `'intxlon'`, which is the interval between longitudes to label. The main point of this parameter is to avoid the wrong labels (see plot a) when shifting the lons.

Status:

[https://earth.bsc.es/gitlab/es/s2dv/-/tree/develop\\_PlotEquiMap](https://earth.bsc.es/gitlab/es/s2dv/-/tree/develop_PlotEquiMap)

```
xlonshft=10,intxlon=60,xlabels=c(paste0(seq(170, 50, by = -60), 'w'), paste0(seq(10, 190, by = 60), 'e'))
```





# PlotEquiMap shift map

The current PlotEquiMap() can only plot map with either [0, 360] or [-180, 180]. The new development allows flexible longitude range.

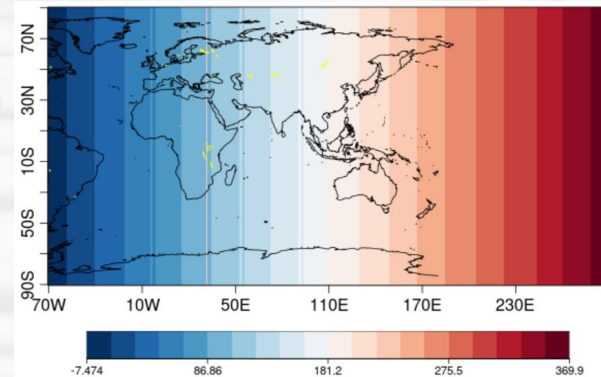
Status:

[https://earth.bsc.es/gitlab/es/s2dv/-/tree/develop\\_PlotEquiMap](https://earth.bsc.es/gitlab/es/s2dv/-/tree/develop_PlotEquiMap)

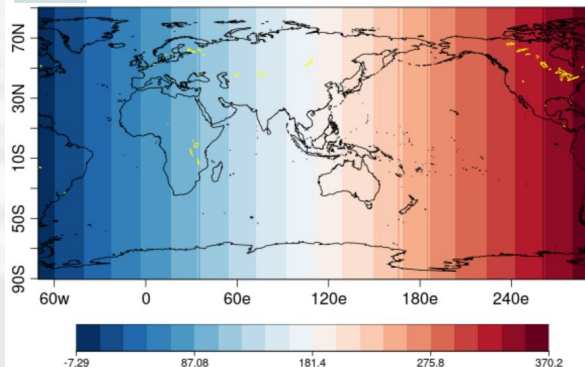
**\*\*Since the code of plotting map is changed quite a lot, it is recommended testing it with your script before it's officially included in the package.**

```
data <- array(1:360, dim = c(lon = 360, lat = 181))
lon <- -180:179 + 110 #shift 110deg
lat <- -90:90
PlotEquiMap(data, lon, lat, filled.continents=F,
country.borders=F, filled.ocean=F, lake_color='yellow',
xlonshft=10, intxlon=60,
xlabels=c('60w', '0', paste0(seq(60, 300, by=60), 'e')))
```

old



new



# Load(): Bugfix for R\_4.1.2

Load() returns error in R\_4.1.2:

*Error in prod(dim\_exp) : invalid 'type' (list) of argument*

Status: Master branch

- s2dverification won't be fixed. So it is only available for R < 4.0.0.
- CStools::CS\_Load uses s2dverification::Load for now. Before it is changed to s2dv, it can't perform well in R\_4.1.2 either.

# ClimProjDiags

# WeightedMean() bugfix

dlon calculation correction

<https://earth.bsc.es/gitlab/es/ClimProjDiags/-/issues/7>

Status: Master branch

Use it now by

```
source("https://earth.bsc.es/gitlab/es/ClimProjDiags/-/raw/master/R/WeightedMean.R")
```

# General R

# General R

- Start using **module R/4.1.2** on Workstation and Nord3v2, and report problems if found.
- RStudio on WS:  
[https://earth.bsc.es/wiki/doku.php?id=computing:workstations#using\\_rstudio-server\\_in\\_ws](https://earth.bsc.es/wiki/doku.php?id=computing:workstations#using_rstudio-server_in_ws)

