

Supplementary Information: Paleoclimate Modeling

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Introduction

We'll be comparing paleoclimate model estimates of temperature and precipitation over three points in the west Mediterranean to global paleoclimate proxies.

Setup

Load all the packages we'll need for this analysis.

```
library(ncdf4) # import GCM data
library(rgdal) # read GCM data
library(raster) # process GCM data
library(rasterVis) # plotting GCM data
library(tidyverse) # data management and plotting
library(magrittr) # pipes for code readability
library(EMD) # calculate trends in the data
```

You'll need to have the netCDF libraries already installed on your system for ncdf4 to work.

Climate Model

Sample Locations

Create a matrix with the coordinates for the three locations of interest in the west Mediterranean.

```
samp.pts <- matrix(c(0, 40, 4, 44, 12, 46, 14, 43),
                     ncol = 2, byrow = T)

#12, 43
```

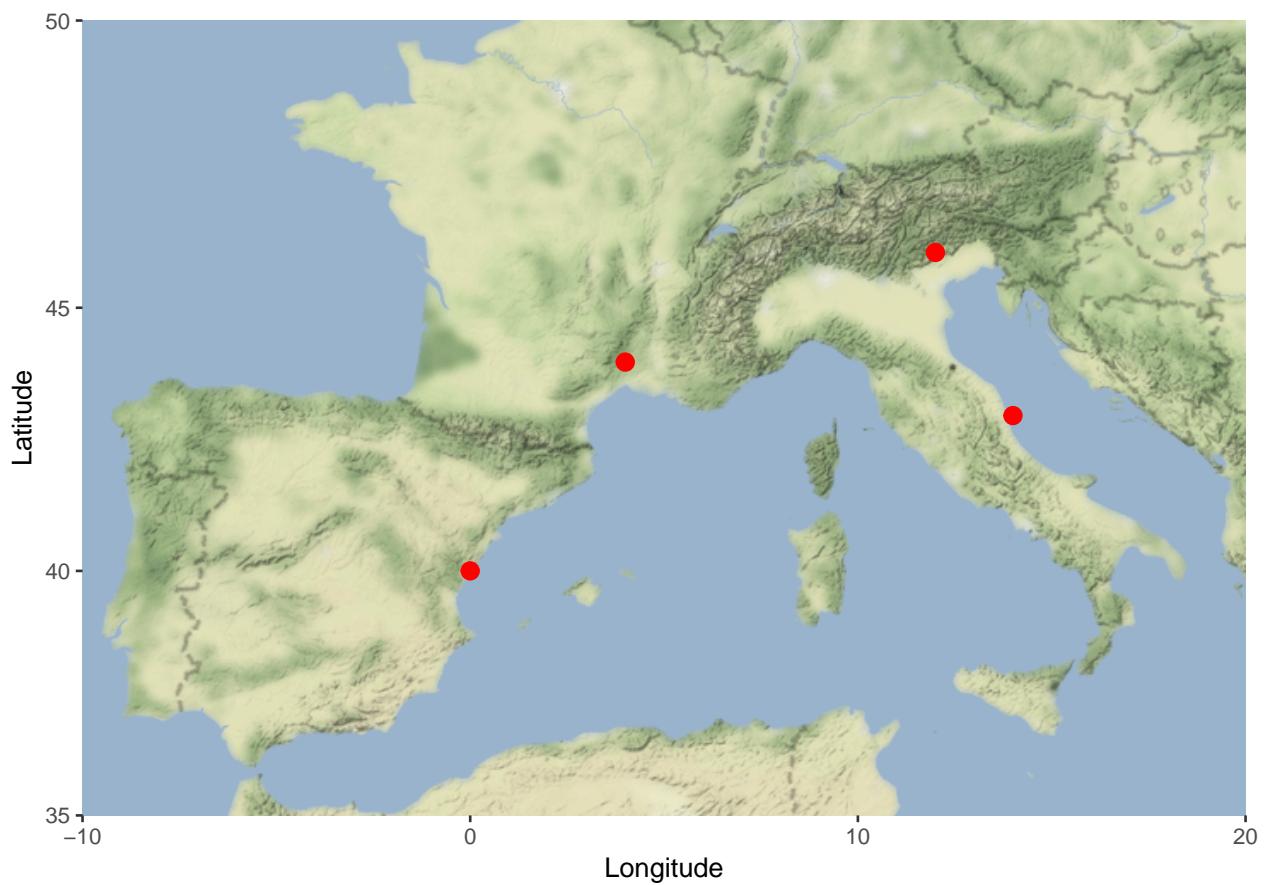
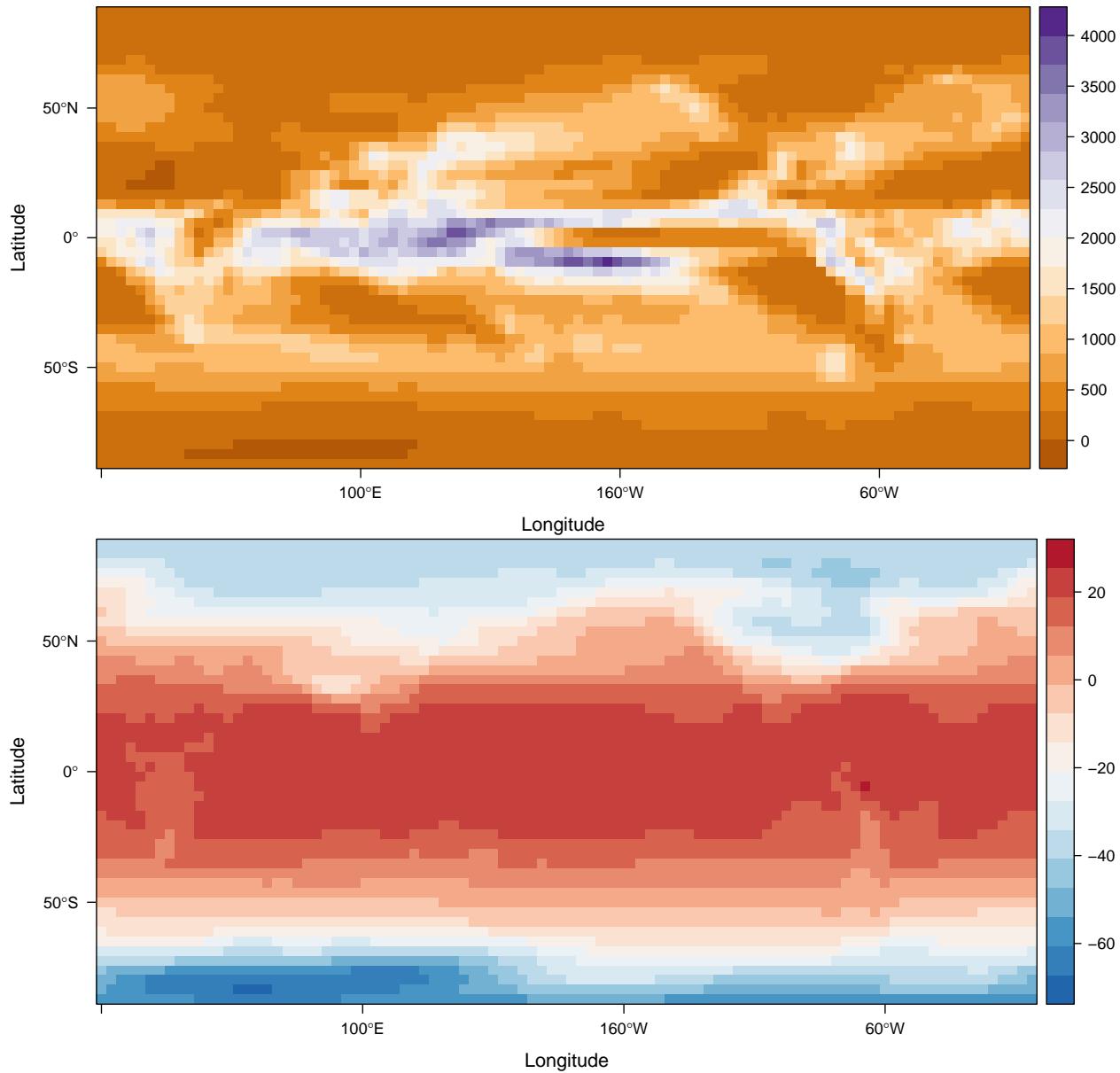


