

## **Wikiprint Book**

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## Analysing model drift in South-western Iberia

We will take as reference for measuring the drift the lead-month 0 forecast (i.e. the July initializations)

```
> ref <- loadECOMS(dataset = "CFSv2_seasonal_16", var = "tas", lonLim = c(-10,-1), latLim = c(36,40), season = 7, years = 2006)
> plotMeanField(ref)
> ref.field <- apply(ref$Data, MARGIN = c(4,3), FUN = mean, na.rm = TRUE)
```



Next, we will load the forecast of the target variable recursively for lead month values from 1 to 6 (i.e., the initializations from January to June). The different objects are arranged in a list:

```
> cfs.list <- lapply(1:6, function(lead.month) {
  loadECOMS(dataset = "CFSv2_seasonal_16", var = "tas", lonLim = c(-10,-1), latLim = c(36,40), season = 7, years = 2006)
})
```

In order to visualize the departures of each lead month from the reference in the same range of values, we will use the `splot` method for plotting spatial objects of the library `sp`. To this aim, we will first compute the multi-member spatial mean for each lead month forecast, and then we will arrange the data in a matrix of 6 columns (one for each month), and `x*y` coordinates rows:

```
> require(sp)
> # Matrix of anomalies between lead month and reference
> aux.mat <- matrix(ncol = length(cfs.list), nrow = length(ref$xyCoords$x)*length(ref$xyCoords$y))
+ for (i in 1:length(cfs.list)) {
+   mm.field <- apply(cfs.list[[i]]$Data, MARGIN = c(4,3), FUN = mean, na.rm = TRUE)
+   aux.mat[,i] <- mm.field - ref.field
+ }
> # 2D coordinates
> xy <- expand.grid(ref$xyCoords$x, ref$xyCoords$y)
> xy.coords <- coordinates(points2grid(points = SpatialPoints(xy), tolerance = .003))
> df <- cbind.data.frame(xy.coords, aux.mat)
> names(df) <- c("x", "y", paste("LeadMonth_", 1:6, sep = ""))
> str(df)
'data.frame': 55 obs. of 8 variables:
 $ x      : num -10.31 -9.38 -8.44 -7.5 -6.56 ...
 $ y      : num 40.2 40.2 40.2 40.2 40.2 ...
 $ LeadMonth_1: num 0.0398 0.0691 0.1427 0.116 0.1935 ...
 $ LeadMonth_2: num 0.386 0.451 0.472 0.311 0.161 ...
 $ LeadMonth_3: num 0.0428 0.0787 0.0707 -0.125 -0.2751 ...
 $ LeadMonth_4: num 0.261 0.258 0.131 -0.177 -0.464 ...
 $ LeadMonth_5: num 0.0576 0.154 0.1327 -0.1282 -0.3963 ...
 $ LeadMonth_6: num 0.0789 0.192 0.0989 -0.2667 -0.6457 ...
> Conversion of the data.frame to a Spatial gridded object:
> coordinates(df) <- c(1,2)
> class(df)
[1] "SpatialPixelsDataFrame"
attr(,"package")
[1] "sp"
```

In the next lines we use the `splot` method of package `sp`, generating a lattice-type map. In first place, we will also load a `SpatialLines` dataset remotely stored at Santander Met Group server, in order to represent the geographical lines in the lattice map generated:

```
> load(url("http://meteo.unican.es/work/downscaler/aux/wlines.rda"), verbose = TRUE)
Loading objects:
 wlines
> ll <- list("sp.lines", wlines)
> splot(df, as.table = TRUE, col.regions = colorRampPalette(c("blue", "white", "red")), at = seq(-2.5, 2.5, .25), scales = list())
```



The results show how a increasing lead month leads to a negative bias of the forecast, demonstrating that the mean state of a variable of a forecast is not stationary through time.