RStudio should have the same environment (same modules loaded, same directories, etc.) as your workstation. If you find them different, restart RStudio by:

- Press command/ctrl + shift + F10 (with fn)
- Type ``.rs.restartR()``

# CSDownscale



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# Statistical downscaling with CSDownscale

Jaume Ramon

*With inputs from Llorenç, Carlos D., Lluís, Núria, Raül, An-Chi, Alba...*

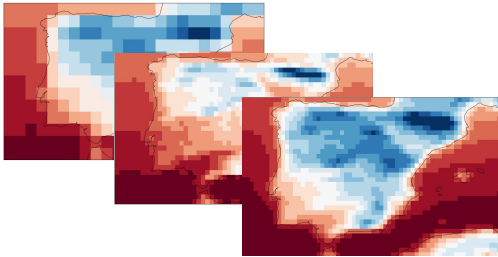
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R user meeting



# CSDownscale

- Will be an **open source** R package.
- The user should be able to easily **compare** the performance of different **downscaling methods**.

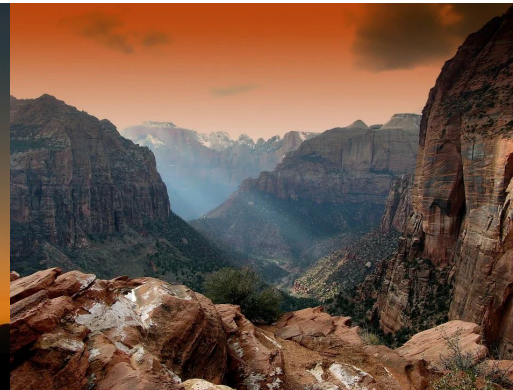


## Downscaling methods

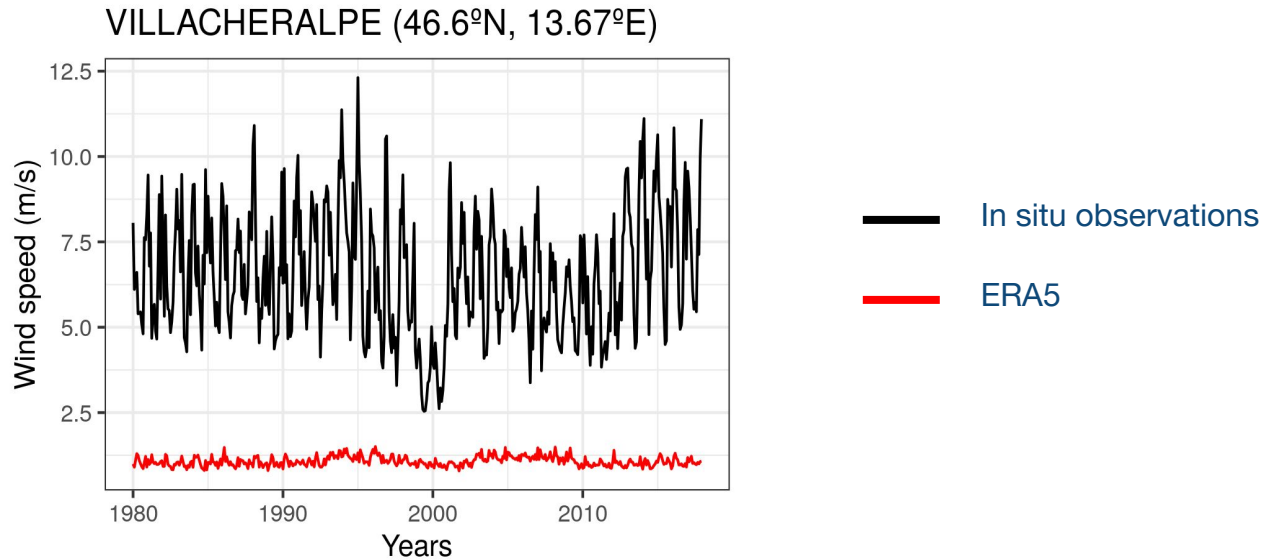
- 1) Interpolation
- 2) Interpolation plus bias adjustment
- 3) Interpolation plus linear regression
- 4) Large-scale predictors and local climate variables
- 5) Stencil
- 6) Analogs
- 7) Logistic regression

# Why downscaling?

- Downscaling is any procedure to infer **high-resolution** information from **low-resolution** variables.
- Climate predictions are delivered on **grids** of thousands of square kilometres, which is little useful for **local applications** (e.g. predictions at a wind farm, river basin or mountain valley).
- Go from coarse to fine grids, or even to a **point scale**.