Figure 1: Locations of 3 sample points.

TraCE-21k



First, import data from the TraCE-21k paleoclimate simulation. Then extract temperature and precipitation values at three locations in the west Mediterranean. Use the `brick` function from **raster** to import decadal averages from the simulation. Put the coordinates for the three locations in a matrix, and use that matrix to and **raster**'s `extract` function to get the values from the climate model brick. Convert the precipitation values to mm/year and temperature values to degrees Celsius. Finally, name the columns for each region appropriately.

Now pull all the TraCE data into one data frame, with one row per year, and one column per variable/location combination. First `rbind` the two sets of TraCE data and `transpose` the results, turning the 6 rows into 6 columns. Add a column for the Year (in ka BP), and use to select only the entries earlier than 6,000 BP.

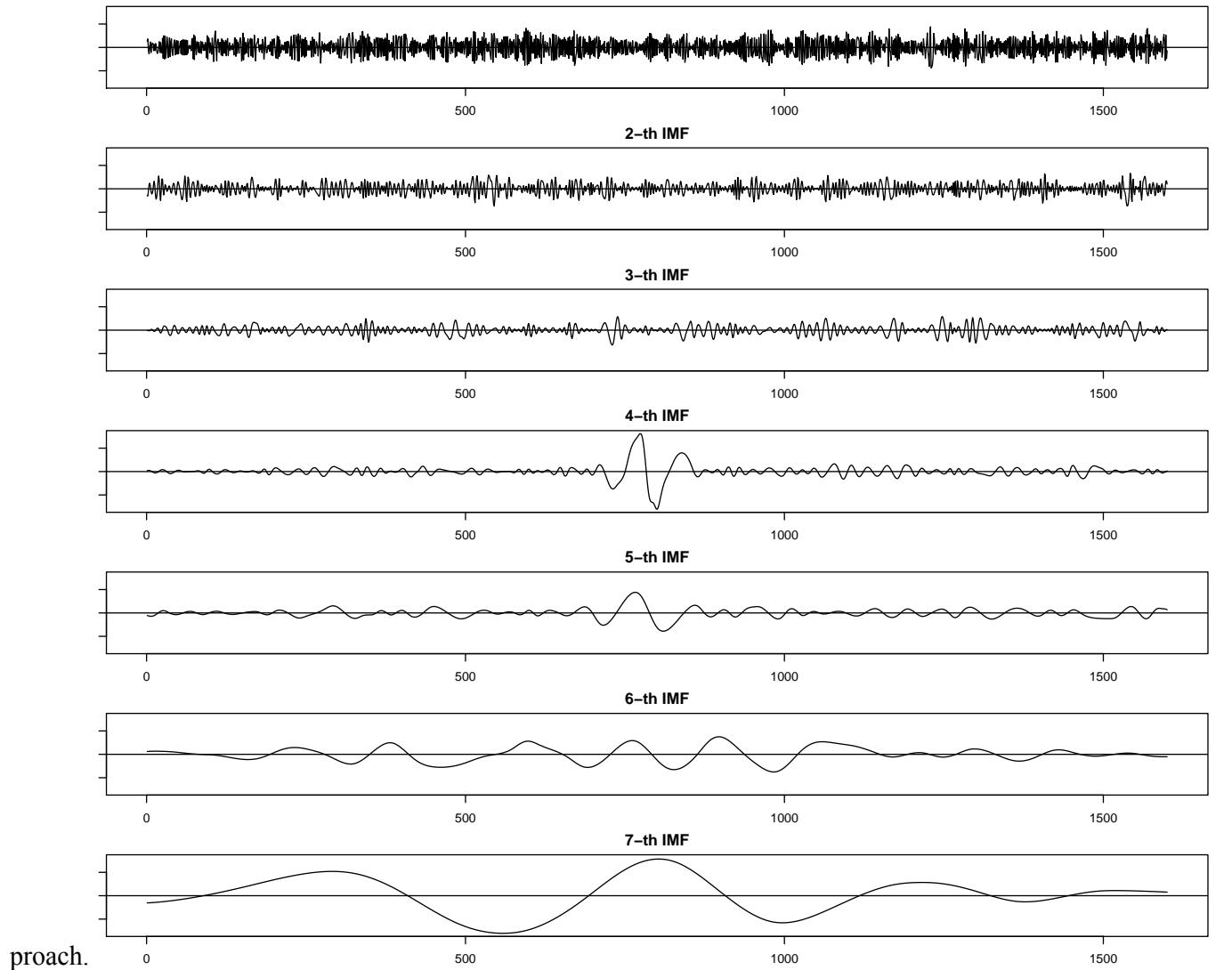
```

trace.dat <- rbind(
  brick('trace.01-36.22000BP.cam2.TREFHT.22000BP_decavg_400BCE.nc') %>%
    raster::extract(samp pts) %>%
    subtract(273.15), # convert from kelvin to C
  brick('trace.01-36.22000BP.cam2.PRECT.22000BP_decavg_400BCE.nc') %>%
    raster::extract(samp pts) %>% # extract data at these coordinates
    multiply_by(3.154e+10)) %>% # convert to mm/year
  t %>% # transpose
  as.data.frame %>%
  set_colnames(c('tmp,Southwest', 'tmp,North Central', 'tmp,Northeast', 'tmp,Southeast',
    'prc,Southwest', 'prc,North Central', 'prc,Northeast', 'prc,Southeast')) %>%
  rownames_to_column('Year') %>%
  mutate(Year = as.numeric(substr(Year, 3))) %>%
  filter(Year > 6) # get all the decades up to 6ka BP

```

Trend Analysis

Let's use the **EMD** package to calculate actual trend lines using the empirical mode decomposition approach



Now organize the temperature and precipitation data to make plotting easier using functions from **tidyverse**.

```
trace.plot <- trace.dat %>%
  gather(key, value, -Year) %>%
  separate(key, c('Variable', 'Region'), ',') %>%
  mutate(Region = factor(Region, levels = c('Southwest', 'North Central', 'Northeast', 'Southeast')) # turn into factor
  Variable = ifelse(
    Variable == 'tmp', 'Temperature (°C)', 'Precipitation (mm)')

emd.res <- function(x) emd(x, boundary = 'wave')$residue
trace.emd <- trace.dat %>%
  mutate_at(vars(-Year), emd.res) %>%
```

Replace the variable names to make facet naming easier too.

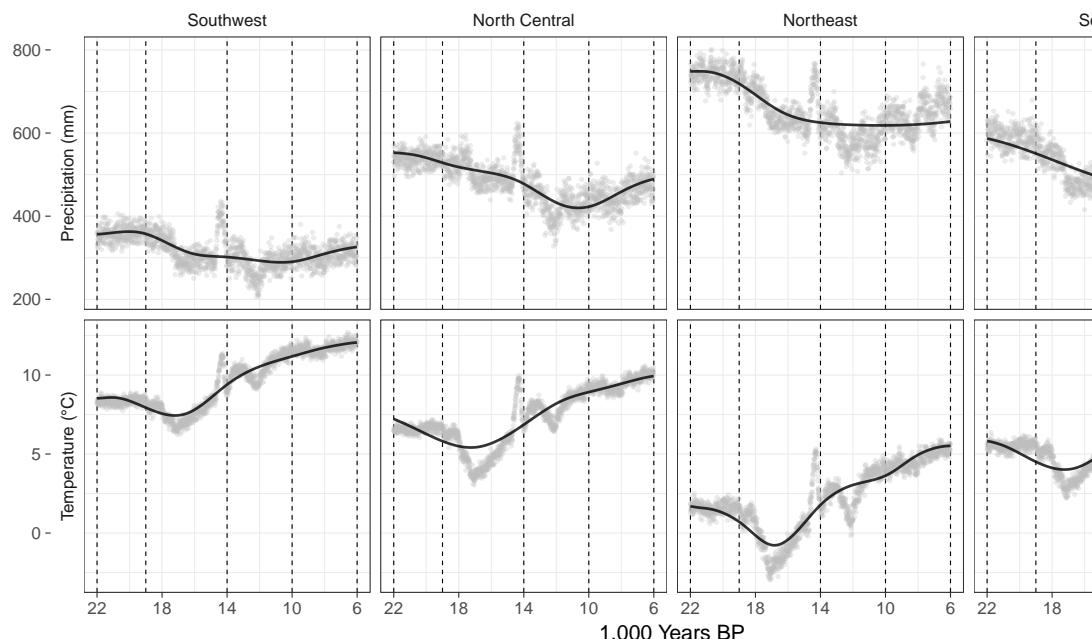
```

gather(key, value, - Year) %>%
separate(key, c('Variable', 'Region'), ',') %>%
mutate(Region = factor(Region, levels = c('Southwest', 'North Central', 'Northeast', 'Southeast')),
       Variable = ifelse(
         Variable == 'tmp', 'Temperature (°C)', 'Precipitation (mm)'))

#trace.plot %>% mutate(Raw = value, Trend = trace.emd[,4], value = NULL) %>% write_csv('~/Downloads/trace_plo

```

Plotting

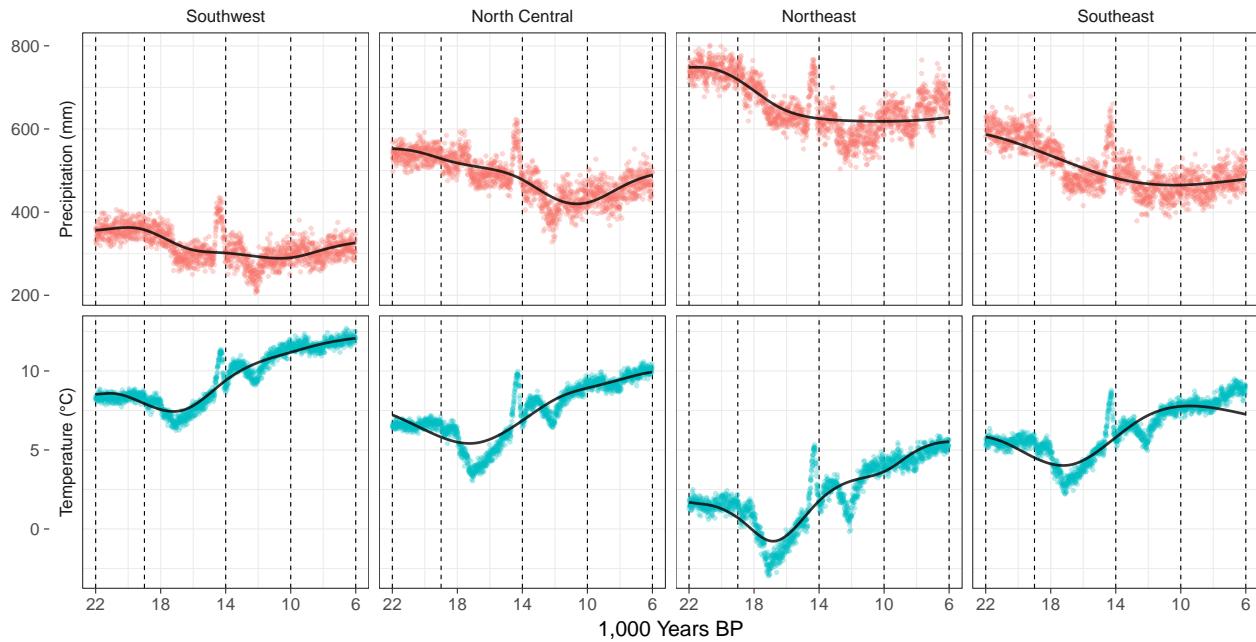


Plot everything with **ggplot2**.

```

ggplot(data = trace.plot, aes(x = Year, y = value)) +
  facet_grid(Variable ~ Region, switch = 'y', scale = 'free_y') +
  geom_vline(xintercept = c(22, 19, 14, 10, 6), lty = 2) +
  geom_point(aes(color = Variable), alpha = .3) +
  geom_line(data = trace.emd, size = 1.2, color = "black", alpha = .8) +
  scale_x_reverse(breaks = seq(6,22,4)) +
  labs(x = '1,000 Years BP', y = '') +
  guides(color = "none") +
  theme_bw(base_size = 20) +
  theme(strip.background = element_blank())