# Why downscaling?



# Functions in CSDownscale

```
Interpolation <- function(exp, points = NULL, method_remap = NULL, target_grid = NULL, lat_dim = "lat",  
                          lon_dim = "lon", remap_region = NULL, method_point_interp = NULL)
```

```
Intbc <- function(exp, obs, target_grid, int_method, bc_method, points = NULL, method_point_interp = NULL,  
                 lat_dim = "lat", lon_dim = "lon", sdate_dim = "sdate", member_dim = "member", ncores = 1)
```

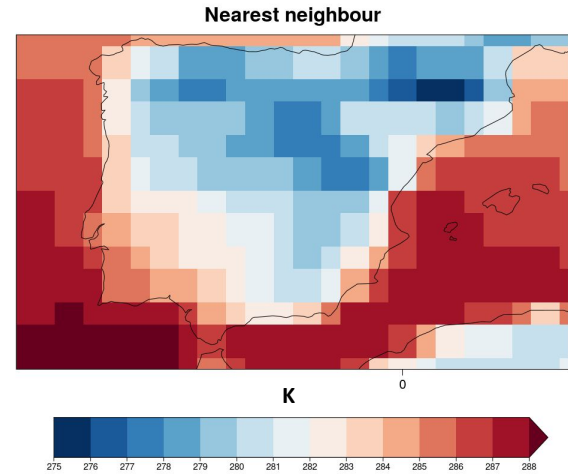
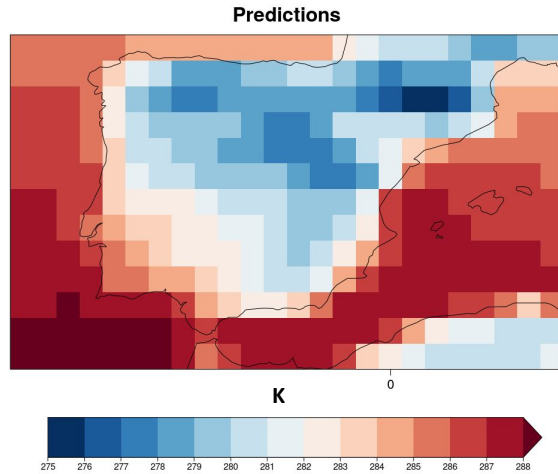
```
Intlr <- function(exp, obs, target_grid, int_method, lr_method, predictors = NULL, lat_dim = "lat",  
                 lon_dim = "lon", sdate_dim = "sdate", time_dim = "time",  
                 large_scale_predictor_dimname = 'vars', loocv = FALSE, ncores = 1)
```

```
Analogs <- function(exp, obs, nanalogs = 3, fun_analog = "mean", lat_dim = "lat", lon_dim = "lon",  
                   sdate_dim = "sdate", time_dim = "time", grid_exp = NULL, ncores = 1)
```

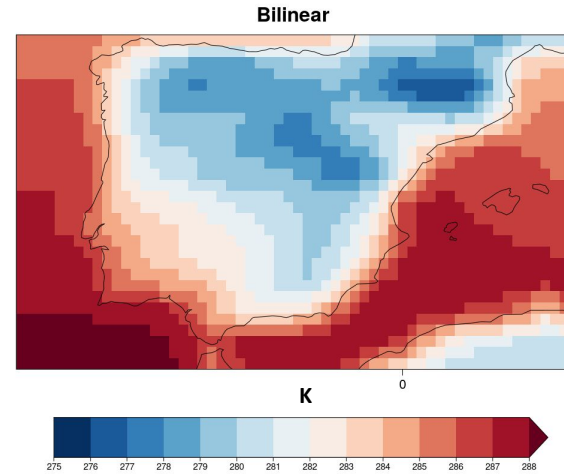
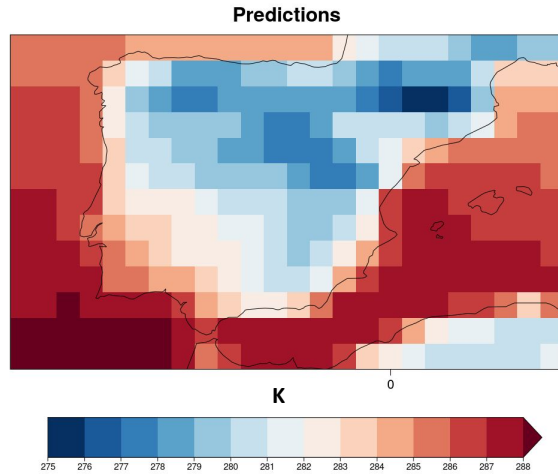