```



Calculate the detrended variances.

```
emd.dat <- trace.dat %>%
  mutate_at(vars(-Year), emd.res)

(trace.dat - emd.dat) %>%
  select(-Year) %>%
  cbind(Year = trace.dat$Year, .) %>%
  mutate(Period = cut(Year, c(22, 19, 14, 10, 6))) %>%
  group_by(Period) %>%
  summarise_each(funs(var)) %>%
  select(-Year) %>%
  print(width = Inf)

## # A tibble: 4 × 9
##   Period `tmp,Southwest` `tmp,North Central` `tmp,Northeast`
##   <fctr>      <dbl>            <dbl>           <dbl>
## 1 (6,10]     0.08704209    0.0922012    0.1853124
## 2 (10,14]    0.27096878    0.3907428    0.8121861
## 3 (14,19]    0.40989953    1.4185010    1.3565404
## 4 (19,22]    0.10400222    0.2620887    0.2029233
##   `tmp,Southeast` `prc,Southwest` `prc,North Central` `prc,Northeast`
##   <dbl>          <dbl>            <dbl>           <dbl>
## 1 0.4481085    488.1919       778.4795      1230.0957
## 2 0.3543841    772.4411       1015.4651     1191.1768
## 3 0.9364233    1317.6964      1308.8443     1567.4188
## 4 0.2716865    313.4281       391.5337      531.6333
##   `prc,Southeast`
```

```

##          <dbl>
## 1      739.1441
## 2      936.9639
## 3     2021.5821
## 4      649.1395

How do these compare to the overall regional variance?

bbox <- extent(c(-10, 20, 35, 47))

trace.reg.avg <- rbind(
  brick('trace.01-36.22000BP.cam2.TREFHT.22000BP_decavg_400BCE.nc') %>%
    raster::extract(bbox, fun = mean) %>%
    subtract(273.15), # convert from kelvin to C
  brick('trace.01-36.22000BP.cam2.PRECT.22000BP_decavg_400BCE.nc') %>%
    raster::extract(bbox, fun = mean) %>%
    multiply_by(3.154e+10)) %>% # convert to mm/year
t %>% # transpose
as.data.frame %>%
  set_colnames(c('tmp,StudyArea', 'prc,StudyArea')) %>%
  rownames_to_column('Year') %>%
  mutate(Year = as.numeric(substring(Year, 3))) %>%
  filter(Year > 6)

emd.reg.avg <- trace.reg.avg %>%
  mutate_at(vars(-Year), emd.res)

(trace.dat - emd.dat) %>%
  select(-Year) %>%
  cbind(Year = trace.dat$Year, .) %>%
  mutate(Period = cut(Year, c(22, 19, 14, 10, 6))) %>%
  group_by(Period) %>%
  summarise_each(funs(var)) %>%
  select(-Year) %>%
  subtract(((trace.reg.avg - emd.reg.avg) %>%
    magrittr::extract(c(2,2,2,2,3,3,3,3)) %>%
    cbind(Year = trace.dat$Year, .)) %>%
    mutate(Period = cut(Year, c(22, 19, 14, 10, 6))) %>%
    group_by(Period) %>%
    summarise_each(funs(var)) %>%
    select(-Year))) %>%
  print(width = Inf)

## Warning in Ops.factor(left, right): '-' not meaningful for factors

##   Period tmp,Southwest tmp,North Central tmp,Northeast tmp,Southeast
## 1      NA   -0.03803870      -0.03287960     0.06023160     0.32302766

```

```

## 2     NA -0.16399338      -0.04421932    0.37722393   -0.08057806
## 3     NA -0.58309403       0.42550744    0.36354680   -0.05657024
## 4     NA -0.06821138       0.08987510    0.03070968    0.09947292
## prc,Southwest prc,North Central prc,Northeast prc,Southeast
## 1     127.1965        417.4842    869.1004    378.1488
## 2     288.8544        531.8784    707.5902    453.3773
## 3     109.8887        101.0366    359.6112    813.7744
## 4     102.7977        180.9033    321.0029    438.5091

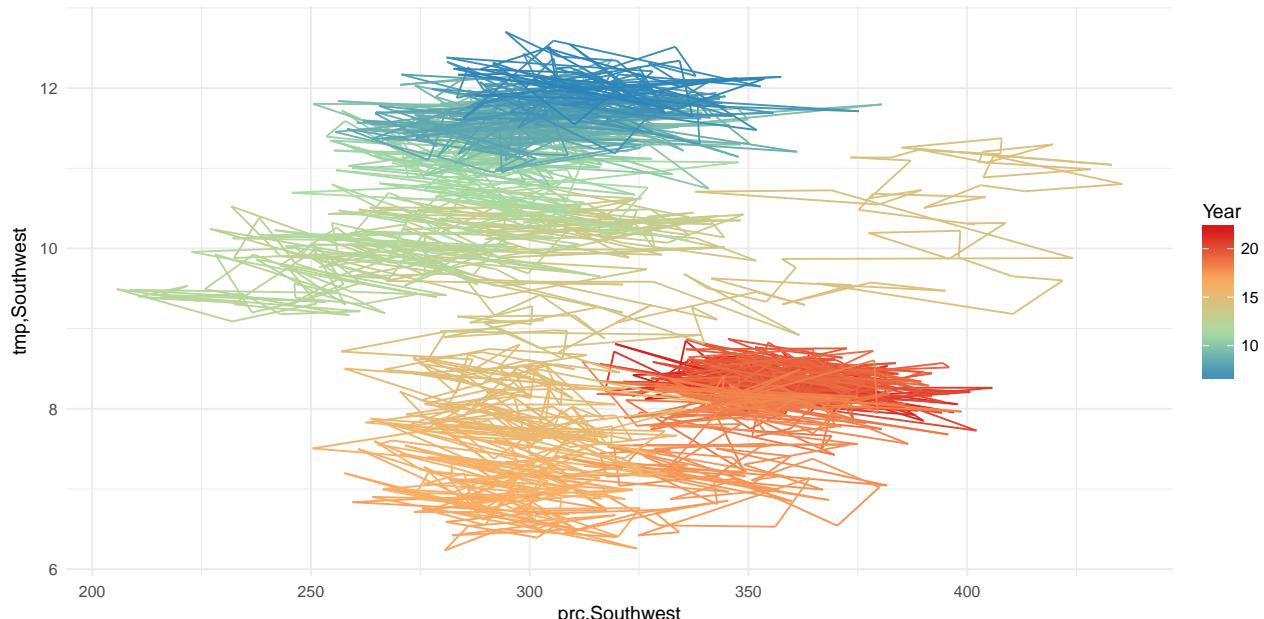
```