# Interpolation(..., method = "nearest\_neighbour")

- Regrid from coarse to fine grid.

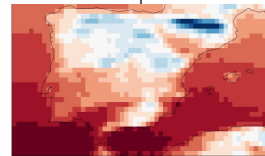
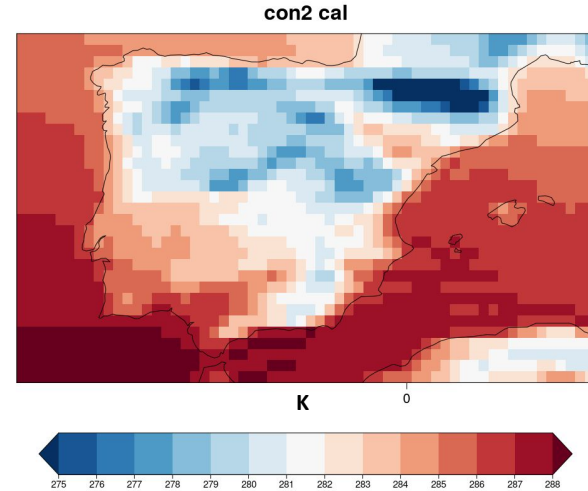
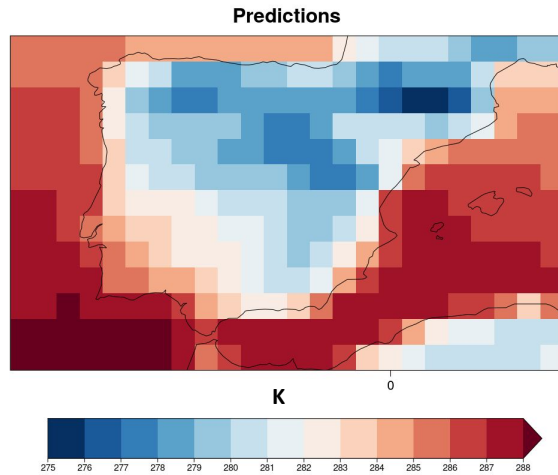


# Interpolation(..., method = "bilinear")



# Intbc(..., bc\_method = “calibration”)

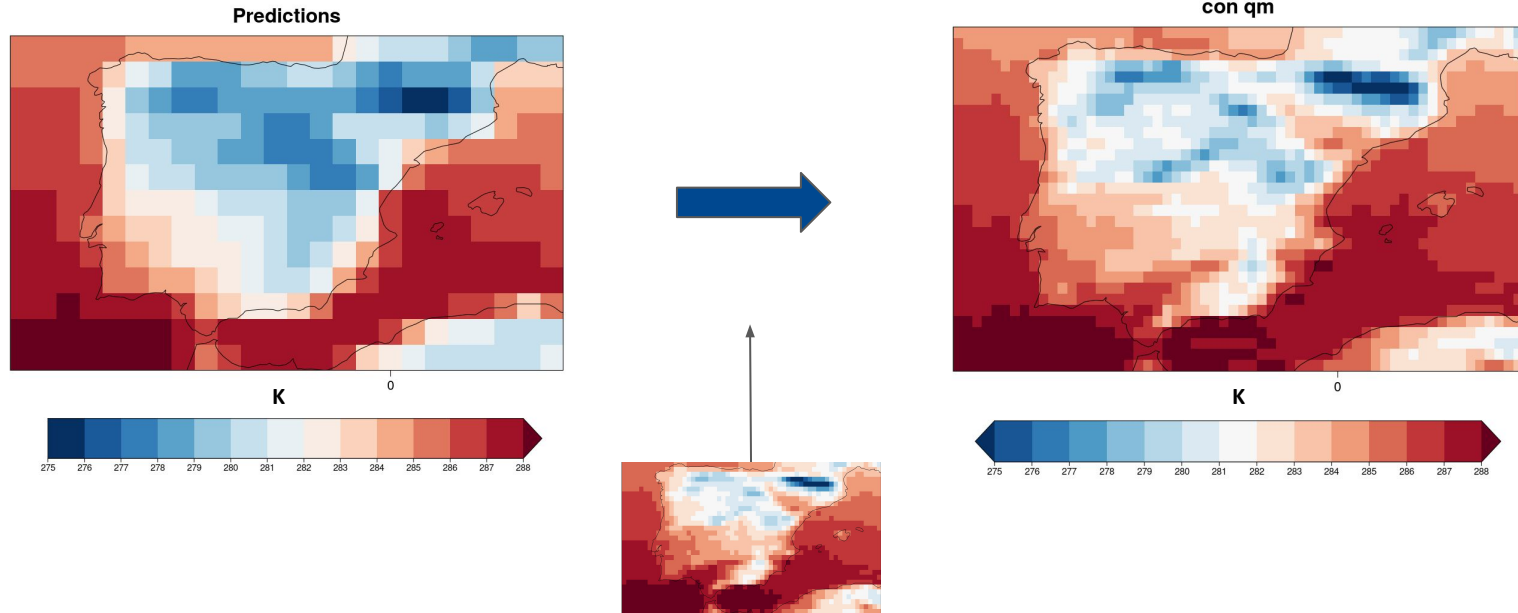
- First, interpolate from coarse to fine grid. Then, do a bias adjustment.



High-res observations

# Intbc(..., bc\_method = “quantile\_mapping”)

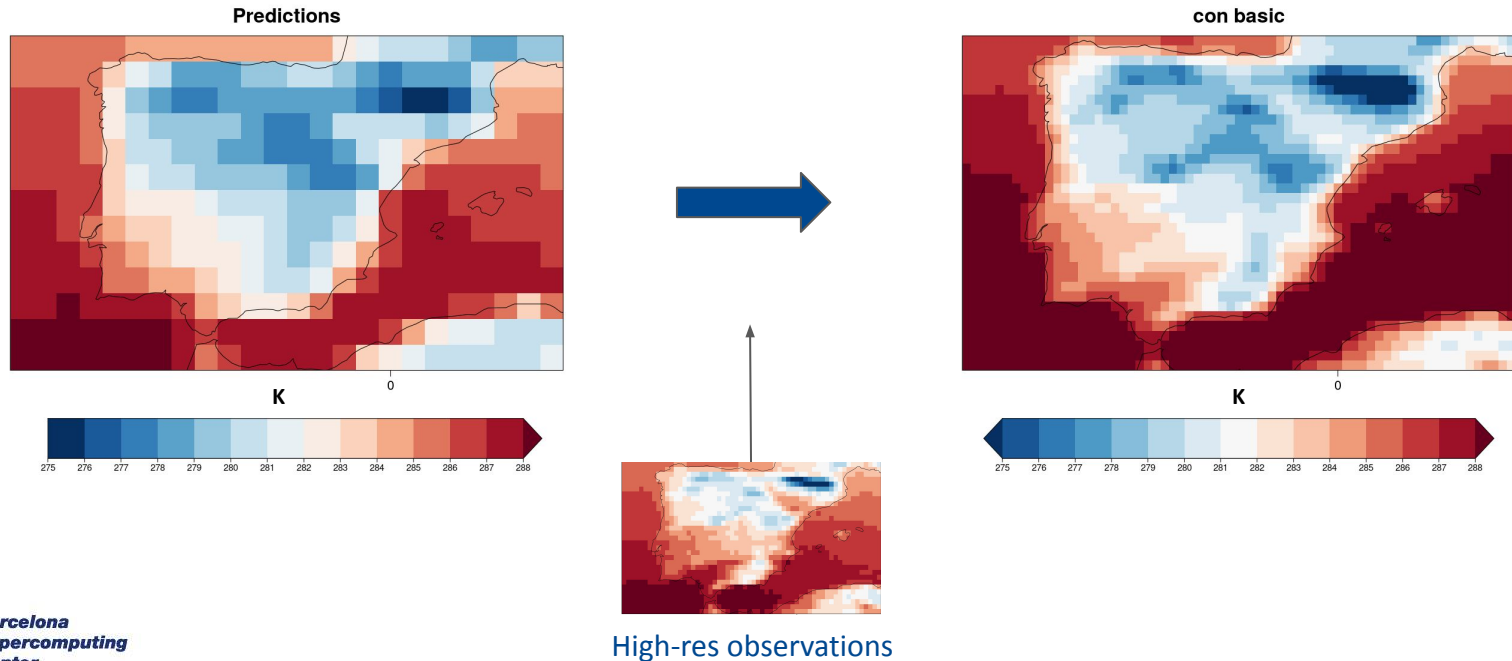
- First, interpolate from coarse to fine grid. Then, do a bias adjustment.