Climate trajectories

```

myPalette <- colorRampPalette(rev(brewer.pal(4, "Spectral")))
ggplot(trace.dat, aes(x = 'prc,Southwest', y = 'tmp,Southwest', color = Year)) +
  geom_path(linejoin = 'round') +
  scale_color_gradientn(colors = myPalette(100)) +
  theme_minimal()

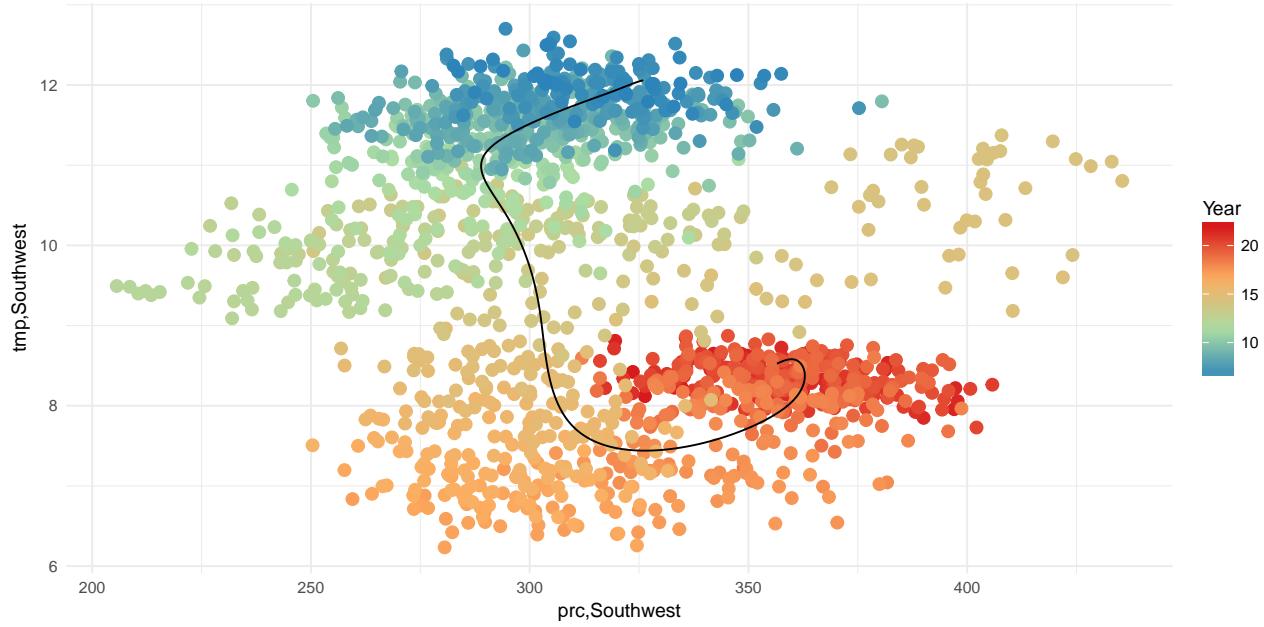
```



```

ggplot(trace.dat, aes(x = 'prc,Southwest', y = 'tmp,Southwest', color = Year)) +
  geom_point(size = 3) +
  geom_path(data = emd.dat, color = 'black') +
  scale_color_gradientn(colors = myPalette(100)) +
  theme_minimal()

```



Spatial patterns

Now let's import previously-downscaled ensemble equilibrium simulations of the Last Glacial Maximum and Mid Holocene, to estimate how the spatial patterns of climate variability have changed over time, and to test for consistency with the transient TraCE simulation.

First import the downscaled data.

```
bbox <- extent(c(-10, 20, 35, 47))

lgm.prc <- brick('Downscaled/ensemble_prc_lgm.tif') %>%
  crop(bbox)
mh.prc <- brick('Downscaled/ensemble_prc_mh6k.tif') %>%
  crop(bbox)

lgm.tmp <- brick('Downscaled/ensemble_tmn_lgm.tif') %>%
  crop(bbox) %>%
  add(brick('Downscaled/ensemble_tmx_lgm.tif')) %>% crop(bbox)) %>%
  divide_by(2)
mh.tmp <- brick('Downscaled/ensemble_tmn_mh6k.tif') %>%
  crop(bbox) %>%
  add(brick('Downscaled/ensemble_tmx_mh6k.tif')) %>% crop(bbox)) %>%
  divide_by(2)
```