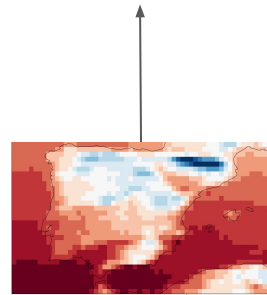
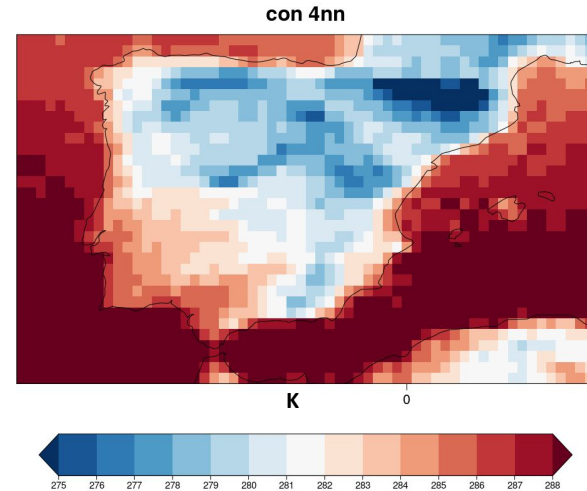
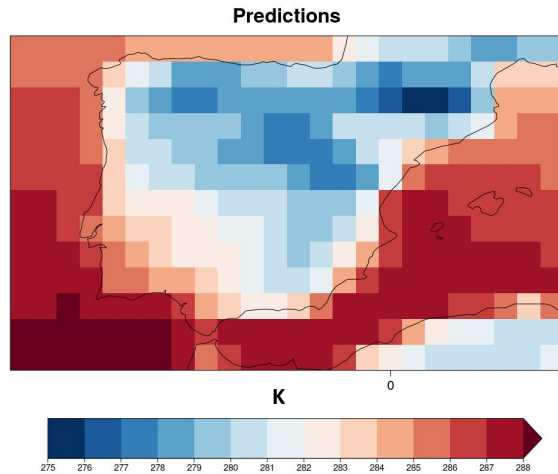
# Intlr(..., lr\_method = "basic")

- First, interpolate from coarse to fine grid. Then, adjust the models with a point-wise linear regression with the high-res observations as predictands.



# Intlr(..., lr\_method = "4nn")

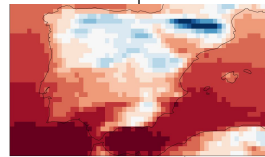
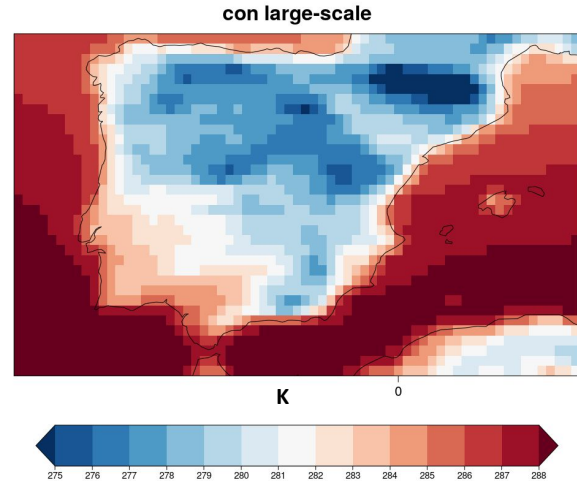
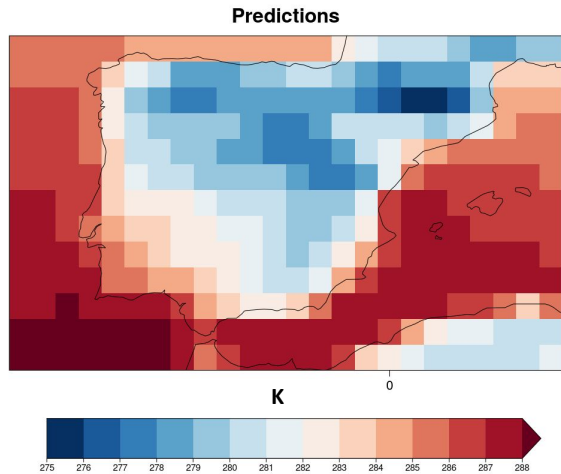
- First, interpolate from coarse to fine grid. Then, adjust the models with a point-wise multi-linear regression with the four NN as predictors.



High-res observations

# Intlr(..., lr\_method = "large-scale")

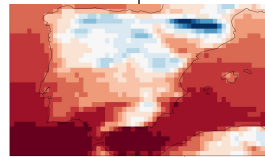
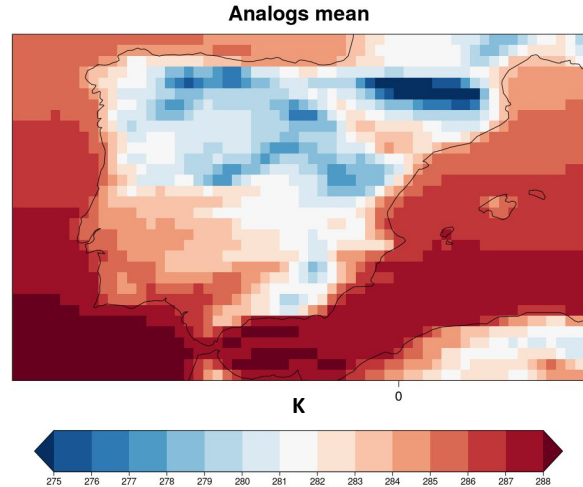
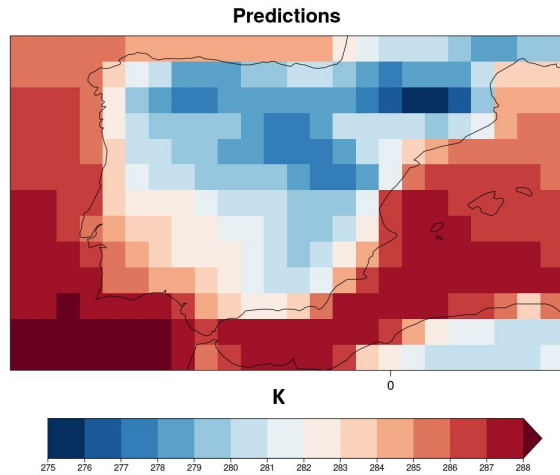
- First, interpolate from coarse to fine grid. Then, adjust the models with a point-wise multi-linear regression with large-scale predictors (e.g. teleconnection indices).



High-res observations

# Analogs(..., nanalogs = 3, fun\_analogs = 'mean')

- Interpolate high-res observations to model grid. Then, compute the Euclidean distance between the model and observations' fields. Select the best analogs in the high-res observational fields.



High-res observations

# Q & A

# Thanks for joining

Next meeting: 7th July 2022 (11 am)