Calculate changes in seasonal precipitation and temperature.

```
bySeason <- function(x, season, var){
  if(season == 'djf') {ids <- c(1,2,12)}
```

```

if(season == 'jja') {ids <- c(6,7,8)}

if(var == 'tmp') return(mean(x[[ids]]))
if(var == 'prc') return(sum(x[[ids]]))
}

prc.change.map <- brick(c(
  bySeason(mh.prc, 'djf', 'prc') - bySeason(lgm.prc, 'djf', 'prc'),
  bySeason(mh.prc, 'jja', 'prc') - bySeason(lgm.prc, 'jja', 'prc')))

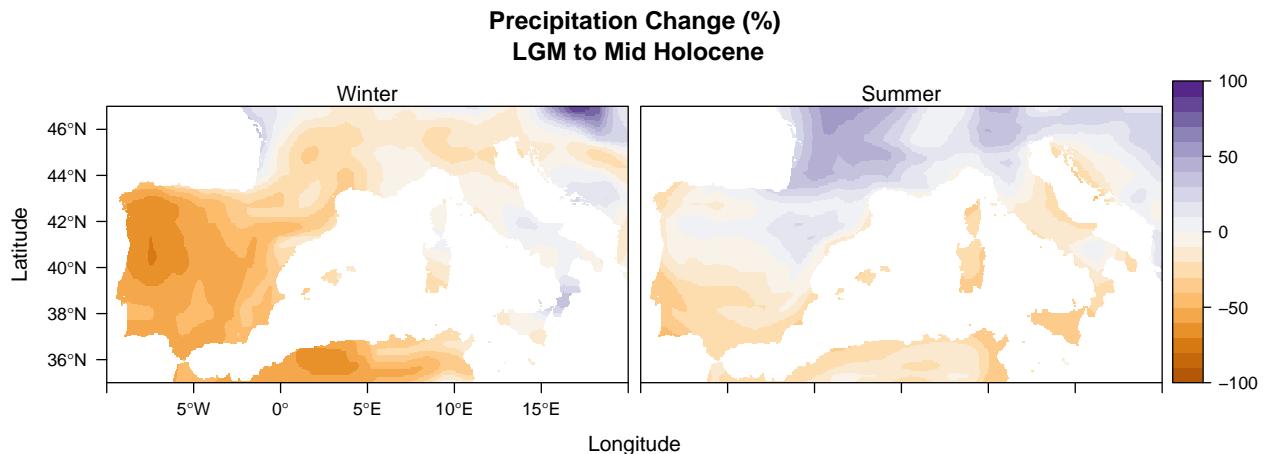
prc.change.map[prc.change.map < -200] <- -200 # so the plot isn't washed out by large precip values

prc.change.map.percent <- brick(c(
  (bySeason(mh.prc, 'djf', 'prc') - bySeason(lgm.prc, 'djf', 'prc')) * 100 / bySeason(lgm.prc, 'djf', 'prc'),
  (bySeason(mh.prc, 'jja', 'prc') - bySeason(lgm.prc, 'jja', 'prc')) * 100 / bySeason(lgm.prc, 'jja', 'prc')))

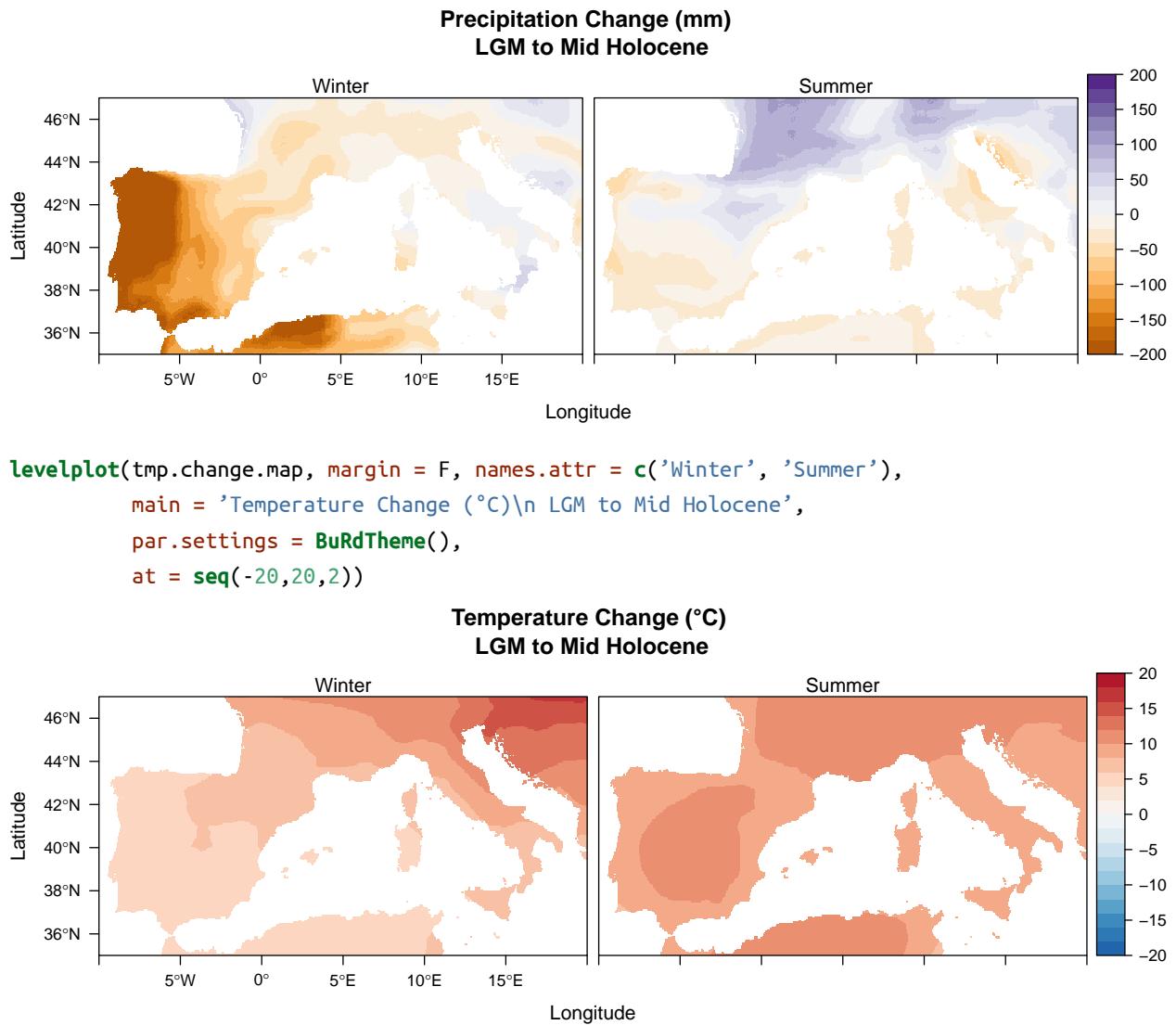
tmp.change.map <- brick(c(
  bySeason(mh.tmp, 'djf', 'tmp') - bySeason(lgm.tmp, 'djf', 'tmp'),
  bySeason(mh.tmp, 'jja', 'tmp') - bySeason(lgm.tmp, 'jja', 'tmp')))
```

Plot the results

```
levelplot(prc.change.map.percent, margin = F, names.attr = c('Winter', 'Summer'),
          main = 'Precipitation Change (%)\\n LGM to Mid Holocene',
          par.settings = PuOrTheme(),
          at = seq(-100,100,10))
```



```
# note in the next plot all values less than -200 have been turned into -200
levelplot(prc.change.map, margin = F, names.attr = c('Winter', 'Summer'),
          main = 'Precipitation Change (mm) \\n LGM to Mid Holocene',
          par.settings = PuOrTheme(),
          at = seq(-200,200,20))
```



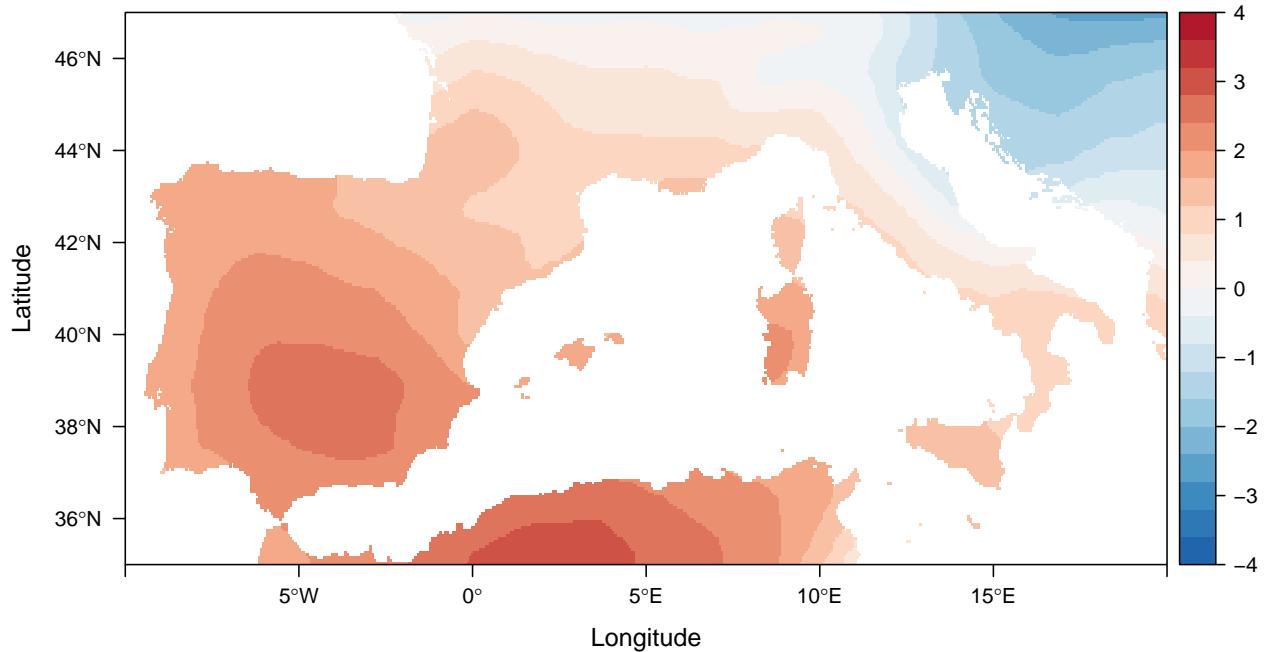
Now we can calculate changes in seasonality. For temperature, this is just the standard deviation of all 12 monthly averages. For precipitation, we will use the coefficient of variation.

```
tmp.seasonality <- calc(mh.tmp, sd) - calc(lgm.tmp, sd)
prc.seasonality <- cv(mh.prc) - cv(lgm.prc)
```

Plot the results.

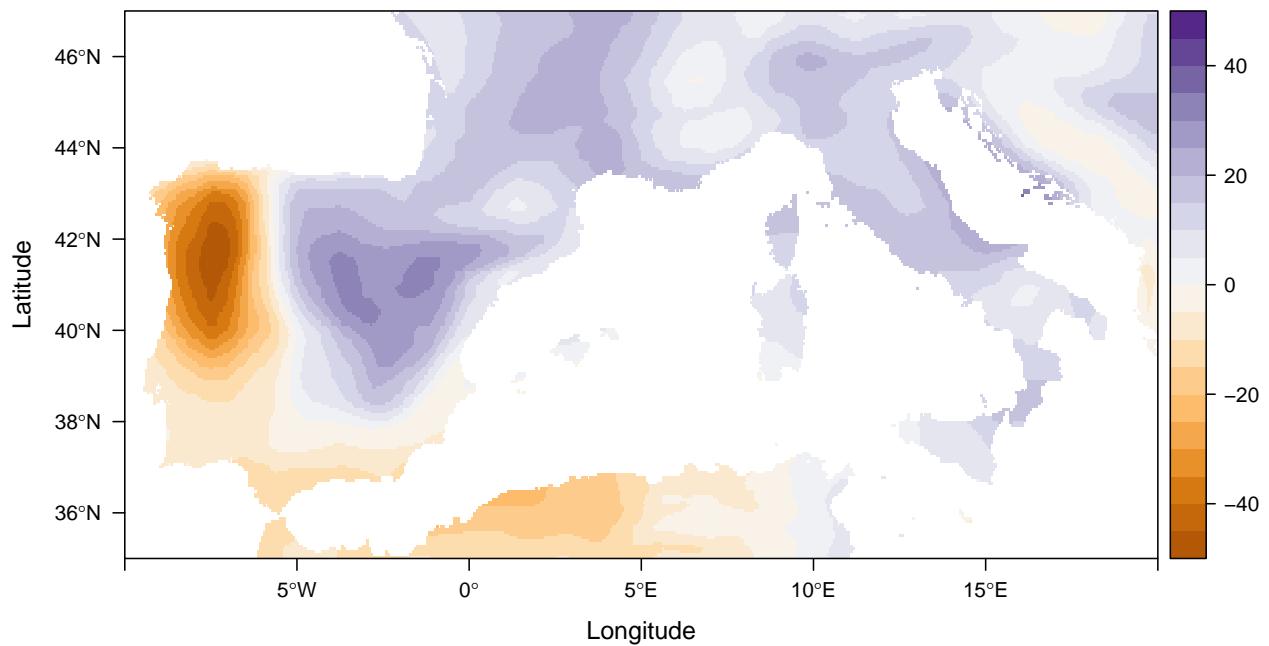
```
levelplot(tmp.seasonality, margin = F,
          main = 'Change in temperature seasonality (SD)\n LGM to Mid Holocene',
          par.settings = BuRdTheme(),
          at = seq(-4, 4, .4))
```

Change in temperature seasonality (SD) LGM to Mid Holocene



```
levelplot(prc.seasonality, margin = F,
  main = 'Change in precipitation seasonality (CV)\n LGM to Mid Holocene',
  par.settings = PuOrTheme(),
  at = seq(-50, 50, 5))
```

Change in precipitation seasonality (CV) LGM to Mid Holocene



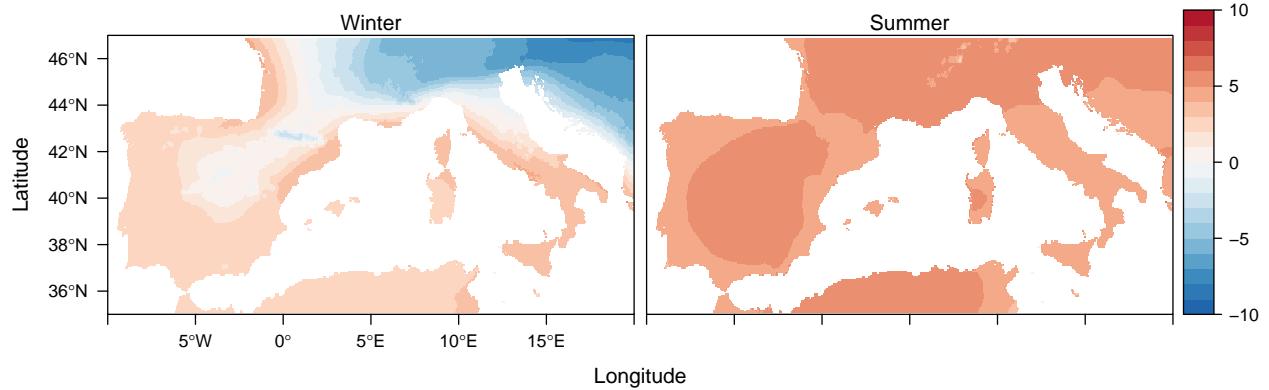
What about changes in spatial heterogeneity?

```
wts <- matrix(c(0,0,1,0,0,0,1,1,1,0,1,1,1,1,1,0,1,1,1,0,0,0,1,0,0), nrow = 5)

tmp.hetero <- brick(c(
  bySeason(mh.tmp, 'djf', 'tmp') %>%
    focal(w = wts, sd, na.rm = T) %>%
    subtract(
      bySeason(lgm.tmp, 'djf', 'tmp') %>%
        focal(w = wts, sd, na.rm = T)),
  bySeason(mh.tmp, 'jja', 'tmp') %>%
    focal(w = wts, sd, na.rm = T) %>%
    subtract(
      bySeason(lgm.tmp, 'jja', 'tmp') %>%
        focal(w = wts, sd, na.rm = T)))) %>%
mask(mh.tmp[[1]]) # clip buffer added by window

levelplot(tmp.hetero, margin = F, names.attr = c('Winter', 'Summer'),
  main = 'Temperature heterogeneity (SD in 25km radius) change\n LGM to Mid Holocene',
  par.settings = BuRdTheme(), at = seq(-10, 10, 1))

  Temperature heterogeneity (SD in 25km radius) change
  LGM to Mid Holocene
```



Same for precipitation.

```
prc.hetero.sd <- brick(c(
  bySeason(mh.prc, 'djf', 'prc') %>%
    focal(w = wts, sd, na.rm = T) %>%
    subtract(
      bySeason(lgm.prc, 'djf', 'prc') %>%
        focal(w = wts, sd, na.rm = T)),
  bySeason(mh.prc, 'jja', 'prc') %>%
    focal(w = wts, sd, na.rm = T) %>%
    subtract(
      bySeason(lgm.prc, 'jja', 'prc') %>%
        focal(w = wts, sd, na.rm = T)))) %>%
mask(mh.prc[[1]]) # clip buffer added by window
```

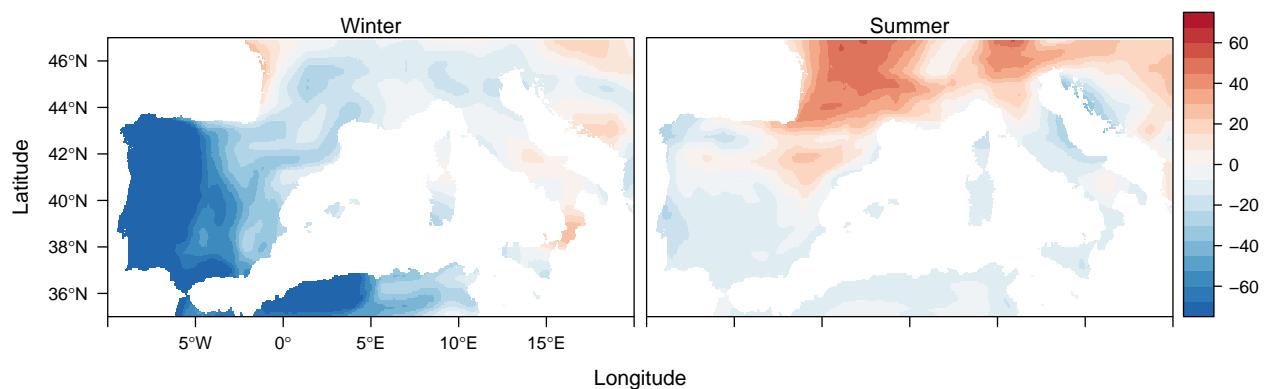
```
#levelplot(prc.hetero.sd, margin = F, names.attr = c('Winter', 'Summer'),
#          main = 'Precipitation heterogeneity (SD in 25km radius) change\n LGM to Mid Holocene',
#          par.settings = BuRdTheme(), at = seq(-500, 500, 50))
```

#capped at -75

```
prc.hetero.sd[prc.hetero.sd < -75] <- -75
```

```
levelplot(prc.hetero.sd, margin = F, names.attr = c('Winter', 'Summer'),
          main = 'Precipitation heterogeneity (SD in 25km radius) change\n LGM to Mid Holocene',
          par.settings = BuRdTheme(), at = seq(-75, 75, 7.5))
```

Precipitation heterogeneity (SD in 25km radius) change
LGM to Mid Holocene

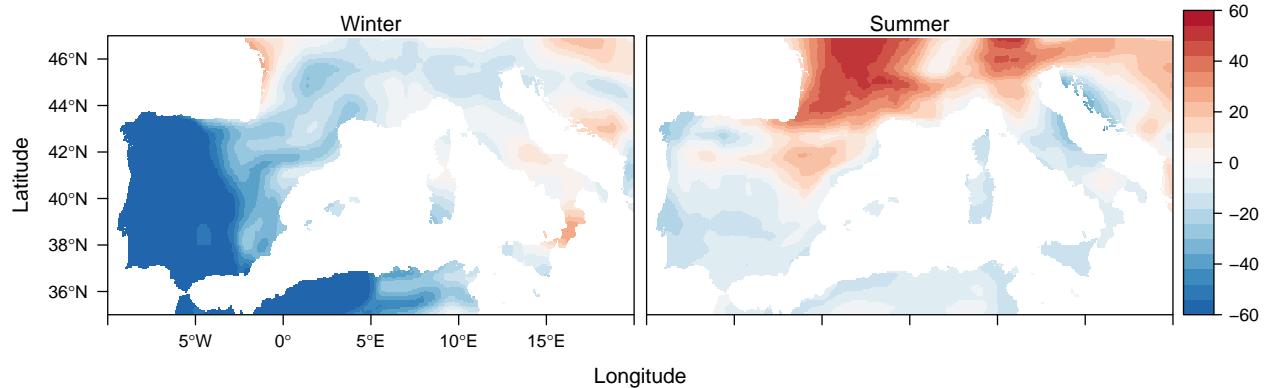


capped at -60

```
prc.hetero.sd[prc.hetero.sd < -60] <- -60
```

```
levelplot(prc.hetero.sd, margin = F, names.attr = c('Winter', 'Summer'),
          main = 'Precipitation heterogeneity (SD in 25km radius) change\n LGM to Mid Holocene',
          par.settings = BuRdTheme(), at = seq(-60, 60, 6))
```

Precipitation heterogeneity (SD in 25km radius) change
LGM to Mid Holocene



now cv

```
prc.hetero.cv <- brick(c(
  bySeason(mh.prc, 'djf', 'prc') %>%
  focal(w = wts, cv, na.rm = T) %>%
```

